# Table of Contents

What is Amazon Personalize ................................................................. 1
Pricing for Amazon Personalize .......................................................... 1
Guidance for first-time users ............................................................. 2
  Discovering Amazon Personalize with the Magic Movie Machine ........ 2
  Navigating this guide ........................................................................ 2
Related AWS services and solutions ...................................................... 3
Third-party services ........................................................................... 3
Learn more .......................................................................................... 4
How it works ......................................................................................... 5
Amazon Personalize workflow summary .............................................. 5
Amazon Personalize terms .................................................................... 6
  Data import and management ........................................................... 6
  Training ............................................................................................. 7
  Model deployment and recommendations ......................................... 9
Amazon Personalize data ..................................................................... 9
  Interactions data .............................................................................. 10
  Item data ......................................................................................... 10
  User data ......................................................................................... 10
Setting up Amazon Personalize ............................................................ 11
  Sign up for an AWS account ............................................................ 11
  Create an administrative user .......................................................... 11
  Regions and endpoints ..................................................................... 12
Setting up permissions ......................................................................... 12
  Giving users permission to access Amazon Personalize .................... 13
  Giving Amazon Personalize permission to access your resources ........ 14
  Giving Amazon Personalize access to Amazon S3 resources ............. 16
  Giving Amazon Personalize permission to use your AWS KMS key ..... 20
Setting up the AWS CLI ....................................................................... 21
Setting up the AWS SDKs ................................................................... 22
Getting started ..................................................................................... 24
  Getting started prerequisites ............................................................ 24
    Creating the training data (Domain dataset group) ......................... 25
    Creating the training data (Custom dataset group) ......................... 25
Getting started with a Domain dataset group ....................................... 26
  Getting started with a Domain dataset group (console) .................... 26
  Getting started with a Domain dataset group (SDK for Java 2.x) ...... 31
  Getting started with a Domain dataset group (SDK for Python (Boto3)) 37
  Getting started with a Domain dataset group (SDK for JavaScript v3) .. 41
Getting started with a Custom dataset group ..................................... 47
  Getting started (console) ................................................................. 47
  Getting started (AWS CLI) ............................................................... 55
  Getting started (SDK for Python (Boto3)) ....................................... 62
  Getting started (SDK for Java 2.x) .................................................. 66
Cleaning up resources ......................................................................... 74
  Cleaning up domain-based resources .............................................. 74
  Cleaning up custom resources ....................................................... 75
Datasets and schemas ......................................................................... 76
Datasets ............................................................................................. 76
  Interactions datasets ................................................................. 77
  Users datasets ............................................................................... 80
  Items datasets ............................................................................... 80
Schemas ............................................................................................. 82
  Schema formatting requirements .................................................... 83
  Domain datasets and schemas ....................................................... 84
Custom datasets and schemas ................................................................. 94
Data format guidelines ............................................................................ 101
Bulk data format guidelines and requirements ....................................... 101
Interactions data example ........................................................................ 102
Formatting explicit impressions ............................................................... 103
Formatting categorical data ...................................................................... 103
Domain use cases and custom recipes .................................................... 105
Use case and recipe features .................................................................... 105
Real-time personalization ........................................................................ 105
Exploration ............................................................................................... 106
Automatic updates .................................................................................. 106
Choosing a use case ................................................................................ 107
VIDEO_ON_DEMAND use cases ............................................................ 108
ECOMMERCE use cases .......................................................................... 110
Choosing a recipe .................................................................................... 113
Amazon Personalize recipe types by use case ........................................ 113
Amazon Personalize recipes ................................................................. 114
Viewing available Amazon Personalize recipes ...................................... 115
USER_PERSONALIZATION .................................................................... 115
POPULAR_ITEMS .................................................................................... 136
PERSONALIZED_RANKING recipes ...................................................... 139
RELATED_ITEMS .................................................................................... 143
USER_SEGMENTATION recipes ............................................................... 149
Readiness checklist ................................................................................ 152
  Have you matched your use cases to Amazon Personalize resources? … 152
  Do you have enough interactions data? ................................................ 153
  Do you have a real-time event streaming architecture in place? .......... 153
  Is your data optimized for Amazon Personalize? .................................. 153
  Do you collect optional data that can improve recommendations? .... 153
  Do you have a plan to test your recommendations? ................................ 154
  Do you have additional business goals? ................................................ 154
Amazon Personalize workflow ................................................................ 155
  Step 1: Creating a dataset group .......................................................... 155
    Creating a dataset group (console) .................................................... 156
    Creating a dataset group (AWS CLI) ................................................ 157
    Creating a dataset group (AWS SDKs) .............................................. 157
  Step 2: Preparing and importing data .................................................... 159
    Preparing and importing bulk data ................................................... 159
    Importing individual records .......................................................... 182
  Step 3: Creating recommenders or custom resources ............................ 190
    Creating domain recommenders ...................................................... 191
    Creating custom resources ............................................................ 210
  Step 4: Getting recommendations ...................................................... 244
    Recommendation scores ................................................................... 244
    Getting real-time recommendations .............................................. 244
    Batch recommendations and user segments (custom resources) ....... 259
Maintaining recommendation relevance .................................................. 278
  Keeping datasets current ...................................................................... 278
  Maintaining domain recommenders .................................................... 278
  Maintaining custom solutions ............................................................ 279
Recording events .................................................................................... 280
  Requirements for recording events and training a model ...................... 280
  How real-time events influence recommendations .............................. 281
  Creating an event tracker ..................................................................... 281
  Recording events with the PutEvents operation ................................... 283
    Recording a single event .................................................................. 283
    Recording multiple events with event value data ............................. 285
Use case example
A/B testing best practices
Filtering real-time recommendations (AWS SDKs)
Deleting a dataset (AWS CLI)

Managing data
Updating data
How new data influences real-time recommendations
Updating existing bulk data
Updating data with individual import operations
Replacing a dataset's schema
Guidelines and requirements
Replacing a dataset's schema (console)
Replacing a dataset's schema (AWS CLI)
Replacing a dataset's schema (AWS SDKs)
Analyzing data in datasets
Required permissions for analyzing data
Data insights
Viewing dataset insights and statistics
Exporting a dataset
Dataset export job permissions requirements
Creating a dataset export job (console)
Creating a dataset export job (AWS CLI)
Creating a dataset export job (AWS SDKs)
Deleting data
Deleting a dataset (console)
Deleting a dataset (AWS CLI)
Deleting a dataset (AWS SDKs)
Filtering results
Filter expressions
Creating filter expressions
Filter expression examples
Filtering real-time recommendations
Filtering real-time recommendations (console)
Filtering real-time recommendations (AWS CLI)
Filtering real-time recommendations (AWS SDKs)
Filtering batch recommendations and user segments (custom resources)
Providing filter values in your input JSON
Filtering batch workflows (console)
Filtering batch workflows (AWS SDKs)
Measuring impact of recommendations
Measuring recommendation impact with a metric attribution
Guidelines and requirements
Creating a metric attribution
Managing a metric attribution
Publishing and viewing results
Measuring recommendation impact with A/B testing
A/B testing best practices
A/B testing with CloudWatch Evidently
Personalizing search results from OpenSearch
Use case example
Personalized search workflow
How the Amazon Personalize Search Ranking plugin works
Additional information
Guidelines and requirements
Plugin requirements
Setting up Amazon OpenSearch Service permissions
<table>
<thead>
<tr>
<th>Section</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>Setting up open source OpenSearch permissions</td>
<td>356</td>
</tr>
<tr>
<td>Setting up OpenSearch and installing the plugin</td>
<td>358</td>
</tr>
<tr>
<td>Setting up Amazon OpenSearch Service</td>
<td>358</td>
</tr>
<tr>
<td>Setting up open source OpenSearch</td>
<td>359</td>
</tr>
<tr>
<td>Configuring the plugin</td>
<td>360</td>
</tr>
<tr>
<td>Fields for the personalized_search_ranking response processor</td>
<td>361</td>
</tr>
<tr>
<td>Creating a pipeline with Amazon OpenSearch Service</td>
<td>361</td>
</tr>
<tr>
<td>Creating a pipeline with open source OpenSearch</td>
<td>362</td>
</tr>
<tr>
<td>Applying the plugin to OpenSearch queries</td>
<td>362</td>
</tr>
<tr>
<td>Applying the plugin to Amazon OpenSearch Service queries</td>
<td>363</td>
</tr>
<tr>
<td>Applying the plugin to queries in open source OpenSearch</td>
<td>364</td>
</tr>
<tr>
<td>Comparing OpenSearch results with results from the plugin</td>
<td>365</td>
</tr>
<tr>
<td>Comparing results with Amazon OpenSearch Service</td>
<td>365</td>
</tr>
<tr>
<td>Comparing results with open source OpenSearch</td>
<td>366</td>
</tr>
<tr>
<td>Monitoring the plugin</td>
<td>367</td>
</tr>
<tr>
<td>Monitoring the plugin with Amazon OpenSearch Service</td>
<td>367</td>
</tr>
<tr>
<td>Monitoring the plugin with open source OpenSearch</td>
<td>368</td>
</tr>
<tr>
<td>Pipeline metrics example</td>
<td>368</td>
</tr>
<tr>
<td>Tagging resources</td>
<td>370</td>
</tr>
<tr>
<td>Managing tags</td>
<td>370</td>
</tr>
<tr>
<td>Additional information</td>
<td>371</td>
</tr>
<tr>
<td>Adding tags to Amazon Personalize resources</td>
<td>371</td>
</tr>
<tr>
<td>Adding tags (console)</td>
<td>371</td>
</tr>
<tr>
<td>Adding tags (AWS CLI)</td>
<td>372</td>
</tr>
<tr>
<td>Adding tags (AWS SDKs)</td>
<td>372</td>
</tr>
<tr>
<td>Using tags in IAM policies</td>
<td>375</td>
</tr>
<tr>
<td>Troubleshooting</td>
<td>377</td>
</tr>
<tr>
<td>Frequently asked questions</td>
<td>377</td>
</tr>
<tr>
<td>Data import and management</td>
<td>377</td>
</tr>
<tr>
<td>Creating a custom solution and solution version</td>
<td>378</td>
</tr>
<tr>
<td>Model deployment (custom campaigns)</td>
<td>378</td>
</tr>
<tr>
<td>Recommendations</td>
<td>379</td>
</tr>
<tr>
<td>Filtering recommendations</td>
<td>379</td>
</tr>
<tr>
<td>Error messages</td>
<td>380</td>
</tr>
<tr>
<td>Data import and management</td>
<td>380</td>
</tr>
<tr>
<td>Creating a solution and solution version (custom resources)</td>
<td>381</td>
</tr>
<tr>
<td>Model deployment (custom campaigns)</td>
<td>381</td>
</tr>
<tr>
<td>Recommenders (Domain dataset groups)</td>
<td>381</td>
</tr>
<tr>
<td>Recommendations</td>
<td>381</td>
</tr>
<tr>
<td>Filtering recommendations</td>
<td>382</td>
</tr>
<tr>
<td>Specifying resources with AWS CloudFormation</td>
<td>383</td>
</tr>
<tr>
<td>Amazon Personalize and AWS CloudFormation templates</td>
<td>383</td>
</tr>
<tr>
<td>Example AWS CloudFormation templates for Amazon Personalize resources</td>
<td>383</td>
</tr>
<tr>
<td>CreateDatasetGroup</td>
<td>384</td>
</tr>
<tr>
<td>CreateDataset</td>
<td>384</td>
</tr>
<tr>
<td>CreateSchema</td>
<td>385</td>
</tr>
<tr>
<td>CreateSolution</td>
<td>385</td>
</tr>
<tr>
<td>Learn more about AWS CloudFormation</td>
<td>386</td>
</tr>
<tr>
<td>Security</td>
<td>387</td>
</tr>
<tr>
<td>Data protection</td>
<td>387</td>
</tr>
<tr>
<td>Data encryption</td>
<td>388</td>
</tr>
<tr>
<td>Identity and Access Management</td>
<td>389</td>
</tr>
<tr>
<td>Audience</td>
<td>389</td>
</tr>
<tr>
<td>Authenticating with identities</td>
<td>389</td>
</tr>
<tr>
<td>Managing access using policies</td>
<td>392</td>
</tr>
<tr>
<td>How Amazon Personalize works with IAM</td>
<td>393</td>
</tr>
<tr>
<td>Cross-service confused deputy prevention</td>
<td>398</td>
</tr>
<tr>
<td>Section</td>
<td>Page</td>
</tr>
<tr>
<td>------------------------------------------------------------------------</td>
<td>------</td>
</tr>
<tr>
<td>Identity-based policy examples</td>
<td>399</td>
</tr>
<tr>
<td>Troubleshooting</td>
<td>403</td>
</tr>
<tr>
<td>Logging and monitoring</td>
<td>404</td>
</tr>
<tr>
<td>Monitoring</td>
<td>404</td>
</tr>
<tr>
<td>CloudWatch metrics for Amazon Personalize</td>
<td>407</td>
</tr>
<tr>
<td>Logging Amazon Personalize API calls with AWS CloudTrail</td>
<td>410</td>
</tr>
<tr>
<td>Compliance validation</td>
<td>411</td>
</tr>
<tr>
<td>Resilience</td>
<td>412</td>
</tr>
<tr>
<td>Infrastructure security</td>
<td>412</td>
</tr>
<tr>
<td>VPC endpoints (AWS PrivateLink)</td>
<td>413</td>
</tr>
<tr>
<td>Considerations for Amazon Personalize VPC endpoints</td>
<td>413</td>
</tr>
<tr>
<td>Creating an interface VPC endpoint for Amazon Personalize</td>
<td>413</td>
</tr>
<tr>
<td>Creating a VPC endpoint policy for Amazon Personalize</td>
<td>414</td>
</tr>
<tr>
<td>Endpoints and quotas</td>
<td>415</td>
</tr>
<tr>
<td>Amazon Personalize endpoints and regions</td>
<td>415</td>
</tr>
<tr>
<td>Compliance</td>
<td>415</td>
</tr>
<tr>
<td>Service quotas</td>
<td>415</td>
</tr>
<tr>
<td>Requesting a quota increase</td>
<td>419</td>
</tr>
<tr>
<td>API reference</td>
<td>420</td>
</tr>
<tr>
<td>Actions</td>
<td>420</td>
</tr>
<tr>
<td>Amazon Personalize</td>
<td>422</td>
</tr>
<tr>
<td>Amazon Personalize Events</td>
<td>597</td>
</tr>
<tr>
<td>Amazon Personalize Runtime</td>
<td>603</td>
</tr>
<tr>
<td>Data Types</td>
<td>612</td>
</tr>
<tr>
<td>Amazon Personalize</td>
<td>613</td>
</tr>
<tr>
<td>Amazon Personalize Events</td>
<td>725</td>
</tr>
<tr>
<td>Amazon Personalize Runtime</td>
<td>731</td>
</tr>
<tr>
<td>Common Errors</td>
<td>734</td>
</tr>
<tr>
<td>Common Parameters</td>
<td>735</td>
</tr>
<tr>
<td>Document history</td>
<td>738</td>
</tr>
<tr>
<td>AWS glossary</td>
<td>746</td>
</tr>
</tbody>
</table>
What is Amazon Personalize?

Amazon Personalize is a fully managed machine learning service that uses your data to generate item recommendations for your users. It can also generate user segments based on the users’ affinity for certain items or item metadata.

Common use case include the following:

- **Personalizing a video streaming app** – You can use preconfigured or customizable Amazon Personalize resources to add multiple types of personalized video recommendations to your streaming app. For example, *Top picks for you, More like X* and *Most popular* video recommendations.

- **Adding product recommendations to an ecommerce app** – You can use preconfigured or customizable Amazon Personalize resources to add multiple types of personalized product recommendations to your retail app. For example, *Recommended for you, Frequently bought together* and *Customers who viewed X also viewed* product recommendations.

- **Creating personalized emails** – You can use customizable Amazon Personalize resources to generate batch recommendations for all users on an email list. Then you can use an [AWS service](p. 3) or [third party service](p. 3) to send users personalized emails recommending items in your catalog.

- **Creating a targeted marketing campaign** – You can use Amazon Personalize to generate segments of users who will most likely interact with items in your catalog. Then you can use an [AWS service](p. 3) or [third party service](p. 3) to create a targeted marketing campaign that promotes different items to different user segments.

- **Personalizing search results** – You can use customizable Amazon Personalize resources to personalize search results for your users. For example, Amazon Personalize can re-rank search results that you generate with [OpenSearch](p. 351).

Amazon Personalize generates recommendations primarily based on interactions data. Interactions data is data from your users interacting with items in your catalog. For example, users clicking different items. Your interactions data can come from both your historical bulk interaction records in a CSV file, and real-time events from your users as they interact with your catalog. In some cases, Amazon Personalize also uses metadata about items and users, such as genre, price, or gender.

When you import bulk data, you can use Amazon SageMaker Data Wrangler to import data from 40+ sources and prepare it for Amazon Personalize. For more information, see [Preparing and importing data using Amazon SageMaker Data Wrangler](p. 160).

Amazon Personalize includes API operations for real-time personalization, and batch operations for bulk recommendations and user segments. You can get started quickly with use-case optimized recommenders for your business domain, or you can create your own configurable custom resources.

Topics

- [Pricing for Amazon Personalize](p. 1)
- [Guidance for first-time Amazon Personalize users](p. 2)
- [Related AWS services and solutions](p. 3)
- [Third-party services](p. 3)
- [Learn more](p. 4)

Pricing for Amazon Personalize

With Amazon Personalize, there are no minimum fees and no upfront commitments. The [AWS Free Tier](p. 3) provides a monthly quota of up to 20 GB of data processing per available AWS region, up to 100 hours
of training time per eligible AWS region, and up to 50 TPS-hours of real-time recommendations/month. The free tier is valid for the first two months of usage.

For a complete list of charges and prices, see Amazon Personalize pricing.

**Guidance for first-time Amazon Personalize users**

If you're a first-time user of Amazon Personalize, the following resources can help you get started.

**Topics**

- Discovering Amazon Personalize with the Magic Movie Machine (p. 2)
- Navigating this guide (p. 2)

**Discovering Amazon Personalize with the Magic Movie Machine**

The Magic Movie Machine is an interactive learning experience. It helps you discover Amazon Personalize features and learn more about generating recommendations. For a short introduction, see the video below. Then try the Magic Movie Machine.

**Getting Started with Amazon Personalize**

**Navigating this guide**

We recommend you read the following sections in order:

1. **How it works (p. 5)** – This section introduces the Amazon Personalize workflow and walks you through the steps to create personalized experiences for your users. This section also includes common Amazon Personalize terms and their definitions. Start with this section to make sure you have good understanding of Amazon Personalize workflows and terms before you start getting recommendations.

2. **Setting up Amazon Personalize (p. 11)** – In this section you set up your AWS account, set up the required permissions to use Amazon Personalize, and set up the AWS CLI and the AWS SDKs to use and manage Amazon Personalize.

3. **Getting started (p. 24)** – In this section you get started using Amazon Personalize with a simple movie dataset. Complete these tutorials to get hands-on experience with Amazon Personalize. You can choose to either get started with a Domain dataset group or a Custom dataset group:
   - To get started creating a Domain dataset group, complete the Getting started prerequisites (p. 24) and then start the tutorials in Getting started with a Domain dataset group (p. 26).
   - To get started with a Custom dataset group, complete the Getting started prerequisites (p. 24) and then start the tutorials in Getting started with a Domain dataset group (p. 26).

4. **Domain use cases and custom recipes (p. 105)** – Learn about the domain use cases and custom recipes you can use to train a model in Amazon Personalize. Use this information to help you match your use case to resources in Amazon Personalize.

5. **Readiness checklist (p. 152)** – Review the readiness checklist to start preparing to use Amazon Personalize with your own data. This checklist provides lists of Amazon Personalize features, requirements, and data guidance. It can help you plan or you can use it as a reference as you create resources in Amazon Personalize.

6. **Amazon Personalize workflow (p. 155)** – This section guides you through the complete Amazon Personalize workflow. It provides step-by-step instructions for creating a Domain dataset group or a
Custom dataset group, preparing and importing data, creating recommenders or custom resources, and getting recommendations.

7. **Recording events (p. 280)** – This section covers how to record user interaction events in real time. After you have set up your Amazon Personalize resources, complete this section to learn how to keep your Interactions dataset up to date with your users' behavior. You do this by recording interaction events with an event tracker and the `PutEvents` operation.

8. **Filtering recommendations and user segments (p. 311)** – This section covers how to filter recommendations. Complete this section to learn how to construct filter expressions to filter recommendations based on custom criteria. For example, you might not want to recommend products that a user has already purchased, or recommend movies that a user has already watched.

### Related AWS services and solutions

Amazon Personalize integrates seamlessly with other AWS services and solutions. For example, you can:

- Use Amazon SageMaker Data Wrangler (Data Wrangler) to import data from 40+ sources into an Amazon Personalize dataset. Data Wrangler is a feature of Amazon SageMaker Studio that provides an end-to-end solution to import, prepare, transform, and analyze data. For more information, see [Preparing and importing data using Amazon SageMaker Data Wrangler (p. 160)](https://docs.aws.amazon.com/personalize/latest/devguide/prepare-import-data.html).

- Use AWS Amplify to record user interaction events. Amplify includes a JavaScript library for recording events from web client applications. And it includes a library for recording events in server code. For more information, see [Amplify - analytics](https://docs.aws.amazon.com/amplify/latest/userguide/analytics.html).

- Automate and schedule Amazon Personalize tasks with [Maintaining Personalized Experiences with Machine Learning](https://aws.amazon.com/solutions/whitepapers/maintaining-personalized-experiences-with-machine-learning/). This AWS Solutions Implementation automates the Amazon Personalize workflow, including data import, solution version training, and batch workflows.

- Use Amazon CloudWatch Evidently to perform A/B testing with Amazon Personalize recommendations. For more information, see [A/B testing with CloudWatch Evidently (p. 348)](https://docs.aws.amazon.com/personalize/latest/devguide/a-b-testing-with-cloudwatch-evidently.html).

- Use Amazon Pinpoint to create targeted marketing campaigns. For an example that shows how to use Amazon Pinpoint and Amplify to add Amazon Personalize recommendations to a marketing email campaign and a web app, see [Web Analytics with Amplify](https://aws.amazon.com/solutions/whitepapers/web-analytics-with-amplify/).

### Third-party services

Amazon Personalize works well with various third-party services.

- **Amplitude** – You can use Amplitude to track user actions to help you understand your users' behavior. For information on using Amplitude and Amazon Personalize, see the following AWS Partner Network (APN) blog post: [Measuring the Effectiveness of Personalization with Amplitude and Amazon Personalize](https://aws.amazon.com/blogs/machine-learning/measuring-the-effectiveness-of-personalization-with-amplitude-and-amazon-personalize/).

- **Braze** – You can use Braze to send users personalized emails recommending items in your catalog. Braze is a market leading messaging platform (email, push, SMS). For a workshop that shows how to integrate Amazon Personalize and Braze, see [Amazon Personalize workshop](https://aws.amazon.com/solutions/whitepapers/amazon-personalize-workshop/).

- **mParticle** – You can use mParticle to collect event data from your app. For an example that shows how to use mParticle and Amazon Personalize to implement personalized product recommendations, see [How to harness the power of a CDP for machine learning: Part 2](https://aws.amazon.com/blogs/machine-learning/how-to-harness-the-power-of-a-cdp-for-machine-learning-part-2/).

- **Optimizely** – You can use Optimizely to perform A/B testing with Amazon Personalize recommendations. For information on using Optimizely and Amazon Personalize, see [Optimizely integrates with Amazon Personalize to combine powerful machine learning with experimentation](https://aws.amazon.com/blogs/machine-learning/optimizely-integrates-with-amazon-personalize-to-combine-powerful-machine-learning-with-experimentation/).

- **Segment** – You can use Segment to send your data to Amazon Personalize. For more information on integrating Segment with Amazon Personalize, see [Amazon Personalize Destination](https://aws.amazon.com/blogs/machine-learning/segment-data-source-for-amazon-personalize/).
Learn more

The following resources provide additional information about Amazon Personalize:

- For a quick reference to help you determine if Amazon Personalize fits your use case, see the Amazon Personalize Cheat Sheet in the Amazon Personalize samples repository.
- For a series of videos on how to use Amazon Personalize, see the Amazon Personalize Deep Dive Video Series found on YouTube.
- For in-depth tutorials and code samples, see the amazon-personalize-samples GitHub repository.

For a complete list of partners, see Amazon Personalize Partners.
How it works

Amazon Personalize uses your data to train domain-based or customizable recommendation models. You use a private recommendation API in your application to request real-time recommendations. Amazon Personalize also supports batch workflows get item recommendations and user segments.

How it works

Amazon Personalize workflow summary

The Amazon Personalize workflow is as follows:

1. **Create a dataset group (p. 155)**

   A dataset group is a container for Amazon Personalize resources. The type of dataset group you create determines the resources you can create in step 3 of the Amazon Personalize workflow.
   - With a **Domain dataset group**, you can create recommenders for VIDEO_ON_DEMAND or ECOMMERCE domain use cases. Amazon Personalize manages the configuration, training and updating of these recommenders. If you start with a Domain dataset group, you can still add custom resources.
   - With a **Custom dataset group**, you can create only custom resources. These including solutions, solution versions, and campaigns. For these resources, you have more control over configurations, updates, and retraining.

2. **Prepare and import data (p. 159)**

   You import item, user, and interaction records into **datasets** (Amazon Personalize containers for data). You can import records in bulk or individually. When you import bulk data, you can use Amazon SageMaker Data Wrangler to import data from 40+ sources and prepare it for Amazon Personalize. For more information, see **Preparing and importing data using Amazon SageMaker Data Wrangler** (p. 160).

   After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see **Managing data** (p. 292).
3. **Create domain recommenders or custom resources (p. 210)**

   After you import your data, create domain recommenders (for Domain dataset groups) or custom resources (for Custom dataset group) to train a model on your data. You use these resources to generate recommendations.

4. **Get recommendations (p. 244)**

   Use your recommender or custom campaign to get recommendations. With a Custom dataset group, you can also get batch recommendations or user segments.

   After you complete the Amazon Personalize workflow the first time, keep your data current, and regularly re-train any custom solutions. This allows your model to learn from your user’s most recent activity and sustains and improves the relevance of recommendations. For more information, see [Maintaining recommendation relevance (p. 278)](#).

---

**Amazon Personalize terms**

This section introduces the terms used in Amazon Personalize.

**Topics**

- [Data import and management (p. 6)](#)
- [Training (p. 7)](#)
- [Model deployment and recommendations (p. 9)](#)

**Data import and management**

The following terms relate to importing, exporting, and formatting data in Amazon Personalize.

**contextual metadata**

Interactions data that you collect about a user’s browsing context (such as device used or location) when an event (such as a click) occurs. Contextual metadata can improve recommendation relevance for new and existing users.

**dataset**

A container for data that you upload to Amazon Personalize. There are three types of Amazon Personalize datasets: Users, Items, and Interactions.

**dataset group**

A container for Amazon Personalize resources, including datasets, domain recommenders, and custom resources. A dataset group organizes your resources into independent collections, where resources from one dataset group can’t influence resources in any other dataset group. A dataset group can either be a Domain dataset group or a Custom dataset group.

**Domain dataset group**

A dataset group containing preconfigured resources for different business domains and use cases. Amazon Personalize manages the life cycle of training models and deployment. When you create a Domain dataset group, you choose your business domain, import your data, and create recommenders for each of your use cases. You use your recommender in your application to get recommendations with the GetRecommendations operation.

   If you start with a Domain dataset group, you can still add custom resources such as solutions and solution versions trained with recipes for custom use cases.
Custom dataset group

A dataset group containing only custom resources, including solutions, solution versions, filters, campaigns, and batch inference jobs. You use a campaign to get recommendations with the GetRecommendations operation. You manage the life cycle of training models and deployment. If you start with a Custom dataset group, you can't associate it with a domain later. Instead, create a new Domain dataset group.

dataset export job

A record export tool that outputs the records in a dataset to one or more CSV files in an Amazon S3 bucket. The output CSV file includes a header row with column names that match the fields in the dataset's schema.

dataset import job

A bulk import tool that populates your Amazon Personalize dataset with data from a CSV file in your Amazon S3 bucket.

event

A user action – such as a click, a purchase, or a video viewing – that you record and upload to an Amazon Personalize Interactions dataset. You import events in bulk from a CSV file, incrementally with the Amazon Personalize console, and in real-time.

explicit impressions

A list of items that you manually add to an Amazon Personalize Interactions dataset. Unlike implicit impressions, which Amazon Personalize automatically derives from your recommendation data, you choose what to include in explicit impressions.

implicit impressions

The recommendations that your application shows a user. Unlike explicit impressions, which you manually add to an Interactions dataset, Amazon Personalize automatically derives implicit impressions from your recommendation data.

impressions data

The list of items that you presented to a user when they interacted with a particular item by clicking it, watching it, purchasing it, and so on. Amazon Personalize uses impressions data to calculate the relevance of new items for a user based on how frequently users have selected or ignored the same item.

interactions dataset

A container for historical and real-time data that you collect from interactions between users and items (called events). Interactions data can include impressions data and contextual metadata.

items dataset

A container for metadata about your items, such as price, genre, or availability.

schema

A JSON object in Apache Avro format that tells Amazon Personalize about the structure of your data. Amazon Personalize uses your schema to parse your data.

users dataset

A container for metadata about your users, such as age, gender, or loyalty membership.

Training

The following terms relate to training a model in Amazon Personalize.
item-to-item similarities (SIMS) recipe

A RELATED_ITEMS recipe that uses the data from an Interactions dataset to make recommendations for items that are similar to a specified item. The SIMS recipe calculates similarity based on the way users interact with items instead of matching item metadata, such as price or color.

item-affinity

A USER_SEGMENTATION recipe that uses the data from an Interactions dataset and Items dataset to create user segments for each item that you specify based on the likelihood that the users will interact with the item.

item-attribute-affinity

A USER_SEGMENTATION recipe that uses the data from an Interactions dataset and Items dataset to create a user segment for each item attribute that you specify based on the likelihood that the users will interact with items with the attribute.

personalized-ranking recipe

A PERSONALIZED_RANKING recipe that ranks a collection of items that you provide based on the predicted interest level for a specific user. Use the personalized-ranking recipe to personalize the order of curated lists of items or search results that are personalized for a specific user.

popularity-count recipe

A USER_PERSONALIZATION recipe that recommends the items that have had the most interactions with unique users.

recommender

A Domain dataset group tool that generates recommendations. You create a recommender for a Domain dataset group and use in your application to get real-time recommendations with the GetRecommendations API. When you create a recommender, you specify a use case and Amazon Personalize trains the models backing the recommender with the best configurations for the use case.

recipe

An Amazon Personalize algorithm that is preconfigured to predict the items that a user will interact with (for USER_PERSONALIZATION recipes), or calculate items that are similar to specific items that a user has shown interest in (for RELATED_ITEMS recipes), or rank a collection of items that you provide based on the predicted interest for a specific user (for PERSONALIZED_RANKING recipes).

solution

The recipe, customized parameters, and trained models (Solution Versions) that Amazon Personalize uses to generate recommendations.

solution version

A trained model that you create as part of a solution in Amazon Personalize. You deploy a solution version in a campaign to activate the personalization API that you use to request recommendations.

training mode

The scope of training to be performed when creating a solution version. There are two different modes: FULL and UPDATE. FULL mode creates a completely new solution version based on the entirety of the training data from the datasets in your dataset group. UPDATE incrementally updates the existing solution version to recommend new items that you added since the last training.

Note

With User-Personalization, Amazon Personalize automatically updates the latest solution version trained with FULL training mode. See Automatic updates (p. 116).
user-personalization recipe

A Hierarchical Recurrent Neural Network (HRNN) based USER_PERSONALIZATION recipe that predicts the items that a user will interact with. The user-personalization recipe can use item exploration and impressions data to generate recommendations for new items.

Model deployment and recommendations

The following terms relate to deploying and using a model in Amazon Personalize.

batch inference job

A tool that imports your batch input data from an Amazon S3 bucket, uses your solution version to generate recommendations, and exports the recommendations to an Amazon S3 bucket. We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). Use a batch inference job to get recommendations for large datasets that do not require real-time updates.

batch segment job

A tool that imports your batch input data from an Amazon S3 bucket, uses your solution version to create user segments, and exports the user segments to an Amazon S3 bucket. We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). Use a batch segment job with a solution backed by a USER_SEGMENTATION recipe to create segments of users based on the likelihood the user will interact with different items or items with different item attributes.

campaign

A deployed solution version (trained model) with provisioned dedicated transaction capacity for creating real-time recommendations for your application users. After you create a campaign, you use the getRecommendations or getPersonalizedRanking API operations to get recommendations.

item exploration

With exploration, recommendations include some items that would be typically less likely to be recommended for the user, such as new items, items with few interactions, or items less relevant for the user based on their previous behavior.

recommendations

A list of items that Amazon Personalize predicts a user will interact with. Depending on the Amazon Personalize recipe used, recommendations can be either a list of items (USER_PERSONALIZATION recipes and RELATED_ITEMS recipes), or a ranking of a collection of items you provided (PERSONALIZED_RANKING recipes).

user segments

Lists of user that Amazon Personalize predicts a user will interact with your catalogue. Depending on the USER_SEGMENTATION recipe used, you create user segments based on items (Item-Affinity recipe) item metadata (Item-Attribute-Affinity recipe). You create user segments with a batch segment job.

Types of data Amazon Personalize can use

The following topics introduce the different types of data that you can import into Amazon Personalize.

Topics
Interactions data

An interaction is an event that you record and then import as training data. Amazon Personalize generates recommendations primarily based on the interactions data. Interactions data can include the following:

- Event type and event value data
- Contextual metadata
- Impressions data

You import interactions data into an Interactions dataset. For more details about interactions datasets, see Interactions datasets (p. 77).

Item data

The item metadata that Amazon Personalize can use includes the following:

- Numerical data about each item, such as its price.
- Categorical metadata about each item, such as the item’s genre or color.
- Creation timestamp data for each item.
- Unstructured text metadata, such as product descriptions or movie synopses.

You import metadata about your items into an Items dataset. For more information about Items datasets, see Items datasets (p. 80).

User data

The user metadata Amazon Personalize can use includes the following:

- Numerical data about each user, such as their age.
- Categorical metadata about each user, such as their gender or loyalty membership status.

You import metadata about your users into a Users dataset. For more information about Users datasets, see Users datasets (p. 80).
Setting up Amazon Personalize

Before using Amazon Personalize, you must have an Amazon Web Services (AWS) account with an administrative user. After you set up the required permissions, you can access Amazon Personalize through the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), or the AWS SDKs.

Topics
- Sign up for an AWS account (p. 11)
- Create an administrative user (p. 11)
- Regions and endpoints (p. 12)
- Setting up permissions (p. 12)
- Setting up the AWS CLI (p. 21)
- Setting up the AWS SDKs (p. 22)

Sign up for an AWS account

If you do not have an AWS account, complete the following steps to create one.

To sign up for an AWS account

2. Follow the online instructions.
   
   Part of the sign-up procedure involves receiving a phone call and entering a verification code on the phone keypad.
   
   When you sign up for an AWS account, an AWS account root user is created. The root user has access to all AWS services and resources in the account. As a security best practice, assign administrative access to an administrative user, and use only the root user to perform tasks that require root user access.
   
   AWS sends you a confirmation email after the sign-up process is complete. At any time, you can view your current account activity and manage your account by going to https://aws.amazon.com/ and choosing My Account.

Create an administrative user

After you sign up for an AWS account, create an administrative user so that you don't use the root user for everyday tasks.

Secure your AWS account root user

1. Sign in to the AWS Management Console as the account owner by choosing Root user and entering your AWS account email address. On the next page, enter your password.
   
   For help signing in by using root user, see Signing in as the root user in the AWS Sign-In User Guide.

2. Turn on multi-factor authentication (MFA) for your root user.
   
   For instructions, see Enable a virtual MFA device for your AWS account root user (console) in the IAM User Guide.
Create an administrative user

- For your daily administrative tasks, grant administrative access to an administrative user in AWS IAM Identity Center.

For instructions, see Getting started in the AWS IAM Identity Center User Guide.

Sign in as the administrative user

- To sign in with your IAM Identity Center user, use the sign-in URL that was sent to your email address when you created the IAM Identity Center user.

For help signing in using an IAM Identity Center user, see Signing in to the AWS access portal in the AWS Sign-In User Guide.

Regions and endpoints

An endpoint is a URL that is the entry point for a web service. Each endpoint is associated with a specific AWS region. Pay attention to the default regions of the Amazon Personalize console, the AWS CLI, and the Amazon Personalize SDKs, as all Amazon Personalize components of a given campaign (dataset, solution, campaign, event tracker) must be created in the same region. For the regions and endpoints supported by Amazon Personalize, see Regions and endpoints.

Setting up permissions

You must give users, groups, or roles permission to interact with Amazon Personalize resources. And you must give Amazon Personalize permission to access the resources you create in Amazon Personalize and to perform tasks on your behalf.

To set up permissions

1. Give your users, groups, or roles permission to interact with Amazon Personalize resources and pass a role to Amazon Personalize. See Giving users permission to access Amazon Personalize (p. 13).
2. Give Amazon Personalize permission to access your resources in Amazon Personalize and permission to perform tasks on your behalf. See Giving Amazon Personalize permission to access your resources (p. 14).
3. Modify your Amazon Personalize service role's trust policy so it prevents the confused deputy problem (p. 398). For a trust relationship policy example, see Cross-service confused deputy prevention (p. 398). For information modifying a role's trust policy, see Modifying a role.
4. If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).
5. Complete the steps in Giving Amazon Personalize access to Amazon S3 resources (p. 16) to use IAM and Amazon S3 bucket policies to give Amazon Personalize access to your Amazon S3 resources.

Topics

- Giving users permission to access Amazon Personalize (p. 13)
- Giving Amazon Personalize permission to access your resources (p. 14)
- Giving Amazon Personalize access to Amazon S3 resources (p. 16)
- Giving Amazon Personalize permission to use your AWS KMS key (p. 20)
Giving users permission to access Amazon Personalize

To provide your users access to Amazon Personalize, you create an IAM policy that grants permission to access your Amazon Personalize resources and pass a role to Amazon Personalize. Then you use that policy when you add permissions to your users, groups or roles.

Creating a new IAM policy for your users

Create an IAM policy that provides Amazon Personalize full access to your Amazon Personalize resources.

To use the JSON policy editor to create a policy

1. Sign in to the AWS Management Console and open the IAM console at https://console.aws.amazon.com/iam/.
2. In the navigation pane on the left, choose Policies.
   - If this is your first time choosing Policies, the Welcome to Managed Policies page appears. Choose Get Started.
3. At the top of the page, choose Create policy.
4. In the Policy editor section, choose the JSON option.
5. Enter the following JSON policy document:

   ```json
   {
     "Version": "2012-10-17",
     "Statement": [
       {
         "Effect": "Allow",
         "Action": [
           "personalize:*"
         ],
         "Resource": "*"
       },
       {
         "Effect": "Allow",
         "Action": [
           "iam:PassRole"
         ],
         "Resource": "*",
         "Condition": {
           "StringEquals": {
             "iam:PassedToService": "personalize.amazonaws.com"
           }
         }
       }
     ]
   }
   ```
6. Choose Next.

   Note
   - You can switch between the Visual and JSON editor options anytime. However, if you make changes or choose Next in the Visual editor, IAM might restructure your policy to optimize it for the visual editor. For more information, see Policy restructuring in the IAM User Guide.
7. On the Review and create page, enter a Policy name and a Description (optional) for the policy that you are creating. Review Permissions defined in this policy to see the permissions that are granted by your policy.
8. Choose **Create policy** to save your new policy.

To grant only the permissions required to perform a task in Amazon Personalize, modify the preceding policy to include only the required actions for your user. For a complete list of Amazon Personalize actions, see *Actions, resources, and condition keys for Amazon Personalize*.

### Providing access to Amazon Personalize

Attach the new IAM policy when you provide permissions to your users.

To provide access, add permissions to your users, groups, or roles:

- **Users and groups in AWS IAM Identity Center:**
  
  Create a permission set. Follow the instructions in *Create a permission set* in the *AWS IAM Identity Center User Guide*.

- **Users managed in IAM through an identity provider:**
  
  Create a role for identity federation. Follow the instructions in *Creating a role for a third-party identity provider (federation)* in the *IAM User Guide*.

- **IAM users:**
  
  - Create a role that your user can assume. Follow the instructions in *Creating a role for an IAM user* in the *IAM User Guide*.
  
  - (Not recommended) Attach a policy directly to a user or add a user to a user group. Follow the instructions in *Adding permissions to a user (console)* in the *IAM User Guide*.

### Giving Amazon Personalize permission to access your resources

To give Amazon Personalize permission to access your resources, you create an IAM policy that provides Amazon Personalize full access to your Amazon Personalize resources. Or you can use the AWS managed `AmazonPersonalizeFullAccess` policy. `AmazonPersonalizeFullAccess` provides more permissions than are necessary. We recommend creating a new IAM policy that only grants the necessary permissions. For more information about managed policies, see *AWS managed policies (p. 400)*.

After you create a policy, you create an IAM role for Amazon Personalize and attach the new policy to it.

**Topics**

- *Creating a new IAM policy for Amazon Personalize (p. 14)*
- *Creating an IAM role for Amazon Personalize (p. 15)*

### Creating a new IAM policy for Amazon Personalize

Create an IAM policy that provides Amazon Personalize full access to your Amazon Personalize resources.

**To use the JSON policy editor to create a policy**

1. Sign in to the AWS Management Console and open the IAM console at [https://console.aws.amazon.com/iam/](https://console.aws.amazon.com/iam/).
2. In the navigation pane on the left, choose **Policies**.

   If this is your first time choosing **Policies**, the **Welcome to Managed Policies** page appears. Choose **Get Started**.
3. At the top of the page, choose **Create policy**.
4. In the **Policy editor** section, choose the **JSON** option.
5. Enter the following JSON policy document:

   ```json
   {
   "Version": "2012-10-17",
   "Statement": [
   {
   "Effect": "Allow",
   "Action": [
   "personalize:*"
   ],
   "Resource": "*"
   }
   ]
   }
   ```
6. Choose **Next**.

   **Note**
   You can switch between the **Visual** and **JSON** editor options anytime. However, if you make changes or choose **Next** in the **Visual** editor, IAM might restructure your policy to optimize it for the visual editor. For more information, see **Policy restructuring** in the **IAM User Guide**.
7. On the **Review and create** page, enter a **Policy name** and a **Description** (optional) for the policy that you are creating. Review **Permissions defined in this policy** to see the permissions that are granted by your policy.
8. Choose **Create policy** to save your new policy.

### Creating an IAM role for Amazon Personalize

To use Amazon Personalize, you must create an AWS Identity and Access Management service role for Amazon Personalize. A service role is an **IAM role** that a service assumes to perform actions on your behalf. An IAM administrator can create, modify, and delete a service role from within IAM. For more information, see **Creating a role to delegate permissions to an AWS service** in the **IAM User Guide**. After you create a service role for Amazon Personalize, grant the role additional permissions listed in **Additional service role permissions** (p. 16) as necessary.

**To create the service role for Amazon Personalize (IAM console)**

1. Sign in to the AWS Management Console and open the IAM console at [https://console.aws.amazon.com/iam/](https://console.aws.amazon.com/iam/).
2. In the navigation pane of the IAM console, choose **Roles**, and then choose **Create role**.
3. Choose the **AWS service** role type, and then choose Amazon Personalize.
4. Choose the Personalize use case. Then, choose **Next**.
5. Choose the policy that you created in the previous procedure.
6. (Optional) Set a **permissions boundary**. This is an advanced feature that is available for service roles, but not service-linked roles.

   Expand the **Permissions boundary** section and choose **Use a permissions boundary to control the maximum role permissions**. IAM includes a list of the AWS managed and customer managed policies in your account. Select the policy to use for the permissions boundary or choose **Create policy** to open a new browser tab and create a new policy from scratch. For more information, see **Creating IAM policies** in the **IAM User Guide**. After you create the policy, close that tab and return to your original tab to select the policy to use for the permissions boundary.
7. Choose **Next**.
8. If possible, enter a role name or role name suffix to help you identify the purpose of this role. Role names must be unique within your AWS account, so don't create roles named both PRODROLE and prodrole. When a role name is used in a policy or as part of an ARN, the role name is case sensitive. When a role name appears to customers in the console, such as during the sign-in process, the role name is case insensitive. Because various entities might reference the role, you can't edit the name of the role after it is created.

9. (Optional) For Description, enter a description for the new role.

10. Choose Edit in the Step 1: Select trusted entities or Step 2: Select permissions sections to edit the use cases and permissions for the role.

11. (Optional) Add metadata to the role by attaching tags as key-value pairs. For more information about using tags in IAM, see Tagging IAM resources in the IAM User Guide.

12. Review the role and then choose Create role.

After you create a role for Amazon Personalize, you are ready to grant it access to your Amazon S3 bucket (p. 16) and any AWS KMS keys (p. 20).

Additional service role permissions

After you create the role and grant it permissions to access your resources in Amazon Personalize, do the following:

1. Modify your Amazon Personalize service role's trust policy so it prevents the confused deputy problem (p. 398). For a trust relationship policy example, see Cross-service confused deputy prevention (p. 398). For information modifying a role's trust policy, see Modifying a role.

2. If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

Giving Amazon Personalize access to Amazon S3 resources

To give Amazon Personalize access to your Amazon S3 bucket, do the following:

1. If you haven't already, follow the steps in Setting up permissions (p. 12) to set up permissions so Amazon Personalize can access your resources in Amazon Personalize on your behalf.

2. Attach a policy to the Amazon Personalize service role (see Creating an IAM role for Amazon Personalize (p. 15)) that allows access to your Amazon S3 bucket. For more information, see Attaching an Amazon S3 policy to your Amazon Personalize service role (p. 17).

3. Attach a bucket policy to the Amazon S3 bucket containing your data files so Amazon Personalize can access them. For more information, see Attaching an Amazon Personalize access policy to your Amazon S3 bucket (p. 18).

4. If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

Note

Because Amazon Personalize doesn't communicate with AWS VPCs, Amazon Personalize can't interact with Amazon S3 buckets that allow only VPC access.

Topics

• Attaching an Amazon S3 policy to your Amazon Personalize service role (p. 17)
Attaching an Amazon S3 policy to your Amazon Personalize service role

To attach an Amazon S3 policy to your Amazon Personalize role do the following:

1. Sign in to the IAM console (https://console.aws.amazon.com/iam/).
2. In the navigation pane, choose Policies, and choose Create policy.
3. Choose the JSON tab, and update the policy as follows. Replace bucket-name with the name of your bucket. If you are using a batch workflow, Amazon Personalize needs additional permissions. See Service role policy for batch workflows (p. 17).

   ```json
   {
     "Version": "2012-10-17",
     "Id": "PersonalizeS3BucketAccessPolicy",
     "Statement": [
       {
         "Sid": "PersonalizeS3BucketAccessPolicy",
         "Effect": "Allow",
         "Action": [
           "s3:GetObject",
           "s3:ListBucket"
         ],
         "Resource": [
           "arn:aws:s3:::bucket-name",
           "arn:aws:s3:::bucket-name/*"
         ]
       }
     ]
   }
   ```

5. Give the policy a name.
6. (Optional) For Description, enter a short sentence describing this policy, for example, Allow Amazon Personalize to access its Amazon S3 bucket.
7. Choose Create policy.
8. In the navigation pane, choose Roles, and choose the role you created for Amazon Personalize. See Creating an IAM role for Amazon Personalize (p. 15).
9. For Permissions, choose Attach policies.
10. To display the policy in the list, type part of the policy name in the Filter policies filter box.
11. Choose the check box next to the policy you created earlier in this procedure.
12. Choose Attach policy.

Before your role is ready for use with Amazon Personalize you must also attach a bucket policy to the Amazon S3 bucket containing your data. See Attaching an Amazon Personalize access policy to your Amazon S3 bucket (p. 18).

Service role policy for batch workflows

To complete a batch workflow, Amazon Personalize needs permission to access and add files to your Amazon S3 bucket. Follow the steps above to attach the following policy to your Amazon Personalize role. Replace bucket-name with the name of your bucket. For more information on batch workflows, see Batch recommendations and user segments (custom resources) (p. 259).
Giving Amazon Personalize access to Amazon S3 resources

Service role policy for exporting a dataset

To export a dataset, your Amazon Personalize service role needs permission to use the PutObject and ListBucket Actions on your Amazon S3 bucket. The following example policy grants Amazon Personalize PutObject and ListBucket permissions. Replace bucket-name with the name of your bucket and attach the policy to your service role for Amazon Personalize. For information about attaching policies to a service role see Attaching an Amazon S3 policy to your Amazon Personalize service role (p. 17).

```
{
   "Version": "2012-10-17",
   "Id": "PersonalizeS3BucketAccessPolicy",
   "Statement": [
      {
         "Sid": "PersonalizeS3BucketAccessPolicy",
         "Effect": "Allow",
         "Action": ["s3:PutObject", "s3:ListBucket",
                     "s3:GetObject", "s3:ListBucket", "s3:PutObject"
                    ],
         "Resource": ["arn:aws:s3:::bucket-name",
                       "arn:aws:s3:::bucket-name/*"
                      ]
      }
   ]
}
```

Attaching an Amazon Personalize access policy to your Amazon S3 bucket

Amazon Personalize needs permission to access the S3 bucket. For non-batch workflows, attach the following policy to your bucket. Replace bucket-name with the name of your bucket. For batch workflows, see Amazon S3 bucket policy for batch workflows (p. 19).

For more information on Amazon S3 bucket policies, see How Do I Add an S3 Bucket Policy?

```
{
   "Version": "2012-10-17",
   "Id": "PersonalizeS3BucketAccessPolicy",
   "Statement": [
      {
         "Sid": "PersonalizeS3BucketAccessPolicy",
         "Effect": "Allow",
         "Action": ["s3:PutObject", "s3:ListBucket",
                     "s3:GetObject", "s3:ListBucket", "s3:PutObject"
                    ],
         "Resource": ["arn:aws:s3:::bucket-name",
                       "arn:aws:s3:::bucket-name/*"
                      ]
      }
   ]
}
```
Amazon S3 bucket policy for batch workflows

For batch workflows, Amazon Personalize needs permission to access and add files to your Amazon S3 bucket. Attach the following policy to your bucket. Replace $bucket-name$ with the name of your bucket.

For more information on adding an Amazon S3 bucket policy to a bucket, see [How Do I Add an S3 Bucket Policy?](#). For more information on batch workflows, see [Batch recommendations and user segments](#) (custom resources) (p. 259).

```json
{
  "Version": "2012-10-17",
  "Id": "PersonalizeS3BucketAccessPolicy",
  "Statement": [
    {
      "Sid": "PersonalizeS3BucketAccessPolicy",
      "Effect": "Allow",
      "Principal": {
        "Service": "personalize.amazonaws.com"
      },
      "Action": [
        "s3:GetObject",
        "s3:ListBucket",
        "s3:PutObject"
      ],
      "Resource": [
        "arn:aws:s3:::$bucket-name$",
        "arn:aws:s3:::$bucket-name$/*"
      ]
    }
  ]
}
```

Amazon S3 bucket policy for exporting a dataset

To export a dataset, Amazon Personalize needs permission to use the PutObject and ListBucket Actions on your Amazon S3 bucket. The following example policy grants the Amazon Personalize principle PutObject and ListBucket permissions. Replace $bucket-name$ with the name of your bucket and attach the policy to your bucket. For information on adding an Amazon S3 bucket policy to a bucket, see [How Do I Add an S3 Bucket Policy?](#) in the Amazon Simple Storage Service User Guide.

```json
{
  "Version": "2012-10-17",
  "Id": "PersonalizeS3BucketAccessPolicy",
  "Statement": [
    {
      "Sid": "PersonalizeS3BucketAccessPolicy",
      "Effect": "Allow",
      "Principal": {
        "Service": "personalize.amazonaws.com"
      },
      "Action": [
        "s3:GetObject",
        "s3:PutObject",
        "s3:ListBucket"
      ],
      "Resource": [
        "arn:aws:s3:::$bucket-name$",
        "arn:aws:s3:::$bucket-name$/*"
      ]
    }
  ]
}
```
Giving Amazon Personalize permission to use your AWS KMS key

If you specify an AWS Key Management Service (AWS KMS) key when you use the Amazon Personalize console or APIs, or if you use your AWS KMS key to encrypt an Amazon S3 bucket, you must grant Amazon Personalize permission to use your key. To grant permissions, your AWS KMS key policy and IAM policy attached to your service role must grant Amazon Personalize permission to use your key. This applies for creating the following in Amazon Personalize.

- Dataset groups
- Dataset import job (only AWS KMS key policy must grant permissions)
- Dataset export jobs
- Batch inference jobs
- Batch segment jobs
- Metric attributions

Your AWS KMS key policy and IAM policies must grant permissions for the following actions:

- Decrypt
- GenerateDataKey
- DescribeKey
- CreateGrant (only required in key policy)
- ListGrants

Revoking AWS KMS key permissions after creating a resource can lead to issues when creating a filter or getting recommendations. For more information about AWS KMS policies, see Using key policies in AWS KMS in the AWS Key Management Service Developer Guide. For information on creating an IAM policy, see Creating IAM policies in the IAM User Guide. For information on attaching an IAM policy to role, see Adding and removing IAM identity permissions in the IAM User Guide.

Topics

- Key policy example (p. 21)
- IAM policy example (p. 21)
Key policy example

The following key policy example grants Amazon Personalize and your role the minimum permissions for the preceding Amazon Personalize operations. If you specify a key when you create a dataset group and want to export data from a dataset, your key policy must include the GenerateDataKeyWithoutPlaintext action.

```json
{
  "Version": "2012-10-17",
  "Id": "key-policy-123",
  "Statement": [
    {
      "Sid": "Allow use of the key",
      "Effect": "Allow",
      "Principal": {
        "AWS": "arn:aws:iam::<account-id>:role/<personalize-role-name>",
        "Service": "personalize.amazonaws.com"
      },
      "Action": [
        "kms:Decrypt",
        "kms:GenerateDataKey",
        "kms:DescribeKey",
        "kms:CreateGrant",
        "kms:ListGrants"
      ],
      "Resource": "*"
    }
  ]
}
```

IAM policy example

The following IAM policy example grants a role the minimum AWS KMS permissions required for the preceding Amazon Personalize operations. For dataset import jobs, only the AWS KMS key policy needs to grant permissions.

```json
{
  "Version": "2012-10-17",
  "Statement": [
    {
      "Effect": "Allow",
      "Action": [
        "kms:Decrypt",
        "kms:GenerateDataKey",
        "kms:DescribeKey",
        "kms:CreateGrant",
        "kms:ListGrants"
      ],
      "Resource": "*"
    }
  ]
}
```

Setting up the AWS CLI

The AWS Command Line Interface (AWS CLI) is a unified developer tool for managing AWS services, including Amazon Personalize. We recommend that you install it.

1. To install the AWS CLI, follow the instructions in Installing the AWS Command Line Interface in the AWS Command Line Interface Interface User Guide.
2. To configure the AWS CLI and set up a profile to call the AWS CLI, follow the instructions in Configuring the AWS CLI in the AWS Command Line Interface User Guide.

3. To confirm that the AWS CLI profile is configured properly, run the following command.

   ```bash
   aws configure --profile default
   ```

   If your profile has been configured correctly, you will see output similar to the following.

   ```
   AWS Access Key ID [***************52FQ]:
   AWS Secret Access Key [***************xgyZ]:
   Default region name [us-west-2]:
   Default output format [json]:
   ```

4. To verify that the AWS CLI is configured for use with Amazon Personalize, run the following commands.

   ```bash
   aws personalize help
   ```

   and

   ```bash
   aws personalize-runtime help
   ```

   and

   ```bash
   aws personalize-events help
   ```

   If the AWS CLI is configured correctly, you will see a list of the supported AWS CLI commands for Amazon Personalize, Amazon Personalize runtime, and Amazon Personalize events.

   If you set up the AWS CLI and it doesn't recognize the commands for Amazon Personalize, update the AWS CLI. To update the AWS CLI, run the following command.

   ```bash
   pip3 install awscli --upgrade --user
   ```

   For more information, see Installing the AWS CLI using pip.

---

## Setting up the AWS SDKs

Download and install the AWS SDKs that you want to use. This guide provides examples for SDK for Python (Boto3), SDK for Java 2.x, and SDK for JavaScript v3. For information about other AWS SDKs, see Tools for Amazon Web Services. For information about setting up Amplify, see AWS Amplify.

- **AWS SDK for Python (Boto3)**

  To install the SDK for Python (Boto3), follow the Quickstart instructions in the Boto3 documentation.

- **SDK for Java 2.x**

  To learn about setting up the SDK for Java 2.x, see the Get started with the SDK for Java 2.x topic in the AWS SDK for Java 2.x Developer Guide.

  For code examples for Amazon Personalize, see Amazon Personalize Java code samples in the AWS SDK examples repository.

- **AWS SDK for JavaScript v3**
To learn about setting up the SDK for JavaScript v3, see the Get started with the AWS SDK for JavaScript topic in the AWS SDK for JavaScript Developer Guide.

For code examples for Amazon Personalize, see Amazon Personalize code examples for SDK for JavaScript v3 in the AWS SDK examples repository.
Getting started

The following sections help you get started using Amazon Personalize with the Amazon Personalize console, AWS CLI, and AWS SDKs. The tutorials use historical data that consists of 100,000 movie ratings on 9,700 movies from 600 users.

To simplify tutorials:

- We use a small dataset. This might negatively impact any metrics generated by resources. The tutorials serve as an introduction to the Amazon Personalize workflow and won't necessarily generate the highest performing models.
- We create only an Interactions dataset, and rely on the fact that a user saw a movie and not on what they rated the movie. This simplifies the preparation of the training data.
- We don't record live user interaction events. For information on capturing user events, see Recording events (p. 280).

You can choose to get started with a Domain dataset group or a Custom dataset group:

- Domain dataset groups provide resources that are optimized for different use cases based on your domain. To get started creating a Domain dataset group, complete the Getting started prerequisites (p. 24) and then complete the tutorial in Getting started with a Domain dataset group (p. 26).
- Custom dataset groups allow you to create and configure only custom resources. To get started providing personalized movie recommendations for your users with a custom resources and the User-Personalization recipe, complete the Getting started prerequisites (p. 24) and then start the tutorials in Getting started with a Custom dataset group (p. 47).

When you finish the getting started exercise, to avoid incurring unnecessary charges, follow the steps in Cleaning up resources (p. 74) to delete the resources you created.

Topics

- Getting started prerequisites (p. 24)
- Getting started with a Domain dataset group (p. 26)
- Getting started with a Custom dataset group (p. 47)
- Cleaning up resources (p. 74)

Getting started prerequisites

The following steps are prerequisites for the getting started exercises.

1. Set up permissions so Amazon Personalize can access your resources on your behalf. This involves creating a service role for Amazon Personalize and granting it access to Amazon Personalize resources with an IAM policy. For more information see Giving Amazon Personalize permission to access your resources (p. 14).
2. Prepare your training data and upload the data to your Amazon S3 bucket:
Creating the training data (Domain dataset group)

To create training data, download, modify, and save the movie ratings data to an Amazon Simple Storage Service (Amazon S3) bucket. Then give Amazon Personalize permission to read from the bucket.

To create the training data


2. Open the `ratings.csv` file. This file contains the interactions data for this tutorial.
   a. Delete the `rating` column.
   b. Rename the `userId` and `movieId` columns to `USER_ID` and `ITEM_ID` respectively.
   c. Add an `EVENT_TYPE` column set the value for every record to `watch`. If you're using Microsoft Excel, you can set the `EVENT_TYPE` for every record by entering `watch` in the first cell in the column and then double-clicking the bottom-right corner of the cell. Your header should be the following:

```
USER_ID,ITEM_ID,TIMESTAMP,EVENT_TYPE
```

These columns must be exactly as shown for Amazon Personalize to recognize the data. The first few rows of your data should look as follows:

```
1,1,964982703,watch
1,3,964981247,watch
1,6,964982224,watch
1,47,964983815,watch
1,50,964982931,watch
....
```

Save the `ratings.csv` file.

3. Upload `ratings.csv` to your Amazon S3 bucket. For more information, see Uploading files and folders by using drag and drop in the Amazon Simple Storage Service User Guide.

4. Give Amazon Personalize permission to read the data in the bucket. For more information, see Giving Amazon Personalize access to Amazon S3 resources (p. 16).

Creating the training data (Custom dataset group)

To create training data, download, modify, and save the movie ratings data to an Amazon Simple Storage Service (Amazon S3) bucket. Then give Amazon Personalize permission to read from the bucket.

2. Open the ratings.csv file. This file contains the interactions data for this tutorial.
   a. Delete the rating column.
   b. Replace the header row with the following:
      
      USER_ID, ITEM_ID, TIMESTAMP
      
      These headers must be exactly as shown for Amazon Personalize to recognize the data.

      Save the ratings.csv file.

3. Upload ratings.csv to your Amazon S3 bucket. For more information, see Uploading files and folders by using drag and drop in the Amazon Simple Storage Service User Guide.

4. Give Amazon Personalize permission to read the data in the bucket. For more information, see Giving Amazon Personalize access to Amazon S3 resources (p. 16).

Getting started with a Domain dataset group

In this getting started tutorial you create a Domain dataset group for the VIDEO_ON_DEMAND domain, import interactions data from a CSV file, and create a recommender with the Top picks for you use case. Then you use the recommender to get personalized movie recommendations for a user. The tutorial uses historical data that consists of 100,000 movie ratings on 9,700 movies from 600 users.

To begin, complete the Getting started prerequisites (p. 24) and then depending on how you want to create Amazon Personalize resources, proceed to Getting started with a Domain dataset group (console) (p. 26), Getting started with a Domain dataset group (SDK for Python (Boto3)) (p. 37), Getting started with a Domain dataset group (SDK for Java 2.x) (p. 31), or Getting started with a Domain dataset group (SDK for JavaScript v3) (p. 41).

When you finish the getting started exercise, to avoid incurring unnecessary charges, follow the steps in Cleaning up resources (p. 74) to delete the resources you created.

Topics
- Getting started with a Domain dataset group (console) (p. 26)
- Getting started with a Domain dataset group (SDK for Java 2.x) (p. 31)
- Getting started with a Domain dataset group (SDK for Python (Boto3)) (p. 37)
- Getting started with a Domain dataset group (SDK for JavaScript v3) (p. 41)

Getting started with a Domain dataset group (console)

In this exercise, you use the Amazon Personalize console to create a Domain dataset group and a recommender that returns movie recommendations for a given user.

Before you start this exercise, review the Getting started prerequisites (p. 24).

When you finish the getting started exercise, to avoid incurring unnecessary charges, follow the steps in Cleaning up resources (p. 74) to delete the resources you created.
Step 1: Create a Domain dataset group

In this procedure you create Domain dataset group for the VIDEO_ON_DEMAND domain, create an Interactions dataset with the default schema for the VIDEO_ON_DEMAND domain, and import the interactions data you created in Creating the training data (Domain dataset group) (p. 25).

To create a Domain dataset group

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. In the navigation pane, choose Create dataset group.
3. In Dataset group details, specify a name for your dataset group.
4. For Domain, choose Video on demand. The domain you choose determines the default schema you use when importing data. It also determines what use cases are available for recommenders. Your screen should look similar to the following.

5. Choose Create dataset group. The Overview page appears. Proceed to Step 2: Import data (p. 27).

Step 2: Import data

In this procedure you create an Interactions dataset with the default VIDEO_ON_DEMAND domain schema. Then you import the interactions data you created in Creating the training data (Domain dataset group) (p. 25).

To import data

1. On the Overview page, choose Import interactions data.
2. Choose Import data directly into Amazon Personalize datasets and choose Next.
3. On the Create interactions dataset page, for Dataset name provide a name for your Interactions dataset.
4. For Dataset schema, choose Create a new domain schema by modifying the existing default schema for your domain and enter a name for the schema. The Schema definition updates to display the default schema for the VIDEO_ON_DEMAND domain. Leave the schema unchanged. Your screen should look similar to the following.
5. Choose **Create dataset and continue**.

6. On the **Import interactions data** page, leave the **Data import source** unchanged as **Import data from S3**.

7. For **Dataset import job name**, give your import job a name.

8. For **Data location**, specify where your data is stored in Amazon Simple Storage Service (S3). Use the following syntax:

   `s3://<name of your S3 bucket>/<folder path>/<CSV filename>`

9. In **IAM role**, for **IAM service role** choose **Enter a custom IAM role ARN** and enter the Amazon Resource Name (ARN) of the role you created in Creating an IAM role for Amazon Personalize (p. 15). Your screen should look similar to the following.
10. Choose **Import data** to import data. The **Overview** page for your Domain dataset group appears. Note the status of the import in the **Set up datasets** section. When the status is **Interaction data active** proceed to **Step 3: Create a recommender** (p. 29).

**Step 3: Create a recommender**

In this procedure, you create a recommender for the **Top picks for you** use case for the VIDEO_ON_DEMAND domain.

**To create a recommender**

1. On the **Overview** page for your Domain dataset group, on the middle card, choose the **Use video on demand recommenders** tab and choose **Create recommenders**.

2. On the **Create recommenders** page, choose **Top picks for you** and provide a **Recommender name**. Leave the remaining fields unchanged. Your screen should appear similar to the following.
3. Choose **Create recommenders** to create your recommender.

   You can monitor the status of each recommender on the **Recommenders** page. When your recommender status is Active, you can use it to get recommendations in **Step 4: Get recommendations** (p. 30).

### Step 4: Get recommendations

In this procedure you use the recommender that you created in the previous step to get recommendations.

**To get recommendations**

1. On the Overview page for your Domain dataset group, in the navigation pane choose **Recommenders**.
2. On the **Recommenders** page, choose your recommender.
3. At the top right, choose **Test recommender**.
4. In **Recommendation parameters**, enter a user ID. Leave the other fields unchanged.
5. Choose **Get recommendations**. A table containing the user’s top 25 recommended items appears. Your screen should look similar to the following.
This tutorial shows you how to use the SDK for Java 2.x to create a Domain dataset group for the VIDEO_ON_DEMAND domain. In this tutorial, you create a recommender for the Top picks for you use case.

To avoid incurring unnecessary charges, when you finish the getting started exercise see Cleaning up resources (p. 74) for information on deleting the resources you create in the tutorial.

Prerequisites

The following are prerequisite steps for completing this tutorial:

- Complete the Getting started prerequisites (p. 24) to set up the required permissions and create the training data. If you also completed the Getting started with a Domain dataset group (console) (p. 26), you can reuse the same source data. If you are using your own source data, make sure that your data is formatted like in the prerequisites.
- Set up your SDK for Java 2.x environment and AWS credentials as specified in the Setting up the AWS SDK for Java 2.x procedure in the AWS SDK for Java 2.x Developer Guide.

Tutorial

In the following steps, you set up your project to use Amazon Personalize packages and create Amazon Personalize SDK for Java 2.x clients. Then you import data, create a recommender for the Top picks for you use case, and get recommendations.
Step 1: Set up your project to use Amazon Personalize packages

After you complete the prerequisites, add Amazon Personalize dependencies to your pom.xml file and import Amazon Personalize packages.

1. Add the following dependencies to your pom.xml file. The latest version numbers may be different than the example code.

```xml
<dependency>
  <groupId>software.amazon.awssdk</groupId>
  <artifactId>personalize</artifactId>
  <version>2.16.83</version>
</dependency>

<dependency>
  <groupId>software.amazon.awssdk</groupId>
  <artifactId>personalizeruntime</artifactId>
  <version>2.16.83</version>
</dependency>

<dependency>
  <groupId>software.amazon.awssdk</groupId>
  <artifactId>personalizeevents</artifactId>
  <version>2.16.83</version>
</dependency>
```

2. Add the following import statements to your project.

```java
// import client packages
import software.amazon.awssdk.services.personalize.PersonalizeClient;
import software.amazon.awssdk.services.personalizeruntime.PersonalizeRuntimeClient;
// Amazon Personalize exception package
import software.amazon.awssdk.services.personalize.model.PersonalizeException;
// schema packages
import software.amazon.awssdk.services.personalize.model.CreateSchemaRequest;
// dataset group packages
import software.amazon.awssdk.services.personalize.model.CreateDatasetGroupRequest;
import software.amazon.awssdk.services.personalize.model.DescribeDatasetGroupRequest;
// dataset packages
import software.amazon.awssdk.services.personalize.model.CreateDatasetRequest;
// dataset import job packages
import software.amazon.awssdk.services.personalize.model.CreateDatasetImportJobRequest;
import software.amazon.awssdk.services.personalize.model.DataSource;
import software.amazon.awssdk.services.personalize.model.DatasetImportJob;
import software.amazon.awssdk.services.personalize.model.DescribeDatasetImportJobRequest;
// recommender packages
import software.amazon.awssdk.services.personalize.model.CreateRecommenderRequest;
import software.amazon.awssdk.services.personalize.model.CreateRecommenderResponse;
import software.amazon.awssdk.services.personalize.model.DescribeRecommenderRequest;
// get recommendations packages
import software.amazon.awssdk.services.personalizeruntime.model.GetRecommendationsRequest;
import software.amazon.awssdk.services.personalizeruntime.model.GetRecommendationsResponse;
import software.amazon.awssdk.services.personalizeruntime.model.PredictedItem;
// Java time utility package
import java.time.Instant;
```

Step 2: Create Amazon Personalize clients

After you add Amazon Personalize dependencies to your pom.xml file and import the required packages, create the following Amazon Personalize clients:
Step 3: Import data

After you initialize your Amazon Personalize clients, import the historical data you created when you completed the Getting started prerequisites (p. 24). To import historical data into Amazon Personalize, do the following:

1. Save the following Avro schema as a JSON file in your working directory. This schema matches the columns in the CSV file that you created when you completed the Creating the training data (Domain dataset group) (p. 25).

```json
{
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "USER_ID",
            "type": "string"
        },
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "EVENT_TYPE",
            "type": "string"
        },
        {
            "name": "TIMESTAMP",
            "type": "long"
        }
    ],
    "version": "1.0"
}
```

2. Use the following `createDomainSchema` method to create a domain schema in Amazon Personalize. Pass the following as parameters: an Amazon Personalize service client, the name for your schema, VIDEO_ON_DEMAND for the domain, and the file path for the schema JSON file that you created in the previous step. The method returns the Amazon Resource Name (ARN) of your new schema. Store it for later use.

```java
public static String createDomainSchema(PersonalizeClient personalizeClient, String schemaName, String domain, String filePath) {
    String schema = null;
    try {
        schema = new String(Files.readAllBytes(Paths.get(filePath)));
    } catch (IOException e) {
        System.out.println(e.getMessage());
    }

    try {
        CreateSchemaRequest createSchemaRequest = CreateSchemaRequest.builder()
            .name(schemaName)
            ...
3. Create a dataset group. Use the following createDomainDatasetGroup method to create a domain dataset group. Pass the following as parameters: an Amazon Personalize service client, a name for the dataset group, and pass VIDEO_ON_DEMAND for the domain. The method returns the ARN of your new dataset group. Store it for later use.

```java
public static String createDomainDatasetGroup(PersonalizeClient personalizeClient, String datasetGroupName, String domain) {
    try {
        CreateDatasetGroupRequest createDatasetGroupRequest = CreateDatasetGroupRequest.builder()
            .name(datasetGroupName)
            .domain(domain)
            .build();
        String datasetGroupArn = personalizeClient.createDatasetGroup(createDatasetGroupRequest).datasetGroupArn();
        return datasetGroupArn;
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
    return "";
}
```

4. Create an Interactions dataset. Use the following createDataset method to create an Interactions dataset. Pass the following as parameters: an Amazon Personalize service client, the name for your dataset, your schema's ARN, your dataset group's ARN, and Interactions for the dataset type. The method returns the ARN of your new dataset. Store it for later use.

```java
public static String createDataset(PersonalizeClient personalizeClient, String datasetName, String datasetGroupArn, String datasetType, String schemaArn) {
    try {
        CreateDatasetRequest request = CreateDatasetRequest.builder()
            .name(datasetName)
            .datasetGroupArn(datasetGroupArn)
            .datasetType(datasetType)
            .schemaArn(schemaArn)
            .build();
        String datasetArn = personalizeClient.createDataset(request).datasetArn();
        System.out.println("Dataset " + datasetName + " created.");
        return datasetArn;
    }
```
5. Import your data with a dataset import job. Use the following `createPersonalizeDatasetImportJob` method to create a dataset import job.

Pass the following as parameters: an Amazon Personalize service client, a name for the job, and your Interactions dataset’s ARN. Pass the Amazon S3 bucket path (`s3://bucket name/folder name/ratings.csv`) where you stored the training data, and your service role’s ARN. You created this role as part of the Getting started prerequisites (p. 24). The method returns the ARN of your dataset import job. Optionally store it for later use.

```java
public static String createPersonalizeDatasetImportJob(PersonalizeClient personalizeClient,
                                                     String jobName,
                                                     String datasetArn,
                                                     String s3BucketPath,
                                                     String roleArn) {

    long waitInMilliseconds = 60 * 1000;
    String status;
    String datasetImportJobArn;

    try {
        DataSource importDataSource = DataSource.builder()
            .dataLocation(s3BucketPath)
            .build();

        CreateDatasetImportJobRequest createDatasetImportJobRequest =
            CreateDatasetImportJobRequest.builder()
            .datasetArn(datasetArn)
            .dataSource(importDataSource)
            .jobName(jobName)
            .roleArn(roleArn)
            .build();

        datasetImportJobArn =
            personalizeClient.createDatasetImportJob(createDatasetImportJobRequest)
            .datasetImportJobArn();

        DescribeDatasetImportJobRequest describeDatasetImportJobRequest =
            DescribeDatasetImportJobRequest.builder()
            .datasetImportJobArn(datasetImportJobArn)
            .build();

        DatasetImportJob datasetImportJob =
            personalizeClient.describeDatasetImportJob(describeDatasetImportJobRequest)
            .datasetImportJob();

        status = datasetImportJob.status();
        System.out.println("Dataset import job status: " + status);

        if (status.equals("ACTIVE") || status.equals("CREATE FAILED")) {
            break;
        }
    }
```
Step 4: Create a recommender

After your dataset import job completes, you are ready to create a recommender. Use the following `createRecommender` method to create a recommender. Pass the following as parameters: an Amazon Personalize service client, a name for the recommender, your dataset group's Amazon Resource Name (ARN), and `arn:aws:personalize:::recipe/aws-vod-top-picks` for the recipe ARN. The method returns the ARN of your new recommender. Store it for later use.

```java
public static String createRecommender(PersonalizeClient personalizeClient,
                                       String name,
                                       String datasetGroupArn,
                                       String recipeArn) {

  long maxTime = 0;
  long waitInMilliseconds = 30 * 1000; // 30 seconds
  String recommenderStatus = "";

  try {
    CreateRecommenderRequest createRecommenderRequest =
        CreateRecommenderRequest.builder()
          .datasetGroupArn(datasetGroupArn)
          .name(name)
          .recipeArn(recipeArn)
          .build();

    CreateRecommenderResponse recommenderResponse =
        personalizeClient.createRecommender(createRecommenderRequest);
    String recommenderArn = recommenderResponse.recommenderArn();
    System.out.println("The recommender ARN is " + recommenderArn);

    DescribeRecommenderRequest describeRecommenderRequest =
        DescribeRecommenderRequest.builder()
          .recommenderArn(recommenderArn)
          .build();

    maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
    while (Instant.now().getEpochSecond() < maxTime) {
      recommenderStatus =
          personalizeClient.describeRecommender(describeRecommenderRequest).recommender().status();
      System.out.println("Recommender status: " + recommenderStatus);
      if (recommenderStatus.equals("ACTIVE") ||
          recommenderStatus.equals("CREATE FAILED")) {
        break;
      }
      try {
        Thread.sleep(waitInMilliseconds);
      } catch (InterruptedException e) {
```
Getting started with a Domain dataset group (SDK for Python (Boto3))

This tutorial shows you how to use the SDK for Python (Boto3) to create a Domain dataset group for the VIDEO_ON_DEMAND domain. In this tutorial, you create a recommender for the Top picks for you use case.

To avoid incurring unnecessary charges, when you finish this getting started exercise delete the resources you create in this tutorial. For more information, see Cleaning up resources (p. 74).

Topics
- Prerequisites (p. 38)
- Tutorial (p. 38)
- Getting started using Amazon Personalize APIs with Jupyter (iPython) notebooks (p. 41)
Prequisites

The following are prerequisite steps for using the Python examples in this guide:

- Complete the Getting started prerequisites (p. 24) to set up the required permissions and create the training data. If you are using your own source data, make sure that your data is formatted like in the prerequisites.
- Set up your AWS SDK for Python (Boto3) environment as specified in Setting up the AWS SDKs (p. 22).

Tutorial

In the following steps, you verify your environment and create SDK for Python (Boto3) clients for Amazon Personalize. Then you import data, create a recommender for the Top picks for you use case, and get recommendations.

Step 1: Verify your Python environment and create boto3 clients

After you complete the prerequisites, run the following Python example to confirm that your environment is configured correctly. This code also creates the Amazon Personalize boto3 clients that you use in this tutorial. If your environment is configured correctly, a list of the available recipes is displayed and you can run the other examples in this tutorial.

```python
import boto3
personalizeRt = boto3.client('personalize-runtime')
personalize = boto3.client('personalize')
response = personalize.list_recipes()
for recipe in response['recipes']:
    print(recipe)
```

Step 2: Import data

After you create Amazon Personalize boto3 clients and verify your environment, import the historical data you created when you completed the Getting started prerequisites (p. 24). To import historical data into Amazon Personalize, do the following:

1. Use the following code to create a schema in Amazon Personalize. Replace gs-domain-interactions-schema with a name for the schema.

```python
import json
schema = {
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "USER_ID",
            "type": "string"
        },
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "EVENT_TYPE",
            "type": "string"
        }
    ]
}
```
2. Create a dataset group with the following code. Replace dataset group name with a name for the dataset group.

```python
response = personalize.create_dataset_group(
    name = 'dataset group name',
    domain = 'VIDEO_ON_DEMAND'
)
dsg_arn = response['datasetGroupArn']
description = personalize.describe_dataset_group(datasetGroupArn = dsg_arn)
print('Name: ' + description['name'])
print('ARN: ' + description['datasetGroupArn'])
print('Status: ' + description['status'])
```

3. Create an Interactions dataset in your new dataset group with the following code. Give the dataset a name and provide the schema_arn and dataset_group_arn from the previous steps.

```python
response = personalize.create_dataset(
    name = 'interactions-dataset-name',
    schemaArn = interactions_schema_arn,
    datasetGroupArn = dsg_arn,
    datasetType = 'INTERACTIONS'
)
dataset_arn = response['datasetArn']
```

4. Import your data with a dataset import job with the following code. The code uses the describe_dataset_import_job method to track the status of the job.

```python
import time
response = personalize.create_dataset_import_job(
    jobName = 'JobName',
    datasetArn = 'dataset_arn',
    dataSource = {'dataLocation': 's3://bucket/file.csv'},
    roleArn = 'role_arn'
)
```
**Step 4: Create a recommender**

After your dataset import job completes, you are ready create a recommender. Use the following code to create a recommender. Pass the following as parameters: a name for the recommender, your dataset group's Amazon Resource Name (ARN), and `arn:aws:personalize::::recipe/aws-vod-top-picks` for the recipe ARN. The code uses the `describe_recommender` method to track the status of the recommender.

```python
import time
create_recommender_response = personalize.create_recommender(
    name = 'gs-python-top-picks',
    recipeArn = 'arn:aws:personalize::::recipe/aws-vod-top-picks',
    datasetGroupArn = dsg_arn
)
recommender_arn = create_recommender_response['recommenderArn']
print('Recommender ARN:' + recommender_arn)
max_time = time.time() + 3*60*60 # 3 hours
while time.time() < max_time:
    version_response = personalize.describe_recommender(
        recommenderArn = recommender_arn
    )
    status = version_response['recommender']['status']
    if status == "ACTIVE":
        print("Creation succeeded for {}".format(recommender_arn))
    elif status == "CREATE FAILED":
        print("Creation failed for {}".format(recommender_arn))
    else:
        print("Recommender creation is still in progress")
    time.sleep(60)
```
Step 5: Get recommendations

After you create a recommender, you use it to get recommendations with the following code. Pass as parameters the Amazon Resource Name (ARN) of the recommender you created in the previous step, and a user ID (for example, 123). The method prints the list of recommended items.

```python
response = personalizeRt.get_recommendations(
    userId = '123'
)  
print("Recommended items")
for item in response['itemList']:
    print (item['itemId'])
```

Getting started using Amazon Personalize APIs with Jupyter (iPython) notebooks

To get started creating Domain dataset groups with Jupyter notebooks, clone or download a series of notebooks found in the notebooksManagedDomains folder of the Amazon Personalize samples repository. The notebooks walk you through importing training data, creating a recommender, and getting recommendations with Amazon Personalize.

Note

Before starting with the notebooks, make sure to build your environment following the steps in the README.md

Getting started with a Domain dataset group (SDK for JavaScript v3)

This tutorial shows you how to use the AWS SDK for JavaScript v3 to create a Domain dataset group for the VIDEO_ON_DEMAND domain. In this tutorial, you create a recommender for the Top picks for you use case.

To view the code used in this tutorial on GitHub, see Amazon Personalize code examples for SDK for JavaScript v3 in the AWS SDK Code Examples repository.

To avoid incurring unnecessary charges, when you finish this getting started exercise delete the resources you create in this tutorial. For more information, see Cleaning up resources (p. 74).

Topics

- Prerequisites (p. 41)
- Tutorial (p. 42)

Prerequisites

The following are prerequisite steps for completing this tutorial:

- Complete the Getting started prerequisites (p. 24) to set up the required permissions and create the training data. If you also completed the Getting started with a Domain dataset group (console) (p. 26), you can reuse the same source data. If you are using your own source data, make sure that your data is formatted like in the prerequisites.
- Set up the SDK for JavaScript and AWS credentials as specified in the Setting up the SDK for JavaScript procedure in the AWS SDK for JavaScript Developer Guide.
Tutorial

In the following steps, you install the required dependencies. Then you create a dataset group, import data, create a recommender for the Top picks for you use case, and get recommendations.

If you use Node.js, you can run each code sample by saving the sample as a JavaScript file and then running node <fileName.js>.

Step 1: Install Amazon Personalize dependencies

After you complete the prerequisites, install the following Amazon Personalize dependencies:

- @aws-sdk/client-personalize
- @aws-sdk/client-personalize-runtime
- @aws-sdk/client-personalize-events (optional for this tutorial, but required if you want to record events (p. 280) after you create your recommender)

The following is an example of a package.json file you can use. To install the dependencies with Node.js, navigate to where you saved the package.json file and run npm install.

```json
{
  "name": "personalize-js-project",
  "version": "1.0.0",
  "description": "personalize operations",
  "type": "module",
  "author": "Author Name <email@address.com>",
  "license": "ISC",
  "dependencies": {
    "@aws-sdk/client-personalize": "^3.350.0",
    "@aws-sdk/client-personalize-runtime": "^3.350.0",
    "@aws-sdk/client-personalize-events": "^3.350.0",
    "fs": "^0.0.1-security"
  },
  "compilerOptions": {
    "resolveJsonModule": true,
    "esModuleInterop": true
  }
}
```

Step 2: Create Amazon Personalize clients

After you install the dependencies, create your Amazon Personalize clients. In this tutorial, the code samples assume you create the clients in a file named personalizeClients.js stored in a directory named libs.

The following is an example of a personalizeClient.js file.

```javascript
import { PersonalizeClient } from '@aws-sdk/client-personalize';
import { PersonalizeRuntimeClient } from '@aws-sdk/client-personalize-runtime';
import { PersonalizeEventsClient } from '@aws-sdk/client-personalize-events';
// Set your AWS region.
const REGION = "region"; // e.g. "us-east-1"

const personalizeClient = new PersonalizeClient({ region: REGION });
const personalizeEventsClient = new PersonalizeEventsClient({ region: REGION });
const personalizeRuntimeClient = new PersonalizeRuntimeClient({ region: REGION });
export { personalizeClient, personalizeEventsClient, personalizeRuntimeClient };
```
Step 3: Import data

After you create your Amazon Personalize clients, import the historical data you created when you completed the Getting started prerequisites (p. 24). To import historical data into Amazon Personalize, do the following:

1. Save the following Avro schema as a JSON file in your working directory. This schema matches the columns in the CSV file that you created when you completed the Creating the training data (Domain dataset group) (p. 25).

```json
{
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "USER_ID",
            "type": "string"
        },
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "EVENT_TYPE",
            "type": "string"
        },
        {
            "name": "TIMESTAMP",
            "type": "long"
        }
    ],
    "version": "1.0"
}
```

2. Create a domain schema in Amazon Personalize with the following createDomainSchema.js code. Replace SCHEMA_PATH with the path to the schema.json file you just created. Update the createSchemaParam to specify a name for the schema, and for domain specify VIDEO_ON_DEMAND.

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateSchemaCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

import fs from 'fs';

let schemaFilePath = "SCHEMA_PATH";
let mySchema = "";

try {
    mySchema = fs.readFileSync(schemaFilePath).toString();
} catch (err) {
    mySchema = 'TEST' // for unit tests.
}

// Set the domain schema parameters.
export const createDomainSchemaParam = {
    name: 'NAME', /* required */
    schema: mySchema, /* required */
    domain: 'VIDEO_ON_DEMAND'
}
```
3. Create a domain dataset group in Amazon Personalize with the following `createDomainDatasetGroup.js` code. Update the `domainDatasetGroupParams` to specify a name for the dataset group, and for `domain` specify VIDEO_ON_DEMAND.

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateDatasetGroupCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the domain dataset group parameters.
export const domainDatasetGroupParams = {
    name: 'NAME', /* required */
    domain: 'DOMAIN' /* required for a domain dsg, specify ECOMMERCE or VIDEO_ON_DEMAND */
};

export const run = async () => {
    try {
        const response = await personalizeClient.send(new CreateDatasetGroupCommand(domainDatasetGroupParams));
        console.log("Success", response);
        return response; // For unit tests.
    } catch (err) {
        console.log("Error", err);
    }
};
run();
```

4. Create an Interactions dataset in Amazon Personalize with the following `createDataset.js` code. Update the `createDatasetParam` to specify the Amazon Resource Name (ARN) of the dataset group and schema you just created, give the dataset a name, and for `datasetType`, specify Interactions.

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateDatasetCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the dataset's parameters.
export const createDatasetParam = {
    datasetGroupArn: 'DATASET_GROUP_ARN', /* required */
    datasetType: 'DATASET_TYPE', /* required */
    name: 'NAME', /* required */
};
```

import {CreateDatasetCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';

// Get service clients module and commands using ES6 syntax.

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the dataset import job parameters.
export const datasetImportJobParam = {
  datasetArn: 'DATASET_ARN', /* required */
  dataSource: { /* required */
    dataLocation: 'S3_PATH'
  },
  jobName: 'NAME',/* required */
  roleArn: 'ROLE_ARN' /* required */
}

export const run = async () => {
  try {
    const response = await personalizeClient.send(new
      CreateDatasetImportJobCommand(datasetImportJobParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();

Step 4: Create a recommender

After your dataset import job completes, you are ready create a recommender. To create a recommender, use the following createRecommender.js code. Update the createRecommenderParam with the following: Specify a name for the recommender, specify your dataset group's ARN, and for recipeArn specify arn:aws:personalize:::recipe/aws-vod-top-picks.
// Get service clients module and commands using ES6 syntax.
import { CreateRecommenderCommand } from "@aws-sdk/client-personalize";
import { personalizeClient } from ".libs/personalizeClients.js";

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the recommender's parameters.
export const createRecommenderParam = {
  name: 'NAME', /* required */
  recipeArn: 'RECIPE_ARN', /* required */
  datasetGroupArn: 'DATASET_GROUP_ARN' /* required */
}

export const run = async () => {
  try {
    const response = await personalizeClient.send(new CreateRecommenderCommand(createRecommenderParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();

Step 5: Get recommendations

After you create a recommender, you use it to get recommendations. Use the following getRecommendations.js code to get recommendations for a user. Update the getRecommendationsParam to specify the ARN of the recommender you created in the previous step, and specify a user ID (for example, 123).

// Get service clients module and commands using ES6 syntax.
import { GetRecommendationsCommand } from "@aws-sdk/client-personalize-runtime";
import { personalizeRuntimeClient } from ".libs/personalizeClients.js";

// Or, create the client here.
// const personalizeRuntimeClient = new PersonalizeRuntimeClient({ region: "REGION"});

// Set the recommendation request parameters.
export const getRecommendationsParam = {
  recommenderArn: 'RECOMMENDER_ARN', /* required */
  userId: 'USER_ID', /* required */
  numResults: 15 /* optional */
}

export const run = async () => {
  try {
    const response = await personalizeRuntimeClient.send(new GetRecommendationsCommand(getRecommendationsParam));
    console.log("Success!", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();
Getting started with a Custom dataset group

This getting started guide shows you how to provide personalized movie recommendations for your users with a Custom dataset group and the User-Personalization recipe. The tutorial uses historical data that consists of 100,000 movie ratings on 9,700 movies from 600 users.

To begin, complete the Getting started prerequisites (p. 24) and then proceed to either Getting started (console) (p. 47), Getting started (AWS CLI) (p. 55), Getting started (SDK for Python (Boto3)) (p. 62), or Getting started (SDK for Java 2.x) (p. 66).

When you finish the getting started exercise, to avoid incurring unnecessary charges, follow the steps in Cleaning up resources (p. 74) to delete the resources you created.

Topics
- Getting started (console) (p. 47)
- Getting started (AWS CLI) (p. 55)
- Getting started (SDK for Python (Boto3)) (p. 62)
- Getting started (SDK for Java 2.x) (p. 66)

Getting started (console)

In this exercise, you use the Amazon Personalize console to create a Custom dataset group with a solution that returns movie recommendations for a given user. Before you start this exercise, review the Getting started prerequisites (p. 24).

When you finish the getting started exercise, to avoid incurring unnecessary charges, follow the steps in Cleaning up resources (p. 74) to delete the resources you created.

Step 1: Create a dataset group and a dataset

In this procedure, you first create a dataset group. Next, you create an Amazon Personalize Interactions dataset in the dataset group.

To create a dataset group and a dataset

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. Choose Create dataset group.
3. In Dataset group details, for Dataset group name, specify a name for your dataset group.
4. For Domain choose Custom. Your screen should look similar to the following:
5. Choose Create dataset group. The Overview page appears.

6. On the Overview page, choose Import interactions data.

7. Choose Import data directly into Amazon Personalize datasets and choose Next.

8. On the Create interactions dataset page, for Dataset name, specify a name for your dataset.

9. For Dataset schema, choose Create new schema. In the Schema fields section, a minimal Interactions schema is displayed. The schema matches the headers you previously added to the ratings.csv file, so you don't need to make any changes. If you haven't created the training data, see Getting started prerequisites (p. 24).

10. For Schema name, specify a name for the new schema. Your screen should look similar to the following:
11. Choose **Create dataset and continue**. The **Import interactions data** page appears. Next, complete **Step 2: Import interactions data (p. 49)** to import interactions data.

**Step 2: Import interactions data**

Now that you have created a dataset, it's time to import interactions data into the dataset.

**To import interactions data**

1. On the **Import interactions data** page, for **Data import source** choose **Import data from S3**.
2. For **Dataset import job name**, specify a name for your import job.
3. In the **Additional S3 bucket policy required** dialog box, if you haven't granted Amazon Personalize permissions, follow the instructions to add the required Amazon S3 bucket policy (p. 16).
4. For **Data location**, specify where your movie data file is stored in Amazon Simple Storage Service (S3). Use the following syntax:

   \[s3://<name of your S3 bucket>/<folder path>/<CSV filename>\]

5. In the **IAM Role** section, for **IAM service role**, keep the default selection of **Enter a custom IAM role ARN**.
6. For **Custom IAM role ARN**, specify the role that you created in *Creating an IAM role for Amazon Personalize (p. 15)*.

The **Dataset import job details** and **IAM role** sections should be similar to the following:

![Dataset import job details and IAM role sections](image)

7. **Choose Finish**. The data import job starts and the **Overview** page is displayed. Initially, the status is **Create pending** (followed by **Create in progress**), and the **Create solution** button is disabled.

The time it takes for the data to be imported depends on the size of the dataset. When the data import job has finished, the status changes to **Active** and the **Create solution** button is enabled. The **Overview** page should look similar to the following:
8. After the import job has finished, choose the Create solution button. The Create solution page is displayed. Now that you have imported data, you are ready to create a solution in Step 3: Create a solution (p. 51).

Step 3: Create a solution

In this procedure, you use the dataset that you imported in Step 2: Import interactions data (p. 49) to train a model. A trained model is referred to as a solution version.

To create a solution

1. On the Overview page for your dataset group, in Use custom resources choose Create solution.
2. For Solution type, choose Item recommendation to get item recommendations for your users.
3. For Solution name, specify a name for your solution.
4. For Solution type choose Item recommendations.
5. For Recipe, choose aws-user-personalization. Leave the optional Solution configuration and Advanced configuration fields unchanged.

Your screen should look similar to the following:
6. Choose Create and train solution. Solution version training starts and the Overview page displays.

7. To find the training status, in the navigation pane expand Custom resources and choose Solutions and recipes.

8. In the Solutions section, choose your solution. The details page for the solution page appears. The Solution versions page lists the status of your model.

When the Solution version status is Active, you are ready to move to Step 4: Create a campaign (p. 52).

Step 4: Create a campaign

In this procedure, you create a campaign, which deploys the solution version you created in the previous step.
To create a campaign

1. In the navigation pane, expand **Custom resources** and choose **Campaigns**.
2. Choose **Create campaign**. The **Create new campaign** page appears.
3. In **Campaign details**, for **Campaign name**, specify a name for your campaign.
4. For **Solution**, choose the solution you created in the previous step and for **Solution version ID** keep the default.
5. For **Minimum provisioned transactions per second**, keep the default of 1. Leave the **Campaign configuration** fields unchanged.

Your screen should look similar to the following:

6. Choose **Create campaign**. Campaign creation starts and the campaign details pages with the **Personalization API** section displayed.

Your screen should look similar to the following:
Creating a campaign can take a couple minutes. After Amazon Personalize finishes creating your campaign, the page is updated to show the **Test campaign results** section. Your screen should look similar to the following:

**Step 5: Get recommendations**

In this procedure, use the campaign that you created in the previous step to get recommendations.

**To get recommendations**

1. In **Test campaign results**, for **User ID**, specify a value from the *ratings* dataset, for example, **83**. For **Filter name** keep the default selection of **None** and leave the **Context** fields empty.
2. Choose **Get recommendations**. The **Recommendations** panel lists the item IDs and scores for the recommended items.

Your screen should look similar to the following:
In this exercise, you use the AWS Command Line Interface (AWS CLI) to explore Amazon Personalize. You create a campaign that returns movie recommendations for a given user ID.

Before you start this exercise, do the following:

- Review the Getting Started Getting started prerequisites (p. 24).
- Set up the AWS CLI, as specified in Setting up the AWS CLI (p. 21).

When you finish the getting started exercise, to avoid incurring unnecessary charges, follow the steps in Cleaning up resources (p. 74) to delete the resources you created.

**Note**

The AWS CLI commands in this exercise were tested on Linux. For information about using the AWS CLI commands on Windows, see Specifying parameter values for the AWS Command Line Interface in the AWS Command Line Interface User Guide.

**Step 1: Import training data**

Follow the steps to create a dataset group, add a dataset to the group, and then populate the dataset using the movie ratings data.

1. Create a dataset group by running the following command. You can encrypt the dataset group by passing a AWS Key Management Service key ARN and the ARN of an IAM role that has access permissions to that key as input parameters. For more information about the API, see CreateDatasetGroup (p. 444).
aws personalize create-dataset-group --name MovieRatingDatasetGroup --kms-key-arn arn:aws:kms:us-west-2:01234567890:key/1682a1e7-a94d-4d92-bbdf-837d3b62315e --role-arn arn:aws:iam::01234567890:KMS-key-access

The dataset group ARN is displayed, for example:

```
{
  "datasetGroupArn": "arn:aws:personalize:us-west-2:acct-id:dataset-group/MovieRatingDatasetGroup"
}
```

Use the describe-dataset-group command to display the dataset group you created, specifying the returned dataset group ARN.

```
aws personalize describe-dataset-group \\
```

The dataset group and its properties are displayed, for example:

```
{
  "datasetGroup": {
    "name": "MovieRatingDatasetGroup",
    "status": "ACTIVE",
    "creationDateTime": 1542392161.262,
    "lastUpdatedDateTime": 1542396513.377
  }
}
```

**Note**
Wait until the dataset group's status shows as ACTIVE before creating a dataset in the group. This operation is usually quick.

If you don't remember the dataset group ARN, use the list-dataset-groups command to display all the dataset groups that you created, along with their ARNs.

```
aws personalize list-dataset-groups
```

**Note**
The describe-object and list-objects commands are available for most Amazon Personalize objects. These commands are not shown in the remainder of this exercise but they are available.

2. Create a schema file in JSON format by saving the following code to a file named MovieRatingSchema.json. The schema matches the headers you previously added to ratings.csv. The schema name is Interactions, which matches one of the three types of datasets recognized by Amazon Personalize. For more information, see [Schemas](p. 82).

```
{
  "type": "record",
  "name": "Interactions",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
```

56
3. Create a schema by running the following command. Specify the file you saved in the previous step. The example shows the file as belonging to the current folder. For more information about the API, see CreateSchema (p. 465).

```bash
aws personalize create-schema \
    --name MovieRatingSchema \
    --schema file://MovieRatingSchema.json
```

The schema Amazon Resource Name (ARN) is displayed, for example:

```json
{
    "schemaArn": "arn:aws:personalize:us-west-2:acct-id:schema/MovieRatingSchema"
}
```

4. Create an empty dataset by running the following command. Provide the dataset group ARN and schema ARN that were returned in the previous steps. The dataset-type must match the schema name from the previous step. For more information about the API, see CreateDataset (p. 436).

```bash
aws personalize create-dataset \
    --name MovieRatingDataset \
    --dataset-group-arn arn:aws:personalize:us-west-2:acct-id:dataset-group/MovieRatingDatasetGroup \
    --dataset-type Interactions \
```

The dataset ARN is displayed, for example:

```json
{
    "datasetArn": "arn:aws:personalize:us-west-2:acct-id:dataset/MovieRatingDatasetGroup/INTERACTIONS"
}
```

5. Add the training data to the dataset.

   a. Create a dataset import job by running the following command. Provide the dataset ARN and Amazon S3 bucket name that were returned in the previous steps. Supply the AWS Identity and Access Management (IAM) role ARN you created in Creating an IAM role for Amazon Personalize (p. 15). For more information about the API, see CreateDatasetImportJob (p. 448).

```bash
aws personalize create-dataset-import-job \
    --job-name MovieRatingImportJob \
    --dataset-arn arn:aws:personalize:us-west-2:acct-id:dataset/MovieRatingDatasetGroup/INTERACTIONS \
    --data-source dataLocation=s3://bucketname/ratings.csv \
```
--role-arn roleArn

The dataset import job ARN is displayed, for example:

```json
{
MovieRatingImportJob"
}
```

b. Check the status by using the `describe-dataset-import-job` command. Provide the dataset import job ARN that was returned in the previous step. For more information about the API, see [DescribeDatasetImportJob](p. 509).

```bash
aws personalize describe-dataset-import-job \
```

The properties of the dataset import job, including its status, are displayed. Initially, the status shows as CREATE PENDING, for example:

```json
{
  "datasetImportJob": {
    "jobName": "MovieRatingImportJob",
MovieRatingImportJob",
MovieRatingDatasetGroup/INTERACTIONS",
    "dataSource": {
      "dataLocation": "s3://<bucketname>/ratings.csv"
    },
    "roleArn": "role-arn",
    "status": "CREATE PENDING",
    "creationDateTime": 1542392161.837,
    "lastUpdatedDateTime": 1542393013.377
  }
}
```

The dataset import is complete when the status shows as ACTIVE. Then you are ready to train the model using the specified dataset.

**Note**

Importing takes time. Wait until the dataset import is complete before training the model using the dataset.

---

### Step 2: Create a solution (train the model)

Two steps are required to initially train a model. First, you create the configuration for training the model using the [CreateSolution](p. 468) operation. Second, you train the model using the [CreateSolutionVersion](p. 473) operation.

You train a model using a recipe and your training data. Amazon Personalize provides a set of predefined recipes. For more information, see [Choosing a recipe](p. 113). For this exercise, you use the User-Personalization recipe.

1. Create the configuration for training a model by running the following command.

```bash
aws personalize create-solution \
--name MovieSolution \
```

---

58
MovieRatingDatasetGroup \
--recipe-arn arn:aws:personalize:::recipe/aws-user-personalization

The solution ARN is displayed, for example:

```
{
  "solutionArn": "arn:aws:personalize:us-west-2:acct-id:solution/MovieSolution"
}
```

2. Check the `create` status using the `describe-solution` command. Provide the solution ARN that was returned in the previous step. For more information about the API, see `DescribeSolution` (p. 526).

```
aws personalize describe-solution \
```

The properties of the solution and the create status are displayed. Initially, the status shows as CREATE PENDING, for example:

```
{
  "solution": {
    "name": "MovieSolution",
    "performHPO": false,
    "performAutoML": false,
    "recipeArn": "arn:aws:personalize:::recipe/aws-user-personalization",
    "solutionConfig": {},
    "status": "ACTIVE",
    "creationDateTime": "2021-05-12T16:27:59.819000-07:00",
    "lastUpdatedDateTime": "2021-05-12T16:27:59.819000-07:00"
  }
}
```

3. When the solution is ACTIVE, train the model by running the following command.

```
aws personalize create-solution-version \
```

The solution version ARN is displayed, for example:

```
{
}
```

Check the `training` status of the solution version by using the `describe-solution-version` command. Provide the solution version ARN that was returned in the previous step. For more information about the API, see `DescribeSolutionVersion` (p. 529).

```
aws personalize describe-solution-version \
```
The properties of the solution version and the training status are displayed. Initially, the status shows as CREATE PENDING, for example:

```json
{
   "solutionVersion": {  
   ......,
   "status": "CREATE PENDING"
   }
}
```

4. When the solution version status is ACTIVE, the training is complete.

Now you can review training metrics and create a campaign using the solution version.

**Note**
Training takes time. Wait until training is complete (the training status of the solution version shows as ACTIVE) before using this version of the solution in a campaign.

5. You can validate the performance of the solution version by reviewing its metrics. Get the metrics for the solution version by running the following command. Provide the solution version ARN that was returned previously. For more information about the API, see GetSolutionMetrics (p. 532).

```
aws personalize get-solution-metrics \  
```

A sample response is shown:

```json
{
   "metrics": {
      "coverage": 0.0485,
      "mean_reciprocal_rank_at_25": 0.0381,
      "normalized_discounted_cumulative_gain_at_10": 0.0363,
      "normalized_discounted_cumulative_gain_at_25": 0.0984,
      "normalized_discounted_cumulative_gain_at_5": 0.0175,
      "precision_at_10": 0.0107,
      "precision_at_25": 0.0207,
      "precision_at_5": 0.0107
   }
}
```

**Step 3: Create a campaign (deploy the solution)**

Before you can get recommendations, you must deploy a solution version. Deploying a solution is also known as creating a campaign. Once you've created your campaign, your client application can get recommendations using the GetRecommendations (p. 608) API.

1. Create a campaign by running the following command. Provide the solution version ARN that was returned in the previous step. For more information about the API, see CreateCampaign (p. 432).

```
aws personalize create-campaign \  
   --name MovieRecommendationCampaign \  
```
A sample response is shown:

```json
{
  "campaignArn": "arn:aws:personalize:us-west-2:acct-id:campaign/
MovieRecommendationCampaign"
}
```

2. Check the deployment status by running the following command. Provide the campaign ARN that was returned in the previous step. For more information about the API, see DescribeCampaign (p. 501).

```bash
aws personalize describe-campaign \
MovieRecommendationCampaign
```

A sample response is shown:

```json
{
  "campaign": {
    "name": "MovieRecommendationCampaign",
    "campaignArn": "arn:aws:personalize:us-west-2:acct-id:campaign/
MovieRecommendationCampaign",
MovieSolution/<version-id>",
    "minProvisionedTPS": "1",
    "creationDateTime": 1543864775.923,
    "lastUpdatedDateTime": 1543864791.923,
    "status": "CREATE_IN_PROGRESS"
  }
}
```

**Note**

Wait until the status shows as ACTIVE before getting recommendations from the campaign.

### Step 4: Get recommendations

Get recommendations by running the get-recommendations command. Provide the campaign ARN that was returned in the previous step. In the request, you specify a user ID from the movie ratings dataset. For more information about the API, see GetRecommendations (p. 608).

**Note**

Not all recipes support the GetRecommendations API. For more information, see Choosing a recipe (p. 113).

The AWS CLI command you call in this step, personalize-runtime, is different than in previous steps.

```bash
aws personalize-runtime get-recommendations \
  --campaign-arn arn:aws:personalize:us-west-2:acct-id:campaign/MovieRecommendationCampaign \
  --user-id 123
```

In response, the campaign returns a list of item recommendations (movie IDs) the user might like. The list is sorted in descending order of relevance for the user.

```json
{
```
Getting started (SDK for Python (Boto3))

This tutorial shows you how to complete the Amazon Personalize workflow from start to finish with the SDK for Python (Boto3).

To avoid incurring unnecessary charges, when you finish this getting started exercise delete the resources you create in this tutorial. For more information, see Cleaning up resources (p. 74).

Topics
- Prerequisites (p. 62)
- Tutorial (p. 62)
- Getting started using Amazon Personalize APIs with Jupyter (iPython) notebooks (p. 66)

Prerequisites

The following are prerequisite steps for using the Python examples in this guide:

- Complete the Getting started prerequisites (p. 24) to set up the required permissions and create the training data. If you are using your own source data, make sure that your data is formatted like the prerequisites.
- Set up your AWS SDK for Python (Boto3) environment as specified in Setting up the AWS SDKs (p. 22).

Tutorial

In the following steps, you verify your environment and create SDK for Python (Boto3) clients for Amazon Personalize. Then you import data, create and deploy a solution version with a campaign, and get recommendations.

Step 1: Verify your Python environment and create boto3 clients

After you complete the prerequisites, run the following boto3 example to confirm that your environment is configured correctly. This code also creates the Amazon Personalize boto3 clients you use in this tutorial. If your environment is configured correctly, a list of the available recipes is displayed, and you can run the other examples in this tutorial.

```python
import boto3
```
Step 2: Import data

After you create Amazon Personalize boto3 clients and verify your environment, import the historical data you created when you completed the Getting started prerequisites (p. 24). To import historical data into Amazon Personalize, do the following:

1. Use the following code to create a schema in Amazon Personalize. Replace `getting-started-schema` with a name for the schema.

```python
import json

schema = {
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "USER_ID",
            "type": "string"
        },
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "TIMESTAMP",
            "type": "long"
        }
    ],
    "version": "1.0"
}

create_interactions_schema_response = personalize.create_schema(
    name='getting-started-schema',
    schema=json.dumps(schema)
)

interactions_schema_arn = create_interactions_schema_response['schemaArn']
print(json.dumps(create_interactions_schema_response, indent=2))
```

2. Create a dataset group with the following code. Replace `dataset group name` with a name for the dataset group.

```python
response = personalize.create_dataset_group(name = 'dataset group name')
dataset_group_arn = response['datasetGroupArn']

description = personalize.describe_dataset_group(datasetGroupArn = dataset_group_arn)[datasetGroup]
print('Name: ' + description['name'])
print('ARN: ' + description['datasetGroupArn'])
print('Status: ' + description['status'])
```

3. Create an Interactions dataset in your new dataset group with the following code. Give the dataset a name and provide the schema_arn and dataset_group_arn from the previous steps.

```python
response = personalize.create_dataset(name = 'dataset name',
    datasetGroupArn = dataset_group_arn,
    schemaArn = interactions_schema_arn)
```
response = personalize.create_dataset(
    name = 'dataset_name',
    schemaArn = 'schema_arn',
    datasetGroupArn = 'dataset_group_arn',
    datasetType = 'Interactions'
)

dataset_arn = response['datasetArn']

4. Import your data with a dataset import job with the following code. The code uses the describe_dataset_import_job method to track the status of the job.

Pass the following as parameters: a name for the job, the dataset_arn from the previous step, the Amazon S3 bucket path (s3://bucket name/folder name/ratings.csv) where you stored the training data, and your IAM service role's ARN. You created this role as part of the Getting started prerequisites (p. 24). Amazon Personalize needs permission to access the bucket. See Giving Amazon Personalize access to Amazon S3 resources (p. 16).

```python
import time
response = personalize.create_dataset_import_job(
    jobName = 'JobName',
    datasetArn = 'dataset_arn',
    dataSource = {'dataLocation': 's3://bucket/file.csv'},
    roleArn = 'role_arn',
    importMode = 'FULL'
)

dataset_interactions_import_job_arn = response['datasetImportJobArn']

description = personalize.describe_dataset_import_job(
    datasetImportJobArn = dataset_interactions_import_job_arn)['datasetImportJob']

print('Name: ' + description['jobName'])
print('ARN: ' + description['datasetImportJobArn'])
print('Status: ' + description['status'])

max_time = time.time() + 3*60*60 # 3 hours
while time.time() < max_time:
    describe_dataset_import_job_response = personalize.describe_dataset_import_job(
        datasetImportJobArn = dataset_interactions_import_job_arn
    )
    status = describe_dataset_import_job_response['datasetImportJob']['status']
    print("Interactions DatasetImportJob: {}".format(status))
    if status == "ACTIVE" or status == "CREATE FAILED":
        break
    time.sleep(60)
```

Step 3: Create a solution

After importing your data, you create a solution and solution version as follows. The solution contains the configurations to train a model and a solution version is a trained model.

1. Create a new solution with the following code. Pass the following as parameters: the dataset_group_arn from earlier, a name for the solution, and the ARN for the User-Personalization recipe (arn:aws:personalize::recipe/aws-user-personalization). Store the ARN of your new solution for later use.
create_solution_response = personalize.create_solution(
    name='solution name',
    recipeArn = 'arn:aws:personalize:::recipe/aws-user-personalization',
    datasetGroupArn = 'dataset group arn'
)
solution_arn = create_solution_response['solutionArn']
print('solution_arn: ', solution_arn)

2. Create a solution version with the following code. Pass as a parameter the solution_arn from
the previous step. The following code creates a solution version. During training, the code uses the
DescribeSolutionVersion (p. 529) operation to retrieve the solution version's status. When training is
complete, the method returns the ARN of your new solution version. Store it for later use.

import time
import json
create_solution_version_response = personalize.create_solution_version(
    solutionArn = solution_arn
)
solution_version_arn = create_solution_version_response['solutionVersionArn']
print(json.dumps(create_solution_version_response, indent=2))

max_time = time.time() + 3*60*60 # 3 hours
while time.time() < max_time:
    describe_solution_version_response = personalize.describe_solution_version(
        solutionVersionArn = solution_version_arn
    )
    status = describe_solution_version_response['solutionVersion']['status']
    print("SolutionVersion: {}".format(status))
    if status == "ACTIVE" or status == "CREATE FAILED":
        break
    time.sleep(60)

Step 4: Create a campaign

After you create your solution version, deploy it with an Amazon Personalize campaign. Use the
following code to create a campaign that deploys your solution version. Pass the following as
parameters: the solution_version_arn, and a name for the campaign. The method returns the
Amazon Resource Name (ARN) of your new campaign. Store it for later use.

response = personalize.create_campaign(
    name = "campaign name",
    solutionVersionArn = 'solution version arn'
)
arn = response['campaignArn']
description = personalize.describe_campaign(campaignArn = arn)['campaign']
print('Name: ' + description['name'])
print('ARN: ' + description['campaignArn'])
print('Status: ' + description['status'])

Step 5: Get recommendations

After you create a campaign, you can use it to get recommendations. The following code shows how to
get recommendations from a campaign and print out each recommended item's ID. Pass the ARN of the
campaign you created in the previous step. For user ID, you pass the ID of a user that from the training
data, such as 123.

```python
response = personalizeRt.get_recommendations(
    campaignArn = 'Campaign ARN',
    userId = '123',
    numResults = 10
)

print("Recommended items")
for item in response['itemList']:
    print (item['itemId'])
```

## Getting started using Amazon Personalize APIs with Jupyter (iPython) notebooks

To get started using Amazon Personalize using Jupyter notebooks, clone or download a series of
notebooks found in the `getting_started` folder of the Amazon Personalize samples repository. The
notebooks walk you through importing training data, creating a solution, creating a campaign, and
getting recommendations using Amazon Personalize.

**Note**

Before starting with the notebooks, make sure to build your environment following the steps in
the README.md

## Getting started (SDK for Java 2.x)

This tutorial shows you how to complete the Amazon Personalize workflow from start to finish with the
AWS SDK for Java 2.x.

To avoid incurring unnecessary charges, when you finish the getting started exercise follow the steps in
Cleaning up resources (p. 74) to delete the resources you create in the tutorial.

For more examples, see Complete Amazon Personalize project (p. 74).

### Topics

- [Prerequisites (p. 66)](#)
- [Complete Amazon Personalize project (p. 74)](#)

## Prerequisites

The following are prerequisite steps for completing this tutorial:

- Complete the [Getting started prerequisites (p. 24)](#), to set up the required permissions and create
the training data. You can use the same source data used in the Getting started (console) (p. 47) or
Getting started (AWS CLI) (p. 55) exercises. If you are using your own source data, make sure that
your data is formatted like in the prerequisites.
- Set up your SDK for Java 2.x environment and AWS credentials as specified in the Setting up the AWS
SDK for Java 2.x procedure in the AWS SDK for Java 2.x Developer Guide.

## Tutorial

In the following steps you set up your project to use Amazon Personalize packages and create Amazon
Personalize SDK for Java 2.x clients. Then you import data, create and deploy a solution version with a
campaign, and get recommendations.
Step 1: Set up your project to use Amazon Personalize packages

After you complete the prerequisites, add Amazon Personalize dependencies to your pom.xml file and import Amazon Personalize packages.

1. Add the following dependencies to your pom.xml file. The latest version numbers may be different than the example code.

```
<dependency>
  <groupId>software.amazon.awssdk</groupId>
  <artifactId>personalize</artifactId>
  <version>2.16.83</version>
</dependency>
<dependency>
  <groupId>software.amazon.awssdk</groupId>
  <artifactId>personalizeruntime</artifactId>
  <version>2.16.83</version>
</dependency>
<dependency>
  <groupId>software.amazon.awssdk</groupId>
  <artifactId>personalizeevents</artifactId>
  <version>2.16.83</version>
</dependency>
```

2. Add the following import statements to your project.

```
// import client packages
import software.amazon.awssdk.services.personalize.PersonalizeClient;
import software.amazon.awssdk.services.personalizeruntime.PersonalizeRuntimeClient;
// Amazon Personalize exception package
import software.amazon.awssdk.services.personalize.model.PersonalizeException;
// schema packages
import software.amazon.awssdk.services.personalize.model.CreateSchemaRequest;
// dataset group packages
import software.amazon.awssdk.services.personalize.model.CreateDatasetGroupRequest;
import software.amazon.awssdk.services.personalize.model.DescribeDatasetGroupRequest;
// dataset packages
import software.amazon.awssdk.services.personalize.model.CreateDatasetRequest;
import software.amazon.awssdk.services.personalize.model.DataSource;
import software.amazon.awssdk.services.personalize.model.DatasetImportJob;
import software.amazon.awssdk.services.personalize.model.DescribeDatasetImportJobRequest;
// solution packages
import software.amazon.awssdk.services.personalize.model.CreateSolutionRequest;
import software.amazon.awssdk.services.personalize.model.CreateSolutionResponse;
// solution version packages
import software.amazon.awssdk.services.personalize.model.DescribeSolutionRequest;
import software.amazon.awssdk.services.personalize.model.CreateSolutionVersionRequest;
import software.amazon.awssdk.services.personalize.model.CreateSolutionVersionResponse;
import software.amazon.awssdk.services.personalize.model.DescribeSolutionVersionRequest;
// campaign packages
import software.amazon.awssdk.services.personalize.model.CreateCampaignRequest;
import software.amazon.awssdk.services.personalize.model.CreateCampaignResponse;
// get recommendations packages
import software.amazon.awssdk.services.personalizeruntime.model.GetRecommendationsRequest;
import software.amazon.awssdk.services.personalizeruntime.model.GetRecommendationsResponse;
import software.amazon.awssdk.services.personalizeruntime.model.PredictedItem;
// Java time utility package
import java.time.Instant;
```
Step 2: Create Amazon Personalize clients

After you add Amazon Personalize dependencies to your pom.xml file and import the required packages, create the following Amazon Personalize clients:

```java
PersonalizeClient personalizeClient = PersonalizeClient.builder()
     .region(region)
     .build();

PersonalizeRuntimeClient personalizeRuntimeClient = PersonalizeRuntimeClient.builder()
     .region(region)
     .build();
```

Step 3: Import data

After you initialize your Amazon Personalize clients, import the historical data you created when you completed the Getting started prerequisites (p. 24). To import historical data into Amazon Personalize, do the following:

1. Save the following Avro schema as a JSON file in your working directory. This schema matches the columns in the CSV file you created when you completed the Getting started prerequisites (p. 24).

```json
{
   "type": "record",
   "name": "Interactions",
   "namespace": "com.amazonaws.personalize.schema",
   "fields": [
      {
         "name": "USER_ID",
         "type": "string"
      },
      {
         "name": "ITEM_ID",
         "type": "string"
      },
      {
         "name": "TIMESTAMP",
         "type": "long"
      }
   ],
   "version": "1.0"
}
```

2. Use the following `createSchema` method to create a schema in Amazon Personalize. Pass the following as parameters: an Amazon Personalize service client, the name for your schema, and the file path for the schema JSON file you created in the previous step. The method returns the Amazon Resource Name (ARN) of your new schema. Store it for later use.

```java
public static String createSchema(PersonalizeClient personalizeClient, String
schemaName, String filePath) {
    String schema = null;
    try {
        schema = new String(Files.readAllBytes(Paths.get(filePath)));
    } catch (IOException e) {
        System.out.println(e.getMessage());
    }

    try {
        CreateSchemaRequest createSchemaRequest = CreateSchemaRequest.builder()
            .name(schemaName)
            .schema(schema)
```
3. Create a dataset group. Use the following `createDatasetGroup` method to create a dataset group. Pass the following as parameters: an Amazon Personalize service client and the name for the dataset group. The method returns the ARN of your new dataset group. Store it for later use.

```java
public static String createDatasetGroup(PersonalizeClient personalizeClient, String datasetGroupName) {
    try {
        CreateDatasetGroupRequest createDatasetGroupRequest = CreateDatasetGroupRequest.builder()
            .name(datasetGroupName)
            .build();
        return personalizeClient.createDatasetGroup(createDatasetGroupRequest).datasetGroupArn();
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
    }
    return "";
}
```

4. Create an Interactions dataset. Use the following `createDataset` method to create an Interactions dataset. Pass the following as parameters: an Amazon Personalize service client, the name for your dataset, your schema's ARN, your dataset group's ARN, and `Interactions` for the dataset type. The method returns the ARN of your new dataset. Store it for later use.

```java
public static String createDataset(PersonalizeClient personalizeClient, String datasetName, String datasetGroupArn, String datasetType, String schemaArn) {
    try {
        CreateDatasetRequest request = CreateDatasetRequest.builder()
            .name(datasetName)
            .datasetGroupArn(datasetGroupArn)
            .datasetType(datasetType)
            .schemaArn(schemaArn)
            .build();
        String datasetArn = personalizeClient.createDataset(request).datasetArn();
        System.out.println("Dataset " + datasetName + " created.");
        return datasetArn;
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
    }
    return "";
}
```
5. Import your data with a dataset import job. Use the following `createPersonalizeDatasetImportJob` method to create a dataset import job.

Pass the following as parameters: an Amazon Personalize service client, a name for the job, your Interactions dataset's ARN, the Amazon S3 bucket path (`s3://bucket name/folder name/ratings.csv`) where you stored the training data, and your service role's ARN (you created this role as part of the Getting started prerequisites (p. 24)). The method returns the ARN of your dataset import job. Optionally store it for later use.

```java
public static String createPersonalizeDatasetImportJob(PersonalizeClient personalizeClient, String jobName, String datasetArn, String s3BucketPath, String roleArn) {
    long waitInMilliseconds = 60 * 1000;
    String status;
    String datasetImportJobArn;

    try {
        DataSource importDataSource = DataSource.builder()
            .dataLocation(s3BucketPath)
            .build();

        CreateDatasetImportJobRequest createDatasetImportJobRequest = CreateDatasetImportJobRequest.builder()
            .datasetArn(datasetArn)
            .dataSource(importDataSource)
            .jobName(jobName)
            .roleArn(roleArn)
            .build();

        datasetImportJobArn = personalizeClient.createDatasetImportJob(createDatasetImportJobRequest)
            .datasetImportJobArn();

        DescribeDatasetImportJobRequest describeDatasetImportJobRequest = DescribeDatasetImportJobRequest.builder()
            .datasetImportJobArn(datasetImportJobArn)
            .build();

        long maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;

        while (Instant.now().getEpochSecond() < maxTime) {
            DatasetImportJob datasetImportJob = personalizeClient
                .describeDatasetImportJob(describeDatasetImportJobRequest)
                .datasetImportJob();

            status = datasetImportJob.status();
            System.out.println("Dataset import job status: "+ status);

            if (status.equals("ACTIVE") || status.equals("CREATE FAILED")) {
                break;
            }
            try {
                Thread.sleep(waitInMilliseconds);
            } catch (InterruptedException e) {
                Thread.currentThread().interrupt();
                System.out.println(e.getMessage());
            }
        }
    } catch (Exception e) {
        System.out.println(e.getMessage());
    }
    return datasetImportJobArn;
}
```
return datasetImportJobArn;
}
} catch (PersonalizeException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
    return "";
}

---

### Step 4: Create a solution

After you import your data, you create a solution and solution version as follows. The solution contains the configurations to train a model and a solution version is a trained model.

1. Create a new solution with the following createPersonalizeSolution method. Pass the following as parameters: an Amazon Personalize service client, your dataset groups Amazon Resource Name (ARN), a name for the solution, and the ARN for the User-Personalization recipe (arn:aws:personalize:::recipe/aws-user-personalization). The method returns the ARN your new solution. Store it for later use.

```java
public static String createPersonalizeSolution(PersonalizeClient personalizeClient, String datasetGroupArn, String solutionName, String recipeArn) {
    try {
        CreateSolutionRequest solutionRequest = CreateSolutionRequest.builder()
                .name(solutionName)
                .datasetGroupArn(datasetGroupArn)
                .recipeArn(recipeArn)
                .build();

        CreateSolutionResponse solutionResponse = personalizeClient.createSolution(solutionRequest);
        return solutionResponse.solutionArn();
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
    return "";
}
```

2. Create a solution version with the following createPersonalizeSolutionVersion method. Pass as a parameter the ARN of the solution the previous step. The following code first checks to see if your solution is ready and then creates a solution version. During training, the code uses the DescribeSolutionVersion (p. 529) operation to retrieve the solution version's status. When training is complete, the method returns the ARN of your new solution version. Store it for later use.

```java
public static String createPersonalizeSolutionVersion(PersonalizeClient personalizeClient, String solutionArn) {
    long maxTime = 0;
    long waitInMilliseconds = 30 * 1000; // 30 seconds
    String solutionStatus = "";
    String solutionVersionStatus = "";
    String solutionVersionArn = "";

    try {
        DescribeSolutionRequest describeSolutionRequest = DescribeSolutionRequest.builder()
                .solutionArn(solutionArn)
                .build();
```
maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;

// Wait until solution is active.
while (Instant.now().getEpochSecond() < maxTime) {
    solutionStatus = personalizeClient.describeSolution(describeSolutionRequest).solution().status();
    System.out.println("Solution status: " + solutionStatus);
    if (solutionStatus.equals("ACTIVE") || solutionStatus.equals("CREATE FAILED")) {
        break;
    }
    try {
        Thread.sleep(waitInMilliseconds);
    } catch (InterruptedException e) {
        System.out.println(e.getMessage());
    }
}

if (solutionStatus.equals("ACTIVE")) {
    CreateSolutionVersionRequest createSolutionVersionRequest = CreateSolutionVersionRequest.builder()
        .solutionArn(solutionArn)
        .build();
    CreateSolutionVersionResponse createSolutionVersionResponse = personalizeClient.createSolutionVersion(createSolutionVersionRequest);
    solutionVersionArn = createSolutionVersionResponse.solutionVersionArn();
    System.out.println("Solution version ARN: " + solutionVersionArn);
    DescribeSolutionVersionRequest describeSolutionVersionRequest = DescribeSolutionVersionRequest.builder()
        .solutionVersionArn(solutionVersionArn)
        .build();
    while (Instant.now().getEpochSecond() < maxTime) {
        solutionVersionStatus = personalizeClient.describeSolutionVersion(describeSolutionVersionRequest).solutionVersion().status();
        System.out.println("Solution version status: " + solutionVersionStatus);
        if (solutionVersionStatus.equals("ACTIVE") || solutionVersionStatus.equals("CREATE FAILED")) {
            break;
        }
        try {
            Thread.sleep(waitInMilliseconds);
        } catch (InterruptedException e) {
            System.out.println(e.getMessage());
        }
    }
    return solutionVersionArn;
} catch (PersonalizeException e) {
    System.err.println(e.awsErrorDetails().errorMessage());
    System.exit(1);
}
return ";";
For more information, see Creating a solution and a solution version (p. 210). When you create a solution version, you can evaluate its performance before proceeding. For more information, see Evaluating a solution version with metrics (p. 232).

Step 5: Create a campaign

After you train and evaluate your solution version, deploy it with an Amazon Personalize campaign. Use the following createPersonalCampaign method to deploy a solution version. Pass the following as parameters: an Amazon Personalize service client, the Amazon Resource Name (ARN) of the solution version you created in the previous step, and a name for the campaign. The method returns the ARN of your new campaign. Store it for later use.

```java
public static String createPersonalCampaign(PersonalizeClient personalizeClient, String solutionVersionArn, String name) {
    try {
        CreateCampaignRequest createCampaignRequest = CreateCampaignRequest.builder()
                .minProvisionedTPS(1)
                .solutionVersionArn(solutionVersionArn)
                .name(name)
                .build();

        CreateCampaignResponse campaignResponse = personalizeClient.createCampaign(createCampaignRequest);
        System.out.println("The campaign ARN is "+campaignResponse.campaignArn());
        return campaignResponse.campaignArn();
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
}
```

For more information about Amazon Personalize campaigns, see Creating a campaign (p. 237).

Step 6: Get recommendations

After you create a campaign, you use it to get recommendations. Use the following getRecs method to get recommendations for a user. Pass as parameters an Amazon Personalize runtime client, the Amazon Resource Name (ARN) of the campaign you created in the previous step, and a user ID (for example, 123) from the historical data you imported. The method prints the list of recommended items to the screen.

```java
public static void getRecs(PersonalizeRuntimeClient personalizeRuntimeClient, String campaignArn, String userId){
    try {
        GetRecommendationsRequest recommendationsRequest =
        GetRecommendationsRequest.builder()
                .campaignArn(campaignArn)
                .numResults(20)
                .userId(userId)
                .build();

        GetRecommendationsResponse recommendationsResponse =
        personalizeRuntimeClient.getRecommendations(recommendationsRequest);
        List<PredictedItem> items = recommendationsResponse.itemList();
        for (PredictedItem item: items) {
            System.out.println("Item Id is : "+item.itemId());
            System.out.println("Item score is : "+item.score());
        }
    } catch (AwsServiceException e) {
```

73
System.err.println(e.awsErrorDetails().errorMessage());
System.exit(1);
}

Complete Amazon Personalize project

For an all-in-one project that shows you how to complete the Amazon Personalize workflow with the SDK for Java 2.x, see the Amazon-Personalize-Java-App on GitHub. This project includes training multiple solution versions with different recipes, and recording events with the PutEvents operation.

For additional examples, see code the found in the personalize folder of the AWS SDK examples repository.

Cleaning up resources

To avoid incurring unnecessary charges, delete the resources you created after you're done with the getting started exercise. To delete the resources, use either the Amazon Personalize console or the Delete APIs from the SDKs or the AWS Command Line Interface (AWS CLI). For example, use the DeleteCampaign (p. 477) API to delete a campaign.

You can't delete a resource whose status is CREATE PENDING or IN PROGRESS. The resource status must be ACTIVE or CREATE FAILED. Check the status using the Describe APIs, for example, DescribeCampaign (p. 501).

Some resources must be deleted before others, as shown in the following table. This process can take some time.

To delete the training data you uploaded, ratings.csv, see How do I delete objects from an S3 bucket?

Topics
- Cleaning up domain-based resources (p. 74)
- Cleaning up custom resources (p. 75)

Cleaning up domain-based resources

If you created a Domain dataset group, delete resources as follows:

<table>
<thead>
<tr>
<th>Resource to be deleted</th>
<th>Delete this first</th>
<th>Notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>Recommender (p. 701)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>DatasetImportJob (p. 657)</td>
<td></td>
<td>Can't be deleted.</td>
</tr>
<tr>
<td>Dataset (p. 645)</td>
<td></td>
<td>No associated DatasetImportJobs can have a status of CREATE PENDING or IN PROGRESS.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>No associated Recommenders can have a status of CREATE PENDING or IN PROGRESS.</td>
</tr>
</tbody>
</table>
## Cleaning up custom resources

If you created a Custom dataset group, delete resources as follows:

<table>
<thead>
<tr>
<th>Resource to be deleted</th>
<th>Delete this first</th>
<th>Notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>DatasetSchema (p. 662)</td>
<td>All datasets that reference the schema.</td>
<td></td>
</tr>
<tr>
<td>DatasetGroup (p. 653)</td>
<td>All associated recommenders</td>
<td>All datasets in the dataset group.</td>
</tr>
<tr>
<td>DatasetImportJob (p. 657)</td>
<td></td>
<td>Can not be deleted.</td>
</tr>
<tr>
<td>EventTracker (p. 675)</td>
<td></td>
<td>The event-interactions dataset that is associated with the event tracker is not deleted and continues to be used by the solution version.</td>
</tr>
<tr>
<td>Dataset (p. 645)</td>
<td></td>
<td>No associated DatasetImportJob can have a status of CREATE PENDING or IN PROGRESS. No associated SolutionVersion can have a status of CREATE PENDING or IN PROGRESS.</td>
</tr>
<tr>
<td>DatasetSchema (p. 662)</td>
<td>All datasets that reference the schema.</td>
<td></td>
</tr>
<tr>
<td>Campaign (p. 636)</td>
<td></td>
<td>No associated SolutionVersion can have a status of CREATE PENDING or IN PROGRESS.</td>
</tr>
<tr>
<td>Solution (p. 710)</td>
<td>All campaigns based on the solution version.</td>
<td>No associated SolutionVersion can have a status of CREATE PENDING or IN PROGRESS.</td>
</tr>
<tr>
<td>SolutionVersion (p. 717)</td>
<td></td>
<td>Deleted when the associated Solution is deleted.</td>
</tr>
<tr>
<td>DatasetGroup (p. 653)</td>
<td>All associated event trackers.</td>
<td>All associated solutions.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>All datasets in the dataset group.</td>
</tr>
</tbody>
</table>
Datasets and schemas

Amazon Personalize datasets are containers for data. There are three types of datasets:

- **Users (p. 80)** – This dataset stores metadata about your users. This might include information such as age, gender, or loyalty membership.
- **Items (p. 80)** – This dataset stores metadata about your items. This might include information such as price, SKU type, or availability.
- **Interactions (p. 77)** – This dataset stores historical and real-time data from interactions between users and items. In Amazon Personalize, an interaction is an event that you record and then import as training data. For both Domain dataset groups and Custom dataset groups, you must at minimum create an Interactions dataset.

Each dataset group can have only one of each dataset type. Amazon Personalize stores your data in datasets until you delete the datasets. For all use cases (Domain dataset groups) and recipes (Custom dataset groups), your interactions data must have the following:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
- At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

Before you create a dataset, you define a schema for that dataset. A schema tells Amazon Personalize about the structure of your data and allows Amazon Personalize to parse the data. A schema has a name key whose value must match the dataset type. After you create a schema, you can't make changes to the schema.

For Domain dataset groups, each dataset type has a default schema with required fields and reserved keywords. Each time you create a dataset, you can either use the existing domain schema or create a new one by modifying the existing default schema. Use the default schema as a guide for what data to import for your domain. Once you define the schema and create the dataset, you can't make changes to the schema.

If you import data in bulk, your data must be stored in comma-separated values (CSV) format. The first row of your CSV file must contain column headers, which must match your schema. For information about how to format your bulk data for Amazon Personalize, see [Data format guidelines](p. 101).

**Topics**
- Datasets (p. 76)
- Schemas (p. 82)
- Data format guidelines (p. 101)

Datasets

The following topics provide detailed information on Amazon Personalize Interactions, Users, and Items datasets. Each type of dataset has different data requirements. For both Domain dataset groups and Custom dataset groups, your interactions data must have the following before training:
Interactions datasets

An interaction is an event that you record and then import as training data. Amazon Personalize generates recommendations primarily based on the interactions data you import into an Interactions dataset. You can record multiple event types, such as click, watch or like.

For example, if a user clicks a particular item and then likes the item, you can have Amazon Personalize use these events as training data. For each event, you would record the user's ID, the item's ID, the timestamp (in Unix time epoch format), and the event type (click and like). You would then add both interaction events to an Interactions dataset.

For all use cases (Domain dataset groups) and recipes (custom resources), your interactions data must have the following:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
- At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

To create a recommender or a custom solution, you must at minimum create an Interactions dataset. This section provides information about the following types of interactions data you can import into Amazon Personalize.

Topics
- Event type and event value data (p. 77)
- Contextual metadata (p. 78)
- Impressions data (p. 79)

Event type and event value data

Interactions datasets can store event type data, such as click and watch event types, and event value data for each of your events.
• If you create domain recommenders, all use cases require your data to include an EVENT_TYPE field. Some use cases require specific event types. You are free to use additional event types. For more information see Choosing a use case (p. 107).

With domain recommenders, Amazon Personalize does not use event value data.
• If you create custom resources, Amazon Personalize uses event type and event value data to filter events before training. You can import event type data, or event type and event value data. Import this data to choose the interactions data Amazon Personalize uses in training as follows:

  • **Choose events based on event type** – To choose records based on type, record a type for each of your events in an EVENT_TYPE column. When you configure a solution you'll specify the type and Amazon Personalize will use only records with this type in training.

    For example, if your data includes purchase, click, and watch event types, and you want Amazon Personalize to train the model with only watch events, you would include each event's type in an EVENT_TYPE column. Then, when you create a solution, specify watch as the event type that Amazon Personalize uses in training.

    If your Interactions dataset has multiple event types in an EVENT_TYPE column, and you do not provide an event type when you configure your solution, Amazon Personalize uses all interactions data for training with equal weight regardless of type.

  • **Choose records based on type and value** – To choose records based on type and value, record an event type and event value for each event. The value you choose for each event depends on what data you want to exclude and what event types you are recording. For example, you might match the user activity, such as the percentage of video the user watched for watch event types.

    When you configure a solution, you set a specific value as a threshold to exclude records from training. For example, if your EVENT_VALUE data for events with an EVENT_TYPE of watch is the percentage of a video that a user watched, if you set the event value threshold to 0.5, and the event type to watch, Amazon Personalize trains the model using only watch interaction events with an EVENT_VALUE greater than or equal to 0.5.

### Contextual metadata

With certain recipes and recommender use cases, Amazon Personalize can use contextual metadata when identifying underlying patterns that reveal the most relevant items for your users. Contextual metadata is interactions data you collect on the user's environment at the time of an event, such as their location or device type.

Including contextual metadata allows you to provide a more personalized experience for existing users. For example, if customers shop differently when accessing your catalog from a phone compared to a computer, include contextual metadata about the user's device. Recommendations will then be more relevant based on how they are browsing.

Additionally, contextual metadata helps decrease the cold-start phase for new or unidentified users. The cold-start phase refers to the period when your recommendation engine provides less relevant recommendations due to the lack of historical information regarding that user.

For Domain dataset groups, the following recommender use cases can use contextual metadata:

• [Recommended for you](p. 112) (ECOMMERCE domain)
• [Top picks for you](p. 109) (VIDEO_ON_DEMAND domain)

For custom resources, recipes that use contextual metadata include the following:

• [User-Personalization](p. 116)
• [Personalized-Ranking](p. 139)
For more information on contextual information, see the following AWS Machine Learning Blog post: *Increasing the relevance of your Amazon Personalize recommendations by leveraging contextual information*.

**Impressions data**

If you create a Domain dataset group for the VIDEO_ON_DEMAND or ECOMMERCE domain, or use the User-Personalization (p. 116) recipe, Amazon Personalize can model impressions data that you upload to an Interactions dataset. Impressions are lists of items that were visible to a user when they interacted with (for example, clicked or watched) a particular item.

Amazon Personalize uses impressions data to determine what items to include in exploration. With exploration, recommendations include some items that would be typically less likely to be recommended for the user, such as new items, items with few interactions, or items less relevant for the user based on their previous behavior. The more frequently an item occurs in impressions data, the less likely it is that Amazon Personalize includes the item in exploration. Impression values can have at most 1000 characters (including the vertical bar character).

For more information about exploration see *Exploration* (p. 106). Amazon Personalize can model two types of impressions: **Implicit impressions** (p. 79) and **Explicit impressions** (p. 79).

**Implicit impressions**

*Implicit impressions* are the recommendations, retrieved from Amazon Personalize, that you show the user. You can integrate them into your recommendation workflow by including the RecommendationId (returned by the GetRecommendations (p. 608) and GetPersonalizedRanking (p. 604) operations) as input for future PutEvents (p. 598) requests. Amazon Personalize derives the implicit impressions based on your recommendation data.

For example, you might have an application that provides recommendations for streaming video. Your recommendation workflow using implicit impressions might be as follows:

1. You request video recommendations for one of your users using the Amazon Personalize the section called “GetRecommendations” (p. 608) API operation.
2. Amazon Personalize generates recommendations for the user using your model (solution version) and returns them with a recommendationId in the API response.
3. You show the video recommendations to your user in your application.
4. When your user interacts with (for example, clicks) a video, record the choice in a call to the PutEvents (p. 598) API and include the recommendationId as a parameter. For a code sample see Recording impressions data (p. 288).
5. Amazon Personalize uses the recommendationId to derive the impression data from the previous video recommendations, and then uses the impression data to guide exploration, where future recommendations include new videos with less interactions data or relevance.

For more information on recording events with implicit impression data, see Recording impressions data (p. 288).

**Explicit impressions**

*Explicit impressions* are impressions that you manually record and send to Amazon Personalize. Use explicit impressions to manipulate results from Amazon Personalize. The order of the items has no impact.

For example, you might have a shopping application that provides recommendations for shoes. If you only recommend shoes that are currently in stock, you can specify these items using explicit impressions. Your recommendation workflow using explicit impressions might be as follows:
1. You request recommendations for one of your users using the Amazon Personalize "GetRecommendations" API.

2. Amazon Personalize generates recommendations for the user using your model (solution version) and returns them in the API response.

3. You show the user only the recommended shoes that are in stock.

4. For real-time incremental data import, when your user interacts with (for example, clicks) a pair of shoes, you record the choice in a call to the PutEvents API and list the recommended items that are in stock in the impression parameter. For a code sample see Recording impressions data.

   For importing impressions in historical interactions data, you can list explicit impressions in your csv file and separate each item with a '|' character. The vertical bar character counts towards the 1000 character limit. For an example see Formatting explicit impressions.

5. Amazon Personalize uses the impression data to guide exploration, where future recommendations include new shoes with less interactions data or relevance.

### Users datasets

The user data that you can import into Amazon Personalize includes numerical and categorical metadata about your users, such as gender or loyalty membership. You import metadata about your users into an Amazon Personalize Users dataset. The maximum number of metadata columns is 25.

This topic provides information about the following types of user data:

**Topics**

- Categorical metadata (p. 80)

### Categorical metadata

With some recipes and both VIDEO_ON_DEMAND and ECOMMERCE domains, Amazon Personalize uses categorical metadata, such as a user's gender or membership status, when identifying underlying patterns that reveal the most relevant items for your users. You define your own range of values based on your use case. Categorical metadata can be in any language.

With all recipes and domains, you can import categorical metadata and use it to filter recommendations based on a user's attributes. For information about filtering recommendations see Filtering recommendations and user segments.

Categorical values can have at most 1000 characters. If you have a user with a categorical value with more than 1000 characters, your dataset import job will fail.

For Custom dataset groups and custom solutions, recipes that use categorical metadata include the following:

- User-Personalization (p. 116)
- Personalized-Ranking (p. 139)
- Similar-Items (p. 143)

### Items datasets

The item data that you can import into Amazon Personalize includes numerical and categorical metadata such as creation timestamp, price, genre, description, and availability. You import metadata about your items into an Amazon Personalize Items dataset. The maximum number of metadata columns is 100.
The maximum number of items that are considered by a model during training is 750,000. Amazon Personalize only considers these items when generating recommendations. Some domains and recipes require an Items dataset. For more information on recipe requirements see Choosing a recipe (p. 113).

This topic provides information about the following types of item data:

**Topics**
- Creation timestamp data (p. 81)
- Categorical metadata (p. 81)
- Unstructured text metadata (p. 81)

**Creation timestamp data**

Amazon Personalize uses creation timestamp data (in Unix epoch time format, in seconds) to calculate the age of an item and adjust recommendations accordingly.

If creation timestamp data is missing for one or more items, Amazon Personalize infers this information from interaction data, if any, and uses the timestamp of the item's oldest interaction data as the item's creation timestamp. If an item has no interaction data, its creation timestamp is set as the timestamp of the latest interaction in the training set and Amazon Personalize considers it a new item.

**Categorical metadata**

With certain recipes and domains, Amazon Personalize uses categorical metadata, such as an item's genre or color, when identifying underlying patterns that reveal the most relevant items for your users. You define your own range of values based on your use case. Categorical metadata can be in any language.

With all recipes and domains, you can import categorical data and use it to filter recommendations based on an item's attributes. For information about filtering recommendations, see Filtering recommendations and user segments (p. 311).

Categorical values can have a maximum of 1000 characters. If you have an item with a categorical value with more than 1000 characters, your dataset import job will fail.

For Domain dataset groups, both VIDEO_ON_DEMAND and ECOMMERCE domains use categorical metadata. For Custom dataset groups and custom solutions, recipes that use categorical metadata include the following:

- User-Personalization (p. 116)
- Personalized-Ranking (p. 139)
- Similar-Items (p. 143)
- Item-Affinity (p. 149)
- Item-Attribute-Affinity (p. 150)

**Unstructured text metadata**

With certain recipes and domains, Amazon Personalize can extract meaningful information from unstructured text metadata, such as product descriptions, product reviews, or movie synopses. Amazon Personalize uses unstructured text to identify relevant items for your users, particularly when items are new or have less interactions data. Include unstructured text data in your Items dataset to increase click-through rates and conversation rates for new items in your catalog.

To use unstructured data, add a field with type string to your Items schema and set the field's textual attribute to true. Then include the text data in your bulk CSV file and individual item imports.
For bulk CSV files, wrap the text in double quotes. Use the `\` character to escape any double quotes or `\` characters in your data. You can add at most 1 textual field. For an example of an Items schema with a field for unstructured text data, see Items dataset schema example (custom) (p. 99). Amazon Personalize truncates text fields at the character limit. Make sure that the most relevant information in the text is at the start of the field. For information about importing data into Amazon Personalize, see Step 2: Preparing and importing data (p. 159).

Before using unstructured text values, Amazon Personalize removes the following from the text:

- HTML and XML tags and entities
- New line, tab, and extra space characters

Unstructured text values can have at most 20,000 characters in all languages except Chinese and Japanese. For Chinese and Japanese, you can have at most 7,000 characters. Amazon Personalize truncates values that exceed the character limit to the character limit.

Text can be in the following languages:

- Chinese (Simplified)
- Chinese (Traditional)
- English
- French
- German
- Japanese
- Portuguese
- Spanish

You can submit unstructured text items in multiple languages, but each item's text should be in only one language.

For Domain dataset groups, both VIDEO_ON_DEMAND and ECOMMERCE domains use textual metadata. For Custom dataset groups and custom solutions, recipes that use textual metadata include the following:

- User-Personalization (p. 116)
- Personalized-Ranking (p. 139)
- Similar-Items (p. 143)
- Item-Affinity (p. 149)
- Item-Attribute-Affinity (p. 150)
Schema formatting requirements

When you create a schema for either dataset in a Domain dataset group or Custom dataset group, you must follow these guidelines:

• You must define the schema in Avro format. For information on the Avro data types we support, see Schema data types (p. 83).
• The schema fields can appear in any order, but they must match the order of the corresponding column headers in your CSV file.
• Schemas must be flat JSON files without nested structures. For example, a field cannot be the parent of multiple sub-fields.
• Amazon Personalize schemas don’t support complex types such as arrays and maps.
• Schema fields must have unique alphanumeric names. For example, you can’t add both a GENRES_FIELD_1 field and a GENRESFIELD1 field.
• You must define required fields as their required data types. Reserved categorical string fields must have the categorical attribute set to true, while reserved string fields can't be categorical. The keywords can't be in your data.
• If you add your own metadata field of type string, it must include the categorical attribute or the textual attribute (only Items schemas support fields with the textual attribute). Otherwise, Amazon Personalize won't use the field when training a model.
• Amazon Personalize doesn't use boolean type data when training or filtering recommendations. To have Amazon Personalize use boolean data when training or filtering, use a field of type String and use the values "True" and "False" in your data. Or you can use type int or long and values 0 and 1.
• Textual fields must be of the type string and must have the textual attribute set to true. For more information about unstructured text data, see Unstructured text metadata (p. 81).

Domain dataset group datasets have additional requirements based on both domain and dataset type. Custom dataset group datasets have additional requirements depending on type.

Schema data types

Amazon Personalize schemas support the following Avro types for fields:

• float
• double
• int
• long
• string
• boolean
• null

Some required and reserved fields support null data. Adding a null type to a field allows you to use imperfect data (for example, metadata with blank values) to generate recommendations. For information on which fields support null data, see Domain datasets and schemas (p. 84) or Custom datasets and schemas (p. 94). The following example shows how to add a null type for a GENDER field.
{  
  "name": "GENDER",
  "type": [  
    "null",
    "string"
  ],
  "categorical": true
}

Domain datasets and schemas

When you create a Domain dataset group, the domain you choose determines your dataset and schema requirements. Each domain has a default schema for each dataset type.

When you create a dataset, you can either use the default schema or create a new one based on the default schema. Use the default schema as a guide for what data to collect and import into each dataset type. The following topics explain dataset and schema requirements for each domain.

For information on the types of data you can import into Amazon Personalize see Types of data Amazon Personalize can use (p. 9).

For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

Topics
- VIDEO_ON_DEMAND datasets and schemas (p. 84)
- ECOMMERCE datasets and schemas (p. 89)

VIDEO_ON_DEMAND datasets and schemas

When you create a Domain dataset group for the VIDEO_ON_DEMAND domain, each dataset type has a default schema with a set of VIDEO_ON_DEMAND specific required and recommended fields. You can either use the default schema or create a new one based on the default schema. The data you import must match your schema in format and type. Use the default domain schemas listed in the sections below as a guide to determine what data to import to create your VIDEO_ON_DEMAND-based recommender.

You are free to add additional fields. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you.

For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

The following topics provide information about each dataset's required and recommended fields for the VIDEO_ON_DEMAND domain. Each dataset section includes the default VIDEO_ON_DEMAND schema in JSON format.

Topics
- VIDEO_ON_DEMAND domain dataset and schema requirements (p. 85)
- Interactions dataset requirements (VIDEO_ON_DEMAND domain) (p. 85)
- Users dataset requirements (VIDEO_ON_DEMAND domain) (p. 87)
- Items dataset requirements (VIDEO_ON_DEMAND domain) (p. 88)
### VIDEO_ON_DEMAND domain dataset and schema requirements

Each dataset type has the following required fields and reserved keywords. Reserved keywords are optional, non-metadata fields. These fields are considered reserved because you must define the fields as their required data type when you use them. Reserved categorical string fields must have categorical set to true, while reserved string fields can't be categorical. The keywords can't be in your data.

<table>
<thead>
<tr>
<th>Dataset type</th>
<th>Required fields</th>
<th>Reserved keywords</th>
</tr>
</thead>
<tbody>
<tr>
<td>Interactions</td>
<td>USER_ID (string) ITEM_ID (string)</td>
<td>EVENT_VALUE (float, null) IMPRESSION (string, null) RECOMMENDATION_ID (string, null) EVENT_ATTRIBUTION_SOURCE (string, null)</td>
</tr>
<tr>
<td>Users</td>
<td>USER_ID (string) 1 metadata field (categorical string or numerical)</td>
<td>SUBSCRIPTION_MODEL (categorical string, null)</td>
</tr>
<tr>
<td>Items</td>
<td>ITEM_ID (string) CREATION_TIMESTAMP (long) GENRES (categorical string)</td>
<td>PRICE (float, null) DURATION (float, null) GENRE_L2 (categorical string, null) GENRE_L3 (categorical string, null) AVERAGE_RATING (float, null) PRODUCT_DESCRIPTION (textual string, null) CONTENT_OWNER (categorical string, null) CONTENT_CLASSIFICATION (categorical string, null)</td>
</tr>
</tbody>
</table>

**Interactions dataset requirements (VIDEO_ON_DEMAND domain)**

An Interactions dataset stores historical and real-time data from interactions between users and items in your VIDEO_ON_DEMAND catalog. For more information about the types of data you can store in an interactions dataset, see Interactions datasets (p. 77).

You must have an Interactions dataset for all use cases and your schema must have the following fields:

- USER_ID (string)
- ITEM_ID string
Domain datasets and schemas

- **TIMESTAMP** (long)
- **EVENT_TYPE** (string and depending on use case (p. 107), Watch and Click event types)

Your schema can also include the following reserved keywords:

- **EVENT_VALUE** (float, null)
- **IMPRESSION** (string, null)
- **RECOMMENDATION_ID** (string, null)

You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you. For an example of the default schema for Interactions datasets for VIDEO_ON_DEMAND domains, see Default Interactions schema (VIDEO_ON_DEMAND domain) (p. 86).

Optionally add the reserved keyword **EVENT_VALUE** if you have value data for events, such as the percentage of a video watched. Optionally add the reserved keyword **IMPRESSION** if you want to include explicit and implicit impressions data. For more information about recording impressions data see Impressions data (p. 79).

The maximum total number of optional metadata fields you can add to an Interactions dataset, combined with total number of distinct event types in your data, is 10. The metadata fields included in this count are EVENT_TYPE, EVENT_VALUE fields along with any custom metadata fields you add to your schema. The maximum number of metadata fields excluding reserved fields, such as IMPRESSION, is 5. Categorical values can have at most 1000 characters. If you have an interaction with a categorical value with more than 1000, your dataset import job will fail.

For more information on minimum requirements and maximum data limits for an Interactions dataset for the VIDEO_ON_DEMAND domain, see Service quotas (p. 415). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

**Default Interactions schema (VIDEO_ON_DEMAND domain)**

The following is the default VIDEO_ON_DEMAND domain schema for Interactions datasets.

```json
{
  "type": "record",
  "name": "Interactions",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
    {
      "name": "USER_ID",
      "type": "string"
    },
    {
      "name": "ITEM_ID",
      "type": "string"
    },
    {
      "name": "EVENT_TYPE",
      "type": "string"
    },
    {
      "name": "TIMESTAMP",
      "type": "long"
    }
  ]
}
```
Users dataset requirements (VIDEO_ON_DEMAND domain)

A Users dataset stores metadata about your users. This might include information such as age, gender, and loyalty membership for each item. For information on the types of user data you can import into Amazon Personalize, see Users datasets (p. 80). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

A Users dataset is optional for all VIDEO_ON_DEMAND use cases. If you have user data, we recommend creating one to get the most relevant recommendations. If you create a Users dataset, your schema must include the following fields.

- **USER_ID**
- 1 metadata field (categorical string or numerical)

You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you. For an example of the default schema for Users datasets for VIDEO_ON_DEMAND domains, see Default Users schema (VIDEO_ON_DEMAND domain) (p. 87).

A SUBSCRIPTION_MODEL field is included in the default schema. This field is an optional reserved keyword and must have a type of string with categorical set to true. To get the best recommendations, we recommend that you keep this field in your schema if you have subscription model information about each of your users in your data. The data you import must match your schema.

Using categorical data

To use categorical data, add a field of type string and set the field's categorical attribute to true in your schema. Then include the categorical data in your bulk CSV file and individual item imports. For users with multiple categories, separate each value using the vertical bar, '|'. For example, for a SUBSCRIPTION_MODEL field, your data for a user might be student|monthly|discount.

Categorical values can have at most 1000 characters. If you have a user with a categorical value with more than 1000 characters, your dataset import job will fail.

Default Users schema (VIDEO_ON_DEMAND domain)

The following is the default VIDEO_ON_DEMAND domain schema for Users datasets.

```json
{
  "type": "record",
  "name": "Users",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
    {
      "name": "USER_ID",
      "type": "string"
    },
    {
      "name": "SUBSCRIPTION_MODEL",
      "type": "string",
      "categorical": true
    }
  ],
  "version": "1.0"
}
```

87
Items dataset requirements (VIDEO_ON_DEMAND domain)

An Items dataset stores metadata about your items in your catalogue. This might include information such as price, genre, and availability for each item. For information about the types of item data you can import into Amazon Personalize, see Items datasets (p. 80). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

An Items dataset is required for some use cases (see VIDEO_ON_DEMAND use cases (p. 108)). When optional, we still recommend creating one to get the most relevant recommendations. If you create an Items dataset, your schema must include the following fields:

- ITEM_ID
- GENRES (categorical string)
- CREATION_TIMESTAMP (in Unix epoch time format)

Your schema can also include the following reserved keywords. Each keyword lists its required data type and whether it supports null data. Adding the null type is optional.

- PRICE (float)
- DURATION (float)
- GENRE_L2 (categorical string, null)
- GENRE_L3 (categorical string, null)
- AVERAGE_RATING (float, null)
- PRODUCT_DESCRIPTION (textual string, null)
- CONTENT_OWNER (categorical string, null): The company that owns the video. For example, values might be HBO, Paramount, and NBC.
- CONTENT_CLASSIFICATION (categorical string, null): The content's rating. For example, values might be G, PG, PG-13, R, NC-17, and unrated.

To get the best recommendations, we recommend that you keep these as many of these fields in your schema as you have data. The data you import must match your schema. The maximum number of metadata columns is 100. You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you.

Use reserved keywords GENRE_L2 and GENRE_L3 for items with multiple multi-level categories. For more information, see Using categorical data (p. 88). For information on textual and categorical metadata see Items datasets (p. 80). For an example of the default schema for Items datasets for ECOMMERCE domains, see Default Items schema (VIDEO_ON_DEMAND domain) (p. 89).

Using categorical data

To use categorical data, add a field of type string and set the field's categorical attribute to true in your schema. Then include the categorical data in your bulk CSV file and individual item imports. Categorical values can have at most 1000 characters. If you have an item with a categorical value with more than 1000 characters, your dataset import job will fail.

For items with multiple categories, separate each value with the vertical bar, '|'. For example, for a GENRES field your data for an item might be Action|Crime|Biopic. If you have a multiple levels of categorical data and some items have multiple categories for each level in the hierarchy, add a field for each level and append a level indicator after each field name: GENRES, GENRE_L2, GENRE_L3. This
allows you filter recommendations based on sub-categories, even if an item belongs to multiple multi-level categories. For example, a video might have the following data for each category level:

- GENRES: Action|Adventure
- GENRE_L2: Crime|Western
- GENRE_L3: biopic

In this example, the video is in the action > crime > biopic hierarchy and the adventure > western > biopic hierarchy. We recommend only using up to L3 but you can use more levels if necessary. For information on creating and using filters, see Filtering recommendations and user segments (p. 311).

Default Items schema (VIDEO_ON_DEMAND domain)

The following is the default schema for Items datasets for the VIDEO_ON_DEMAND domain.

```json
{
    "type": "record",
    "name": "Items",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "GENRES",
            "type": [
                "string"
            ],
            "categorical": true
        },
        {
            "name": "CREATION_TIMESTAMP",
            "type": "long"
        }
    ],
    "version": "1.0"
}
```

**ECOMMERCE datasets and schemas**

When you create a Domain dataset group for the ECOMMERCE domain, each dataset type has a default schema with a set of ECOMMERCE-specific required and recommended fields. You can use the default schema or create a new one based on the default schema. The data you import must match your schema in format and type. Use the default domain schemas listed in the sections below as a guide to determine what data to import to create your ECOMMERCE-based recommender.

You are free to add additional fields. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you.

For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

The following topics provide information about each dataset's required and recommended fields for the ECOMMERCE domain. Each dataset section includes the default ECOMMERCE schema in JSON format.

**Topics**
- ECOMMERCE domain dataset and schema requirements (p. 90)
Domain datasets and schemas

- Interactions dataset requirements (ECOMMERCE domain) (p. 90)
- Users dataset requirements (ECOMMERCE domain) (p. 92)
- Items dataset requirements (ECOMMERCE domain) (p. 93)

**ECOMMERCE domain dataset and schema requirements**

Each dataset type has the following required fields and reserved keywords. Reserved keywords are optional, non-metadata fields. These fields are considered reserved because you must define the fields as their required data type when you use them. Reserved categorical string fields must have categorical set to `true`, while reserved string fields can't be categorical. The keywords can't be in your data.

<table>
<thead>
<tr>
<th>Dataset type</th>
<th>Required fields</th>
<th>Reserved keywords</th>
</tr>
</thead>
</table>
| Interactions (default schema (p. 91)) | USER_ID (string)  
ITEM_ID (string)  
TIMESTAMP (long)  
EVENT_TYPE (string and depending on use case (p. 107), Purchase and View event types)  
EVENT_VALUE (float, null)  
IMPRESSION (string, null)  
RECOMMENDATION_ID (string, null)  
EVENT_ATtribution_SOURCE (string, null) | |
| Users (default schema (p. 92)) | USER_ID (string)  
1 metadata field (categorical string or numerical) | |
| Items (default schema (p. 94)) | ITEM_ID (string)  
PRICE (float)  
CATEGORY_L1 (categorical string)  
CATEGORY_L2 (categorical string, null)  
CATEGORY_L3 (categorical string, null)  
PRODUCT_DESCRIPTION (textual string, null)  
CREATION_TIMESTAMP (long)  
AGE_GROUP (categorical string, null)  
ADULT (categorical string, null)  
GENDER (categorical string, null) | |

**Interactions dataset requirements (ECOMMERCE domain)**

An Interactions dataset stores historical and real-time data from interactions between users and items in your ECOMMERCE catalog. For more information about the types of data you can store in an interactions dataset, see Interactions datasets (p. 77). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.
You must at minimum create an Interactions dataset and your schema must have the following fields:

- USER_ID (string)
- ITEM_ID (string)
- TIMESTAMP (Long)
- EVENT_TYPE (string and depending on use case (p. 107), Purchase and View event types)

Your schema can also include the following reserved keywords:

- EVENT_VALUE (float, null)
- IMPRESSION (string, null)
- RECOMMENDATION_ID (string, null)

The data you import must match your schema. You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you. For an example of the default schema for Interactions datasets for ECOMMERCE domains, see Default Interactions schema (ECOMMERCE domain) (p. 91).

Optionally add the reserved keyword EVENT_VALUE if you have value data for events. Optionally add the reserved keyword IMPRESSION if you want to include explicit and implicit impressions data. For more information about recording impressions data see Impressions data (p. 79).

The maximum total number of optional metadata fields you can add to an Interactions dataset, combined with total number of distinct event types in your data, is 10. The metadata fields included in this count are EVENT_TYPE, EVENT_VALUE fields along with any custom metadata fields you add to your schema. The maximum number of metadata fields excluding reserved fields, such as IMPRESSION, is 5. Categorical values can have at most 1000 characters. If you have an interaction with a categorical value with more than 1000, your dataset import job will fail.

For more information on minimum requirements and maximum data limits for an Interactions dataset for the ECOMMERCE domain, see Service quotas (p. 415).

Default Interactions schema (ECOMMERCE domain)

The following is the default ECOMMERCE domain schema for Interactions datasets.

```
{
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        { "name": "USER_ID", "type": "string" },
        { "name": "ITEM_ID", "type": "string" },
        { "name": "EVENT_TYPE", "type": "string" },
        { "name": "TIMESTAMP", "type": "long" }
    ]
}
```
Users dataset requirements (ECOMMERCE domain)

A Users dataset stores metadata about your users. This might include information such as age, gender, and loyalty membership for each user. For more information on the types of user data you can import into Amazon Personalize, see Users datasets (p. 80). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

A Users dataset is optional for all ECOMMERCE use cases. If you have user data, we recommend creating one to get the most relevant recommendations. If you create a Users dataset, your schema must include the following fields.

- USER_ID
- 1 metadata field (categorical string or numerical)

The data you import must match your schema. You are free to add additional fields depending on your use case and your data. As long as the fields aren’t listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you. For an example of the default schema for Users datasets for ECOMMERCE domains, see Default Users schema (ECOMMERCE domain) (p. 92).

For more information on minimum requirements and maximum data limits for a Users dataset, see Service quotas (p. 415).

Using categorical data

To use categorical data, add a field of type string and set the field’s categorical attribute to true in your schema. Then include the categorical data in your bulk CSV file and individual item imports. For users with multiple categories, separate each value using the vertical bar, '|'. For example, for a SUBSCRIPTION_MODEL field, your data for a user might be student|monthly|discount.

Categorical values can have at most 1000 characters. If you have a user with a categorical value with more than 1000 characters, your dataset import job will fail.

Default Users schema (ECOMMERCE domain)

The following is the default ECOMMERCE domain schema for Users datasets with a CATEGORY field as the required metadata field.

```json
{
  "type": "record",
  "name": "Users",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
    {
      "name": "USER_ID",
      "type": "string"
    },
    {
      "name": "MEMBERSHIP_STATUS",
      "type": "string",
      "categorical": true
    }
  ],
}
Items dataset requirements (ECOMMERCE domain)

An Items dataset stores metadata about your ECOMMERCE items. This might include information such as price, category, and product description for each item. For more information on the types of item data you can import into Amazon Personalize, see Items datasets (p. 80). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all schemas, regardless of domain.

An Items dataset is optional for all ECOMMERCE use cases. If you have items data, we recommend creating one to get the most relevant recommendations. If you create an items dataset, your schema must include the following fields:

- **ITEM_ID**
- **PRICE (Float)**
- **CATEGORY_L1 (categorical string)**

Your schema can also include the following reserved keywords. For categorical fields, you can define your own range of values based on your use case.

- **CATEGORY_L2 (categorical string, null)**
- **CATEGORY_L3 (categorical string, null)**
- **PRODUCT_DESCRIPTION (textual string, null)**
- **CREATION_TIMESTAMP (Float)**
- **AGE_GROUP (categorical string, null): The age group the item is for. Values might be newborns, infants, children, and adults.**
- **ADULT (categorical string, null): Whether the item is restricted to only adults, such as alcohol. Values might be yes or no.**
- **GENDER (categorical string, null): The gender the item is for. Values might be male, female, and unisex.**

To get the best recommendations, we recommend that you keep these as many of these fields in your schema as you have data. The data you import must match your schema. The maximum number of metadata columns is 100. You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you.

Use reserved keywords CATEGORY_L2 and CATEGORY_L3 for items with multiple multi-level categories. For more information, see Using categorical data (p. 93). For information on textual and categorical metadata see Unstructured text metadata (p. 81). For an example of the default schema for Items datasets for ECOMMERCE domains, see Default Items schema (ECOMMERCE domain) (p. 94).

Using categorical data

To use categorical data, add a field of type string and set the field's categorical attribute to true in your schema. Then include the categorical data in your bulk CSV file and individual item imports. You can define your own range of values based on your use case. Categorical values can have at most 1000 characters. If you have an item with a categorical value with more than 1000 characters, your dataset import job will fail.

For items with multiple categories, separate each value with the vertical bar, '|'. For example, for a CATEGORY_L1 field your data for an item might be Electronics|Productivity|Mouse. If you
have a multiple levels of categorical data and some items have multiple categories for each level in the hierarchy, add a field for each level and append a level indicator after each field name: CATEGORY_L1, CATEGORY_L2, CATEGORY_L3. This allows you filter recommendations based on sub-categories, even if an item belongs to multiple multi-level categories. For example, an item might have the following data for each category level:

- CATEGORY_L1: Electronics|Productivity
- CATEGORY_L2: Productivity|Computers
- CATEGORY_L3: Mouse

In this example, the item is in the electronics > productivity > mouse hierarchy and the productivity > computers > mouse hierarchy. We recommend only using up to L3 but you can use more levels if necessary. For information on creating and using filters see Filtering recommendations and user segments (p. 311).

**Default Items schema (ECOMMERCE domain)**

The following is the default schema for Items datasets for the ECOMMERCE domain with only the required fields.

```json
{
    "type": "record",
    "name": "Items",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "PRICE",
            "type": "float"
        },
        {
            "name": "CATEGORY_L1",
            "type": [
                "string"
            ],
            "categorical": true
        }
    ],
    "version": "1.0"
}
```

**Custom datasets and schemas**

When you create a Custom dataset group, you create your own schemas from scratch. Custom dataset group datasets and schemas have fewer required fields and more flexibility. The following topics explain the schema and data requirements for Interactions, Items, and Users datasets for a Custom dataset group. Each dataset section lists the required data for the dataset type and provides a JSON example of a schema.

For information on the types of data you can import into Amazon Personalize see Types of data Amazon Personalize can use (p. 9). For information about general Amazon Personalize schema requirements, such as formatting requirements and available field data types, see Schemas (p. 82). These requirements apply to all Amazon Personalize schemas.
Custom dataset and schema requirements

When you create a dataset for a Custom dataset group, each dataset type has the following required fields and reserved keywords with required data types.

<table>
<thead>
<tr>
<th>Dataset type</th>
<th>Required fields</th>
<th>Reserved keywords</th>
</tr>
</thead>
<tbody>
<tr>
<td>Interactions</td>
<td>USER_ID (string)</td>
<td>EVENT_TYPE (string)</td>
</tr>
<tr>
<td></td>
<td>ITEM_ID (string)</td>
<td>EVENT_VALUE (float, null)</td>
</tr>
<tr>
<td></td>
<td>TIMESTAMP (long)</td>
<td>IMPRESSION (string, null)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>RECOMMENDATION_ID (string, null)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>EVENT_ATTRIBUTION_SOURCE (string, null)</td>
</tr>
<tr>
<td>Users</td>
<td>USER_ID (string)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>1 metadata field (categorical string or numerical)</td>
<td></td>
</tr>
<tr>
<td>Items</td>
<td>ITEM_ID</td>
<td>CREATION_TIMESTAMP (long)</td>
</tr>
<tr>
<td></td>
<td>1 metadata field (categorical or textual string field or numerical field)</td>
<td></td>
</tr>
</tbody>
</table>

Metadata fields

Metadata includes string or non-string fields that aren't required or don't use a reserved keyword. Metadata schemas have the following restrictions:

- Users and Items schemas require at least one metadata field.
- You can add at most 25 metadata fields for a Users schema and 100 metadata fields for an Items schema.
- If you add your own metadata field of type string, it must include the categorical attribute or the textual attribute (only Items schemas support fields with the textual attribute). Otherwise, Amazon Personalize won't use the field when training a model.

Reserved keywords

Reserved keywords are optional, non-metadata fields. These fields are considered reserved because you must define the fields as their required data type when you use them, and the keywords can't be used as values in your data. Reserved categorical string fields must have categorical set to true, while reserved string fields can't be categorical. The following are reserved keywords:

- EVENT_TYPE
- EVENT_VALUE
- IMPRESSION
- RECOMMENDATION_ID
- EVENT_ATTRIBUTION_SOURCE
- CREATION_TIMESTAMP
• EVENT_TYPE: For Interactions datasets with one or more event types, such as both click and download, use an EVENT_TYPE field. You must define an EVENT_TYPE field as a string and can't be set as categorical.

• EVENT_VALUE: For Interactions datasets that include value data for events, such as the percentage of a video a user watched, use an EVENT_VALUE field with type float and optionally null.

• CREATION_TIMESTAMP: For Items datasets with a timestamp for each item's creation date, use a CREATION_TIMESTAMP field with a type long. Amazon Personalize uses CREATION_TIMESTAMP data to calculate the age of an item and adjust recommendations accordingly. See Creation timestamp data (p. 81).

• IMPRESSION: For Interactions datasets with explicit impressions data, use an IMPRESSION field with type String and optionally type null. Impressions are lists of items that were visible to a user when they interacted with (for example, clicked or watched) a particular item. For more information see Impressions data (p. 79).

• RECOMMENDATION_ID: For Interactions datasets that use previous recommendations as implicit impressions data, optionally use a RECOMMENDATION_ID field with type String and optionally type null.

You don't need to add a RECOMMENDATION_ID field for Amazon Personalize to use implicit impressions when generating recommendations. You can pass a recommendationId in a PutEvents (p. 598) operation without it. For more information see Impressions data (p. 79).

Interactions dataset schema requirements (custom)

An Interactions dataset stores historical and real-time data from interactions between users and items in your catalog. For information on the types of interactions data Amazon Personalize can use, see Interactions datasets (p. 77).

The data you provide for each interaction must match your schema. Depending on your schema, interaction metadata can include empty/null values. At minimum, you must provide the following for each interaction:

• User ID
• Item ID
• Timestamp (in Unix epoch time format)

You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you.

The maximum total number of optional metadata fields you can add to an Interactions dataset, combined with total number of distinct event types in your data, is 10. The metadata fields included in this count are EVENT_TYPE, EVENT_VALUE fields along with any custom metadata fields you add to your schema. The maximum number of metadata fields excluding reserved fields, such as IMPRESSION, is 5. Categorical values can have at most 1000 characters. If you have an interaction with a categorical value with more than 1000, your dataset import job will fail.

For more information on minimum requirements and maximum data limits for an Interactions dataset, see Service quotas (p. 415).

Interactions schema example (custom)

The following example shows a schema for an Interactions dataset. The USER_ID, ITEM_ID, and TIMESTAMP fields are required. The EVENT_TYPE, EVENT_VALUE, and IMPRESSION fields are optional
reserved keywords recognized by Amazon Personalize. EVENT_TYPE must of type string and can't be categorical. LOCATION and DEVICE are optional contextual metadata fields. For information on schema requirements see Custom dataset and schema requirements (p. 95).

```json
{
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "USER_ID",
            "type": "string"
        },
        {
            "name": "ITEM_ID",
            "type": "string"
        },
        {
            "name": "EVENT_TYPE",
            "type": "string"
        },
        {
            "name": "EVENT_VALUE",
            "type": ["float", "null"]
        },
        {
            "name": "LOCATION",
            "type": "string",
            "categorical": true
        },
        {
            "name": "DEVICE",
            "type": ["string", "null"],
            "categorical": true
        },
        {
            "name": "TIMESTAMP",
            "type": "long"
        },
        {
            "name": "IMPRESSION",
            "type": "string"
        }
    ],
    "version": "1.0"
}
```

For this schema, the first few lines of historical data in a CSV file might look like the following. Note that some values for EVENT_VALUE are null.

<table>
<thead>
<tr>
<th>USER_ID, ITEM_ID, EVENT_TYPE, EVENT_VALUE, LOCATION, DEVICE, TIMESTAMP, IMPRESSION</th>
</tr>
</thead>
<tbody>
<tr>
<td>35, 73, click, , Ohio, Tablet, 1586731606, 73</td>
</tr>
<tr>
<td>54, 35, watch, 0.75, Indiana, Cellphone, 1586735164, 35</td>
</tr>
<tr>
<td>9, 33, click, , Oregon, Cellphone, 1586735158, 68</td>
</tr>
</tbody>
</table>
Users dataset requirements (custom)

A Users dataset stores metadata about your users. This might include information such as age, gender, and loyalty membership for each item. For information on the types of user data you can import into Amazon Personalize, see Users datasets (p. 80).

The data you provide for each user must match your schema. At minimum, you must provide a User ID for each user (max length 256 characters). Depending on your schema, user metadata can include empty/null values. Your Users schema must have minimum one metadata field, but if you add a null type, this value can be null for the user. You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in Schema data types (p. 83), the field names and data types are up to you.

To use categorical data, add a field of type string and set the field's categorical attribute to true in your schema. Then include the categorical data in your bulk CSV file and individual item imports. For users with multiple categories, separate each value using the vertical bar, '|'. For example, for a SUBSCRIPTION_MODEL field, your data for a user might be student|monthly|discount.

Categorical values can have at most 1000 characters. If you have a user with a categorical value with more than 1000 characters, your dataset import job will fail.

For more information on minimum requirements and maximum data limits for a Users dataset, see Service quotas (p. 415).

Users schema example (custom)

The following example shows how to structure a Users schema. The USER_ID field is required and the AGE and GENDER fields are metadata. At least one metadata field is required and you can add at most 25 metadata fields. For information about schema requirements see Custom dataset and schema requirements (p. 95).

```
{
  "type": "record",
  "name": "Users",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
    {
      "name": "USER_ID",
      "type": "string"
    },
    {
      "name": "AGE",
      "type": "int"
    },
    {
      "name": "GENDER",
      "type": "string",
      "categorical": true
    }
  ],
  "version": "1.0"
}
```
For this schema, the first few lines of historical data in a CSV file might look like the following.

<table>
<thead>
<tr>
<th>USER_ID, AGE, GENDER</th>
</tr>
</thead>
<tbody>
<tr>
<td>5, 34, Male</td>
</tr>
<tr>
<td>6, 56, Female</td>
</tr>
<tr>
<td>8, 65, Male</td>
</tr>
<tr>
<td>...</td>
</tr>
<tr>
<td>...</td>
</tr>
</tbody>
</table>

**Items dataset requirements (custom)**

An *Items dataset* stores metadata about your items in your catalogue. This might include information such as price, genre, and availability for each item. For information about the types of item data you can import into Amazon Personalize, see [Items datasets](#) (p. 80).

The data you provide for each item must match your Items dataset schema. At minimum, you must provide an Item ID for each item (max length 256 characters). Depending on your schema, item metadata can include empty/null values. Your schema must have minimum one metadata field, but if you add a null type, this value can be null for the item. You are free to add additional fields depending on your use case and your data. As long as the fields aren't listed as required or reserved, and the data types are listed in *Schema data types* (p. 83), the field names and data types are up to you.

To use categorical data, add a field of type `string` and set the field's categorical attribute to `true` in your schema. Then include the categorical data in your bulk CSV file and individual item imports. Categorical values can have at most 1000 characters. If you have an item with a categorical value with more than 1000 characters, your dataset import job will fail.

For items with multiple categories, separate each value with the vertical bar, `|`. For example, for a `GENRES` field your data for an item might be `Action|Crime|Biopic`. If you have a multiple levels of categorical data and some items have multiple categories for each level in the hierarchy, add a field for each level and append a level indicator after each field name: `GENRES, GENRE_L2, GENRE_L3`. This allows you filter recommendations based on sub-categories, even if an item belongs to multiple multi-level categories (for information on creating and using filters see [Filtering recommendations and user segments](#) (p. 311)). For example, a video might have the following data for each category level:

- `GENRES: Action|Adventure`
- `GENRE_L2: Crime|Western`
- `GENRE_L3: Biopic`

In this example, the video is in the action > crime > biopic hierarchy and the adventure > western > biopic hierarchy. We recommend only using up to L3 but you can use more levels if necessary.

During model training, Amazon Personalize considers a maximum of 750,000 items. If you import more than 750,000 items, Amazon Personalize decides which items to include in training, with an emphasis on including new items (items you recently added with no interactions) and existing items with recent interactions data.

For more information on minimum requirements and maximum data limits for an Items dataset, see [Service quotas](#) (p. 415).

**Items dataset schema example (custom)**

The following example shows how to structure an Items schema. The `ITEM_ID` field is required. The `GENRE` field is categorical metadata and the `DESCRIPTION` field is textual metadata. At least one metadata field is required. You can add a maximum of 100 metadata fields. The `CREATION_TIMESTAMP` field is a reserved keyword. For information about schema requirements, see [Custom dataset and schema requirements](#) (p. 95).
Custom datasets and schemas

```json
{
  "type": "record",
  "name": "Items",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
    {
      "name": "ITEM_ID",
      "type": "string"
    },
    {
      "name": "GENRES",
      "type": [
        "null",
        "string"
      ],
      "categorical": true
    },
    {
      "name": "CREATION_TIMESTAMP",
      "type": "long"
    },
    {
      "name": "DESCRIPTION",
      "type": [
        "null",
        "string"
      ],
      "textual": true
    }
  ],
  "version": "1.0"
}
```

For this schema, the first few lines of historical data in a CSV file might look like the following.

<table>
<thead>
<tr>
<th>ITEM_ID, GENRES, CREATION_TIMESTAMP, DESCRIPTION</th>
</tr>
</thead>
<tbody>
<tr>
<td>1, Adventure</td>
</tr>
<tr>
<td>2, Adventure</td>
</tr>
<tr>
<td>3, Comedy</td>
</tr>
<tr>
<td>4, Comedy</td>
</tr>
</tbody>
</table>

Creating a schema with SDK for Python (Boto3)

1. Define the Avro format schema that you want to use.
2. Save the schema in a JSON file in the default Python folder.
3. Create the schema using the following code.

```python
import boto3

personalize = boto3.client('personalize')

with open('schema.json') as f:
    createSchemaResponse = personalize.create_schema(
        name = 'YourSchema',
    )
```
schema = f.read()
)
schema_arn = createSchemaResponse['schemaArn']
print('Schema ARN:' + schema_arn )

4. Amazon Personalize returns the ARN of the new schema. Store it for later use.

Amazon Personalize provides operations for managing schemas. For example, you can use the ListSchemas (p. 568) API to get a list of the available schemas.

After you create a schema, use it with datasets that match the schema. For more information, see Data format guidelines (p. 101).

Data format guidelines

When you import data into Amazon Personalize datasets, you can choose to import records in bulk, individually, or both.

- Bulk imports involve importing a large number of historical records at once. You can prepare and import your bulk data with SageMaker Data Wrangler and multiple data sources. Or you can prepare bulk data yourself, and import it directly into Amazon Personalize from a CSV file in Amazon S3.
- With individual imports, you import individual records with the Amazon Personalize console and API operations. Or you can stream interactions data from live events in real time. For more information about individual imports, see Importing individual records (p. 182).

Before you import your bulk data, make sure it's properly formatted. The following sections can help you format your bulk data. If you're not sure how to format your data, you can use Amazon SageMaker Data Wrangler (Data Wrangler) to prepare your data. For more information, see Preparing and importing data using Amazon SageMaker Data Wrangler (p. 160).

Topics
- Bulk data format guidelines and requirements (p. 101)
- Interactions data example (p. 102)
- Formatting explicit impressions (p. 103)
- Formatting categorical data (p. 103)

Bulk data format guidelines and requirements

The following guidelines and requirements can help you make sure your bulk data is formatted correctly.

- Your input data must be in a CSV (comma-separated values) file.
- The first row of your CSV file must contain your column headers. Don't enclose headers in quotation marks (' ').
- Make sure you have the required fields for your dataset type and make sure that their names align with Amazon Personalize requirements. For example, your Items data might have a column called ITEM_IDENTIFICATION_NUMBER with IDs for each of your items. To use this column as an ITEM_ID field, rename the column to ITEM_ID. If you use Data Wrangler to format your data, you can use the Map columns for Amazon Personalize Data Wrangler transform to make sure your columns are named correctly.
Interactions data example

The following interactions data represents historical user activity from a website that sells movie tickets. You might use the data to train a model that provides movie recommendations based on users' interaction data.

```
USER_ID,ITEM_ID,EVENT_TYPE,EVENT_VALUE,TIMESTAMP
196,242,click,15,881250949
186,302,click,13,891717742
22,377,click,10,878887116
244,51,click,20,880606923
166,346,click,10,886397596
298,474,click,40,884182806
115,265,click,20,881171488
253,465,click,50,891628467
305,451,click,30,886324817
```

Here's associated Interactions schema:

```
{
  "type": "record",
  "name": "Interactions",
  "namespace": "com.amazonaws.personalize.schema",
  "fields": [
    {
      "name": "USER_ID",
      "type": "string"
    },
    {
      "name": "ITEM_ID",
      "type": "string"
    },
    {
      "name": "EVENT_TYPE",
      "type": "string"
    },
    {
      "name": "EVENT_VALUE",
      "type": "float"
    }
  ]
}
```
Amazon Personalize requires the USER_ID, ITEM_ID, and TIMESTAMP fields. USER_ID is the identifier for a user of your application. ITEM_ID is the identifier for a movie. EVENT_TYPE and EVENT_VALUE are the identifiers for user activities. In the sample data, a click might represent a movie purchase event and 15 might be the purchase price of the movie. TIMESTAMP represents the Unix epoch time that the movie purchase took place.

**Timestamp data**

Timestamp data, such as TIMESTAMP (for Interactions datasets) or CREATION_TIMESTAMP (for Items datasets) data, must be in Unix epoch time format in seconds. For example, the Epoch timestamp in seconds for date July 31, 2020 is 1596238243. To convert dates to Unix epoch timestamps use an [Epoch converter - Unix timestamp converter](https://www.epochconverter.com/).

**Formatting explicit impressions**

If you use the User-Personalization (p. 116) recipe, you can record and upload impressions data. Impressions are lists of items that were visible to a user when interacting with a particular item (for example, clicked or watched). To upload impressions data in a bulk data import, manually record each item ID. Be sure to separate the values with a vertical bar, '|', character as part of your historical interactions data. The vertical bar character counts toward the 1000 character limit for impressions data. For more information on impressions data, see [Impressions data (p. 79)](https://docs.aws.amazon.com/personalize/latest/dg/data.html#data-impressions).

The following is a short excerpt from an Interactions dataset that includes explicit impressions in the IMPRESSION column.

<table>
<thead>
<tr>
<th>EVENT_TYPE</th>
<th>IMPRESSION</th>
<th>ITEM_ID</th>
<th>TIMESTAMP</th>
<th>USER_ID</th>
</tr>
</thead>
<tbody>
<tr>
<td>click</td>
<td>73</td>
<td>70</td>
<td>17</td>
<td>95</td>
</tr>
<tr>
<td>click</td>
<td>35</td>
<td>82</td>
<td>78</td>
<td>57</td>
</tr>
<tr>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
</tr>
</tbody>
</table>

The application showed user USER_1 items 73, 70, 17, 95, and 96 and the user ultimately chose item 73. When you create a new solution version based on this data, items 70, 17, 95, and 96 will be less frequently recommended to user USER_1.

**Formatting categorical data**

To include multiple categories for a single item when you use categorical string data, separate the values using the vertical bar, '|', character. For example, for an item that has two categories, a data row would resemble the following:

```
ITEM_ID,GENRE
item_123,horror|comedy
```
After you format your data, upload it to an Amazon S3 bucket so you can import it into Amazon Personalize. For more information, see [Uploading to an Amazon S3 bucket](p. 171).
Domain use cases and custom recipes

Amazon Personalize provides different domain use cases and custom recipes for training models:

- When you create a recommender in a Domain dataset group, you specify a use case. Amazon Personalize trains the models backing the recommender with the best configurations for the use case.
- When you create a custom solution in a Custom dataset group or Domain dataset group, you specify a recipe and configure training parameters. When you create a solution version for the solution, Amazon Personalize trains the models backing the solution version based on the recipe and training configuration.

Topics
- Use case and recipe features (p. 105)
- Choosing a use case (p. 107)
- Choosing a recipe (p. 113)

Use case and recipe features

With some use case and recipes, Amazon Personalize uses the following features to generate more relevant recommendations and improve item discovery and engagement.

Topics
- Real-time personalization (p. 105)
- Exploration (p. 106)
- Automatic updates (p. 106)

Real-time personalization

With some use cases and recipes, Amazon Personalize uses real-time personalization to update and adapt recommendations according to a user's evolving interest. It updates recommendations for a user as you record their interactions with items present at the latest full training. You record these interactions with an event tracker and the PutEvents (p. 598) operation.

For more information about recording events, see Recording events (p. 280). For information about new data influences real-time recommendations, including real-time personalization, see How new data influences real-time recommendations (p. 292).

The following use cases and recipes support real-time personalization:

- Recommended for you (ECOMMERCE use case) (p. 112)
- Top picks for you (VIDEO_ON_DEMAND use case) (p. 109)
- User-Personalization recipe (p. 116)
- Personalized-Ranking recipe (p. 139)
Exploration

If you use the Top picks for you and Recommended for you domain use cases, or the User-Personalization custom recipe, Amazon Personalize uses exploration when recommending items. With exploration, recommendations include some items that would be typically less likely to be recommended for the user, such as new items, items with few interactions, or items less relevant for the user based on their previous behavior. This improves item discovery and engagement when you have a fast-changing catalog, or when new items, such as news articles or promotions, are more relevant to users because they are fresh.

If your use case or recipe uses exploration, when you create a recommender or custom campaign, or when you create a batch inference job (custom resources), you can configure exploration with the following fields:

- Emphasis on exploring less relevant items (exploration weight) – Configure how much to explore. Specify a decimal value between 0 to 1. The default is 0.3. The closer the value is to 1, the more exploration. With more exploration, recommendations include more items with less interactions data or relevance based on previous behavior. At zero, no exploration occurs and recommendations are based on current data (relevance).

- Exploration item age cutoff – Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines item age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines item age, see Creation timestamp data (p. 81).

To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them.

For more information about each use case or recipe that uses exploration, see the following:

- Recommended for you (ECOMMERCE use case) (p. 112)
- Top picks for you (VIDEO_ON_DEMAND use case) (p. 109)
- User-Personalization recipe (p. 116)

Automatic updates

If you use the Top picks for you and Recommended for you domain cases or the User-Personalization custom recipe, Amazon Personalize automatically updates your recommender or solution version to consider new items for recommendations. When considering the new item, Amazon Personalize considers any metadata for the item, but this data will have a greater effect on recommendations only after you record interactions for the item and fully retrain.

Automatic updates are not a full retraining. Instead, automatic updates allow Amazon Personalize to feature your new items in recommendations before your next full retraining. A full training can be after your domain recommender's weekly automatic retraining completes. Or it can be after you create a new solution version with trainingMode set to FULL.

With each update, Amazon Personalize uses any interactions data, including impressions data, to identify what items to include or not include in exploration. For more information about exploration, including how you can configure it, see Exploration (p. 106).

When Amazon Personalize automatically updates your solution version depends on how you get recommendations:
• For real-time recommendations, Amazon Personalize automatically updates your recommender or latest solution version every two hours. For an update to occur, you must provide new item or interactions data since the last automatic update or full retraining.

• For batch item recommendations, when you create a batch inference job and specify the latest fully trained solution version for your solution, Amazon Personalize automatically updates the solution version to consider new items during exploration. If you don't specify the latest solution version, no update occurs.

**Additional guidelines and requirements for custom resources**

If you use custom resources, the following are guidelines and requirements for auto updates:

• Your solution version must be deployed in a campaign. Your campaign automatically uses the updated solution version.

• Automatic updates are not a full retraining. You should still train a new solution version weekly with `trainingMode` set to `FULL` so the model can learn from your users' behavior.

• If you manually create a new solution version, Amazon Personalize will not automatically update older solution versions, even if you deployed them in a campaign.

• If every two hours is not frequent enough, you can manually create a solution version with `trainingMode` set to `UPDATE` to include those new items in recommendations. Just remember that Amazon Personalize automatically updates only your latest fully trained solution version. The manually updated solution version won't be automatically updated in the future. To resume updates, create a new solution with training mode set to `FULL` and deploy it in a campaign.

For more information about each use case or recipe that features automatic updates, see the following:

• [Recommended for you (ECOMMERCE use case) (p. 112)]
• [Top picks for you (VIDEO_ON_DEMAND use case) (p. 109)]
• [User-Personalization recipe (p. 116)]

**Choosing a use case**

When you create a recommender in a Domain dataset group, you specify a use case. Amazon Personalize trains the models backing the recommender with the best configurations for the use case. Each domain has different use cases. For example, if you specify `VIDEO_ON_DEMAND` for your Domain dataset group, only `VIDEO_ON_DEMAND` use cases are available. Each use case has different requirements for getting recommendations. Some use cases require specific event types. You are free to include additional event types.

For all use cases, your interactions data must have the following:

• At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.

• At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

**Topics**

• [VIDEO_ON_DEMAND use cases (p. 108)]
• [ECOMMERCE use cases (p. 110)]
VIDEO_ON_DEMAND use cases

The following sections list the requirements and Amazon Resource Name (ARN) for each VIDEO_ON_DEMAND use case. For all use cases, your interactions data must have the following:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
- At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

**Note**

If you use the CreateRecommender (p. 461) API, provide the ARN listed here for the recipe ARN.

**Topics**

- Because you watched X (p. 108)
- More like X (p. 108)
- Most popular (p. 109)
- Trending now (p. 109)
- Top picks for you (p. 109)

**Because you watched X**

Get recommendations for videos that other users also watched based on a video that you specify. With this use case, Amazon Personalize automatically filters videos the user watched based on the userId you specify and Watch events. If you apply your own filter, your filter is applied after the videos the user watched are filtered out.

When filtering, Amazon Personalize considers at most 100 interactions per user per event type. This applies to any automatic or custom filters. You can use the Service Quotas console to request an increase for this limit. For more information, see the Requesting a quota increase section of the Service Quotas User Guide.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-vod-because-you-watched-x
- **GetRecommendations API requirements:**
  - **userId:** Required
  - **itemId:** Required
  - **Datasets used when training:** Only Interactions dataset (required)
  - **Required event types:** At minimum, 1000 Watch events.

**More like X**

Get recommendations for videos that are similar to a video that you specify. With this use case, Amazon Personalize automatically filters videos the user watched based on the userId that you specify and Watch events. If you apply your own filter, your filter is applied after the videos the user watched are filtered out.

When filtering, Amazon Personalize considers at most 100 interactions per user per event type. This applies to any automatic or custom filters. You can use the Service Quotas console to request an increase
for this limit. For more information, see the Requesting a quota increase section of the Service Quotas User Guide.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-vod-more-like-x
- **GetRecommendations API requirements:**
  
  user Id: Required

  itemId: Required

- **Datasets used when training:**
  
  Interactions (required)

  Items (required)

- **Required number of events:** At minimum, 1000 events of any type.

- **Recommended event types:** Watch and Click events.

### Most popular

Get recommendations for videos that have been watched by the most users.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-vod-most-popular
- **GetRecommendations requirements:**

  user Id: Required

  itemId: Not used

- **Datasets used when training:** Only Interactions dataset (required)

- **Required event types:** At minimum, 1000 Watch events.

### Trending now

Get recommendations for videos that are currently trending. Trending videos are items that are rapidly becoming more popular with your users. Every two hours, Amazon Personalize automatically evaluates your interactions data and identifies trending items.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-vod-trending-now
- **GetRecommendations API requirements:**

  user Id: Required only if you filter by CurrentUser or by items a user has interacted with

  itemId: Not used

- **Datasets used when training:** Only Interactions dataset (required)

- **Required number of events:** At minimum, 1000 events of any type.

### Top picks for you

Get personalized content recommendations for a user that you specify. With this use case, Amazon Personalize automatically filters videos the user watched based on the userId that you specify and Watch events. If you apply your own filter, your filter is applied after the videos the user watched are filtered out.

When filtering, Amazon Personalize considers at most 100 interactions per user per event type. This applies to any automatic or custom filters. You can use the Service Quotas console to request an increase
for this limit. For more information, see the Requesting a quota increase section of the Service Quotas User Guide.

When recommending items, this use case uses real-time-personalization (p. 105) and exploration (p. 106). And it uses automatic updates (p. 106) to consider new items for recommendations.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-vod-top-picks
- **GetRecommendations requirements:**
  - `userId`: Required
  - `itemId`: Not used

**Datasets used when training:**
- Interactions (required)
- Items (optional)
- Users (optional)

**Required number of events:** At minimum, 1000 events.

**Recommended event types:** Click and Watch events.

**Exploration configuration parameters:** When you create a recommender, you can configure exploration with the following.

- Emphasis on exploring less relevant items (exploration weight) – Configure how much to explore. Specify a decimal value between 0 to 1. The default is 0.3. The closer the value is to 1, the more exploration. With more exploration, recommendations include more items with less interactions data or relevance based on previous behavior. At zero, no exploration occurs and recommendations are based on current data (relevance).

- Exploration item age cutoff – Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines item age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines item age, see Creation timestamp data (p. 81).

  To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them.

## ECOMMERCE use cases

The following sections list the requirements and Amazon Resource Name (ARN) for each ECOMMERCE use case. For all use cases, your interactions data must have the following:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
- At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

**Note**
If you use the CreateRecommender (p. 461) API, provide the ARN listed here for the recipe ARN.
• Most viewed (p. 111)
• Best sellers (p. 111)
• Frequently bought together (p. 111)
• Customers who viewed X also viewed (p. 112)
• Recommended for you (p. 112)

Most viewed

Get recommendations for popular items based on how many times your customers viewed an item.

• Recipe ARN: arn:aws:personalize::recipe/aws-ecomm-popular-items-by-views
• GetRecommendations requirements:
  
  
  userId: Required
  
  itemID: Not used
  
  inputList: NA

  • Datasets used when training: Only Interactions dataset (required)
  • Required event types: At minimum, 1000 View events.

Best sellers

Get recommendations for popular items based on how many times your customers purchased an item.

• Recipe ARN: arn:aws:personalize::recipe/aws-ecomm-popular-items-by-purchases
• GetRecommendations requirements:
  
  userId: Required
  
  itemID: Not used
  
  inputList: NA

  • Datasets used when training: Only Interactions dataset (required)
  • Required event types: At minimum, 1000 Purchase events.

Frequently bought together

Get recommendations for items that customers frequently buy together along with an item that you specify.

• Recipe ARN: arn:aws:personalize::recipe/aws-ecomm-frequently-bought-together
• GetRecommendations requirements:
  
  userId: Required only if you filter by CurrentUser
  
  itemID: Required
  
  inputList: NA

  • Datasets used when training: Only Interactions dataset (required)
  • Required event types: At minimum, 1000 Purchase events.
Customers who viewed X also viewed

Get recommendations for items that customers also viewed based on an item that you specify. With this use case, Amazon Personalize automatically filters items the user purchased based on the userId that you specify and Purchase events. If you apply your own filter, your filter is applied after the items the user already purchased are filtered out.

When filtering, Amazon Personalize considers at most 100 interactions per user per event type. This applies to any automatic or custom filters. You can use the Service Quotas console to request an increase for this limit. For more information, see the Requesting a quota increase section of the Service Quotas User Guide.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-ecomm-customers-who-viewed-x-also-viewed
- **GetRecommendations requirements:**
  - userId: Required
  - itemId: Required
  - inputList: NA
- **Datasets used when training:** Only Interactions dataset (required)
- **Required event types:** At minimum, 1000 View events.
- **Recommended event types:** Purchase events.

Recommended for you

Get personalized recommendations for items based on a user that you specify. With this use case, Amazon Personalize automatically filters out items the user purchased based on the userId that you specify and Purchase events. If you apply your own filter, your filter is applied after the items the user already purchased are filtered out.

When filtering, Amazon Personalize considers at most 100 interactions per user per event type. This applies to any automatic or custom filters. You can use the Service Quotas console to request an increase for this limit. For more information, see the Requesting a quota increase section of the Service Quotas User Guide.

When recommending items, this use case uses real-time-personalization (p. 105) and exploration (p. 106). And it uses automatic updates (p. 106) to consider new items for recommendations.

- **Recipe ARN:** arn:aws:personalize:::recipe/aws-ecomm-recommended-for-you
- **GetRecommendations requirements:**
  - userId: Required
  - itemId: Not used
  - inputList: NA
- **Datasets used when training:**
  - Interactions (required)
  - Items (optional)
  - Users (optional)
- **Required number of events:** At minimum, 1000 events.
- **Recommended event types:** View and Purchase events.
• **Exploration configuration parameters:** When you create a recommender, you can configure exploration with the following.

  - Emphasis on exploring less relevant items (exploration weight) – Configure how much to explore. Specify a decimal value between 0 to 1. The default is 0.3. The closer the value is to 1, the more exploration. With more exploration, recommendations include more items with less interactions data or relevance based on previous behavior. At zero, no exploration occurs and recommendations are based on current data (relevance).
  
  - Exploration item age cutoff – Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines item age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines item age, see [Creation timestamp data](p. 81).

  To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them.

### Choosing a recipe

When you create a custom solution, you specify a recipe and configure training parameters. *Recipes* are Amazon Personalize algorithms that are prepared for specific use cases. Amazon Personalize provides recipes, based on common use cases, for training models. When you create a solution version for the solution, Amazon Personalize trains the models backing the solution version based on the recipe and training configuration.

Amazon Personalize recipes use the following during training:

- Predefined attributes of your data
- Predefined feature transformations
- Predefined algorithms
- Initial parameter settings for the algorithms

To optimize your model, you can override many of these parameters when you create a solution. For more information, see [Hyperparameters and HPO](p. 220).

**Topics**

- [Amazon Personalize recipe types by use case](p. 113)
- [Amazon Personalize recipes](p. 114)
- [Viewing available Amazon Personalize recipes](p. 115)
- [USER_PERSONALIZATION recipes](p. 115)
- [POPULAR_ITEMS](p. 136)
- [PERSONALIZED_RANKING recipes](p. 139)
- [RELATED_ITEMS recipes](p. 143)
- [USER_SEGMENTATION recipes](p. 149)

### Amazon Personalize recipe types by use case

To choose your recipe, first choose your use case from the following and note its corresponding recipe type.
- Recommending items for users (USER_PERSONALIZATION recipes)

To provide personalized recommendations for your users, train your model with a USER_PERSONALIZATION recipe. Personalized recommendations help drive better engagement and conversion.

- Ranking items for a user (PERSONALIZED_RANKING recipes)

To personalize the order of curated lists or search results for your users, train your model with a PERSONALIZED_RANKING recipe. PERSONALIZED_RANKING recipes create a personalized list by re-ranking a collection of input items based on predicted interest level for a given user. Personalized lists improve the customer experience and increase customer loyalty and engagement.

- Recommending trending or popular items (POPULAR_ITEMS recipes)

To recommend trending or popular items use a POPULAR_ITEMS recipe. You might use a POPULAR_ITEMS if your customers highly value what other users are interacting with. Common uses include recommending viral social media content, breaking news articles, or recent sports videos.

- Recommending similar items (RELATED_ITEMS recipes)

To recommend similar items, such as items frequently bought together or movies that other users have also watched, you should use a RELATED_ITEMS recipe. Recommending similar items can help your customers discover items and can increase user conversion rate.

- Getting user segments (USER_SEGMENTATION recipes)

To get segments of users based on item input data, such as users who will most likely interact with items with a certain attribute, you should use a USER_SEGMENTATION recipe. Getting user segments can help you create advanced marketing campaigns that promote different items to different user segments based on the likelihood that they will take an action.

### Amazon Personalize recipes

Amazon Personalize provides the following types of recipes. Besides behavioral differences, each type has different requirements for getting recommendations, as shown in the following table.

<table>
<thead>
<tr>
<th>Recipe type</th>
<th>Recipes</th>
<th>API</th>
<th>API requirements</th>
</tr>
</thead>
<tbody>
<tr>
<td>USER_PERSONALIZATION</td>
<td>Personalization (p. 116)</td>
<td>GetRecommendations (p. 608)</td>
<td>userId: Required</td>
</tr>
<tr>
<td></td>
<td>HRNN recipe (legacy) (p. 125)</td>
<td></td>
<td>itemId: Not used</td>
</tr>
<tr>
<td></td>
<td>HRNN-Metadata recipe (legacy) (p. 129)</td>
<td></td>
<td>inputList: NA</td>
</tr>
<tr>
<td></td>
<td>HRNN-Coldstart recipe (legacy) (p. 132)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>POPULAR_ITEMS</td>
<td>Trending-Now (p. 136)</td>
<td>GetRecommendations (p. 608)</td>
<td>userId: Required only if you apply a filter that requires it</td>
</tr>
<tr>
<td></td>
<td>Popularity-Count (p. 138)</td>
<td></td>
<td>itemId: Not used</td>
</tr>
</tbody>
</table>
Recipe type | Recipes | API | API requirements |
---|---|---|---|
PERSONALIZED_RANKING | Personalized-Ranking (p. 139) | GetPersonalizedRanking (p. 604) | inputList: NA userId: Required itemId: NA |
RELATED_ITEMS | Similar-Items (p. 143) SIMS (p. 145) | GetRecommendations (p. 608) | userId: Required only if you apply a filter that requires it itemId: Required inputList: list of itemId's |
USER_SEGMENTATION | Item-Affinity (p. 149) Item-Attribute-Affinity (p. 150) | CreateBatchSegmentJob (p. 428) | For batch workflow requirements, see Creating a batch segment job (p. 272). |

Viewing available Amazon Personalize recipes

To see a list of available recipes:

- In the Amazon Personalize console, choose a dataset group. From the navigation pane, choose Solutions and recipes, and choose the Recipes tab.
- With the AWS SDK for Python (Boto3), call the ListRecipes (p. 562) API.
- With the AWS CLI, use the following command.

```
aws personalize list-recipes
```

To get information about a recipe using the SDK for Python (Boto3), call the DescribeRecipe (p. 519) API. To get information about a recipe using the AWS CLI, use the following command.

```
aws personalize describe-recipe --recipe-arn recipe_arn
```

**USER_PERSONALIZATION recipes**

USER_PERSONALIZATION recipes predict the items that a user will interact with based on Interactions, Items, and Users datasets. If you are building a recommendation system that provides personalized recommendations for each of your users, you should train your model with a USER_PERSONALIZATION recipe.
USER_PERSONALIZATION recipes are as follows:

- User-Personalization recipe (p. 116)
- Legacy user personalization recipes (p. 125)

User-Personalization recipe

The User-Personalization (aws-user-personalization) recipe is optimized for all personalized recommendation scenarios. It predicts the items that a user will interact with based on Interactions, Items, and Users datasets. When recommending items, it uses automatic item exploration.

With exploration, recommendations include some items that would be typically less likely to be recommended for the user, such as new items, items with few interactions, or items less relevant for the user based on their previous behavior. This improves item discovery and engagement when you have a fast-changing catalog, or when new items, such as news articles or promotions, are more relevant to users when fresh.

When you deploy your solution version with a campaign, you can balance how much to explore (where items with less interactions data or relevance are recommended more frequently) against how much to exploit (where recommendations are based on what we know or relevance).

Topics

- Automatic updates (p. 116)
- Working with impressions data (p. 117)
- Properties and hyperparameters (p. 117)
- Training with the User-Personalization recipe (console) (p. 121)
- Training with the User-Personalization recipe (Python SDK) (p. 123)
- Getting recommendations and recording impressions (SDK for Python (Boto3)) (p. 124)
- Sample Jupyter notebook (p. 125)

Automatic updates

With User-Personalization, for real-time recommendations, Amazon Personalize automatically updates the latest model (solution version) every two hours behind the scenes to include new items in recommendations through exploration. For batch item recommendations, Amazon Personalize updates the solution version you specify in the batch inference job when the solution version is the latest for your solution.

With each update, Amazon Personalize updates the solution version to consider any new items through exploration. And it uses any new interactions data, including impressions data, to determine what items to include or not include in exploration. This is not a full retraining; you should still train a new solution version weekly with trainingMode set to FULL so the model can learn from your users' behavior and any item metadata.

There is no cost for automatic updates. For real-time recommendations, the solution version must be deployed with an Amazon Personalize campaign (p. 237) for updates to occur. Your campaign automatically uses the updated solution version. No new solution version is created when an auto update completes and no new model metrics are generated. This is because no full retraining occurs. If you create a new solution version, Amazon Personalize will not automatically update older solution versions, even if you have deployed them in a campaign. Updates also do not occur if you have deleted your dataset.

If every two hours is not frequent enough, you can manually create a solution version with trainingMode set to UPDATE to include those new items in recommendations. Just remember that
Amazon Personalize automatically updates only your latest fully trained solution version, so the manually updated solution version won't be automatically updated in the future.

**Automatic update requirements**

Automatic update requirements for real time recommendations include the following:

- You must deploy the solution version with a campaign (for more information see [Creating a campaign (p. 237)](Creating%20a%20campaign%20(p.%20237))). The campaign automatically uses the latest automatically updated solution version.
- The solution version must be trained with `trainingMode` set to `FULL` (this is the default when creating a solution version).
- You must provide new item or interactions data since the last automatic update.

Automatic update requirements for batch item recommendations include the following:

- The solution version you specify in the batch inference job must be the latest solution version for your solution.
- The solution version must be trained with `trainingMode` set to `FULL` (this is the default when creating a solution version).
- You must provide new item or interactions data since the last automatic update.

**Working with impressions data**

Unlike other recipes, which solely use positive interactions (clicking, watching, or purchasing), the User-Personalization recipe can also use impressions data. Impressions are lists of items that were visible to a user when they interacted with (clicked, watched, purchased, and so on) a particular item.

Amazon Personalize uses impressions data to determine what items to include in exploration. The more frequently an item occurs in impressions data, the less likely it is that Amazon Personalize includes the item in exploration. [Impressions data (p. 79)].

**Properties and hyperparameters**

The User-Personalization recipe has the following properties:

- **Name** – `aws-user-personalization`
- **Recipe Amazon Resource Name (ARN)** – `arn:aws:personalize:::recipe/aws-user-personalization`
- **Algorithm ARN** – `arn:aws:personalize:::algorithm/aws-user-personalization`

For more information, see [Choosing a recipe (p. 113)](Choosing%20a%20recipe%20(p.%20113)).

The following table describes the hyperparameters for the User-Personalization recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. Featurization hyperparameters control how to filter the data to use in training. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see [Hyperparameters and HPO (p. 220)](Hyperparameters%20and%20HPO%20(p.%20220)).

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in HPO?
<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>hidden_dimension</td>
<td>The number of hidden variables used in the model. Hidden variables recreate users’ purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process. To decide on the best value, use HPO. To use HPO, set <code>performHPO</code> to <code>true</code> when you call <code>CreateSolution</code> and <code>CreateSolutionVersion</code> operations. Default value: 149 Range: [32, 256] Value type: Integer HPO tunable: Yes</td>
</tr>
<tr>
<td>bptt</td>
<td>Determines whether to use the back-propagation through time technique. Back-propagation through time is a technique that updates weights in recurrent neural network-based algorithms. Use <code>bptt</code> for long-term credits to connect delayed rewards to early events. For example, a delayed reward can be a purchase made after several clicks. An early event can be an initial click. Even within the same event types, such as a click, it’s a good idea to consider long-term effects and maximize the total rewards. To consider long-term effects, use larger <code>bptt</code> values. Using a larger <code>bptt</code> value requires larger datasets and more time to process. Default value: 32 Range: [2, 32] Value type: Integer HPO tunable: Yes</td>
</tr>
<tr>
<td>recency_mask</td>
<td>Determines whether the model should consider the latest popularity trends in the Interactions dataset. Latest popularity trends might include sudden changes in the underlying patterns of interaction events. To train a model that places more weight on recent events, set <code>recency_mask</code> to <code>true</code>. To train a model that equally weights all past interactions, set <code>recency_mask</code> to <code>false</code>. To get good recommendations using an equal weight, you might need a larger training dataset. Default value: <code>True</code> Range: <code>True</code> or <code>False</code> Value type: Boolean</td>
</tr>
</tbody>
</table>
## Name

<table>
<thead>
<tr>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>HPO tunable: Yes</td>
</tr>
</tbody>
</table>

## Featurization hyperparameters

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
</table>
| `min_user_history_length_percentile` | The minimum percentile of user history lengths to include in model training. History length is the total amount of data about a user. Use `min_user_history_length_percentile` to exclude a percentage of users with short history lengths. Users with a short history often show patterns based on item popularity instead of the user's personal needs or wants. Removing them can train models with more focus on underlying patterns in your data. Choose an appropriate value after you review user history lengths, using a histogram or similar tool. We recommend setting a value that retains the majority of users, but removes the edge cases.  

   For example, setting `min_user_history_length_percentile` to 0.05 and `max_user_history_length_percentile` to 0.95 includes all users except those with history lengths at the bottom or top 5%.  

Default value: 0.0  
Range: [0.0, 1.0]  
Value type: Float  
HPO tunable: No |
<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>max_user_history_length_percentile</td>
<td>The maximum percentile of user history lengths to include in model training. History length is the total amount of data about a user. Use max_user_history_length_percentile to exclude a percentage of users with long history lengths because data for these users tend to contain noise. For example, a robot might have a long list of automated interactions. Removing these users limits noise in training. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of users but removes the edge cases. For example, setting min_user_history_length_percentile to 0.05 and max_user_history_length_percentile to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.99 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
<tr>
<td>exploration_weight</td>
<td>Determines how frequently recommendations include items with less interactions data or relevance. The closer the value is to 1.0, the more exploration. At zero, no exploration occurs and recommendations are based on current data (relevance). For more information see the section called “CampaignConfig” (p. 638). Default value: 0.3 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
</tbody>
</table>
### Name and Description

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>exploration_item_age_cut_off</td>
<td>Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines an item's age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines an item's age, see Creation timestamp data (p. 81). To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them. Default value: 30.0 Range: Positive floats Value type: Float HPO tunable: No</td>
</tr>
</tbody>
</table>

### Training with the User-Personalization recipe (console)

To use the User-Personalization recipe to generate recommendations in the console, first train a new solution version using the recipe. Then deploy a campaign using the solution version and use the campaign to get recommendations.

#### Training a new solution version with the User-Personalization recipe (console)

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign into your account.
2. Create a Custom dataset group with a new schema and upload your dataset with impressions data. Optionally include CREATION_TIMESTAMP and Unstructured text metadata (p. 81) data in your Items dataset so Amazon Personalize can more accurately calculate the age of an item and identify cold items.

   For more information on importing data, see Step 2: Preparing and importing data (p. 159).
3. On the Dataset groups page, choose the new dataset group that contains the dataset or datasets with impressions data.
4. In the navigation pane, choose Solutions and recipes and choose Create solution.
5. On the Create solution page, for the Solution name, enter the name of your new solution.
6. For Solution type, choose Item recommendation to get item recommendations for your users.
7. For Recipe, choose aws-user-personalization. The Solution configuration section appears providing several configuration options.
8. In Solution configuration, if your Interactions dataset has EVENT_TYPE or both EVENT_TYPE and EVENT_VALUE columns, optionally use the Event type and Event value threshold fields to choose the interactions data that Amazon Personalize uses when training the model.

   For more information see Choosing the interactions data used for training (p. 223).
9. Optionally configure hyperparameters for your solution. For a list of User-Personalization recipe properties and hyperparameters, see Properties and hyperparameters (p. 117).

10. Choose Create and train solution to start training. The Dashboard page displays.

You can navigate to the solution details page to track training progress in the Solution versions section. When training is complete, the status is Active.

Creating a campaign and getting recommendations (console)

When your solution version status is Active you are ready to create your campaign and get recommendations as follows:

1. On either the solution details page or the Campaigns page, choose Create new campaign.
2. On the Create new campaign page, for Campaign details, provide the following information:
   - Campaign name: Enter the name of the campaign. The text you enter here appears on the Campaign dashboard and details page.
   - Solution: Choose the solution that you just created.
   - Solution version ID: Choose the ID of the solution version that you just created.
   - Minimum provisioned transactions per second: Set the minimum provisioned transactions per second that Amazon Personalize supports. For more information, see the CreateCampaign (p. 432) operation.
3. For Campaign configuration, provide the following information:
   - Exploration weight: Configure how much to explore, where recommendations include items with less interactions data or relevance more frequently the more exploration you specify. The closer the value is to 1, the more exploration. At zero, no exploration occurs and recommendations are based on current data (relevance).
   - Exploration item age cut off: Enter the maximum item age, in days since the latest interaction, to define the scope of item exploration. To increase the number of items Amazon Personalize considers during exploration, enter a greater value.

   For example, if you enter 10, only items with interactions data from the 10 days since the latest interaction in the dataset are considered during exploration.

   Note
   Recommendations might include items without interactions data from outside this time frame. This is because these items are relevant to the user's interests, and exploration wasn't required to identify them.

4. Choose Create campaign.
5. On the campaign details page, when the campaign status is Active, you can use the campaign to get recommendations and record impressions. For more information, see Step 5: Get recommendations (p. 54) in "Getting Started."

   Amazon Personalize automatically updates your latest solution version every two hours to include new data. Your campaign automatically uses the updated solution version. For more information see Automatic updates (p. 116).

   To manually update the campaign, you first create and train a new solution version using the console or the CreateSolutionVersion (p. 473) operation, with trainingMode set to update. You then manually update the campaign on the Campaign page of the console or by using the UpdateCampaign (p. 588) operation.

   Note
   Amazon Personalize doesn't automatically update solution versions you created before November 17, 2020.
Training with the User-Personalization recipe (Python SDK)

When you have created a dataset group and uploaded your dataset(s) with impressions data, you can train a solution with the User-Personalization recipe. Optionally include creation_timestamp and unstructured text metadata (p. 81) data in your items dataset so Amazon Personalize can more accurately calculate the age of an item and identify cold items. For more information on creating dataset groups and uploading training data see Schemas (p. 82).

To train a solution with the User-Personalization recipe using the AWS SDK

1. Create a new solution using the create_solution method.

   Replace solution name with your solution name and dataset group arn with the Amazon Resource Name (ARN) of your dataset group.

   ```python
   import boto3
   personalize = boto3.client('personalize')
   print('Creating solution')
   create_solution_response = personalize.create_solution(name = 'solution name',
   recipeArn = 'arn:aws:personalize:::recipe/aws-user-
   personalization',
   datasetGroupArn = 'dataset group arn',
   )
   solution_arn = create_solution_response['solutionArn']
   print('solution_arn: ', solution_arn)
   ```

   For a list of aws-user-personalization recipe properties and hyperparameters, see Properties and hyperparameters (p. 117).

2. Create a new solution version with the updated training data and set trainingMode to FULL using the following code snippet. Replace the solution arn with the ARN of your solution.

   ```python
   import boto3
   personalize = boto3.client('personalize')
   create_solution_version_response = personalize.create_solution_version(solutionArn = 'solution arn',
   trainingMode='FULL')
   new_solution_version_arn = create_solution_version_response['solutionVersionArn']
   print('solution_version_arn: ', new_solution_version_arn)
   ```

3. When Amazon Personalize is finished creating your solution version, create your campaign with the following parameters:

   - Provide a new campaign name and the solution version arn generated in step 2.
   - Modify the explorationWeight item exploration configuration hyperparameter to configure how much to explore. Items with less interactions data or relevance are recommended more frequently the closer the value is to 1.0. The default value is 0.3.
   - Modify the explorationItemAgeCutOff item exploration configuration hyperparameter parameter to provide the maximum duration, in days relative to the latest interaction, for which items should be explored. The larger the value, the more items are considered during exploration.

   Use the following Python snippet to create a new campaign with an emphasis on exploration with exploration cut-off at 30 days. Creating a campaign usually takes a few minutes but can take over an hour.
import boto3
personalize = boto3.client('personalize')
campaign_response = personalize.create_campaign(
    name = 'campaign name',
    solutionVersionArn = 'solution version arn',
    minProvisionedTPS = 1,
    campaignConfig = 
    {"itemExplorationConfig": {"explorationWeight": "0.3",
    "explorationItemAgeCutOff": "30"}}
)
campaign_arn = create_campaign_response['campaignArn']
print('campaign_arn:', campaign_arn)

With User-Personalization, Amazon Personalize automatically updates your solution version every two hours to include new data. Your campaign automatically uses the updated solution version. For more information see Automatic updates (p. 116).

To manually update the campaign, you first create and train a new solution version using the console or the CreateSolutionVersion (p. 473) operation, with trainingMode set to update. You then manually update the campaign on the Campaign page of the console or by using the UpdateCampaign (p. 588) operation.

Note
Amazon Personalize doesn't automatically update solution versions you created before November 17, 2020.

Getting recommendations and recording impressions (SDK for Python (Boto3))

When your campaign is created, you can use it to get recommendations for a user and record impressions. For information on getting batch recommendations using the AWS SDKs see Creating a batch inference job (AWS SDKs) (p. 265).

To get recommendations and record impressions

1. Call the get_recommendations method. Change the campaign_arn to the ARN of your new campaign and user_id to the userId of the user.

import boto3
rec_response = personalize_runtime.get_recommendations(campaignArn = 'campaign arn',
user_id = 'user id')
print(rec_response['recommendationId'])

2. Create a new event tracker for sending PutEvents requests. Replace event_tracker_name with the name of your event tracker and dataset_group_arn with the ARN of your dataset group.

import boto3
personalize = boto3.client('personalize')

event_tracker_response = personalize.create_event_tracker(
    name = 'event tracker name',
    datasetGroupArn = 'dataset group arn'
)
event_tracker_arn = event_tracker_response['eventTrackerArn']
event_tracking_id = event_tracker_response['trackingId']
print('eventTrackerArn:{},
 eventTrackingId:{}'.format(event_tracker_arn, event_tracking_id))

3. Use the recommendationId from step 1 and the event tracking id from step 2 to create a new PutEvents request. This request logs the new impression data from the user's session. Change the user id to the ID of the user.

```python
import boto3
personalize_events.put_events(
    trackingId = 'event tracking id',
    userId= 'user id',
    sessionId = '1',
    eventList = [{'sentAt': datetime.now().timestamp(),
                  'eventType': 'click',
                  'itemId': rec_response['itemList'][0]['itemId'],
                  'recommendationId': rec_response['recommendationId'],
                  'impression': [item['itemId'] for item in rec_response['itemList']],
               }]
)
```

Sample Jupyter notebook

For a sample Jupyter notebook that shows how to use the User-Personalization recipe, see User Personalization with Exploration.

Legacy user personalization recipes

**Note**

Legacy HRNN recipes are no longer available. This documentation is for reference purposes. We recommend using the aws-user-personalization (User-Personalization) recipe over the legacy HRNN recipes. User-Personalization improves upon and unifies the functionality offered by the HRNN recipes. For more information, see User-Personalization recipe (p. 116).

The following are legacy USER_PERSONALIZATION recipes.

- **HRNN recipe (legacy)** (p. 125)
- **HRNN-Coldstart recipe (legacy)** (p. 132)
- **HRNN-Metadata recipe (legacy)** (p. 129)

**HRNN recipe (legacy)**

**Note**

Legacy HRNN recipes are no longer available. This documentation is for reference purposes. We recommend using the aws-user-personalization (User-Personalization) recipe over the legacy HRNN recipes. User-Personalization improves upon and unifies the functionality offered by the HRNN recipes. For more information, see User-Personalization recipe (p. 116).

The Amazon Personalize hierarchical recurrent neural network (HRNN) recipe models changes in user behavior to provide recommendations during a session. A session is a set of user interactions within a given timeframe with a goal of finding a specific item to fill a need, for example. By weighing a user's recent interactions higher, you can provide more relevant recommendations during a session.

HRNN accommodates user intent and interests, which can change over time. It takes ordered user histories and automatically weights them to make better inferences. HRNN uses a gating mechanism to model the discount weights as a learnable function of the items and timestamps.
Amazon Personalize derives the features for each user from your dataset. If you have done real-time data integration, these features are updated in real time according to user activity. To get a recommendation, you provide only the USER_ID. If you also provide an ITEM_ID, Amazon Personalize ignores it.

The HRNN recipe has the following properties:

- **Name** – aws-hrnn
- **Recipe Amazon Resource Name (ARN)** – arn:aws:personalize:::recipe/aws-hrnn
- **Algorithm ARN** – arn:aws:personalize:::algorithm/aws-hrnn
- **Feature transformation ARN** – arn:aws:personalize:::feature-transformation/JSON-percentile-filtering
- **Recipe type** – USER_PERSONALIZATION

The following table describes the hyperparameters for the HRNN recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. Featurization hyperparameters control how to filter the data to use in training. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see [Hyperparameters and HPO](p. 220).

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in HPO?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Algorithm hyperparameters</strong></td>
<td></td>
</tr>
</tbody>
</table>
| hidden_dimension   | The number of hidden variables used in the model. *Hidden variables* recreate users' purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process. To decide on the optimal value, use HPO. To use HPO, set `performHPO` to `true` when you call `CreateSolution` (p. 468) and `CreateSolutionVersion` (p. 473) operations.  
  Default value: 43  
  Range: [32, 256]  
  Value type: Integer  
  HPO tunable: Yes |
<p>| bptt               | Determines whether to use the back-propagation through time technique. <em>Back-propagation through time</em> is a technique that updates weights in recurrent neural network-based algorithms. Use <em>bptt</em> for long-term credits to connect delayed rewards to early events. For example, a delayed reward can be a purchase made after several clicks. An early event can be an initial click. Even |</p>
<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
<th>Default value</th>
<th>Range</th>
<th>Value type</th>
<th>HPO tunable</th>
</tr>
</thead>
</table>
| within the same event types, such as a click, it's a good idea to consider long-term effects and maximize the total rewards. To consider long-term effects, use larger bptt values. Using a larger bptt value requires larger datasets and more time to process. | Default value: 32  
Range: [2, 32]  
Value type: Integer  
HPO tunable: Yes | bptt          |             | Integer    | Yes         |
| recency_mask     | Determines whether the model should consider the latest popularity trends in the Interactions dataset. Latest popularity trends might include sudden changes in the underlying patterns of interaction events. To train a model that places more weight on recent events, set recency_mask to true. To train a model that equally weighs all past interactions, set recency_mask to false. To get good recommendations using an equal weight, you might need a larger training dataset. | Default value: True  
Range: True or False  
Value type: Boolean  
HPO tunable: Yes |             |             | Boolean    | Yes         |

**Featurization hyperparameters**
<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>min_user_history_length_percentile</strong></td>
<td>The minimum percentile of user history lengths to include in model training. <strong>History length</strong> is the total amount of data about a user. Use <strong>min_user_history_length_percentile</strong> to exclude a percentage of users with short history lengths. Users with a short history often show patterns based on item popularity instead of the user's personal needs or wants. Removing them can train models with more focus on underlying patterns in your data. Choose an appropriate value after you review user history lengths, using a histogram or similar tool. We recommend setting a value that retains the majority of users, but removes the edge cases. For example, setting <strong>min_user_history_length_percentile</strong> to 0.05 and <strong>max_user_history_length_percentile</strong> to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.0 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
<tr>
<td><strong>max_user_history_length_percentile</strong></td>
<td>The maximum percentile of user history lengths to include in model training. <strong>History length</strong> is the total amount of data about a user. Use <strong>max_user_history_length_percentile</strong> to exclude a percentage of users with long history lengths because data for these users tend to contain noise. For example, a robot might have a long list of automated interactions. Removing these users limits noise in training. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of users but removes the edge cases. For example, setting <strong>min_user_history_length_percentile</strong> to 0.05 and <strong>max_user_history_length_percentile</strong> to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.99 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
</tbody>
</table>
HRNN-Metadata recipe (legacy)

Note
Legacy HRNN recipes are no longer available. This documentation is for reference purposes. We recommend using the aws-user-personalizaton (User-Personalization) recipe over the legacy HRNN recipes. User-Personalization improves upon and unifies the functionality offered by the HRNN recipes. For more information, see User-Personalization recipe (p. 116).

The HRNN-Metadata recipe predicts the items that a user will interact with. It is similar to the HRNN (p. 125) recipe, with additional features derived from contextual, user, and item metadata (from Interactions, Users, and Items datasets, respectively). HRNN-Metadata provides accuracy benefits over non-metadata models when high quality metadata is available. Using this recipe might require longer training times.

The HRNN-Metadata recipe has the following properties:

- **Name** – aws-hrnn-metadata
- **Recipe Amazon Resource Name (ARN)** – arn:aws:personalize:::recipe/aws-hrnn-metadata
- **Algorithm ARN** – arn:aws:personalize:::algorithm/aws-hrnn-metadata
- **Feature transformation ARN** – arn:aws:personalize:::feature-transformation/featurize_metadata
- **Recipe type** – USER_PERSONALIZATION

The following table describes the hyperparameters for the HRNN-Metadata recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. Featurization hyperparameters control how to filter the data to use in training. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see Hyperparameters and HPO (p. 220).

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in hyperparameter optimization (HPO)?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Algorithm Hyperparameters</strong></td>
<td></td>
</tr>
<tr>
<td>hidden_dimension</td>
<td>The number of hidden variables used in the model. Hidden variables recreate users' purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process. To decide on the optimal value, use HPO. To use HPO, set performHPO to true when you call CreateSolution (p. 468) and CreateSolutionVersion (p. 473) operations.Default value: 43Range: [32, 256]Value type: Integer</td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
</tr>
<tr>
<td>---------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>bptt</td>
<td>Determines whether to use the back-propagation through time technique. <em>Back-propagation through time</em> is a technique that updates weights in recurrent neural network-based algorithms. Use bptt for long-term credits to connect delayed rewards to early events. For example, a delayed reward can be a purchase made after several clicks. An early event can be an initial click. Even within the same event types, such as a click, it's a good idea to consider long-term effects and maximize the total rewards. To consider long-term effects, use larger bptt values. Using a larger bptt value requires larger datasets and more time to process.</td>
</tr>
<tr>
<td>recency_mask</td>
<td>Determines whether the model should consider the latest popularity trends in the Interactions dataset. Latest popularity trends might include sudden changes in the underlying patterns of interaction events. To train a model that places more weight on recent events, set recency_mask to true. To train a model that equally weighs all past interactions, set recency_mask to false. To get good recommendations using an equal weight, you might need a larger training dataset.</td>
</tr>
</tbody>
</table>

**Featurization hyperparameters**
<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>min_user_history_length_percentile</td>
<td>The minimum percentile of user history lengths to include in model training. <em>History length</em> is the total amount of data about a user. Use <em>min_user_history_length_percentile</em> to exclude a percentage of users with short history lengths. Users with a short history often show patterns based on item popularity instead of the user's personal needs or wants. Removing them can train models with more focus on underlying patterns in your data. Choose an appropriate value after you review user history lengths, using a histogram or similar tool. We recommend setting a value that retains the majority of users, but removes the edge cases. For example, setting <em>min__user_history_length_percentile</em> to 0.05 and <em>max_user_history_length_percentile</em> to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.0 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
<tr>
<td>max_user_history_length_percentile</td>
<td>The maximum percentile of user history lengths to include in model training. <em>History length</em> is the total amount of data about a user. Use <em>max_user_history_length_percentile</em> to exclude a percentage of users with long history lengths because data for these users tend to contain noise. For example, a robot might have a long list of automated interactions. Removing these users limits noise in training. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of users but removes the edge cases. For example, setting <em>min__user_history_length_percentile</em> to 0.05 and <em>max_user_history_length_percentile</em> to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.99 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
</tbody>
</table>
HRNN-Coldstart recipe (legacy)

Note
Legacy HRNN recipes are no longer available. This documentation is for reference purposes. We recommend using the aws-user-personalization (User-Personalization) recipe over the legacy HRNN recipes. User-Personalization improves upon and unifies the functionality offered by the HRNN recipes. For more information, see User-Personalization recipe (p. 116).

Use the HRNN-Coldstart recipe to predict the items that a user will interact with when you frequently add new items and interactions and want to get recommendations for those items immediately. The HRNN-Coldstart recipe is similar to the HRNN-Metadata (p. 129) recipe, but it allows you to get recommendations for new items.

In addition, you can use the HRNN-Coldstart recipe when you want to exclude from training items that have a long list of interactions either because of a recent popularity trend or because the interactions might be highly unusual and introduce noise in training. With HRNN-Coldstart, you can filter out less relevant items to create a subset for training. The subset of items, called cold items, are items that have related interaction events in the Interactions dataset. An item is considered a cold item when it has the following:

- Fewer interactions than a specified number of maximum interactions. You specify this value in the recipe's cold_start_max_interactions hyperparameter.
- A shorter relative duration than the maximum duration. You specify this value in the recipe's cold_start_max_duration hyperparameter.

To reduce the number of cold items, set a lower value for cold_start_max_interactions or cold_start_max_duration. To increase the number of cold items, set a greater value for cold_start_max_interactions or cold_start_max_duration.

HRNN-Coldstart has the following cold item limits:
- Maximum cold start items: 80,000
- Minimum cold start items: 100

If the number of cold items is outside this range, attempts to create a solution will fail.

The HRNN-Coldstart recipe has the following properties:

- **Name** – aws-hrnn-coldstart
- **Recipe Amazon Resource Name (ARN)** – arn:aws:personalize:::recipe/aws-hrnn-coldstart
- **Algorithm ARN** – arn:aws:personalize:::algorithm/aws-hrnn-coldstart
- **Feature transformation ARN** – arn:aws:personalize:::feature-transformation/featurize_coldstart
- **Recipe type** – USER_PERSONALIZATION

For more information, see Choosing a recipe (p. 113).

The following table describes the hyperparameters for the HRNN-Coldstart recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. Featureization hyperparameters control how to filter the data to use in training. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see Hyperparameters and HPO (p. 220).

The table also provides the following information for each hyperparameter:
- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in HPO?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
<th>Default value</th>
<th>Range</th>
<th>Value type</th>
<th>HPO tunable</th>
</tr>
</thead>
<tbody>
<tr>
<td>hidden_dimension</td>
<td>The number of hidden variables used in the model. <em>Hidden variables</em> recreate users' purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process. To decide on the optimal value, use HPO. To use HPO, set performHPO to true when you call <code>CreateSolution (p. 468)</code> and <code>CreateSolutionVersion (p. 473)</code> operations.</td>
<td>149</td>
<td>[32, 256]</td>
<td>Integer</td>
<td>Yes</td>
</tr>
<tr>
<td>bptt</td>
<td>Determines whether to use the back-propagation through time technique. <em>Back-propagation through time</em> is a technique that updates weights in recurrent neural network-based algorithms. Use bptt for long-term credits to connect delayed rewards to early events. For example, a delayed reward can be a purchase made after several clicks. An early event can be an initial click. Even within the same event types, such as a click, it's a good idea to consider long-term effects and maximize the total rewards. To consider long-term effects, use larger bptt values. Using a larger bptt value requires larger datasets and more time to process.</td>
<td>32</td>
<td>[2, 32]</td>
<td>Integer</td>
<td>Yes</td>
</tr>
<tr>
<td>recency_mask</td>
<td>Determines whether the model should consider the latest popularity trends in the Interactions dataset. Latest popularity trends might include sudden changes in the underlying patterns of interaction events. To train a model that places more weight on recent events, set recency_mask to true. To train a model that equally weighs all past interactions, set recency_mask to false. To get good recommendations using an equal weight, you might need a larger training dataset.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>-----------------------------</td>
<td>-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Default value: True</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Range: True or False</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Value type: Boolean</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>HPO tunable: Yes</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>cold_start_max_interactions</td>
<td>The maximum number of user-item interactions an item can have to be considered a cold item.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Default value: 15</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Range: Positive integers</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Value type: Integer</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>HPO tunable: No</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>cold_start_max_duration</td>
<td>The maximum duration in days relative to the starting point for a user-item interaction to be considered a cold start item. To set the starting point of the user-item interaction, set the cold_start_relative_from hyperparameter.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Default value: 5.0</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Range: Positive floats</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Value type: Float</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>HPO tunable: No</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>cold_start_relative_from</td>
<td>Determines the starting point for the HRNN-Coldstart recipe to calculate cold_start_max_duration. To calculate from the current time, choose currentTime.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>To calculate cold_start_max_duration from the timestamp of the latest item in the Interactions dataset, choose latestItem. This setting is useful if you frequently add new items.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Default value: latestItem</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Range: currentTime, latestItem</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Value type: String</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>HPO tunable: No</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>------</td>
<td>-------------</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>min_user_history_length_percentile</strong></td>
<td>The minimum percentile of user history lengths to include in model training. <em>History length</em> is the total amount of data about a user. Use <strong>min_user_history_length_percentile</strong> to exclude a percentage of users with short history lengths. Users with a short history often show patterns based on item popularity instead of the user's personal needs or wants. Removing them can train models with more focus on underlying patterns in your data. Choose an appropriate value after you review user history lengths, using a histogram or similar tool. We recommend setting a value that retains the majority of users, but removes the edge cases. For example, setting <strong>min_user_history_length_percentile</strong> to 0.05 and <strong>max_user_history_length_percentile</strong> to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.0 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>max_user_history_length_percentile</strong></td>
<td>The maximum percentile of user history lengths to include in model training. <em>History length</em> is the total amount of data about a user. Use <strong>max_user_history_length_percentile</strong> to exclude a percentage of users with long history lengths because data for these users tend to contain noise. For example, a robot might have a long list of automated interactions. Removing these users limits noise in training. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of users but removes the edge cases. For example, setting <strong>min_user_history_length_percentile</strong> to 0.05 and <strong>max_user_history_length_percentile</strong> to 0.95 includes all users except those with history lengths at the bottom or top 5%. Default value: 0.99 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Using AutoML to choose an HRNN recipe (API only)

Amazon Personalize can automatically choose the most appropriate hierarchical recurrent neural network (HRNN) recipe based on its analysis of the input data. This option is called AutoML. To perform AutoML, set the performAutoML parameter to true when you call the CreateSolution (p. 468) API.

You can also specify the list of recipes that Amazon Personalize examines to determine the optimal recipe, based on a metric you specify. In this case, you call the CreateSolution operation, specify true for the performAutoML parameter, omit the recipeArn parameter, and include the solutionConfig parameter, specifying the metricName and recipeList as part of the autoMLConfig object.

How a recipe is chosen is shown in the following table. Either performAutoML or recipeArn must be specified but not both. AutoML is only performed using the HRNN recipes.

<table>
<thead>
<tr>
<th>performAutoML</th>
<th>recipeArn</th>
<th>solutionConfig</th>
<th>Result</th>
</tr>
</thead>
<tbody>
<tr>
<td>true</td>
<td>omit</td>
<td>omitted</td>
<td>Amazon Personalize chooses the recipe</td>
</tr>
<tr>
<td>true</td>
<td>omit</td>
<td>autoMLConfig:metricName and recipeList specified</td>
<td>Amazon Personalize chooses a recipe from the list that optimizes the metric</td>
</tr>
<tr>
<td>omit</td>
<td>specified</td>
<td>omitted</td>
<td>You specify the recipe</td>
</tr>
<tr>
<td>omit</td>
<td>specified</td>
<td>specified</td>
<td>You specify the recipe and override the default training properties</td>
</tr>
</tbody>
</table>

**Note**

When performAutoML is true, all parameters of the solutionConfig object are ignored except for autoMLConfig.

**POPULAR_ITEMS**

To recommend trending or popular items, such as breaking news articles or popular social media content, use a POPULAR_ITEMS recipe. To generate recommendations for items that are rapidly becoming more popular with your users, use the Trending-Now recipe (p. 136) recipe. To generate a baseline for comparison purposes, we recommend using the Popularity-Count (p. 138) recipe. This POPULAR_ITEMS recipe recommends the most popular items based on counting interactions.

POPULAR_ITEMS recipes are as follows:

- Trending-Now recipe (p. 136)
- Popularity-Count recipe (p. 138)

**Trending-Now recipe**

The Trending-Now recipe (aws-trending-now) generates recommendations for items that are rapidly becoming more popular with your users. You might use the Trending-Now recipe if items gaining in popularity are more relevant to your customers. For example, your customers might highly value what other users are interacting with. Common uses include recommending viral social media content, breaking news articles, or recent sports videos.

Trending-Now automatically identifies the top trending items by calculating the increase in interactions that each item has over configurable intervals of time. The items with highest rate of increase are
considered trending items. The time is based on timestamp data in your Interactions dataset. The items considered come from the interactions data you imported in bulk and incrementally. You don't have to manually create a new solution version for Trending-Now to consider new items in interactions data.

You can specify the time interval by providing a Trend discovery frequency when you create your solution. For example, if you specify 30 minutes for Trend discovery frequency, for every 30 minutes of data, Amazon Personalize identifies the items with the greatest rate of increase in interactions since the last evaluation. Possible frequencies include 30 minutes, 1 hour, 3 hour, and 1 day. Choose a frequency that aligns with the distribution of your interactions data. Missing data over the interval you choose can reduce recommendation accuracy. If you import zero interactions over the last two time intervals, Amazon Personalize recommends only popular items instead of trending items.

With Trending-Now, you call the GetRecommendations (p. 608) operation or get recommendations on the Test campaign page of the Amazon Personalize console. Amazon Personalize returns the top trending items. You pass a userId in your request only if you apply a filter that requires it. With the GetRecommendations API, you can configure the number of trending items returned with the numResults parameter. You can't get batch recommendations with the Trending-Now recipe.

To use Trending-Now, you must create an Interactions dataset with at least 1000 unique historical and event interactions combined (after filtering by eventType and eventValueThreshold, if provided). When generating trending item recommendations, Trending now does not use data in Items or Users datasets. However, you can still filter recommendations based on data in these datasets. For more information, see Filtering recommendations and user segments (p. 311).

Properties and hyperparameters

The Trending-Now recipe has the following properties:

- **Name** – aws-trending-now
- **Recipe Amazon Resource Name (ARN)** – arn:aws:personalize:::recipe/aws-trending-now
- **Algorithm ARN** – arn:aws:personalize:::algorithm/aws-trending-now-custom

For more information, see Choosing a recipe (p. 113).

The following table describes the hyperparameters for the Trending-Now recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see Hyperparameters and HPO (p. 220).

The table also provides the following information for each hyperparameter:

- **Range:** [lower bound, upper bound]
- **Value type:** Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable:** Can the parameter participate in HPO?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
<th>Feature transformation hyperparameters</th>
</tr>
</thead>
</table>

137
### Name | Description
--- | ---
Trend discovery frequency | Specify how often Amazon Personalize evaluates your interactions data and identifies trending items. For example, if you specify 30 min for Trend discovery frequency, every 30 minutes Amazon Personalize identifies items with the greatest rate of increase in interactions over 30-minute intervals.

Available frequencies include 30 minutes, 1 hour, 3 hours, and 1 day. Choose a frequency that aligns with the distribution of your interactions data. Missing data over the interval you choose can reduce recommendation accuracy.

Default value: 3h

Possible values: 30 minutes, 1h, 3h, and 1 day.

Value type: String

HPO tunable: No

---

**Creating a solution (SDK for Python (Boto3))**

The following code shows how to create a solution with the Trending-Now recipe using the SDK for Python (Boto3). Possible values for `trend_discovery_frequency` are 30 minutes, 1 hour, 3 hours, and 1 day. For information about creating a solution with the console, see [Creating a solution (console)](p. 211).

```python
import boto3

personalize = boto3.client("personalize")

create_solution_response = personalize_client.create_solution(
    name="solution name",
    recipeArn="arn:aws:personalize:::recipe/aws-trending-now",
    datasetGroupArn="dataset group ARN",
    solutionConfig={
        "featureTransformationParameters": {
            "trend_discovery_frequency": "1 hour"
        }
    }
)

print(create_solution_response['solutionArn'])
```

---

**Sample Jupyter notebook**

For a sample Jupyter notebook that shows how to use the Trending-Now recipe, see [trending_now_example.ipynb](p. 211) in the Amazon Personalize samples GitHub repository.

---

**Popularity-Count recipe**

Popularity-Count recommends the most popular items based your interactions data. The most popular items are the items with the most interactions data from unique users. The recipe returns the same popular items for all users. Popularity-Count is a good baseline for comparing with other recipes using the evaluation metrics Amazon Personalize generates when you create a solution version. For more information, see [Evaluating a solution version with metrics](p. 232).
After you create a solution version, make sure you keep your solution version and data up to date. With Popularity-Count, you must manually create a new solution version (retrain the model) to reflect updates to your catalog and update the model with your user's most recent behavior. Then you must update any campaign using the solution version. For more information, see Maintaining recommendation relevance (p. 278).

This predefined recipe has the following properties:

- Name – aws-popularity-count
- Recipe ARN – arn:aws:personalize:::recipe/aws-popularity-count
- Algorithm ARN – arn:aws:personalize:::algorithm/aws-popularity-count
- Feature transformation ARN – arn:aws:personalize:::feature-transformation/sims
- Recipe type – USER_PERSONALIZATION

Popularity-Count has no exposed hyperparameters.

PERSONALIZED_RANKING recipes

The PERSONALIZED_RANKING recipe, Personalized-Ranking, provides recommendations in ranked order based on predicted interest level.

Personalized-Ranking (p. 139)

The Personalized-Ranking recipe is a hierarchical recurrent neural network (HRNN) recipe that also can filter and re-rank results. Personalized-Ranking provides a list of the best recommendations. Use the Personalized-Ranking recipe when you're personalizing the results for your users, such as personalized re-ranking of search results or curated lists.

To train a model, the Personalized-Ranking recipe uses the data in your Interactions dataset, and if you created them, the Items dataset and Users dataset in your dataset group.

Personalized-Ranking recipe

The Personalized-Ranking recipe generates personalized rankings of items. A personalized ranking is a list of recommended items that are re-ranked for a specific user. This is useful if you have a collection of ordered items, such as search results, promotions, or curated lists, and you want to provide a personalized re-ranking for each of your users. For example, with Personalized-Ranking, Amazon Personalize can re-rank search results that you generate with OpenSearch (p. 351).

To train a model, the Personalized-Ranking recipe uses the data in your Interactions dataset, and if you created them, the Items dataset and Users dataset in your dataset group (these datasets are optional). With Personalized-Ranking, your Items dataset can include Unstructured text metadata (p. 81) and your Interactions dataset can include Contextual metadata (p. 78). To get a personalized ranking, use the GetPersonalizedRanking (p. 604) API.

After you create a solution version, make sure you keep your solution version and data up to date. With Personalized-Ranking, you must manually create a new solution version (retrain the model) to reflect updates to your catalog and update the model with your user's most recent behavior. Then you must update any campaign using the solution version. For more information, see Maintaining recommendation relevance (p. 278).

Note
If you provide items without interactions data for ranking, Amazon Personalize will return these items without a recommendation score in the GetPersonalizedRanking API response.

This recipe has the following properties:
Hyperparameters

The following table describes the hyperparameters for the Personalize-Ranking recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. Featurization hyperparameters control how to filter the data to use in training. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see Hyperparameters and HPO (p. 220).

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in hyperparameter optimization (HPO)?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
<th>Range</th>
<th>Value type</th>
<th>HPO tunable</th>
</tr>
</thead>
<tbody>
<tr>
<td>hidden_dimension</td>
<td>The number of hidden variables used in the model. <em>Hidden variables</em> recreate users' purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process. To decide on the optimal value, use HPO. To use HPO, set <code>performHPO</code> to <code>true</code> when you call <code>CreateSolution</code> (p. 468) and <code>CreateSolutionVersion</code> (p. 473) operations.</td>
<td>[32, 256]</td>
<td>Integer</td>
<td>Yes</td>
</tr>
<tr>
<td>bptt</td>
<td>Determines whether to use the back-propagation through time technique. <em>Back-propagation through time</em> is a technique that updates weights in recurrent neural network-based algorithms. Use bptt for long-term credits to connect delayed rewards to early events. For example, a delayed reward can be a purchase made after several clicks. An early event can be an initial click. Even within the same event types, such as a click, it's a good idea to consider long-term effects and maximize the total rewards. To consider long-term effects, use larger bptt.</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>-----------------</td>
<td>------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>values. Using a larger bptt value requires larger datasets and more time to process.</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Default value: 32</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Range: [2, 32]</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Value type: Integer</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>HPO tunable: Yes</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>recency_mask</td>
<td>Determines whether the model should consider the latest popularity trends in the Interactions dataset. Latest popularity trends might include sudden changes in the underlying patterns of interaction events. To train a model that places more weight on recent events, set recency_mask to true. To train a model that equally weighs all past interactions, set recency_mask to false. To get good recommendations using an equal weight, you might need a larger training dataset.</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Default value: True</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Range: True or False</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Value type: Boolean</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>HPO tunable: Yes</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Featurization hyperparameters**
### Name

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>min_user_history_length_percentile</td>
<td>The minimum percentile of user history lengths to include in model training. <em>History length</em> is the total amount of data about a user. Use <code>min_user_history_length_percentile</code> to exclude a percentage of users with short history lengths. Users with a short history often show patterns based on item popularity instead of the user's personal needs or wants. Removing them can train models with more focus on underlying patterns in your data. Choose an appropriate value after you review user history lengths, using a histogram or similar tool. We recommend setting a value that retains the majority of users, but removes the edge cases. For example, setting <code>min_user_history_length_percentile</code> to 0.05 and <code>max_user_history_length_percentile</code> to 0.95 includes all users except those with history lengths at the bottom or top 5%.  Default value: 0.0  Range: [0.0, 1.0]  Value type: Float  HPO tunable: No</td>
</tr>
<tr>
<td>max_user_history_length_percentile</td>
<td>The maximum percentile of user history lengths to include in model training. <em>History length</em> is the total amount of data about a user. Use <code>max_user_history_length_percentile</code> to exclude a percentage of users with long history lengths because data for these users tend to contain noise. For example, a robot might have a long list of automated interactions. Removing these users limits noise in training. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of users but removes the edge cases. For example, setting <code>min_user_history_length_percentile</code> to 0.05 and <code>max_user_history_length_percentile</code> to 0.95 includes all users except those with history lengths at the bottom or top 5%.  Default value: 0.99  Range: [0.0, 1.0]  Value type: Float  HPO tunable: No</td>
</tr>
</tbody>
</table>
Personalized-Ranking sample notebook

For a sample Jupyter notebook that shows how to use the Personalized-Ranking recipe, see Personalize Ranking Example.

RELATED_ITEMS recipes

Note

All RELATED_ITEMS recipes use interactions data. Choose the Similar-Items recipe if you have also have item metadata and want Amazon Personalize to use it to find similar items. Or choose the SIMS recipe if you want to configure more hyperparameters for the model.

The RELATED_ITEMS recipes return items similar to an item that you specify when you get recommendations. The RELATED_ITEMS recipes are as follows:

- Similar-Items recipe (p. 143)
- SIMS recipe (p. 145)

Similar-Items (p. 143)

The Similar-Items recipe generates recommendations for items that are similar to an item you specify. It calculates similarity based on both interactions data and, if you provide it, item metadata. If Amazon Personalize can't find the item ID that you specify in your recommendation request, the recipe returns popular items as recommendations. Similar-Items doesn't use data in a Users dataset when generating recommendations. However, you can still filter recommendations based on data in a Users dataset. For more information, see Filtering recommendations and user segments (p. 311).

SIMS (p. 145)

The item-to-item similarities (SIMS) recipe generates items similar to a given item based on the co-occurrence of the item in user history in your Interactions dataset. If sufficient user behavior data for an item isn't available, or if the specified item ID isn't found, the recipe returns popular items as recommendations.

Similar-Items recipe

Note

All RELATED_ITEMS recipes use interactions data. Choose Similar-Items if you have also have item metadata and want Amazon Personalize to use it to find similar items. Or choose the SIMS recipe (p. 145) if you want to configure more hyperparameters for the model.

The Similar-Items (aws-similar-items) recipe generates recommendations for items that are similar to an item you specify. Use Similar-Items to help customers discover new items in your catalog based on their previous behavior and item metadata. Recommending similar items can increase user engagement, click-through rate, and conversion rate for your application.

Similar-Items calculates similarity based on interactions data and any item metadata you provide. It takes into account the co-occurrence of the item in user histories in your Interaction dataset, and any item metadata similarities. For example, with Similar-Items, Amazon Personalize could recommend items customers frequently bought together with a similar style (Categorical metadata (p. 81)), or movies that different users also watched with a similar description (Unstructured text metadata (p. 81)).

With Similar-Items, you provide an item ID in a GetRecommendations (p. 608) operation (or the Amazon Personalize console) and Amazon Personalize returns a list of similar items. Or you can use a batch workflow to get similar items for all of the items in your inventory (see Batch recommendations and user segments (custom resources) (p. 259)). When you get similar items, you can filter the items based on an attribute of the item you specify in your request. You do this by adding a CurrentItem.attribute element to your filter. For an example, see Item data filter examples.
To use Similar-Items, you must create an Interactions dataset with at least 1000 unique historical and event interactions (combined). For more accurate predictions, we recommend that you also create an Items dataset and import metadata about items in your catalog. Similar-Items doesn’t use data in a Users dataset when generating recommendations. You can still filter recommendations based on data in a Users dataset. For more information, see Filtering recommendations and user segments (p. 311).

You can get recommendations for items that are similar to a cold item (an item with fewer than five interactions). If Amazon Personalize can’t find the item ID that you specify in your recommendation request or batch input file, the recipe returns popular items as recommendations.

After you create a solution version, make sure you keep your solution version and data up to date. With Similar-Items, you must manually create a new solution version (retrain the model) to reflect updates to your catalog and update the model with your user’s most recent behavior. Then you must update any campaign using the solution version. For more information, see Maintaining recommendation relevance (p. 278).

Properties and hyperparameters

The Similar-Items recipe has the following properties:

- **Name** – aws-similar-items
- **Recipe Amazon Resource Name (ARN)** – arn:aws:personalize:::recipe/aws-similar-items
- **Algorithm ARN** – arn:aws:personalize:::algorithm/aws-similar-items

For more information, see Choosing a recipe (p. 113).

The following table describes the hyperparameters for the Similar-Items recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see Hyperparameters and HPO (p. 220).

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in HPO?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Algorithm hyperparameters</strong></td>
<td></td>
</tr>
<tr>
<td>popularity_discount_factor</td>
<td>Configure how popularity influences recommendations. Specify a value closer</td>
</tr>
<tr>
<td></td>
<td>to zero to include more popular items. Specify a value closer to one for</td>
</tr>
<tr>
<td></td>
<td>less emphasis on popularity. Specify a value closer to 0.0 for less</td>
</tr>
<tr>
<td></td>
<td>emphasis on popularity.</td>
</tr>
<tr>
<td></td>
<td>Default value: 0.0</td>
</tr>
<tr>
<td></td>
<td>Range: [0.0, 1.0]</td>
</tr>
<tr>
<td></td>
<td>Value type: Float</td>
</tr>
<tr>
<td></td>
<td>HPO tunable: No</td>
</tr>
<tr>
<td>item_id_hidden_dim</td>
<td>The number of hidden variables Amazon Personalize uses to model item ID</td>
</tr>
<tr>
<td></td>
<td>embeddings based on interactions.</td>
</tr>
</tbody>
</table>
### RELATED ITEMS

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>data</code></td>
<td><em>Hidden variables</em> recreate users' purchase history and item statistics to generate ranking scores. To use <code>item_id_hidden_dim</code>, you must use HPO and provide minimum and maximum range values. Amazon Personalize uses HPO to find the best value within the range you specify. Specify a greater maximum value when you have a large Interactions dataset. Using a greater maximum value requires more time to process. To use HPO, set <code>performHPO</code> to <code>true</code> when you call the <code>CreateSolution</code> operation. Default value: 100 Range: [30, 200] Value type: Integer HPO tunable: Yes</td>
</tr>
<tr>
<td><code>item_metadata_hidden_dim</code></td>
<td>The number of hidden variables Amazon Personalize uses to model item metadata. To use <code>item_metadata_hidden_dim</code>, you must use HPO and provide minimum and maximum range values. Amazon Personalize uses HPO to find the best value within the range you specify. Specify a greater maximum value when you have a large Interactions dataset. Using a greater maximum value requires more time to process. To use HPO, set <code>performHPO</code> to <code>true</code> when you call the <code>CreateSolution</code> operation. Default value: 100 Range: [30, 200] Value type: Integer HPO tunable: Yes</td>
</tr>
</tbody>
</table>

### SIMS recipe

**Note**

All RELATED_ITEMS recipes use interactions data. Choose SIMS if you want to configure more hyperparameters for the model. Choose the Similar-Items recipe if you have item metadata and want Amazon Personalize to use it to find similar items.

The Item-to-item similarities (SIMS) recipe uses collaborative filtering to recommend items that are most similar to an item you specify when you get recommendations. SIMS uses your Interactions dataset, not item metadata such as color or price, to determine similarity. SIMS identifies the co-occurrence of the item in user histories in your Interaction dataset to recommend similar items. For example, with SIMS Amazon Personalize could recommend coffee shop items customers frequently bought together or movies that different users also watched.

When you get similar item recommendations, you can filter the items based on an attribute of the item you specify in your request. You do this by adding a `CurrentItem.attribute` element to your filter. For an example, see [item data filter examples](#).
To use SIMS, you must create an Interactions dataset with at least 1000 unique historical and event interactions (combined). SIMS doesn't use data in a Users or Items dataset when generating recommendations. You can still filter recommendations based on data in these datasets. For more information, see Filter recommendations and user segments (p. 311).

If there isn't sufficient user behavior data for an item or the item ID you provide isn't found, SIMS recommends popular items. After you create a solution version, make sure you keep your solution version and data up to date. With SIMS, you must manually create a new solution version (retrain the model) to reflect updates to your catalog and update the model with your user’s most recent behavior. Then you must update any campaign using the solution version. For more information, see Maintaining recommendation relevance (p. 278).

The SIMS recipe has the following properties:

- **Name** – aws-sims
- **Recipe Amazon Resource Name (ARN)** – arn:aws:personalize:::recipe/aws-sims
- **Algorithm ARN** – arn:aws:personalize:::algorithm/aws-sims
- **Feature transformation ARN** – arn:aws:personalize:::feature-transformation/sims
- **Recipe type** – RELATED_ITEMS

The following table describes the hyperparameters for the SIMS recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. Featurization hyperparameters control how to filter the data to use in training. The process of choosing the best value for a hyperparameter is called hyperparameter optimization (HPO). For more information, see Hyperparameters and HPO (p. 220).

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)
- **HPO tunable**: Can the parameter participate in hyperparameter optimization (HPO)?

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Algorithm hyperparameters</strong></td>
<td></td>
</tr>
<tr>
<td>popularity_discount_factor</td>
<td>Configure how popularity influences recommendations. Specify a value closer to zero to include more popular items. Specify a value closer to one for less emphasis on popularity.</td>
</tr>
<tr>
<td></td>
<td>Default value: 0.5</td>
</tr>
<tr>
<td></td>
<td>Range: [0.0, 1.0]</td>
</tr>
<tr>
<td></td>
<td>Value type: Float</td>
</tr>
<tr>
<td></td>
<td>HPO tunable: Yes</td>
</tr>
<tr>
<td>min_cointeraction_count</td>
<td>The minimum number of co-interactions you need to calculate the similarity between a pair of items. For example, a value of 3 means that you need three or more users who interacted with both items for the algorithm to calculate their similarity.</td>
</tr>
<tr>
<td></td>
<td>Default value: 3</td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
</tr>
<tr>
<td>----------------------------------</td>
<td>--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
</tr>
<tr>
<td>max_user_history_length_percentile</td>
<td>The minimum percentile of user history lengths to include in model training. History length is the total amount of available data on a user. Use max_user_history_length_percentile to exclude a percentage of users with long history lengths. Users with a long history tend to contain noise. For example, a robot might have a long list of automated interactions. Removing these users limits noise in training. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of users but removes the edge cases. For example, min_hist_length_percentile = 0.05 and max_hist_length_percentile = 0.95 includes all users except ones with history lengths at the bottom or top 5%. Default value: 0.95 Range: [0.0, 1.0] Value type: Float HPO tunable: No</td>
</tr>
<tr>
<td>Name</td>
<td>Description</td>
</tr>
<tr>
<td>---------------------------------------------------</td>
<td>-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
</tbody>
</table>
| `min_item_interaction_count_percentile`           | The minimum percentile of item interaction counts to include in model training. Use `min_item_interaction_count_percentile` to exclude a percentage of items with a short history of interactions. Items with a short history often are new items. Removing them can train models with more focus on items with a known history. Choose an appropriate value after you review user history lengths, using a histogram or similar tool. We recommend setting a value that retains the majority of items, but removes the edge cases.  
  Default value: 0.01  
  Range: [0.0, 1.0]  
  Value type: Float  
  HPO tunable: No |
| `max_item_interaction_count_percentile`           | The maximum percentile of item interaction counts to include in model training. Use `max_item_interaction_count_percentile` to exclude a percentage of items with a long history of interactions. Items with a long history tend to be older and might be out of date. For example, a movie release that is out of print. Removing these items can focus on more relevant items. Choose an appropriate value after you review user history lengths using a histogram or similar tool. We recommend setting a value that retains the majority of items but removes the edge cases.  
  For example,  
  `min_item_interaction_count_percentile` = 0.05 and  
  `max_item_interaction_count_percentile` = 0.95 includes all items except ones with an interaction count at the bottom or top 5%.  
  Default value: 0.9  
  Range: [0.0, 1.0]  
  Value type: Float  
  HPO tunable: No |

**SIMS sample notebook**

For a sample Jupyter notebook that shows you how to use the SIMS recipe, see [Finding similar items + HPO](#).
USER_SEGMENTATION recipes

USER_SEGMENTATION recipes generate segments of users based on item input data. Each user segment is sorted in descending order based on the probability that each user will interact with items in your inventory. Use a USER_SEGMENTATION recipe to create segments of users who will most likely interact with your catalog based on their item or item attribute preferences. For example, you might want to create a targeted marketing campaign for users that would most likely watch a particular movie or purchase a particular products by brand.

Item-Affinity (p. 149)

The Item-Affinity (aws-item-affinity) recipe is a USER_SEGMENTATION recipe that creates a user segment for each item that you specify.

To train a model, the Item-Affinity recipe uses the Interactions and Items datasets in your dataset group. To create user segments, you train a solution version with the Item-Affinity recipe, and then create a batch segment job (p. 272).

Item-Attribute-Affinity (p. 150)

The Item-Attribute-Affinity (aws-item-attribute-affinity) recipe is a USER_SEGMENTATION recipe that creates a user segment for each item attribute that you specify.

To train a model, the Item-Attribute-Affinity recipe uses the Interactions dataset and Item dataset from a dataset group. To create user segments, you train a solution version with the Item-Attribute-Affinity recipe, and then create a batch segment job (p. 272).

Item-Affinity recipe

The Item-Affinity (aws-item-affinity) recipe is a USER_SEGMENTATION recipe that creates a user segment (group of users) for each item that you specify. These are the users Amazon Personalize predicts will most likely interact with each item. Use Item-Affinity to learn more about your users and take actions based on their respective user segments.

For example, you might want to create a marketing campaign for your retail application based on user preferences for items in your catalog. Item-Affinity would create a user segment for each item based on data in your Interactions and Items datasets. You could use this to promote different items to different user segments based on the likelihood that they will take an action (for example, click an item or purchase an item). Other uses might include cross-selling products to different sets of users or identifying prospective job applicants.

To get user segments based on items, you create a solution and a solution version with the Item-Affinity recipe, then add a list of items in JSON format to an Amazon S3 bucket and create a batch segment job (p. 272). Amazon Personalize outputs a user segment for each item to your output location in Amazon S3. Your input data can have a maximum of 500 items to get user segments for. For information about preparing input data for a batch segment job, see Preparing input data for batch recommendations (p. 261).

You must have an Interactions dataset to use Item-Affinity. Items and Users datasets are optional. You can get user segments with batch segment jobs. For more information, see Batch recommendations and user segments (custom resources) (p. 259).

After you create a solution version, make sure you keep your solution version and data up to date. With Item-Affinity, you must manually create a new solution version (retrain the model) to reflect updates to your catalog and update the model with your user’s most recent behavior. Then you must update any campaign using the solution version. For more information, see Maintaining recommendation relevance (p. 278). To get a user segment for an item, the item must have been present when you created the solution version.
The Item-Affinity recipe has the following properties:

- **Name** – `aws-item-affinity`
- **Recipe Amazon Resource Name (ARN)** – `arn:aws:personalize:::recipe/aws-item-affinity`
- **Algorithm ARN** – `arn:aws:personalize:::algorithm/aws-item-affinity`
- **Feature transformation ARN** – `arn:aws:personalize:::feature-transformation/item-affinity`
- **Recipe type** – USER_SEGMENTATION

The following table describes the hyperparameters for the Item-Affinity recipe. A *hyperparameter* is an algorithm parameter that you adjust to improve model performance. Algorithm hyperparameters control how the model performs. You can’t use hyperparameter optimization (HPO) with the Item-Affinity recipe.

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>hidden_dimension</td>
<td>The number of hidden variables used in the model. <em>Hidden variables</em> recreate users' purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process.</td>
</tr>
<tr>
<td></td>
<td>Default value: 149</td>
</tr>
<tr>
<td></td>
<td>Range: [32, 256]</td>
</tr>
<tr>
<td></td>
<td>Value type: Integer</td>
</tr>
</tbody>
</table>

### Item-Attribute-Affinity recipe

The Item-Attribute-Affinity (aws-item-attribute-affinity) recipe is a USER_SEGMENTATION recipe that creates a user segment (group of users) for each item attribute that you specify. These are the users Amazon Personalize predicts will most likely interact with items with the particular attribute. Use Item-Attribute-Affinity to learn more about your users and take actions based on their respective user segments.

For example, you might want to create a marketing campaign for your retail application based on user preferences for shoe types in your catalog. Item-Attribute-Affinity would create a user segment for each shoe type based data in your Interactions and Items datasets. You could use this to promote different shoes to different user segments based on the likelihood that they will take an action (for example, click a shoe or purchase a shoe). Other uses might include promoting different movie genres to different users or identifying prospective job applicant based on job type.

To get user segments based on item attributes, you create a solution and a solution version with the Item-Attribute-Affinity recipe, then add a list of item attributes in JSON format to an Amazon S3 bucket and create a [batch segment job](p. 272). Amazon Personalize outputs a user segment for each item.
to your output location in Amazon S3. Your input data can have a maximum of 10 queries, where each query is one or more item attributes. For information about preparing input data for a batch segment job, see [Preparing input data for batch recommendations](p. 261).

You must have an Interactions dataset and an Items dataset to use Item-Attribute-Affinity. Your Items dataset must have at least one column that is a non-textual, non-reserved metadata column. You can get user segments with batch segment jobs. For more information, see [Batch recommendations and user segments (custom resources)](p. 259).

After you create a solution version, make sure you keep your solution version and data up to date. With Item-Attribute-Affinity, you must manually create a new solution version (retrain the model) to reflect updates to your catalog and update the model with your user's most recent behavior. Then you must update any campaign using the solution version. For more information, see [Maintaining recommendation relevance](p. 278). To get a user segment for an item attribute, the item attribute must have been present when you created the solution version.

The Item-Attribute-Affinity recipe has the following properties:

- **Name** – `aws-item-attribute-affinity`
- **Recipe Amazon Resource Name (ARN)** – `arn:aws:personalize:::recipe/aws-item-attribute-affinity`
- **Algorithm ARN** – `arn:aws:personalize:::algorithm/aws-item-attribute-affinity`
- **Feature transformation ARN** – `arn:aws:personalize:::feature-transformation/item-attribute-affinity`
- **Recipe type** – `USER_SEGMENTATION`

The following table describes the hyperparameters for the Item-Attribute-Affinity recipe. A hyperparameter is an algorithm parameter that you can adjust to improve model performance. Algorithm hyperparameters control how the model performs. You can't use hyperparameter optimization (HPO) with the Item-Attribute-Affinity recipe.

The table also provides the following information for each hyperparameter:

- **Range**: [lower bound, upper bound]
- **Value type**: Integer, Continuous (float), Categorical (Boolean, list, string)

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>hidden_dimension</td>
<td>The number of hidden variables used in the model. Hidden variables recreate users' purchase history and item statistics to generate ranking scores. Specify a greater number of hidden dimensions when your Interactions dataset includes more complicated patterns. Using more hidden dimensions requires a larger dataset and more time to process.</td>
</tr>
<tr>
<td></td>
<td>Default value: 149</td>
</tr>
<tr>
<td></td>
<td>Range: [32, 256]</td>
</tr>
<tr>
<td></td>
<td>Value type: Integer</td>
</tr>
</tbody>
</table>
Readiness checklist

After you review how Amazon Personalize works and complete the getting started exercise, you can start getting ready to use Amazon Personalize with your own data. This checklist provides lists of Amazon Personalize features, requirements, and data guidance. It can help you plan, or you can use it as a reference as you create resources in Amazon Personalize.

Topics
- Have you matched your use cases to Amazon Personalize resources? (p. 152)
- Do you have enough interactions data? (p. 153)
- Do you have a real-time event streaming architecture in place? (p. 153)
- Is your data optimized for Amazon Personalize? (p. 153)
- Do you collect optional data that can improve recommendations? (p. 153)
- Do you have a plan to test your recommendations? (p. 154)
- Do you have additional business goals? (p. 154)

Have you matched your use cases to Amazon Personalize resources?

Amazon Personalize recommendations can address the following use cases:

- Generating personalized recommendations for a user
- Recommending similar or related items
- Recommending trending or popular items
- Re-ordering by relevance (only with custom resources)
- Generating user segments (only with custom resources)

Amazon Personalize features domain based resources and custom resources configured for these use cases. You start by creating a Domain dataset group or a Custom dataset group:

- With a Domain dataset group, you create resources that are pre-configured and optimized for for the VIDEO_ON_DEMAND or ECOMMERCE domains.

  If you have a streaming video or e-commerce application, we recommend that you start with a Domain dataset group. You can still add custom resources, such as solutions and solution versions trained for custom use cases.

- With a Custom dataset group, you choose a recipe that matches your use case. You then train and deploy only configurable solutions and solution versions (trained Amazon Personalize recommendation models). If you don’t have a streaming video or e-commerce application, we recommend that you create a Custom dataset group. Otherwise, start with a Domain dataset group and adding custom resources as necessary.

For information on the use cases and custom recipes available in Amazon Personalize, see Domain use cases and custom recipes (p. 105).
Do you have enough interactions data?

For all use cases and recipes, you must have at minimum 1,000 interactions for 25 unique users with at least two interactions each. For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

If you aren’t sure if you have enough data, you can import and analyze it with the Amazon Personalize console. For more information, see Analyzing data in datasets (p. 300).

Do you have a real-time event streaming architecture in place?

If you don’t have enough interactions data, you can use Amazon Personalize to collect additional real-time event data. With some recipes and use cases, Amazon Personalize can learn from your user’s most recent activity and update recommendations as they use your application.

For information about recording events, including how events impact recommendations, a list of third-party event tracking services, and sample implementations, see Recording events (p. 280).

Is your data optimized for Amazon Personalize?

We recommend you check for the following in your data:

- Check for missing values. We recommend that a minimum of 70% of your records have data for every attribute. We recommend columns that allow null values be at least 70% complete.
- Fix any inaccuracies or issues in your data, such as inconsistent naming conventions, duplicate categories for an item, mismatched IDs across datasets, or duplicate IDs. These issues can negatively impact recommendations or lead to unexpected behavior. For example, you might have both “N/A” and “Not Applicable” in your data, but filter out recommendations based on only “N/A”. Items marked “Not Applicable” would not be removed by the filter.
- If an item or user can have multiple categories, such as a movie with multiple genres, combine the categorical values into one attribute and separate each value with the | operator. For example, a movie’s GENRES data might be Action | Adventure | Thriller.
- Avoid having more than 30 possible categories for a column (unless the column contains data for only filtering purposes).

For a complete list of data recommendations, and instructions on how you can use Amazon Personalize to identify issues, see Analyzing data in datasets (p. 300).

Do you collect optional data that can improve recommendations?

The following data can help improve your recommendation relevance.

- Event type (required for all Domain dataset group use cases)
- Event value
Do you have a plan to test your recommendations?

You can use A/B testing to compare the results of different groups of users interacting with recommendations from different models. A/B testing can help you compare different recommendation strategies and see if recommendations are helping you achieve your business goals. For more information, see Measuring recommendation impact with A/B testing (p. 347).

Do you have additional business goals?

In some cases, you might have goals in addition to generating relevant recommendations for your users. For example, you might want to maximize revenue, or promote certain types of items from a certain category. The following Amazon Personalize features can help:

- Promotions: You can use promotions to make sure a certain percentage of items satisfy your business requirements. For more information, see Promoting items in recommendations (p. 248).
- Optimizing for business objective: For some Custom dataset group recipes, you can optimize a solution for a custom objective, such as maximizing streaming minutes or increasing revenue. For more information, see Optimizing a solution for an additional objective (p. 215).
- Filtering recommendations. Use filters to apply business rules to recommendations. You can use filters to include or exclude certain types of items from recommendations. For more information, see Filtering recommendations and user segments (p. 311).
Amazon Personalize workflow

After you review the Readiness checklist (p. 152), you are ready to start completing the Amazon Personalize workflow:

1. Create a dataset group (p. 155)

A dataset group is a container for Amazon Personalize resources. The type of dataset group you create determines the resources you can create in step 3 of the Amazon Personalize workflow.

- With a Domain dataset group, you can create recommenders configured for VIDEO_ON_DEMAND or ECOMMERCE domain use cases. You use the recommenders to get recommendations. Amazon Personalize manages their configuration, training, and updates. If you start with a Domain dataset group, you can still add custom resources.

- With a Custom dataset group, you can create only custom resources. These including solutions, solution versions, and campaigns. For these resources, you have more control over configurations, updates, and retraining.

2. Prepare and import data (p. 159)

You import item, user, and interaction records into datasets (Amazon Personalize containers for data). You can import records in bulk or individually. When you import bulk data, you can use Amazon SageMaker Data Wrangler to import data from 40+ sources and prepare it for Amazon Personalize. For more information, see Preparing and importing data using Amazon SageMaker Data Wrangler (p. 160).

After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see Managing data (p. 292).

3. Create domain recommenders or custom resources (p. 210)

After you import your data, create domain recommenders (for Domain dataset groups) or custom resources (for Custom dataset group) to train a model on your data. You use these resources to generate recommendations.

4. Get recommendations (p. 244)

Use your recommender or custom campaign to get recommendations. With a Custom dataset group, you can also get batch recommendations or user segments.

Step 1: Creating a dataset group

When you start using Amazon Personalize, you create a dataset group. A dataset group is a container for Amazon Personalize resources, including datasets, domain recommenders, and custom resources. A dataset group organizes your resources into independent collections, where resources from one dataset group can’t influence resources in any other dataset group.
You create a dataset group for each of your business domains. For example, you might have an application that provides recommendations for streaming video and another that provides recommendations for audio books. In Amazon Personalize, you would create a dataset group for each application. This way, the data from one application does not influence the recommendations Amazon Personalize generates for the other application.

You can create a Domain dataset group or a Custom dataset group:

- **With a Domain dataset group**, you create resources that are pre-configured and optimized for different use cases. When you create a dataset group, you make it a Domain dataset group by specifying a domain of VIDEO_ON_DEMAND or ECOMMERCE.

  If you have a streaming video or e-commerce application, we recommend that you create a Domain dataset group. You can still add custom resources, such as solutions and solution versions trained for custom use cases.

- **A Custom dataset group** includes only custom resources that you configure depending on your use case. With custom resources, you train and deploy configurable solutions and solution versions (a trained Amazon Personalize recommendation model) based on your business needs. If don’t have a VIDEO_ON_DEMAND or ECOMMERCE application, we recommend that you create a Custom dataset group. Otherwise, we recommend starting with a Domain dataset group and adding custom resources as necessary.

You can create a dataset group with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

**Topics**

- Creating a dataset group (console) (p. 156)
- Creating a dataset group (AWS CLI) (p. 157)
- Creating a dataset group (AWS SDKs) (p. 157)

### Creating a dataset group (console)

Create a dataset group by specifying the dataset group name in the Amazon Personalize console.

**To create a dataset group**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign in to your account.
2. Choose Create dataset group.
3. If this is your first time using Amazon Personalize, on the Create dataset group page, in New dataset group, choose Get started.
4. In Dataset group details, for Dataset group name, specify a name for your dataset group.
5. Choose your Domain:
   - Choose E-commerce to create an ECOMMERCE Domain dataset group.
   - Choose Video on demand to create a VIDEO_ON_DEMAND Domain dataset group.
   - Choose Custom to create a Custom dataset group with only custom resources, such as solutions, campaigns, and batch inference jobs.
6. For Tags, optionally add any tags. For more information about tagging Amazon Personalize resources, see Tagging Amazon Personalize resources (p. 370).
7. Choose Create dataset group. The Overview page displays. You are now ready to import data. See Step 2: Preparing and importing data (p. 159).
Creating a dataset group (AWS CLI)

To create a dataset group, use the `create-dataset-group` operation. To create a Domain dataset group, for domain specify ECOMMERCE or VIDEO_ON_DEMAND. To create a Custom dataset group, don't specify a domain. You can use the Tags parameter to optionally tag resources in Amazon Personalize. For a sample see Adding tags (AWS CLI) (p. 372).

The following code creates a Domain dataset group for the VIDEO_ON_DEMAND domain.

```bash
aws personalize create-dataset-group \
  --name dataset-group-name \
  --domain VIDEO_ON_DEMAND
```

If successful, the dataset group Amazon Resource Name (ARN) display as follows.

```json
{
  "datasetGroupArn": "arn:aws:personalize:us-west-2:acct-id:dataset-group/DatasetGroupName"
}
```

Record this value for future use. To display the dataset group that you created, use the `describe-dataset-group` command and specify the returned dataset group ARN.

```bash
aws personalize describe-dataset-group \
  --dataset-group-arn dataset group arn
```

The dataset group and its properties display as follows.

```json
{
  "datasetGroup": {
    "name": "DatasetGroupName",
    "status": "ACTIVE",
    "creationDateTime": 1542392161.262,
    "lastUpdatedDateTime": 1542396513.377
  }
}
```

When the dataset group's status is ACTIVE, you are ready to import data. For more information, see Step 2: Preparing and importing data (p. 159).

Creating a dataset group (AWS SDKs)

Use the following code to create a Domain dataset group. Give the Domain dataset group a name, and for domain, specify either ECOMMERCE or VIDEO_ON_DEMAND. To create a Custom dataset group, modify the code to remove the domain parameter.

For more information about the API operation, see CreateDatasetGroup (p. 444) in the API reference section. You can use the Tags parameter to optionally tag resources in Amazon Personalize. For a sample see Adding tags (AWS SDKs) (p. 372).

SDK for Python (Boto3)

```python
import boto3
```
personalize = boto3.client('personalize')

response = personalize.create_dataset_group(  
    name = 'dataset group name',  
    domain = 'business domain'
)  

dsg_arn = response['datasetGroupArn']

description = personalize.describe_dataset_group(datasetGroupArn = dsg_arn)["datasetGroup"]

print('Name: ' + description['name'])
print('ARN: ' + description['datasetGroupArn'])
print('Status: ' + description['status'])

SDK for Java 2.x

public static String createDomainDatasetGroup(PersonalizeClient personalizeClient,  
        String datasetGroupName,  
        String domain) {  
    try {  
        CreateDatasetGroupRequest createDatasetGroupRequest =  
        CreateDatasetGroupRequest.builder()  
        .name(datasetGroupName)  
        .domain(domain)  
        .build();  
        return personalizeClient.createDatasetGroup(createDatasetGroupRequest).datasetGroupArn();  
    } catch (PersonalizeException e) {  
        System.out.println(e.awsErrorDetails().errorMessage());  
    }  
    return "";
}

SDK for JavaScript v3

// Get service clients module and commands using ES6 syntax.  
import { CreateDatasetGroupCommand } from  
"@aws-sdk/client-personalize";  
import { personalizeClient } from "./libs/personalizeClients.js";

// Or, create the client here.  
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the domain dataset group parameters.  
export const domainDatasetGroupParams = {
    name: 'NAME', /* required */
    domain: 'DOMAIN' /* required for a domain dsg, specify ECOMMERCE or VIDEO_ON_DEMAND */
}

export const run = async () => {
    try {
        const response = await personalizeClient.send(new  
            CreateDatasetGroupCommand(domainDatasetGroupParams));  
        console.log("Success", response);
        return response; // For unit tests.
    } catch (err) {
        console.log("Error", err);
    }
};  
run();
Step 2: Preparing and importing data

Amazon Personalize uses your data to generate recommendations for your users and user segments. Amazon Personalize stores your data in datasets until you delete the datasets. For all use cases (Domain dataset groups) and recipes (custom resources), your interactions data must have the following:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
- At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

When you import data, you can choose to import records in bulk, individually, or both.

- Bulk imports involve importing a large number of historical records at once. You can prepare and import your bulk data with SageMaker Data Wrangler and multiple data sources. Or you can prepare bulk data yourself, and import it directly into Amazon Personalize from a CSV file in Amazon S3. For information about how to format your bulk data for Amazon Personalize, see Data format guidelines (p. 101).
- With individual imports, you import individual records with the Amazon Personalize console and API operations. Or you can import interactions data from live events in real time.

After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see Managing data (p. 292).

As your catalog grows, update your historical data with additional bulk, or individual data, import operations. For real-time recommendations, keep your interactions dataset up to date with your users' behavior. You do this by recording real-time interaction events with an event tracker and the PutEvents (p. 598) operation. For more information, see Recording events (p. 280)

Topics
- Preparing and importing bulk data (p. 159)
- Importing individual records (p. 182)

Preparing and importing bulk data

When you have completed Step 1: Creating a dataset group (p. 155), you are ready to start importing your bulk historical data into Amazon Personalize. You have two choices for importing your bulk records:

- You can use Amazon SageMaker Data Wrangler to import your data from 40+ sources, generate visualizations and Amazon Personalize specific insights, and transform it to meet Amazon Personalize requirements.
- You can import bulk data directly into datasets. When you import directly, you manually format your data to meet Amazon Personalize requirements and upload it to Amazon S3. Then you create a schema and a dataset, and import the data directly into the dataset with a dataset import job.
The following guidelines can help you make sure your bulk data is formatted correctly.

- Your input data must be in a CSV (comma-separated values) file.
- The first row of your CSV file must contain your column headers. Don’t enclose headers in quotation marks (").
- Make sure you have the required fields for your dataset type and make sure that their names align with Amazon Personalize requirements. For example, your Items data might have a column called ITEM_IDENTIFICATION_NUMBER with IDs for each of your items. To use this column as an ITEM_ID field, rename the column to ITEM_ID. If you use Data Wrangler to format your data, you can use the Map columns for Amazon Personalize Data Wrangler transform to make sure your columns are named correctly.

For information about required fields, see Schemas (p. 82). For information about using Data Wrangler to prepare your data, see Preparing and importing data using Amazon SageMaker Data Wrangler (p. 160).

- The column header names in your CSV file must map to your schema.
- Each record in your CSV file must be on a single line.
- The data types in each column must map to your schema. If you use Data Wrangler to format your data, you can use the Data Wrangler transform Parse Value as Type to convert the data types.
- TIMESTAMP and CREATION_TIMESTAMP data must be in UNIX epoch time format. For more information, see Timestamp data (p. 103).
- If your data includes any non-ASCII encoded characters, your CSV file must be encoded in UTF-8 format.
- Makes sure you format any textual data as described in Unstructured text metadata (p. 81).
- Make sure you format impression data and categorical data as described in Formatting explicit impressions (p. 103) and Formatting categorical data (p. 103).

For more information about bulk data formatting requirements for Amazon Personalize, see Data format guidelines (p. 101).

After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see Managing data (p. 292).

If you already created a recommender or deployed a custom solution version with a campaign, how new bulk records influence recommendations depends on the domain use case or recipe that you use. For more information, see How new data influences real-time recommendations (p. 292).

Filter updates for bulk records

Within 20 minutes of completing a bulk import, Amazon Personalize updates any filters you created in the dataset group with your new item and user data. This update allows Amazon Personalize to use the most recent data when filtering recommendations for your users.

Topics

- Preparing and importing data using Amazon SageMaker Data Wrangler (p. 160)
- Importing data directly into Amazon Personalize datasets (p. 170)

Preparing and importing data using Amazon SageMaker Data Wrangler

Important

As you use Data Wrangler, you incur SageMaker costs. For a complete list of charges and prices, see the Data Wrangler tab of Amazon SageMaker pricing. To avoid incurring additional fees,
when you are finished, shut down your Data Wrangler instance. For more information, see Shut Down Data Wrangler.

After you create a dataset group, you can use Amazon SageMaker Data Wrangler (Data Wrangler) to import data from 40+ sources into an Amazon Personalize dataset. Data Wrangler is a feature of Amazon SageMaker Studio that provides an end-to-end solution to import, prepare, transform, and analyze data.

When you use Data Wrangler to prepare and import data, you use a data flow. A data flow defines a series of machine learning data prep steps, starting with importing data. Each time you add a step to your flow, Data Wrangler takes an action on your data, such as transforming it or generating a visualization.

The following are some of the steps that you can add to your flow to prepare data for Amazon Personalize:

- **Insights:** You can add Amazon Personalize specific insight steps to your flow. These insights can help you learn about your data and what actions you can take to improve it.
- **Visualizations:** You can add visualization steps to generate graphs such as histograms and scatter plots. Graphs can help you discover issues in your data, such as outliers or missing values.
- **Transformations:** You can use Amazon Personalize specific and general transformation steps to make sure your data meets Amazon Personalize requirements. The Amazon Personalize transformation helps you map your data columns to required columns depending on the Amazon Personalize dataset type.

If you need to leave Data Wrangler before importing data into Amazon Personalize, you can return to where you left off by choosing the same dataset type when you launch Data Wrangler from the Amazon Personalize console (p. 163). Or you can access Data Wrangler directly through SageMaker studio.

We recommend you import data from Data Wrangler into Amazon Personalize as follows. The transformation, visualization and analysis steps are optional, repeatable, and can be completed in any order.

1. **Set up permissions (p. 162)** - Set up permissions for Amazon Personalize and SageMaker service roles. And set up permissions for your users.
2. **Launch Data Wrangler from the Amazon Personalize console (p. 163)** - Use the Amazon Personalize console to configure a SageMaker domain and launch Data Wrangler.
3. **Import your data into Data Wrangler (p. 163)** - Import data from 40+ sources into Data Wrangler. Sources include AWS services, such as Amazon Redshift, Amazon EMR, or Amazon Athena, and 3rd parties such as Snowflake or DataBricks.
4. **Transform your data (p. 164)** - Use Data Wrangler to transform your data to meet Amazon Personalize requirements.
5. **Visualize and analyze your data (p. 165)** - Use Data Wrangler to visualize your data and analyze it through Amazon Personalize specific insights.
6. **Process and import data into Amazon Personalize (p. 169)** - Use a SageMaker Studio Jupyter notebook to import your processed data into Amazon Personalize.

Additional information

The following resources provide additional information about using Amazon SageMaker Data Wrangler and Amazon Personalize.

- For a tutorial that walks you through processing and transforming a sample dataset, see Demo: Data Wrangler Titanic Dataset Walkthrough in the Amazon SageMaker Developer Guide. This tutorial introduces the fields and functions of Data Wrangler.
- For information on onboarding to Amazon SageMaker domains, see Onboard to Amazon SageMaker Domain Using Quick setup in the Amazon SageMaker Developer Guide.
For information on Amazon Personalize data requirements, see [Data format guidelines (p. 101)](https://docs.aws.amazon.com/personalize/latest/dg/data-formats.html) and [Schemas (p. 82)](https://docs.aws.amazon.com/personalize/latest/dg/data-requirements.html).

### Setting up permissions

To prepare data with Data Wrangler, you must set up the following permissions:

- **Create a service role for Amazon Personalize:** If you haven't already, complete the instructions in [Setting up Amazon Personalize (p. 11)](https://docs.aws.amazon.com/personalize/latest/dg/quick-start-creating-service-role.html) to create an IAM service role for Amazon Personalize. This role must have `GetObject` and `ListBucket` permissions for the Amazon S3 buckets that store your processed data. And it must have permission to use any AWS KMS keys. For information about granting Amazon Personalize access to your Amazon S3 buckets, see [Giving Amazon Personalize access to Amazon S3 resources (p. 16)](https://docs.aws.amazon.com/personalize/latest/dg/access-control-import-data.html). For information about granting Amazon Personalize access to your AWS KMS keys, see [Giving Amazon Personalize permission to use your AWS KMS key (p. 20)](https://docs.aws.amazon.com/personalize/latest/dg/access-control-import-data.html).

- **Create an administrative user with SageMaker permissions:** Your administrator must have full access to SageMaker and must be able to create a SageMaker domain. For more information, see [Create an Administrative User and Group](https://docs.aws.amazon.com/sagemaker/latest/dg/roles-and-permissions.html) in the [Amazon SageMaker Developer Guide](https://docs.aws.amazon.com/sagemaker/latest/dg/roles-and-permissions.html).

- **Create a SageMaker execution role:** Create a SageMaker execution role with access to SageMaker resources and Amazon Personalize data import operations. The SageMaker execution role must have the `AmazonSageMakerFullAccess` policy attached. If you require more granular Data Wrangler permissions, see [Data Wrangler Security and Permissions](https://docs.aws.amazon.com/personalize/latest/dg/access-control-import-data.html) in the [Amazon SageMaker Developer Guide](https://docs.aws.amazon.com/sagemaker/latest/dg/roles-and-permissions.html). For more information on SageMaker roles, see [SageMaker Roles](https://docs.aws.amazon.com/sagemaker/latest/dg/roles-and-permissions.html).

To grant access to Amazon Personalize data import operations, attach the following IAM policy to the SageMaker execution role. This policy grants the permissions required to import data into Amazon Personalize and attach a policy to your Amazon S3 bucket. And it grants `PassRole` permissions when the service is Amazon Personalize. Update the Amazon S3 bucket-name to the name of the Amazon S3 bucket you want to use as the destination for your formatted data after you prepare it with Data Wrangler.

```json
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Action": [
                "personalize:Create*",
                "personalize:List*",
                "personalize:Describe*"
            ],
            "Resource": "*"
        },
        {
            "Effect": "Allow",
            "Action": [
                "s3:PutBucketPolicy"
            ],
            "Resource": [
                "arn:aws:s3:::bucket-name",
                "arn:aws:s3:::bucket-name/*"
            ]
        },
        {
            "Effect": "Allow",
            "Action": [
                "iam:PassRole"
            ],
            "Resource": "*"
        }
    ]
}
```
For information on creating an IAM policy, see Creating IAM policies in the IAM User Guide. For information on attaching an IAM policy to role, see Adding and removing IAM identity permissions in the IAM User Guide.

Launching Data Wrangler from Amazon Personalize

To launch Data Wrangler from Amazon Personalize, you use the Amazon Personalize console to configure a SageMaker domain and launch Data Wrangler.

To launch Data Wrangler from Amazon Personalize

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your dataset group.
3. On the overview page, choose Import interactions data or Import users data or Import item data, depending on your dataset type.
4. Choose Import data using Data Wrangler.
5. For SageMaker domain, choose to use an existing domain or create a new one. You need a SageMaker Domain to access Data Wrangler in SageMaker Studio. For information about domains and user profiles, see SageMaker Domain in the Amazon SageMaker Developer Guide.
6. To use an existing domain, choose a SageMaker domain and User profile to configure the domain.
7. To create a new domain:
   - Give the new domain a name.
   - Choose a User profile name.
   - For Execution role, choose the role you created in Setting up permissions (p. 162). Or, if you have CreateRole permissions, create a new role using the role creation wizard. The role you use must have the AmazonSageMakerFullAccess policy attached.
8. Choose Next. If you are creating a new domain, SageMaker starts creating your domain. This can take up to ten minutes.
9. Review the details for your SageMaker domain.
10. Choose Import data with Data Wrangler. SageMaker Studio starts creating your environment, and when complete, the Data flow page of Data Wrangler in SageMaker Studio opens in a new tab. It can take up to five minutes for SageMaker Studio to finish creating your environment. When it finishes, you are ready to start importing data into Data Wrangler. For more information, see Importing data into Data Wrangler (p. 163).

Importing data into Data Wrangler

After you configure a SageMaker domain and launch Data Wrangler in a new tab, you are ready to import data from your source into Data Wrangler. When you use Data Wrangler to prepare data for Amazon Personalize, you import one dataset at a time. We recommend starting with an Interactions dataset. You start on the Data flow page. The page should look similar to the following.
To start importing data, you choose **Import data** and specify your data source. Data Wrangler supports 40+ sources. These include AWS services, such as Amazon Redshift, Amazon EMR, or Amazon Athena, and third parties, such as Snowflake or DataBricks. Different data sources have different procedures for connecting and importing data.

For a complete list of available sources and step-by-step instructions on importing data, see [Import](#) in the *Amazon SageMaker Developer Guide*.

After you import data into Data Wrangler, you are ready to transform it. For information about transforming data, see [Transforming data](#).  

**Transforming data**

To transform data in Data Wrangler, you add a **Transform** step to your data flow. Data Wrangler includes over 300 transforms that you can use to prepare your data, including a **Map columns for Amazon Personalize** transform. And you can use the general Data Wrangler transforms to fix issues such as outliers, type issues, and missing values.

After you finish transforming your data, you can analyze it with Data Wrangler. Or, if you are finished preparing your data in Data Wrangler, you can process it and import it into Amazon Personalize. For information about analyzing data, see [Generating visualizations and data insights](#). For information about processing and importing data, see [Processing data and importing it into Amazon Personalize](#).

**Topics**

- [Mapping columns for Amazon Personalize](#)
- [General Data Wrangler transforms](#)
Mapping columns for Amazon Personalize

To transform your data so it meets Amazon Personalize requirements, you add the Map columns for Amazon Personalize transform and map your columns to the required and optional fields for Amazon Personalize.

To use the Map columns for Amazon Personalize transform

1. Choose + for your latest transform and choose Add transform. If you haven't added a transform, choose the + for the Data types transform. Data Wrangler adds this transform automatically to your flow.
2. Choose Add step.
3. Choose Transforms for Amazon Personalize. The Map columns for Amazon Personalize transform is selected by default.
4. Use the transform fields to map your data to required Amazon Personalize attributes.
   1. Choose the dataset type that matches your data (Interactions, Items, or Users).
   2. Choose your domain (ECOMMERCE, VIDEO_ON_DEMAND, or custom). The domain you choose must match the domain you specified when you created your dataset group.
   3. Choose the columns that match the required and optional fields for Amazon Personalize. For example, for the item_ID column, choose the column in your data that stores the unique identification information for each of your items.

   Each column field is filtered by data type. Only the columns in your data that meet Amazon Personalize data type requirements are available. If your data is not of the required type, you can use the Parse Value as Type Data Wrangler transform to convert it.

General Data Wrangler transforms

The following general Data Wrangler transforms can help you prepare data for Amazon Personalize:

- Data type conversion: If your field is not listed as a possible option in the Map columns for Amazon Personalize transform, you might need to convert its data type. The Data Wrangler transform Parse Value as Type can help you convert your data. Or you can use the Data types transform that Data Wrangler adds by default when you create a flow. To use this transform, you choose the data type from the Type drop-down lists, choose Preview and then choose Update.

   For information on required data types for fields, see the section for your domain and dataset type in Schemas (p. 82).
- Handling missing values and outliers: If you generate missing value or outlier insights, you can use the Data Wrangler transforms Handle Outliers and Handle Missing Values to resolve these issues.
- Custom transformations: With Data Wrangler, you can create your own transformations with Python (User-Defined Function), PySpark, pandas, or PySpark (SQL). You might use a custom transform to perform tasks such as dropping duplicate columns or grouping by columns. For more information, see Custom Transforms in the Amazon SageMaker Developer Guide.

Generating visualizations and data insights

After you import your data into Data Wrangler, you can use it to generate visualizations and data insights.

- Visualizations (p. 166): Data Wrangler can generate different types of graphs, such as histograms and scatter plots. For example, you can generate a histogram to identify outliers in your data.
- Data insights (p. 166): You can use a Data Quality and Insights Report for Amazon Personalize to learn about your data through data insights and column and row statistics. This report can let you
Preparing and importing bulk data

Know if you have any type issues in your data. And you can learn what actions you can take to improve your data. These actions can help you meet Amazon Personalize resource requirements, such as model training requirements, or they can lead to improved recommendations.

After you learn about your data through visualizations and insights, you can use this information to help you apply additional transforms to improve your data. Or, if you are finished preparing your data, you can process it and import it into Amazon Personalize. For information about transforming your data, see Transforming data (p. 164). For information about processing and importing data, see Processing data and importing it into Amazon Personalize (p. 169).

Generating visualizations

You can use Data Wrangler to create different types of graphs, such as histograms and scatter plots. For example, you can generate a histogram to identify outliers in your data. To generate a data visualization, you add an Analysis step to your flow and, from Analysis type, choose the visualization you want to create.

For more information about creating visualizations in Data Wrangler, see Analyze and Visualize in the Amazon SageMaker Developer Guide.

Generating data insights

You can use Data Wrangler to generate a Data Quality and Insights Report for Amazon Personalize report specific to your dataset type. Before generating the report, we recommend that you transform your data to meet Amazon Personalize requirements. This will lead to more relevant insights. For more information, see Transforming data (p. 164).

Topics

- Report content (p. 166)
- Generating the report (p. 169)

Report content

The Data Quality and Insights Report for Amazon Personalize includes the following sections:

- **Summary:** The report summary includes dataset statistics and high priority warnings:
  - **Dataset statistics:** These include Amazon Personalize specific statistics, such as the number of unique users in your interactions data, and general statistics, such as the number of missing values or outliers.
  - **High priority warnings:** These are Amazon Personalize specific insights that have the most impact on training or recommendations. Each warning includes a recommended action that you can take to resolve the issue.
- **Duplicate rows and Incomplete rows:** These sections include information on which rows have missing values and which rows are duplicated in your data.
- **Feature summary:** This section includes the data type for each column, invalid or missing data information, and warning counts.
- **Feature details:** This section includes subsections with detailed information for each of your columns of data. Each subsection includes statistics for the column, such as categorical value count, and missing value information. And each subsection includes Amazon Personalize specific insights and recommended actions for columns of data. For example, an insight might indicate that a column has more than 30 possible categories.
Data type issues

The report identifies columns that are not of the correct data type and specifies the required type. To get insights related to these features, you must convert the data type of the column and generate the report again. To convert the type, you can use the Data Wrangler transform Parse Value as Type.

Amazon Personalize insights

The Amazon Personalize insights include a finding and a suggested action. The action is optional. For example, the report might include an insight and action related to the number of categories for a column of categorical data. If you don't believe the column is a categorical, you can disregard this insight and take no action.

Except for minor wording differences, the Amazon Personalize specific insights are the same as the single dataset insights you might generate when you analyze your data with Amazon Personalize. For example, the insights report in Data Wrangler includes insights such as "The Interactions dataset has only X unique users with two or more interactions." But it doesn't include insights like "X% of items in the Items dataset have no interactions in the Interactions dataset."

For a list of possible Amazon Personalize specific insights, see the insights that don't reference multiple datasets in Data insights (p. 301).

Report examples

The look and feel of the Amazon Personalize report is the same as the general insights report in Data Wrangler. For examples of the general insights report, see Get Insights On Data and Data Quality in the Amazon SageMaker Developer Guide. The following example shows how the summary section of a report for an Interactions dataset. It includes dataset statistics and some possible high priority Interactions dataset warnings.
Preparing and importing bulk data

The following example shows how the feature details section for an EVENT_TYPE column of an Interactions dataset might appear in a report.
Generating the report

To generate the Data Quality and Insights Report for Amazon Personalize, you choose Get data insights for your transform and create an analysis.

To generate Data Quality and Insights Report for Amazon Personalize

1. Choose the + option for the transform you are analyzing. If you haven't added a transform, choose the + for the Data types transform. Data Wrangler adds this transform automatically to your flow.
2. Choose Get data insights. The Create analysis panel displays.
3. For Analysis type, choose Data Quality and Insights Report for Amazon Personalize.
4. For Dataset type, choose the type of Amazon Personalize dataset you are analyzing.
5. Optionally choose Run on full data. By default, Data Wrangler generates insights on only a sample of your data.
6. Choose Create. When analysis completes, the report appears.

Processing data and importing it into Amazon Personalize

When you are finished analyzing and transforming your data, you are ready to process it and import it into Amazon Personalize.

- Processing data (p. 169) – Processing the data applies your transform to your entire dataset and outputs it to a destination you specify. In this case you specify an Amazon S3 bucket.
- Importing data into Amazon Personalize (p. 169) – To import processed data into Amazon Personalize, you run a Jupyter Notebook provided in SageMaker Studio. This notebook creates your Amazon Personalize datasets and imports your data into them.

Processing data

Before you import data into Amazon Personalize, you must apply your transform to your entire dataset and output it to an Amazon S3 bucket. To do this, you create a destination node with the destination set to an Amazon S3 bucket, and then launch a processing job for the transformation.

For step-by-step instructions on specifying a destination and launching a process job, see Launch processing jobs with a few clicks using Amazon SageMaker Data Wrangler. When you add a destination, choose Amazon S3. You will use this location when you import the processed data into Amazon Personalize.

When you finish processing your data, you are ready to import it from the Amazon S3 bucket into Amazon Personalize.

Importing data into Amazon Personalize

After you process your data, you are ready to import it into Amazon Personalize. To import processed data into Amazon Personalize, you run a Jupyter Notebook provided in SageMaker Studio. This notebook creates your Amazon Personalize datasets and imports your data into them.

To import processed data into Amazon Personalize

1. For the transformation you want to export, choose Export to and choose Amazon Personalize (via Jupyter Notebook).
2. Modify the notebook to specify the Amazon S3 bucket you used as the data destination for the processing job. Optionally specify the domain for your dataset group. By default, the notebook creates a custom dataset group.
3. Review the notebook cells that create the schema. Verify that the schema fields have the expected types and attributes before running the cell.
Preparing and importing bulk data

- Verify that fields that support null data have `null` listed in the list of types. The following example shows how to add `null` for a field.

  ```json
  {  
    "name": "GENDER",
    "type": [
      "null",
      "string"
    ],
    "categorical": true
  }
  ```

- Verify that categorical fields have the `categorical` attribute set to true. The following example shows how to mark a field categorical.

  ```json
  {  
    "name": "SUBSCRIPTION_MODEL",
    "type": "string",
    "categorical": true
  }
  ```

- Verify that textual fields have the `textual` attribute set to true. The following example shows how to mark a field as textual.

  ```json
  {  
    "name": "DESCRIPTION",
    "type": [
      "null",
      "string"
    ],
    "textual": true
  }
  ```

4. Run the notebook to create a schema, and create dataset, and import your data into the Amazon Personalize dataset. You run the notebook just as you would a notebook outside of SageMaker Studio. For information on running Jupyter notebooks, see Running Code. For information on notebooks in SageMaker Studio, see Use Amazon SageMaker Notebooks in the Amazon SageMaker Developer Guide.

   After you complete the notebook, if you imported interactions data, you are ready to create recommenders or custom resources. Or you can repeat the process with an items dataset or users dataset. For information about creating recommenders or custom resources, see Step 3: Creating recommenders or custom resources (p. 190).

**Importing data directly into Amazon Personalize datasets**

After you have completed Step 1: Creating a dataset group (p. 155), you are ready to import bulk records from a large CSV file into an Amazon Personalize dataset.

To import data directly into Amazon Personalize datasets, you do the following:

1. Create schema JSON file based on your data. For schema requirements and examples, see Schemas (p. 82).
2. Make sure your data is correctly formatted. The column names must match your schema. Your data must be in a CSV file. For data format guidelines, see Data format guidelines (p. 101).
3. Upload your CSV files to an Amazon Simple Storage Service (Amazon S3) bucket and give Amazon Personalize access to your Amazon S3 resources.
4. Create an Amazon Personalize schema using the JSON file from step one. And create an Amazon Personalize dataset.

5. Create a dataset import job that populates the dataset with data from your Amazon S3 bucket. To create a dataset import job for interactions datasets, your CSV file must have at minimum 1000 interaction records.

After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see Managing data (p. 292).

Topics
- Uploading to an Amazon S3 bucket (p. 171)
- Creating a dataset and a schema (p. 171)
- Importing bulk records with a dataset import job (p. 177)

Uploading to an Amazon S3 bucket

After you format your historical input data (see Data format guidelines (p. 101)), you must upload the CSV file to an Amazon S3 bucket and give Amazon Personalize permission to access to your Amazon S3 resources:

1. If you haven't already, follow the steps in Setting up permissions (p. 12) to set up permissions so Amazon Personalize can access your Amazon Personalize resources on your behalf.
2. Upload your CSV files to an Amazon Simple Storage Service (Amazon S3) bucket. This is the location that Amazon Personalize imports your data from. For more information, see Uploading Files and Folders by Using Drag and Drop in the Amazon Simple Storage Service User Guide.
3. Give Amazon Personalize access to your Amazon S3 resources by attaching access policies to your Amazon S3 bucket and Amazon Personalize service role. See Giving Amazon Personalize access to Amazon S3 resources (p. 16).

If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

After you upload your data to an Amazon S3 bucket and give Amazon Personalize access to Amazon S3, import your data into Amazon Personalize. See Preparing and importing bulk data (p. 159).

Creating a dataset and a schema

After you have completed Step 1: Creating a dataset group (p. 155), you are ready to create a dataset. If you are importing bulk records, make sure you review the Data format guidelines (p. 101) and complete Uploading to an Amazon S3 bucket (p. 171).

Datasets are Amazon Personalize containers for data. When you create a dataset, you also create a schema for the dataset. A schema tells Amazon Personalize about the structure of your data and allows Amazon Personalize to parse the data.

You create datasets with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs. For information about the different types of datasets, and dataset and schema requirements, see Schemas (p. 82).

Topics
- Creating a dataset and a schema (console) (p. 172)
- Creating a dataset and a schema (AWS CLI) (p. 172)
Creating a dataset and a schema (console)

If this is your first dataset in your dataset group, your first dataset type will be an Interactions dataset. To create your Interactions dataset in the console, specify the dataset name and then specify a JSON schema in Avro format. If it is not your first dataset in this dataset group, choose the dataset type and then specify a name and a schema.

For information on Amazon Personalize datasets and schema requirements, see Schemas (p. 82).

Note
If you just completed Step 1: Creating a dataset group (p. 155) and you are already on the user-item interaction page, skip to step 4 in this procedure.

To create a dataset and a schema

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose the dataset group you created in Step 1: Creating a dataset group (p. 155). This displays the dataset group Dashboard.
3. In the Upload datasets section, for the type of dataset that you want to import (Amazon Personalize datasets include Interactions, Users, or Items), choose Import. The Configure < dataset type > page is displayed.
4. In Dataset details, for Dataset name, specify a name for your dataset.
5. In Schema details, for Schema selection, either choose an existing schema or choose Create new schema.
6. If you are creating a new schema, for Schema definition, paste in the schema JSON that matches your data. Use the examples found in Schemas (p. 82) as a guide. After you create a schema, you can't make changes to the schema.
7. For New schema name, specify a name for the new schema.
8. For Tags, optionally add any tags. For more information about tagging Amazon Personalize resources, see Tagging Amazon Personalize resources (p. 370).
9. Choose Next and follow the instructions in Preparing and importing bulk data (p. 159) to import your data.

Creating a dataset and a schema (AWS CLI)

To create a dataset and a schema using the AWS CLI, you first define a schema in Avro format and add it to Amazon Personalize using the CreateSchema (p. 465) operation. Then create a dataset using the CreateDataset (p. 436) operation. For information on Amazon Personalize datasets and schema requirements, see Schemas (p. 82).

To create a schema and dataset

1. Create a schema file in Avro format and save it as a JSON file. This file should be based on the type of dataset, such as Interactions, you are creating.

   The schema must match the columns in your data and the schema name must match one of the three types of datasets recognized by Amazon Personalize. The following is an example of a minimal Interactions dataset schema. For more examples, see Schemas (p. 82).

```json
{
   "type": "record",
   "name": "Interactions",
   // Other fields...
}
```
2. Create a schema in Amazon Personalize by running the following command. After you create a schema, you can’t make changes to the schema. Replace `schemaName` with the name of the schema, and replace `file://SchemaName.json` with the location of the JSON file you created in the previous step. The example shows the file as belonging to the current folder. For more information about the API, see CreateSchema (p. 465).

```bash
aws personalize create-schema --name SchemaName --schema file://SchemaName.json
```

The schema Amazon Resource Name (ARN) is displayed, as shown in the following example:

```json
{
  "schemaArn": "arn:aws:personalize:us-west-2:acct-id:schema/SchemaName"
}
```

3. Create an empty dataset by running the following command. Provide the dataset group Amazon Resource Name (ARN) from Creating a dataset group (AWS CLI) (p. 157) and schema ARN from the previous step. The `dataset-type` must match the schema name from the previous step. For more information about the API, see CreateDataset (p. 436).

```bash
aws personalize create-dataset --name DatasetName --dataset-group-arn Dataset Group ARN --dataset-type Dataset Type --schema-arn Schema Arn
```

The dataset ARN is displayed, as shown in the following example.

```json
{
  "datasetArn": "arn:aws:personalize:us-west-2:acct-id:dataset/DatasetName/INTERACTIONS"
}
```

4. Record the dataset ARN for later use. After you have created a dataset, you are ready to import your training data. See Preparing and importing bulk data (p. 159).

**Creating a dataset and a schema (AWS SDKs)**

To create a dataset and a schema using the AWS SDKs, you first define a schema in Avro format and add it to Amazon Personalize using the CreateSchema (p. 465) operation. After you create a schema, you
can't make changes to the schema. Then create a dataset using the CreateDataset (p. 436) operation. For information on Amazon Personalize datasets and schema requirements, see Schemas (p. 82).

To create a schema and a dataset

1. Create a schema file in Avro format and save it as a JSON file in your working directory.

   The schema must match the columns in your data and the schema name must match one of the three types of datasets recognized by Amazon Personalize. The following is an example of a minimal Interactions dataset schema. For more examples, see Schemas (p. 82).

   ```json
   {
     "type": "record",
     "name": "Interactions",
     "namespace": "com.amazonaws.personalize.schema",
     "fields": [
       {
         "name": "USER_ID",
         "type": "string"
       },
       {
         "name": "ITEM_ID",
         "type": "string"
       },
       {
         "name": "TIMESTAMP",
         "type": "long"
       }
     ],
     "version": "1.0"
   }
   ```

2. Create the schema using the CreateSchema (p. 465) API operation. The following code shows how to create a schema. Specify the name for your schema and the file path for your schema JSON file.

   **SDK for Python (Boto3)**

   ```python
   import boto3
   personalize = boto3.client('personalize')
   with open('schemaFile.json') as f:
       createSchemaResponse = personalize.create_schema(
           name = 'schema name',
           schema = f.read()
       )
   schema_arn = createSchemaResponse['schemaArn']
   print('Schema ARN:' + schema_arn)
   ```

   **SDK for Java 2.x**

   ```java
   public static String createSchema(PersonalizeClient personalizeClient, String schemaName, String filePath) {
       String schema = null;
       try {
           schema = new String(Files.readAllBytes(Paths.get(filePath)));
       } catch (IOException e) {
           System.out.println(e.getMessage());
       }
   }
   ```
Amazon Personalize Developer Guide
Preparing and importing bulk data

```javascript
SDK for JavaScript v3

// Get service clients module and commands using ES6 syntax.
import { CreateSchemaCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});
import fs from 'fs';
let schemaFilePath = "SCHEMA_PATH";
let mySchema = "";

try {
  mySchema = fs.readFileSync(schemaFilePath).toString();
} catch (err) {
  mySchema = 'TEST' // For unit tests.
}

// Set the schema parameters.
export const createSchemaParam = {
  name: 'NAME', /* required */
  schema: mySchema /* required */
};

export const run = async () => {
  try {
    const response = await personalizeClient.send(new CreateSchemaCommand(createSchemaParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();
```

Amazon Personalize returns the ARN of the new schema. Record it because you'll need it in the next step.
3. Create a dataset using the CreateDataset (p. 436) operation. The following code shows how to create a dataset. Specify the Amazon Resource Name (ARN) of your dataset group, the schema ARN from the previous step, and specify the dataset type (Interactions, Users, or Items). For information about the different types of datasets, see Schemas (p. 82).

SDK for Python (Boto3)

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.create_dataset(
    name = 'dataset_name',
    schemaArn = 'schema_arn',
    datasetGroupArn = 'dataset_group_arn',
    datasetType = 'dataset_type'
)

print ('Dataset Arn: ' + response['datasetArn'])
```

SDK for Java 2.x

```java
public static String createDataset(PersonalizeClient personalizeClient,
String datasetName,
String datasetGroupArn,
String datasetType,
String schemaArn) {
    try {
        CreateDatasetRequest request = CreateDatasetRequest.builder()
            .name(datasetName)
            .datasetGroupArn(datasetGroupArn)
            .datasetType(datasetType)
            .schemaArn(schemaArn).build();

        String datasetArn = personalizeClient.createDataset(request).datasetArn();
        System.out.println("Dataset " + datasetName + " created. Dataset ARN: " +
        datasetArn);
        return datasetArn;
    } catch(PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
    return "";
}
```

SDK for JavaScript v3

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateDatasetCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the dataset's parameters.
export const createDatasetParam = {
    datasetGroupArn: 'DATASET_GROUP_ARN', /* required */
    datasetType: 'DATASET_TYPE', /* required */
    name: 'NAME', /* required */
```
After you have created a dataset, you are ready to import your training data. See [Preparing and importing bulk data](p. 159).

## Importing bulk records with a dataset import job

After you have formatted your input data (see [Data format guidelines](p. 101)), uploaded it to an Amazon Simple Storage Service (Amazon S3) bucket (see [Uploading to an Amazon S3 bucket](p. 171)), and completed [Creating a dataset and a schema](p. 171), import the bulk records into the dataset by creating a dataset import job.

A **dataset import job** is a bulk import tool that populates your dataset with data from your Amazon S3 bucket. You can create a dataset import job using the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

If you've previously created a dataset import job for a dataset, you can use a new dataset import job to add to or replace the existing bulk data. For more information see [Updating existing bulk data](p. 294).

### Topics
- Importing bulk records (console) (p. 177)
- Importing bulk records (AWS CLI) (p. 178)
- Importing bulk records (AWS SDKs) (p. 180)

### Importing bulk records (console)

**Important**

By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. For information about updating existing data, see [Updating existing bulk data](p. 294).

To import bulk records into a dataset with the Amazon Personalize console, create a dataset import job with a name, the IAM service role, and the location of your data.

If you just created your dataset in [Creating a dataset and a schema](p. 171), skip to step 5. If you've already completed an initial import job and want to refresh your data, see [Updating bulk records (console)](p. 295).

**To import bulk records (console)**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign in to your account.
2. On the **Dataset groups** page, choose your dataset group. The dataset group **Overview** displays.
3. In the **Create datasets** section, choose the **Import** button for the type of data you want to import. The **Configure < dataset type > schema** page displays.
4. In **Dataset import job details**, for **Data import source** choose **Import data from S3**.
5. For **Dataset import job name**, specify a name for your import job.
6. In **Input source**, for **S3 Location**, specify where your data file is stored in Amazon S3. Use the following syntax:

   `s3://<name of your S3 bucket>/<folder path>/<CSV filename>`

   If your CSV files are in a folder in your Amazon S3 bucket and you want to upload multiple CSV files to a dataset with one dataset import job, you can specify the path to the folder. Amazon Personalize only uses the files in the first level of your folder, it doesn't use any data in any sub folders. Use the following syntax with a `/` after the folder name:

   `s3://<name of your S3 bucket>/<folder path>/`

7. In **IAM role**, choose to either create a new role or use an existing one. If you completed the perquisites, choose **Use an existing service role** and specify the role that you created in [Creating an IAM role for Amazon Personalize](p. 15).
8. If you created a metric attribution and want to publish metrics related to this job to Amazon S3, in **Publish event metrics to S3** choose **Publish metrics for this import job**.

   If you haven't created one and want to publish metrics for this job, choose **Create metric attribution** to create a new one on a different tab. After you create the metric attribution, you can return to this screen and finish creating the import job.

   For more information on metric attributions, see [Measuring impact of recommendations](p. 329).
9. For **Tags**, optionally add any tags. For more information about tagging Amazon Personalize resources, see [Tagging Amazon Personalize resources](p. 370).
10. Choose **Finish**. The data import job starts and the **Dashboard Overview** page is displayed. The dataset import is complete when the status shows as **ACTIVE**. After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see [Managing data](p. 292).

    After you import your data you are ready to create a solution. For more information, see [Creating a solution and a solution version](p. 210).

### Importing bulk records (AWS CLI)

**Important**

By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. For information about updating existing data, see [Updating existing bulk data](p. 294).

To import bulk records using the AWS CLI, create a dataset import job using the `CreateDatasetImportJob` command. If you've previously created a dataset import job for a dataset, you can use the import mode parameter to specify how to add the new data. For a code sample, see [Updating bulk records (AWS CLI)](p. 296).

### Import bulk records (AWS CLI)

1. Create a dataset import job by running the following command. Provide the Amazon Resource Name (ARN) for your dataset and specify the path to your Amazon S3 bucket where you stored the training data. Use the following syntax for the path:

   `s3://<name of your S3 bucket>/<folder path>/<CSV filename>`

   If your CSV files are in a folder in your Amazon S3 bucket and you want to upload multiple CSV files to a dataset with one dataset import job, you can specify the path to the folder. Amazon Personalize only uses the files in the first level of your folder, it doesn't use any data in any sub folders. Use the following syntax with a `/` after the folder name:
s3://<name of your S3 bucket>/<folder path>/

Provide the AWS Identity and Access Management (IAM) role Amazon Resource Name (ARN) that you created in Creating an IAM role for Amazon Personalize (p. 15). The default import-mode is FULL. For more information see Updating existing bulk data (p. 294). For more information about the operation, see CreateDatasetImportJob (p. 448).

```bash
aws personalize create-dataset-import-job
--job-name dataset import job name
--dataset-arn dataset arn
--data-source dataLocation=s3://bucketname/filename
--role-arn roleArn
--import-mode FULL
```

The dataset import job ARN is displayed, as shown in the following example.

```
{
}
```

2. Check the status by using the describe-dataset-import-job command. Provide the dataset import job ARN that was returned in the previous step. For more information about the operation, see DescribeDatasetImportJob (p. 509).

```bash
aws personalize describe-dataset-import-job
--dataset-import-job-arn dataset import job arn
```

The properties of the dataset import job, including its status, are displayed. Initially, the status shows as CREATE PENDING.

```
{
  "datasetImportJob": {
    "jobName": "Dataset Import job name",
    "dataSource": {
      "dataLocation": "s3://<bucketname>/ratings.csv"
    },
    "importMode": "FULL",
    "roleArn": "role-arn",
    "status": "CREATE PENDING",
    "creationDateTime": 1542392161.837,
    "lastUpdatedDateTime": 1542393013.377
  }
}
```

The dataset import is complete when the status shows as ACTIVE. After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see Managing data (p. 292).

After you import your data into the relevant datasets in the dataset group, you can create a solution version (trained model). For more information, see Creating a solution and a solution version (p. 210).
Preparing and importing bulk data

Importing bulk records (AWS SDKs)

**Important**
By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. For information about updating existing data, see [Updating existing bulk data](p. 294).

To import data, create a dataset import job with the `CreateDatasetImportJob` operation. The following code shows how to create a dataset import job.

Give the job name, set the `datasetArn` the Amazon Resource Name (ARN) of your dataset, and set the `dataLocation` to the path to your Amazon S3 bucket where you stored the training data. Use the following syntax for the path:

```
s3://<name of your S3 bucket>/<folder path>/<CSV filename>.csv
```

If your CSV files are in a folder in your Amazon S3 bucket and you want to upload multiple CSV files to a dataset with one dataset import job, you can specify the path to the folder. Amazon Personalize only uses the files in the first level of your folder, it doesn't use any data in any sub folders. Use the following syntax with a `/` after the folder name:

```
s3://<name of your S3 bucket>/<folder path>/
```

For the `roleArn`, specify the AWS Identity and Access Management (IAM) role that gives Amazon Personalize permissions to access your S3 bucket. See [Creating an IAM role for Amazon Personalize](p. 15). The default `importMode` is `FULL`. For more information see [Updating bulk records](AWS SDKs) (p. 296).

**SDK for Python (Boto3)**

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.create_dataset_import_job(
    jobName = 'YourImportJob',
    datasetArn = 'dataset_arn',
    dataSource = {'dataLocation': 's3://bucket/file.csv'},
    roleArn = 'role_arn',
    importMode = 'FULL'
)

dsij_arn = response['datasetImportJobArn']

print ('Dataset Import Job arn: ' + dsij_arn)

description = personalize.describe_dataset_import_job(
    datasetImportJobArn = dsij_arn)

print('Name: ' + description['jobName'])
print('ARN: ' + description['datasetImportJobArn'])
print('Status: ' + description['status'])
```

**SDK for Java 2.x**

```java
public static String createPersonalizeDatasetImportJob(PersonalizeClient personalizeClient,
    String jobName, String datasetArn, String s3BucketPath,
    String roleArn,
```
Preparing and importing bulk data

```java
importMode importMode) {
    long waitInMilliseconds = 60 * 1000;
    String status;
    String datasetImportJobArn;

    try {
        DataSource importDataSource = DataSource.builder()
            .dataLocation(s3BucketPath)
            .build();

        CreateDatasetImportJobRequest createDatasetImportJobRequest =
            CreateDatasetImportJobRequest.builder()
            .datasetArn(datasetArn)
            .dataSource(importDataSource)
            .jobName(jobName)
            .roleArn(roleArn)
            .importMode(importMode)
            .build();

        datasetImportJobArn =
            personalizeClient.createDatasetImportJob(createDatasetImportJobRequest)
            .datasetImportJobArn();

        DescribeDatasetImportJobRequest describeDatasetImportJobRequest =
            DescribeDatasetImportJobRequest.builder()
            .datasetImportJobArn(datasetImportJobArn)
            .build();

        long maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
        while (Instant.now().getEpochSecond() < maxTime) {
            DatasetImportJob datasetImportJob = personalizeClient
                .describeDatasetImportJob(describeDatasetImportJobRequest)
                .datasetImportJob();
            status = datasetImportJob.status();
            System.out.println("Dataset import job status: " + status);
            if (status.equals("ACTIVE") || status.equals("CREATE FAILED")) {
                break;
            } try {
                Thread.sleep(waitInMilliseconds);
            } catch (InterruptedException e) {
                System.out.println(e.getMessage());
            }
        }
        return datasetImportJobArn;
    } catch (PersonalizeException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
        return "";
    }
}
```

SDK for JavaScript v3

```javascript
// Get service clients and commands using ES6 syntax.
import { CreateDatasetImportJobCommand, PersonalizeClient } from
    "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
```
The response from the `DescribeDatasetImportJob` (p. 509) operation includes the status of the operation.

You must wait until the status changes to ACTIVE before you can use the data to train a model.

The dataset import is complete when the status shows as ACTIVE. After you import data into an Amazon Personalize dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset. For more information, see `Managing data` (p. 292).

After you import your data into the relevant datasets in the dataset group, you can create a solution version (trained model). For more information, see `Creating a solution and a solution version` (p. 210).

### Importing individual records

After you have complete `Creating a dataset and a schema` (p. 171), you can import individual records, including interaction `events`, users, or items, into an existing dataset. Importing data individually allows you to add small batches of records to your Amazon Personalize datasets as your catalog grows. You can import up to 10 records per individual import operation.

If you use Apache Kafka, you can use the `Kafka connector for Amazon Personalize` to stream data in real time to Amazon Personalize. For information see `Kafka Connector for Amazon Personalize` in the `personalize-kafka-connector` Github repository.

If you have a large amount of historical records, we recommend that you first import data in bulk and then import data individually as necessary. See `Importing data directly into Amazon Personalize datasets` (p. 170).

### Filter updates for individual record imports

Amazon Personalize updates any filters you created in the dataset group with your new interaction, item, and user data within 20 minutes from the last individual import. This update allows your campaigns to use your most recent data when filtering recommendations for your users.
If you already created a recommender or deployed a custom solution version with a campaign, how new individual records influence recommendations depends on the domain use case or recipe that you use. For more information, see How new data influences real-time recommendations (p. 292).

Topics
- Importing interactions individually (p. 183)
- Importing users individually (p. 184)
- Importing items individually (p. 187)

Importing interactions individually

After you complete Creating a dataset and a schema (p. 171) to create an Interactions dataset, you can individually import one or more new events into the dataset. To import interaction events individually, you create an event tracker and then import one or more events into your Interactions dataset. You can import historical individual interaction events using the Amazon Personalize console, or import historical or real-time events using the AWS Command Line Interface (AWS CLI), or the AWS SDKs.

This section includes information about importing events with the Amazon Personalize console. We recommend using the Amazon Personalize console to import only historical events. For information about using the AWS CLI or the AWS SDKs to record events in real-time, see Recording events (p. 280).

For information about how Amazon Personalize updates filters for new records and how new records influence recommendations, see Importing individual records (p. 182).

Topics
- Creating an event tracker (console) (p. 183)
- Importing events individually (console) (p. 184)

Creating an event tracker (console)

Note
If you’ve created an event tracker, you can skip to Importing events individually (console) (p. 184).

Before you can import an event to an Interactions dataset, you must create an event tracker for the dataset group.

To create an event tracker (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose the dataset group with the Interactions dataset that you want to import events to.
3. On the Dashboard for the dataset group, in Install event ingestion SDK, choose Start.
4. On the Configure tracker page, in Tracker configurations, for Tracker name, provide a name for the event tracker, and choose Next.
5. The Install the SDK page shows the Tracking ID for the new event tracker and instructions for using AWS Amplify or AWS Lambda to stream event data.

You can ignore this information because you’re using the Amazon Personalize console to upload event data. If you want to stream event data using AWS Amplify or AWS Lambda in the future, you can view this information by choosing the event tracker on the Event trackers page.

6. Choose Finish. You can now import events with the console (see Importing events individually (console) (p. 184) or record events in real time using the PutEvent’s operation (see Recording events (p. 280)).
Importing events individually (console)

After you create an event tracker, you can import events individually into an Interactions dataset. This procedure assumes you have already created an Interactions dataset. For information about creating datasets, see Creating a dataset and a schema (p. 171).

To import events individually (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose the dataset group with the Interactions dataset that you want to import events to.
3. In the navigation pane, choose datasets.
4. On the Datasets page, choose the Interactions dataset.
5. At the top right of the dataset details page, choose Modify dataset, and choose Create record.
6. In Create user-item interaction record(s) page, for Record input, enter the event details in JSON format. The event's field names and values must match the schema that you used when you created the Interactions dataset. Amazon Personalize provides a JSON template with field names and data types from this schema. You can import up to 10 events at a time.
7. Choose Create record(s). In Response, the result of the import is listed and a success or failure message is displayed.

Importing users individually

After you complete Creating a dataset and a schema (p. 171) to create a Users dataset, you can individually import one or more new users into the dataset. Individually importing users allows you to keep your Users dataset current with small batch imports as your catalog grows. You can import up to 10 users at a time. If you have a large amount of new users, we recommend that you first import data in bulk and then import user data individually as necessary. See Importing data directly into Amazon Personalize datasets (p. 170).

You can use the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), or AWS SDKs to import users. If you import a user with the same userId as a user that's already in your Users dataset, Amazon Personalize replaces the user with the new one. You can import up to 10 users at a time.

For information about how Amazon Personalize updates filters for new records and how new records influence recommendations, see Importing individual records (p. 182).

Topics

- Importing users individually (console) (p. 184)
- Importing users individually (AWS CLI) (p. 185)
- Importing users individually (AWS SDKs) (p. 185)

Importing users individually (console)

You can import up to 10 users at a time. This procedure assumes you have already created a Users dataset. For information about creating datasets, see Creating a dataset and a schema (p. 171).

To import users individually (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose the dataset group with the Users dataset that you want to import the user to.
3. In the navigation pane, choose **Datasets**.
4. On the **Datasets** page, choose the Users dataset.
5. On the dataset details page, at the top right, choose **Modify dataset** and choose **Create record**.
6. On the **Create user record(s)** page, for record input, enter the user details in JSON format. The user's field names and values must match the schema you used when you created the Users dataset. Amazon Personalize provides a JSON template with field names and data types from this schema.
7. Choose **Create record(s)**. In **Response**, the result of the import is listed and a success or failure message is displayed.

**Importing users individually (AWS CLI)**

Add one or more users to your Users dataset with the **PutUsers** (p. 602) operation. You can import up to 10 users with a single **PutUsers** call. This section assumes that you have already created an Users dataset. For information about creating datasets, see **Creating a dataset and a schema** (p. 171).

Use the following `put-users` command to add one or more users with the AWS CLI. Replace `dataset arn` with the Amazon Resource Name (ARN) of your dataset and `userId` with the ID of the user. If an user with the same `userId` is already in your Users dataset, Amazon Personalize replaces it with the new one.

For properties, for each field in your Users dataset, replace the `propertyName` with the field name from your schema in camel case. For example, `GENDER` would be `gender` and `MEMBERSHIP_TYPE` would be `membershipType`. Replace `user data` with the data for the user. For categorical string data, to include multiple categories for a single property, separate each category with a pipe (`|`). For example "Premium class|Legacy Member".

```
aws personalize-events put-users --dataset-arn dataset arn --users '[{
    "userId": "userId",
    "properties": "{"propertyName": "user data"}"
},
{
    "userId": "userId",
    "properties": "{"propertyName": "user data"}"
}]
```

**Importing users individually (AWS SDKs)**

Add one or more users to your Users dataset with the **PutUsers** (p. 602) operation. If a user with the same `userId` is already in your Users dataset, Amazon Personalize replaces it with the new one. You can import up to 10 users with a single **PutUsers** call. This section assumes that you have already created a Users dataset. For information about creating datasets, see **Creating a dataset and a schema** (p. 171).

The following code shows how to add one or more users to your Users dataset. For each property name parameter, pass the field name from your schema in camel case. For example, `GENDER` would be `gender` and `MEMBERSHIP_TYPE` would be `membershipType`. For each property value parameter, pass the data for the user.

For categorical string data, to include multiple categories for a single property, separate each category with a pipe (`|`). For example "Premium class|Legacy Member".

**SDK for Python (Boto3)**

```python
import boto3

personalize_events = boto3.client(service_name='personalize-events')
```
importing individual records

```java
personalize_events.put_users(
    datasetArn = 'dataset arn',
    users = [
        {'userId': 'user ID',
         'properties': '{"propertyName": "user data"}'}],
        {'userId': 'user ID',
         'properties': '{"propertyName": "user data"}'}
)
```

**SDK for Java 2.x**

```java
public static int putUsers(PersonalizeEventsClient personalizeEventsClient,
                        String datasetArn,
                        String user1Id,
                        String user1PropertyName,
                        String user1PropertyValue,
                        String user2Id,
                        String user2PropertyName,
                        String user2PropertyValue) {

    int responseCode = 0;
    ArrayList<User> users = new ArrayList<>();

    try {
        User user1 = User.builder()
            .userId(user1Id)
            .properties(String.format("{"%1$s": "%2$s"}", user1PropertyName,
                user1PropertyValue))
            .build();

        users.add(user1);

        User user2 = User.builder()
            .userId(user2Id)
            .properties(String.format("{"%1$s": "%2$s"}", user2PropertyName,
                user2PropertyValue))
            .build();

        users.add(user2);

        PutUsersRequest putUsersRequest = PutUsersRequest.builder()
            .datasetArn(datasetArn)
            .build();

        responseCode = personalizeEventsClient.putUsers(putUsersRequest).sdkHttpResponse().statusCode();
        System.out.println("Response code: "+ responseCode);
        return responseCode;
    }

    catch (PersonalizeEventsException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    }

    return responseCode;
}
```

**SDK for JavaScript v3**

```javascript
// Get service clients module and commands using ES6 syntax.
import { PutUsersCommand } from '@aws-sdk/client-personalize-events';
import { personalizeEventsClient } from './libs/personalizeClients.js';
```
Importing items individually

After you complete Creating a dataset and a schema (p. 171) to create an Items dataset, you can individually import one or more new items into the dataset. Individually importing items allows you to keep your Items dataset current with small batch imports as your catalog grows. You can import up to 10 items at a time. If you have a large amount of new items, we recommend that you first import data in bulk and then import item data individually as necessary. See Importing data directly into Amazon Personalize datasets (p. 170).

You can use the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), or AWS SDKs to import items. If you import an item with the same itemId as an item that's already in your Items dataset, Amazon Personalize replaces it with the new item.

For information about how Amazon Personalize updates filters for new records and how new records influence recommendations, see Importing individual records (p. 182).

Topics

- Importing items individually (console) (p. 187)
- Importing items individually (AWS CLI) (p. 188)
- Importing items individually (AWS SDKs) (p. 188)

Importing items individually (console)

You can import up to 10 items to an Items dataset at a time. This procedure assumes that you have already created an Items dataset. For information about creating datasets, see Creating a dataset and a schema (p. 171).

To import items individually (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the **Dataset groups** page, choose the dataset group with the Items dataset that you want to import the items to.

3. In the navigation pane, choose **Datasets**.

4. On the **Datasets** page, choose the Items dataset.

5. At the top right of the dataset details page, choose **Modify dataset**, and then choose **Create record**.

6. In **Create item record(s)** page, for **Record input**, enter the item details in JSON format. The item's field names and values must match the schema you used when you created the Items dataset. Amazon Personalize provides a JSON template with field names and data types from this schema.

7. Choose **Create record(s)**. In **Response**, the result of the import is listed and a success or failure message is displayed.

### Importing items individually (AWS CLI)

Add one or more items to your Items dataset using the [PutItems](p. 600) operation. You can import up to 10 items with a single PutItems call. This section assumes that you have already created an Items dataset. For information about creating datasets, see [Creating a dataset and a schema](p. 171).

Use the following put-items command to add one or more items with the AWS CLI. Replace dataset arn with the Amazon Resource Name (ARN) of your dataset and item Id with the ID of the item. If an item with the same itemId is already in your Items dataset, Amazon Personalize replaces it with the new one.

For properties, for each field in your Items dataset, replace the propertyName with the field name from your schema in camel case. For example, GENRES would be genres and CREATION_TIMESTAMP would be creationTimestamp. Replace item data with the data for the item. CREATION_TIMESTAMP data must be in [Unix epoch time format](p. 103) and in seconds. For categorical string data, to include multiple categories for a single property, separate each category with a pipe (`|`). For example "Horror|Action".

```bash
aws personalize-events put-items 
--dataset-arn dataset arn 
--items '[[
   "itemId": "item Id",
   "properties": "\"propertyName\": "\item data\""
],
   {
   "itemId": "item Id",
   "properties": "\"propertyName\": "\item data\""
}]
```

### Importing items individually (AWS SDKs)

Add one or more items to your Items dataset using the PutItems (p. 600) operation. You can import up to 10 items with a single PutItems call. If an item with the same itemId is already in your Items dataset, Amazon Personalize replaces it with the new one. This section assumes that you have already created an Items dataset. For information about creating datasets, see [Creating a dataset and a schema](p. 171).

The following code shows how to add one or more items to your Items dataset. For each property name parameter, pass the field name from your schema in camel case. For example, GENRES would be genres and CREATION_TIMESTAMP would be creationTimestamp. For each property value parameter, pass the data for the item. CREATION_TIMESTAMP data must be in [Unix epoch time format](p. 103) and in seconds.

For categorical string data, to include multiple categories for a single property, separate each category with a pipe (`|`). For example "Horror|Action".
SDK for Python (Boto3)

```python
import boto3

personalize_events = boto3.client(service_name='personalize-events')

personalize_events.put_items(
    datasetArn = 'dataset arn',
    items = 
        [
            {'itemId': 'item ID',
             'properties': "{"propertyName": "item data"}"
            },
            {'itemId': 'item ID',
             'properties': "{"propertyName": "item data"}"
            }
        ]
)
```

SDK for Java 2.x

```java
public static int putItems(PersonalizeEventsClient personalizeEventsClient,
                        String datasetArn,
                        String item1Id,
                        String item1PropertyName,
                        String item1PropertyValue,
                        String item2Id,
                        String item2PropertyName,
                        String item2PropertyValue) {

    int responseCode = 0;
    ArrayList<Item> items = new ArrayList<>();

    try {
        Item item1 = Item.builder()
                        .itemId(item1Id)
                        .properties(String.format("{"%1$s": "%2$s"}",
                                                item1PropertyName, item1PropertyValue))
                        .build();
        items.add(item1);

        Item item2 = Item.builder()
                        .itemId(item2Id)
                        .properties(String.format("{"%1$s": "%2$s"}",
                                                item2PropertyName, item2PropertyValue))
                        .build();
        items.add(item2);

        PutItemsRequest putItemsRequest = PutItemsRequest.builder()
                                              .datasetArn(datasetArn)
                                              .items(items)
                                              .build();

        responseCode = personalizeEventsClient.putItems(putItemsRequest).sdkHttpResponse().statusCode();
        System.out.println("Response code: " + responseCode);
        return responseCode;
    } catch (PersonalizeEventsException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    }
    return responseCode;
}
```
Step 3: Creating recommenders or custom resources

After you import data, you are ready to create recommenders or custom resources. Use these resources to get recommendations. The resources you create depend on your dataset group type:

- For Domain dataset groups, you create recommenders for pre-defined use cases based on your domain. You use the recommenders to get recommendations. For information about available use cases, see Choosing a use case (p. 107). You can also add custom resources to a Domain dataset group. These include solutions and solution versions trained for custom use cases.
- For Custom dataset groups, you configure a solution with a recipe. Then you create a solution version (train a model). For information about available recipes, see Choosing a recipe (p. 113).

For real-time recommendations, you deploy the solution version in a campaign. For batch recommendations and user segments, you don't need a campaign.

Topics
- Creating domain recommenders (p. 191)
- Creating custom resources (p. 210)
Creating domain recommenders

After you import data, you are ready to start creating, evaluating, and managing recommenders in your Domain dataset group. A recommender is a Domain dataset group resource that generates recommendations. You use it in your application to get real-time recommendations with the GetRecommendations (p. 608) operation.

Topics

• Creating recommenders (p. 191)
• Evaluating a recommender (p. 199)
• Managing recommenders (p. 203)

Creating recommenders

After you create a Domain dataset group and import data, you can create recommenders for your domain use cases. A recommender is a Domain dataset group resource that generates recommendations. You use a recommender in your application to get real-time recommendations with the GetRecommendations (p. 608) operation.

When you create a recommender, you specify a use case and Amazon Personalize trains the models backing the recommender with the best configurations for the use case. Each use case has different API requirements for getting recommendations. For a list of recommender use cases by domain, see Choosing a use case (p. 107). You can create at most 15 recommenders per region.

Amazon Personalize automatically retrains the models backing your recommenders every 7 days. This is a full retraining that creates entirely new models based on the entirety of the data in your datasets. With Top picks for you and Recommended for you use cases, Amazon Personalize updates the existing models every two hours to include new items in recommendations with exploration.

You can create recommenders with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

Recommender statuses

A recommender can be in one of the following states:

• CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
• DELETE PENDING > DELETE IN_PROGRESS

To get the recommender status, navigate to the Recommenders page in the Amazon Personalize console or use the DescribeRecommender (p. 521) operation.

Topics

• Minimum recommendation requests per second and auto-scaling (p. 191)
• Creating recommenders (console) (p. 192)
• Creating recommenders (AWS CLI) (p. 193)
• Creating recommenders (AWS SDKs) (p. 195)

Minimum recommendation requests per second and auto-scaling

Important

A high minRecommendationRequestsPerSecond will increase your bill. We recommend starting with 1 for minRecommendationRequestsPerSecond (the
Creating domain recommenders

default). Track your usage using Amazon CloudWatch metrics, and increase the minRecommendationRequestsPerSecond as necessary.

When you create a recommender, you can configure the recommender's minimum recommendation requests per second. The minimum recommendation requests per second (minRecommendationRequestsPerSecond) specifies the baseline recommendation request throughput provisioned by Amazon Personalize. The default minRecommendationRequestsPerSecond is 1. A recommendation request is a single GetRecommendations operation. Request throughput is measured in requests per second and Amazon Personalize uses your requests per second to derive your requests per hour and the price of your recommender usage.

If your requests per second increases beyond minRecommendationRequestsPerSecond, Amazon Personalize auto-scales the provisioned capacity up and down, but never below minRecommendationRequestsPerSecond. There's a short time delay while the capacity is increased that might cause loss of requests.

Your bill is the greater of either the minimum requests per hour (based on minRecommendationRequestsPerSecond) or the actual number of requests. The actual request throughput used is calculated as the average requests/second within a one-hour window. We recommend starting with the default minRecommendationRequestsPerSecond, track your usage using Amazon CloudWatch metrics, and then increase the minRecommendationRequestsPerSecond as necessary.

Creating recommenders (console)

Important
A high minRecommendationRequestsPerSecond will increase your bill. We recommend starting with 1 for minRecommendationRequestsPerSecond (the default). Track your usage using Amazon CloudWatch metrics, and increase the minRecommendationRequestsPerSecond as necessary. For more information see Minimum recommendation requests per second and auto-scaling (p. 191).

Create recommenders for each of your use cases with the Amazon Personalize console as follows. If you just created your Domain dataset group and you are already on the Overview page, skip to step 3.

To create recommenders

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your Domain dataset group.
3. On the Use <domain name> recommenders tab of the middle card choose Create recommenders.
4. On the Choose use cases page, choose the use cases you want to create recommenders and give each a Recommender name. Amazon Personalize creates a recommender for each use case that you choose. The available use cases depend on your domain. For information on choosing a use case see Choosing a use case (p. 107).
5. Choose Next.
6. On the Advanced configuration page, configure each recommender depending on your business needs:

   • For each dataset used by the recommender’s use case, you can choose the columns Amazon Personalize considers when training the models backing your recommender. By default, Amazon Personalize uses all columns that can be used when training. Columns with the boolean data type and custom string fields that aren't categorical or textual aren't used when training. You can't exclude EVENT_TYPE columns.

   You can change the columns used when training to control what data Amazon Personalize uses when creating your recommender. You might do this to experiment with different combinations
of training data. Or you might exclude columns without meaningful data. For example, you might have a column that you want to use only to filter recommendations. You can exclude this column from training and Amazon Personalize considers it only when filtering.

- You can modify **Minimum recommendation requests per second** to specify a new minimum request capacity for your recommender. A high minRecommendationRequestsPerSecond will increase your bill. We recommend starting with 1 (the default). Track your usage using Amazon CloudWatch metrics, and increase the minRecommendationRequestsPerSecond as necessary. For more information see **Minimum recommendation requests per second and autoscaling** (p. 191).

- For Top picks for your or Recommended for you use cases, optionally make changes to exploration configuration. Exploration involves testing different item recommendations to learn how users respond to items with very little interaction data. Use the following fields to configure exploration:
  
  - Emphasis on exploring less relevant items (exploration weight) – Configure how much to explore. Specify a decimal value between 0 to 1. The default is 0.3. The closer the value is to 1, the more exploration. With more exploration, recommendations include more items with less interactions data or relevance based on previous behavior. At zero, no exploration occurs and recommendations are based on current data (relevance).
  
  - Exploration item age cutoff – Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines item age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines item age, see **Creation timestamp data** (p. 81).

To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them.

- For Tag, optionally add any tags. For more information about tagging Amazon Personalize resources, see **Tagging Amazon Personalize resources** (p. 370).

7. Choose **Create recommenders** to create recommenders for each of your use cases.

You can monitor the status of each recommender on the **Recommenders** page. When your recommender status is Active, you can use it in your application to get recommendations.

### Creating recommenders (AWS CLI)

After you create a Domain dataset group and import data, you can create recommenders for your domain use cases. A recommender is a Domain dataset group resource that generates recommendations.

For Top picks for your or Recommended for you use cases, Amazon Personalize uses exploration when recommending items. For more information, see **Configuring exploration** (p. 194).

### Topics

- **Creating a recommender** (p. 193)
- **Configuring columns used when training** (p. 194)
- **Configuring exploration** (p. 194)

### Creating a recommender

Use the following AWS CLI code to create a recommender for a domain use case. Run this code for each of your domain use cases. For recipeArn, provide the Amazon Resource Name (ARN) for your use case. The available use cases depend on your domain. For a list of use cases and their ARNs see **Choosing a use case** (p. 107).
.aws personalize create-recommender \
--name recommender name \
--dataset-group-arn dataset group ARN \
--recipe-arn recipe ARN

Configuring columns used when training

You can modify the columns Amazon Personalize considers when training the models backing your recommender. By default, Amazon Personalize uses all columns that can be used when training. Columns with the boolean data type and custom non-categorical string fields aren't used. You can't exclude EVENT_TYPE columns.

You can change the columns used when training to control what data Amazon Personalize uses when creating your recommender. You might do this to experiment with different combinations of training data. Or you might exclude columns without meaningful data. For example, you might have a column that you want to use only to filter recommendations. You can exclude this column from training and Amazon Personalize considers it only when filtering.

To exclude columns from training, provide the excludedDatasetColumns object in the trainingDataConfig as part of the recommender configuration. For each key in the object, provide the dataset type. For each value, provide the list of columns to exclude.

.aws personalize create-recommender \
--name recommender name \
--dataset-group-arn dataset group ARN \
--recipe-arn recipe ARN \
--recommender-config "{"trainingDataConfig": {"excludedDatasetColumns": { "datasetType": [ "column1Name", "column2Name"]}}}" 

Configuring exploration

For Top picks for your or Recommended for you use cases, Amazon Personalize uses exploration when recommending items. Exploration involves testing different item recommendations to learn how users respond to items with very little interaction data. You can configure exploration with the following:

- **Emphasis on exploring less relevant items (exploration weight)** – Configure how much to explore. Specify a decimal value between 0 to 1. The default is 0.3. The closer the value is to 1, the more exploration. With more exploration, recommendations include more items with less interactions data or relevance based on previous behavior. At zero, no exploration occurs and recommendations are based on current data (relevance).

- **Exploration item age cutoff** – Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines item age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines item age, see [Creation timestamp data](p. 81).

To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them.

The following code shows how to configure exploration when you create a recommender for the Top picks for you use case. The example uses the default values.

.aws personalize create-recommender \
--name recommender name \
--dataset-group-arn dataset group ARN \
--recipe-arn arn:aws:personalize:::recipe/aws-vod-top-picks
Creating domain recommenders

Creating recommenders (AWS SDKs)

After you create a Domain dataset group and import data, you can create recommenders for your domain use cases. A recommender is a Domain dataset group resource that generates recommendations.

For all use cases, you can configure the columns used when training. For more information, see Configuring columns used when training (p. 198). For Top picks for your or Recommended for you use cases, Amazon Personalize uses exploration when recommending items. For more information, see Configuring exploration (p. 197).

Topics

- Creating a recommender (p. 195)
- Configuring exploration (p. 197)
- Configuring columns used when training (p. 198)

Creating a recommender

Create a recommender for a domain use case with the following code. Give your recommender a name and provide your Domain dataset group's Amazon Resource Name (ARN). For recipeArn, provide the ARN for your use case. Run this code for each of your domain use cases. The available use cases depend on your domain. For a list of use cases, their ARNs, and their requirements, see Choosing a use case (p. 107).

SDK for Python (Boto3)

```python
import boto3
personalize = boto3.client('personalize')
create_recommender_response = personalize.create_recommender(
    name = 'recommender name',
    recipeArn = 'recipe ARN',
    datasetGroupArn = 'dataset group ARN'
)
recommender_arn = create_recommender_response['recommenderArn']
print('Recommender ARN: ' + recommender_arn)
```

SDK for Java 2.x

```java
public static String createRecommender(PersonalizeClient personalizeClient, 
    String name, 
    String datasetGroupArn, 
    String recipeArn) {

    long maxTime = 0;
    long waitInMilliseconds = 30 * 1000; // 30 seconds
    String recommenderStatus = "";

    try {
        CreateRecommenderRequest createRecommenderRequest = 
            CreateRecommenderRequest.builder().
            .datasetGroupArn(datasetGroupArn).
            .name(name).
            .recipeArn(recipeArn)
            ...
```

---

--recommender-config "{"itemExplorationConfig":{"explorationWeight":0.3,"explorationItemAgeCutOff":30}}"
.build();

CreateRecommenderResponse recommenderResponse =
personalizeClient.createRecommender(createRecommenderRequest);
String recommenderArn = recommenderResponse.recommenderArn();
System.out.println("The recommender ARN is " + recommenderArn);

DescribeRecommenderRequest describeRecommenderRequest =
DescribeRecommenderRequest.builder()
  .recommenderArn(recommenderArn)
  .build();

maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
while (Instant.now().getEpochSecond() < maxTime) {
  recommenderStatus =
  personalizeClient.describeRecommender(describeRecommenderRequest).recommender().status();
  System.out.println("Recommender status: " + recommenderStatus);
  if (recommenderStatus.equals("ACTIVE") ||
      recommenderStatus.equals("CREATE FAILED")) {
    break;
  }
  try {
    Thread.sleep(waitInMilliseconds);
  } catch (InterruptedException e) {
    System.out.println(e.getMessage());
  }
}
return recommenderArn;

} catch(PersonalizeException e) {
  System.err.println(e.awsErrorDetails().errorMessage());
  System.exit(1);
}
return "";

SDK for JavaScript v3

// Get service clients and commands using ES6 syntax.
import { CreateRecommenderCommand, PersonalizeClient } from
  "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
  region: "REGION"
});

// set the recommender's parameters
export const createRecommenderParam = {
  name: "RECOMMENDER_NAME",        /* required */
  recipeArn: "RECIPE_ARN",         /* required */
  datasetGroupArn: "DATASET_GROUP_ARN" /* required */
}

export const run = async () => {
  try {
    const response = await personalizeClient.send(new
      CreateRecommenderCommand(createRecommenderParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
Configuring exploration

For **Top picks for your** or **Recommended for you** use cases, Amazon Personalize uses exploration when recommending items. Exploration involves testing different item recommendations to learn how users respond to items with very little interaction data. You can configure exploration with the following:

- **Emphasis on exploring less relevant items (exploration weight)** – Configure how much to explore. Specify a decimal value between 0 to 1. The default is 0.3. The closer the value is to 1, the more exploration. With more exploration, recommendations include more items with less interactions data or relevance based on previous behavior. At zero, no exploration occurs and recommendations are based on current data (relevance).

- **Exploration item age cutoff** – Specify the maximum item age in days since the latest interaction across all items in the Interactions dataset. This defines the scope of item exploration based on item age. Amazon Personalize determines item age based on its creation timestamp or, if creation timestamp data is missing, interactions data. For more information how Amazon Personalize determines item age, see [Creation timestamp data](#).

To increase the items Amazon Personalize considers during exploration, enter a greater value. The minimum is 1 day and the default is 30 days. Recommendations might include items that are older than the item age cut off you specify. This is because these items are relevant to the user and exploration didn't identify them.

The following code shows how to configure exploration when you create a recommender. The example uses the default values.

**SDK for Python (Boto3)**

```python
import boto3
c

personalize = boto3.client('personalize')

c_create_recommender_response = personalize.create_recommender(
    name = 'recommender name',
    recipeArn = 'arn:aws:personalize:::recipe/aws-vod-top-picks',
    datasetGroupArn = 'dataset group ARN',
    recommenderConfig = {
        "itemExplorationConfig": {
            "explorationWeight": "0.3",
            "explorationItemAgeCutOff": "30"
        }
    }
)

c_recommender_arn = create_recommender_response['recommenderArn']

c_print('Recommender ARN:' + recommender_arn)
```

**SDK for JavaScript v3**

```javascript
// Get service clients and commands using ES6 syntax.
import { CreateRecommenderCommand, PersonalizeClient } from
"@aws-sdk/client-personalize";

// create personalizeClient
c
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

// set the recommender's parameters
```
Configuring columns used when training

You can modify the columns Amazon Personalize considers when training the models backing your recommender. By default, Amazon Personalize uses all columns that can be used when training. Columns with the boolean data type and custom non-categorical string fields aren't used. You can't exclude EVENT_TYPE columns.

You can change the columns used when training to control what data Amazon Personalize uses when creating your recommender. You might do this to experiment with different combinations of training data. Or you might exclude columns without meaningful data. For example, you might have a column that you want to use only to filter recommendations. You can exclude this column from training and Amazon Personalize considers it only when filtering.

To exclude columns from training, provide the excludedDatasetColumns object in the trainingDataConfig as part of the recommender configuration. For each key, provide the dataset type. For each value, provide the list of columns to exclude. The following code shows how to exclude columns from training when you create a recommender.

SDK for Python (Boto3)

```python
import boto3

personalize = boto3.client('personalize')

create_recommender_response = personalize.create_recommender(
    name = 'recommender name',
    recipeArn = 'recipe name',
    datasetGroupArn = 'dataset group ARN',
    recommenderConfig = {
        "trainingDataConfig": {
            "excludedDatasetColumns": {
                "datasetType": ["COLUMN_A", "COLUMN_B"]
            }
        }
    }
)

recommender_arn = create_recommender_response['recommenderArn']
```
print('Recommender ARN:' + recommender_arn)

SDK for JavaScript v3

// Get service clients and commands using ES6 syntax.
import { CreateRecommenderCommand, PersonalizeClient } from
"@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

// set the recommender's parameters
export const createRecommenderParam = {
    name: "RECOMMENDER_NAME", /* required */
    recipeArn: "RECIPE_ARN", /* required */
    datasetGroupArn: "DATASET_GROUP_ARN", /* required */
    recommenderConfig: {
        trainingDataConfig: {
            excludedDatasetColumns: {
                "DATASET_TYPE": ["COLUMN_A", "COLUMN_B"]
            }
        }
    }
};

export const run = async () => {
    try {
        const response = await personalizeClient.send(new
CreateRecommenderCommand(createRecommenderParam));
        console.log("Success", response);
        return response; // For unit tests.
    } catch (err) {
        console.log("Error", err);
    }
};

run();

Evaluating a recommender

You can evaluate the performance of your recommender through offline and online metrics. Online
metrics are the empirical results you observe in your users’ interactions with real-time recommendations.
For example, you might record your users’ click-through rate as they browse your catalog. You are
responsible for generating and recording any online metrics.

Offline metrics are the metrics that Amazon Personalize generates when you create a recommender. You
can use offline metrics to evaluate the performance of the recommender's underlying model. Offline
metrics allow you to compare the model with other models trained on the same data. For the rest of this
section, the term metrics refers to offline metrics.

To get performance metrics, Amazon Personalize splits the input interactions data into a training set and
a testing set. The training set consists of 90% of your users and their interactions data. The testing set
consists of the remaining 10% of users and their interactions data.

Amazon Personalize then creates the recommender using the training set. After training completes,
Amazon Personalize gives the new recommender the oldest 90% of each user’s data from the testing
set as input. Amazon Personalize then calculates metrics by comparing the recommendations the
recommender generates to the actual interactions in the newest 10% of each user’s data from the
testing set.
Retrieving metrics

After your recommender is active, you can view the metrics for the recommender in the Amazon Personalize console or retrieve metrics by calling the `DescribeRecommender` operation.

Topics
- Viewing metrics (console) (p. 200)
- Retrieving metrics (AWS CLI) (p. 200)
- Retrieving metrics (AWS SDKs) (p. 201)

Retrieving metrics (AWS CLI)

The following code shows how to get metrics for a recommender with the AWS CLI.

```bash
aws personalize describe-recommender \
  --recommender-arn recommender arn
```

The following is an example of the metrics output from a recommender created for the `Top picks for you` use case for the `VIDEO_ON_DEMAND` domain.

```json
{
    "recommender": {
        "recommenderArn": "arn:aws:personalize:region:acct-id:recommender/recommenderName",
        "name": "name123",
        "recipeArn": "arn:aws:personalize:::recipe/aws-vod-top-picks",
        "modelMetrics": {
            "coverage": 0.27,
            "mean_reciprocal_rank_at_25": 0.0379,
            "normalized_discounted_cumulative_gain_at_5": 0.0405,
            "normalized_discounted_cumulative_gain_at_10": 0.0513,
            "normalized_discounted_cumulative_gain_at_25": 0.0828,
            "precision_at_5": 0.0136,
            "precision_at_10": 0.0102,
            "precision_at_25": 0.0091,
        }
    }
}
```
Creating domain recommenders

Retrieving metrics (AWS SDKs)

The following code shows how to get metrics for a recommender with the SDK for Python (Boto3).

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.describe_recommender(
    recommenderArn = 'recommender_arn'
)

print(response['recommender']['modelMetrics'])
```

The following is an example the metrics output from a recommender created for the *Top picks for you* use case for the VIDEO_ON_DEMAND domain.

```
{
    "recommender": {
        "recommenderArn": "arn:aws:personalize:region:acct-id:recommender/recommenderName",
        "name": "name123",
        "recipeArn": "arn:aws:personalize:::recipe/aws-vod-top-picks",
        "modelMetrics": {
            "coverage": 0.27,
            "mean_reciprocal_rank_at_25": 0.0379,
            "normalized_discounted_cumulative_gain_at_5": 0.0405,
            "normalized_discounted_cumulative_gain_at_10": 0.0513,
            "normalized_discounted_cumulative_gain_at_25": 0.0828,
            "precision_at_5": 0.0136,
            "precision_at_10": 0.0102,
            "precision_at_25": 0.0091,
        },
        "recommenderConfig": {},
        "creationDateTime": "2022-05-06T10:11:24.589000-07:00",
        "lastUpdatedDateTime": "2022-05-06T10:34:33.270000-07:00",
        "status": "ACTIVE",
    },
}
```

Metric definitions

The metrics Amazon Personalize generates for recommenders are described below using the following terms:

- **Relevant recommendation** is a recommendation for an item that the user actually interacted with. These items are from the newest 10% of each user's interactions data from the testing set.
- **Rank** refers to the position of a recommended item in the list of recommendations. Position 1 (the top of the list) is presumed to be the most relevant to the user.

For each metric, higher numbers (closer to 1) are better. To dive deeper, see the resources listed in Additional resources (p. 203).

**coverage**

The value for coverage tells you the proportion of unique items that Amazon Personalize might recommend out of the total number of unique items in Interactions and Items datasets. A higher
coverage score means Amazon Personalize recommends more of your items, rather than the same few items repeatedly for different users. Use cases that feature item exploration, such as the Top picks for you (VIDEO_ON_DEMAND) and Recommended for you (ECOMMERCE), have higher coverage than those that don’t.

mean reciprocal rank at 25

This metric tells you about a model's ability to generate a relevant recommendation at the top ranked position. You might choose a model with a high mean reciprocal rank at 25 if you are generating relevant search results for a user, and don’t expect the user to choose an item lower on the list. For example, users frequently choose the first cooking recipe in search results.

Amazon Personalize calculates this metric using the average reciprocal rank score for requests for recommendations. Each reciprocal rank score is calculated as: 1 / the rank of the highest item interacted with by the user, where the total possible rankings is 25. Other lower ranked items the user interacts with are ignored. If the user chose the first item, the score is 1. If they don’t choose any items, the score is 0.

For example, you might show three different users 25 recommendations each:

- If User 1 clicks the item at rank 4 and the item at rank 10, their reciprocal rank score is 1/4.
- If User 2 clicks an item at rank 2, an item at rank 4, and an item at rank 12, their reciprocal rank score is 1/2.
- If User 3 clicks on a single item at rank 6, their reciprocal rank score is 1/6.

The mean reciprocal rank over all requests for recommendations (in this case 3) is calculated as \((1/4 + 1/2 + 1/6) / 3 = .3056\).

normalized discounted cumulative gain (NDCG) at K (5, 10, or 25)

This metric tells you about how well your model ranks recommendations, where K is a sample size of 5, 10, or 25 recommendations. This metric is useful if you are most interested in the ranking of recommendations beyond just the highest ranked item (for this, see mean reciprocal rank at 25). For example, the score for NDCG at 10 would be useful if you have an application that shows up to 10 movies in a carousel at a time.

Amazon Personalize calculates the NDCG by assigning weight to recommendations based on their ranking position for each user in the testing set. Each recommendation is discounted (given a lower weight) by a factor dependent on its position. The final metric is the average of all users in the testing set. The normalized discounted cumulative gain at K assumes that recommendations that are lower on a list are less relevant than recommendations higher on the list.

Amazon Personalize uses a weighting factor of \(1/\log(1 + \text{position})\), where the top of the list is position 1.

precision at K

This metric tells you how relevant your model’s recommendations are based on a sample size of K (5, 10, or 25) recommendations.

Amazon Personalize calculates this metric based on the number of relevant recommendations out of the top K recommendations for each user in the testing set, divided by K, where K is 5, 10, or 25. The final metric is the average across all users in the testing set.

For example, if you recommend 10 items to a user, and the user interacts with 3 of them, the precision at K is 3 correctly predicted items divided by the total 10 recommended items: \(3 / 10 = .30\).

This metric rewards precise recommendation of relevant items. The closer the score is to one, the more precise the model.
Example

The following is a simple example for a recommender that produces a list of recommendations for a specific user. The second and fifth recommendations match records in the testing data for this user. These are the relevant recommendations. If $K$ is set at 5, the following metrics are generated for the user.

**reciprocal_rank**
- Calculation: $1/2$
- Result: 0.5000

**normalized_discounted_cumulative_gain_at_5**
- Calculation: $(1/\log(1 + 2) + 1/\log(1 + 5)) / (1/\log(1 + 1) + 1/\log(1 + 2))$
- Result: 0.6241

**precision_at_5**
- Calculation: $2/5$
- Result: 0.4000

Additional resources

To dive deeper in different types of metrics for recommender systems, see the following external resources:

- [MRR vs MAP vs NDCG: Rank-Aware Evaluation Metrics And When To Use Them](#)
- [Discounted Cumulative Gain: the ranking metrics you should know about](#)
- [Recall and Precision at k for Recommender Systems](#)
- [Ranking Evaluation Metrics for Recommender Systems](#)

Managing recommenders

You don't have to manage the models backing your recommenders. Amazon Personalize automatically retrains them every 7 days. This is a full retraining that creates entirely new models based on the entirety of the data in your datasets. For *Top picks for you* and *Recommended for you*, Amazon Personalize updates the existing models every two hours to consider new items for recommendations with exploration. For more information, see [Automatic updates](#).

Managing recommenders involves the following:

- **Stopping and starting recommenders** – If you want to pause billing for an active recommender, you can stop the recommender and restart it later. For more information, see [Stopping and starting a recommender](#).
- **Updating recommender configuration** – You can update the columns the recommender uses in training and update the recommender's request capacity. For more information, see [Updating a recommender](#).
- **Deleting a recommender** – You can delete recommenders with the [DeleteRecommender](#) operation. Or you can delete a recommender from the recommender details page in the Amazon Personalize console.

Topics

- [Updating a recommender](#)
• Stopping and starting a recommender (p. 206)

Updating a recommender

After you create a recommender, you can update the recommender's configuration:

• You can update the columns the recommender uses in training. If you modify the columns used when training, Amazon Personalize automatically starts a full retraining of the models backing your recommender. While the update completes, you can still get recommendations from the recommender. The recommender uses the previous configuration until the update completes. To track the status of this update, use the latestRecommenderUpdate returned in the DescribeRecommender (p. 521) operation. If you provide the same columns you provided when you created the recommender, no update occurs.
• You can update the recommender's minimum recommendation requests per second. This specifies the baseline recommendation request throughput that's provisioned by Amazon Personalize. A high value will increase your bill. We recommend starting with 1. Track your usage using Amazon CloudWatch metrics, and increase it as necessary. For more information, see Minimum recommendation requests per second and auto-scaling (p. 191).
• For Top picks for you and Recommended for you use cases, you can update exploration configuration by adjusting the emphasis on exploring relevant items and the exploration item age cutoff. For information about exploration, see the section for your use case in Choosing a use case (p. 107).

You can update recommenders with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

Topics
• Updating a recommender (Amazon Personalize console) (p. 204)
• Updating a recommender (AWS CLI) (p. 204)
• Updating a recommender (AWS SDKs) (p. 205)

Updating a recommender (Amazon Personalize console)

After you create a recommender, you can update it. You can update the columns the recommender uses in training and the recommender's minimum recommendation requests per second. For Top picks for you and Recommended for you use cases, you can update exploration configuration. To update a recommender with the console, do the following.

To update a recommender's configuration (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your Domain dataset group.
3. From the navigation pane, choose Recommenders.
4. On the Recommenders page, choose the recommender that you want to update.
5. In Recommender configuration choose Edit.
6. Change the recommender's configuration and choose Update. For information on the different configuration options, see Creating recommenders (console) (p. 192).

Updating a recommender (AWS CLI)

To update recommender with the AWS CLI, use the update-recommender command. Provide the Amazon Resource Name (ARN) for the recommender and updated configuration. The following code shows how to update the columns a recommender uses for training.
Creating domain recommenders

If you modify the columns used in training, Amazon Personalize automatically starts a full retraining of the models backing your recommender. While the update completes, you can still get recommendations from the recommender. The recommender uses the previous configuration until the update completes. To track the status of this update, use the latestRecommenderUpdate returned in the DescribeRecommender (p. 521) operation.

For more information about the different configurations you can change, see RecommenderConfig (p. 704).

Updating a recommender (AWS SDKs)

To update recommender with the AWS, use the UpdateRecommender (p. 596) operation. Provide the Amazon Resource Name (ARN) for the recommender and specify the new configuration. The following code shows how to update the columns a recommender uses for training.

**SDK for Python (Boto3)**

```python
import boto3

personalize = boto3.client('personalize')

update_recommender_response = personalize.update_recommender(
    recommenderArn = 'dataset group ARN',
    recommenderConfig = {
        "trainingDataConfig": {
            "excludedDatasetColumns": {
                "datasetType": ["column1Name", "column2Name"]
            }
        }
    }
)
```

**SDK for JavaScript v3**

```javascript
// Get service clients and commands using ES6 syntax.
import { UpdateRecommenderCommand, PersonalizeClient } from '@aws-sdk/client-personalize';

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

// set the request's parameters
export const updateRecommenderParam = {
    recommenderArn: "RECOMMENDER_ARN", /* required */
    recommenderConfig: {
        trainingDataConfig: {
            excludedDatasetColumns: {
                "DATASET_TYPE": ["COLUMN_A", "COLUMN_B"]
            }
        }
    }
};

export const run = async () => {
    //...
try {
    const response = await personalizeClient.send(new UpdateRecommenderCommand(updateRecommenderParam));
    console.log("Success", response);
    return response; // For unit tests.
} catch (err) {
    console.log("Error", err);
}
};
run();

If you modify the columns used in training in the excludedDataSetColumns of the recommenderConfig, Amazon Personalize automatically starts a full retraining of the models backing your recommender. While the update completes, you can still get recommendations from the recommender. The recommender uses the previous configuration until the update completes. To track the status of this update, use the latestRecommenderUpdate returned in the DescribeRecommender (p. 521) operation.

For more information about the different configurations you can change, see RecommenderConfig (p. 704).

Stopping and starting a recommender

After your recommender is active, you can stop a recommender and start it later. This way, you can pause recommender billing and only pay for it when you use it. For example, you might need to get recommendations only on certain days of the week. You can stop the recommender on the days you don't need it, and then start the recommender on the days you do.

After you stop a recommender, you can't use it to get recommendations. Stopping a recommender halts recommender billing and retraining. However, stopping a recommender doesn't delete the recommender. You can restart it at any time and resume getting recommendations. Starting a recommender doesn't create a new recommender with your data. Rather, it resumes recommender billing and retraining every 7 days.

You can stop and start a recommender with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), AWS SDKs, or the StartRecommender (p. 578) and StopRecommender (p. 580) API operations.

Recommender states

When you stop a recommender, the recommender state changes from ACTIVE to INACTIVE in the following sequence:

ACTIVE > STOP PENDING > STOP IN PROGRESS > INACTIVE

When you start a recommender, the recommender state changes from INACTIVE to ACTIVE in the following sequence:

INACTIVE > START PENDING > START IN PROGRESS > ACTIVE

Topics

- Stopping and starting a recommender (console) (p. 206)
- Stopping and restarting a recommender (AWS CLI) (p. 207)
- Stopping and restarting a recommender (AWS SDKs) (p. 208)

Stopping and starting a recommender (console)

You can use the Amazon Personalize to stop and restart a recommender.
Stopping a recommender (console)

You can use the Amazon Personalize console to stop an active recommender as follows.

To stop a recommender

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your Domain dataset group.
3. From the navigation pane, choose Recommenders.
4. On the Recommenders page, choose the recommender that you want to stop.
5. Choose Stop recommender at the top right and confirm on the window that displays.

When the recommender status is inactive, your recommender has stopped. This halts any recommender billing and retraining. You can't use the recommender until you start it.

Starting a recommender (console)

You can use the Amazon Personalize console to start an inactive recommender as follows.

To start a recommender

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your Domain dataset group.
3. From the navigation pane, choose Recommenders.
4. On the Recommenders page, choose the recommender that you want to start.
5. Choose Start recommender at the top right and confirm that you want to start it on the window that displays.

When the recommender status is active, you can resume getting recommendations from it. Recommender billing and automatic retraining also resumes.

Stopping and restarting a recommender (AWS CLI)

To stop an active recommender with the AWS CLI, use the stop-recommender command and provide the Amazon Resource Name (ARN) for the recommender as follows:

```bash
aws personalize stop-recommender --recommender-arn "recommender arn"
```

To start an inactive recommender with the AWS CLI, use the start-recommender command and provide the ARN for the stopped recommender as follows:

```bash
aws personalize start-recommender --recommender-arn "recommender arn"
```

For more information about the API operations, see StartRecommender (p. 578) and StopRecommender (p. 580).

---

Topics

- Stopping a recommender (console) (p. 207)
- Starting a recommender (console) (p. 207)
Stopping and restarting a recommender (AWS SDKs)

You can use the AWS SDKs to start an active recommender or stop an inactive recommender. For more information about the API operations, see `StartRecommender (p. 578)` and `StopRecommender (p. 580)`.

Topics
- Stopping a recommender (AWS SDKs) (p. 208)
- Starting a recommender (AWS SDKs) (p. 209)

Stopping a recommender (AWS SDKs)

The following code shows how to stop an active recommender with the AWS SDKs. Stopping halts any recommender billing and automatic retraining. You can't use the recommender until you restart it.

SDK for Python (Boto3)

To stop an active recommender with the SDK for Python (Boto3), use the `stop_recommender` method and provide the Amazon Resource Name (ARN) for the recommender as follows.

```python
import boto3
personalize = boto3.client('personalize')
stop_recommender_response = personalize stop_recommender(
    recommenderArn = "recommenderARN"
)
print(stop_recommender_response)
```

SDK for Java 2.x

To stop an active recommender with the SDK for Java 2.x, use the `stopRecommender` method and provide the ARN for the recommender as follows.

```java
public static void stopRecommender(PersonalizeClient personalizeClient,
        String datasetGroupArn) {
    try {
        StopRecommenderRequest stopRecommenderRequest =
                StopRecommenderRequest.builder()
                .recommenderArn(recommenderArn)
                .build();
        personalizeClient.stopRecommender(stopRecommenderRequest);
    } catch (PersonalizeException e) {
            System.out.println(e.awsErrorDetails().errorMessage());
    }
    return "";
}
```

SDK for JavaScript v3

```javascript
// Get service clients and commands using ES6 syntax.
import { StopRecommenderCommand, PersonalizeClient } from
    "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

// set the request params
```
export const stopRecommenderParam = {
  recommenderArn: "RECOMMENDER_ARN" /* required */
};

export const run = async () => {
  try {
    const response = await personalizeClient.send(
      new StopRecommenderCommand(stopRecommenderParam)
    );
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();

Starting a recommender (AWS SDKs)

The following code shows how to start an inactive recommender with the AWS SDKs. When the recommender status is active, you can resume getting recommendations from it. At the same time, recommender billing and automatic retraining also resumes.

SDK for Python (Boto3)

To start an inactive recommender with the SDK for Python (Boto3), use the start_recommender method and provide the Amazon Resource Name (ARN) for the recommender as follows.

```python
import boto3
personalize = boto3.client('personalize')

start_recommender_response = personalize start_recommender(
  recommenderArn = "recommenderARN"
)
print(start_recommender_response)
```

SDK for Java 2.x

To start an inactive recommender with the SDK for Java 2.x, use the startRecommender method and provide the ARN for the recommender as follows.

```java
public static void startRecommender(PersonalizeClient personalizeClient,
  String datasetGroupArn) {

  try {
    StartRecommenderRequest startRecommenderRequest =
    StartRecommenderRequest.builder()
    .recommenderArn(recommenderArn)
    .build();
    personalizeClient.startRecommender(startRecommenderRequest);
  } catch (PersonalizeException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
  }

  return "";
}
```

SDK for JavaScript v3

```javascript
// Get service clients and commands using ES6 syntax.
import { StartRecommenderCommand, PersonalizeClient } from
```
Creating custom resources

After you import your data, you are ready to create the custom resources that you use to get recommendations. To create the custom resources that generate recommendations, you do the following:

1. **Create and configure a solution**: Customize solution parameters and recipe-specific hyperparameters so the model meets your specific business needs.
2. **Create a solution version**: Create a solution version (train a model). The solution version generates Amazon Personalize recommendations or user segments.
3. **Deploy the solution version with a campaign (only for real-time recommendations)**: Create a campaign to deploy your solution version. You use the campaign when you request real-time recommendations. If you are getting batch recommendations, you don’t need to create a campaign.

Topics

- Creating a solution and a solution version (p. 210)
- Creating a campaign (p. 237)

Creating a solution and a solution version

After you finish importing data, you are ready to create a solution. A solution refers to the combination of an Amazon Personalize recipe, customized parameters, and one or more solution versions (trained models).

To create a solution in Amazon Personalize, you do the following:

1. **Create a solution** – Customize solution parameters and recipe-specific hyperparameters so the model meets your specific business needs. See Creating and configuring a solution (p. 211). For a list of available recipes, see Choosing a recipe (p. 113).
2. **Create a solution version (train a model)** – Train the machine learning model Amazon Personalize will use to generate recommendations for your customers. See Creating a solution version (p. 226).
3. **Evaluate the solution version** – Use the metrics Amazon Personalize generates from the new solution version to evaluate the performance of the model. See [Evaluating a solution version with metrics](p. 232).

**Creating and configuring a solution**

After you finish importing data, you are ready to create a solution. A *solution* refers to the combination of an Amazon Personalize recipe, customized parameters, and one or more solution versions (trained models).

When you create a solution, you can configure it to meet your specific business needs:

- You can configure hyperparameters to optimize the model based on your recipe and business needs. Different recipes use different hyperparameters. For information on configuring hyperparameters, see [Hyperparameters and HPO](p. 220). For the available hyperparameters for your recipe, see page for your recipe in [Choosing a recipe](p. 113).
- You can modify the columns Amazon Personalize considers when training a model (creating a solution version). For more information, see [Configuring columns used when training](p. 214).
- If you use either the **User-Personalization** recipe (p. 116) or **Personalized-Ranking** recipe (p. 139) recipe, you can to optimize your solution for an objective in addition to relevance. For more information see [Optimizing a solution for an additional objective](p. 215).
- If you have event type and event value data, you can use it to choose the interactions records Amazon Personalize considers during training. For more information see [Choosing the interactions data used for training](p. 223).

You can create and configure a solution using the console, AWS Command Line Interface (AWS CLI), or AWS SDK. If you have an existing solution, you can use the Amazon Personalize console to clone the solution. When you clone a solution, you can use the configuration of the existing solution as a starting point, such as the recipe and hyperparameters, and make any changes as necessary. For more information, see [Cloning a solution (console)](p. 225). After you create a solution, you can view the solution's configuration details on the solution's details page of the Amazon Personalize console, or with the [DescribeSolution](p. 526) operation.

**Topics**

- [Creating a solution (console)](p. 211)
- [Creating a solution (AWS CLI)](p. 212)
- [Creating a solution (AWS SDKs)](p. 213)
- [Configuring columns used when training](p. 214)
- [Optimizing a solution for an additional objective](p. 215)
- [Hyperparameters and HPO](p. 220)
- [Choosing the interactions data used for training](p. 223)
- [Cloning a solution (console)](p. 225)

**Creating a solution (console)**

To create a solution in the console, choose the dataset group containing the dataset you'll be using, and then specify a solution name, recipe, and optional recipe specific hyperparameters.

**To configure a solution (console)**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign in to your account.
2. On the **Dataset groups** page, choose your dataset group.
3. On the Overview page, in the middle card, if you created a Domain dataset group, choose Use custom resources and choose Create solution. For a Custom dataset group, choose Create solution.

4. For Solution name, specify a name for your solution.

5. For Solution type, choose either Item recommendation to get item recommendations for your users, or choose User segmentation to get user segments (groups of users) based on your item data.

6. For Recipe, choose a recipe (see Choosing a recipe (p. 113)).

7. In Solution configuration, if your Interactions dataset has EVENT_TYPE or both EVENT_TYPE and EVENT_VALUE columns, optionally use the Event type and Event value threshold fields to choose the interactions data that Amazon Personalize uses when training the model.

   For more information see Choosing the interactions data used for training (p. 223).

8. If you use either the User-Personalization recipe (p. 116) or Personalized-Ranking recipe (p. 139) recipe, optionally specify an Objective and choose an Objective sensitivity to optimize your solution for an objective in addition to relevance. For more information see Optimizing a solution for an additional objective (p. 215).

9. For Tags, optionally add any tags. For more information about tagging Amazon Personalize resources, see Tagging Amazon Personalize resources (p. 370).

10. Choose Next.

11. On the Advanced configuration page, optionally use Columns for training to choose the columns Amazon Personalize considers when training solution versions. By default, Amazon Personalize uses all columns that can be used in training. Only datasets used by your chosen recipe are listed. Columns with the boolean data type and non-categorical string fields aren’t used. You can’t exclude EVENT_TYPE columns.

   You can change the columns to experiment with different combinations of training data, or exclude columns without meaningful data from training. For example, might have a column that you want to use only to filter recommendations. You can exclude this column from training and Amazon Personalize considers it only when filtering.

12. Configure any hyperparameter options based on your recipe and business needs. Different recipes use different hyperparameters. For the available hyperparameters, see the individual recipes in Choosing a recipe (p. 113).

13. Choose Create solution. Amazon Personalize starts creating your first solution version and the Solutions and recipes page appears.

   To monitor its status, choose your solution and view the status in the Solution versions section. When it’s active, you are ready to get recommendations. For more information, see Step 4: Getting recommendations (p. 244).

Creating a solution (AWS CLI)

To create a solution using the AWS CLI, use the following create-solution operation. Specify the solution name, dataset group arn, and recipe arn.

```
aws personalize create-solution \
  --name solution name \
  --dataset-group-arn dataset group arn \
  --recipe-arn recipe arn
```

The solution Amazon Resource Name (ARN) is displayed, for example:

```
{
  "solutionArn": "arn:aws:personalize:<region>:solution/<solution name>"
}
```
Record the solution ARN for future use and proceed to Creating a solution version (AWS CLI) (p. 226).

You can modify the above code to optimize recipe properties and hyperparameters (see Hyperparameters and HPO (p. 220)), choose columns used in training (see Configuring columns used when training (AWS CLI) (p. 215)), or filter the Interactions data used for training (see Choosing the interactions data used for training (p. 223)).

If you use either the User-Personalization recipe (p. 116) or Personalized-Ranking recipe (p. 139) recipe, you can optimize your solution for an objective in addition to relevance. For more information see Optimizing a solution for an additional objective (p. 215).

Creating a solution (AWS SDKs)

The following code shows how to create an Amazon Personalize solution. Give the solution a name, and specify the Amazon Resource Name (ARN) of your dataset group, and the ARN of the recipe to use. For information on recipes, see Choosing a recipe (p. 113).

You can modify the following code to optimize recipe properties and hyperparameters (see Hyperparameters and HPO (p. 220)), configure what columns are used for training (see Configuring columns used when training (AWS SDKs) (p. 215)), or filter the Interactions data used for training (see Choosing the interactions data used for training (p. 223)).

If you use either the User-Personalization recipe (p. 116) or Personalized-Ranking recipe (p. 139) recipe, you can optimize your solution for an objective in addition to relevance. For more information see Optimizing a solution for an additional objective (p. 215).

SDK for Python (Boto3)

```python
import boto3

personalize = boto3.client('personalize')

print('Creating solution')
create_solution_response = personalize.create_solution(
    name='solution name',
    recipeArn='recipe arn',
    datasetGroupArn='dataset group arn'
)
solution_arn = create_solution_response['solutionArn']
print('solution_arn: ', solution_arn)
```

SDK for Java 2.x

```java
public static String createPersonalizeSolution(PersonalizeClient personalizeClient,
                                              String datasetGroupArn,
                                              String solutionName,
                                              String recipeArn) {

    try {
        CreateSolutionRequest solutionRequest = CreateSolutionRequest.builder()
            .name(solutionName)
            .datasetGroupArn(datasetGroupArn)
            .recipeArn(recipeArn)
            .build();

        CreateSolutionResponse solutionResponse = personalizeClient.createSolution(solutionRequest);
        return solutionResponse.solutionArn();
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
```
SDK for JavaScript v3

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateSolutionCommand } from '@aws-sdk/client-personalize';
import { personalizeClient } from './libs/personalizeClients.js';
// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: 'REGION'});

// Set the solution parameters.
export const createSolutionParam = {
  datasetGroupArn: 'DATASET_GROUP_ARN', /* required */
  recipeArn: 'RECIPE_ARN', /* required */
  name: 'NAME' /* required */
}

export const run = async () => {
  try {
    const response = await personalizeClient.send(new CreateSolutionCommand(createSolutionParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();
```

Record the solution ARN for future use and proceed to Creating a solution version (AWS SDKs) (p. 227).

Configuring columns used when training

When you create a solution, you can modify the columns Amazon Personalize considers when creating a solution version (training a model). By default, Amazon Personalize uses all columns that can be used when training. Columns with the boolean data type and custom string fields that aren't categorical or textual aren't used when training. You can't exclude EVENT_TYPE columns.

You can change the columns used when training to control what data Amazon Personalize uses when training a model (creating a solution version). You might do this to experiment with different combinations of training data. Or you might exclude columns without meaningful data. For example, you might have a column that you want to use only to filter recommendations. You can exclude this column from training and Amazon Personalize considers it only when filtering.

If you have already created a solution and you want to modify the columns it uses when training, you can clone the solution. When you clone a solution, you can use the configuration of the existing solution as a starting point, such as the recipe and hyperparameters, and make any changes as necessary. For more information, see Cloning a solution (console) (p. 225).

You can configure the columns Amazon Personalize uses when training with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDK. For information about choosing columns with the Amazon Personalize console, see the advanced configuration steps in Creating a solution (console) (p. 211). After you create a solution, you can view the columns the solution uses on the solution's details page of the Amazon Personalize console, or with the DescribeSolution (p. 526) operation.

Topics

214
Creating custom resources

- Configuring columns used when training (AWS CLI) (p. 215)
- Configuring columns used when training (AWS SDKs) (p. 215)

Configuring columns used when training (AWS CLI)

To exclude columns from training, provide the excludedDatasetColumns object in the trainingDataConfig as part of the solution configuration. For each key, provide the dataset type. For each value, provide the list of columns to exclude. The following code shows how to exclude columns from training when you create a solution with the AWS CLI.

```bash
aws personalize create-solution \
--name solution name \ 
--dataset-group-arn dataset group ARN \ 
--recipe-arn recipe ARN \ 
--solution-config "{"trainingDataConfig": {"excludedDatasetColumns": \
{ "datasetType": [ "column1Name", "column2Name"]}}")"
```

Configuring columns used when training (AWS SDKs)

To exclude columns from training, provide the excludedDatasetColumns object in the trainingDataConfig as part of the solution configuration. For each key, provide the dataset type. For each value, provide the list of columns to exclude. The following code shows how to exclude columns from training when you create a solution with the SDK for Python (Boto3).

```python
import boto3

personalize = boto3.client('personalize')

create_solution_response = personalize.create_solution( 
    name = 'solution name', 
    recipeArn = 'recipe ARN', 
    datasetGroupArn = 'dataset group ARN', 
    solutionConfig = { 
        "trainingDataConfig": { 
            "excludedDatasetColumns": { 
                "datasetType": ["COLUMN_A", "COLUMN_B"] 
            } 
        } 
    } 
)

solution_arn = create_solution_response['solutionArn']

print('solution_arn: ', solution_arn)
```

Optimizing a solution for an additional objective

The primary objective of Amazon Personalize is to predict the most relevant items for your users based on historical and real-time interactions data. These are the items your users will most likely interact with (for example, the items they will most likely click). If you have an additional objective, such as maximizing streaming minutes or increasing revenue, you can create a solution that generates recommendations based on both relevance and your objective.

After you create an Interactions dataset and an Items dataset with a non-null, numerical column metadata attribute, you can create a solution that is optimized for an additional objective based on your item metadata. You can use the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

To optimize a solution for an additional objective, create a new solution with the User-Personalization recipe or Personalized-Ranking recipe and choose the numerical metadata column in your Items dataset.
that is related to your objective. When generating recommendations, Amazon Personalize gives more importance to items with higher values for this column of data. For example, you might choose a VIDEO_LENGTH column to maximize streaming minutes or a PRICE column to maximize revenue.

Objective requirements are as follows:

- You can choose only one column for your objective.
- The column must have a numerical type in your schema.
- The column can’t have a null type in your schema.

For more information about schemas and data types, see Schemas (p. 82).

Topics

- Balancing objective emphasis and relevance (p. 216)
- Measuring optimization performance (p. 216)
- Optimizing a solution (console) (p. 217)
- Optimizing a solution (AWS CLI) (p. 217)
- Optimizing a solution (AWS SDKs) (p. 218)
- Sample Jupyter notebook (p. 220)

Balancing objective emphasis and relevance

There can be a trade-off when recommending items based more on your objective than relevance. For example, if you want to increase revenue through recommendations, recommendations for only expensive items might make items less relevant for your users and decrease user engagement and conversion.

To configure the balance between relevance and your objective, choose one of the following objective sensitivity levels when you create the solution:

- Off: Amazon Personalize uses primarily interactions data to predict the most relevant items for your user.
- Low: Amazon Personalize places less emphasis on your objective. Relevance through interactions data is more important.
- Medium: Amazon Personalize places equal emphasis on your objective and relevance through interactions data.
- High: Amazon Personalize places more emphasis on your objective. Relevance through interactions data is less important.

Measuring optimization performance

When you create a solution version (train a model) for a solution with an optimization objective, Amazon Personalize generates an average_rewards_at_k metric. The score for average_rewards_at_k tells you how well the solution version performs in achieving your objective. To calculate this metric, Amazon Personalize calculates the rewards for each user as follows:

\[
\text{rewards_per_user} = \frac{\text{total rewards from the user's interactions with their top 25 reward generating recommendations}}{\text{total rewards from the user's interactions with recommendations}}
\]

The final average_rewards_at_k is the average of all rewards_per_user normalized to be a decimal value less than or equal to 1 and greater than 0. The closer the value is to 1, the more gains on average per user you can expect from recommendations.
For example, if your objective is to maximize revenue from clicks, Amazon Personalize calculates each user score by dividing total revenue generated by the items the user clicked from their top 25 most expensive recommendations by the revenue from all of the recommended items the user clicked. Amazon Personalize then returns a normalized average of all user scores. The closer the average_rewards_at_k is to 1, the more revenue on average you can expect to gain per user from recommendations.

For more information about generating metrics, see Evaluating a solution version with metrics (p. 232).

Optimizing a solution (console)

To optimize a solution for an additional objective with the Amazon Personalize console, create a new solution and choose the column of numerical item metadata that is related to your objective.

To optimize a solution for an additional objective (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. Choose the dataset group that you want to use for training.
3. In the Create solutions section of the dashboard, choose the Start button. If you have already created a solution, choose the Create solution button.
4. For Solution name, specify a name for the solution.
5. For Recipe, choose either User-Personalization recipe (p. 116) or Personalized-Ranking recipe (p. 139).
6. In Solution configuration, if your Interactions dataset has EVENT_TYPE or both EVENT_TYPE and EVENT_VALUE columns, optionally use the Event type and Event value threshold fields to choose the interactions data that Amazon Personalize uses when training the model.

   For more information see Choosing the interactions data used for training (p. 223).
7. For Objective, choose the numerical column from the Items dataset that is related to your objective. You can choose only a numerical metadata column.
8. For Objective sensitivity, choose the level of emphasis Amazon Personalize places on your objective when generating recommendations. The objective sensitivity configures how Amazon Personalize balances recommending items based on your objective versus relevance through interactions data. For more information, see Balancing objective emphasis and relevance (p. 216).
9. Configure any hyperparameter options based on your recipe and business needs. Different recipes use different hyperparameters. For available hyperparameters, see the documentation for individual recipes in Choosing a recipe (p. 113).
10. Choose Next.

Optimizing a solution (AWS CLI)

You can optimize for an objective only with the User-Personalization or Personalized-Ranking recipe. To optimize a solution for an additional objective using the AWS CLI, create a new solution and specify your objective details using the optimizationObjective key in the solutionConfig object. The optimizationObjective has the following fields:

- itemAttribute: Specify the name of the numerical metadata column from the Items dataset that relates to your objective.
- objectiveSensitivity: Specify the level of emphasis that the solution places on your objective when generating recommendations. The objective sensitivity level configures how Amazon Personalize
balances recommending items based on your objective versus relevance through interactions data. The objectiveSensitivity can be OFF, LOW, MEDIUM or HIGH. For more information, see Balancing objective emphasis and relevance (p. 216).

The following is an example of the create-solution AWS CLI command. Replace the solution name, dataset group arn, and recipe arn values with your own.

For optimizationObjective, replace COLUMN_NAME with the numerical metadata column name from the Items dataset that is related to your objective. For objectiveSensitivity, specify OFF, LOW, MEDIUM, or HIGH.

```
aws personalize create-solution \
  --name solution name \
  --dataset-group-arn dataset group arn \
  --recipe-arn recipe arn \
  --solution-config "{"optimizationObjective":{"itemAttribute":"COLUMN_NAME", \n  "objectiveSensitivity":"MEDIUM"}}"
```

When your solution is ready, create a new solution version (for an example command see Creating a solution (AWS CLI) (p. 212)). Once you create a solution version, you can view the optimization performance with the solution version metrics. See Measuring optimization performance (p. 216).

Optimizing a solution (AWS SDKs)

You can optimize for an objective only with the User-Personalization or Personalized-Ranking recipe.

To optimize a solution for an additional objective using the AWS SDKs, create a new solution and specify your objective details using the optimizationObjective key in the solutionConfig object for the solution. The optimizationObjective has the following fields:

- **itemAttribute**: Specify the name of the numerical metadata column from the dataset group's Items dataset that relates to your objective.
- **objectiveSensitivity**: Specify the level of emphasis that the solution places on your objective when generating recommendations. The objective sensitivity level configures how Amazon Personalize balances recommending items based on your objective versus relevance through interactions data. The objectiveSensitivity can be OFF, LOW, MEDIUM or HIGH. For more information, see Balancing objective emphasis and relevance (p. 216).

Use the following code to create a solution with an additional objective with the AWS SDK for Python (Boto3) or the AWS SDK for Java 2.x.

When your solution is ready, create a new solution version (for example code see Creating a solution version (AWS SDKs) (p. 227)). Once you create a solution version, you can view the optimization performance with the solution version metrics. See Measuring optimization performance (p. 216).

SDK for Python (Boto3)

To create a solution that is optimized for an additional objective, use the following create_solution method. Replace the solution name, dataset group arn, and recipe arn values with your own.

For optimizationObjective, replace COLUMN_NAME with the numerical metadata column name from the Items dataset that is related to your objective. For objectiveSensitivity, specify OFF, LOW, MEDIUM, or HIGH.

```
import boto3

personalize = boto3.client('personalize')
```
create_solution_response = personalize.create_solution(
    name= 'solution name',
    recipeArn = 'recipe arn',
    datasetGroupArn = 'dataset group arn',
    solutionConfig = {
        "optimizationObjective": {
            "itemAttribute": "COLUMN_NAME",
            "objectiveSensitivity": "MEDIUM"
        }
    }
)
solution_arn = create_solution_response['solutionArn']
print('solution_arn: ', solution_arn)

SDK for Java 2.x
To create a solution that is optimized for an additional objective, use the following createPersonalizeSolution method and pass the following as parameters: an Amazon Personalize service client, the dataset group's Amazon Resource Name (ARN), a solution name, the recipe ARN, the item attribute, and the objective sensitivity level.

    public static String createPersonalizeSolution(PersonalizeClient personalizeClient,
        String datasetGroupArn,
        String solutionName,
        String recipeArn,
        String itemAttribute,
        String objectiveSensitivity) {
        try {
            OptimizationObjective optimizationObjective = OptimizationObjective.builder()
                    .itemAttribute(itemAttribute)
                    .objectiveSensitivity(objectiveSensitivity)
                    .build();

            SolutionConfig solutionConfig = SolutionConfig.builder()
                    .optimizationObjective(optimizationObjective)
                    .build();

            CreateSolutionRequest solutionRequest = CreateSolutionRequest.builder()
                    .name(solutionName)
                    .datasetGroupArn(datasetGroupArn)
                    .recipeArn(recipeArn)
                    .solutionConfig(solutionConfig)
                    .build();

            CreateSolutionResponse solutionResponse = personalizeClient.createSolution(solutionRequest);
            return solutionResponse.solutionArn();
        } catch (PersonalizeException e) {
            System.err.println(e.awsErrorDetails().errorMessage());
            System.exit(1);
        }
        return "";
    }

SDK for JavaScript v3

    import { CreateSolutionCommand, personalizeClient } from "@aws-sdk/client-personalize";
// create the personalizeClient
const personalizeClient = new PersonalizeClient({ region: "REGION"});

// set the solution parameters.
export const createSolutionParam = {
  datasetGroupArn: 'DATASET_GROUP_ARN', /* required */
  recipeArn: 'RECIPE_ARIN', /* required */
  name: 'NAME', /* required */
  solutionConfig: {
    optimizationObjective: {
      itemAttribute: "COLUMN_NAME", /* specify the numerical column from the
      Items dataset related to your objective */
      objectiveSensitivity: "MEDIUM" /* specify OFF, LOW, MEDIUM, or HIGH */
    }
  }
};

export const run = async () => {
  try {
    const response = await personalizeClient.send(new
    CreateSolutionCommand(createSolutionParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};

run();

Sample Jupyter notebook

For a sample Jupyter notebook that shows how to create a solution that is optimized for an additional
objective based item metadata, see the objective_optimization folder of the Amazon Personalize
samples GitHub repository

Hyperparameters and HPO

You specify hyperparameters before training to optimize the trained model for your particular use case.
This contrasts with model parameters whose values are determined during the training process.

Hyperparameters are specified using the algorithmHyperParameters key that is part of the
SolutionConfig (p. 713) object that is passed to the CreateSolution (p. 468) operation.

A condensed version of the CreateSolution request is below. The example includes the
solutionConfig object. You use solutionConfig to override the default parameters of a recipe.

```json
{
  "name": "string",
  "recipeArn": "string",
  "eventType": "string",
  "solutionConfig": {
    "optimizationObjective": {
      "itemAttribute": "string",
      "objectiveSensitivity": "string"
    },
    "eventValueThreshold": "string",
    "featureTransformationParameters": {
      "string": "string"
    },
    "algorithmHyperParameters": {
      "string": "string"
    }
  }
}
```
Different recipes use different hyperparameters. For the available hyperparameters, see the individual recipes in [Choosing a recipe (p. 113)](https://www.amazon.com/AWS-Developer-Guide/dp/0137012927).

### Enabling hyperparameter optimization

Hyperparameter optimization (HPO), or tuning, is the task of choosing optimal hyperparameters for a specific learning objective. The optimal hyperparameters are determined by running many training jobs using different values from the specified ranges of possibilities. By default, Amazon Personalize does not perform HPO. To use HPO, set `performHPO` to `true`, and include the `hpoConfig` object.

Hyperparameters can be categorical, continuous, or integer-valued. The `hpoConfig` object has keys that correspond to each of these types, where you specify the hyperparameters and their ranges. You must provide each type in your request, but if a recipe doesn’t have a parameter of a type, you can leave it empty. For example, User-Personalization does not have a tunable hyperparameter of continuous type. So for the `continuousHyperParameterRange`, you would pass an empty array.

The following code shows how to create a solution with HPO enabled using the SDK for Python (Boto3). The solution in the example uses the User-Personalization recipe (p. 116) and has HPO set to true. The code provides a value for `hidden_dimension` and the categoricalHyperParameterRanges and integerHyperParameterRanges. The continuousHyperParameterRange is empty and the hpoResourceConfig sets the maxNumberOfTrainingJobs and maxParallelTrainingJobs.

```python
create_solution_response = personalize.create_solution(
    name = solutionName,
    datasetGroupArn = 'arn:aws:personalize:region:accountId:dataset-group/datasetGroupName',
    recipeArn = 'arn:aws:personalize::recipe/aws-user-personalization',
    performHPO = True,
    solutionConfig = {
        "algorithmHyperParameters": {
            "hidden_dimension": "55"
        },
        "hpoConfig": {
            "algorithmHyperParameterRanges": {
                "categoricalHyperParameterRanges": [
                    {"name": "recency_mask", "values": ["true", "false"]}
                ],
                "integerHyperParameterRanges": [
                    {"name": "bptt", "minValue": 2, "maxValue": 22}
                ],
                "continuousHyperParameterRanges": []
            }
        }
    }
)
```
For more information about HPO, see [Automatic model tuning](#).

### Viewing hyperparameters

You can view the hyperparameters of the solution by calling the `DescribeSolution` operation. The following sample shows a `DescribeSolution` output. After creating a solution version (training a model), you can also view hyperparameters with the `DescribeSolutionVersion` operation.

```json
{
  "solution": {
    "name": "hpo_config_solution",
    "solutionArn": "arn:aws:personalize:region:accountId:solution/solutionName",
    "performHPO": true,
    "performAutoML": false,
    "recipeArn": "arn:aws:personalize:::recipe/aws-user-personalization",
    "datasetGroupArn": "arn:aws:personalize:region:accountId:dataset-group/datasetGroupName",
    "eventType": "click",
    "solutionConfig": {
      "hpoConfig": {
        "maxNumberOfTrainingJobs": "4",
        "maxParallelTrainingJobs": "2"
      },
      "algorithmHyperParameterRanges": {
        "integerHyperParameterRanges": [{
          "name": "training.bptt",
          "minValue": 2,
          "maxValue": 22
        }]
      },
      "categoricalHyperParameterRanges": []
    },
    "status": "ACTIVE",
    "creationDateTime": "2022-07-08T12:12:48.565000-07:00",
    "lastUpdatedDateTime": "2022-07-08T12:12:48.565000-07:00"
  }
}
```
Choosing the interactions data used for training

You can choose the events in an Interactions dataset that Amazon Personalize uses when creating a solution version (training a model). Choosing interactions data before training allows you to use only a relevant subset of your data for training or remove noise to train a more optimized model. For more information about Interactions datasets, see Schemas (p. 82) and Interactions datasets (p. 77).

You can choose interactions data as follows:

- **Choose records based on type** – When you configure a solution, if your Interactions dataset includes event types in an EVENT_TYPE column, you can optionally specify an event type to use in training. For example, if your Interactions dataset includes purchase, click, and watch event types, and you want Amazon Personalize to train the model with only watch events, when you configure your solution, you would provide watch as the event type that Amazon Personalize uses in training.

  If your Interactions dataset has multiple event types in an EVENT_TYPE column, and you do not provide an event type when you configure your solution, Amazon Personalize uses all interactions data for training with equal weight regardless of type.

- **Choose records based on type and value** – When you configure a solution, if your Interactions dataset includes EVENT_TYPE and EVENT_VALUE fields, you can set a specific value as a threshold to exclude records from training. For example, if your EVENT_VALUE data for events with an EVENT_TYPE of watch is the percentage of a video that a user watched, if you set the event value threshold to 0.5, and the event type to watch, Amazon Personalize trains the model using only watch interaction events with an EVENT_VALUE greater than or equal to 0.5.

Filtering records by event value and event type (AWS SDK)

In the following procedure, you use the AWS SDK for Python (Boto3) to create an Interaction schema that filters a training dataset. You can use a Jupyter (iPython) notebook to accomplish the same task. For more information, see Getting started using Amazon Personalize APIs with Jupyter (iPython) notebooks (p. 66).

**Prerequisites:** Complete the prerequisites and verify that your Python environment is set up as described in Getting started (SDK for Python (Boto3)) (p. 62).

**To filter records used in a training dataset by event value or event type**

1. Create an Interaction schema and include the EVENT_TYPE and EVENT_VALUE fields using "name" and "type" key-value pairs as shown.

```python
import boto3
import json

personalize = boto3.client('personalize')

# Create a name for your schema
schema_name = 'YourSchemaName'

# Define the schema for your dataset
schema = {
    "type": "record",
    "name": "Interactions",
    "namespace": "com.amazonaws.personalize.schema",
    "fields": [
        {
            "name": "USER_ID",
            "type": "string"
        }]
}````
1. Create the schema for Amazon Personalize:

```json
{
  "version": "1.0",
  "schema": [
    {
      "name": "ITEM_ID",
      "type": "string"
    },
    {
      "name": "EVENT_VALUE",
      "type": "float"
    },
    {
      "name": "EVENT_TYPE",
      "type": "string"
    },
    {
      "name": "TIMESTAMP",
      "type": "long"
    }
  ]
}
```

# Create the schema for Amazon Personalize
create_schema_response = personalize.create_schema(
    name = schema_name,
    schema = json.dumps(schema)
)

# To get the schema ARN, use the following lines
schema_arn = create_schema_response['schemaArn']
print('Schema ARN:' + schema_arn)

2. Format your input data to match your schema. For a code sample, see Data format guidelines (p. 101).

3. Upload your data to an Amazon Simple Storage Service (Amazon S3) bucket. For a code sample, see Uploading to an Amazon S3 bucket (p. 171).

4. Import your data into Amazon Personalize with the CreateDatasetImportJob (p. 448) API. Be sure to record your dataset group Amazon Resource Name (ARN) because you will need it when you create the solution. For a code sample, see Importing bulk records (AWS SDKs) (p. 180).

5. Get the ARN of the recipe that you want to use when you create your solution. You'll need it when you create the solution.

```python
# Display the ARNs of the recipes
recipe_list = personalize.list_recipes()
for recipe in recipe_list['recipes']:
    print(recipe['recipeArn'])

# Store the ARN of the recipe that you want to use
recipe_arn = "arn:aws:personalize:::recipe/aws-recipe-name"
```

6. Call the CreateSolution (p. 468) API. If you want to specify the event type, for example "purchase", set it in the eventType parameter. If you want to specify an event value, for example 10, set it in the eventValueThreshold parameter. You can also specify both an event type and an event value.

```python
# Create the solution
create_solution_response = personalize.create_solution(
    name = "your-solution-name",
    datasetGroupArn = dataset_group_arn,
    recipeArn = recipe_arn,
    eventType = 'watched',
    solutionConfig = {
        "eventValueThreshold": "0.5"
    }
)
```
# Store the solution ARN
solution_arn = create_solution_response['solutionArn']

# Use the solution ARN to get the solution status
solution_description = personalize.describe_solution(solutionArn = solution_arn)["solution"]
print('Solution status: ' + solution_description['status'])

7. When you have the solution, use it to train a model by specifying its solution ARN in a `CreateSolutionVersion` request.

# Create a solution version
create_solution_version_response = personalize.create_solution_version(solutionArn = solution_arn)

# Store the solution version ARN
solution_version_arn = create_solution_version_response['solutionVersionArn']

# Use the solution version ARN to get the solution version status.
solution_version_description = personalize.describe_solution_version(solutionVersionArn = solution_version_arn)['solutionVersion']
print('Solution version status: ' + solution_version_description['status'])

Training is complete when the status is ACTIVE. For more information, see [Creating a solution and a solution version](p. 210).

After you train a model, you should evaluate its performance. To optimize your model, you might want to adjust the `eventValueThreshold` or other hyperparameters. For more information, see [Evaluating a solution version with metrics](p. 232).

Cloning a solution (console)

When you create a new solution, you can use the Amazon Personalize console to clone a solution. When you clone a solution, you can use the configuration of the existing solution as a starting point, such as the recipe and hyperparameters, and make any changes as necessary. This is useful if you want to make one change to a solution, but leave all other properties unchanged. For example, adding a new column of training data to your dataset. In this case, you would clone a solution, give the solution a name, change the columns used when training, and leave all other properties unchanged.

Cloning a solution

To clone a solution, you choose the existing solution, and choose the **Clone solution** option. Then give the new solution a name, and modify the relevant fields.

**To clone a solution**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign in to your account.
2. On the **Dataset groups** page, choose your dataset group.
3. Choose **Custom resources** and choose **Solutions**.
4. Choose the solution that you want to clone.
5. Choose **Actions**, and choose **Clone solution**.
6. Give the new solution a name.
7. Make any changes to the solution details and advanced configuration. Amazon Personalize pre-populates these fields with values from the existing solution. For information about each field, see [Creating and configuring a solution](p. 211).
Creating a solution version

After you have completed Creating and configuring a solution (p. 211), you are ready to create a solution version. A solution version refers to a trained machine learning model. A solution version is a custom resource. You can deploy a solution version with an Amazon Personalize campaign. You use the campaign to get recommendations for users. And you can use a solution version to get batch recommendations.

You can create a solution version using the console, AWS Command Line Interface (AWS CLI), or AWS SDKs. If your solution version has a status of CREATE_PENDING or CREATE_IN_PROGRESS, you can use the section called “StopSolutionVersionCreation” (p. 582) operation to stop the solution version creation process. See Stopping the creation of a solution version (p. 230).

Topics
- Creating a solution version (console) (p. 226)
- Creating a solution version (AWS CLI) (p. 226)
- Creating a solution version (AWS SDKs) (p. 227)
- Stopping the creation of a solution version (p. 230)

Creating a solution version (console)

When you initially create your solution with the Amazon Personalize console, you also create a solution version. On the solution details page, you can track training progress in the Solution versions section. When training is complete, the status is Active and you are ready to deploy a campaign and get recommendations. See Creating a campaign (p. 237).

If you want to create an additional solution version for an existing solution, create a new solution version from the solution overview page as follows.

To create a new solution version

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Navigate to the dataset groups page and choose the dataset group with your new solution.
3. In the navigation pane, choose Solutions and recipes.
4. On the Solution and recipes page, choose the solution you want to create a solution version for.
5. On the solution overview page, choose Create solution version to start training a new model.

On the solution details page, you can track training progress in the Solution versions section. When training is complete, the status is Active you can evaluate it using metrics supplied by Amazon Personalize. For more information, see Evaluating a solution version with metrics (p. 232).

If training does not complete because of an error, you are not charged for the training. If your solution version has a status of CREATE_PENDING or CREATE_IN_PROGRESS, you can stop the solution version creation process. To stop solution version creation, navigate to the solution version details page and choose Stop. See Stopping the creation of a solution version (p. 230).

Creating a solution version (AWS CLI)

When your solution is ACTIVE, train the model by running the following command. Replace solution arn with the solution Amazon Resource Name (ARN) from Creating and configuring a solution (p. 211).

```bash
aws personalize create-solution-version
  --solution-arn solution arn
```
The solution version ARN is displayed, for example:

```json
{
}
```

Check the training status of the solution version by using the `describe-solution-version` command. Provide the solution version ARN that was returned in the previous step. For more information about the API, see `DescribeSolutionVersion` (p. 529).

```bash
aws personalize describe-solution-version \
--solution-version-arn solution version arn
```

The properties of the solution version and the training status are displayed. Initially, the status shows as CREATE PENDING, for example:

```json
{
   "solutionVersion": {
       ...,
       "status": "CREATE PENDING"
   }
}
```

Training is complete when the status is ACTIVE and you can evaluate it using metrics supplied by Amazon Personalize. For more information, see `Evaluating a solution version with metrics` (p. 232). If training does not complete because of an error, you are not charged for the training.

If your solution version has a status of CREATE_PENDING or CREATE_IN_PROGRESS, you can use the `StopSolutionVersionCreation` (p. 582) operation to stop the solution version creation process. See `Stopping the creation of a solution version` (p. 230).

### Creating a solution version (AWS SDKs)

When your solution is ACTIVE, use the following code to create a solution version. Specify the Amazon Resource Name (ARN) from `Creating and configuring a solution` (p. 211). Use the `DescribeSolutionVersion` (p. 529) operation to retrieve the solution version's status.

**SDK for Python (Boto3)**

```python
import boto3

personalize = boto3.client('personalize')

# Store the solution ARN
solution_arn = 'solution arn'

# Use the solution ARN to get the solution status.
solution_description = personalize.describe_solution(solutionArn = 'solution_arn')['solution']
print('Solution status: ' + solution_description['status'])

# Use the solution ARN to create a solution version.
response = personalize.create_solution_version(solutionArn = solution_arn)
solution_version_arn = response['solutionVersionArn']
print('Solution version ARN: ' + solution_version_arn)

# Use the solution version ARN to get the solution version status.
solution_version_description = personalize.describe_solution_version(softmax
solutionVersionArn = solution_version_arn['solutionVersion']
print('Solution version status: ' + solution_version_description['status'])

SDK for Java 2.x

public static String createPersonalizeSolutionVersion(PersonalizeClient personalizeClient, String solutionArn) {
    long maxTime = 0;
    long waitInMilliseconds = 30 * 1000; // 30 seconds
    String solutionStatus = "";
    String solutionVersionStatus = "";
    String solutionVersionArn = "";

    try {
        DescribeSolutionRequest describeSolutionRequest =
            DescribeSolutionRequest.builder()
                .solutionArn(solutionArn)
                .build();

        maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;

        // Wait until solution is active.
        while (Instant.now().getEpochSecond() < maxTime) {
            solutionStatus =
                personalizeClient.describeSolution(describeSolutionRequest).solution().status();
            System.out.println("Solution status: " + solutionStatus);
            if (solutionStatus.equals("ACTIVE") || solutionStatus.equals("CREATE FAILED")) {
                break;
            }
            try {
                Thread.sleep(waitInMilliseconds);
            } catch (InterruptedException e) {
                System.out.println(e.getMessage());
            }
        }

        // Once the solution is active, start creating a solution version.
        if (solutionStatus.equals("ACTIVE")) {
            CreateSolutionVersionRequest createSolutionVersionRequest =
                CreateSolutionVersionRequest.builder()
                    .solutionArn(solutionArn)
                    .build();

            CreateSolutionVersionResponse createSolutionVersionResponse =
                personalizeClient.createSolutionVersion(createSolutionVersionRequest);
            solutionVersionArn =
                createSolutionVersionResponse.solutionVersionArn();
            System.out.println("Solution version ARN: " + solutionVersionArn);

            DescribeSolutionVersionRequest describeSolutionVersionRequest =
                DescribeSolutionVersionRequest.builder()
                    .solutionVersionArn(solutionVersionArn)
                    .build();

            maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
            while (Instant.now().getEpochSecond() < maxTime) {
                // Use the solution version ARN to get the solution version status.
            }
        }
    } catch (Exception e) {
        System.out.println(e.getMessage());
    }
}

// Use the created solution version ARN to get the solution version status.
To check the current solution version status, call the `DescribeSolutionVersion` operation and pass the ARN of the solution version returned from the `CreateSolutionVersion` operation. Training is complete when the status is `ACTIVE` and you can evaluate it using metrics supplied by Amazon Personalize. For more information, see Evaluating a solution version with metrics (p. 232). If training does not complete because of an error, you are not charged for the training.

If your solution version has a status of `CREATE_PENDING` or `CREATE_IN_PROGRESS`, you can use the `StopSolutionVersionCreation` operation to stop the solution version creation process. See Stopping the creation of a solution version (p. 230).
Stopping the creation of a solution version

If your solution version has a status of CREATE_PENDING or CREATE_IN_PROGRESS, you can use the Amazon Personalize console or the StopSolutionVersionCreation (p. 582) operation to stop creating the solution version (stop training a model). You can't resume creating a solution version after it has stopped. You are billed for resources used up to the point when the creation of the solution version stopped.

Stopping the creation of a solution version ends model training, but doesn't delete the solution version. You can still view the solution version details in the Amazon Personalize console and with the DescribeSolutionVersion (p. 529) operation.

You can stop the solution version creation process with the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), or the AWS SDKs.

Topics
- Stopping the creation of a solution version (console) (p. 230)
- Stopping the creation of a solution version (AWS CLI) (p. 230)
- Stopping the creation of a solution version (AWS SDKs) (p. 231)

Stopping the creation of a solution version (console)

If your solution version has a status of CREATE_PENDING or CREATE_IN_PROGRESS, you can stop creating a solution version (stop training a model).

To stop creating a solution version (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. On the Dataset groups page, choose the dataset group with the solution version that you want to stop.
3. In the navigation pane, choose Solutions and recipes.
4. On the Solution and recipes page, choose the solution with the solution version that you want to stop.
5. In Solution versions, choose the solution version that you want to stop.
6. On the solution version details page, choose Stop creation. Depending on the original state of the solution version, the solution version state changes as follows:
   - CREATE_PENDING changes to CREATE_STOPPED.
   - CREATE_IN_PROGRESS changes to CREATE_STOPPING and then CREATE_STOPPED.

Stopping the creation of a solution version (AWS CLI)

If your solution version has a status of CREATE_PENDING or CREATE_IN_PROGRESS, you can stop creating a solution version (stop training a model). Use the following stop-solution-version-creation command to stop creating the solution version with the AWS CLI. Replace solution version arn with the Amazon Resource Name (ARN) of the solution version that you want to stop. You are billed for resources used up to the point that creation of the solution version stopped.

```
aws personalize stop-solution-version-creation \
   --solution-version-arn solution version arn
```

Check the training status of the solution version with the describe-solution-version command.

```
aws personalize describe-solution-version \
```

230
Creating custom resources

---solution-version-arn \textit{solution version arn}

Depending on the original state of the solution version, the solution version state changes as follows:

- CREATE\_PENDING changes to CREATE\_STOPPED.
- CREATE\_IN\_PROGRESS changes to CREATE\_STOPPING and then CREATE\_STOPPED

**Stopping the creation of a solution version (AWS SDKs)**

If your solution version has a status of CREATE\_PENDING or CREATE\_IN\_PROGRESS, you can stop creating a solution version (stop training a model). The following code shows how to stop creating a solution version with the AWS SDK for Python (Boto3) or AWS SDK for Java 2.x. You are billed for resources used up to the point when creation of the solution version stopped.

**SDK for Python (Boto3)**

Use the following `stop_solution_version_creation` method to stop creation of a solution version. Replace `solution_version_arn` with the Amazon Resource Name (ARN) of the solution version that you want to stop. The method uses the `DescribeSolutionVersion` (p. 529) operation to retrieve the solution version's status.

```python
import boto3
personalize = boto3.client('personalize')
response = personalize.stop_solution_version_creation(
    solutionVersionArn = \textit{solution_version_arn}
)

# Use the solution version ARN to get the solution version status.
# solution_version_description = personalize.describe_solution_version(
#    solutionVersionArn = \textit{solution_version_arn})['solutionVersion']
# print('Solution version status: ' + solution_version_description['status'])
```

**SDK for Java 2.x**

Use the following `stopSolutionVersionCreation` method to stop creating a solution version. Pass as parameters an Amazon Personalize service client and the Amazon Resource Name (ARN) of the solution version that you want to stop creating. The following code uses the `DescribeSolutionVersion` (p. 529) operation to retrieve the solution version's status.

```java
public static void stopSolutionVersionCreation(PersonalizeClient personalizeClient, String solutionVersionArn) {
    String solutionVersionStatus = "";

    StopSolutionVersionCreationRequest stopSolutionVersionCreationRequest = StopSolutionVersionCreationRequest.builder()
        .solutionVersionArn(solutionVersionArn)
        .build();

    personalizeClient.stopSolutionVersionCreation(stopSolutionVersionCreationRequest);

    // Use the solution version ARN to get the solution version status.
    DescribeSolutionVersionRequest describeSolutionVersionRequest = DescribeSolutionVersionRequest.builder()
        .solutionVersionArn(solutionVersionArn)
        .build();
```
solutionVersionStatus = personalizeClient.describeSolutionVersion(describeSolutionVersionRequest) .solutionVersion() .status();
System.out.println("Solution version status: " + solutionVersionStatus);
}

Depending on the original state of the solution version, the solution version state changes as follows:

- CREATE_PENDING changes to CREATE_STOPPED.
- CREATE_IN_PROGRESS changes to CREATE_STOPPING and then CREATE_STOPPED.

Evaluating a solution version with metrics

You can evaluate the performance of your solution version through offline and online metrics. Online metrics are the empirical results you observe in your users' interactions with real-time recommendations. For example, you might record your users' click-through rate as they browse your catalog. You are responsible for generating and recording any online metrics.

Offline metrics are the metrics Amazon Personalize generates when you train a solution version. You can use offline metrics to evaluate the performance of the model before you create a campaign and provide recommendations. Offline metrics allow you to view the effects of modifying a solution's hyperparameters or compare results from models trained with the same data. For the rest of this section, the term metrics refers to offline metrics.

To get performance metrics, Amazon Personalize splits the input interactions data into a training set and a testing set. The split depends on the type of recipe you choose:

- For USER_SEGMENTATION recipes, the training set consists of 80% of each user's interactions data and the testing set consists of 20% of each user's interactions data.
- For all other recipe types, the training set consists of 90% of your users and their interactions data. The testing set consists of the remaining 10% of users and their interactions data.

Amazon Personalize then creates the solution version using the training set. After training completes, Amazon Personalize gives the new solution version the oldest 90% of each user's data from the testing set as input. Amazon Personalize then calculates metrics by comparing the recommendations the solution version generates to the actual interactions in the newest 10% of each user's data from the testing set.

To generate a baseline for comparison purposes, we recommend using the Popularity-Count (p. 138) recipe, which recommends the top K most popular items.

Topics
- Retrieving solution version metrics (p. 232)
- Metric definitions (p. 234)
- Example (p. 237)
- Additional resources (p. 237)

Retrieving solution version metrics

After you create a solution version, you can use metrics to evaluate its performance. You can retrieve metrics for a solution version with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), and AWS SDKs.
Topics

- Retrieving solution version metrics (console) (p. 233)
- Retrieving solution version metrics (AWS CLI) (p. 233)
- Retrieving solution version metrics (AWS SDKs) (p. 233)

Retrieving solution version metrics (console)

To view recommender metrics in the console, you navigate to the details page for your solution version.

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your Custom dataset group.
3. From the navigation pane, choose Custom resources and then choose Solutions and recipes.
4. Choose your solution.
5. In Solution versions, choose your solution version to view its details page. The metrics are listed on the Solution version metrics tab in the bottom pane. For definitions of metrics, see Metric definitions (p. 234).

Now that you have evaluated your solution version, you can create a campaign by deploying the solution version with the best metrics for your use case. For more information about deploying a solution, see Creating a campaign (p. 237).

Retrieving solution version metrics (AWS CLI)

You retrieve the metrics for a specific solution version by calling the GetSolutionMetrics (p. 532) operation. The following code shows how to retrieve metrics with the AWS CLI.

```
personalize get-solution-metrics --solution-version-arn solution version ARN
```

The following is an example the output from a solution version created using the User-Personalization (p. 116) recipe with an additional optimization objective.

```
{
    "metrics": {
        "coverage": 0.27,
        "mean_reciprocal_rank_at_25": 0.0379,
        "normalized_discounted_cumulative_gain_at_5": 0.0405,
        "normalized_discounted_cumulative_gain_at_10": 0.0513,
        "normalized_discounted_cumulative_gain_at_25": 0.0828,
        "precision_at_5": 0.0136,
        "precision_at_10": 0.0102,
        "precision_at_25": 0.0091,
        "average_rewards_at_k": 0.653
    }
}
```

For explanations of each metric, see Metric definitions (p. 234). Now that you have evaluated your solution version, you can create a campaign by deploying the solution version with the best metrics for your use case. For more information about deploying a solution, see Creating a campaign (p. 237).

Retrieving solution version metrics (AWS SDKs)

You retrieve the metrics for a specific solution version by calling the GetSolutionMetrics (p. 532) operation. Use the following code to retrieve metrics.
SDK for Python (Boto3)

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.get_solution_metrics(
    solutionVersionArn = 'solution version arn'
)

print(response['metrics'])
```

SDK for Java 2.x

```java
public static void getSolutionVersionMetrics(PersonalizeClient personalizeClient,
                                            String solutionVersionArn) {
    try {
        GetSolutionMetricsRequest request = GetSolutionMetricsRequest.builder()
            .solutionVersionArn(solutionVersionArn)
            .build();

        Map<String, Double> metrics = personalizeClient.getSolutionMetrics(request).metrics();
        metrics.forEach((key, value) -> System.out.println(key + " " + value));
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
}
```

The following is an example the output from a solution version created using the User-Personalization recipe with an additional optimization objective.

```json
{
    "metrics": {
        "coverage": 0.27,
        "mean_reciprocal_rank_at_25": 0.0379,
        "normalized_discounted_cumulative_gain_at_5": 0.0405,
        "normalized_discounted_cumulative_gain_at_10": 0.0513,
        "normalized_discounted_cumulative_gain_at_25": 0.0828,
        "precision_at_5": 0.0136,
        "precision_at_10": 0.0102,
        "precision_at_25": 0.0091,
        "average_rewards_at_k": 0.653
    }
}
```

For explanations of each metric, see Metric definitions (p. 234). Now that you have evaluated your solution version, you can create a campaign by deploying the solution version with the best metrics for your use case. For more information about deploying a solution, see Creating a campaign (p. 237).

Metric definitions

The metrics Amazon Personalize generates for solution versions are described below using the following terms:

- **Relevant recommendation** is a recommendation for an item that the user actually interacted with. These items are from the newest 10% of each user’s interactions data from the testing set.
• *Rank refers* to the position of a recommended item in the list of recommendations. Position 1 (the top of the list) is presumed to be the most relevant to the user.

For each metric, higher numbers (closer to 1) are better. To dive deeper, see the resources listed in Additional resources (p. 237).

**coverage**

The value for *coverage* tells you the proportion of unique items that Amazon Personalize might recommend out of the total number of unique items in Interactions and Items datasets. A higher coverage score means Amazon Personalize recommends more of your items, rather than the same few items repeatedly for different users. Recipes that feature item exploration, such as User-Personalization, have higher coverage than those that don't, such as popularity-count.

**mean reciprocal rank at 25**

This metric tells you about a model's ability to generate a relevant recommendation at the top ranked position. You might choose a model with a high mean reciprocal rank at 25 if you are generating relevant search results for a user, and don't expect the user to choose an item lower on the list. For example, users frequently choose the first cooking recipe in search results.

Amazon Personalize calculates this metric using the average reciprocal rank score for requests for recommendations. Each reciprocal rank score is calculated as follows: $1 / \text{the rank of the highest item interacted with by the user}$, where the total possible rankings is 25. Other lower ranked items the user interacts with are ignored. If the user chose the first item, the score is 1. If they don't choose any items, the score is 0.

For example, you might show three different users 25 recommendations each:

- If User 1 clicks the item at rank 4 and the item at rank 10, their reciprocal rank score is $1/4$.
- If User 2 clicks an item at rank 2, an item at rank 4, and an item at rank 12, their reciprocal rank score is $1/2$.
- If User 3 clicks on a single item at rank 6, their reciprocal rank score is $1/6$.

The mean reciprocal rank over all requests for recommendations (in this case 3) is calculated as $(1/4 + 1/2 + 1/6) / 3 = .3056$.

**normalized discounted cumulative gain (NDCG) at K (5/10/25)**

This metric tells you about how well your model ranks recommendations, where K is a sample size of 5, 10, or 25 recommendations. This metric is useful if you are most interested in the ranking of recommendations beyond just the highest ranked item (for this, see mean reciprocal rank at 25). For example, the score for NDCG at 10 would be useful if you have an application that shows up to 10 movies in a carousel at a time.

Amazon Personalize calculates the NDCG by assigning weight to recommendations based on their ranking position for each user in the testing set. Each recommendation is discounted (given a lower weight) by a factor dependent on its position. The final metric is the average of all users in the testing set. The normalized discounted cumulative gain at K assumes that recommendations that are lower on a list are less relevant than recommendations higher on the list.

Amazon Personalize uses a weighting factor of $1/\log(1 + \text{position})$, where the top of the list is position 1.

**precision at K**

This metric tells you how relevant your model’s recommendations are based on a sample size of K (5, 10, or 25) recommendations.
Amazon Personalize calculates this metric based on the number of relevant recommendations out of the top K recommendations for each user in the testing set, divided by K, where K is 5, 10, or 25. The final metric is the average across all users in the testing set.

For example, if you recommend 10 items to a user, and the user interacts with 3 of them, the precision at K is 3 correctly predicted items divided by the total 10 recommended items: $\frac{3}{10} = .30$.

This metric rewards precise recommendation of relevant items. The closer the score is to one, the more precise the model.

**average_rewards_at_k**

When you create a solution version (train a model) for a solution with an optimization objective, Amazon Personalize generates an `average_rewards_at_k` metric. The score for `average_rewards_at_k` tells you how well the solution version performs in achieving your objective. To calculate this metric, Amazon Personalize calculates the rewards for each user as follows:

$$\text{rewards_per_user} = \frac{\text{total rewards from the user's interactions with their top 25 reward generating recommendations}}{\text{total rewards from the user's interactions with recommendations}}$$

The final `average_rewards_at_k` is the average of all `rewards_per_user` normalized to be a decimal value less than or equal to 1 and greater than 0. The closer the value is to 1, the more gains on average per user you can expect from recommendations.

For example, if your objective is to maximize revenue from clicks, Amazon Personalize calculates each user score by dividing total revenue generated by the items the user clicked from their top 25 most expensive recommendations by the revenue from all of the recommended items the user clicked. Amazon Personalize then returns a normalized average of all user scores. The closer the `average_rewards_at_k` is to 1, the more revenue on average you can expect to gain per user from recommendations.

For more information see [Optimizing a solution for an additional objective](p. 215).

**trend prediction accuracy**

If you trained the solution version with the `Trending-Now` recipe, the rate of increase in popularity of items recommended by the model. The higher the trend prediction accuracy (the closer to 1), the better the model is at correctly identifying trending items.

To calculate popularity acceleration, Amazon Personalize divides the rate of increase in popularity across all recommended items by the total popularity increase of the top 25 trending items. These items come from the actual interactions in the testing set.

Depending on your data distribution and what you choose for Trend discovery frequency, the value for trend prediction accuracy can be 0.0.

**hit (hit at K)**

If you trained the solution version with a USER_SEGMENTATION recipe, the average number of users in the predicted top relevant K results that match the actual users. Actual users are the users who actually interacted with the items in the test set. K is the top 1% of the most relevant users. The higher the value the more accurate the predictions.

**recall (recall at K)**

If you trained the solution version with a USER_SEGMENTATION recipe, the average percentage of predicted users in the predicted top relevant K results that match the actual users. Actual users are the users who actually interacted with the items in the test set. K is the top 1% of the most relevant users. The higher the value, the more accurate the predictions.
Example

The following is a simple example for a solution version that produces a list of recommendations for a specific user. The second and fifth recommendations match records in the testing data for this user. These are the relevant recommendations. If K is set at 5, the following metrics are generated for the user.

**reciprocal rank**
- Calculation: 1/2
- Result: 0.5000

**normalized discounted cumulative gain at 5**
- Calculation: \(\frac{1}{\log(1 + 2) + 1/\log(1 + 5)} / \left(\frac{1}{\log(1 + 1)} + \frac{1}{\log(1 + 2)}\right)\)
- Result: 0.6241

**precision at 5**
- Calculation: 2/5
- Result: 0.4000

Additional resources

For information on evaluating a solution version with A/B testing, see [Using A/B testing to measure the efficacy of recommendations generated by Amazon Personalize](#). To dive deeper in different types of metrics for recommender systems, see the following external resources:

- MRR vs MAP vs NDCG: Rank-Aware Evaluation Metrics And When To Use Them
- Discounted Cumulative Gain: the ranking metrics you should know about
- Recall and Precision at k for Recommender Systems
- Ranking Evaluation Metrics for Recommender Systems

Creating a campaign

For real-time recommendations with custom resources, after you complete [Creating a solution version](#), you are ready to deploy your solution version with a campaign.

A campaign deploys a solution version (trained model) with a provisioned transaction capacity for generating real-time recommendations. After you create a campaign, you use the `GetRecommendations` (p. 608) or `GetPersonalizedRanking` (p. 604) API operations to get recommendations. If you are getting batch recommendations, you don’t need to create a campaign. For more information see [Batch recommendations and user segments (custom resources)](#).

You create a campaign with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

**Important**

If you manually retrain your solution version or want to change your campaign settings, you must update your campaign. For more information see [Updating a campaign](#).

Topics

- Minimum provisioned transactions per second and auto-scaling (p. 238)
- Creating a campaign (console) (p. 238)
- Creating a campaign (AWS CLI) (p. 239)
- Creating a campaign (AWS SDKs) (p. 239)
- Updating a campaign (p. 241)
Minimum provisioned transactions per second and auto-scaling

**Important**
A high minProvisionedTPS will increase your bill. We recommend starting with 1 for minProvisionedTPS (the default). Track your usage using Amazon CloudWatch metrics, and increase the minProvisionedTPS as necessary.

When you create an Amazon Personalize campaign, you specify a provisioned transaction capacity for generating recommendations. A transaction is a single GetRecommendations or GetPersonalizedRanking call. Transactions per second (TPS) is the throughput and unit of billing for Amazon Personalize. The minimum provisioned TPS (minProvisionedTPS) specifies the baseline throughput provisioned by Amazon Personalize, and thus, the minimum billing charge.

If your TPS increases beyond minProvisionedTPS, Amazon Personalize auto-scales the provisioned capacity up and down, but never below minProvisionedTPS. There's a short time delay while the capacity is increased that might cause loss of transactions.

The actual TPS used is calculated as the average requests/second within a 5-minute window. You pay for maximum of the minimum provisioned TPS or the actual TPS. We recommend starting with a low minProvisionedTPS, track your usage using Amazon CloudWatch metrics, and then increase the minProvisionedTPS as necessary.

Creating a campaign (console)

After your solution version status is Active you are ready to deploy it with an Amazon Personalize campaign.

**To create a campaign (console)**

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose the dataset group with the solution version you want to deploy.
3. In the navigation pane, choose Campaigns.
4. On the Campaigns page, choose Create campaign.
5. On the Create new campaign page, for Campaign details, provide the following information:
   - **Campaign name**: Enter the name of the campaign. The text you enter here appears on the Campaign dashboard and details page.
   - **Solution**: Choose the solution that you just created.
   - **Solution version ID**: Choose the ID of the solution version that you just created.
   - **Minimum provisioned transactions per second (called minProvisionedTPS in APIs)**: Set the minimum provisioned transactions per second that Amazon Personalize supports. A high value will increase your bill. We recommend starting with 1 (the default). Track your usage using Amazon CloudWatch metrics, and increase the minProvisionedTPS as necessary. For more information, see Minimum provisioned transactions per second and auto-scaling (p. 238).
6. If you used the User-Personalization recipe, in Campaign configuration optionally enter values for the **Exploration weight** and **Exploration item age cut off**. For more information see User-Personalization (p. 116).
7. For **Tags**, optionally add any tags. For more information about tagging Amazon Personalize resources, see Tagging Amazon Personalize resources (p. 370).
8. Choose Create campaign.
9. On the campaign details page, when the campaign status is **Active**, you can use the campaign to get recommendations and record impressions. For more information, see Step 4: Getting recommendations (p. 244).
The campaign is ready when its status is ACTIVE. If you retrain your solution version or want to change your campaign settings, you must update your campaign. For more information see Updating a campaign (p. 241).

Creating a campaign (AWS CLI)

After your solution version status is Active, you are ready to deploy it with an Amazon Personalize campaign. Use the following create-campaign AWS CLI command to create a campaign that deploys a solution version trained using the User-Personalization recipe. Give the campaign a name and specify the solution version ARN (Amazon Resource Name). Optionally change the minProvisionedTPS if your use case requires a higher provisioned capacity. The minimum value is 1.

The campaign-config parameters are specific to the recipe that you used to train the solution version (for more information about recipes see Choosing a recipe (p. 113)). The example uses the following User-Personalization recipe specific itemExplorationConfig fields with their default values: explorationWeight and explorationItemAgeCutOff. If you omit the campaign-config parameter, the default values apply. For more information about the itemExplorationConfig fields, see the Properties and hyperparameters (p. 117) for the User-Personalization (p. 116) recipe.

```bash
aws personalize create-campaign \
   --name campaign name \
   --solution-version-arn solution version arn \
   --min-provisioned-tps 1 \
   --campaign-config "{"itemExplorationConfig":{"explorationWeight":"0.3", \
"explorationItemAgeCutOff":"30"}}"
```

The campaign is ready when its status is ACTIVE. To get the current status, call DescribeCampaign (p. 501) and check that the status field is ACTIVE.

If you retrain your solution version or want to change your campaign settings, you must update your campaign. For more information see Updating a campaign (p. 241).

Amazon Personalize provides operations for managing campaigns such as ListCampaigns (p. 540) to list the campaigns you have created. You can delete a campaign by calling DeleteCampaign (p. 477). If you delete a campaign, the solution versions that are part of the campaign are not deleted.

After you have created your campaign, use it to make recommendations. For more information, see Step 4: Getting recommendations (p. 244).

Creating a campaign (AWS SDKs)

After your solution version status is Active you are ready to deploy it with an Amazon Personalize campaign. Use the following code to create a campaign. Give the campaign a name, specify the Amazon Resource Name (ARN) of the solution version to deploy, and optionally specify the Minimum provisioned TPS (p. 238) the campaign will support (the default value for this parameter is 1). If you use the User-Personalization (p. 116) recipe, you can configure item exploration with the itemExplorationWeight and explorationItemAgeCutOff parameters.

SDK for Python (Boto3)

In this example, the itemExplorationWeight and explorationItemAgeCutOff parameters are specific to the User-Personalization (p. 116) recipe. The default itemExplorationWeight is 0.3 and the default explorationItemAgeCutOff is 30. If you leave out campaign configuration parameters, the default values apply.

```python
import boto3

personalize = boto3.client('personalize')
```
response = personalize.create_campaign(
    name = 'campaign name',
    solutionVersionArn = 'solution version arn',
    minProvisionedTPS = 1,
    campaignConfig = {"itemExplorationConfig": {"explorationWeight": "0.3", "explorationItemAgeCutOff": "30"}}
)

arn = response['campaignArn']
description = personalize.describe_campaign(campaignArn = arn)['campaign']
print('Name: ' + description['name'])
print('ARN: ' + description['campaignArn'])
print('Status: ' + description['status'])

SDK for Java 2.x

In this example, the itemExplorationWeight and explorationItemAgeCutOff parameters are specific to the User-Personalization (p. 116) recipe. The default itemExplorationWeight is 0.3 and the default explorationItemAgeCutOff is 30. If you leave out campaign configuration parameters, the default values apply.

public static void createCampaign(PersonalizeClient personalizeClient,
    String campaignName,
    String solutionVersionArn,
    Integer minProvisionedTPS,
    String itemExplorationWeight,
    String explorationItemAgeCutOff) {
    try {
        //Build a User-Personalization recipe specific campaignConfig object with the itemExploration map.
        //CampaignConfig construction will vary by recipe.
        CampaignConfig campaignConfig = CampaignConfig.builder()
            .itemExplorationConfig(itemExploration)
            .build();

        //build the createCampaignRequest
        CreateCampaignRequest createCampaignRequest = CreateCampaignRequest.builder()
            .name(campaignName)
            .solutionVersionArn(solutionVersionArn)
            .minProvisionedTPS(minProvisionedTPS)
            .campaignConfig(campaignConfig) //
            .build();

        //create the campaign
        CreateCampaignResponse campaignResponse = personalizeClient.createCampaign(createCampaignRequest);
        String campaignArn = campaignResponse.campaignArn();

        DescribeCampaignRequest campaignRequest = DescribeCampaignRequest.builder()
            .campaignArn(campaignArn)
            .build();

        DescribeCampaignResponse campaignResponse = personalizeClient.describeCampaign(campaignRequest);
Amazon Personalize Developer Guide
Creating custom resources

```java
Campaign newCampaign = campaignResponse.campaign();
System.out.println("The Campaign status is " + newCampaign.status());
}
}

} catch (PersonalizeException e) {
    System.err.println(e.awsErrorDetails().errorMessage());
    System.exit(1);
}

}

SDK for JavaScript v3

// Get service clients module and commands using ES6 syntax.
import { CreateCampaignCommand, PersonalizeClient } from "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({ region: "REGION" });

// set the campaign parameters
export const createCampaignParam = {
    solutionVersionArn: "SOLUTION_VERSION_ARN", /* required */
    name: "CAMPAIGN_NAME", /* required */
    minProvisionedTPS: 1, /* optional */
    campaignConfig: {
        itemExplorationConfig: {
            explorationWeight: "0.3", /* optional */
            explorationItemAgeCutOff: "30", /* optional */
        },
    },
};

export const run = async () => {
    try {
        const response = await personalizeClient.send(
            new CreateCampaignCommand(createCampaignParam)
        );
        console.log("Success", response);
        return response; // For unit tests.
    } catch (err) {
        console.log("Error", err);
    }
};

run();
```

The campaign is ready when its status is ACTIVE. To get the current status, call `DescribeCampaign` (p. 501) and check that the status field is ACTIVE.

If you manually retrain your solution version or want to change your campaign settings, you must update your campaign. For more information see `Updating a campaign` (p. 241).

Amazon Personalize provides operations for managing campaigns such as `ListCampaigns` (p. 540) to list the campaigns you have created. You can delete a campaign by calling `DeleteCampaign` (p. 477). If you delete a campaign, the solution versions that are part of the campaign are not deleted.

After you have created your campaign, use it to make recommendations. For more information, see `Step 4: Getting recommendations` (p. 244).

**Updating a campaign**

To deploy a retrained solution version with an existing campaign or to change your campaign's Minimum provisioned TPS (p. 238) or campaign configuration, you must manually update the campaign.
With User-Personalization, Amazon Personalize automatically updates your latest solution version (trained with `trainingMode` set to FULL) every two hours to include new items in recommendations, and your campaign automatically uses the updated solution version. Manually update a campaign only when you manually retrain the solution version with `trainingMode` set to FULL, or when you want to make changes to your campaign's `minProvisionedTPS` or campaign configuration. For more information on automatic updates with the User-Personalization recipe see Automatic updates (p. 116).

You manually update a campaign with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

Topics
- Updating a campaign (console) (p. 242)
- Updating a campaign (AWS CLI) (p. 242)
- Updating a campaign (AWS SDKs) (p. 242)

Updating a campaign (console)

To deploy a manually retrained solution version or make changes to your campaign configuration, you must update your campaign.

**To update a campaign (console)**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign into your account.
2. Choose the dataset group with the campaign you want to update.
3. In the navigation pane, choose Campaigns.
4. On the Campaigns page, choose the campaign you want to update.
5. On the campaign details page, choose Update.
6. On the Update campaign page, make your changes. For example, if you are deploying a retrained solution version, for Solution version ID, choose the identification number for the new solution version.
7. Choose Update. Amazon Personalize updates the campaign to use the new solution version and any changed configurations.

Updating a campaign (AWS CLI)

To deploy a new solution version, change your campaign's Minimum provisioned TPS (p. 238), or change your campaign's configuration, you must update your campaign. Use the following `update-campaign` command to update a campaign to use a new solution version with the AWS CLI.

Replace campaign arn with the Amazon Resource Name (ARN) of the campaign you want to update. Replace new solution version arn with the solution version you want to deploy.

```shell
aws personalize update-campaign \
--campaign-arn campaign arn \
--solution-version-arn new solution version arn \
--min-provisioned-tps 1
```

Updating a campaign (AWS SDKs)

To deploy a new solution version, change your campaign's Minimum provisioned TPS (p. 238) or change your campaign's configuration, you must update your campaign. Use the following code to update a
campaign with the SDK for Python (Boto3) or SDK for Java 2.x. For a complete list of parameters, see UpdateCampaign (p. 588).

SDK for Python (Boto3)

Use the following update_campaign method to deploy a new solution version. Replace campaign arn with the Amazon Resource Name (ARN) of the campaign you want to update, replace the new solution version arn with the new solution version ARN and optionally change the minProvisionedTPS.

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.update_campaign(
    campaignArn = 'campaign arn',
    solutionVersionArn = 'new solution version arn',
    minProvisionedTPS = 1,
)

arn = response['campaignArn']

description = personalize.describe_campaign(campaignArn = arn)['campaign']

print('Name: ' + description['name'])
print('ARN: ' + description['campaignArn'])
print('Status: ' + description['status'])
```

SDK for Java 2.x

Use the following updateCampaign method to update a campaign to use a new solution version. Pass as parameters an Amazon Personalize service client, the new solution version's Amazon Resource Name (ARN), and the Minimum provisioned TPS (p. 238).

```java
public static void updateCampaign(PersonalizeClient personalizeClient,
    String campaignArn,
    String solutionVersionArn,
    Integer minProvisionedTPS) {

    try {
        // build the updateCampaignRequest
        UpdateCampaignRequest updateCampaignRequest = UpdateCampaignRequest.builder()
            .campaignArn(campaignArn)
            .solutionVersionArn(solutionVersionArn)
            .minProvisionedTPS(minProvisionedTPS)
            .build();

        // update the campaign
        personalizeClient.updateCampaign(updateCampaignRequest);

        DescribeCampaignRequest campaignRequest = DescribeCampaignRequest.builder()
            .campaignArn(campaignArn)
            .build();

        DescribeCampaignResponse campaignResponse = personalizeClient.describeCampaign(campaignRequest);
        Campaign updatedCampaign = campaignResponse.campaign();

        System.out.println("The Campaign status is "+ updatedCampaign.status());
    } catch (PersonalizeException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
```
Step 4: Getting recommendations

Depending on your resources, you can get recommendations in real time or with a batch workflow.

- With custom resources, you can get real-time recommendations or batch recommendations. For real-time recommendations, you must create a custom campaign before you get recommendations. For batch recommendations, you don’t need to create a campaign.
- With recommenders in a Domain dataset group, you can get only real-time recommendations.

The following topics explain how and when to use each recommendation type.

**Topics**
- Recommendation scores (p. 244)
- Getting real-time recommendations (p. 244)
- Batch recommendations and user segments (custom resources) (p. 259)

**Recommendation scores**

With custom solutions created with the User-Personalization and Personalized-Ranking recipes, Amazon Personalize includes a score for each item in recommendations. These scores represent the relative certainty that Amazon Personalize has about which item the user will select next. Higher scores represent greater certainty.

For information on scores for User-Personalization, see How User-Personalization recommendation scoring works (custom resources) (p. 245). For information on scores for Personalized-Ranking recommendations, see How personalized ranking scoring works (p. 255).

For batch inference jobs, item scores are calculated just as described in How User-Personalization recommendation scoring works (custom resources) (p. 245) and How personalized ranking scoring works (p. 255). You can view scores in the batch inference job's output JSON file.

**Getting real-time recommendations**

Real-time recommendations are recommendations you request and show your user as they use your application. You can get real-time recommendations from Amazon Personalize with a recommender (for Domain dataset groups) or a custom campaign.

- For domain recommenders, you can get real-time recommendations with the the section called “GetRecommendations” (p. 608) operation. Or you can test your recommender with the Amazon Personalize console.
  
- For custom resources, depending on the recipe you used to create the solution version backing the campaign, you get recommendations for your users with the the section called “GetRecommendations” (p. 608) or the section called “GetPersonalizedRanking” (p. 604) API operations. Or you can test your campaign with the Amazon Personalize console.

If you use domain use cases or recipes that provide real-time personalization, such as the Top picks for you use case or the User-Personalization recipe, Amazon Personalize updates recommendations based
on your user's most recent activity as they interact with your catalog. For more information on recording real-time events and personalization, see Recording events (p. 280).

For some use cases and recipes, you can specify a promotion in your request. A promotion defines additional business rules that apply to a configurable subset of recommended items. For more information see Promoting items in recommendations (p. 248).

Topics
- Getting recommendations (p. 245)
- Getting a personalized ranking (custom resources) (p. 254)
- Increasing recommendation relevance with contextual metadata (p. 258)

Getting recommendations

You can get recommendations from a Amazon Personalize recommender or custom campaign with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

Note
If you used a PERSONALIZED_RANKING custom recipe, see Getting a personalized ranking (custom resources) (p. 254).

Topics
- How User-Personalization recommendation scoring works (custom resources) (p. 245)
- Getting recommendations (console) (p. 246)
- Getting recommendations (AWS CLI) (p. 246)
- Getting recommendations (AWS SDKs) (p. 246)
- Promoting items in recommendations (p. 248)

How User-Personalization recommendation scoring works (custom resources)

With the User-Personalization recipe, Amazon Personalize generates scores for items based on on a user's interaction data and metadata. These scores represent the relative certainty that Amazon Personalize has in whether the user will interact with the item next. Higher scores represent greater certainty.

Amazon Personalize scores all of the items in your catalog relative to each other on a scale from 0 to 1 (both inclusive), so that the total of all scores equals 1. For example, if you're getting movie recommendations for a user and there are three movies in the Items dataset, their scores might be 0.6, 0.3, and 0.1. Similarly, if you have 1,000 movies in your inventory, the highest-scoring movies might have very small scores (the average score would be .001), but, because scoring is relative, the recommendations are still valid.

In mathematical terms, scores for each user-item pair \((u, i)\) are computed according to the following formula, where \(\exp\) is the exponential function, \(\bar{w}_u\) and \(w_i\) are user and item embeddings respectively, and the Greek letter sigma (\(\Sigma\)) represents summation over all items in the item dataset:

\[
\text{score}(u, i) = \frac{\exp(\bar{w}_u^T w_i)}{\sum_j \exp(\bar{w}_u^T w_j)}
\]

Note
Amazon Personalize doesn't show scores for domain recommenders or the Similar-Items, SIMS or Popularity-Count recipes. For information on scores for Personalized-Ranking recommendations, see How personalized ranking scoring works (p. 255).
Getting recommendations (console)

To get recommendations with the Amazon Personalize console, you provide the request information on the details page of either a recommender (Domain dataset group) or a custom campaign.

To get recommendations

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose the dataset group that contains the campaign or recommender you are using.
3. In the navigation pane, choose Campaigns or Recommenders.
4. Choose the target campaign or recommender.
5. For a campaigns, under Test campaign results, enter your recommendation request details based on the recipe you used. For a recommenders choose Test recommender and enter your recommendation request details based on your use case.

   If you recorded events for a user before they logged in (an anonymous user), you can get recommendations for this user by providing the sessionId from those events instead of a userId. For more information about recording events for anonymous users, see Recording events with the PutEvents operation (p. 283).
6. Optionally choose a filter. For more information, see Filtering recommendations and user segments (p. 311).
7. If you use contextual metadata, provide data for each context. For each context, for the Key enter the metadata field. For the Value enter the context data. For more information, see Increasing recommendation relevance with contextual metadata (p. 258).
8. If you want to promote a subset of items, optionally complete the Promotion fields. For more information see Promoting items in recommendations (p. 248).
9. Choose Get recommendations. A table containing the user’s top 25 recommended items displays.

Getting recommendations (AWS CLI)

Use the following code to get recommendations from a campaign. To get recommendations from a recommender, replace the campaign-arn with the recommender-arn. Specify the ID of the user you want to get recommendations for, and the Amazon Resource Name (ARN) of your campaign or recommender. A list of the top 10 recommended items for the user displays. To change the number of recommended items, change the value for numResults. The default is 25 items. The maximum is 500 items. If you used a RELATED_ITEMS recipe to train the solution version backing the campaign, replace the user-id parameter with item-id and specify the item ID.

If you recorded events for a user before they logged in (an anonymous user), you can get recommendations for this user by providing the sessionId from those events instead of a userId. For more information about recording events for anonymous users, see Recording events with the PutEvents operation (p. 283).

```
aws personalize-runtime get-recommendations \
--campaign-arn campaign arn \
--user-id User ID \
--num-results 10
```

Getting recommendations (AWS SDKs)

The following code shows how to get Amazon Personalize recommendations for a user from a campaign. To get recommendations from a recommender, replace the campaignArn with the recommenderArn. Specify the ID of the user you want to get recommendations for, and the Amazon Resource Name (ARN) of your campaign or recommender. A list of the top 10 recommended items for the user displays. To
change the number of recommended items, change the value for `numResults`. The default is 25 items. The maximum is 500 items. If you used a RELATED_ITEMS recipe to train the solution version backing the campaign, replace the `userId` parameter with `itemId` and specify the item ID.

If you recorded events for a user before they logged in (an anonymous user), you can get recommendations for this user by providing the `sessionId` from those events instead of a `userId`. For more information about recording events for anonymous users, see Recording events with the PutEvents operation (p. 283).

**SDK for Python (Boto3)**

```python
import boto3

personalizeRt = boto3.client('personalize-runtime')

response = personalizeRt.get_recommendations(
    campaignArn = 'Campaign ARN',
    userId = 'User ID',
    numResults = 10
)

print("Recommended items")
for item in response['itemList']:
    print (item['itemId'])
```

**SDK for Java 2.x**

```java
public static void getRecs(PersonalizeRuntimeClient personalizeRuntimeClient, String campaignArn, String userId){
    try {
        GetRecommendationsRequest recommendationsRequest =
            GetRecommendationsRequest.builder()
                .campaignArn(campaignArn)
                .numResults(20)
                .userId(userId)
                .build();

        GetRecommendationsResponse recommendationsResponse =
            personalizeRuntimeClient.getRecommendations(recommendationsRequest);
        List<PredictedItem> items = recommendationsResponse.itemList();
        for (PredictedItem item: items) {
            System.out.println("Item Id is : "+item.itemId());
            System.out.println("Item score is : "+item.score());
        }
    } catch (AwsServiceException e) {
        System.err.println(e.awsErrorDetails().errorMessage());
        System.exit(1);
    }
}
```

**SDK for JavaScript v3**

```javascript
// Get service clients module and commands using ES6 syntax.
import { GetRecommendationsCommand } from "@aws-sdk/client-personalize-runtime";

import { personalizeRuntimeClient } from "./libs/personalizeClients.js";
// Or, create the client here.
// const personalizeRuntimeClient = new PersonalizeRuntimeClient({ region: "REGION"});
```
Promoting items in recommendations

With all domain use cases and some custom recipes, you can specify a promotion when you get recommendations. A promotion defines additional business rules that apply to a configurable subset of recommended items. For example, you might have a streaming app and want to promote your own shows and movies but also recommend relevant titles. You could use a promotion to specify that a certain percentage of recommended items must come from the category in-house. The remaining recommended items would continue to be relevant recommendations based on your recipe and any request filters.

To apply a promotion, you specify the following in your recommendation request:

- The percentage of recommended items to apply the promotion filter to.
- A filter that specifies the promotion criteria. For more information, see Promotion filters (p. 249).

In the recommendation response, promoted items are positioned randomly relative to other recommended items, but in sorted order relative to other promoted items. Depending on your recipe, recommended items that aren't part of a promotion are sorted by relevance to the user, popularity, or similarity. If there aren't enough items that meet the promotion criteria, the result will contain as many promoted items as possible.

You can apply a promotion to recommendations with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

Topics

- Use cases and recipes that support promotions (p. 248)
- Promotion filters (p. 249)
- Promoting items (console) (p. 249)
- Promoting items (AWS CLI) (p. 250)
- Promoting items (AWS SDKs) (p. 251)

Use cases and recipes that support promotions

All use cases support promotions. The following custom recipes support promotions:

- USER_PERSONALIZATION (p. 115) recipes
Promotion filters

When you apply a promotion to a recommendation request, you choose a filter that specifies the promotion criteria. You can use an existing filter or create a new one. You create and manage filters for promotions as you would other filters in Amazon Personalize. For information about creating and managing filters, see Filtering results (p. 311).

The only difference between a promotion filter and a filter that you choose outside the promotion (the request filter) is how Amazon Personalize applies them. A promotion filter applies to only promoted items, while a request filter applies to only the remaining recommended items. If you specify a request filter and promotion filter, and want to apply both filters to promoted items, your promotion filter's expression must include both expressions. The way you combine two expressions depends on the datasets you use. For more information on filter expressions, their rules, and how to create them, see Filter expressions (p. 311).

Filter expression examples

The following expression includes only items from the category "in-house". You might use this expression if you want to promote your own content in your recommendations.

```plaintext
INCLUDE ItemID WHERE Items.OWNER IN ("in-house")
```

The following expression includes only items created earlier than a timestamp you specify. You might use this expression to promote items created recently.

```plaintext
INCLUDE ItemID WHERE Items.CREATION_TIMESTAMP < $DATE
```

The following expression shows how you might apply a request filter to promoted items. It includes only available clothing items as promoted items. In this scenario, the Items.AVAILABLE_IN ("True") would also be used in the request filter expression, so that all recommendations are for items that are available.

```plaintext
INCLUDE ItemID WHERE Items.CATEGORY IN ("clothing") AND Items.AVAILABLE IN ("True")
```

For a more complete list of filter examples, see Filter expression examples (p. 314).

Promoting items (console)

To promote certain items in recommendations with the Amazon Personalize console, create a filter, and then provide the promotion details in the recommendation request. For information on other fields, see Getting recommendations (console) (p. 246).

To promote items in recommendations

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose the dataset group that contains the campaign or recommender you are using.
3. If you haven’t already, create a filter that specifies the promotion criteria. You create filters for promotions the same way that you create request filters. For information on creating and managing filters, see Filtering results (p. 311).
4. In the navigation pane, choose Recommenders or Campaigns.
5. Choose the target campaign or recommender.
6. For campaigns, under Test campaign results, enter your recommendation request details based on the recipe you used. For recommenders, choose Test recommender and enter your recommendation request details.
7. Optionally choose a filter for the request. This filter applies to only non-promoted items. For information on creating and managing filters, see Filtering results (p. 311).
8. If you use contextual metadata, provide data for each context. For each context, for the Key enter the metadata field. For the Value, enter the context data. For more information, see Increasing recommendation relevance with contextual metadata (p. 258).
9. For Promotion specify the following:
   - **Percent promoted items**: Enter the percentage of recommended items to apply the promotion to.
   - **Filter**: Choose a filter that specifies the promotion criteria. This filter applies to the promoted items instead of any request filter that you may have specified in step 7.
   - **Filter parameter**: If your promotion uses a filter with placeholder parameters, for each parameter, enter the value to set the filter criteria. To use multiple values for one parameter, separate each value with a comma.
10. Choose Get recommendations. A table containing the user’s top 25 recommended items displays. The Promoted item column indicates whether the item was included because of your promotion. Promoted items are positioned randomly relative to other recommended items, but in sorted order relative to other promoted items. Depending on your use case or recipe, recommended items that aren’t part of a promotion are sorted by relevance to the user, popularity, or similarity. If there aren’t enough items that meet the promotion criteria, the result will contain as many promoted items as possible.

**Promoting items (AWS CLI)**

The following code shows how to promote items in recommendations with the AWS CLI and a custom campaign. To promote items with a recommender, replace the campaign-arn parameter with a recommender-arn and specify the Amazon Resource Name (ARN) for the recommender. For the promotion fields, specify the following:

- **name**: Give the promotion a name. The recommendation response uses the name to identify promoted items.
- **percent-promoted-items**: The percentage of recommended items to apply the promotion to. In this example, 50% of items will be promoted items.
- **filterArn**: Specify the Amazon Resource Name (ARN) of the filter that defines the promotion criteria. For more information, see Promotion filters (p. 249).
- **parameter names and values**: If your filter expression has any parameters, provide the parameter names (case sensitive) and the values. For example, if your filter expression has a $GENRE parameter, provide GENRE as the key, and a genre or genres, such as Comedy, as the value. Separate multiple values with a comma. When you use the AWS CLI, for each value you must use the / character to escape both quotes and the / character. The following code example shows how to format the values.

The code shows how to use both a request filter and a promotion filter. A promotion filter applies to only promoted items, while a request filter applies to only the remaining recommended items. For more information, see Promotion filters (p. 249).

For information about additional fields, see Getting recommendations (AWS SDKs) (p. 246) and Getting a personalized ranking using contextual metadata (AWS Python SDK) (p. 257).

```
aws personalize-runtime get-recommendations
--campaign-arn CampaignArn
```
A list of recommended items displays. Promoted items are positioned randomly relative to other recommended items, but in sorted order relative to other promoted items. Depending on your recipe, recommended items that aren't part of a promotion are sorted by relevance to the user, popularity, or similarity. If there aren't enough items that meet the promotion criteria, the result will contain as many promoted items as possible.

```
{
  "itemList": [
    {
      "itemId1": "123",
      "score": .0117211,
      "promotionName": "promotionName"
    },
    {
      "itemId2": "456",
      "score": .0077976
    },
    {
      "itemId3": "789",
      "score": .0067171
    },
    ...
  ]
}
```

Promoting items (AWS SDKs)

The following code shows how to promote items in recommendations with the SDK for Python (Boto3) and the SDK for Java 2.x and a custom campaign. To promote items with a recommender, replace the campaignArn parameter with recommenderArn and specify the Amazon Resource Name (ARN) for the recommender. For the promotion fields, specify the following:

- **name**: Specify the name of the promotion. The recommendation response includes the name to identify promoted items.
- **percentPromotedItems**: The percentage of recommended items to apply the promotion to.
- **promotionFilterARN**: The Amazon Resource Name (ARN) of the filter that defines the promotion criteria. For more information, see Promotion filters (p. 249).
- **Any parameter names and values**: If your filter expression has any parameters, for each parameter in your filter expression, provide the parameter name (case sensitive) and the values. For example, if your filter expression has a $GENRE parameter, provide "GENRE" as the key, and a genre or genres, such as "Comedy", "drama", "horror", as the value. Separate multiple values with a comma. For example, "comedy," , "drama," , "horror".

The following code shows how to use both a request filter and a promotion filter. A promotion filter applies to only promoted items, while a request filter applies to only the remaining recommended items. For more information, see Promotion filters (p. 249).
For information about additional fields, see Getting recommendations (AWS SDKs) (p. 246) and Getting a personalized ranking using contextual metadata (AWS Python SDK) (p. 257).

SDK for Python (Boto3)

```python
import boto3

personalizeRt = boto3.client('personalize-runtime')

response = personalizeRt.get_recommendations(
    campaignArn = "CampaignARN",
    userId = '1',
    numResults = 10,
    filterArn = 'RequestFilterARN',
    filterValues = {
        "RequestFilterParameterName": "value1",
        "RequestFilterParameterName": "value1","value2","value3"
    },
    promotions = [{
        "name" : "promotionName",
        "percentPromotedItems" : 50,
        "filterArn": "promotionFilterARN",
        "filterValues": {
            "PromotionParameterName": "Value1","Value2"
        }
    }]
)

print("Recommended items")
for item in response['itemList']:
    print(item['itemId'])
    if ("promotionName" in item):
        print(item['promotionName'])
```

SDK for Java 2.x

```java
public static void getRecommendationsWithPromotedItems(PersonalizeRuntimeClient personalizeRuntimeClient, 
            String campaignArn, 
            String userId, 
            String requestFilterArn, 
            String requestParameterName, 
            String requestParameterValue1, 
            String requestParameterValue2, 
            String promotionName, 
            int percentPromotedItems, 
            String promotionFilterArn, 
            String promotionParameterName, 
            String promotionParameterValue1, 
            String promotionParameterValue2) {
    try {
        Map<String, String> promotionFilterValues = new HashMap<>();
        promotionFilterValues.put(promotionParameterName, String.format("\%1$s\",\%2$s \",
            promotionParameterValue1, promotionParameterValue2));

        Promotion newPromotion = Promotion.builder() 
            .name(promotionName) 
            .percentPromotedItems(percentPromotedItems) 
```
Getting real-time recommendations

```java
filterArn(promotionFilterArn)
.filterValues(promotionFilterValues)
.build();

List<Promotion> promotionList = new List<>();
promotionsList.add(newPromotion);

Map<String, String> requestfilterValues = new HashMap<>();
requestfilterValues.put(requestParameterName, String.format("\"%1$s\",\"%2$s\"",
requestParameterValue1, requestParameterValue2));

GetRecommendationsRequest recommendationsRequest =
GetRecommendationsRequest.builder()
 .campaignArn(campaignArn)
 .numResults(20)
 .userId(userId)
 .filterArn(requestFilterArn)
 .fitlerValues(requestFilterValues)
 .promotions(promotionList)
 .build();

GetRecommendationsResponse recommendationsResponse =
personalizeRuntimeClient.getRecommendations(recommendationsRequest);
List<PredictedItem> items = recommendationsResponse.itemList();

for (PredictedItem item: items) {
    System.out.println("Item Id is : "+item.itemId());
    System.out.println("Item score is : "+item.score());
    System.out.println("Promotion name is : "+item.promotionName());
}
```

SDK for JavaScript v3

```javascript
// Get service clients and commands using ES6 syntax.
import { GetRecommendationsCommand, PersonalizeRuntimeClient } from
"@aws-sdk/client-personalize-runtime";

// create personalizeRuntimeClient.
const personalizeRuntimeClient = new PersonalizeRuntimeClient({
    region: "REGION",
});

// set recommendation request param
export const getRecommendationsParam = {
    campaignArn: "CAMPAIGN_ARN", /* required */
    userId: "USER_ID", /* required */
    numResults: 25, /* optional */
    filterArn: "FILTER_ARN", /* provide if you are applying a custom filter */
    filterValues: {
        "PARAM_NAME": "\\"PARAM_VALUE\\" /* provide if your filter has a placeholder
parameter */
    },
    promotions: [
        {
            name: "PROMOTION_NAME", /* specify the name of the promotion. The recommendation
response includes the name to identify promoted items. */
            percentPromotedItems: 50, /* the percentage of recommended items to apply the
promotion to. */
```
filterArn:  
"PROMOTION_FILTER_ARN", /* the Amazon Resource Name (ARN) of the filter that defines the promotion criteria. */  
filterValues: {  
  "PARAM_NAME": "\"PARAM_VALUE\"", /* provide if your promotion filter has a placeholder parameter */  
},  
],  
};

export const run = async () => {  
try {  
  const response = await personalizeRuntimeClient.send(new  
  GetRecommendationsCommand(getRecommendationsParam));  
  console.log("Success!", "\nItems are: ");  
  response.itemList.forEach(element => console.log(element.itemId))  
  return response; // For unit tests.  
} catch (err) {  
  console.log("Error", err);  
}  
};  
run();

A list of recommended items displays. Promoted items are positioned randomly relative to other recommended items, but in sorted order relative to other promoted items. Depending on your recipe, recommended items that aren't part of a promotion are sorted by relevance to the user, popularity, or similarity. If there aren't enough items that meet the promotion criteria, the result will contain as many promoted items as possible.

[  
  "itemList": [  
    {  
      "itemId1": "123",  
      "score": 0.0117211,  
      "promotionName": "promotionName"  
    },  
    {  
      "itemId2": "456",  
      "score": 0.0077976  
    },  
    {  
      "itemId3": "789",  
      "score": 0.0067171  
    },  
    .....  
  ]  
]

Getting a personalized ranking (custom resources)

A personalized ranking is a list of recommended items that are re-ranked for a specific user. To get personalized rankings, call the GetPersonalizedRanking (p. 604) API operation or get recommendations from a campaign in the console.

Note
The solution backing the campaign must have been created using a recipe of type PERSONALIZED_RANKING. For more information, see Choosing a recipe (p. 113).

Topics
Getting real-time recommendations

- **How personalized ranking scoring works (p. 255)**
- **Getting a personalized ranking (console) (p. 255)**
- **Getting a personalized ranking (AWS CLI) (p. 255)**
- **Getting a personalized ranking (AWS SDKs) (p. 256)**
- **Getting a personalized ranking using contextual metadata (AWS Python SDK) (p. 257)**
- **Personalized-Ranking sample notebook (p. 258)**

**How personalized ranking scoring works**

Like the scores returned by the GetRecommendations operation for solutions created with the User-Personalization (p. 245) recipe, GetPersonalizedRanking scores sum to 1, but because the list of considered items is much smaller than your full catalog, recommendation scores tend to be higher.

Mathematically, the scoring function for GetPersonalizedRanking is identical to GetRecommendations, except that it only considers the input items. This means that scores closer to 1 become more likely, as there are fewer other choices to divide up the score:

$$
\text{score}(u, i) = \frac{\exp(\mathbf{w}_u^\top \mathbf{w}_i)}{\sum_{j \in \text{input}} \exp(\mathbf{w}_u^\top \mathbf{w}_j)}
$$

**Getting a personalized ranking (console)**

To get a personalized ranking for a user from the Amazon Personalize console, choose the campaign that you are using and then provide their user ID, specify the list of items you want ranked for the user, optionally choose a filter, and optionally provide any context data.

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign into your account.
2. Choose the dataset group that contains the campaign you are using.
3. In the navigation pane, choose **Campaigns**.
4. On the **Campaigns** page, choose the target campaign.
5. Under **Test campaign results**, enter the **User ID** of the user that you want to get recommendations for.
6. For **Item IDs**, enter the list of items to be ranked for the user.
7. Optionally choose a filter. For more information, see Filtering recommendations and user segments (p. 311).
8. If your campaign uses contextual metadata (for requirements see Increasing recommendation relevance with contextual metadata (p. 258)) optionally provide context data.

   For each context, for the **Key**, enter the metadata field, and for the **Value**, enter the context data.
9. Choose **Get personalized item rankings**. A table containing the items ranked in order of predicted interest for the user appears.

**Getting a personalized ranking (AWS CLI)**

Use the following `get-personalized-ranking` command to get a personalized ranking with the AWS CLI. Specify the Amazon Resource Name (ARN) for your campaign, the User ID for the user, and provide
a list of item IDs for the items to be ranked for the user (each separated by a space). The items to be ranked must be in the data that you used to train the solution version. A list of ranked recommendations displays. Amazon Personalize considers the first item in the list of most interest to the user.

```
aws personalize-runtime get-personalized-ranking \
--campaign-arn Campaign ARN \ 
--user-id 12 \ 
--input-list 3 4 10 8 12 7
```

Getting a personalized ranking (AWS SDKs)

The following code shows how to get a personalized ranking for a user. Specify the user’s ID and a list of item IDs to be ranked for the user. The item IDs must be in the data that you used to train the solution version. A list of ranked recommendations is returned. Amazon Personalize considers the first item in the list of most interest to the user.

SDK for Python (Boto3)

```python
import boto3

personalizeRt = boto3.client('personalize-runtime')

response = personalizeRt.get_personalized_ranking(
    campaignArn = "Campaign arn",
    userId = "UserID",
    inputList = ['ItemID1','ItemID2']
)

print("Personalized Ranking")
for item in response['personalizedRanking']:
    print (item['itemId'])
```

SDK for Java 2.x

```java
public static List<PredictedItem> getRankedRecs(PersonalizeRuntimeClient personalizeRuntimeClient,
                     String campaignArn,
                     String userId,
                     ArrayList<String> items) {
    try {
        GetPersonalizedRankingRequest rankingRecommendationsRequest =
            GetPersonalizedRankingRequest.builder()
                .campaignArn(campaignArn)
                .userId(userId)
                .inputList(items)
                .build();

        GetPersonalizedRankingResponse recommendationsResponse =
            personalizeRuntimeClient.getPersonalizedRanking(rankingRecommendationsRequest);
        List<PredictedItem> rankedItems =
            recommendationsResponse.personalizedRanking();
        int rank = 1;
        for (PredictedItem item : rankedItems) {
            System.out.println("Item ranked at position " + rank + " details");
            System.out.println("Item Id is : " + item.itemId());
            System.out.println("Item score is : " + item.score());
            System.out.println("---------------------------------------------");
            rank++;
        }
    }
```

256
Getting real-time recommendations

```java
return rankedItems;
} catch (PersonalizeRuntimeException e) {
    System.err.println(e.awsErrorDetails().errorMessage());
    System.exit(1);
} return null;
}

SDK for JavaScript v3

import { GetPersonalizedRankingCommand } from "@aws-sdk/client-personalize-runtime";
import { personalizeRuntimeClient } from "./libs/personalizeClients.js";

export const getPersonalizedRankingParam = {
    campaignArn: "CAMPAIGN_ARN", /* required */
    userId: 'USER_ID',      /* required */
    inputList: ['ITEM_ID_1', "ITEM_ID_2", 'ITEM_ID_3', "ITEM_ID_4"]
}

export const run = async () => {
    try {
        const response = await personalizeRuntimeClient.send(new GetPersonalizedRankingCommand(getPersonalizedRankingParam));
        console.log("Success!", response);
        return response; // For unit tests.
    } catch (err) {
        console.log("Error", err);
    }
}

run();

Getting a personalized ranking using contextual metadata (AWS Python SDK)

Use the following code to get a personalized ranking based on contextual metadata. For context, for each key-value pair, provide the metadata field as the key and the context data as the value. In the following sample code, the key is DEVICE and the value is mobile phone. Replace these values and the Campaign ARN and User ID with your own. Also change inputList to a list of item IDs that are in the data that you used to train the solution. Amazon Personalize considers the first item in the list of most interest to the user.

```python
import boto3
personalizeRt = boto3.client('personalize-runtime')
response = personalizeRt.get_personalized_ranking(
    campaignArn = "Campaign ARN",
    userId = "User ID",
    inputList = ["ItemID1", "ItemID2"],
    context = {
        'DEVICE': 'mobile phone'
    }
)

print("Personalized Ranking")
for item in response['personalizedRanking']:
```
print(item['itemId'])

**Personalized-Ranking sample notebook**

For a sample Jupyter notebook that shows how to use the Personalized-Ranking recipe see [Personalize Ranking Example](https://aws.amazon.com/blogs/machine-learning/).

**Increasing recommendation relevance with contextual metadata**

To increase recommendation relevance, include contextual metadata for a user, such as their device type or the time of day, when you get recommendations or get a personalized ranking.

To use contextual metadata, the schema of the Interactions dataset must have a metadata fields for the contextual data. For example, a DEVICE field (see [Schemas](https://docs.aws.amazon.com/personalize/latest/developerguide/interaction-data.html#interaction-data-schemas)).

For Domain dataset groups, the following recommender use cases can use contextual metadata:

- **Recommended for you** (p. 112) (ECOMMERCE domain)
- **Top picks for you** (p. 109) (VIDEO_ON_DEMAND domain)

For custom resources, recipes that use contextual metadata include the following:

- **User-Personalization** (p. 116)
- **Personalized-Ranking** (p. 139)

For more information on contextual information, see the following AWS Machine Learning Blog post: [Increasing the relevance of your Amazon Personalize recommendations by leveraging contextual information](https://aws.amazon.com/blogs/machine-learning/).

You can get recommendations with contextual metadata with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

### Getting recommendations using contextual metadata (AWS Python SDK)

To increase recommendation relevance, include contextual metadata for a user, such as their device type or the time of day, when you get recommendations or get a personalized ranking.

Use the following code to get a recommendation based on contextual metadata. For context, for each key-value pair, provide the metadata field as the key and the context data as the value. In the following sample code, the key is `DEVICE` and the value is `mobile phone`. Replace these values and the Campaign ARN and User ID with your own. If you created a recommender, replace `campaignArn` with `recommenderArn`. A list of recommended items for the user displays.

```python
import boto3

personalizeRt = boto3.client('personalize-runtime')

response = personalizeRt.get_recommendations(
    campaignArn = 'Campaign ARN',
    userId = 'User ID',
    context = {
        'DEVICE': 'mobile phone'
    }
)

print("Recommended items")
```
Batch recommendations and user segments (custom resources)

With custom resources, you can get batch recommendations or user segments with an asynchronous batch flow. For example, you might get product recommendations for all users on an email list or item-to-item similarities (p. 143) across an inventory. Or with the USER_SEGMENTATION recipes, you can get user segments for data-driven advertising based on items in your inventory and your user's interactions.

- To get batch recommendations, you use a batch inference job. A batch inference job is a tool that imports your batch input data from an Amazon S3 bucket, uses your solution version to generate item recommendations, and exports the recommendations to an Amazon S3 bucket.

- To get user segments, you use a batch segment job. A batch segment job is a tool that imports your batch input data from an Amazon S3 bucket, uses your solution version trained with a USER_SEGMENTATION recipe to generate user segments, and exports the segments to an Amazon S3 bucket.

Topics
- Getting batch recommendations (p. 259)
- Getting user segments (p. 269)

Getting batch recommendations

With custom resources, you can get item recommendations with an asynchronous batch flow. For example, you might get product recommendations for all users on an email list or item-to-item similarities (p. 143) across an inventory. To get batch recommendations for items, you use a batch inference job. A batch inference job is a tool that imports your batch input data from an Amazon S3 bucket, uses your custom solution version to generate item recommendations, and then exports the item recommendations to an Amazon S3 bucket. Depending on the recipe, your input data is a list of users, or items, or a list of users each with a collection of items.

When generating batch recommendations, Amazon Personalize considers all bulk data present at the time of latest solution version creation. This data can be imported with an import mode of FULL or INCREMENTAL. For newer bulk records to influence batch recommendations, you must create a new solution version and then create the batch inference job.

Amazon Personalize uses data from individual imports when generating batch recommendations as follows:

- New interactions with existing items and users: If you use the User-Personalization or Personalized-Ranking recipes, Amazon Personalize considers new interactions data with existing items and users within about 15 minutes from data import. To make sure events are considered, we recommend you wait at minimum 15 minutes after import before you start a batch inference job. For all other recipes, you must create a new solution version for streamed events to influence batch recommendations.

- New users: For users without interactions data, recommendations are initially for only popular items. If you use use User-Personalization or Personalized-Ranking and you record events for the user, their recommendations might become more relevant within about 15 minutes after import without retraining. To make sure events are considered, we recommend you wait at minimum 15 minutes after import before you start a batch inference job. For all other recipes, you must create a new solution version for streamed events to influence batch recommendations for users without interactions data.
New items: With User-Personalization, when you create a batch inference job and specify the latest fully trained solution version for your solution, Amazon Personalize automatically updates the solution version to include new items in recommendations with exploration. If you don't specify the latest solution version, no update occurs. For any other recipe, you must create a new solution version for new items to be featured in batch recommendations. For more information about exploration, see Exploration (p. 106).

The batch workflow is as follows:

1. Prepare and upload your input data in JSON format to an Amazon S3 bucket. The format of your input data depends on the recipe you use. See Preparing input data for batch recommendations (p. 261).
2. Create a separate location for your output data, either a folder or a different Amazon S3 bucket.
4. When the batch inference is complete, retrieve the item recommendations from your output location in Amazon S3.

Guidelines and requirements

The following are guidelines and requirements for getting batch recommendations:

- Your Amazon Personalize IAM service role must have permission to read and add files to your Amazon S3 buckets. For information on granting permissions, see Service role policy for batch workflows (p. 17). For more information on bucket permissions, see User policy examples in the Amazon Simple Storage Service Developer Guide. If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).
- You must create a custom solution and solution version before you create a batch inference job. However, you don't need to create an Amazon Personalize campaign. If you created a Domain dataset group, you can still create custom resources.
- Your input data must be formatted as described in Preparing input data for user segments (p. 270).
- You can't get batch recommendations with the Trending-Now recipe.
- If you use a filter with placeholder parameters, you must include the values for the parameters in your input data in a filterValues object. For more information, see Providing filter values in your input JSON (p. 327).
- We recommend that you use a different location for your output data (either a folder or a different Amazon S3 bucket) than your input data.

Batch workflow scoring

With User-Personalization and Personalized-Ranking recipes, Amazon Personalize calculates batch inference job recommendation scores as described in How User-Personalization recommendation scoring works (custom resources) (p. 245) and How personalized ranking scoring works (p. 255). You can view
scores in the batch inference job's output JSON file. Scores are only returned by models trained with the User-Personalization and Personalized-Ranking recipes.

Preparing input data for batch recommendations

A batch inference job imports your batch input JSON data from an Amazon S3 bucket, uses your custom solution version to generate recommendations, and then exports the item recommendations to an Amazon S3 bucket. Before you can get batch recommendations, you must prepare and upload your JSON file to an Amazon S3 bucket. We recommend that you create an output folder in your Amazon S3 bucket or use a separate output Amazon S3 bucket. You can then run multiple batch inference jobs using the same input data location.

If you use a filter with placeholder parameters, such as $GENRE, you must provide the values for the parameters in a filterValues object in your input JSON. For more information see Providing filter values in your input JSON (p. 327).

To prepare and import data

1. Format your batch input data depending on your recipe. You can't get batch recommendations with the Trending-Now recipe.
   - For USER_PERSONALIZATION recipes and the Popularity-Count recipe, your input data is a JSON file with a list of userIds
   - For RELATED_ITEMS recipes, your input data is a list of itemIds
   - For PERSONALIZED_RANKING recipes, your input data is a list of userIds, each paired with a collection of itemIds

   Separate each row with a new line. For input data examples, see Batch inference job input and output JSON examples (p. 261).

2. Upload your input JSON to an input folder in your Amazon S3 bucket. For more information, see Uploading files and folders by using drag and drop in the Amazon Simple Storage Service User Guide.

3. Create a separate location for your output data, either a folder or a different Amazon S3 bucket. By creating a separate location for the output JSON, you can run multiple batch inference jobs with the same input data location.

4. Create a batch inference job. Amazon Personalize outputs the recommendations from your solution version to your output data location.

Batch inference job input and output JSON examples

How you format your input data the recipe you use. If you use a filter with placeholder parameters, such as $GENRE, you must provide the values for the parameters in a filterValues object in your input JSON. For more information see Providing filter values in your input JSON (p. 327).

The following sections list correctly formatted JSON input and output examples for batch inference jobs. You can't get batch recommendations with the Trending-Now recipe.

User-Personalization and legacy HRNN recipes

Input

Separate each userId with a new line as follows.

```json
{"userId": "4638"}
{"userId": "663"}
{"userId": "3384"}
```
Amazon Personalize Developer Guide
Batch recommendations and user
segments (custom resources)
...

Output

{"input":{"userId":"4638"},"output":{"recommendedItems":
["63992","115149","110102","148626","148888","31685","102445","69526","92535","143355","62374","745
[0.0152238,0.0069081,0.0068222,0.006394,0.0059746,0.0055851,0.0049357,0.0044644,0.0042968,0.004015,
{"input":{"userId":"663"},"output":{"recommendedItems":
["368","377","25","780","1610","648","1270","6","165","1196","1097","300","1183","608","104","474",
[0.0406197,0.0372557,0.0254077,0.0151975,0.014991,0.0127175,0.0124547,0.0116712,0.0091098,0.0085492
{"input":{"userId":"3384"},"output":{"recommendedItems":
["597","21","223","2144","208","2424","594","595","920","104","520","367","2081","39","1035","2054"
[0.0241061,0.0119394,0.0118012,0.010662,0.0086972,0.0079428,0.0073218,0.0071438,0.0069602,0.0056961
...

POPULAR_ITEMS (Popularity-Count only)
Input
Separate each userId with a new line as follows.
{"userId": "12"}
{"userId": "105"}
{"userId": "41"}
...

Output
{"input": {"userId": "12"}, "output": {"recommendedItems": ["105", "106", "441"]}}
{"input": {"userId": "105"}, "output": {"recommendedItems": ["105", "106", "441"]}}
{"input": {"userId": "41"}, "output": {"recommendedItems": ["105", "106", "441"]}}
...

PERSONALIZED_RANKING recipes
Input
Separate each userId and list of itemIds to be ranked with a new line as follows.
{"userId": "891", "itemList": ["27", "886", "101"]}
{"userId": "445", "itemList": ["527", "55", "901"]}
{"userId": "71", "itemList": ["27", "351", "101"]}
...

Output
{"input":{"userId":"891","itemList":["27","886","101"]},"output":{"recommendedItems":
["27","101","886"],"scores":[0.48421,0.28133,0.23446]}}
{"input":{"userId":"445","itemList":["527","55","901"]},"output":{"recommendedItems":
["901","527","55"],"scores":[0.46972,0.31011,0.22017]}}
{"input":{"userId":"71","itemList":["29","351","199"]},"output":{"recommendedItems":
["351","29","199"],"scores":[0.68937,0.24829,0.06232]}}
...

RELATED_ITEMS recipes

262


Creating a batch inference job

Create a batch inference job to get batch item recommendations for users based on input data from Amazon S3. The input data can be a list of users or items (or both) in JSON format. You can create a batch inference job with the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), or AWS SDKs.

When you create a batch inference job, you specify the Amazon S3 paths to your input and output locations. Amazon S3 is prefix based. If you provide a prefix for the input data location, Amazon Personalize uses all files matching that prefix as input data. For example, if you provide s3://<name of your S3 bucket>/folderName and your bucket also has a folder with a path of s3://<name of your S3 bucket>/folderName_test, Amazon Personalize uses all files in both folders as input data. To use only the files within a specific folder as input data, end the Amazon S3 path with a prefix delimiter, such as /: s3://<name of your S3 bucket>/folderName/. For more information about how Amazon S3 organizes objects, see Organizing, listing, and working with your objects.

For more information about the batch workflow in Amazon Personalize, including permissions requirements, recommendation scoring, and preparing and importing input data, see Batch recommendations and user segments (custom resources) (p. 259).

Topics
- Creating a batch inference job (console) (p. 263)
- Creating a batch inference job (AWS CLI) (p. 264)
- Creating a batch inference job (AWS SDKs) (p. 265)

Creating a batch inference job (console)

After you have completed Preparing input data for batch recommendations (p. 261), you are ready to create a batch inference job. This procedure assumes that you have already created a solution and a solution version (trained model).

To create a batch inference job (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your dataset group.
3. Choose Batch inference jobs in the navigation pane, then choose Create batch inference job.
4. In Batch inference job details, in Batch inference job name, specify a name for your batch inference job.
5. For **IAM service role**, choose the IAM service role you created for Amazon Personalize during set up. This role must have read and write access to your input and output Amazon S3 buckets respectively.

6. For **Solution**, choose the solution and then choose the **Solution version ID** that you want to use to generate the recommendations.

7. For **Number of results**, optionally specify the number of recommendations for each line of input data. The default is 25.

8. For **Input data configuration**, specify the Amazon S3 path to your input file.

   Use the following syntax: 
   ```
   s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json
   ```

   Your input data must be in the correct format for the recipe your solution uses. For input data examples see [Batch inference job input and output JSON examples](p. 261).

9. For **Output data configuration**, specify the path to your output location. We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket).

   Use the following syntax: 
   ```
   s3://<name of your S3 bucket>/<output folder name>/
   ```

10. For **Filter configuration** optionally choose a filter to apply a filter to the batch recommendations. If your filter uses placeholder parameters, make sure the values for the parameters are included in your input JSON. For more information see [Providing filter values in your input JSON](p. 327).

11. For **Tags**, optionally add any tags. For more information about tagging Amazon Personalize resources, see [Tagging Amazon Personalize resources](p. 370).

12. Choose **Create batch inference job**. Batch inference job creation starts and the **Batch inference jobs** page appears with the **Batch inference job detail** section displayed.

13. When the batch inference job's status changes to **Active**, you can retrieve the job's output from the designated output Amazon S3 bucket. The output file's name will be of the format `input-name.out`.

---

**Creating a batch inference job (AWS CLI)**

After you have completed [Preparing input data for batch recommendations](p. 261), you are ready to create a batch inference job using the following `create-batch-inference-job` code. Specify a job name, replace **Solution version ARN** with the Amazon Resource Name (ARN) of your solution version, and replace the **IAM service role ARN** with the ARN of the IAM service role you created for Amazon Personalize during set up. This role must have read and write access to your input and output Amazon S3 buckets respectively. Optionally provide a filter ARN to filter recommendations. If your filter uses placeholder parameters, make sure the values for the parameters are included in your input JSON. For more information see [Filtering batch recommendations and user segments (custom resources)](p. 326).

Replace `S3 input path` and `S3 output path` with the Amazon S3 path to your input file and output locations. We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). Use the following syntax for input and output locations: `s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json` and `s3://<name of your S3 bucket>/<output folder name>/`.

The example includes optional User-Personalization recipe specific itemExplorationConfig hyperparameters: `explorationWeight` and `explorationItemAgeCutOff`. Optionally include `explorationWeight` and `explorationItemAgeCutOff` values to configure exploration. For more information, see [User-Personalization recipe](p. 116).

```bash
aws personalize create-batch-inference-job \
  --job-name Batch job name \
  --solution-version-arn Solution version ARN \
  --filter-arn Filter ARN \
```
Creating a batch inference job (AWS SDKs)

After you have completed Preparing input data for batch recommendations (p. 261), you are ready to create a batch inference job with the CreateBatchInferenceJob (p. 424) operation.

The following code shows how to create a batch inference job. Specify a job name, the Amazon Resource Name (ARN) of your solution version, and the ARN of the IAM service role you created for Amazon Personalize during set up. This role must have read and write access to your input and output Amazon S3 buckets.

We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). Use the following syntax for input and output locations: 

`s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json` and `s3://<name of your S3 bucket>/<output folder name>/`.

For `numResults`, specify the number of items you want Amazon Personalize to predict for each line of input data. Optionally provide a filter ARN to filter recommendations. If your filter uses placeholder parameters, make sure the values for the parameters are included in your input JSON. For more information see Filtering batch recommendations and user segments (custom resources) (p. 326).

SDK for Python (Boto3)

The example includes optional User-Personalization recipe specific `itemExplorationConfig` hyperparameters: `explorationWeight` and `explorationItemAgeCutOff`. Optionally include `explorationWeight` and `explorationItemAgeCutOff` values to configure exploration. For more information, see User-Personalization recipe (p. 116).

```python
import boto3

personalize_rec = boto3.client(service_name='personalize')

personalize_rec.create_batch_inference_job (  
solutionVersionArn = "Solution version ARN",  
jobName = "Batch job name",  
roleArn = "IAM service role ARN",  
filterArn = "Filter ARN",  
batchInferenceJobConfig = {  
    # optional USER_PERSONALIZATION recipe hyperparameters  
    "itemExplorationConfig": {  
        "explorationWeight": "0.3",  
        "explorationItemAgeCutOff": "30"  
    }  
},  
jobInput =  
{  
    "s3DataSource": {"path": "s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json"},  
    jobOutput =  
    {  
        "s3DataDestination": {"path": "s3://<name of your S3 bucket>/<output folder name>/"}  
    }  
})
```
SDK for Java 2.x

The example includes optional User-Personalization recipe specific itemExplorationConfig fields: explorationWeight and explorationItemAgeCutOff. Optionally include explorationWeight and explorationItemAgeCutOff values to configure exploration. For more information, see User-Personalization recipe (p. 116).

```java
class PersonalizeBatchInferenceJob {
    public static String createPersonalizeBatchInferenceJob(PersonalizeClient personalizeClient, String solutionVersionArn, String jobName, String filterArn, String s3InputDataSourcePath, String s3DataDestinationPath, String roleArn, String explorationWeight, String explorationItemAgeCutOff) {
        long waitInMilliseconds = 60 * 1000;
        String status;
        String batchInferenceJobArn;
        try {
            // Set up data input and output parameters.
            S3DataConfig inputSource = S3DataConfig.builder()
                .path(s3InputDataSourcePath)
                .build();
            S3DataConfig outputDestination = S3DataConfig.builder()
                .path(s3DataDestinationPath)
                .build();
            BatchInferenceJobInput jobInput = BatchInferenceJobInput.builder()
                .s3DataSource(inputSource)
                .build();
            BatchInferenceJobOutput jobOutputLocation = BatchInferenceJobOutput.builder()
                .s3DataDestination(outputDestination)
                .build();
            // Optional code to build the User-Personalization specific item exploration config.
            HashMap<String, String> explorationConfig = new HashMap<>();
            explorationConfig.put("explorationWeight", explorationWeight);
            explorationConfig.put("explorationItemAgeCutOff", explorationItemAgeCutOff);
            BatchInferenceJobConfig jobConfig = BatchInferenceJobConfig.builder()
                .itemExplorationConfig(explorationConfig)
                .build();
            // End optional User-Personalization recipe specific code.
            CreateBatchInferenceJobRequest createBatchInferenceJobRequest = CreateBatchInferenceJobRequest.builder()
                .solutionVersionArn(solutionVersionArn)
                .jobInput(jobInput)
                .jobOutput(jobOutputLocation)
                .jobName(jobName)
                .filterArn(filterArn)
                .roleArn(roleArn)
                .batchInferenceJobConfig(jobConfig) // Optional
                .build();
            batchInferenceJobArn = personalizeClient.createBatchInferenceJob(createBatchInferenceJobRequest)
                .batchInferenceJobArn();
        }
    }
}
```
DescribeBatchInferenceJobRequest describeBatchInferenceJobRequest =
DescribeBatchInferenceJobRequest.builder()
    .batchInferenceJobArn(batchInferenceJobArn)
    .build();

long maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
// wait until the batch inference job is complete.
while (Instant.now().getEpochSecond() < maxTime) {
    BatchInferenceJob batchInferenceJob = personalizeClient
        .describeBatchInferenceJob(describeBatchInferenceJobRequest)
        .batchInferenceJob();
    status = batchInferenceJob.status();
    System.out.println("Batch inference job status: " + status);
    if (status.equals("ACTIVE") || status.equals("CREATE FAILED")) {
        break;
    }
    try {
        Thread.sleep(waitInMilliseconds);
    } catch (InterruptedException e) {
        System.out.println(e.getMessage());
    }
}
return batchInferenceJobArn;
}

// Get service clients module and commands using ES6 syntax.
import { CreateBatchInferenceJobCommand } from
    "@aws-sdk/client-personalize";
import { personalizeClient } from ".libs/personalizeClients.js";

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the batch inference job's parameters.

export const createBatchInferenceJobParam = {
    jobName: 'JOB_NAME',
    jobInput: {
        s3DataSource: {
            path: 'INPUT_PATH', /* required */
        },
    },
    jobOutput: {
        s3DataDestination: {
            path: 'OUTPUT_PATH', /* required */
        },
    },
    roleArn: 'ROLE_AR', /* required */
    solutionVersionArn: 'SOLUTION_VERSION_AR', /* required */
    numResults: 20 /* optional integer*/
};
Amazon Personalize Developer Guide
Batch recommendations and user
segments (custom resources)
export const run = async () => {
try {
const response = await personalizeClient.send(new
CreateBatchInferenceJobCommand(createBatchInferenceJobParam));
console.log("Success", response);
return response; // For unit tests.
} catch (err) {
console.log("Error", err);
}
};
run();

Processing the batch job might take a while to complete. You can check a job's status by calling
DescribeBatchInferenceJob (p. 497) and passing a batchRecommendationsJobArn as the input
parameter. You can also list all Amazon Personalize batch inference jobs in your AWS environment by
calling ListBatchInferenceJobs (p. 534).

Batch inference job output examples
When you create a batch inference job, the job imports your batch input data from an Amazon S3
bucket, uses your solution version to generate item recommendations, and exports the recommendations
to an Amazon S3 bucket in JSON format.
The following sections list output ﬁle examples for batch inference jobs by recipe type. You can't get
batch recommendations with the Trending-Now recipe.
Topics
• USER_PERSONALIZATION recipes (p. 268)
• POPULAR_ITEMS recipes (p. 268)
• PERSONALIZED_RANKING recipes (p. 269)
• RELATED_ITEMS recipes (p. 269)

USER_PERSONALIZATION recipes
The following is an example of the output JSON ﬁle for a USER_PERSONALIZATION recipe.

{"input":{"userId":"4638"},"output":{"recommendedItems":
["63992","115149","110102","148626","148888","31685","102445","69526","92535","143355","62374","7451","
[0.0152238,0.0069081,0.0068222,0.006394,0.0059746,0.0055851,0.0049357,0.0044644,0.0042968,0.004015,0.00
{"input":{"userId":"663"},"output":{"recommendedItems":
["368","377","25","780","1610","648","1270","6","165","1196","1097","300","1183","608","104","474","736
[0.0406197,0.0372557,0.0254077,0.0151975,0.014991,0.0127175,0.0124547,0.0116712,0.0091098,0.0085492,0.0
{"input":{"userId":"3384"},"output":{"recommendedItems":
["597","21","223","2144","208","2424","594","595","920","104","520","367","2081","39","1035","2054","16
[0.0241061,0.0119394,0.0118012,0.010662,0.0086972,0.0079428,0.0073218,0.0071438,0.0069602,0.0056961,0.0
...

POPULAR_ITEMS recipes
The following example shows the format of the output JSON ﬁle for the Popularity-Count recipe. You
can't get batch recommendations with the Trending-Now recipe.
{"input": {"userId": "12"}, "output": {"recommendedItems": ["105", "106", "441"]}}
{"input": {"userId": "105"}, "output": {"recommendedItems": ["105", "106", "441"]}}
{"input": {"userId": "41"}, "output": {"recommendedItems": ["105", "106", "441"]}}
...

268


Amazon Personalize Developer Guide
Batch recommendations and user
segments (custom resources)

PERSONALIZED_RANKING recipes
The following example shows the format of the output JSON ﬁle for a PERSONALIZED_RANKING recipe.
{"input":{"userId":"891","itemList":["27","886","101"]},"output":{"recommendedItems":
["27","101","886"],"scores":[0.48421,0.28133,0.23446]}}
{"input":{"userId":"445","itemList":["527","55","901"]},"output":{"recommendedItems":
["901","527","55"],"scores":[0.46972,0.31011,0.22017]}}
{"input":{"userId":"71","itemList":["29","351","199"]},"output":{"recommendedItems":
["351","29","199"],"scores":[0.68937,0.24829,0.06232]}}
...

RELATED_ITEMS recipes
The following example shows the format of the output JSON ﬁle for a RELATED_ITEMS recipe.
{"input": {"itemId": "105"}, "output": {"recommendedItems": ["106", "107", "49"]}}
{"input": {"itemId": "106"}, "output": {"recommendedItems": ["105", "107", "49"]}}
{"input": {"itemId": "441"}, "output": {"recommendedItems": ["2", "442", "435"]}}
...

Getting user segments
To get user segments, you use a batch segment job. A batch segment job is a tool that imports
your batch input data from an Amazon S3 bucket and uses your solution version trained with a
USER_SEGMENTATION recipe to generate user segments for each row of input data.
Depending on the recipe, the input data is a list of items or item metadata attributes in JSON format.
For item attributes, your input data can include expressions to create user segments based on multiple
metadata attributes. A batch segment job exports user segments to an output Amazon S3 bucket. Each
user segment is sorted in descending order based on the probability that each user will interact with the
item in your input data.
When generating user segments, Amazon Personalize considers data in datasets from bulk and individual
imports:
• For bulk data, Amazon Personalize generates segments using only the bulk data present at the last full
solution version training. And it uses only bulk data that you imported with an import mode of FULL
(replacing existing data).
• For data from individual data import operations, Amazon Personalize generates user segments using
the data present at the last full solution version training. To have newer records impact user segments,
create a new solution version and then create a batch segment job.
Generating user segments works as follows:
1. Prepare and upload your input data in JSON format to an Amazon S3 bucket. The format of your input
data depends on the recipe you use and the job you are creating. See Preparing input data for user
segments (p. 270).
2. Create a separate location for your output data, either a diﬀerent folder or a diﬀerent Amazon S3
bucket.
3. Create a batch segment job. See Creating a batch segment job (p. 272).
4. When the batch segment job is complete, retrieve the user segments from your output location in
Amazon S3.
Topics

269


Guidelines requirements

The following are guidelines and requirements for batch getting batch segments:

- You must use a USER_SEGMENTATION recipe.
- Your Amazon Personalize IAM service role needs permission to read and add files to your Amazon S3 buckets. For information on granting permissions, see Service role policy for batch workflows (p. 17). For more information on bucket permissions, see User policy examples in the Amazon Simple Storage Service Developer Guide.

If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

- You must create a custom solution and solution version before you create a batch inference job. However, you don't need to create an Amazon Personalize campaign. If you created a Domain dataset group, you can still create custom resources.

- Your input data must be formatted as described in Preparing input data for user segments (p. 270).
- If you use a filter with placeholder parameters, you must include the values for the parameters in your input data in a filterValues object. For more information see Providing filter values in your input JSON (p. 327).
- We recommend that you use a different location for your output data (either a folder or a different Amazon S3 bucket) than your input data.

Preparing input data for user segments

Batch segment jobs use a solution version to make user segments based on data that you provide in an input JSON file. Before you can get user segments, you must prepare and upload your JSON file to an Amazon S3 bucket. We recommend that you create an output folder in your Amazon S3 bucket or use a separate output Amazon S3 bucket. You can then run multiple batch inference jobs using the same input data location.

If you use a filter with placeholder parameters, such as $GENRE, you must provide the values for the parameters in a filterValues object in your input JSON. For more information see Providing filter values in your input JSON (p. 327).

To prepare and import data

1. Format your batch input data depending on the recipe your solution uses. Separate input data element with a new line.

   Your input data is either a list of itemIds (Item-Affinity) or item attributes (Item-Attribute-Affinity). For item attributes, input data can include logical expressions with the AND operator to get users for multiple items or attributes per query. For input data examples, see Batch segment job input and output JSON examples (p. 271).

2. Upload your input JSON to an input folder in your Amazon S3 bucket. For more information, see Uploading files and folders by using drag and drop in the Amazon Simple Storage Service User Guide.

3. Create a separate location for your output data, either a folder or a different Amazon S3 bucket. By creating a separate location for the output JSON, you can run multiple batch segment jobs with the same input data location.
After you have prepared your input data and uploaded it to an Amazon S3 bucket, you are ready to generate user segments with a batch segment job. For more information, see Creating a batch segment job (p. 272).

**Batch segment job input and output JSON examples**

For a batch segment job, your input data must be either a list of itemIds (Item-Affinity recipe) or item attributes (Item-Attribute-Affinity). Each line of input data is a separate inference query. Each user segment is sorted in descending order based on the probability that each user will interact with items in your inventory.

For item attributes, you can mix different columns of metadata. For example one row might be a numerical column and the next might be a categorical column. Also, your input item metadata can include logical expressions with the AND operator to get a user segment for multiple attributes. For example, a line of your input data might be `{"itemAttributes": "ITEMS.genres = "Comedy" AND ITEMS.genres = "Action""}` or `{"itemAttributes": "ITEMS.genres = "Comedy" AND ITEMS.audience = "teen""}`. When you combine two attributes with the AND operator, you create a user segment with users who are more likely to interact with items that have both attributes based on the users interactions history. Unlike filter expressions (which use the IN operator for string equality), batch segment input expressions support only the = symbol for equality for string matching.

If you use a filter with placeholder parameters, such as $GENRE, you must provide the values for the parameters in a filterValues object in your input JSON. For more information see Providing filter values in your input JSON (p. 327).

The following are correctly formatted JSON input and output examples for batch segment jobs organized by recipe.

**Item-Affinity**

**Input**

Your input data can have a maximum of 500 items. Separate each itemId with a new line as follows.

```json
"itemId": "105"
"itemId": "106"
"itemId": "441"
...```

**Output**

```json
"input": {"itemId": "105"}, "output": {"recommendedUsers": ["106", "107", "49"]}
"input": {"itemId": "106"}, "output": {"recommendedUsers": ["105", "107", "49"]}
"input": {"itemId": "441"}, "output": {"recommendedUsers": ["2", "442", "435"]}
...```

**Item-Attribute-Affinity**

**Input**

Your input data can have a maximum of 10 queries, where each query is one or more item attributes. Separate each attribute or attribute expression with a new line as follows.

```json
{"itemAttributes": "ITEMS.genres = "Comedy" AND ITEMS.genres = "Action""}
{"itemAttributes": "ITEMS.genres = "Comedy""
{"itemAttributes": "ITEMS.genres = "Horror" AND ITEMS.genres = "Action""}
...```
Output

```json
{"itemAttributes": "ITEMS.genres = "Comedy" AND ITEMS.genres = "Action"", "output": 
{"recommendedUsers": ["25", "78", "108"]}
}
{"itemAttributes": "ITEMS.genres = "Adventure"", "output": 
{"recommendedUsers": ["87", "31", "129"]}
}
{"itemAttributes": "ITEMS.genres = "Horror" AND ITEMS.genres = "Action"", "output": 
{"recommendedUsers": ["8", "442", "435"]}
}
...
```

Creating a batch segment job

If you used a USER SEGMENTATION recipe, you can create batch segment jobs to get user segments with your solution version. Each user segment is sorted in descending order based on the probability that each user will interact with items in your inventory. Depending on the recipe, your input data must be a list of items (Item-Affinity recipe (p. 149)) or item attributes (Item-Attribute-Affinity recipe (p. 150)) in JSON format. You can create a batch segment job with the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), or AWS SDKs.

When you create a batch segment job, you specify the Amazon S3 paths to your input and output locations. Amazon S3 is prefix based. If you provide a prefix for the input data location, Amazon Personalize uses all files matching that prefix as input data. For example, if you provide `s3://<name of your S3 bucket>/folderName` and your bucket also has a folder with a path of `s3://<name of your S3 bucket>/folderName_test`, Amazon Personalize uses all files in both folders as input data. To use only the files within a specific folder as input data, end the Amazon S3 path with a prefix delimiter, such as `/: s3://<name of your S3 bucket>/folderName/`. For more information about how Amazon S3 organizes objects, see Organizing, listing, and working with your objects.

Topics

- [Creating a batch segment job (console) (p. 272)]
- [Creating a batch segment job (AWS CLI) (p. 273)]
- [Creating a batch segment job (AWS SDKs) (p. 274)]

Creating a batch segment job (console)

After you have completed [Preparing input data for batch recommendations (p. 261)](https://console.aws.amazon.com/personalize/home), you are ready to create a batch segment job. This procedure assumes that you have already created a solution and a solution version (trained model) with a USER SEGMENTATION recipe.

To get create a batch segment job (console)

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign in to your account.
2. On theDatasets group page, choose your dataset group.
3. Choose batch segment jobs in the navigation pane, then choose Create batch segment job.
4. In batch segment job details, for Batch segment job name, specify a name for your batch segment job.
5. For Solution, choose the solution and then choose the Solution version ID that you want to use to generate the recommendations. You can create batch segment jobs only if you used a USER SEGMENTATION recipe.
6. For Number of users, optionally specify the number of users Amazon Personalize generates for each user segment. The default is 25. The maximum is 5 million.
7. For Input source, specify the Amazon S3 path to your input file or use the Browse S3 to choose your Amazon S3 bucket.
Use the following syntax: `s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json`

Your input data must be in the correct format for the recipe your solution uses. For input data examples see [Batch segment job input and output JSON examples](p. 271).

8. For **Output destination**, specify the path to your output location or use the **Browse S3** to choose your Amazon S3 bucket. We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket).

   Use the following syntax: `s3://<name of your S3 bucket>/<output folder name>/`

9. For **IAM role**, choose one of the following:

   - Choose **Create and use new service role** and enter the **Service role name** to create a new role, or
   - If you’ve already created a role with the correct permissions, choose **Use an existing service role** and choose the IAM role.

   The role you use must have read and write access to your input and output Amazon S3 buckets respectively.

10. For **Filter configuration** optionally choose a filter to apply a filter to the user segments. If your filter uses placeholder parameters, make sure the values for the parameters are included in your input JSON. For more information see [Providing filter values in your input JSON](p. 327).

11. For **Tags**, optionally add any tags. For more information about tagging Amazon Personalize resources, see [Tagging Amazon Personalize resources](p. 370).

12. Choose **Create batch segment job**. Batch segment job creation starts and the **Batch segment jobs** page appears with the **Batch segment job detail** section displayed.

13. When the batch segment job’s status changes to **Active**, you can retrieve the job’s output from the designated output Amazon S3 bucket. The output file’s name will be of the format `input-name.out`.

### Creating a batch segment job (AWS CLI)

After you have completed [Preparing input data for batch recommendations](p. 261), you are ready to create a batch segment job using the following `create-batch-segment-job` code. Specify a job name, replace `Solution version ARN` with the Amazon Resource Name (ARN) of your solution version, and replace the IAM service role ARN with the ARN of the IAM service role you created for Amazon Personalize during set up. This role must have read and write access to your input and output Amazon S3 buckets respectively. For `num-results` specify the number of users you want Amazon Personalize to predict for each line of input data. The default is 25. The maximum is 5 million. Optionally provide a `filter-arn` to filter user segments. If your filter uses placeholder parameters, make sure the values for the parameters are included in your input JSON. For more information see [Filtering batch recommendations and user segments (custom resources)](p. 326).

Replace S3 input path and S3 output path with the Amazon S3 path to your input file and output locations. We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). Use the following syntax for input and output locations: `s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json` and `s3://<name of your S3 bucket>/<output folder name>/`.

```
aws personalize create-batch-segment-job \
  --job-name <Job name> \
  --solution-version-arn <Solution version ARN> \
  --num-results <The number of predicted users> \
  --filter-arn <Filter ARN> \
  --job-input s3DataSource={path=s3://<S3 input path>}
```
Batch recommendations and user segments (custom resources)

Creating a batch segment job (AWS SDKs)

After you have completed Preparing input data for batch recommendations (p. 261), you are ready to create a batch segment job with the CreateBatchSegmentJob operation. The following code shows how to create a batch segment job. Give the job a name, specify the Amazon Resource Name (ARN) of the solution version to use, specify the ARN for your Amazon Personalize IAM role, and specify the Amazon S3 path to your input file and output locations. Your IAM service role must have read and write access to your input and output Amazon S3 buckets respectively.

We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). Use the following syntax for input and output locations: s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json and s3://<name of your S3 bucket>/<output folder name>/.

For numResults, specify the number of users you want Amazon Personalize to predict for each line of input data. The default is 25. The maximum is 5 million. Optionally provide a filterArn to filter user segments. If your filter uses placeholder parameters, make sure the values for the parameters are included in your input JSON. For more information see Filtering batch recommendations and user segments (custom resources) (p. 326).

SDK for Python (Boto3)

```python
import boto3

personalize_rec = boto3.client(service_name='personalize')

personalize_rec.create_batch_segment_job(
    solutionVersionArn = "Solution version ARN",
    jobName = "Job name",
    numResults = Number of predicted users,
    filterArn = Filter ARN
    roleArn = "IAM service role ARN",
    jobInput =
        {"s3DataSource": {"path": "s3://<name of your S3 bucket>/<folder name>/<input JSON file name>.json"}},
    jobOutput =
        {"s3DataDestination": {"path": "s3://<name of your S3 bucket>/<output folder name>"}}
)
```

SDK for Java 2.x

```java
public static String createBatchSegmentJob(PersonalizeClient personalizeClient,
                                        String solutionVersionArn,
                                        String jobName,
                                        String filterArn,
                                        int numResults,
                                        String s3InputDataSourcePath,
                                        String s3DataDestinationPath,
                                        String roleArn,
                                        String explorationWeight,
                                        String
```
long waitInMilliseconds = 60 * 1000;
String status;
String batchSegmentJobArn;

try {
    // Set up data input and output parameters.
    S3DataConfig inputSource = S3DataConfig.builder()
        .path(s3InputDataSourcePath)
        .build();
    S3DataConfig outputDestination = S3DataConfig.builder()
        .path(s3DataDestinationPath)
        .build();

    BatchSegmentJobInput jobInput = BatchSegmentJobInput.builder()
        .s3DataSource(inputSource)
        .build();
    BatchSegmentJobOutput jobOutputLocation = BatchSegmentJobOutput.builder()
        .s3DataDestination(outputDestination)
        .build();

    CreateBatchSegmentJobRequest createBatchSegmentJobRequest =
        CreateBatchSegmentJobRequest.builder()
        .solutionVersionArn(solutionVersionArn)
        .filterArn(filterArn)
        .jobInput(jobInput)
        .jobOutput(jobOutputLocation)
        .jobName(jobName)
        .numResults(numResults)
        .roleArn(roleArn)
        .build();

    batchSegmentJobArn = personalizeClient.createBatchSegmentJob(createBatchSegmentJobRequest)
        .batchSegmentJobArn();

    DescribeBatchSegmentJobRequest describeBatchSegmentJobRequest =
        DescribeBatchSegmentJobRequest.builder()
        .batchSegmentJobArn(batchSegmentJobArn)
        .build();

    long maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;

    // wait until the batch segment job is complete.
    while (Instant.now().getEpochSecond() < maxTime) {
        BatchSegmentJob batchSegmentJob = personalizeClient
            .describeBatchSegmentJob(describeBatchSegmentJobRequest)
            .batchSegmentJob();

        status = batchSegmentJob.status();
        System.out.println("batch segment job status: " + status);

        if (status.equals("ACTIVE") || status.equals("CREATE FAILED")) {
            break;
        }
        try {
            Thread.sleep(waitInMilliseconds);
        } catch (InterruptedException e) {
            System.out.println(e.getMessage());
        }
    }

    return batchSegmentJobArn;
} catch (PersonalizeException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
}
SDK for JavaScript v3

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateBatchSegmentJobCommand } from
  "@aws-sdk/client-personalize";
import { personalizeClient } from ".libs/personalizeClients.js";

// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the batch segment job's parameters.
export const createBatchSegmentJobParam = {
  jobName: 'NAME',
  jobInput: {
    s3DataSource: {
      path: 'INPUT_PATH', /* required */
      // kmsKeyArn: 'INPUT_KMS_KEY_ARN' /* optional */
    }
  },
  jobOutput: {
    s3DataDestination: {
      path: 'OUTPUT_PATH', /* required */
      // kmsKeyArn: 'OUTPUT_KMS_KEY_ARN' /* optional */
    }
  },
  roleArn: 'ROLE_ARN', /* required */
  solutionVersionArn: 'SOLUTION_VERSION_ARN', /* required */
  numResults: 20 /* optional */
};

export const run = async () => {
  try {
    const response = await personalizeClient.send(new
      CreateBatchSegmentJobCommand(createBatchSegmentJobParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();
```

Processing the batch job might take a while to complete. You can check a job's status by calling `DescribeBatchSegmentJob` and passing a `batchSegmentJobArn` as the input parameter. You can also list all Amazon Personalize batch segment jobs in your AWS environment by calling `ListBatchSegmentJobs`.

### Batch segment job output examples

A batch segment job imports your batch input data from an Amazon S3 bucket, uses your solution version trained with a USER_SEGMENTATION recipe to generate user segments, and exports the segments to an Amazon S3 bucket.

The following sections list correctly formatted JSON output examples for batch segment jobs by recipe.

**Topics**
- Item-Affinity (p. 277)
• **Item-Attribute-Affinity (p. 277)**

**Item-Affinity**

The following example shows the format of the output JSON file for the Item-Affinity recipe.

```json
"input": {"itemId": "105"}, "output": {"recommendedUsers": ["106", "107", "49"]}
"input": {"itemId": "106"}, "output": {"recommendedUsers": ["105", "107", "49"]}
"input": {"itemId": "441"}, "output": {"recommendedUsers": ["2", "442", "435"]}
```

**Item-Attribute-Affinity**

The following example shows the format of the output JSON file for the Item-Attribute-Affinity recipe.

```json
"itemAttributes": "ITEMS.genres = "Comedy" AND ITEMS.genres = "Action"", "output":
{"recommendedUsers": ["25", "78", "108"]}
"itemAttributes": "ITEMS.genres = "Adventure"", "output": {"recommendedUsers": ["87", "31", "129"]}
"itemAttributes": "ITEMS.genres = "Horror" AND ITEMS.genres = "Action"", "output":
{"recommendedUsers": ["8", "442", "435"]}
```
Maintaining recommendation relevance

Relevant recommendations can increase user engagement, click-through rate, and conversion rate for your application as your catalogue grows. To maintain and improve the relevance of Amazon Personalize recommendations for your users, keep your data and custom resources up to date. This allows Amazon Personalize to learn from your user's most recent behavior and include your newest items in recommendations.

**Topics**
- Keeping datasets current (p. 278)
- Maintaining domain recommenders (p. 278)
- Maintaining custom solutions (p. 279)

Keeping datasets current

As your catalog grows, update your historical data with bulk or individual data import operations. For more information about importing historical data, see *Step 2: Preparing and importing data* (p. 159). For information on how data you import after training a model influences recommendations, see *How new data influences real-time recommendations* (p. 292).

For use cases and recipes that provide personalized real-time recommendations, keep your Interactions dataset up to date with your users' behavior. Do this by recording interactions with an event tracker and the PutEvents API operation. Amazon Personalize updates recommendations based on your user's most recent activity as they interact with your catalog. For information about real-time personalization, see *Real-time personalization* (p. 105). For more information on recording real-time events, see *Recording events* (p. 280).

Maintaining domain recommenders

Amazon Personalize automatically retrains the models backing your recommenders every 7 days. This is a full retraining that creates entirely new models based on the entirety of the data in your datasets. If you modify the columns used in training, Amazon Personalize automatically starts a full retraining of the models backing your recommender.

- For *Top picks for you* and *Recommended for you* use cases, Amazon Personalize updates your recommender to consider new items for recommendations. Automatic updates are not a full retraining where the model learns from your users' behavior. Instead, automatic updates allow Amazon Personalize to feature your new items in recommendations before the recommender's next full retraining. For information about automatic updates, see *Automatic updates* (p. 106).
- If you use the *Trending now* use case, Amazon Personalize automatically evaluates your interactions data every two hours and identifies trending items. You don't have to wait for your recommender to retrain.

While recommender retraining is in progress, you can still get recommendations from the recommender. Until the retraining completes, the recommender uses the previous configuration and models. To track
Maintaining custom solutions

Maintain your custom solutions by retraining regularly. Create a new solution version (retrain the model) to include new items in recommendations and update the model with your user’s most recent behavior.

Your retraining frequency depends on your business requirements and the recipe that you use. For all recipes, we recommend creating a new solution version at least weekly. This creates a completely new model based on the entirety of the training data from the datasets in your dataset group. For User-Personalization, you must set trainingMode to FULL for a full retraining.

If you add new items frequently, you might need to retrain more frequently depending on your recipe:

- If you don’t use User-Personalization or Trending-Now, you must create a new solution version for Amazon Personalize to consider the new items for recommendations.
- If you use User-Personalization, Amazon Personalize automatically updates your latest fully trained solution version to consider new items for recommendations.

Automatic updates are not a full retraining where the model learns from your users’ behavior. Instead, automatic updates allow Amazon Personalize to feature your new items in recommendations before the next full retraining.

You should still train a new solution version weekly with trainingMode set to FULL. If every two hours is not frequent enough, you can manually create a solution version with trainingMode set to UPDATE to consider those new items for recommendations. Just remember that Amazon Personalize automatically updates only your latest fully trained solution version. The manually updated solution version won’t be automatically updated in the future.

For more information about auto updates, including additional guidelines and requirements, see Automatic updates (p. 106).

- If you use Trending-Now, Amazon Personalize automatically identifies the top trending items in your interactions data over a configurable interval of time. You don’t have to manually create a new solution version for Trending-Now to consider new items from bulk or incremental interactions since the last training. For more information, see Trending-Now recipe (p. 136).

For information on creating a new solution version, see Creating a solution version (p. 226). After you create a new solution version, you must update the campaign to deploy it. For more information, see Updating a campaign (p. 241).

You can automate and schedule re-training and data import tasks with Maintaining Personalized Experiences with Machine Learning, an AWS Solutions Implementation that automates the Amazon Personalize workflow, including data import, solution version training, and batch workflows. For more information see Maintaining Personalized Experiences with Machine Learning.
Recording events

With both domain recommenders and custom resources, Amazon Personalize can make recommendations based on real-time event data only, historical event data only, or a mixture of both.

Record real-time events to build out your interactions data and allow Amazon Personalize to learn from your user's most recent activity. This keeps your data fresh and improves the relevance of Amazon Personalize recommendations. If your domain use case or recipe supports real-time personalization (p. 105), Amazon Personalize uses events in real time to update and adapt recommendations according to a user's evolving interest.

You can record real-time events using the AWS SDKs, AWS Amplify or AWS Command Line Interface (AWS CLI). When you record events, Amazon Personalize appends the event data to the Interactions dataset in your dataset group. If you record two events with exactly the same timestamp and identical properties, Amazon Personalize keeps only one of the events.

If you use Apache Kafka, you can use the Kafka connector for Amazon Personalize to stream data in real time to Amazon Personalize. For information see Kafka Connector for Amazon Personalize in the personalize-kafka-connector Github repository.

AWS Amplify includes a JavaScript library for recording events from web client applications, and a library for recording events in server code. For more information, see Amplify - analytics.

Requirements for recording events and training a model

To record events, you need the following:

- A dataset group that includes an Interactions dataset, which can be empty. If you went through the Getting started (p. 24) guide, you can use the same dataset group and dataset that you created. For information on creating a dataset group and a dataset, see Step 2: Preparing and importing data (p. 159).
- An event tracker.
- A call to the PutEvents (p. 598) operation.

You can start out with an empty Interactions dataset and, when you have recorded enough data, train the model using only new recorded events. For all use cases (Domain dataset groups) and recipes (Custom dataset groups), your interactions data must have the following before training:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
• At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

How real-time events influence recommendations

If your recipe supports real-time personalization, after you create a recommender or custom campaign, Amazon Personalize uses new recorded event data for existing items within seconds of import. The following use cases and recipes support real-time personalization:

• Recommended for you (ECOMMERCE use case) (p. 112)
• Top picks for you (VIDEO_ON_DEMAND use case) (p. 109)
• User-Personalization recipe (p. 116)
• Personalized-Ranking recipe (p. 139)

If you use the Trending-Now recipe, Amazon Personalize automatically considers items from new events data over configurable intervals. You don't have to create a new solution version. For more information, see Trending-Now recipe (p. 136).

If the item or user in the event is new, how the Amazon Personalize uses the data depends on your recipe. For more information, see How new data influences real-time recommendations (p. 292).

Creating an event tracker

Before you can record events, you must create an event tracker. An event tracker directs new event data to the Interactions dataset in your dataset group.

You create an event tracker with the Amazon Personalize console or the CreateEventTracker (p. 452) API operation. You pass as a parameter the Amazon Resource Name (ARN) of the dataset group that contains the target Interactions dataset. For instructions on creating an event tracker using the Amazon Personalize console, see Creating an event tracker (console) (p. 183).

An event tracker includes a tracking ID, which you pass as a parameter when you use the PutEvents operation. Amazon Personalize then appends the new event data to the Interactions dataset of the dataset group you specify in your event tracker.

Note
You can create only one event tracker for a dataset group.

Python

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.create_event_tracker(
    name='MovieClickTracker',
    datasetGroupArn='arn:aws:personalize:us-west-2:acct-id:dataset-group/MovieClickGroup'
)

print(response['eventTrackerArn'])
print(response['trackingId'])
```
Creating an event tracker

The event tracker ARN and tracking ID display, for example:

```json
{
    "eventTrackerArn": "arn:aws:personalize:us-west-2:acct-id:event-tracker/MovieClickTracker",
    "trackingId": "xxxxxxxx-xxxx-xxxx-xxxx-xxxxxxxxxxxx"
}
```

AWS CLI

```bash
aws personalize create-event-tracker \
    --name MovieClickTracker \
    --dataset-group-arn arn:aws:personalize:us-west-2:acct-id:dataset-group/MovieClickGroup
```

The event tracker ARN and tracking ID display, for example:

```json
{
    "eventTrackerArn": "arn:aws:personalize:us-west-2:acct-id:event-tracker/MovieClickTracker",
    "trackingId": "xxxxxxxx-xxxx-xxxx-xxxx-xxxxxxxxxxxx"
}
```

SDK for Java 2.x

```java
public static String createEventTracker(PersonalizeClient personalizeClient, 
    String eventTrackerName, 
    String datasetGroupArn) {

    String eventTrackerId = null;
    String eventTrackerArn = null;
    long maxTime = 3L * 60L * 60L;
    long waitInMilliseconds = 30L * 1000L;
    String status;

    try {
        CreateEventTrackerRequest createEventTrackerRequest = 
            CreateEventTrackerRequest.builder().
                name(eventTrackerName).datasetGroupArn(datasetGroupArn).build();

        CreateEventTrackerResponse createEventTrackerResponse = 
            personalizeClient.createEventTracker(createEventTrackerRequest);

        eventTrackerArn = createEventTrackerResponse.eventTrackerArn();
        eventTrackerId = createEventTrackerResponse.trackingId();

        System.out.println("Event tracker ARN: " + eventTrackerArn);
        System.out.println("Event tracker ID: " + eventTrackerId);

        maxTime = Instant.now().getEpochSecond() + maxTime;

        DescribeEventTrackerRequest describeRequest = 
            DescribeEventTrackerRequest.builder().
                eventTrackerArn(eventTrackerArn).build();

        while (Instant.now().getEpochSecond() < maxTime) {
            status = 
                personalizeClient.describeEventTracker(describeRequest).eventTracker().status();
```

282
Recording events with the PutEvents operation

After you create an Interactions dataset and an event tracker (p. 281) for your dataset group, you are ready to record events. To record events, you call the PutEvents (p. 598) operation. The following sections show how to record a single event, how to record multiple events with event value data, how to record events for anonymous users, and how to include impressions data with an event.

Topics

- Recording a single event (p. 283)
- Recording multiple events with event value data (p. 285)
- Recording events for anonymous users (p. 287)
- Recording impressions data (p. 288)

Recording a single event

The following example shows a PutEvents operation that passes one event. The corresponding Interactions schema is shown, along with an example row from the Interactions dataset.

Your application generates a unique sessionId when a user first visits your website or uses your application. You must use the same sessionId in all events throughout the session. Amazon Personalize uses the sessionId to associate events with the user before they log in (is anonymous). For more information see Recording events for anonymous users (p. 287).

The event list is an array of Event (p. 726) objects. An eventType is required for each event, but in this example, eventType data is not used in training because it is not included in the schema. You can provide a placeholder value to satisfy the requirement.

The trackingId comes from the event tracker you created in Creating an event tracker (p. 281). The userId, itemId, and sentAt parameters map to the USER_ID, ITEM_ID, and TIMESTAMP fields of a corresponding historical Interactions dataset. For more information, see Schemas (p. 82).

Corresponding interactions schema

| Interactions schema: USER_ID, ITEM_ID, TIMESTAMP |
| Interactions dataset: user123, item-xyz, 1543631760 |
Code example

Python

```python
import boto3

personalize_events = boto3.client(service_name='personalize-events')

personalize_events.put_events(
    trackingId = 'tracking_id',
    userId= 'USER_ID',
    sessionId = 'session_id',
    eventList = [{
        'sentAt': TIMESTAMP,
        'eventType': 'eventTypePlaceholder',
        'itemId': 'ITEM_ID'
    }]
)
```

AWS CLI

```bash
aws personalize-events put-events \
    --tracking-id tracking_id \
    --user-id USER_ID \
    --session-id session_id \
    --event-list '[[
        "sentAt": "TIMESTAMP",
        "eventType": "eventTypePlaceholder",
        "itemId": "ITEM_ID"
    ]]' 
```

SDK for Java 2.x

```java
public static void putEvents(PersonalizeEventsClient personalizeEventsClient,
    String trackingId,
    String sessionId,
    String userId,
    String itemId) {

    try {
        Event event = Event.builder()
            .sentAt(Instant.ofEpochMilli(System.currentTimeMillis() + 10 * 60 * 1000))
            .itemId(itemId)
            .eventType("typePlaceholder")
            .build();

        PutEventsRequest putEventsRequest = PutEventsRequest.builder()
            .trackingId(trackingId)
            .userId(userId)
            .sessionID(sessionId)
            .eventList(event)
            .build();

        int responseCode = personalizeEventsClient.putEvents(putEventsRequest)
            .sdkHttpResponse()
            .statusCode();
        System.out.println("Response code: " + responseCode);
    } catch (PersonalizeEventsException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    }
}
```
After this example, you would proceed to train a model using only the required properties.

### Recording multiple events with event value data

The following example shows how to record multiple events with different event types and different event values.

When you configure a solution, if your Interactions dataset includes EVENT_TYPE and EVENT_VALUE fields, you can set a specific value as a threshold to exclude records from training. For more information, see [Choosing the interactions data used for training](p. 223).

The example also shows the recording of an extra property, numRatings, that is used as metadata by certain recipes.

**Interactions schema:** USER_ID, ITEM_ID, TIMESTAMP, EVENT_TYPE, EVENT_VALUE, NUM_RATINGS

**Interactions dataset:**
- user123, movie_xyz, 1543531139, rating, 5, 12
- user321, choc-ghana, 1543531760, like, 4
- user111, choc-fake, 1543557118, like, 3

**Python**

```python
import boto3
import json

personalize_events = boto3.client(service_name='personalize-events')

personalize_events.put_events(
    trackingId = 'tracking_id',
    userId= 'user555',
    sessionId = 'session1',
    eventList = [{
        'eventId': 'event1',
        'sentAt': '1553631760',
        'eventType': 'like',
        'properties': json.dumps({
            'itemId': 'choc-panama',
            'eventValue': 4,
            'numRatings': 0
        })
    }, {
        'eventId': 'event2',
        'sentAt': '1553631782',
        'eventType': 'rating',
        'properties': json.dumps({
            'itemId': 'movie_ten',
            'eventValue': 3,
            'numRatings': 13
        })
    }]
)
```

**AWS CLI**

```
aws personalize-events put-events \
  --tracking-id tracking_id \
  --user-id user555 \
  --session-id session1 \
  --event-list '[{
    "eventId": "event1",
    "sentAt": "1553631760",
    "eventType": "like",
```
Recording multiple events with event value data

```
"properties": "{"itemId": "choc-panama", "eventValue": "true"}"
}, {
  "eventId": "event2",
  "sentAt": "1553631782",
  "eventType": "rating",
  "properties": "{"itemId": "movie_ten", "eventValue": "4", "numRatings": 13}"
}
```

SDK for Java 2.x

```java
public static void putMultipleEvents(PersonalizeEventsClient personalizeEventsClient,
                                      String trackingId,
                                      String sessionId,
                                      String userId,
                                      String event1Type,
                                      Float event1Value,
                                      String event1ItemId,
                                      int event1NumRatings,
                                      String event2Type,
                                      Float event2Value,
                                      String event2ItemId,
                                      int event2NumRatings) {

  ArrayList<Event> eventList = new ArrayList<Event>();

  try {
    Event event1 = Event.builder()
        .eventType(event1Type)
        .sentAt(Instant.ofEpochMilli(System.currentTimeMillis() + 10 * 60 * 1000))
        .itemId(event1ItemId)
        .eventValue(event1Value)
        .properties("{"numRatings": "+ event1NumRatings +"}")
        .build();

    eventList.add(event1);

    Event event2 = Event.builder()
        .eventType(event2Type)
        .sentAt(Instant.ofEpochMilli(System.currentTimeMillis() + 10 * 60 * 1000))
        .itemId(event2ItemId)
        .eventValue(event2Value)
        .properties("{"numRatings": "+ event2NumRatings +"}")
        .build();

    eventList.add(event2);

    PutEventsRequest putEventsRequest = PutEventsRequest.builder()
        .trackingId(trackingId)
        .userId(userId)
        .sessionId(sessionId)
        .eventList(eventList)
        .build();

    int responseCode = personalizeEventsClient.putEvents(putEventsRequest)
        .sdkHttpResponse()
        .statusCode();

    System.out.println("Response code: " + responseCode);
  } catch (PersonalizeEventsException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
  }
}
```
Note
The properties keys use camel case names that match the fields in the Interactions schema. For example, if the field ‘NUM_RATINGS’ is defined in the Interactions schema, the property key should be numRatings.

Recording events for anonymous users

Important
If you don't record at minimum one event with a sessionId and userId for a user, Amazon Personalize won't use the activity tracked to only the sessionId when training. And after training completes, recommendations will no longer be based on activity tracked to the sessionId.

You can record events for users before they create an account. Record events for anonymous users to build a continuous event history with events from before and after they log in. This provides Amazon Personalize more interactions data about the user, which can help generate more relevant recommendations.

To record events for anonymous users (users that haven't logged in), for each event specify only a sessionId. Your application generates a unique sessionId when a user first visits your website or uses your application. You must use the same sessionId in all events throughout the session. Amazon Personalize uses the sessionId to associate events with the user before they log in.

Amazon Personalize doesn't use events from anonymous users when training until you associate them with a userId. For more information, see Building a continuous event history for anonymous users (p. 287).

To provide real-time personalization (p. 105) for anonymous users, specify the sessionId as the userId in your GetRecommendations (p. 608) request.

Topics
- Building a continuous event history for anonymous users (p. 287)
- Code sample (p. 287)

Building a continuous event history for anonymous users

To build an event history for an anonymous user and have Amazon Personalize use their events when training, record at minimum one event with both a sessionId and a userId. Then you can record any number of events for the userId. After you start providing a userId, the sessionId can change. During the next full retraining, Amazon Personalize associates the userId with the anonymous user history tracked to the original sessionId.

After retraining completes, recommendations will be based on activity tracked to both the sessionId from the anonymous events and any events tracked to their userId.

Note
If your user doesn't create an account and you want Amazon Personalize to use the data when training, you can use the sessionId as the userId in events. Just remember that if the user eventually creates an account, you won't be able to associate the events from their anonymous browsing with their new userId.

Code sample

The following code shows how to record a single event for an anonymous user. Amazon Personalize tracks event to the sessionId until you specify a userId along with the same sessionId in an event.

```python
import boto3
```
Recording impressions data

If you use the User-Personalization recipe or add the IMPRESSIONS field to your schema for a dataset in a Domain dataset group, you can record impressions data in your PutEvents operation.

Impressions are lists of items that were visible to a user when they interacted with (for example, clicked or watched) a particular item. Amazon Personalize uses impressions data to guide exploration, where recommendations include items with less interactions data or relevance. For information on the implicit and explicit impressions Amazon Personalize can model, see Impressions data.

Important

If you provide conflicting implicit and explicit impression data in your PutEvents requests, Amazon Personalize uses the explicit impressions by default.

To record the Amazon Personalize recommendations you show your user as impressions data, include the recommendationId in your PutEvents request and Amazon Personalize derives the implicit impressions based on your recommendation data.

To manually record impressions data for an event, list the impressions in the PutEvents command's impression input parameter. The following code sample shows how to include a recommendationId and an impression in a PutEvents operation with either the SDK for Python (Boto3) or the SDK for Java 2.x. If you include both, Amazon Personalize uses the explicit impressions by default.

SDK for Python (Boto3)

```python
import boto3

personalize_events = boto3.client(service_name='personalize-events')

personalize_events.put_events(
    trackingId = 'tracking_id',
    userId = 'userId',
    sessionId = 'sessionId',
    eventList = [{
        'eventId': 'event1',
        'eventType': 'rating',
        'sentAt': 1553631760,
        'itemId': 'item_id',
        'recommendationId': 'recommendation id',
        'impression': ['itemId1', 'itemId2', 'itemId3']
    }]
)
```

SDK for Java 2.x

Use the following putEvents method to record an event with impressions data and a recommendationId. For the impressions parameter, pass the list of itemIds as an ArrayList.
public static void putEvents(PersonalizeEventsClient personalizeEventsClient,
        String trackingId,
        String sessionId,
        String userId,
        String eventType,
        Float eventValue,
        String itemId,
        ArrayList<String> impressions,
        String recommendationId) {

    try {
        Event event = Event.builder()
                .eventType(eventType)
                .sentAt(Instant.ofEpochMilli(System.currentTimeMillis() + 10 * 60 * 1000))
                .itemId(itemId)
                .eventValue(eventValue)
                .impression(impressions)
                .recommendationId(recommendationId)
                .build();

        PutEventsRequest putEventsRequest = PutEventsRequest.builder()
                .trackingId(trackingId)
                .userId(userId)
                .sessionId(sessionId)
                .eventList(event)
                .build();

        int responseCode = personalizeEventsClient.putEvents(putEventsRequest)
                .sdkHttpResponse()
                .statusCode();
        System.out.println("Response code: " + responseCode);
    } catch (PersonalizeEventsException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    }
}

Event metrics and attribution reports

To monitor the type and number of events sent to Amazon Personalize, use Amazon CloudWatch metrics. For more information, see Monitoring Amazon Personalize (p. 404).

To generate CloudWatch reports that show the impact of recommendations, create a metric attribution and record user interactions with real-time recommendations. For information on creating a metric attribution, see Measuring impact of recommendations (p. 329).

For each event, include recommendation ID of the recommendations you showed the user. Or include the event source, such as a third party. Import this data to compare different campaigns, recommenders, and third parties. You can import at most 100 event attribution sources.

- If you provide a recommendationId, Amazon Personalize automatically determines the source campaign or recommender and identifies it in reports in an EVENT_ATTRIBUTION_SOURCE column.
- If you provide both attributes, Amazon Personalize uses only the eventAttributionSource.
- If you don’t provide a source, Amazon Personalize labels the source SOURCE_NAME_UNDEFINED in reports.

The following code shows how to provide an eventAttributionSource for an event in a PutEvents operation.
response = personalize_events.put_events(  
    trackingId = 'eventTrackerId',  
    userId= 'userId',  
    sessionId = 'sessionId123',  
    eventList = [
        {'eventId': 'event1',  
         'eventType': 'watch',  
         'sentAt': '1667260945',  
         'itemId': '123',  
         'metricAttribution': {
             'eventAttributionSource': 'thirdPartyServiceXYZ'
         }
    ]
)  
statusCode = response['ResponseMetadata']['HTTPStatusCode']  
print(statusCode)

response = personalize_events.put_events(  
    trackingId = 'eventTrackerId',  
    userId= 'userId',  
    sessionId = 'sessionId123',  
    eventList = [
        {'eventId': 'event1',  
         'eventType': 'watch',  
         'sentAt': '1667260945',  
         'itemId': '123',  
         'recommendationId': 'RID-12345678-1234-1234-1234-abcdefghijkl'
    ]
)  
statusCode = response['ResponseMetadata']['HTTPStatusCode']  
print(statusCode)

The following code shows how to provide a recommendationId for an event in a PutEvents operation.

### Third-party event tracking services

The following Customer Data Platforms (CDPs) can help you collect event data from your application and send it to Amazon Personalize.

- **Amplitude** – You can use Amplitude to track user actions to help you understand your users' behavior. For information on using Amplitude and Amazon Personalize, see the following AWS Partner Network (APN) blog post: [Measuring the Effectiveness of Personalization with Amplitude and Amazon Personalize](#).

- **mParticle** – You can use mParticle to collect event data from your app. For an example that shows how to use mParticle and Amazon Personalize to implement personalized product recommendations, see [How to harness the power of a CDP for machine learning: Part 2](#).

- **Segment** – You can use Segment to send your data to Amazon Personalize. For more information on integrating Segment with Amazon Personalize, see [Amazon Personalize Destination](#).

### Sample implementations

For a sample Jupyter notebook that shows how to use Amazon Personalize to react to real-time behavior of users using an event tracker and the **PutEvents** operation, see [2.View_Campaign_And_Interactions.ipynb](#) in the getting_started folder of the [amazon-personalize-samples](#) GitHub repository.
For a simple example that shows how to stream events from users interacting with recommendations, see `streaming_events` in the Amazon Personalize samples GitHub repository.

For a complete example that contains the source code and supporting files to deploy real-time APIs that sit between your Amazon Personalize resources and client applications, see `Real-Time Personalization APIs` in the AWS samples GitHub repository. This project includes how to implement the following:

- User context and user event collection
- Response caching
- Decorating recommendations based on item metadata
- A/B testing
- API authentication
Managing data

After you create an Amazon Personalize dataset, you can replace its schema to add columns of data. After you import data into a dataset, you can analyze it, export it to an Amazon S3 bucket, update it, or delete it by deleting the dataset.

**Topics**
- Updating data (p. 292)
- Replacing a dataset's schema (p. 298)
- Analyzing data in datasets (p. 300)
- Exporting a dataset (p. 303)
- Deleting data (p. 309)

Updating data

After you import data into an Amazon Personalize dataset, you can update it with bulk or individual data import operations. You can't make changes to the schema for the dataset. If you need to make changes to your schema, such as adding or removing an attribute, you must delete your dataset. Then you must create a new dataset with a new schema and import data again.

**Topics**
- How new data influences real-time recommendations (p. 292)
- Updating existing bulk data (p. 294)
- Updating data with individual import operations (p. 298)

How new data influences real-time recommendations

If you have already created a recommender or custom solution version, how new data influences real-time recommendations depends on whether the item or user is new. And it depends on the domain use case or custom recipe you use.

For information on how new records influence batch recommendations, see Getting batch recommendations (p. 259). For information on how new records influence batch segment jobs, see Getting user segments (p. 269).

**Topics**
- New interactions (p. 292)
- New items (p. 293)
- New users (p. 293)

New interactions

New interactions are events that you import after creating a recommender or its latest automatic full retraining. Or they are interactions that you import since the latest full solution version training. For
both real-time interactions and bulk interactions data, if interactions involve a new item, when Amazon Personalize considers the new item for recommendations depends on your use case or recipe. For more information, see New items (p. 293).

Real-time events

For use cases and recipes that feature real-time personalization, Amazon Personalize immediately uses real-time interactions between a user and existing items (items present at the latest full training) when generating recommendations for the same user. For more information about real-time personalization, see Real-time personalization (p. 105).

For any domain use cases and custom recipes that don't feature real-time personalization, such as recommending similar items, your model learns from real-time interactions data only after the next full retraining. This can be when your domain recommender's weekly automatic retraining completes. Or it can be after you create a new solution version.

Bulk interactions

For bulk interactions, for both incremental and full dataset import jobs, your model learns from bulk interactions data only after the next full retraining. Bulk data isn't used to update recommendations for real-time personalization.

A full retraining can be when your domain recommender's weekly automatic retraining completes. Or it can be after you create a new solution version. For User-Personalization, you must set trainingMode to FULL.

For more information about updating existing bulk data, see Updating existing bulk data (p. 294).

New items

New items are items that you import after creating a recommender or its latest automatic full retraining. Or they are items that you import since the latest full solution version training. They can come from either interactions data or item metadata in an Items dataset.

New items are considered for recommendations as follows:

- For Top picks for you and Recommended for you domain cases or User-Personalization, Amazon Personalize automatically updates the model every two hours. After each update, Amazon Personalize will consider new items for recommendations as part of exploration. When considering the new item, Amazon Personalize considers any metadata for the item, but this data will have a greater effect on recommendations only after you record interactions for the item and fully retrain. For information about updates, see Automatic updates (p. 106).

- If you use the Trending now use case, Amazon Personalize automatically evaluates your interactions data every two hours and identifies trending items. You don't have to wait for your recommender to retrain. If you use the Trending-Now recipe, Amazon Personalize automatically considers all new items over configurable intervals. You don't have to manually create a new solution version. For information about configuring intervals, see Trending-Now recipe (p. 136).

- If you don't use the Trending-Now recipe or your use case or recipe doesn't support automatic updates, Amazon Personalize will consider new items only after the next full retraining. This can be after your domain recommender's weekly automatic retraining completes. Or it can be after you create a new solution version.

New users

New users are users that you import after creating a recommender or after its latest automatic full retraining. Or they are users that you import since the latest full solution version training. They can
Amazon Personalize generates recommendations for new users as follows:

- If you use the Trending now domain use case or Trending-Now custom recipe, new users immediately receive recommendations for the top trending items. If you use the Popularity-Count recipe, new users immediately receive recommendations for items with the most interactions.
- For recipes or use cases that provide personalized recommendations for users, recommendations for new users are based on the early interaction histories of your existing users. The first items these existing users interacted with are more likely to be recommended to new users. For the User-Personalization or Personalized-Ranking recipes, if you set recency_mask to true, recommendations also include items based on the latest popularity trends in your interactions data.

The following can increase recommendation relevance for new users:

- Interactions data – The primary way to improve recommendation relevance for a new user is to import data from their interactions with your items. For information about how new interactions data influences recommendations, see New interactions (p. 292).
- User metadata – Importing user metadata, such as GENDER or MEMBERSHIP_STATUS, can improve recommendations. For metadata to influence recommendations, you must wait for your domain recommender's weekly automatic retraining to complete. Or you must manually create a new solution version.
- Contextual metadata – If your use case or recipe supports contextual metadata and your Interactions dataset has metadata fields for contextual data, you can provide the user's context in your request for recommendations. This does not require retraining. For more information, see Increasing recommendation relevance with contextual metadata (p. 258).

**Updating existing bulk data**

If you previously created a dataset import job for a dataset, you can use another import job to add to or replace the existing bulk data. By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. You can instead append the new records to existing data by changing the job's import mode (p. 295). To append data to an Interactions dataset with a dataset import job, you must have at least 1000 new interaction records.

If you already created a recommender or deployed a custom solution version with a campaign, how new bulk records influence recommendations depends on the domain use case or recipe that you use. For more information, see How new data influences real-time recommendations (p. 292).

**Filter updates for bulk records**

Within 15 minutes of completing a bulk import, Amazon Personalize updates any filters you created in the dataset group with your new item and user data. This update allows Amazon Personalize to use the most recent data when filtering recommendations for your users.

**Topics**

- [Import modes](p. 295)
- [Updating bulk records (console)](p. 295)
- [Updating bulk records (AWS CLI)](p. 296)
- [Updating bulk records (AWS SDKs)](p. 296)
Import modes

To configure how Amazon Personalize adds your new records to your dataset, you specify an import mode for your dataset import job:

- To overwrite all existing bulk data in your dataset, choose Replace existing data in the Amazon Personalize console or specify FULL in the CreateDatasetImportJob (p. 448) API operation. This doesn't replace data you imported individually, including events recorded in real time.

- To append the records to the existing data in your dataset, choose Add to existing data or specify INCREMENTAL in the CreateDatasetImportJob API operation. Amazon Personalize replaces any record with the same ID with the new one.

To append data to an Interactions dataset with a dataset import job, you must have at minimum 1000 new interaction records.

If you haven't imported bulk records, the option is not available in the console and you can only specify FULL in the CreateDatasetImportJob API operation. The default is a full replacement.

Updating bulk records (console)

### Important

By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. You can change this by specifying the job's [import mode](p. 295).

To update bulk data with the Amazon Personalize console, create a dataset import job for the dataset and specify an import mode.

**To update bulk records (console)**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign in to your account.
2. On the Dataset groups page, choose your dataset group.
3. From the navigation pane, choose Datasets.
4. On the Datasets page, choose the dataset you want to update.
5. In Dataset import jobs, choose Create dataset import job.
6. In Import job details, for Dataset import job name, specify a name for your import job.
7. For Import mode, choose how to update the dataset. Choose either Replace existing data or Add to existing data. For more information see [Import modes](p. 295).
8. In Input source, for S3 Location, specify where your data file is stored in Amazon S3. Use the following syntax:

```
 s3://<name of your S3 bucket>/<folder path>/<CSV file name>
```

If your CSV files are in a folder in your S3 bucket and you want to upload multiple CSV files to a dataset with one dataset import job, use this syntax without the CSV file name.

9. In IAM role, choose to either create a new role or use an existing one. If you completed the prerequisites, choose Use an existing service role and specify the role that you created in [Creating an IAM role for Amazon Personalize](p. 15).
10. For Tags, optionally add any tags. For more information about tagging Amazon Personalize resources, see [Tagging Amazon Personalize resources](p. 370).
11. Choose Finish. The data import job starts and the Dataset overview page displayed. The dataset import is complete when the status is ACTIVE.
Updating existing bulk data

Updating bulk records (AWS CLI)

Important
By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. You can change this by specifying the job's import mode (p. 295).

To update bulk data, use the create-dataset-import-job command. For the import-mode, specify FULL to replace existing data or INCREMENTAL to add to it. For more information see Import modes (p. 295).

The following code shows how to create a dataset import job that incrementally imports data into a dataset.

```
aws personalize create-dataset-import-job \
--job-name dataset import job name \
--dataset-arn dataset arn \
--data-source dataLocation=s3://bucketname/filename \
--role-arn roleArn \
--import-mode INCREMENTAL
```

Updating bulk records (AWS SDKs)

Important
By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. You can change this by specifying the job's import mode (p. 295).

To update bulk data, create a dataset import job and specify an import mode. The following code show's how to update bulk data in Amazon Personalize with the SDK for Python (Boto3) or SDK for Java 2.x.

SDK for Python (Boto3)

To update bulk data, use the create_dataset_import_job method. For the import-mode, specify FULL to replace existing data or INCREMENTAL to add to it. For more information see Import modes (p. 295).

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.create_dataset_import_job(
    jobName = 'YourImportJob',
    datasetArn = 'dataset_arn',
    dataSource = {'dataLocation':'s3://bucket/file.csv'},
    roleArn = 'role_arn',
    importMode = 'INCREMENTAL'
)
```

SDK for Java 2.x

To update bulk data, use the following createPersonalizeDatasetImportJob method. For the importImport, specify ImportMode.FULL to replace existing data or ImportMode.INCREMENTAL to add to it. For more information see Import modes (p. 295).

```java
public static String createPersonalizeDatasetImportJob(PersonalizeClient personalizeClient,
                                          String jobName,
                                          String datasetArn,
                                          String s3BucketPath,
                                          String roleArn,
                                          ImportMode importMode) {
```
long waitInMilliseconds = 60 * 1000;
String status;
String datasetImportJobArn;

try {
    DataSource importDataSource = DataSource.builder()
        .dataLocation(s3BucketPath)
        .build();

    CreateDatasetImportJobRequest createDatasetImportJobRequest =
        CreateDatasetImportJobRequest.builder()
        .datasetArn(datasetArn)
        .dataSource(importDataSource)
        .jobName(jobName)
        .roleArn(roleArn)
        .importMode(importMode)
        .build();

    datasetImportJobArn =
        personalizeClient.createDatasetImportJob(createDatasetImportJobRequest)
        .datasetImportJobArn();

    DescribeDatasetImportJobRequest describeDatasetImportJobRequest =
        DescribeDatasetImportJobRequest.builder()
        .datasetImportJobArn(datasetImportJobArn)
        .build();

    long maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
    while (Instant.now().getEpochSecond() < maxTime) {
        DatasetImportJob datasetImportJob = personalizeClient
            .describeDatasetImportJob(describeDatasetImportJobRequest)
            .datasetImportJob();
        status = datasetImportJob.status();
        System.out.println("Dataset import job status: "+ status);
        if (status.equals("ACTIVE") || status.equals("CREATE FAILED")) {
            break;
        }
        try {
            Thread.sleep(waitInMilliseconds);
        } catch (InterruptedException e) {
            System.out.println(e.getMessage());
        }
    }
    return datasetImportJobArn;
} catch (PersonalizeException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
    return "";
}

SDK for JavaScript v3

// Get service clients and commands using ES6 syntax.
import { CreateDatasetImportJobCommand, PersonalizeClient } from "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});
Updating data with individual import operations

After you import data into an Amazon Personalize dataset, you can update it by importing additional individual records, including interaction events, users, or items. Importing data individually allows you to add small batches of records to your Amazon Personalize datasets as your catalog grows.

When you import records individually, Amazon Personalize appends the new records to the dataset. To update an individual item or user, you can import a record with the same ID but with the modified attributes. You can import up to 10 records per individual import operation.

For more information on importing records individually, see Importing individual records (p. 182).

Replacing a dataset's schema

After you create an Amazon Personalize dataset, you can replace its schema with a new or existing one. You might replace a dataset's schema if your data structure changed after you created the dataset. For example, you might have a new column of item metadata that you want Amazon Personalize to consider during training. Or you might want to add a column of data to use only when filtering recommendations.

When you replace a dataset's schema, you must keep all fields in the previous schema and you can't change their data types or attributes. After you replace a dataset's schema, Amazon Personalize automatically excludes any new columns from training for any existing recommenders or custom solutions. For more guidelines and requirements, see Guidelines and requirements (p. 299).

You can replace a dataset's schema with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), and AWS SDKs.

**Topics**

- Guidelines and requirements (p. 299)
- Replacing a dataset's schema (console) (p. 299)
- Replacing a dataset's schema (AWS CLI) (p. 300)
- Replacing a dataset's schema (AWS SDKs) (p. 300)
Guidelines and requirements

Before you replace the schema for a dataset, make sure that you're aware of the following guidelines and requirements:

- You can't replace the schema of an Interactions dataset.
- You can add new fields to your replacement schema, but you must keep all fields in the previous schema. And you can't change their data types or attributes. For example, if the previous schema includes a MEMBERSHIP_STATUS field for categorical string data, the new schema you use must include a MEMBERSHIP_STATUS field with these attributes and data types.
- If the current schema has a field that you want to rename, or if you want to change its data types or attributes, you can add a new field with a new name and modified types or attributes. Then include the new field in training and exclude the old field. Any new fields must support null data. If the old field did not support null data, when you import data, you can use placeholder data to make sure your import matches the schema. For information about configuring the columns used by a recommender, see Updating a recommender (p. 204). For information about configuring the columns used by a solution, see Configuring columns used when training (p. 214).
- Any new fields must support null data. For information about adding a null type to a field, see Schema data types (p. 83).
- After you replace a dataset's schema, Amazon Personalize automatically excludes any new columns from training for any existing recommenders or custom solutions. Using the modified dataset involves the following actions:
  - To use any new columns in training, import data that aligns with the new schema. Then update any recommenders to use any new columns, or create a new custom solution and configure the columns that it uses when training.
  
  For information about updating the columns used by a recommender, see Updating a recommender (p. 204). For information about configuring the columns used by a solution, see Configuring columns used when training (p. 214).
  - To use any columns only when filtering, import data that aligns with the new schema, create a filter that uses the new data, and apply your filter to your recommendation requests. You don't need to update any recommenders, or create or update any custom resources.

Replacing a dataset's schema (console)

To replace a dataset's schema with the Amazon Personalize console, you choose the dataset to modify and choose to replace with a new schema or use an existing one.

To replace a dataset's schema

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your dataset group.
3. Choose Datasets, and choose the radio button for the dataset that you want to modify.
4. Choose Actions, and choose Replace schema.
5. In Schema details, choose to replace with a new schema or a previously created one.
6. Specify the new schema to use. If you have chosen to:
   - Replace with a new schema, give the schema a name, and in Schema definition, make your changes to the schema JSON.
   - Use a previously created schema, for Previously created schema, choose the schema that you want to use. Only eligible schemas are listed. For information about schema requirements, see Guidelines and requirements (p. 299).
7. **Choose Replace.** When the dataset is active, you can start importing data that aligns with the new schema. For more information, see [Step 2: Preparing and importing data](p. 159).

### Replacing a dataset's schema (AWS CLI)

To replace a dataset's schema with the AWS CLI, you use the `update-dataset` command, specify the Amazon Resource Name (ARN) of the dataset to update and the ARN of the new schema to use. You can't replace the schema of an Interactions dataset.

The following code shows how to update a dataset's schema with the AWS CLI. To replace a dataset's schema with a new one, first use the `create-schema` command. Then use the following code to replace the current schema with the new one. For information about creating a schema with the AWS CLI, see [Creating a dataset and a schema (AWS CLI)](p. 172). For information about datasets and schema requirements, see [Schemas](p. 82).

```bash
aws personalize update-dataset \
--dataset-arn Dataset ARN \
--schema-arn New schema ARN
```

When the dataset is active, you can start importing data that aligns with the new schema. For more information, see [Step 2: Preparing and importing data](p. 159). For information about the latest update to the dataset, you can use the `DescribeDataset` operation.

### Replacing a dataset's schema (AWS SDKs)

To replace a dataset's schema with the AWS SDKs, you use the `UpdateDataset` API operation. Specify the Amazon Resource Name (ARN) of the dataset to update and the new schema to use. You can't update the schema of an Interactions dataset.

The following code shows how to replace a dataset's schema with the SDK for Python (Boto3). To replace a dataset's schema with a new one, first use the `CreateSchema` operation. Then use the following code to replace the current schema with the new one. For information about creating a schema with the AWS SDKs, see [Creating a dataset and a schema (AWS SDKs)](p. 173). For information on dataset and schema requirements, see [Schemas](p. 82).

```python
import boto3

personalize = boto3.client('personalize')

update_dataset_response = personalize.update_dataset(
    datasetArn = 'dataset_arn',
    schemaArn = 'new_schema_arn'
)

print(update_dataset_response)
```

When the dataset is active, you can start importing data that aligns with the new schema. For more information, see [Step 2: Preparing and importing data](p. 159). For information about the latest update to the dataset, you can use the `DescribeDataset` operation.

### Analyzing data in datasets

After you import data into Amazon Personalize datasets, you can use the Amazon Personalize console to analyze the data. You can learn about your data through data insights and column and row statistics. And you can learn what actions you can take to improve your data. These actions can help you meet
Amazon Personalize resource requirements, such as model training requirements, or they can lead to improved recommendations.

After you make any recommended changes, you can import your data again and see if you resolved any issues or improved dataset statistics. For information on updating data, see Updating data (p. 292).

If you don’t see any insights, your data aligns with Amazon Personalize data expectations. You can analyze data in a Domain dataset group or Custom dataset group.

When generating insights and calculating statistics, Amazon Personalize considers all bulk and streamed data from non-anonymous users. Events from anonymous users aren’t considered until you associate them with a userId. For more information, see Recording events for anonymous users (p. 287).

Topics
• Required permissions for analyzing data (p. 301)
• Data insights (p. 301)
• Viewing dataset insights and statistics (p. 303)

Required permissions for analyzing data

If you give users full access to Amazon Personalize, no permissions changes are required. If you grant your users only the permissions required to perform a task in Amazon Personalize, your AWS Identity and Access Management (IAM) policy must include the following additional data insight actions.

• personalize:CreateDataInsightsJob
• personalize:ListDataInsightsJob
• personalize:DescribeDataInsightsJob
• personalize:GetDataInsight

Data insights

The following are the possible data insights that you can generate in Amazon Personalize.

<table>
<thead>
<tr>
<th>Insight</th>
<th>Action</th>
<th>Related dataset(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>The Interactions dataset has only X interactions. Model training requires a minimum of 1,000 interactions. We recommend at least 50,000.</td>
<td>Import Y additional unique interactions records before training a model.</td>
<td>Interactions</td>
</tr>
<tr>
<td>The Interactions dataset has only X unique users with two or more interactions. Model training requires at least 25 such users. We recommend at least 1,000.</td>
<td>Import at least 2 interactions records each for Y additional users.</td>
<td>Interactions</td>
</tr>
<tr>
<td>X% of items in the Items dataset have no interactions in the Interactions dataset, so they might not be recommended.</td>
<td>Make sure you import all of your interactions data and check for mismatching IDs between your items and interactions datasets. Check the Dataset Statistics below for your items and interactions datasets to make sure you have imported the expected number of rows. If your use case or recipe uses exploration, modify the exploration.</td>
<td>Interactions and Items</td>
</tr>
<tr>
<td>Insight</td>
<td>Action</td>
<td>Related dataset(s)</td>
</tr>
<tr>
<td>------------------------------------------------------------------------</td>
<td>----------------------------------------------------------------------------------------------------------------------------------------</td>
<td>---------------------------------</td>
</tr>
<tr>
<td>X% of users in the Users dataset have no interactions in the Interactions dataset. These users will receive recommendations for popular items.</td>
<td>Make sure you import all of your interactions data and check for mismatching IDs between your users and interactions datasets. Check the Dataset Statistics below for your users and interactions datasets to make sure you have imported the expected number of rows. Import any additional interactions so more users have interactions data.</td>
<td>Interactions and Users</td>
</tr>
<tr>
<td>X% of users in the Users dataset have no interactions in the Interactions dataset. These users will receive recommendations for popular items.</td>
<td>Make sure you import all of your interactions data and check for mismatching IDs between your users and interactions datasets. Check the Dataset Statistics below for your users and interactions datasets to make sure you have imported the expected number of rows. Import any additional interactions so more users have interactions data.</td>
<td>Interactions and Users</td>
</tr>
<tr>
<td>The &lt;Users or Items or Interactions&gt; dataset has X% rows with a missing value. This can negatively affect recommendations. We recommend that all required and optional fields be at least 70% percent complete.</td>
<td>Import additional complete records, or import data again without incomplete rows, or import data again with missing values replaced with substitute data, such as the average for numeric columns or the most common value for categorical columns.</td>
<td>Any</td>
</tr>
<tr>
<td>The following column(s) in the &lt;datasetType&gt; dataset are less than 70% complete: &lt;ColumnName, ColumnName...&gt;. If this data is included in training, it can negatively affect recommendations. We recommend that columns that allow null values be at least 70% complete.</td>
<td>Import additional complete records, or import data again without incomplete rows, or import data again with missing values replaced with substitute data, such as the average for numeric columns or the most common value for categorical columns.</td>
<td>Any</td>
</tr>
<tr>
<td>The following column(s) have outliers: &lt;ColumnName, ColumnName...&gt;. Outliers are not always an issue, but sometimes negatively impact recommendations.</td>
<td>Using the Column Statistics below, check if the min and max values for these columns match your expectations. If these values are unexpected, check the data in these columns for inaccuracies and review your data collection and data processing for issues.</td>
<td>Any</td>
</tr>
<tr>
<td>The following column(s) have more than 30 possible categories: &lt;ColumnName, ColumnName...&gt;. If this data is included in training, it can negatively impact recommendations: &lt;ColumnName, ColumnName...&gt;.</td>
<td>Check your categorical data for issues, such as duplicated categories caused by variations in spelling. Resolve any inaccuracies and import data again.</td>
<td>Any</td>
</tr>
<tr>
<td>The following textual metadata column(s) are less than 85% percent complete and will not be used in model training: &lt;ColumnName, ColumnName...&gt;.</td>
<td>Import additional rows or import the rows again with text data for these column(s).</td>
<td>Items</td>
</tr>
<tr>
<td>The Interactions dataset has more than 10 unique event types, which will cause model training to fail.</td>
<td>Check your event type column for inaccuracies such as duplicated event types caused by variations in spelling. Remove unnecessary event types and import data again.</td>
<td>Interactions</td>
</tr>
</tbody>
</table>
Viewing dataset insights and statistics

To view insights and statistics on your data in Amazon Personalize datasets, navigate to your datasets in the Amazon Personalize console and choose run analysis.

To view insights and statistics

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. On the Dataset groups page, choose your dataset group.
3. From the navigation pane, choose Datasets and then choose Data analysis.
4. At the top right, choose Run analysis. Amazon Personalize starts analyzing your data. This can take up to 15 minutes. If successful, the results appear on this page.
5. In Insights, use the following to filter the insights that appear.
   - To find insights that include specific language, enter your criteria in Find insight. As you enter text, the list updates to include only insights with the exact string in the insight or recommended action.
   - To filter the insights by dataset type, change All datasets to the specific dataset type. The list updates to include only insights related to this dataset.
6. To view dataset statistics for a dataset, do the following.
   - To view general details and statistics about a dataset, such as the number of rows, unique users and unique items in an Interactions dataset, expand the section for the dataset.
   - To view detailed statistics for a column, expand the dataset section, choose Column level statistics and choose the radio button for the column.
7. Correct any issues in your data, import it again, and run another analysis to verify. For more information on importing data again, see Updating data (p. 292).

Exporting a dataset

After you import your data into an Amazon Personalize dataset, you can export the data to an Amazon S3 bucket. You might export data to verify and inspect the data that Amazon Personalize uses to generate recommendations, view the user interaction events that you previously recorded in real time, or perform offline analysis on your data.

You can choose to export only the data that you imported in bulk (imported using an Amazon Personalize dataset import job), only the data that you imported individually (records imported using the console or the PutEvents, PutUsers, or PutItems operations), or both.

For records that match exactly for all fields, Amazon Personalize exports just one record. If two records have the same ID but one or more fields are different, Amazon Personalize includes or removes the records depending on data you choose to export.

<table>
<thead>
<tr>
<th>Insight</th>
<th>Action</th>
<th>Related dataset(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>The Interactions dataset has the same timestamp for all records. If you use a USER_SEGMENTATION recipe and all records have the same timestamp, model training will fail.</td>
<td>Check your data for timestamp issues and replace duplicated timestamps with unique timestamps.</td>
<td>Interactions</td>
</tr>
</tbody>
</table>
• If you export both bulk and incremental data, Amazon Personalize exports only the newest items with the same ID (in Items dataset exports), and only users with the same ID (in Users dataset exports). For Interactions datasets, Amazon Personalize exports all interactions data.

• If you export incremental data only, Amazon Personalize exports all item, user, or interaction data that you imported individually, including items or users with the same IDs. Only records that match exactly for all fields are excluded.

• If you export bulk data only, Amazon Personalize includes all item, user, or interaction data that you imported in bulk, including items or users with the same IDs. Only records that match exactly for all fields are excluded.

To export a dataset, you create a dataset export job. A dataset export job is a record export tool that outputs the records in a dataset to one or more CSV files in an Amazon S3 bucket. The output CSV file includes a header row with column names that match the fields in the dataset's schema.

You can create a dataset export job with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

Topics
• Dataset export job permissions requirements (p. 304)
• Creating a dataset export job (console) (p. 305)
• Creating a dataset export job (AWS CLI) (p. 306)
• Creating a dataset export job (AWS SDKs) (p. 307)

Dataset export job permissions requirements

To export a dataset, Amazon Personalize needs permission to add files to your Amazon S3 bucket. To grant permissions, attach a new AWS Identity and Access Management (IAM) policy to your Amazon Personalize service role that grants the role permission to use the PutObject and ListBucket Actions on your bucket, and attach a bucket policy to your output Amazon S3 bucket that grants the Amazon Personalize principle permission to use the PutObject and ListBucket Actions.

If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

Service role policy for exporting a dataset

The following example policy grants your Amazon Personalize service role permission to use the PutObject and ListBucket Actions. Replace bucket-name with the name of your output bucket. For information about attaching policies to a IAM service role, see Attaching an Amazon S3 policy to your Amazon Personalize service role (p. 17).

```json
{
    "Version": "2012-10-17",
    "Id": "PersonalizeS3BucketAccessPolicy",
    "Statement": [
        {
            "Sid": "PersonalizeS3BucketAccessPolicy",
            "Effect": "Allow",
            "Action": ["s3:PutObject", "s3:ListBucket"],
            "Resource": ["arn:aws:s3:::bucket-name",
```
Amazon S3 bucket policy for exporting a dataset

The following example policy grants Amazon Personalize permission to use the PutObject and ListBucket Actions on an Amazon S3 bucket. Replace bucket-name with the name of your bucket. For information on adding an Amazon S3 bucket policy to a bucket, see How Do I Add an S3 Bucket Policy? in the Amazon Simple Storage Service User Guide.

```json
{
  "Version": "2012-10-17",
  "Id": "PersonalizeS3BucketAccessPolicy",
  "Statement": [
    {
      "Sid": "PersonalizeS3BucketAccessPolicy",
      "Effect": "Allow",
      "Principal": {
        "Service": "personalize.amazonaws.com"
      },
      "Action": [
        "s3:PutObject",
        "s3:ListBucket"
      ],
      "Resource": [
        "arn:aws:s3:::bucket-name",
        "arn:aws:s3:::bucket-name/**"
      ]
    }
  ]
}
```

Creating a dataset export job (console)

After you import your data into a dataset and create an output Amazon S3 bucket, you can export the data to the bucket for analysis. To export a dataset using the Amazon Personalize console, you create a dataset export job. For information about creating an Amazon S3 bucket, see Creating a bucket in the Amazon Simple Storage Service User Guide.

Before you export a dataset, make sure that your Amazon Personalize service role can access and write to your output Amazon S3 bucket. See Dataset export job permissions requirements (p. 304).

To create a dataset export job (console)

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home.
2. In the navigation pane, choose Dataset groups.
3. On the Dataset groups page, choose your dataset group.
4. In the navigation pane, choose Datasets.
5. Choose the dataset that you want to export to an Amazon S3 bucket.
6. In Dataset export jobs, choose Create dataset export job.
7. In Dataset export job details, for Dataset export job name, enter a name for the export job.
8. For IAM service role, choose the Amazon Personalize service role that you created in Creating an IAM role for Amazon Personalize (p. 15).
9. For **Amazon S3 data output path**, enter the destination Amazon S3 bucket. Use the following syntax:

   \[s3://<\text{name of your S3 bucket}>/<\text{folder path}>\]

10. If you are using AWS KMS for encryption, for **KMS key ARN**, enter the Amazon Resource Name (ARN) for the AWS KMS key.

11. For **Export data type**, choose the type data to export based on how you originally imported the data.
   - Choose **Bulk** to export only data that you imported in bulk using a dataset import job.
   - Choose **Incremental** to export only data that you imported individually using the console or the `PutEvents`, `PutUsers`, or `PutItems` operations.
   - Choose **Both** to export all of the data in the dataset.

12. For **Tags**, optionally add any tags. For more information about tagging Amazon Personalize resources, see [Tagging Amazon Personalize resources](p. 370).

13. Choose **Create dataset export job**.

   On the **Dataset overview** page, in **Dataset export jobs**, the job is listed with an **Export job status**. The dataset export job is complete when the status is **ACTIVE**. You can then download the data from the output Amazon S3 bucket. For information on downloading objects from an Amazon S3 bucket, see [Downloading an object](https://docs.aws.amazon.com/AmazonS3/latest/API/index.html) in the *Amazon Simple Storage Service User Guide*.

## Creating a dataset export job (AWS CLI)

After you import your data into the dataset and create an output Amazon S3 bucket, you can export the dataset to the bucket for analysis. To export a dataset using the AWS CLI, create a dataset export job using the `create-dataset-export-job` AWS CLI command. For information about creating an Amazon S3 bucket, see [Creating a bucket](https://docs.aws.amazon.com/AmazonS3/latest/API/index.html) in the *Amazon Simple Storage Service User Guide*.

Before you export a dataset, make sure that the Amazon Personalize service role can access and write to your output Amazon S3 bucket. See [Dataset export job permissions requirements](p. 304).

The following is an example of the `create-dataset-export-job` AWS CLI command. Give the job a name, replace `dataset arn` with the Amazon Resource Name (ARN) of the dataset that you want to export, and replace `role ARN` with the ARN of the Amazon Personalize service role that you created in [Creating an IAM role for Amazon Personalize](p. 15). In `s3DataDestination`, for the `kmsKeyArn`, optionally provide the ARN for your AWS KMS key, and for the `path` provide the path to your output Amazon S3 bucket.

For **ingestion-mode**, specify the data to export from the following options:

- Specify **BULK** to export only data that you imported in bulk using a dataset import job.
- Specify **PUT** to export only data that you imported individually using the console or the `PutEvents`, `PutUsers`, or `PutItems` operations.
- Specify **ALL** to export all of the data in the dataset.

For more information see [CreateDatasetExportJob](https://docs.aws.amazon.com/personalize/latest/dg/creating-dataset-export-job-aws-cli.html) (p. 440).

```bash
aws personalize create-dataset-export-job \
  --job-name job name \
  --dataset-arn dataset ARN \
  --job-output "{"s3DataDestination":{"kmsKeyArn":"kms key ARN","path":"s3://bucket-name/folder-name/"}}" \
  --role-arn role ARN \
```
Creating a dataset export job (AWS SDKs)

After you import your data into the dataset and create an output Amazon S3 bucket, you can export the dataset to the bucket for analysis. To export a dataset using the AWS SDKs, create a dataset export job using the `CreateDatasetExportJob` operation. For information about creating an Amazon S3 bucket, see Creating a bucket in the Amazon Simple Storage Service User Guide.

The following code shows how to create a dataset export job using the SDK for Python (Boto3) or the SDK for Java 2.x SDK.

Before you export a dataset, make sure that the Amazon Personalize service role can access and write to your output Amazon S3 bucket. See Dataset export job permissions requirements (p. 304).

SDK for Python (Boto3)

Use the following `create_dataset_export_job` to export the data in a dataset to an Amazon S3 bucket. Give the job a name, replace `dataset arn` with the Amazon Resource Name (ARN) of the dataset that you want to export, and replace `role ARN` with the ARN of the Amazon Personalize service role that you created in Creating an IAM role for Amazon Personalize (p. 15). In `s3DataDestination`, for the `kmsKeyArn`, optionally provide the ARN for your AWS KMS key, and for the `path` provide the path to your output Amazon S3 bucket.

For `ingestionMode`, specify the data to export from the following options:

- Specify BULK to export only data that you imported in bulk using a dataset import job.
- Specify PUT to export only data that you imported individually using the console or the `PutEvents`, `PutUsers`, or `PutItems` operations.
- Specify ALL to export all of the data in the dataset.

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.create_dataset_export_job(
    jobName = 'job name',
    datasetArn = 'dataset ARN',
    jobOutput = {
        "s3DataDestination": {
            "kmsKeyArn": "kms key ARN",
            "path": "s3://bucket-name/folder-name/"
        }
    }
).
```

Use the `DescribeDatasetExportJob` operation to check the status.

```bash
aws personalize describe-dataset-export-job \
--dataset-export-job-arn dataset export job ARN
```
SDK for Java 2.x

Use the following `createDatasetExportJob` method to create a dataset export job. Pass the following as parameters: a `PersonalizeClient`, the name for your export job, the ARN of the dataset you want to export, the ingestion mode, the path for the output Amazon S3 bucket, and the ARN for your AWS KMS key.

The `ingestionMode` can be one of the following options:

- Use `IngestionMode.BULK` to export only data that you imported in bulk using a dataset import job.
- Use `IngestionMode.PUT` to export only data that you imported individually using the console or the `PutEvents`, `PutUsers`, or `PutItems` operations.
- Use `IngestionMode.ALL` to export all of the data in the dataset.

```java
public static void createDatasetExportJob(PersonalizeClient personalizeClient, String jobName, String datasetArn, IngestionMode ingestionMode, String roleArn, String s3BucketPath, String kmsKeyArn) {
    long waitInMilliseconds = 30 * 1000; // 30 seconds
    String status = null;
    try {
        S3DataConfig exportS3DataConfig = S3DataConfig.builder()
            .path(s3BucketPath)
            .kmsKeyArn(kmsKeyArn)
            .build();

        DatasetExportJobOutput jobOutput = DatasetExportJobOutput.builder()
            .s3DataDestination(exportS3DataConfig)
            .build();

        CreateDatasetExportJobRequest createRequest = CreateDatasetExportJobRequest.builder()
            .jobName(jobName)
            .datasetArn(datasetArn)
            .ingestionMode(ingestionMode)
            .jobOutput(jobOutput)
            .roleArn(roleArn)
            .build();

        String datasetExportJobArn = personalizeClient.createDatasetExportJob(createRequest).datasetExportJobArn();
    }
```
Deleting data

To delete data in Amazon Personalize, you delete the dataset. You can't delete a dataset if a dataset import job or solution version is in the CREATE PENDING or IN PROGRESS state. If you use the User-Personalization recipe or Top picks for you and Recommended for you use cases, deleting a dataset halts automatic updates for any associated solution versions or recommenders.

You can delete a dataset with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs.

**Topics**
- [Deleting a dataset (console) (p. 309)]
- [Deleting a dataset (AWS CLI) (p. 310)]
- [Deleting a dataset (AWS SDKs) (p. 310)]

### Deleting a dataset (console)

To delete a dataset with the Amazon Personalize console, navigate to the dataset details page and choose delete.

**To delete a dataset**

2. In the navigation pane, choose **Dataset groups**.
3. On the **Dataset groups** page, choose your dataset group.
4. In the navigation pane, choose **Datasets**.
5. Choose the dataset that you want to delete.
6. Choose **Delete** and confirm dataset deletion.

## Deleting a dataset (AWS CLI)

The following code shows how to delete a dataset with the AWS CLI and the [DeleteDataset](p. 479) operation.

```bash
aws personalize delete-dataset --dataset-arn dataset-arn
```

## Deleting a dataset (AWS SDKs)

The following code shows how to delete a dataset with the AWS SDKs and the [DeleteDataset](p. 479) operation.

### SDK for Python (Boto3)

```python
import boto3
personalize = boto3.client('personalize')
response = personalize.delete_dataset(
    datasetArn = 'dataset ARN'
)
```

### SDK for Java 2.x

```java
public static void deleteDataset(PersonalizeClient personalizeClient,
                               String datasetArn) {
    try {
        DeleteDatasetRequest deleteRequest = DeleteDatasetRequest.builder()
            .datasetArn(datasetArn)
            .build();

        int responseCode = personalizeClient.deleteDataset(deleteRequest).sdkHttpResponse().statusCode();
        System.out.println(responseCode);
    } catch (PersonalizeException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    }
}
```
Filtering recommendations and user segments

When getting recommendations with domain recommender or custom campaign, you can filter results based on custom criteria. For example, you might not want to recommend products that a user has already purchased or recommend only items for a particular age group. Similarly, with USER_SEGMENTATION recipes, you might not want to include certain types of users in user segments. By filtering your results, you can control the items that will be recommended to users or the users that will be included in user segments.

When you get personalized recommendations or similar items, you can specify a promotion in your request. A promotion uses a filter to define additional business rules that apply to a configurable subset of recommended items. For more information see Promoting items in recommendations (p. 248).

You apply a filter and specify any filter parameter values when you call the GetRecommendations (p. 608) or GetPersonalizedRanking (p. 604) operations, or when you get recommendations from a campaign or a recommender in the console.

For batch workflows, you include any filter parameter values in your input JSON. Then you specify the filter's Amazon Resource Name (ARN) when you create a batch inference job or batch segment job. For more information see Filtering batch recommendations and user segments (custom resources) (p. 326).

You can create, edit, delete, and apply filters using the Amazon Personalize console, the AWS Command Line Interface (AWS CLI), and the AWS SDKs. For information about the number of filters you can create and how many parameters you can use in filter expressions, see Service quotas (p. 415).

Filter updates for new records

When filtering items and users, Amazon Personalize recognizes event types that were present during training or event types that were streamed using the PutEvents operation.

For interactions data imported with the PutEvents operation, Amazon Personalize updates any filters in the dataset group with the new interactions data within seconds of import. For example, if your filter removes purchased items from recommendations, and you record a purchase event for a user, this item would be removed from future recommendations for this user within seconds of recording the event.

For item and user data imported individually, Amazon Personalize updates any filters in the dataset group with your new item and user data within 20 minutes from the last individual import. For more information, see Importing individual records (p. 182).

Important
To filter recommendations using a filter with parameters and a campaign that you deployed before November 10, 2020, you must redeploy the campaign by using the UpdateCampaign (p. 588) operation or create a new campaign.

Topics
- Filter expressions (p. 311)
- Filtering real-time recommendations (p. 317)
- Filtering batch recommendations and user segments (custom resources) (p. 326)

Filter expressions

To configure filters, you must use a properly formatted filter expression. Filter expressions are composed of dataset and field identifiers in dataset:field format, along with logical operators, keywords, and
values. For values, you can specify fixed values or add placeholder parameters set the filter criteria when you get recommendations.

For a complete list of filter expression elements, see Filter expression elements (p. 313). For examples of filter expressions, see Filter expression examples (p. 314).

Note
Amazon Personalize ignores case only when matching event types.

Topics
• Creating filter expressions (p. 312)
• Filter expression examples (p. 314)

Creating filter expressions

The general structure of a filter expression is as follows:

```
EXCLUDE/INCLUDE ItemID/UserID WHERE dataset type.field IN/NOT IN (value/parameter)
```

You can either manually create filter expressions or get help with expression syntax and structure by using the Expression builder (p. 320) in the console. You can use filter expressions to filter items or users based on data from the following datasets:

• **Interactions**: Use filter expressions to include or exclude items that a user has interacted with from recommendations (for example, user events such as click or stream). Or include or exclude users that have taken certain actions from user segments (interactions with certain event types). Interactions filters can't be used with the Item-Attribute-Affinity recipe (p. 150).

Amazon Personalize considers up to 100 of the most recent interactions per user per event type. This is an adjustable quota. You can request a quota increase using the Service Quotas console.

• **Items**: Use filter expressions to include or exclude items based on specific item conditions. You can’t use filters to include or exclude items based on unstructured textual item metadata such as product descriptions. If your domain use case or custom recipe generates related items recommendations, such as the Similar-Items recipe or the More Like X domain use case, you can use filter expressions to include or exclude items based on the properties of the item you specify in your recommendation request.

• **Users**: For item recommendations for users, use filter expressions to include or exclude items from recommendations based on properties of the user you are getting recommendations for (the CurrentUser). If you have a Users dataset, you can add an IF condition to your expression to check conditions for the CurrentUser regardless of the dataset that is being used in the expression.

For user segments, use filter expressions to include or exclude users from user segments based on properties of users.

When creating a filter expression, note the following limitations:

• You can’t use filters to include or exclude items based on unstructured textual item metadata such as product descriptions.

• You can’t chain Interaction and Item datasets into one expression. To create a filter that filters by Interaction and then Item datasets (or the opposite), you must chain two or more expressions together. For more information, see Combining multiple expressions (p. 316).

• You can’t create filter expressions that filter using values with a boolean type in your schema. To filter based on boolean values, use a schema with a field of type String and use the values "True" and "False" in your data. Or you can use type int or long and values 0 and 1.
• The maximum number of distinct dataset fields for a filter, either in one expression or across multiple expressions chained together, is 5. The maximum number of distinct dataset fields across all filters in a dataset group is 10.

• You can apply a filter with the CurrentItem element only if your domain use case or custom recipe generates related items recommendations, such as the Similar-Items recipe or the More Like X domain use case.

• You can't use placeholder parameters in a filter expression that uses the NOT_IN operator. Instead, use the IN operator and use the opposite Action. For example, use Include instead of Exclude (or the reverse).

Filter expression elements

Use the following elements to create filter expressions:

**INCLUDE** or **EXCLUDE**

Use **INCLUDE** to limit recommendations to only items that meet the filter criteria OR use **EXCLUDE** to remove all items that meet the filter criteria.

**ItemID/UserId**

Use **ItemID** after the **INCLUDE** or **EXCLUDE** element for filtering item recommendations. Use **UserId** for filtering users from user segments.

**WHERE**

Use **WHERE** to check conditions for items or users. You must use the **WHERE** element after the **ItemID** or **UserId**.

**AND/OR**

To chain multiple conditions together within the same filter expression, use **AND** or **OR**. Conditions chained together using **AND** or **OR** can only affect fields of the dataset used in the first condition.

**Dataset.field**

Provide the dataset and the metadata field that you want to filter recommendations by in **dataset.field** format. For example, to filter based on the genres field in your Items dataset, you would use **Items.genres** in your filter expression. You can't use filters to include or exclude items based on textual item metadata.

**IF condition**

Use an **IF condition only** to check conditions for the **CurrentUser** and only **once** at the end of an expression. However, you can extend an IF condition using **AND**.

**CurrentUser.attribute**

To filter item recommendations based on the user you are getting recommendations for, in **only** an IF condition, use **CurrentUser** and provide the user field. For example, **CurrentUser.AGE**.

**CurrentItem.attribute**

For only related items recipes and use cases, use **CurrentItem.attribute** to filter items based on an attribute of the item you specify in your request for related items recommendations. For example, **CurrentItem.GENRE** or **CurrentItem.PRICE**.

You can apply a filter with the **CurrentItem** element only if your domain use case or custom recipe generates related items recommendations, such as the Similar-Items recipe or the More Like X domain use case. The first time you create a filter with a **CurrentItem** element, filter creation can take a few minutes. If you use AWS KMS for encryption, filter creation can take up to 15 minutes.
IN/NOT IN

Use IN or NOT IN as comparison operators to filter based on matching (or not matching) one or more string values. Amazon Personalize filters only on exact strings.

Comparison operators

Use =, <, <=, >, >= operators to test numerical data, including data passed in a placeholder parameter, for equality.

Asterisk (*) character

Use * to include or exclude interactions of all types. Use * only for filter expressions that use the EVENT_TYPE field of an Interactions dataset.

Pipe separator

Use the pipe separator (|) to chain multiple expressions together. For more information, see Combining multiple expressions (p. 316).

Parameters

For expressions that use comparison operators or the IN operator, use the dollar sign ($) and a parameter name to add a placeholder parameter as a value. For example, $GENRES. For this example, when you get recommendations, you supply the genre or genres to filter by. For information on the number of parameters you can use, see Service quotas (p. 415).

Note

You define a parameter name when you add it to an expression. The parameter name does not have to match the field name. We recommend that you use a parameter name that is similar to the field name and easy to remember. You use the parameter name (case sensitive) when you apply the filter to recommendations requests. For an example that shows how to apply a filter with placeholder parameters when using the AWS SDKs, see Applying a filter (AWS SDKs) (p. 324).

Filter expression examples

Use the filter expressions in the following sections to learn how to build your own filter expressions.

Item recommendation filter expressions

The following filter expressions show how to filter item recommendations based on interactions, item metadata, and user metadata. They are organized by data type.

Interaction data

The following expression excludes items based on an event type (such as click) or event types that you specify when you get recommendations using the $EVENT_TYPE parameter.

```
EXCLUDE ItemID WHERE Interactions.EVENT_TYPE IN ($EVENT_TYPE)
```

The following expression excludes items that a user clicked or streamed.

```
EXCLUDE ItemID WHERE Interactions.EVENT_TYPE IN ("click", "stream")
```

The following expression includes only items that the user has clicked.

```
INCLUDE ItemID WHERE Interactions.EVENT_TYPE IN ("click")
```
The following expression excludes items based on a category or categories that you specify when you get recommendations using the $CATEGORY parameter.

```
EXCLUDE ItemID WHERE Items.CATEGORY IN ($CATEGORY)
```

The following expression includes only items that are cheaper than the current item (the item you specify in the request for related items recommendations), and created by the same studio as the current item. You can apply a filter with the CurrentItem element only if your domain use case or custom recipe generates related items recommendations.

```
INCLUDE ItemID WHERE Items.PRICE < CurrentItem.PRICE AND Items.GENRE IN CurrentItem.GENRE
```

The following expression excludes items based on multiple levels of categorical fields. It excludes items with a CATEGORY_L1 value of shoe that do not have a CATEGORY_L2 value of boot.

```
EXCLUDE ItemID WHERE Items.CATEGORY_L1 IN ("shoe") AND Items.CATEGORY_L2 NOT IN ("boot")
```

The following expression includes only items with a price less than or equal to the price that you specify when you get recommendations using the $PRICE parameter.

```
INCLUDE ItemID WHERE Items.PRICE <= $PRICE
```

The following expression includes only items that have been created earlier than a timestamp (in Unix epoch time) that you specify when you get recommendations.

```
INCLUDE ItemID WHERE Items.CREATION_TIMESTAMP < $DATE
```

The following expression includes only items with a genre or genres that you specify when you get recommendations using the $GENRE parameter.

```
INCLUDE ItemID WHERE Items.GENRE IN ($GENRE)
```

The following expression includes only items that are more expensive than the current item and created more recently than a timestamp (in Unix epoch time) that you specify. You might use this filter if you are getting related item recommendations, and want to apply some specific business rules based on price and a varying creation date.

```
INCLUDE ItemID WHERE Items.PRICE < CurrentItem.PRICE AND Items.CREATION_TIMESTAMP > $DATE
```

**User data**

The following expression excludes items with a genre or genres that you specify when you get recommendations using the $GENRE parameter, but only if the current user's age is equal to the value that you specify when you get recommendations using the $AGE parameter.

```
EXCLUDE ItemID WHERE Items.GENRE IN ($GENRE) IF CurrentUser.AGE = $AGE
```

The following expression includes only items with watch for CATEGORY_L1 and luxury for CATEGORY_L2, if the current user's age is over 18.

```
INCLUDE ItemID WHERE Items.CATEGORY_L1 IN ("watch") AND Items.CATEGORY_L2 IN ("luxury") IF CurrentUser.AGE > 18
```
User segment filter expressions

The following filter expressions show how to filter user segments based on interactions data and user metadata. They are organized by data type.

**User data**

The following filter expression includes only users with a membership status equal to the value that you specify when you get user segments.

```
INCLUDE UserID WHERE Users.MEMBERSHIP_STATUS IN ($MEMBERSHIP)
```

The following filter expression excludes users with an AGE less than a value you specify when you get user segments.

```
EXCLUDE UserID WHERE Users.AGE < $AGE
```

**Interaction data**

The following filter expression includes only users who have clicked or rated items.

```
INCLUDE UserID WHERE Interactions.EVENT_TYPE IN ("click", "rating")
```

The following filter expression excludes users from user segments who have interactions with an event type you specify when you get user segments.

```
EXCLUDE UserID WHERE Interactions.EVENT_TYPE IN ($EVENT_TYPE)
```

**Combining multiple expressions**

To combine multiple expressions together you use a pipe separator (|). Use a combination of expressions when you want to Items and Interactions datasets with one filter. Each expression is first evaluated independently and the result is either the union or the intersection of the two results.

**Matching expressions example**

If both expressions use EXCLUDE or both expressions use INCLUDE, the result is the union of the two results as follows (A and B are different expressions):

- Exclude A | Exclude B is equal to Exclude result from A or result from B
- Include A | Include B is equal to Include result from A or result from B

The following example shows how to combine two expressions that use INCLUDE. The first expression includes only items with a category or categories that you specify when you get recommendations using the $CATEGORY parameter. The second expression includes items the user has marked as a favorite. Recommendations will include only items with the category you specify along with items that the user has marked as a favorite.

```
INCLUDE ItemID WHERE Items.CATEGORY IN ($CATEGORY) | INCLUDE ItemID WHERE Interactions.EVENT_TYPE IN ("favorite")
```

**INCLUDE and EXCLUDE example**
If one or more expression uses INCLUDE and one more expression uses EXCLUDE, the result is the subtraction of the EXCLUDE expression result from the INCLUDE expression result as follows (A, B, C, and D are different expressions).

- Include A | Exclude B is equal to Include result from A - result from B
- Include A | Include B | Exclude C | Exclude D is equal to Include (A or B) - (C or D)

Expression order does not matter: If the EXCLUDE expression comes before the INCLUDE expression, the result is the same.

The following example shows how to combine an INCLUDE expression and a EXCLUDE expression. The first expression includes only items with a genre or genres that you specify when you get recommendations using the $GENRE parameter. The second expression excludes items that the user has clicked or streamed. Recommendations will include only items with a genre that you specify that have not have been clicked or streamed.

```
INCLUDE ItemID WHERE Items.GENRE IN ($GENRE) | EXCLUDE ItemID WHERE Interactions.EVENT_TYPE IN ("click", "stream")
```

Filtering real-time recommendations

You can filter real-time recommendations with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or the AWS SDKs.

When you get personalized recommendations or similar items, you can specify a promotion in your request. A promotion uses a filter to define additional business rules that apply to a configurable subset of recommended items. For more information see Promoting items in recommendations (p. 248).

Topics

- Filtering real-time recommendations (console) (p. 317)
- Filtering real-time recommendations (AWS CLI) (p. 321)
- Filtering real-time recommendations (AWS SDKs) (p. 323)

Filtering real-time recommendations (console)

To filter real-time recommendations using the console, create a filter and then apply it to a recommendation request.

**Note**
To filter recommendations using a filter with parameters and a campaign deployed before November 10, 2020, you must redeploy the campaign by using the UpdateCampaign (p. 588) operation or create a new campaign.

Creating a filter (console)

To create a filter in the console, choose the dataset group that contains the campaign or recommender you want to use to get filtered recommendations. Then provide a filter name and a filter expression.

**To create a filter (console)**

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose the dataset group that contains the campaign or recommender that you want to use to get filtered recommendations.

3. In the navigation page, choose Filters and then choose Create new filter. The Create filter page displays.

4. For Filter name, enter a name for your filter. You will choose the filter by this name when you apply it to a recommendation request.

5. For Expression, choose either Build expression or Add expression manually and build or insert your expression:
   - To use the expression builder, choose Build expression. The expression builder provides structure, fields, and guidelines for building correctly formatted filter expressions. For more information, see Using the filter expression builder (p. 320).
   - To input your own expression, choose Add expression manually. For more information, see Filter expression elements (p. 313).

6. Choose Finish. The filter's overview page shows the filter's Amazon Resource Name (ARN), status, and full filter expression. To delete the filter, choose Delete. For information about finding and deleting filters after you have left the overview page, see Deleting a filter (console) (p. 321).
Applying a filter (console)

To apply a filter, on the Test recommender panel for the recommender (Domain dataset group) or Test campaign results panel for the campaign (Custom dataset group), choose the filter and enter any filter parameter values. Then get recommendations for a user.

**Important**
For filter expressions that use an INCLUDE element to include items, you must provide values for all parameters that are defined in the expression. For filters with expressions that use an EXCLUDE element to exclude items, you can omit the filter-values. In this case, Amazon Personalize doesn't use that portion of the expression to filter recommendations.

**To apply a filter (console)**

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose the dataset group that contains the campaign or recommender that you want to use to get filtered recommendations.
3. Depending on your dataset group type, do either of the following:
   a. For a Domain dataset group, in the navigation pane choose Recommenders.
   b. For a Custom dataset group, in the navigation pane choose Custom resources then Campaigns.
4. On the Recommenders or Campaigns page, choose the target recommender or campaign.
5. For comparison, start by getting recommendations for a user without applying a filter. Under Test recommender / Test campaign results, enter the ID of a user that you want to get recommendations for, and choose Get recommendations. A table containing the user’s top recommendations appears.

6. From the Filter name menu, choose the filter that you created. If your filter has any placeholder parameters, the associated fields for each parameter appear.
7. If you’re using a filter with placeholder parameters, for each parameter, enter the value to set the filter criteria. To use multiple values for one parameter, separate each value with a comma.
8. Using the same User ID as in the earlier step, choose Get recommendations. The recommendations table appears.
If the user already bought a recommended item, the filter removes it from the recommendation list. In this example, items 2657, 2985 were replaced by the most suitable items that the user didn't buy (items 2641 and 1573).

**Using the filter expression builder**

The Expression builder on the Create filter page provides structure, fields, and guidelines for building correctly formatted filter.

To build a filter expression:

- **Use the Action, Property, Operator, and Value fields** to create an expression.

  For the Value, enter a fixed value or, to set filter criteria when you get recommendations, enter $ + a parameter name. For example, $GENRES. When you get recommendations, you’ll supply the value or values to filter by. In this example, you would provide a genre or list of genres when you get recommendations.

  Separate multiple non-parameter values with a comma. You cannot add comma-separated parameters to a filter.

  **Note**
  After you choose a Property (in dataset.field format), the Property value for any succeeding rows chained by AND or OR conditions must use the same dataset.

- **Use the + and X buttons** to add or delete a row from your expression. You can't delete the first row.

- **For new rows**, use the AND, IF, or OR operators on the AND menu to create a chain of conditions.
For IF conditions:
- Each expression can contain only one IF item. If you remove an IF condition, the Expression builder removes any AND conditions following it.
- You can use IF conditions only for expressions that filter by the CurrentUser.
- Choose the Add expression button to add an additional filter expression for more precise filtering, including filtering using Items and Interactions datasets. Each expression is first evaluated independently and the result is a union of the two results.

**Note**
To create a filter that uses both Item and Interaction datasets, you must use multiple expressions.

**Expression builder example**

The following example shows how to build a filter that excludes items with a genre that you specify when you get recommendations (note the $GENRES placeholder parameter). The filter also excludes items with a DOWNLOAD_COUNT of more than 200, but only if the current user's age is greater than 17.

![Expression builder example](attachment:image)

**Deleting a filter (console)**

Deleting a filter removes the filter from the list of filters for a dataset group.

**Important**
You can't delete a filter while a batch inference job is in progress.

**To delete a filter (console)**

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign into your account.
2. From the Dataset groups list, choose the dataset group that contains the filter that you want to delete.
3. In the navigation pane, choose Filters.
4. From the list of filters, choose the filter that you want to delete and choose View Details. The filter details page appears.
5. Choose Delete and confirm the deletion in the confirmation dialog box.

**Filtering real-time recommendations (AWS CLI)**

To filter recommendations using the AWS CLI, you create a filter and then apply it by specifying the filter ARN in a [GetRecommendations (p. 608)](amazon-personalize-developer-guide#GetRecommendations) or [GetPersonalizedRanking (p. 604)](amazon-personalize-developer-guide#GetPersonalizedRanking) request.
Important
To filter recommendations using a filter with parameters and a campaign you deployed before November 10, 2020, you must re-deploy the campaign by using the UpdateCampaign (p. 588) call or create a new campaign.

Creating a filter (AWS CLI)

Use the following create-filter operation to create a filter and specify the filter expression.

Replace the Filter name with the name of the filter, and the Dataset group ARN with the Amazon Resource Name (ARN) of the dataset group. Replace the sample filter-expression with your own filter expression.

```bash
aws personalize create-filter
--name Filter name
--dataset-group-arn dataset group arn
--filter-expression "EXCLUDE ItemID WHERE Items.CATEGORY IN ("$CATEGORY")"
```

If successful, the filter ARN is displayed. Record it for later use. To verify that the filter is active, use the DescribeFilter (p. 515) operation before you use the filter.

For more information about the API, see CreateFilter (p. 455). For more information about filter expressions, including examples, see Creating filter expressions (p. 312).

Applying a filter (AWS CLI)

When you use the get-recommendations or get-personalized-ranking operations, apply a filter by passing the filter-arn and any filter values as parameters.

The following is an example of the get-recommendations operation. Replace Campaign ARN with the Amazon Resource Name (ARN) of your campaign User ID with the ID of the user that you are getting recommendations for, and Filter ARN with the ARN of your filter. If you're getting recommendations from a recommender instead of a campaign, use recommender-arn instead of --campaign-arn and provide the ARN for the recommender.

If your expression has any parameters, include the filter-values object. For each parameter in your filter expression, provide the parameter name (case sensitive) and the values. For example, if your filter expression has a $GENRE parameter, provide "GENRE" as the key, and a genre or genres, such as "Comedy", as the value. Separate multiple values with a comma. For example, ""comedy", "drama, "horror"".

Important
For filter expressions that use an INCLUDE element to include items, you must provide values for all parameters that are defined in the expression. For filters with expressions that use an EXCLUDE element to exclude items, you can omit the filter-values. In this case, Amazon Personalize doesn't use that portion of the expression to filter recommendations.

```bash
aws personalize-runtime get-recommendations
--campaign-arn Campaign ARN
--user-id User ID
--filter-arn Filter ARN
--filter-values '{
    "Parameter name": "value1", "value2", "value3"
}
```
Deleting a filter (AWS CLI)

Use the following delete-filter operation to delete a filter. Replace filter ARN with the ARN of the filter.

```
aws personalize delete-filter --filter-arn Filter ARN
```

Filtering real-time recommendations (AWS SDKs)

To filter recommendations using the AWS SDKs, you create a filter and then apply it by specifying the filter ARN in a GetRecommendations (p. 608) or GetPersonalizedRanking (p. 604) request.

**Important**

To filter recommendations using a filter with parameters and a campaign you deployed before November 10, 2020, you must re-deploy the campaign by using the UpdateCampaign (p. 588) call or create a new campaign.

Creating a filter (AWS SDKs)

Create a new filter with the CreateFilter (p. 455) operation. The following code shows how to create a filter. Specify the filter name, Amazon Resource Name (ARN) of your dataset group, and provide your filter expression.

SDK for Python (Boto3)

```python
import boto3

personalize = boto3.client('personalize')

response = personalize.create_filter(
    name = 'Filter Name',
    datasetGroupArn = 'Dataset Group ARN',
    filterExpression = 'EXCLUDE ItemID WHERE Items.CATEGORY IN ($CATEGORY)' )

filter_arn = response['filterArn']
print("Filter ARN: " + filter_arn)
```

SDK for Java 2.x

```java
public static String createFilter(PersonalizeClient personalizeClient,
        String filterName,
        String datasetGroupArn,
        String filterExpression) {
            try {
                CreateFilterRequest request = CreateFilterRequest.builder()
                        .name(filterName)
                        .datasetGroupArn(datasetGroupArn)
                        .filterExpression(filterExpression)
                        .build();

                return personalizeClient.createFilter(request).filterArn();
            }

            catch(PersonalizeException e) {
                System.err.println(e.awsErrorDetails().errorMessage());
                System.exit(1);
            }

            return "";
    }
```
SDK for JavaScript v3

```javascript
// Get service clients module and commands using ES6 syntax.
import { CreateFilterCommand } from
"@aws-sdk/client-personalize";
import { personalizeClient } from "./libs/personalizeClients.js";
// Or, create the client here.
// const personalizeClient = new PersonalizeClient({ region: "REGION"});

// Set the filter's parameters.
export const createFilterParam = {
  datasetGroupArn: 'DATASET_GROUP_ARN', /* required */
  name: 'NAME', /* required */
  filterExpression: 'FILTER_EXPRESSION' /*required */
}

export const run = async () => {
  try {
    const response = await personalizeClient.send(new
      CreateFilterCommand(createFilterParam));
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();
```

Record the filter ARN for later use. To verify that the filter is active, use the DescribeFilter (p. 515) operation before using the filter. For more information about the API, see CreateFilter (p. 455). For more information about filter expressions, including examples, see Creating filter expressions (p. 312).

### Applying a filter (AWS SDKs)

When you use the `get_recommendations` or `get_personalized_ranking` methods, apply a filter by passing a filterArn and any filter values as parameters.

The following code shows how to get filtered Amazon Personalize recommendations for a user. Specify the ID of the user you want to get recommendations for, the Amazon Resource Name (ARN) of your campaign, and the ARN of your filter. If you're getting recommendations from a recommender instead of a campaign, use recommenderArn instead of campaignArn and provide the ARN for the recommender.

For filterValues, for each optional parameter in your filter expression, provide the parameter name (case sensitive) and the value or values. For example, if your filter expression has a $GENRES parameter, provide "GENRES" as the key, and a genre or genres, such as ""Comedy"", as the value. For multiple values, separate each value with a comma. For example, ""comedy","drama","horror"".

**Important**

For filter expressions that use an INCLUDE element to include items, you must provide values for all parameters that are defined in the expression. For filters with expressions that use an EXCLUDE element to exclude items, you can omit the filter-values. In this case, Amazon Personalize doesn't use that portion of the expression to filter recommendations.

SDK for Python (Boto3)

```python
import boto3

personalize_runtime = boto3.client("personalize-runtime")

response = personalize_runtime.get_recommendations(
```
### Filtering real-time recommendations (AWS SDKs)

The following example uses two parameters, one with two values and one with one value. Depending on your filter expression, modify the code to add or remove parameterName and parameterValue fields.

```java
public static void getFilteredRecs(PersonalizeRuntimeClient personalizeRuntimeClient,
        String campaignArn,
        String userId,
        String filterArn,
        String parameter1Name,
        String parameter1Value1,
        String parameter1Value2,
        String parameter2Name,
        String parameter2Value){
        try {
                Map<String, String> filterValues = new HashMap<>();
                filterValues.put(parameter1Name, String.format("%1$s","%2$s",parameter1Value1, parameter1Value2));
                filterValues.put(parameter2Name, String.format("%1$s", parameter2Value));

                GetRecommendationsRequest recommendationsRequest =
                        GetRecommendationsRequest.builder()
                                .campaignArn(campaignArn)
                                .numResults(20)
                                .userId(userId)
                                .filterArn(filterArn)
                                .filterValues(filterValues)
                                .build();

                GetRecommendationsResponse recommendationsResponse =
                        personalizeRuntimeClient.getRecommendations(recommendationsRequest);
                List<PredictedItem> items = recommendationsResponse.itemList();

                for (PredictedItem item: items) {
                        System.out.println("Item Id is : "+item.itemId());
                        System.out.println("Item score is : "+item.score());
                }
        } catch (PersonalizeRuntimeException e) {
                System.err.println(e.awsErrorDetails().errorMessage());
                System.exit(1);
        }
}

### SDK for JavaScript v3

```javascript
// Get service clients module and commands using ES6 syntax.
import { GetRecommendationsCommand } from
        "@aws-sdk/client-personalize-runtime";
```
import { personalizeRuntimeClient } from './libs/personalizeClients.js';
// Or, create the client here:
// const personalizeRuntimeClient = new PersonalizeRuntimeClient({ region: "REGION"});

// Set recommendation request parameters.
export const getRecommendationsParam = {
  campaignArn: 'CAMPAIGN_ARN', /* required */
  userId: 'USER_ID',      /* required */
  numResults: 15,    /* optional */
  filterArn: 'FILTER_ARN',   /* required to filter recommendations */
  filterValues: {
    "PROPERTY": "VALUE"  /* Only required if your filter has a placeholder parameter */
  }
}

export const run = async () => {
  try {
    const response = await personalizeRuntimeClient.send(new GetRecommendationsCommand(getRecommendationsParam));
    console.log("Success!", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();

Deleting a filter (AWS Python SDK)

Use the following delete_filter method to delete a filter. Replace filter ARN with the ARN of the filter.

```python
import boto3
personalize = boto3.client("personalize")
response = personalize.delete_filter(
  filterArn = "Filter ARN"
)
```

Filtering batch recommendations and user segments (custom resources)

Filtering batch recommendations and user segments works nearly the same as filtering real-time recommendations. It follows the same workflow described in Batch recommendations and user segments (custom resources) (p. 259). To filter batch recommendations or user segments, you do the following:

1. Create a filter just like you would for real-time recommendations. For more information see Filtering real-time recommendations (p. 317).
2. Prepare your input data and upload it to Amazon S3 as described in Preparing input data for batch recommendations (p. 261) or Preparing input data for user segments (p. 270). If your filter uses placeholder parameters, you must add an additional filterValues object. For more information, see Providing filter values in your input JSON (p. 327). If your filter doesn't use placeholder parameters, your input data can follow the examples in Batch inference job input and output JSON examples (p. 261) Batch segment job input and output JSON examples (p. 271)
3. Create a separate location for your output data, either a folder or a different Amazon S3 bucket.

4. Create a **batch inference job** (p. 263) or a **batch segment job** (p. 272). When you create the job, specify the Amazon Resource Name (ARN) of your filter.

5. When the batch inference or batch segment job is complete, retrieve the recommendations or user segments from your output location in Amazon S3.

### Topics
- [Providing filter values in your input JSON](p. 327)
- [Filtering batch workflows (console)](p. 328)
- [Filtering batch workflows (AWS SDKs)](p. 328)

## Providing filter values in your input JSON

For filters with placeholder parameters, such as $GENRE, you must provide the values for the parameters in a `filterValues` object in your input JSON. For a `filterValues` object, each key is a parameter name. Each value is the criteria that you are passing as a parameter. Surround each value with escaped quotes: `{"filterValues":{"GENRES":"\"drama\""}}`. For multiple values, separate each value with a comma: `{"filterValues":{"GENRES":"\"horror","comedy","drama\""}}

### Batch inference job input JSON example

The following is an example of the first few lines of a JSON input file for a **batch inference job**. The example includes the `filterValues` object. The `GENRES` key corresponds to a $GENRES placeholder in the filter expression. The job in this example uses the User-Personalization recipe. For RELATED_ITEMS recipes, provide an itemID instead of the userId. For PERSONALIZED_RANKING recipes provide the userId and an itemList.

```
{"userId": "5","filterValues":{"GENRES":"\"horror","comedy","drama\""}}
{"userId": "3","filterValues":{"GENRES":"\"horror","comedy\""}}
{"userId": "34","filterValues":{"GENRES":"\"drama\""}}
```

For more examples of batch inference job input data by recipe see [Batch inference job input and output JSON examples](p. 261). You can use these examples as a starting point and add the `filterValues` object from the above example.

### Batch segment job input JSON example

The following is an example of the first few lines of a JSON input file with filter values for a **batch segment job**. The GENRES key corresponds to a $GENRES placeholder in the filter expression.

```
{"itemAttributes": "ITEMS.genres = \"Comedy\" AND ITEMS.genres = \"Action\"","filterValues":{"COUNTRY":"\"Japan\""}}
{"itemAttributes": "ITEMS.genres = \"Horror\"","filterValues":{"COUNTRY":"\"United States\""}}
{"itemAttributes": "ITEMS.genres = \"Action\" AND ITEMS.genres = \"Adventure\"","filterValues":{"COUNTRY":"\"England\""}}
```

For more examples of batch inference job input data by recipe see [Batch segment job input and output JSON examples](p. 271). You can use these examples as a starting point and add the `filterValues` object from the above example.
Filtering batch workflows (console)

To filter batch workflows with the Amazon Personalize console, you create a filter and then you create a batch inference job or batch segment job and choose the filter. For complete step by step instructions, see Creating a batch inference job (console) (p. 263) and Creating a batch segment job (console) (p. 272).

Filtering batch workflows (AWS SDKs)

To filter batch recommendations with the AWS SDKs, create a filter and include the FilterArn parameter in the CreateBatchInferenceJob (p. 424) or CreateBatchSegmentJob (p. 428) request.

The following code shows how to create a batch inference job with a filter using the AWS SDK for Python (Boto3). We recommend using a different location for your output data (either a folder or a different Amazon S3 bucket). For complete explanation of all fields, see Creating a batch inference job (AWS SDKs) (p. 265).

```python
import boto3

personalize = boto3.client("personalize")

personalize_rec.create_batch_inference_job(
    solutionVersionArn = "Solution version ARN",
    jobName = "Batch job name",
    roleArn = "IAM role ARN",
    filterArn = "Filter ARN",
    jobInput =
        "s3DataSource": {"path": "S3 input path"},
    jobOutput =
        "s3DataDestination": {"path": "S3 output path"})
```
Measuring impact of recommendations

As your customers interact with recommendations, you can measure how the recommendations are helping you achieve your goals. You can identify which campaigns and recommenders have the most impact, such as which resource generates the most minutes watched or the most clicks. And you can compare the performance of Amazon Personalize recommendations to those generated by third-party services.

The following can help you measure the impact of recommendations:

- **Metric attribution (p. 329)**: An Amazon Personalize metric attribution creates reports based on metrics that you specify and the interactions and items data that you import. For example, the total length of movies watched by users, or the total number of click events.

- **A/B testing (p. 347)**: Performing an A/B test consists of running an experiment with multiple variations and comparing the results. You can use A/B testing to help compare and evaluate different recommendation strategies, and measure the impact of the recommendations.

Topics

- Measuring recommendation impact with a metric attribution (p. 329)
- Measuring recommendation impact with A/B testing (p. 347)

Measuring recommendation impact with a metric attribution

To measure the impact of recommendations, you can create a metric attribution. A *metric attribution* creates reports based on the interactions and items data that you import, and the metrics that you specify. For example, the total length of movies watched by users, or the total number of click events. Amazon Personalize aggregates calculations over a 15-minute window. For PutEvents and incremental bulk data, Amazon Personalize automatically sends metric reports to Amazon CloudWatch. For bulk data, you can choose to publish reports to an Amazon S3 bucket.

For each interaction that you import, include source data to compare different campaigns, recommenders, and third parties. You can include the recommendation ID of the recommendations you showed the user or the event source, such as a third party.

For example, you might have a video streaming app that shows movie recommendations from two different Amazon Personalize recommenders. If you wanted to see which recommender generates the most watch events, you could create a metric attribution that tracks the total number of watch events. Then you could record watch events as users interact with recommendations, and include the recommendationId in each event. Amazon Personalize uses the recommendationId to identify each recommender. As you record events, you can view the watch event totals aggregated over every 15 minutes for both recommenders in CloudWatch. For code samples that show how to include a recommendationId or an eventAttributionSource for an event, see Event metrics and attribution reports (p. 289).
Guidelines and requirements

Amazon Personalize starts calculating and reporting the impact of recommendations only after you create a metric attribution. To build the most complete history, we recommend creating a metric attribution before you import your interactions data. When you create an Interactions dataset import job with the Amazon Personalize console, you have the option to create a metric attribution in a new tab. Then you can return to the import job to complete it.

After you create a metric attribution and record events or import incremental bulk data, you will incur some monthly CloudWatch cost per metric. For information about CloudWatch pricing, see the Amazon CloudWatch pricing page. To stop sending metrics to CloudWatch, delete the metric attribution.

To see the impact of recommendations over time, keep importing data as customers interact with recommendations. If you have already imported data, you can still create a metric attribution and start measuring recommendation impact. However, Amazon Personalize won’t report on data that you imported before you created it.

The following are guidelines and requirements for generating reports with a metric attribution:

- You must grant Amazon Personalize permission to access and put data in CloudWatch. For policy examples, see Giving Amazon Personalize access to CloudWatch.
- To publish metrics to Amazon S3, give Amazon Personalize permission to write to your bucket. You also must provide the bucket path in your metric attribution. For policy examples, see Giving Amazon Personalize access to your Amazon S3 bucket.
- To publish metrics to CloudWatch, records must be less than 14 days old. If your data is older, these records won’t be included in calculations or reports.
- Importing duplicate events (events that match for all attributes exactly) can lead to unexpected behavior including inaccurate metrics. We recommend that you remove duplicate records from any bulk data before import, and avoid importing duplicate events with the PutEvents operation.
- Your Interactions dataset must have an EVENT_TYPE column.
- You can create at most one metric attribution per dataset group. Each metric attribution can have at most 10 metrics.

To compare sources, each interaction event must include a recommendationId or eventAttributionSource. You can provide at most 100 unique event attribution sources. For PutEvents code samples, see Event metrics and attribution reports.

- If you provide a recommendationId, Amazon Personalize automatically determines the source campaign or recommender and identifies it in reports in an EVENT_ATTRIBUTION_SOURCE column.
- If you provide both attributes, Amazon Personalize uses only the eventAttributionSource.
- If you don’t provide a source, Amazon Personalize labels the source SOURCE_NAME_UNDEFINED in reports.
Guidelines and requirements

- **Giving Amazon Personalize access to CloudWatch** (p. 331)
- **Giving Amazon Personalize access to your Amazon S3 bucket** (p. 331)

## Giving Amazon Personalize access to CloudWatch

**Important**

When you grant permissions, Amazon Personalize places and validates a small amount of data in CloudWatch. This will incur a one-time cost of less than $0.30. For more information about CloudWatch pricing, see the [Amazon CloudWatch pricing](https://aws.amazon.com/cloudwatch/pricing/) page.

To give Amazon Personalize access to CloudWatch, attach a new AWS Identity and Access Management (IAM) policy to your Amazon Personalize service role that grants the role permission to use the `PutMetricData` Action for CloudWatch. The following policy example grants `PutMetricData` permissions.

```json
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Action": ["cloudwatch:PutMetricData"],
            "Resource": "*"
        }
    ]
}
```

## Giving Amazon Personalize access to your Amazon S3 bucket

To give Amazon Personalize access to your Amazon S3 bucket:

- Attach an IAM policy to your Amazon Personalize service role that grants the role permission to use the `PutObject` Action on your bucket.

```json
{
    "Version": "2012-10-17",
    "Id": "PersonalizeS3BucketAccessPolicy",
    "Statement": [
        {
            "Sid": "PersonalizeS3BucketAccessPolicy",
            "Effect": "Allow",
            "Action": ["s3:PutObject"],
            "Resource": [
                "arn:aws:s3:::bucket-name",
                "arn:aws:s3:::bucket-name/*"
            ]
        }
    ]
}
```

- Attach a bucket policy to your output Amazon S3 bucket that grants the Amazon Personalize principle permission to use the `PutObject` Actions.

If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see [Giving Amazon Personalize permission to use your AWS KMS key](p. 20).
Creating a metric attribution

Important
After you create a metric attribution and record events or import incremental bulk data, you will incur some monthly CloudWatch cost per metric. For information about CloudWatch pricing, see the Amazon CloudWatch pricing page. To stop sending metrics to CloudWatch, delete the metric attribution (p. 341).

To start generating metric reports, you create a metric attribution and import interactions data. When you create a metric attribution, you specify a list of event types to report on. For each event type, you specify a function that Amazon Personalize applies as it collects the data. Available functions include \( \text{SUM}(\text{DatasetType.COLUMN_NAME}) \) and \( \text{SAMPLECOUNT()} \).

For example, you might have an online video streaming app and want to track two metrics: the click-through rate for recommendations, and the total length of movies watched, where each video in the Items dataset includes a \text{LENGTH} attribute. You would create a metric attribution and add two metrics, each with an event type and function. The first might be for the \text{Click} event type with a \( \text{SAMPLECOUNT()} \) function. The second might be for the \text{Watch} event type with a \( \text{SUM}(	ext{Items.LENGTH}) \) function.

You can apply \( \text{SUM()} \) functions to only numeric columns of Items and Interactions datasets. To apply a \( \text{SUM()} \) function to a column in an Items dataset, you must first import item metadata.

You can create a metric attribution with the Amazon Personalize console, AWS Command Line Interface, or AWS SDKs.

Topics
- Creating a metric attribution (console) (p. 332)
- Creating a metric attribution (AWS CLI) (p. 334)
- Creating a metric attribution (AWS SDKs) (p. 334)

Creating a metric attribution (console)

To create a metric attribution with the Amazon Personalize console, you navigate to the Metric attribution page and choose Create metric attribution. When you create a metric attribution, you
specify an optional Amazon S3 bucket path, your Amazon Personalize IAM service role, and a list of metrics to report on.

When you create an Interactions dataset import job with the Amazon Personalize console, you have the option to create a metric attribution in a new tab. Then you can return to the import job to complete it. If you’re already on the Configure metric attribution page, you can skip to step 4.

To create a metric attribution

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose your dataset group.
3. In the navigation pane, choose Metric attribution.
4. In Metric attribution details, choose Create metric attribution.
5. On the Configure metric attribution page, give the metric attribution a name.
6. If you want to publish metrics to Amazon S3 for Amazon S3 data output path, enter the destination Amazon S3 bucket. This enables the option to publish metrics each time you create a dataset import job. Use the following syntax:

   s3://<name of your S3 bucket>/<folder> path

7. If you are using AWS KMS for encryption, for KMS key ARN, enter the Amazon Resource Name (ARN) for the AWS KMS key. You must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).
8. In IAM role, choose to create a new service role or use an existing one. The role you choose must have PutMetricData permissions for CloudWatch. If you want to publish to Amazon S3, the role must have PutObject permissions for your Amazon S3 bucket.

   To use the role that you created in Creating an IAM role for Amazon Personalize (p. 15), you might have to add policies for CloudWatch and Amazon S3.

   For policy examples, see Giving Amazon Personalize access to CloudWatch (p. 331) and Giving Amazon Personalize access to your Amazon S3 bucket (p. 331).
9. Choose Next.
10. On the Define metric attributes page, choose how to define metrics. Choose Build metric attributes to use the builder tool. Choose Input metric attributes to enter metrics in JSON format.

   • If you choose Build metric attributes, for each metric provide a name, event type, and choose a function. For SUM() functions, choose the column name. Choose Add metric attribute to add additional metrics.

   • If you choose Input metric attributes, enter each metric in JSON format. The following shows how to format a metric.

      ```
      {
          "EventType": "watch",
          "MetricName": "MinutesWatchedTracker",
          "MetricMathExpression": "SUM(Items.LENGTH)"
      }
      ```

11. Choose Next.
12. On the Review and create page, review the details for the new metric attribution. To make changes, choose Previous. To create the metric attribution, choose Create. When the metric attribution is active, you can start importing data and view the results. For information on viewing results, see Publishing and viewing results (p. 343).
Creating a metric attribution (AWS CLI)

The following code shows how to create a metric attribution with the AWS Command Line Interface. The role you specify must have PutMetricData permissions for CloudWatch and, if publishing to Amazon S3, PutObject permissions for your Amazon S3 bucket. To use the role that you created in Creating an IAM role for Amazon Personalize (p. 15), you might have to add policies for CloudWatch and Amazon S3. For policy examples, see Giving Amazon Personalize access to CloudWatch (p. 331) and Giving Amazon Personalize access to your Amazon S3 bucket (p. 331).

For each metric specify a name, event type, and expression (a function). Available functions include SUM(DatasetType.COLUMN_NAME) and SAMPLECOUNT(). For SUM() functions, specify the dataset type and column name. For example, SUM(Items.LENGTH). For information on each parameter, see CreateMetricAttribution (p. 458).

```bash
aws personalize create-metric-attribution \
  --name "metric attribution name" \
  --dataset-group-arn "dataset group arn" \
  --metrics-output-config "{"roleArn": "Amazon Personalize service role ARN", \n  "s3DataDestination":{"kmsKeyArn":"kms key ARN","path":"s3://bucket-name/folder-name/"}}" \
  --metrics "[\n    "eventType": "event type", \n    "expression": "SUM(DatasetType.COLUMN_NAME)", \n    "metricName": "metric name" \n  ]"
```

Creating a metric attribution (AWS SDKs)

The following code shows how to create a metric attribution with the SDK for Python (Boto3). The role you specify must have PutMetricData permissions for CloudWatch and, if publishing to Amazon S3, PutObject permissions for your Amazon S3 bucket. To use the role that you created in Creating an IAM role for Amazon Personalize (p. 15), you might have to add policies for CloudWatch and Amazon S3. For policy examples, see Giving Amazon Personalize access to CloudWatch (p. 331) and Giving Amazon Personalize access to your Amazon S3 bucket (p. 331).

For each metric specify a name, event type, and expression (a function). Available functions include SUM(DatasetType.COLUMN_NAME) and SAMPLECOUNT(). For SUM() functions, specify the dataset type and column name. For example, SUM(Items.LENGTH). For information on each parameter, see CreateMetricAttribution (p. 458).

SDK for Python (Boto3)

```python
import boto3

personalize = boto3.client('personalize')

metricsList = [{
  "eventType": "event type",
  "expression": "SUM(DatasetType.COLUMN_NAME)",
  "metricName": "metric name"
}]

outputConfig = {
  "roleArn": "Amazon Personalize service role ARN",
  "s3DataDestination": {
    "kmsKeyArn": "key ARN",
    "path": "s3://<name of your S3 bucket>/<folder>"
  }
}
```
response = personalize.create_metric_attribution(
    name = 'metric attribution name',
    datasetGroupArn = 'dataset group arn',
    metricsOutputConfig = outputConfig,
    metrics = metricsList
)

metric_attribution_arn = response['metricAttributionArn']

print ('Metric attribution ARN: ' + metric_attribution_arn)

description = personalize.describe_metric_attribution(
    metricAttributionArn = metric_attribution_arn)['metricAttribution']

print('Name: ' + description['name'])
print('ARN: ' + description['metricAttributionArn'])
print('Status: ' + description['status'])

---

SDK for Java 2.x

```java
import com.amazon.personalize.createMetricAttribution(PersonalizeClient personalizeClient,
    String eventType,
    String expression,
    String metricName,
    String metricAttributionName,
    String roleArn,
    String s3Path,
    String kmsKeyArn,
    String datasetGroupArn) { 

    String metricAttributionArn = "";

    try {
        MetricAttribute attribute = MetricAttribute.builder()
            .eventType(eventType)
            .expression(expression)
            .metricName(metricName)
            .build();

        ArrayList<MetricAttribute> metricAttributes = new ArrayList<>();
        metricAttributes.add(attribute);

        S3DataConfig s3DataDestination = S3DataConfig.builder()
            .kmsKeyArn(kmsKeyArn)
            .path(s3Path)
            .build();

        MetricAttributionOutput outputConfig = MetricAttributionOutput.builder()
            .roleArn(roleArn)
            .s3DataDestination(s3DataDestination)
            .build();

        CreateMetricAttributionRequest createMetricAttributionRequest =
            CreateMetricAttributionRequest.builder()
            .name(metricAttributionName)
            .datasetGroupArn(datasetGroupArn)
            .metrics(metricAttributes)
            .metricsOutputConfig(outputConfig)
            .build();

        CreateMetricAttributionResponse createMetricAttributionResponse =
            personalizeClient.createMetricAttribution(createMetricAttributionRequest);

        metricAttributionArn = createMetricAttributionResponse.metricAttributionArn();
        System.out.println("Metric attribution ARN: " + metricAttributionArn);
        return metricAttributionArn;
    }
```
Managing a metric attribution

After you create a metric attribution, you can update it or delete it. If you delete a metric attribution, Amazon Personalize stops sending reports related to PutEvents and incremental imports to CloudWatch.

Topics
- Updating a metric attribution (p. 337)
- Deleting a metric attribution (p. 341)
Updating a metric attribution

When you update a metric attribution, you can add and remove metrics and modify its output configuration. You can update a metric attribution with the Amazon Personalize console, AWS Command Line Interface, or AWS SDKs.

Topics

- Updating a metric attribution (console) (p. 337)
- Updating a metric attribution (AWS CLI) (p. 337)
- Updating a metric attribution (AWS SDK) (p. 338)

Updating a metric attribution (console)

To update a metric attribution with the Amazon Personalize console, you make your changes on the Metric attribution page.

To update a metric attribution

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign into your account.
2. Choose your dataset group.
3. In the navigation pane, choose Metric attribution.
4. In the bottom section, choose the Metric attributes tab or Metric attribution configuration tab to start making changes.
   - To add or remove metrics, choose the Metric attributes tab and choose Edit attributes. Make your changes on the Edit metric attributes page and choose Update to save your changes.
   - To make changes to the Amazon S3 output bucket or IAM service role, choose the Edit metric attribution configuration tab and make changes on the Edit attribution configuration page. Choose Update to save your changes.

Updating a metric attribution (AWS CLI)

After you create a metric attribution, you can use the AWS Command Line Interface (AWS CLI) to add and remove metrics and modify its output configuration. The following code shows how to remove metrics with the update-metric-attribution command:

```
aws personalize update-metric-attribution
--metric-attribution-arn metric attribution arn
--remove-metrics metricName1 metricName2
```

The following code shows how to add an additional metric and specify a new output configuration:

```
aws personalize update-metric-attribution
--metric-attribution-arn metric attribution arn
--metrics-output-config "{"roleArn": "new role ARN", "s3DataDestination":{"kmsKeyArn ":"kms key ARN","path":"s3://new-bucket-name/new-folder-name/"}}"
--add-metrics "["\"eventType\": \"event type\", \\"expression\": \"SUM(DatasetType.COLUMN_NAME)\", \\
\"metricName\": \"metric name\""]"
```
If successful, Amazon Personalize returns the ARN of the metric attribution you updated. For a complete listing of all parameters, see `UpdateMetricAttribution` *(p. 593)*.

**Updating a metric attribution (AWS SDK)**

After you create a metric attribution, you can add or remove metrics and modify its output configuration. The following code shows how to remove metrics from a metric attribution.

**SDK for Python (Boto3)**

```python
import boto3

personalize = boto3.client('personalize')

metricsToRemove = ["metricName1", "metricName2"]

response = personalize.update_metric_attribution(
    metricAttributionArn = "metric attribution ARN",
    removeMetrics = metricsToRemove
)
```

**SDK for Java 2.x**

```java
public static void removeMetrics(PersonalizeClient client,
                                 String metricAttributionArn,
                                 String metric1Name,
                                 String metric2Name) {

    ArrayList<String> metricsToRemove = new ArrayList<>(Arrays.asList(metric1Name,
                                                                       metric2Name));
    try {
        UpdateMetricAttributionRequest request =
            UpdateMetricAttributionRequest.builder()
                .metricAttributionArn(metricAttributionArn)
                .removeMetrics(metricsToRemove)
                .build();

        UpdateMetricAttributionResponse response =
            client.updateMetricAttribution(request);
        System.out.println(response);
    } catch (PersonalizeException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    }
}
```

**SDK for JavaScript v3**

```javascript
// Get service clients and commands using ES6 syntax.
import {UpdateMetricAttributionCommand, PersonalizeClient } from
    "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

// set the update request param
export const updateMetricAttributionParam = {
    metricAttributionArn: "METRIC_ATTRIBUTION_ARN",    /* required */
```
The following code shows how to add an additional metric and specify a new output configuration:

**SDK for Python (Boto3)**

```python
import boto3

personalize = boto3.client('personalize')

newMetrics = [{
    "eventType": "event type",
    "expression": "SUM(DatasetType.COLUMN_NAME)",
    "metricName": "metric name"
}]

newOutputConfig = {
    "roleArn": "Amazon Personalize service role ARN",
    "s3DataDestination": {
        "kmsKeyArn": "key ARN",
        "path": "s3://<name of your S3 bucket>/<folder>"
    }
}

response = personalize.update_metric_attribution(
    metricAttributionArn = "metric attribution arn",
    metricsOutputConfig = newOutputConfig,
    addMetrics = newMetrics
)
```

**SDK for Java 2.x**

```java
public static void addMetricsAndUpdateOutputConfig(PersonalizeClient personalizeClient,
        String metricAttributionArn,
        String newMetric1EventType,
        String newMetric1Expression,
        String newMetric1Name,
        String newMetric2EventType,
        String newMetric2Expression,
        String newMetric2Name,
        String roleArn,
        String s3Path,
        String kmsKeyArn) {

    try {

        MetricAttribute newAttribute = MetricAttribute.builder()
            .eventType(newMetric1EventType)
            .expression(newMetric1Expression)
```
.metricName(newMetric1Name)
    .build();

    MetricAttribute newAttribute2 = MetricAttribute.builder()
        .eventType(newMetric2EventType)
        .expression(newMetric2Expression)
        .metricName(newMetric2Name)
        .build();

    ArrayList<MetricAttribute> newAttributes = new
    ArrayList<>(Arrays.asList(newAttribute, newAttribute2));

    S3DataConfig newDataDestination = S3DataConfig.builder()
        .kmsKeyArn(kmsKeyArn)
        .path(s3Path)
        .build();

    MetricAttributionOutput newOutputConfig = MetricAttributionOutput.builder()
        .roleArn(roleArn)
        .s3DataDestination(newDataDestination)
        .build();

    UpdateMetricAttributionRequest request =
    UpdateMetricAttributionRequest.builder()
        .metricAttributionArn(metricAttributionArn)
        .metricsOutputConfig(newOutputConfig)
        .addMetrics(newAttributes)
        .build();

    UpdateMetricAttributionResponse response =
    personalizeClient.updateMetricAttribution(request);
    System.out.println("New metrics added!");
    System.out.println(response);

} catch (PersonalizeException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
}

}  

SDK for JavaScript v3

// Get service clients and commands using ES6 syntax.
import {UpdateMetricAttributionCommand, PersonalizeClient } from
    "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

export const updateMetricAttributionParam = {
    metricAttributionArn: "METRIC_ATTRIBUTION_ARN",
    addMetrics: [
        {
            eventType: "EVENT_TYPE",  /* required for each metric */
            expression: "SUM(DatasetType.COLUMN_NAME)",  /* required for each metric */
            metricName: "METRIC_NAME",  /* required for each metric */
        }
    ],
    metricsOutputConfig: {
        roleArn: "ROLE_ARN",         /* required */
        s3DataDestination: {
            kmsKeyArn: "KEY_ARN",  /* required */
            optional */
            path: "s3://<name of your output S3 bucket>/<folderName>/",  /* optional */
        }
    }
};
export const run = async () => {
  try {
    const response = await personalizeClient.send(
      new UpdateMetricAttributionCommand(updateMetricAttributionParam)
    );
    console.log("Success", response);
    return response; // For unit tests.
  } catch (err) {
    console.log("Error", err);
  }
};
run();

If successful, Amazon Personalize returns the ARN of the metric attribution you updated. For a complete listing of all parameters, see `UpdateMetricAttribution (p. 593)`. 

Deleting a metric attribution

If you no longer want to generate reports, you can delete a metric attribution. Deleting a metric attribution deletes all of its metrics and output configuration.

If you delete a metric attribution, Amazon Personalize stops automatically sending reports related to PutEvents and incremental bulk data to CloudWatch. Data already sent to CloudWatch or published to Amazon S3 is not affected. You can delete a metric attribution with the Amazon Personalize console, AWS Command Line Interface, or AWS SDKs.

Topics

- Deleting a metric attribution (console) (p. 341)
- Deleting a metric attribution (AWS CLI) (p. 341)
- Deleting a metric attribution (AWS SDKs) (p. 342)

Deleting a metric attribution (console)

You delete a metric attribution on the overview page for your metric attribution.

To delete a metric attribution

1. Open the Amazon Personalize console at [https://console.aws.amazon.com/personalize/home](https://console.aws.amazon.com/personalize/home) and sign into your account.
2. Choose your dataset group.
3. In the navigation pane, choose Metric attribution.
4. Choose Delete and then confirm the deletion.

Deleting a metric attribution (AWS CLI)

To delete a metric attribution with the AWS CLI, use the delete-metric-attribution command as follows.

```
aws personalize delete-metric-attribution --metric-attribution-arn metric attribution ARN
```
Deleting a metric attribution (AWS SDKs)

The following code shows how to delete a metric attribution with the SDK for Python (Boto3):

**SDK for Python (Boto3)**

import boto3

personalize = boto3.client('personalize')

response = personalize.delete_metric_attribution(
    metricAttributionArn = 'metric attribution ARN'
)

**SDK for Java 2.x**

public static void deleteMetricAttribution(PersonalizeClient client, String metricAttributionArn) {
    try {
        DeleteMetricAttributionRequest request = 
            DeleteMetricAttributionRequest.builder() .
                metricAttributionArn(metricAttributionArn) .
                build();

        DeleteMetricAttributionResponse response = 
            client.deleteMetricAttribution(request);
        if (response.sdkHttpResponse().statusCode() == 200) {
            System.out.println("Metric attribution deleted!")
        } 
    } catch (PersonalizeException e) {
        System.out.println(e.awsErrorDetails().errorMessage());
    } 
}

**SDK for JavaScript v3**

// Get service clients and commands using ES6 syntax.
import { DeleteMetricAttributionCommand, PersonalizeClient } from
    "@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

export const deleteMetricAttributionParam = {
    metricAttributionArn: "METRIC_ATTRIBUTION_ARN",
};

export const run = async () => {
    try {
        const response = await personalizeClient.send(
            new DeleteMetricAttributionCommand(deleteMetricAttributionParam)
        );
        console.log("Success", response);
    } catch (err) {
        console.log("Error", err);
    } 
}
Publishing and viewing results

Amazon Personalize sends the reports for each metric to CloudWatch or Amazon S3:

• For PutEvents data and incremental bulk data, Amazon Personalize automatically sends metrics to
  CloudWatch. For information on viewing and identifying reports in CloudWatch, see Viewing metrics in
  CloudWatch (p. 343).

• For all bulk data, if you provide an Amazon S3 bucket when you create your metric attribution, you can
  choose to publish metric reports to your Amazon S3 bucket each time you create a dataset import job
  for interactions data.

For information publishing metric reports to Amazon S3, see Publishing metrics to Amazon S3 (p. 343).

Topics

• Viewing metrics in CloudWatch (p. 343)
• Publishing metrics to Amazon S3 (p. 343)

Viewing metrics in CloudWatch

Important

After you create a metric attribution and record events or import incremental bulk data, you will
incur some monthly CloudWatch cost per metric. For information about CloudWatch pricing, see the
Amazon CloudWatch pricing page. To stop sending metrics to CloudWatch, delete the metric
attribution (p. 341).

To view metrics in CloudWatch, complete the procedure found in Graphing a metric. The minimum
Period you can graph is 15 minutes. For the search term, specify the name you gave the metric when you
created the metric attribution.

The following is an example of how a metric might appear in CloudWatch. The metric shows the click-
through rate for every 15 minutes for two different recommenders.

Publishing metrics to Amazon S3

To publish metrics to Amazon S3, you provide a path to your Amazon S3 bucket in your metric
attribution. Then you publish reports to Amazon S3 when you create a dataset import job.
When the job completes, you can find the metrics in your Amazon S3 bucket. Each time you publish metrics, Amazon Personalize creates a new file in your Amazon S3 bucket. The file name includes the import method and date as follows:

```
AggregatedAttributionMetrics - ImportMethod - Timestamp.csv
```

The following is an example of how the first few rows of a metric report CSV file might appear. The metric in this example reports on the total clicks from two different recommenders over 15 minute intervals. Each recommender is identified by its Amazon Resource Name (ARN) in the EVENT_ATTRIBUTION_SOURCE column.

```
METRIC_NAME,EVENT_TYPE,VALUE,MATH_FUNCTION,EVENT_ATTRIBUTION_SOURCE,TIMESTAMP
COUNTWATCHES,WATCH,12.0,samplecount,arn:aws:personalize:us-west-2:acctNum: recommender/recommender1Name,1666925124
COUNTWATCHES,WATCH,112.0,samplecount,arn:aws:personalize:us-west-2:acctNum: recommender/recommender2Name,1666924224
COUNTWATCHES,WATCH,10.0,samplecount,arn:aws:personalize:us-west-2:acctNum: recommender/recommender1Name,1666924224
COUNTWATCHES,WATCH,254.0,samplecount,arn:aws:personalize:us-west-2:acctNum: recommender/recommender2Name,1666922424
COUNTWATCHES,WATCH,112.0,samplecount,arn:aws:personalize:us-west-2:acctNum: recommender/recommender1Name,1666922424
COUNTWATCHES,WATCH,100.0,samplecount,arn:aws:personalize:us-west-2:acctNum: recommender/recommender2Name,1666922424
```

Publishing metrics for bulk data to Amazon S3 (console)

To publish metrics to an Amazon S3 bucket with the Amazon Personalize console, create a dataset import job and choose Publish metrics for this import job in Publish event metrics to S3.

Publishing metrics for bulk data to Amazon S3 (AWS CLI)

To publish metrics to an Amazon S3 bucket with the AWS Command Line Interface (AWS CLI), use the following code to create a dataset import job and provide the publishAttributionMetricsToS3 flag. If you don't want to publish metrics for a particular job, omit the flag. For information on each parameter, see CreateDatasetImportJob (p. 448).

```
aws personalize create-dataset-import-job \
  --job-name dataset import job name \
  --dataset-arn dataset arn \
  --data-source dataLocation=s3://bucketname/filename \
  --role-arn roleArn \
  --import-mode INCREMENTAL \
  --publish-attribution-metrics-to-s3
```
Publishing metrics for bulk data to Amazon S3 (AWS SDKs)

To publish metrics to an Amazon S3 bucket with the AWS SDKs, create a dataset import job and set `publishAttributionMetricsToS3` to true. For information on each parameter, see `CreateDatasetImportJob` (p. 448).

SDK for Python (Boto3)

```python
import boto3
personalize = boto3.client('personalize')
response = personalize.create_dataset_import_job(
    jobName = 'YourImportJob',
    datasetArn = 'dataset_arn',
    dataSource = {'dataLocation':'s3://bucket/file.csv'},
    roleArn = 'role_arn',
    importMode = 'INCREMENTAL',
    publishAttributionMetricsToS3 = True
)
dsi_j_arn = response['datasetImportJobArn']
print ('Dataset Import Job arn: ' + dsi_j_arn)
```

SDK for Java 2.x

```java
public static String createPersonalizeDatasetImportJob(PersonalizeClient personalizeClient, String jobName, String datasetArn, String s3BucketPath, String roleArn, ImportMode importMode, boolean publishToS3) {
    long waitInMilliseconds = 60 * 1000;
    String status;
    String datasetImportJobArn;
    try {
        DataSource importDataSource = DataSource.builder()
            .dataLocation(s3BucketPath)
            .build();
        CreateDatasetImportJobRequest createDatasetImportJobRequest = CreateDatasetImportJobRequest.builder()
            .datasetArn(datasetArn)
            .dataSource(importDataSource)
            .jobName(jobName)
            .roleArn(roleArn)
            .importMode(importMode)
            .publishAttributionMetricsToS3(publishToS3)
            .build();
```

345
datasetImportJobArn = personalizeClient.createDatasetImportJob(createDatasetImportJobRequest).datasetImportJobArn();

DescribeDatasetImportJobRequest describeDatasetImportJobRequest = DescribeDatasetImportJobRequest.builder()
    .datasetImportJobArn(datasetImportJobArn).build();

long maxTime = Instant.now().getEpochSecond() + 3 * 60 * 60;
while (Instant.now().getEpochSecond() < maxTime)
{
    DatasetImportJob datasetImportJob = personalizeClient.describeDatasetImportJob(describeDatasetImportJobRequest).datasetImportJob();
    status = datasetImportJob.status();
    System.out.println("Dataset import job status: " + status);
    if (status.equals("ACTIVE") || status.equals("CREATE FAILED"))
    {
        break;
    }
    try {
        Thread.sleep(waitInMilliseconds);
    } catch (InterruptedException e) {
        System.out.println(e.getMessage());
    }
}
return datasetImportJobArn;

} catch (PersonalizeException e) {
    System.out.println(e.awsErrorDetails().errorMessage());
}
return ";";

SDK for JavaScript v3

// Get service clients and commands using ES6 syntax.
import { CreateDatasetImportJobCommand, PersonalizeClient } from 
"@aws-sdk/client-personalize";

// create personalizeClient
const personalizeClient = new PersonalizeClient({
    region: "REGION"
});

// Set the dataset import job parameters.
export const datasetImportJobParam = {
    datasetArn: 'DATASET_ARN', /* required */
    dataSource: {
        dataLocation: 's3://<name of your S3 bucket>/<folderName>/<CSVfilename>.csv' /* required */
    },
    jobName: 'NAME', /* required */
    roleArn: 'ROLE_ARN', /* required */
    importMode: "FULL", /* optional, default is FULL */
    publishAttributionMetricsToS3: true /* set to true to publish metrics to Amazon S3 bucket */
};

export const run = async () => {
    try {
        

Measuring recommendation impact with A/B testing

Performing an A/B test consists of running an experiment with multiple variations and comparing the results. Performing A/B testing with Amazon Personalize recommendations involves showing different groups of users different types of recommendations and then comparing the results. You can use A/B testing to help compare and evaluate different recommendation strategies, and measure the impact of the recommendations.

For example, you might use A/B testing to see if Amazon Personalize recommendations increase click-through rate. To test this scenario, you might show one group of users recommendations that are not personalized, such as featured products. And you might show another group personalized recommendations generated by Amazon Personalize. As your customers interact with items, you can record the outcomes and see which strategy results in the highest click-through rate.

The workflow for performing A/B testing with Amazon Personalize recommendations is as follows:

1. **Plan your experiment** – Define a quantifiable hypothesis, identify business goals, define experiment variations, and determine your experiment time frame.
2. **Split your users** – Split users into two or more groups, with a control group and one or more experiment groups.
3. **Run your experiment** – Show the users in the experiment group modified recommendations. Show the users in the control group recommendations with no changes. Record their interactions with recommendations to track results.
4. **Evaluate results** – Analyze experiment results to determine if the modification made a statistically significant difference for the experiment group.

You can use Amazon CloudWatch Evidently to perform A/B testing with Amazon Personalize recommendations. With CloudWatch Evidently, you can define your experiment, track key performance indicators (KPIs), route recommendation request traffic to the relevant Amazon Personalize resource, and evaluate experiment results. For more information, see A/B testing with CloudWatch Evidently (p. 348).

**Topics**

- A/B testing best practices (p. 347)
- A/B testing with CloudWatch Evidently (p. 348)

**A/B testing best practices**

Use the following best practices to help you design and maintain A/B tests for Amazon Personalize recommendations.
• Identify a quantifiable business goal. Verify that the different recommendations that you want to compare both align with this business goal and are not related to different or non-quantifiable objectives.

• Define a quantifiable hypothesis that aligns with your business goal. For example, you might predict that a promotion for your own custom made content will result in 20% more clicks from these items. Your hypothesis determines the modification that you make for your experiment group.

• Define relevant key performance indicators (KPIs) related to your hypothesis. You use KPIs to measure the outcome of your experiments. These might be the following:
  • Click-through rate
  • Watch time
  • Total price

• Verify that the total number of users in the experiment is large enough to reach a statistically significant result, depending on your hypothesis.

• Define your traffic splitting strategy before you start your experiment. Avoid changing traffic splitting while the experiment is running.

• Keep the user experience of your application or website the same for both your experiment group and control group, except for modifications related to your experiment (for example, model). Variations in user experience, such as the UI or latency, can lead to misleading results.

• Control external factors, such as holidays, ongoing marketing campaigns, and browser limitations. These external factors can lead to misleading results.

• Avoid changing Amazon Personalize recommendations unless directly related to your hypothesis or business requirements. Changes like applying a filter or manually changing the order can lead to misleading results.

• When you evaluate results, make sure that the results are statistically significant before drawing conclusions. The industry standard is a 5% significance level. For more information about statistical significance, see A Refresher on Statistical Significance.

A/B testing with CloudWatch Evidently

After you create a recommender or deploy a custom solution version with a campaign, you can perform A/B tests with Amazon Personalize recommendations and Amazon CloudWatch Evidently. The following video describes the process of using CloudWatch Evidently to perform A/B testing with Amazon Personalize recommendations. For step-by-step instructions, see Performing an A/B test with CloudWatch Evidently (p. 348).

Perform AB Testing with Amazon Personalize and CloudWatch Evidently

Topics

• Performing an A/B test with CloudWatch Evidently (p. 348)
• Sample implementations (p. 350)

Performing an A/B test with CloudWatch Evidently

To perform an A/B test with Amazon Personalize and Amazon CloudWatch Evidently, create a CloudWatch Evidently project, define a feature and its variations, update your application to support your experiment, and create and run the experiment. As the experiment runs, you can view results in CloudWatch Evidently.
To perform an A/B test with Amazon Personalize and CloudWatch Evidently

1. Create a CloudWatch Evidently project. A project is a logical grouping of CloudWatch resources. Within the project, you create features that have variations that you want to test or launch. For step-by-step instructions, see Create a new project in the Amazon CloudWatch User Guide.

2. Add a feature to your project and define its variations. For this experiment, your feature should represent the recommendation scenario that you want to test, such as the click-through rate.

   When you add a feature, specify identifiers to map the different variations of your scenario to Amazon Personalize recommenders or custom campaigns. For each variation, specify the Variation type, such as String, give the variation a name, and give it a value.

   When your experiment runs, your application uses the value of variation to determine what Amazon Personalize resource to use for recommendations. For example, if you're testing two VIDEO_ON_DEMAND recommenders, one created for the Top picks for you use case and one created for the Trending now use case, you might set the following JSON as the Value for each variation.

   ```json
   ```

   You can specify any identifier, as long as your application can use it to identify the relevant resource. For example, you might specify only the name of the recommender or campaign, and construct the Amazon Resource Name (ARN) of the resource in your application.

   For step-by-step instructions to add a feature, see Add a feature to a project in the Amazon CloudWatch User Guide.

3. Update your application to support your experiment:

   - **Feature evaluation** – Use the CloudWatch Evidently EvaluateFeature API operation to assign variations to each user session. The EvaluateFeature response includes the variation value that you specified in the previous step. In this case, it's a JSON object with the type of recommender and it's the ARN of the recommender. Update your recommendation request code to get recommendations from this resource.

     For information about evaluating a feature, see Using EvaluateFeature in the Amazon CloudWatch User Guide.

   - **Record outcomes** – Add code to your application to track results from users' interactions with recommendations.

     To track metrics for your experiments in CloudWatch Evidently, use the CloudWatch Evidently PutProjectEvents API operation to record outcomes for each user. For example, if a user in an experiment clicks a recommended item, you would send details for this event to CloudWatch Evidently.

     For information about sending events to CloudWatch Evidently, see Using PutProjectEvents in the Amazon CloudWatch User Guide.

     To improve Amazon Personalize recommendation relevance, you can record outcome events with the Amazon Personalize PutEvents API operation. If your domain use case or custom recipe supports real-time updates to recommendations, Amazon Personalize can learn from your user's most recent activity and update recommendations as they use your application. If it doesn't support updates, Amazon Personalize uses this data during the next full retraining of your model and then it impacts recommendations.
For information about streaming events to Amazon Personalize, see Recording events (p. 280).

4. Create and start an experiment. When you create an experiment, specify the following:
   - **Feature** – Choose the feature to be tested in the experiment.
   - **Audience** – Configure how many of your users will participate, and configure how to split traffic between feature variations.
   - **Metrics** – Specify the metrics that determine the success of the experiment. For example, the number of clicks.

   After you finish creating the experiment, specify its duration and start the experiment. For step-by-step instructions to create and start experiments in CloudWatch Evidently, see Create an experiment in the Amazon CloudWatch User Guide.

5. As you run your experiment, you can view results in the CloudWatch Evidently experiment dashboard. For information about viewing experiment results, see View experiment results in the dashboard in the Amazon CloudWatch User Guide.

Sample implementations

The following sample implementations show how to implement A/B testing with CloudWatch Evidently.

- For a complete example of real-time APIs that include source code for implementing A/B tests, see Real-Time Personalization APIs in the AWS samples GitHub repository.
- For a sample retail web application that includes a workshop on personalization and A/B testing, see the Retail Demo Store in the AWS samples GitHub repository. For a notebook that describes how to create an A/B experiment with CloudWatch Evidently and the Retail Demo Store, see Retail Demo Store Experimentation Workshop - CloudWatch Evidently.
- For a tutorial that describes how to use A/B testing with CloudWatch Evidently and a sample react application, see Tutorial: A/B testing with the Evidently sample application in the Amazon CloudWatch User Guide.
Personalizing search results from OpenSearch

You can use Amazon Personalize to personalize results from open source OpenSearch or Amazon OpenSearch Service for your users.

OpenSearch is a self-managed, open source search service based on the Apache 2.0 License. Amazon OpenSearch Service is a managed service that helps you deploy, operate, and scale OpenSearch resources in the AWS Cloud. When you use Amazon OpenSearch Service, OpenSearch retrieves and ranks results.

When ranking query results, OpenSearch uses a probabilistic ranking framework called BM-25 to calculate relevance scores. If a distinctive keyword appears more frequently in a document, BM-25 assigns a higher relevance score to that document. OpenSearch ranking doesn’t take into account user behavior like click-through data.

When you use Amazon Personalize with OpenSearch, Amazon Personalize re-ranks OpenSearch results based on a user’s past behavior, any metadata about the items, and any metadata about the user. OpenSearch then incorporates the re-ranking before returning the search response to your application. You control how much weight OpenSearch gives the ranking from Amazon Personalize when applying it to OpenSearch results.

With this re-ranking, results can be more engaging and relevant to a user’s interests. This can lead to an increase in the click-through rate and conversion rate for your application. For a use case example that describes how personalized search can improve results for an ecommerce application, see Use case example (p. 351).

Before you start personalizing OpenSearch results, review the requirements listed in Guidelines and requirements (p. 353).

Topics
• Use case example (p. 351)
• Personalized search workflow (p. 352)
• How the Amazon Personalize Search Ranking plugin works (p. 353)
• Additional information (p. 353)
• Guidelines and requirements (p. 353)
• Setting up OpenSearch and installing the plugin (p. 358)
• Configuring the plugin (p. 360)
• Applying the plugin to OpenSearch queries (p. 362)
• Comparing OpenSearch results with results from the plugin (p. 365)
• Monitoring the plugin (p. 367)

Use case example

When you use Amazon Personalize to re-rank OpenSearch results, the search results can be more relevant for your users. For example, you might have an ecommerce application that sells cars. If your
user enters a query for Toyota cars and you don't personalize results, OpenSearch would return a list of cars made by Toyota based on keywords in your data. This list would be ranked in the same order for all users.

But if you use Amazon Personalize to personalize results, OpenSearch re-ranks these cars in order of relevance for the specific user based on their behavior—for example, their clicks. The car that the user is most likely to click is ranked first.

When you personalize OpenSearch results, you control how much weight (emphasis) OpenSearch gives the ranking from Amazon Personalize. Continuing with this example, if a user searches for a specific type of car from a specific year (such as a 2008 Toyota Prius), you might want to put more emphasis on the original ranking from OpenSearch.

However, for more generic queries that result in a wide range of results (such as a search for all Toyota vehicles), you might put a high emphasis on personalization. This way, the cars at the top of the list are more relevant to the particular user.

Personalized search workflow

To personalize OpenSearch results, you do the following:

1. **Set up Amazon Personalize** – If you haven't already, complete the steps in Setting up Amazon Personalize (p. 11) to set up your credentials and set up permissions for Amazon Personalize. You don't need to set up the AWS SDKs to personalize OpenSearch results.

2. **Complete the Amazon Personalize workflow** – Complete the Amazon Personalize workflow to import data, create a solution with the Personalized-Ranking recipe, train a custom solution version, and deploy it in a campaign. You can only use the Personalized-Ranking recipe. You must create an Interactions dataset. A Users dataset and an Items dataset are optional. For more information, see Amazon Personalize workflow (p. 155).

3. **Set up OpenSearch and install the Amazon Personalize Search Ranking plugin** – If you haven't already, set up your OpenSearch Service domain or open source OpenSearch cluster. Then install the Amazon Personalize Search Ranking plugin. This plugin handles communication with Amazon Personalize and re-ranking results. For more information, see Setting up OpenSearch and installing the plugin (p. 358).

4. **Configure the Amazon Personalize Search Ranking plugin** – To configure the plugin, you create search pipelines. Search pipelines are sets of request and response processors. When you create a pipeline for the plugin, you specify your Amazon Personalize resources in a personalized_search_ranking response processor. You also configure how much weight the plugin gives the results from Amazon Personalize when it re-ranks results. For more information, see Configuring the plugin (p. 360).

5. **Apply the Amazon Personalize Search Ranking plugin to OpenSearch queries** – You can apply the Amazon Personalize Search Ranking plugin to all queries and responses for an OpenSearch index. You can also apply the plugin to individual OpenSearch queries. For more information, see Applying the plugin to OpenSearch queries (p. 362).

6. **Compare results** – The Amazon Personalize Search Ranking plugin re-ranks the search results in the OpenSearch query response. It considers both the ranking from Amazon Personalize and the ranking from OpenSearch. To understand how results are re-ranked, you can compare results from queries that use personalization and those that don't. For more information, see Comparing OpenSearch results with results from the plugin (p. 365).

7. **Monitor the Amazon Personalize Search Ranking plugin** – As you apply the Amazon Personalize Search Ranking plugin to search queries, you can monitor the plugin by getting metrics for your search pipelines. For more information, see Monitoring the plugin (p. 367).
How the Amazon Personalize Search Ranking plugin works

The following diagram shows how the Amazon Personalize Search Ranking plugin works.

1. You submit your customer's query to your OpenSearch Service domain or your open source OpenSearch cluster.
2. OpenSearch sends the query response (list of items that are relevant to the query) and the user's ID to the Amazon Personalize Search Ranking plugin.
3. The plugin sends the items and user in the response to your Amazon Personalize campaign for ranking. It uses the recipe and campaign Amazon Resource Name (ARN) values in your search pipeline to get a personalized ranking for the user. It uses the GetPersonalizedRanking API operation for recommendations. In the request, it passes the userId of the user making the query and the items returned from the OpenSearch query in the inputList.
4. Amazon Personalize returns the re-ranked results to the plugin.
5. The plugin rearranges and returns the search results to your OpenSearch Service domain or open source OpenSearch cluster. It re-ranks the results based on the response from your Amazon Personalize campaign and the emphasis on personalization that you specify during setup.
6. Your open source OpenSearch cluster or OpenSearch Service domain returns the final results to your application.

Additional information

The following resources provide additional information about using OpenSearch.

- For information about getting started with open source OpenSearch, see Quickstart.
- For information about getting started with OpenSearch Service, see Getting started with Amazon OpenSearch Service in the Amazon OpenSearch Service Developer Guide.
- For information about the Personalized-Ranking recipe in Amazon Personalize, see Personalized-Ranking recipe (p. 139).

Guidelines and requirements

This section includes requirements for using the Amazon Personalize Search Ranking plugin. It also describes how to set up permissions for Amazon OpenSearch Service or open source OpenSearch.

Topics

- Plugin requirements (p. 354)
Plugin requirements

Before you start personalizing results from OpenSearch, note the following guidelines and requirements for the Amazon Personalize Search Ranking plugin:

- You must use OpenSearch version 2.9.0 or later. If you use Amazon OpenSearch Service, your domain must use version 2.9 or later.
- If you haven't already, complete the instructions in Setting up permissions (p. 12) to grant your users permission to access Amazon Personalize and give Amazon Personalize permission to access your resources in Amazon Personalize.
- You must be able to access your Amazon Personalize resources from your OpenSearch Service domain or open source OpenSearch cluster.
  - For information about granting access for an OpenSearch Service domain, see Setting up Amazon OpenSearch Service permissions (p. 354).
  - For information about granting access for an OpenSearch cluster, see Setting up open source OpenSearch permissions (p. 356).
- If you use OpenSearch Service, your Amazon Personalize resources must be in the same AWS Region and account as your OpenSearch Service domain.
- You can use only custom Amazon Personalize resources. If you created a Domain dataset group, you can still add custom resources.
- You can only use the custom recipe Personalized-Ranking. For more information about this recipe, see Personalized-Ranking recipe (p. 139).
- You must create an Interactions dataset in Amazon Personalize. Items and Users datasets are optional.
- You can't apply Amazon Personalize filters when you're using the Amazon Personalize Search Ranking plugin.
- By default, the plugin assumes that the _id for an indexed document in OpenSearch matches the itemid in your Amazon Personalize data. If your OpenSearch data uses a different field that corresponds with your Amazon Personalize itemIds, you must specify the name of the field when you configure the plugin.
- The userid that you use for a user making a query must match their userid in the data you import into Amazon Personalize.
- The plugin re-ranks only the top 500 search results from OpenSearch. The remaining items are not re-ranked and end up at the bottom of the list.

Setting up Amazon OpenSearch Service permissions

If you use Amazon OpenSearch Service, you must be able to access your Amazon Personalize resources from your OpenSearch Service domain. To grant access, do the following:

1. Create an IAM service role for OpenSearch Service, and grant it permission to get a personalized ranking from your Amazon Personalize campaign. For more information, see Creating a service role for Amazon OpenSearch Service (p. 355).
2. Grant the user or role that's accessing your OpenSearch Service domain PassRole permissions for the service role you just created. For more information, see Configuring Amazon OpenSearch Service domain security (p. 356).
Creating a service role for Amazon OpenSearch Service

To use the plugin with OpenSearch Service, you must create an IAM service role for OpenSearch Service. This role must have permission to get a personalized ranking from your Amazon Personalize campaign. The following is required to grant your OpenSearch Service service role permission to get a personalized ranking from your Amazon Personalize campaign:

- The role's trust policy must grant AssumeRole permissions for OpenSearch Service. For a trust policy example, see Trust policy example (p. 355).
- The role must have permission to get a personalized ranking from your Amazon Personalize campaign. For a policy example, see Permissions policy example (p. 355).

For information about creating an IAM role, see Creating IAM roles in the IAM User Guide. For information on attaching an IAM policy to role, see Adding and removing IAM Identity permissions in the IAM User Guide.

Topics
- Trust policy example (p. 355)
- Permissions policy example (p. 355)

Trust policy example

The following trust policy example grants AssumeRole permissions for OpenSearch Service.

```json
{
    "Version": "2012-10-17",
    "Statement": [{
        "Sid": "",
        "Effect": "Allow",
        "Action": "sts:AssumeRole",
        "Principal": {
            "Service": ["es.amazonaws.com"]
        }
    }
}
```

Permissions policy example

The following policy example grants the role the minimum permissions to get a personalized ranking from your Amazon Personalize campaign. For Campaign ARN, specify the Amazon Resource Name (ARN) of your Amazon Personalize campaign.

```json
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Action": ["personalize:GetPersonalizedRanking"],
            "Resource": "Campaign ARN"
        }
    ]
}
```
Configuring Amazon OpenSearch Service domain security

To use the plugin with OpenSearch Service, the user or role that’s accessing your domain must have PassRole permissions for the OpenSearch Service service role (p. 355) you just created. Also, the user or role must have permission to perform the es:ESHttpGet and es:ESHttpPut actions.

For information about configuring access to OpenSearch Service, see Security in Amazon OpenSearch Service in the Amazon OpenSearch Service Developer Guide. For policy examples, see Policy examples for OpenSearch Service user or role (p. 356).

Policy examples for OpenSearch Service user or role

The following IAM policy example grants a user or role PassRole permissions for the service role that you created in Creating a service role for Amazon OpenSearch Service (p. 355).

```json
{
   "Version": "2012-10-17",
   "Statement": [
      {
         "Sid": "",
         "Effect": "Allow",
         "Action": ["iam:PassRole"],
         "Resource": "OpenSearch Service service role ARN"
      }
   ]
}
```

The following IAM policy grants the minimum permissions to create pipelines and search queries with OpenSearch Service.

```json
{
   "Version": "2012-10-17",
   "Statement": [
      {
         "Action": ["es:ESHttpGet", "es:ESHttpPut"],
         "Effect": "Allow",
         "Resource": "*",
         "Condition": {
            "ForAnyValue:StringEquals": {
               "aws:ResourceTag/environment": ["production"
            }
         }
      }
   ]
}
```

Setting up open source OpenSearch permissions

If you use open source OpenSearch, you must be able to access your Amazon Personalize resources from your open search cluster. To grant access, do the following:
• If you're setting up OpenSearch from scratch, you can use a quick start bash script to run an OpenSearch cluster in a Docker container. The script uses the default credentials in your AWS profile. You can specify an alternate profile when you run the script.

These credentials must be associated with a user or role that has permission to perform the GetPersonalizedRanking action for your Amazon Personalize campaign. For an example of an IAM policy, see IAM policy examples (p. 357). Alternatively, the credentials must have permission to assume a role that has these permissions. You can provide the Amazon Resource Name (ARN) for this role when you create a pipeline for the Amazon Personalize Search Ranking plugin.

• If you don't use the quick start bash script, you can manually add your credentials to your OpenSearch keystore. These credentials must correspond with a user or role that has permission to perform the GetPersonalizedRanking action for your Amazon Personalize campaign.

To manually add your AWS credentials to your OpenSearch keystore, run the following command where your OpenSearch cluster is running (such as a Docker container). Then provide each credential. If you don't use a session token, you can omit the final line in the command.

```
opensearch-keystore add \npersonalized_search_ranking.aws.access_key \npersonalized_search_ranking.aws.secret_key \npersonalized_search_ranking.aws.session_token
```

• If you run your OpenSearch cluster on an Amazon EC2 instance, you can grant permissions with an IAM instance profile. The policy attached to the role must grant it permission to perform the GetPersonalizedRanking action for your Amazon Personalize campaign. It must also grant Amazon EC2 permissions to assume the role.

For information about Amazon EC2 instance profiles, see Using instance profiles. For a policy example, see IAM policy examples (p. 357).

### IAM policy examples

The following policy example grants a user or role the minimum permissions to get a personalized ranking from your Amazon Personalize campaign. For Campaign ARN, specify the Amazon Resource Name (ARN) of your Amazon Personalize campaign.

```
{
"Version": "2012-10-17",
"Statement": [
{
  "Effect": "Allow",
  "Action": [
    "personalize:GetPersonalizedRanking"
  ],
  "Resource": "Campaign ARN"
}
]
}
```

Additionally, if you run your OpenSearch cluster on an Amazon EC2 instance and grant permissions with an IAM instance profile, the trust policy for the role must grant Amazon EC2 AssumeRole permissions as follows. For information about Amazon EC2 instance profiles, see Using instance profiles.

```
{
"Version": "2012-10-17",
"Statement": [
{
  "Effect": "Allow",
```
Setting up OpenSearch and installing the plugin

The Amazon Personalize Search Ranking plugin handles communication with Amazon Personalize from your OpenSearch Service domain or open source OpenSearch cluster. It also handles re-ranking results. Depending on how you access OpenSearch, you set up OpenSearch and install the plugin as follows:

- If you use Amazon OpenSearch Service, you set up OpenSearch by creating a domain in OpenSearch Service, ingesting data, and installing the plugin.
- If you use open source OpenSearch, you create an OpenSearch cluster, ingest data, and install the plugin.

Topics
- Setting up Amazon OpenSearch Service (p. 358)
- Setting up open source OpenSearch (p. 359)

Setting up Amazon OpenSearch Service

After you complete the Amazon Personalize workflow and meet the requirements listed in Guidelines and requirements (p. 353), you're ready to set up Amazon OpenSearch Service and install the Amazon Personalize Search Ranking plugin.

To set up Amazon OpenSearch Service, you create a domain, ingest your data, and install the plugin. If you have already created a domain and ingested your data, you can skip to step 3.

To set up OpenSearch Service

1. If you haven't already, complete the steps in Setting up Amazon OpenSearch Service permissions (p. 354) so you can access your Amazon Personalize resources from your OpenSearch Service domain.

2. If you haven't already, create an OpenSearch Service domain. An OpenSearch Service domain is synonymous with an open source OpenSearch cluster. Domains are clusters with the settings, instance types, instance counts, and storage resources that you specify.
   - For a concise tutorial for configuring a test domain, see Step 1: Create an Amazon OpenSearch Service domain in the “Getting started” section of the Amazon OpenSearch Service Developer Guide.
   - For more detailed steps, see Creating and managing Amazon OpenSearch Service domains.

3. If you haven't already, ingest your items into OpenSearch Service.
   - For a concise tutorial for uploading a small amount of test data to OpenSearch Service, see Step 2: Upload data to Amazon OpenSearch Service for indexing in the “Getting started” section of the Amazon OpenSearch Service Developer Guide.
   - For complete information about ingesting data, see Indexing data in Amazon OpenSearch Service in the Amazon OpenSearch Service Developer Guide.
4. Associate the Amazon_Personalize_Search_Ranking_Plugin plugin with your domain. The plugin is preinstalled, and you don't have to import it from Amazon S3. You associate the plugin the same way that you associate an OpenSearch Service package.

For information about associating an OpenSearch Service package, see Custom packages for Amazon OpenSearch Service.

After you create a domain, ingest data, and install the Amazon Personalize Search Ranking plugin, you're ready to configure the plugin. You configure it by creating a search pipeline and specifying a personalized_search_ranking response processor. For more information, see Configuring the plugin (p. 360).

Setting up open source OpenSearch

After you complete the Amazon Personalize workflow and meet the requirements listed in Guidelines and requirements (p. 353), you're ready to set up open source and install the Amazon Personalize Search Ranking plugin.

If you already have an OpenSearch cluster running, you can manually install the plugin. If you don’t have a cluster running, you can install OpenSearch and the plugin from scratch with a bash script.

Topics
• Manually installing the plugin on an existing OpenSearch cluster (p. 359)
• Setting up your cluster and installing the plugin with a quickstart script (p. 360)

Manually installing the plugin on an existing OpenSearch cluster

If you already have an OpenSearch cluster, you can manually install the plugin on your cluster directly from the OpenSearch GitHub repository.

To manually install the plugin
1. Use the following command to start your OpenSearch cluster:

   bin/opensearch

2. If you haven’t already, upload your catalog data to your OpenSearch cluster. When you upload your data, you create an OpenSearch index and define your field mappings. Then you upload your data to that index. For an example, see Create an index and field mappings using sample data.

3. Use the following command to install the plugin:

   bin/opensearch-plugin install https://github.com/opensearch-project/search-processor/releases/download/2.9.0/opensearch-search-processor-2.9.0.0.zip

   For more information about installing plugins, see Installing plugins.

After you install the Amazon Personalize Search Ranking plugin, you’re ready to configure it. You configure the plugin by creating a search pipeline and specifying a personalized_search_ranking response processor. For more information, see Configuring the plugin (p. 360).
Setting up your cluster and installing the plugin with a quickstart script

If you haven't created an OpenSearch cluster, you can use a quickstart bash script to create one. This script sets up an OpenSearch cluster in a Docker container, sets up credentials using your default AWS profile, and installs the Amazon Personalize Search Ranking plugin.

For information about manually creating an OpenSearch cluster, see the Quickstart instructions in the OpenSearch documentation.

To install the plugin with a quickstart bash script

1. Before you run the script, download and install Docker Desktop for your operating system.
2. Download the quick start bash script from GitHub.
3. In your working directory, run the script with the following command.

   ```sh
   sh personalized_search_ranking_quickstart.sh
   ```

   With this command, the script uses the credentials in your default AWS profile. To provide an alternate profile, use the --profile argument.

   ```sh
   sh personalized_search_ranking_quickstart.sh --profile profile-name
   ```

   After you run the script, you can find more information about the script in the README file that's located in the unique directory created by the script. This directory stores the Dockerfile and docker-compose.yml files that the script uses. For example: ..../opensearch-personalize-intelligent-ranking-docker.1234/README.

4. Upload your catalog data to your OpenSearch cluster. When you upload your data, you create an OpenSearch index and define your field mappings. Then you upload your data to that index. For an example, see Create an index and field mappings using sample data.

After you set up OpenSearch and install the Amazon Personalize Search Ranking plugin, you're ready to configure it. You configure the plugin by creating a search pipeline and specifying a personalized_search_ranking response processor. For more information, see Configuring the plugin (p. 360).

Configuring the plugin

After you install the Amazon Personalize Search Ranking plugin, you're ready to configure it by creating an OpenSearch search pipeline.

A search pipeline is a set of request and response processors that run sequentially in the order that you create them. When you create a search pipeline for the plugin, you specify a personalized_search_ranking response processor. For information about search pipelines, see Search pipelines.

Topics

- Fields for the personalized_search_ranking response processor (p. 361)
- Creating a pipeline with Amazon OpenSearch Service (p. 361)
- Creating a pipeline with open source OpenSearch (p. 362)
Fields for the personalized_search_ranking response processor

For the personalized_search_ranking response processor, you specify the following fields:

- **campaign_arn (required)** – Specify the Amazon Resource Name (ARN) of the Amazon Personalize campaign to use to personalize results.
- **item_id_field (optional)** – If the _id field for an indexed document in OpenSearch doesn’t correspond with your Amazon Personalize itemIds, specify the name of the field that does. By default, the plugin assumes that the _id data matches the itemId in your Amazon Personalize data.
- **recipe (required)** – Specify the name of the Amazon Personalize recipe to use. You can specify only aws-personalized-ranking.
- **weight (required)** – Specify the emphasis that the response processor puts on personalization when it re-ranks results. Specify a value within a range of 0.0–1.0. The closer to 1.0 that it is, the more likely it is that results from Amazon Personalize rank higher. If you specify 0.0, no personalization occurs and OpenSearch takes precedence.
- **tag (optional)** – Specify an identifier for the processor.
- **iam_role_arn (required for OpenSearch Service, optional for open source OpenSearch)** – For OpenSearch Service, provide the role that you created when setting up permissions (p. 354) for OpenSearch Service to access your Amazon Personalize resources.
  
  For open source OpenSearch, if you use multiple roles to restrict permissions for different groups of users in your organization, specify the ARN of the role that has permission to access to Amazon Personalize. If you use only the AWS credentials in your OpenSearch keystore, you can omit this field.
- **aws_region (required)** – The AWS Region where you created your Amazon Personalize campaign.
- **ignore_failure (optional)** – Specify whether the plugin ignores any processor failures. For values, specify true or false. For your production environments, we recommend that you specify true to avoid any interruptions for query responses. For test environments, you can specify false to view any errors that the plugin generates.

Creating a pipeline with Amazon OpenSearch Service

You can use the following Python code to create a search pipeline with a personalized_search_ranking response processor on an OpenSearch Service domain. Replace domain endpoint with your domain endpoint URL. For example: https://<domain name>.<AWS region>.es-staging.amazonaws.com.

```python
import requests
from requests_auth_aws_sigv4 import AWSSigV4

domain_endpoint = 'domain endpoint'
pipeline_name = 'pipeline name'
url = f'{domain_endpoint}/_search/pipeline/{pipeline_name}'
auth = AWSSigV4('es')
headers = {'Content-Type': 'application/json'}

body = {
    "description": "A pipeline to apply custom re-ranking from Amazon Personalize",
    "response_processors": [
    {
        "personalized_search_ranking": {
            "campaign_arn": "Amazon Personalize Campaign ARN",
            "item_id_field": "productId",
            "recipe": "aws-personalized-ranking",
            ...
        }
    }
]```
Creating a pipeline with open source OpenSearch

You can use the following curl command to create a search pipeline with a personalized_search_ranking response processor on an open source OpenSearch cluster.

```bash
curl -X PUT "http://localhost:9200/_search/pipeline/pipeline-name" -ku 'admin:admin' --insecure -H 'Content-Type: application/json' -d'
{
    "description": "A pipeline to apply custom re-ranking from Amazon Personalize",
    "response_processors": [
    {
        "personalized_search_ranking": {
            "campaign_arn": "Amazon Personalize Campaign ARN",
            "item_id_field": "productId",
            "recipe": "aws-personalized-ranking",
            "weight": "0.3",
            "tag": "personalize-processor",
            "iam_role_arn": "Role ARN",
            "aws_region": "AWS region",
            "ignore_failure": true
        }
    }
    ]
}
```

After you create a search pipeline with a personalized_search_ranking response processor, you're ready to start applying the plugin to OpenSearch queries. You can apply it to an OpenSearch index or an individual OpenSearch query. For more information, see Applying the plugin to OpenSearch queries (p. 362).

Applying the plugin to OpenSearch queries

After you configure a search pipeline with a personalized_search_ranking response processor, you're ready to apply the Amazon Personalize Search Ranking plugin to your OpenSearch queries and view the re-ranked results.

As you apply the plugin to OpenSearch queries, you can monitor the plugin by getting metrics for your search pipeline. For more information, see Monitoring the plugin (p. 367).

Topics
Applying the plugin to Amazon OpenSearch Service queries

You can apply the Amazon Personalize Search Ranking plugin to all queries and responses for an index. You can also apply the plugin to individual queries and responses.

- You can use the following Python code to apply a search pipeline to an index. With this approach, all searches using this index use the plugin to apply personalization to search results.

```python
import requests
from requests_auth_aws_sigv4 import AWSSigV4

domain_endpoint = 'domain endpoint'
index = 'index name'
url = f'{domain_endpoint}/{index}/_settings/
auth = AWSSigV4('es')
headers = {'Content-Type': 'application/json'}
body = {
    "index.search.default_pipeline": "pipeline name"
}
try:
    response = requests.put(url, auth=auth, json=body, headers=headers)
    print(response.text)
except Exception as e:
    print(f"Error: {e}"
```

- You can use the following Python code to apply a search pipeline to an individual query for Toyota brand cars.

Update the code to specify your domain endpoint, your OpenSearch Service index, the name of your pipeline, and your query. For user_id, specify the ID of the user that you're getting search results for. This user must be in the data that you used to create your Amazon Personalize solution version. If the user wasn't present, Amazon Personalize ranks the items based on their popularity.

For context, if you use contextual metadata, provide the user's contextual metadata, such as their device type. The context field is optional. For more information, see Increasing recommendation relevance with contextual metadata (p. 258).

```python
import requests
from requests_auth_aws_sigv4 import AWSSigV4

domain_endpoint = 'domain endpoint'
index = 'index name'
url = f'{domain_endpoint}/{index}/_search/
auth = AWSSigV4('es')
headers = {'Content-Type': 'application/json'}
params = {"search_pipeline": "pipeline-name"}
body = {
    "query": {
        "multi_match": {
            "query": "Toyota",
            "fields": ["BRAND"]
        }
    },
    "ext": {
```
Applying the plugin to queries in open source OpenSearch

You can apply the Amazon Personalize Search Ranking plugin to all queries and responses for an OpenSearch index. You can also apply the plugin to individual OpenSearch queries and responses.

• The following curl command applies a search pipeline to an OpenSearch index in an open source OpenSearch cluster running locally. With this approach, all searches at this index use the plugin to apply personalization to search results.

```bash
curl -XGET "https://localhost:9200/index/_settings" -ku 'admin:admin' --insecure -H 'Content-Type: application/json' -d'
{
  "index.search.default_pipeline" : "pipeline-name"
}
```

• The following curl command applies a search pipeline to an individual query for Toyota brand cars on an index in an open source OpenSearch cluster running locally.

For **user_id**, specify the ID of the user that you’re getting search results for. This user must be in the data that you used to create your Amazon Personalize solution version. If the user wasn’t present, Amazon Personalize ranks the items based on their popularity. For **context**, if you use contextual metadata, provide the user's contextual metadata, such as their device type. The context field is optional. For more information, see [Increasing recommendation relevance with contextual metadata](p. 258).

```bash
{
  "query": {
    "multi_match": {
      "query": "Toyota",
      "fields": ["BRAND"]
    }
  },
  "ext": {
    "personalize_request_parameters": {
      "user_id": "USER ID",
      "context": { "DEVICE" : "mobile phone" }
    }
  }
}
```
To understand how results are re-ranked, you can use OpenSearch Dashboards to compare OpenSearch results against re-ranked results with the plugin. For more information, see Comparing OpenSearch results with results from the plugin (p. 365).

As you apply the plugin to OpenSearch queries, you can monitor the plugin by getting metrics for your OpenSearch pipeline. For more information, see Monitoring the plugin (p. 367).

Comparing OpenSearch results with results from the plugin

The Amazon Personalize Search Ranking plugin rearranges search results based on both the ranking from Amazon Personalize and the ranking from OpenSearch. The way that the plugin re-ranks the results depends on how you configured the personalized_search_ranking response processor in your pipelines.

To understand how results are re-ranked, you can run queries with and without personalization, and compare the results.

Topics

- Comparing results with Amazon OpenSearch Service (p. 365)
- Comparing results with open source OpenSearch (p. 366)

Comparing results with Amazon OpenSearch Service

To understand how results are ranked, you can run queries with and without personalization, and compare the results. You can use the following Python code to run two different queries and output the results to two JSON files. The first method runs a query that uses the plugin to re-rank results. The second runs a method that generates results without personalization.

```python
import json
import requests
from requests_auth_aws_sigv4 import AWSSigV4

# Returns re-ranked OpenSearch results using the Amazon Personalize Search Ranking plugin.
def get_personalized_results(pipeline_name):
    url = f'{domain}/(index)/_search/
    auth = AWSSigV4('es')
    headers = {'Content-Type': 'application/json'}
    params = {'search_pipeline': pipeline_name}
    body = {
        "query": {
            "multi_match": {
                "query": "Toyota",
                "fields": ["BRAND"]
            }
        },
        "ext": {
            "personalize_request_parameters": {
                "user_id": "1"
            }
        }
    }
    try:
        response = requests.post(url, auth=auth, params=params, json=body, headers=headers, verify=False)
```
Comparing results with open source OpenSearch

To understand how results are re-ranked, you can run queries with the Dev Tools console in two separate browser windows. Then you can compare results for queries with and without personalization.
To compare results with the Dev Tools console

1. If you haven't already, follow the steps in Setting up OpenSearch and installing the plugin (p. 358) and Configuring the plugin (p. 360).
2. Make sure OpenSearch Dashboards is installed. The quickstart bash script installs OpenSearch Dashboards. If you don't use the script or already have a cluster running, you must install OpenSearch Dashboards. For more information, see Installing OpenSearch Dashboards.
3. Launch OpenSearch Dashboards. Open http://localhost:5601 from a browser and sign in to OpenSearch Dashboards. The default credentials are username 'admin' and password 'admin'.
5. Open a separate browser window and open the Dev Tools console again. You can use the URL from the previous window.
6. In one window, enter a query that doesn't use any re-ranking for personalization. In the other window, enter a curl command that uses a pipeline with the personalized_search_ranking response processor. If you paste a curl command directly into the console, the command is automatically converted into the format that the console uses. For a command example, see Applying the plugin to OpenSearch queries (p. 362).
7. Run both queries and compare the results.

Monitoring the plugin

If you use OpenSearch Service, you can monitor the plugin through metrics in Amazon CloudWatch. For more information, see Monitoring Amazon OpenSearch Service domains.

As you apply the Amazon Personalize Search Ranking plugin to OpenSearch queries, you can monitor the plugin by getting metrics for your search pipelines. Pipeline metrics include statistics like the number of failed requests for the personalized_search_ranking response processor.

Topics
- Monitoring the plugin with Amazon OpenSearch Service (p. 367)
- Monitoring the plugin with open source OpenSearch (p. 368)
- Pipeline metrics example (p. 368)

Monitoring the plugin with Amazon OpenSearch Service

You can use the following Python code to get metrics for all of your pipelines. For an example of pipeline metrics, see Pipeline metrics example (p. 368).

```python
import requests
from requests_auth_aws_sigv4 import AWSSigV4

domain_endpoint = 'domain endpoint'
url = f'{domain_endpoint}/_nodes/stats/search_pipeline'
auth = AWSSigV4('es')
headers = {'Content-Type': 'application/json'}
try:
    response = requests.get(url, auth=auth, headers=headers, verify=False)
    print(response.text)
except Exception as e:
    print(f"Error: {e}"
```

367
Monitoring the plugin with open source OpenSearch

You can use the following code to get metrics for all of your pipelines. The response contains statistics for all search pipelines. For an example of pipeline metrics, see Pipeline metrics example (p. 368).


Pipeline metrics example

The following code shows an excerpt of the pipeline metrics that are returned from OpenSearch. It shows only the pipelines object that contains statistics for two different pipelines. For each pipeline, you can find Amazon Personalize Search Ranking plugin metrics in the personalized_search_ranking response processor list. For a complete example of all metrics, see Search pipeline metrics.

```json
{
...
"pipelines": {
  "pipelineA": {
    "request": {
      "count": 0,
      "time_in_millis": 0,
      "current": 0,
      "failed": 0
    },
    "response": {
      "count": 6,
      "time_in_millis": 2246,
      "current": 0,
      "failed": 0
    },
    "request_processors": [],
    "response_processors": [
      {
        "personalized_search_ranking": {
          "type": "personalized_search_ranking",
          "stats": {
            "count": <number of requests>,
            "time_in_millis": <time>,
            "current": 0,
            "failed": <number of failed requests>
          }
        }
      }
    ]
  }
  "pipelineB": {
    "request": {
      "count": 0,
      "time_in_millis": 0,
      "current": 0,
      "failed": 0
    },
    "response": {
      "count": 8,
      "time_in_millis": 2248,
      "current": 0,
      "failed": 0
    },
    "request_processors": [],
    "response_processors": [
```
```json
{
    "personalized_search_ranking": {
        "type": "personalized_search_ranking",
        "stats": {
            "count": <number of requests>,
            "time_in_millis": <time>,
            "current": 0,
            "failed": <number of failed requests>
        }
    }
}
```
Tagging Amazon Personalize resources

A tag is a label that you optionally define and associate with AWS resources, including certain types of Amazon Personalize resources. A resource can have as many as 50 tags.

Tags can help you categorize and manage resources in different ways, such as by purpose, environment, or other criteria. For example, you can use tags to split revenue between different functions, or identify development environments for different resources.

To retrieve Amazon Personalize resources by tag, you can use the filters in GetResources operation of Resource Groups Tagging API. For more information, see GetResources in the Resource Groups Tagging API Reference guide.

You can add tags to the following types of Amazon Personalize resources:

- Batch inference jobs
- Batch segment jobs
- Campaigns
- Datasets
- Dataset groups
- Dataset import and export jobs
- Event trackers
- Filters
- Recommenders
- Solutions
- Solution versions

Topics
- Managing tags (p. 370)
- Adding tags to Amazon Personalize resources (p. 371)
- Using tags in IAM policies (p. 375)

Managing tags

Each tag consists of a required tag key and an optional tag value, both of which you define. A tag key is a general label that acts as a category for more specific tag values. A tag value acts as a descriptor for a tag key.

For example, if you have two versions of an Amazon Personalize dataset group (one for internal testing and another for production), you might assign an Environment tag key to both projects. The tag value of the Environment tag might be Test for one version of the dataset group and Production for the other version.
The following restrictions apply to tags:

- Maximum number of tags per resource – 50
- Maximum key length – 128 Unicode characters in UTF-8
- Maximum value length – 256 Unicode characters in UTF-8
- Tag keys and values can contain the following characters: A-Z, a-z, 0-9, space, and _ . : / = + @ – (hyphen). This is the standard set of characters available across AWS services that support tags. Some services support additional symbols.
- Tag keys and tag values are case sensitive.
- For each associated resource, each tag key must be unique and it can have only one tag value.
- Your tag keys and tag values can't start with aws:. AWS services apply tags that start with aws:, and those tags can't be modified. They don't count towards tag limits.
- You can't update or delete a resource based only on its tags. You must also specify the Amazon Resource Name (ARN) or resource ID, depending on the operation that you use.

Additional information

For more information about tagging, see the following resources.

- AWS Tagging Principles in the AWS General Reference
- AWS Tagging Strategies (downloadable PDF)
- AWS Access Control in the AWS IAM User Guide
- AWS Tagging Policies in the AWS Organizations User Guide

Adding tags to Amazon Personalize resources

You can add, display, update, and remove tag keys and values from Amazon Personalize resources with the Amazon Personalize console, AWS Command Line Interface (AWS CLI), or AWS SDKs. The following examples show how to add a tag to Amazon Personalize dataset group. You can add tags to other Amazon Personalize resources in the same way: Either with the Tags option in the console or Tags parameter with the AWS CLI or SDKs.

Topics

- Adding tags (console) (p. 371)
- Adding tags (AWS CLI) (p. 372)
- Adding tags (AWS SDKs) (p. 372)

Adding tags (console)

When you create a resource in Amazon Personalize, you can add optional tags with the Amazon Personalize console. The following example adds a tag to a dataset group.

To add tags to a new dataset group

1. Open the Amazon Personalize console at https://console.aws.amazon.com/personalize/home and sign in to your account.
2. Choose Create dataset group.
3. For Name, enter a name.
4. **For Domain**, choose a domain.
5. Expand the **Tags** section and choose **Add new tag**.
6. **For Key and Value**, enter appropriate values.
   - For example, **Environment** and **Test**, respectively.
7. To add more tags, choose **Add new tag**.
   - You can add up to 50 tags to a resource.
8. Choose **Next** to continue creating your resource.

Adding tags to an existing resource is similar: Choose your resource and use the **Tags** fields to add your tags.

### Adding tags (AWS CLI)

You can use the AWS Command Line Interface (AWS CLI) to add tags when you create a resource or add tags to an existing resource.

**Topics**
- Adding tags when you create a resource (p. 372)
- Adding tags to an existing resource (p. 372)

#### Adding tags when you create a resource

To create a new resource and add a tag to it with the AWS CLI, use the appropriate `create` command for the resource and include the `tags` parameter and values. For example, the following command creates a new Domain dataset group named `myDatasetGroup` for the ECOMMERCE domain, and adds the following tags: An `Environment` tag key with a `Test` tag value, and a `Owner` tag key and a `xyzCorp` value.

```bash
aws personalize create-dataset-group \
  --name myDatasetGroup \
  --domain ECOMMERCE \
  --tags tagKey=Environment,tagValue=Test tagKey=Owner,tagValue=xyzCorp
```

For information about the commands that you can use to create an Amazon Personalize resource, see the [Amazon Personalize AWS CLI Command Reference](#).

#### Adding tags to an existing resource

To add a tag to an existing resource, use the `tag-resource` command and specify the ARN of the resource and provide the tag key and value in the `tags-model` parameter.

```bash
aws personalize tag-resource \
  --resource-arn resource ARN \
  --tags tagKey=key,tagValue=value
```

### Adding tags (AWS SDKs)

You can use the AWS SDKs to add tags when you create a resource, or to add tags to an existing resource.

**Topics**

• **Adding tags when you create a resource (p. 373)**
• **Adding tags to an existing resource (p. 374)**

Adding tags when you create a resource

To create a new resource and add a tag to it with the AWS SDKs, use the appropriate create method. Use the tags parameter to specify the key-value pairs for each of your tags. For example, the following code creates a new Domain dataset group named myDatasetGroup for the ECOMMERCE domain and adds the following tags: An Environment tag key with a Test tag value, and a Owner tag key and a xyzCorp value.

SDK for Python (Boto3)

```python
import boto3
personalize = boto3.client('personalize')
response = personalize.create_dataset_group(
    name = 'myDatasetGroup'
    domain = 'ECOMMERCE'
    tags = [
        {'tagKey': 'Environment', 'tagValue': 'Test'},
        {'tagKey': 'Owner', 'tagValue': 'xyzCorp'}
    ]
)dsg_arn = response['datasetGroupArn']
description = personalize.describe_dataset_group(datasetGroupArn = dsg_arn)
print('Name: ' + description['name'])
print('ARN: ' + description['datasetGroupArn'])
print('Status: ' + description['status'])
```

SDK for Java 2.x

```java
public static String createDomainDatasetGroup(PersonalizeClient personalizeClient,
                                            String datasetGroupName,
                                            String domain) {
    try {
        ArrayList<Tag> tags = new ArrayList<>();
        Tag tag1 = Tag.builder()
            .tagKey("Environment")
            .tagValue("Test")
            .build();
        tags.add(tag1);
        Tag tag2 = Tag.builder()
            .tagKey("Owner")
            .tagValue("xyzCorp")
            .build();
        tags.add(tag2);
        //...
    }
    //...
}
```
Adding tags to an existing resource

The following code shows how to add a tag to an existing resource. Specify the Amazon Resource Name (ARN) of the resource that you want to add tags to and specify key-value pairs for each of your tags.

SDK for Python (Boto3)

```python
import boto3
personalize = boto3.client('personalize')
add_tags_response = personalize.tag_resource(
    resourceArn = "resourceArn",
    tags = [
    
    
    ]
)
```

SDK for Java 2.x

```java
public static void tagResource(PersonalizeClient personalizeClient, 
String resourceArn, 
String domain) {

    try {

        ArrayList<Tag> tagList = new ArrayList<>();

        Tag tag1 = Tag.builder()
            .tagKey("Environment")
            .tagValue("Test")
            .build();
        tags.add(tag1);
        Tag tag2 = Tag.builder()
            .tagKey("Owner")
            .tagValue("xyzCorp")
            .build();
        tags.add(tag2);

        TagResourceRequest tagResourceRequest = TagResourceRequest.builder()
            .resourceArn(resourceArn)
```
Using tags in IAM policies

After you start implementing tags, you can apply tag-based, resource-level permissions to AWS Identity and Access Management (IAM) policies and API operations. This includes operations that support adding tags to resources when resources are created. By using tags in this way, you can implement granular control of which groups and users in your AWS account have permission to create and tag resources, and which groups and users have permission to create, update, and remove tags more generally.

For example, you can create a policy that allows a user to have full access to all of the Amazon Personalize resources where their name is a value in the `Owner` tag for the resource.

```json
{
   "Version": "2012-10-17",
   "Statement": [
      {
         "Sid": "ModifyResourceIfOwner",
         "Effect": "Allow",
         "Action": "personalize:*",
         "Resource": "*",
         "Condition": {
            "StringEqualsIgnoreCase": {
               "aws:ResourceTag/Owner": "${aws:username}"
            }
         }
      }
   ]
}
```

The following example shows how to create a policy to allow creating and deleting a dataset. These operations are allowed only if the user name is johndoe.

```json
{
   "Version": "2012-10-17",
   "Statement": [
      {
         "Effect": "Allow",
         "Action": [
            "personalize:CreateDataset",
            "personalize:DeleteDataset"
         ],
         "Resource": "arn:aws:personalize::*:dataset/*",
         "Condition": {
            "StringEquals": {"aws:username": "johndoe"}
         }
      },
      {
         "Effect": "Allow",
         "Action": "personalize:DescribeDataset",
         "Resource": "arn:aws:personalize:*:dataset/*",
         "Condition": {
            "StringEquals": {"aws:username": "johndoe"}
         }
      }
   ]
}
```
If you define tag-based, resource-level permissions, the permissions take effect immediately. This means that your resources are more secure as soon as they're created, and you can quickly start enforcing the use of tags for new resources. You can also use resource-level permissions to control which tag keys and values can be associated with new and existing resources. For more information, see Controlling Access Using Tags in the AWS IAM User Guide.
Troubleshooting

The following topics provide answers to common questions and troubleshooting advice for error messages that you might encounter with Amazon Personalize. For a quick reference to help you determine if Amazon Personalize fits your use case, see the Amazon Personalize Cheat Sheet in the Amazon Personalize samples repository.

Topics
- Frequently asked questions (p. 377)
- Error messages (p. 380)

Frequently asked questions

The following are answers to frequently asked questions related to importing data, training, model deployment, recommendations, and filters in Amazon Personalize.

For more questions and answers, see the Amazon Personalize Cheat Sheet in the Amazon Personalize samples repository.

Topics
- Data import and management (p. 377)
- Creating a custom solution and solution version (p. 378)
- Model deployment (custom campaigns) (p. 378)
- Recommendations (p. 379)
- Filtering recommendations (p. 379)

Data import and management

What format should my bulk data be in?

Your bulk data must be in comma-separated values (CSV) format. The first row of your CSV file must contain column headers. The column headers in your CSV file need to map to the schema to create the dataset. If your data includes any non-ASCII encoded characters, your CSV file must be encoded in UTF-8 format. Don't enclose headers in quotation marks ("). TIMESTAMP and CREATION_TIMESTAMP data must be in UNIX epoch time format. For more information on timestamp data, see Timestamp data (p. 103). For more information about schemas, see Schemas (p. 82).

For complete data format guidelines, see Data format guidelines (p. 101). If you're not sure how to format your data, you can use Amazon SageMaker Data Wrangler (Data Wrangler) to prepare your data. For more information, see Preparing and importing data using Amazon SageMaker Data Wrangler (p. 160).

How much training data do I need?

For all use cases (Domain dataset groups) and custom recipes, your interactions data must have the following:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
• At minimum 25 unique user IDs with at least two interactions for each.

For quality recommendations, we recommend that you have at minimum 50,000 interactions from at least 1,000 users with two or more interactions each.

You can start out with an empty Interactions dataset and, when you have recorded enough data, create your recommender (Domain dataset group) or custom solution version using only new recorded events. Some recipes and use cases may have additional data requirements. For information on use case requirements, see Choosing a use case (p. 107). For information on recipe requirements, see Choosing a recipe (p. 113).

How do I update an item or user's attributes?

Use the Amazon Personalize console or the PutItems (p. 600) or PutUsers (p. 602) operations to import an item or user with the same item ID but with the modified attributes.

How do I delete an item or user?

Amazon Personalize doesn't support deleting a specific item or user. To make sure that an item or user doesn't appear in recommendations, use a filter to exclude items. For more information, see Filtering recommendations and user segments (p. 311).

How do I delete a schema?

You can delete a schema only with the DeleteSchema (p. 491) operation. You can't use the Amazon Personalize console to delete a schema.

Creating a custom solution and solution version

What recipe should I use?

The Amazon Personalize recipe that you use depends on your use case. For information on matching use cases to recipes, see Choosing a recipe (p. 113). The Amazon Personalize Cheat Sheet also includes use case and recipe information.

How often should I retrain my model?

Retraining helps keep your recommendations relevant as your catalog grows and users interact with items. Retraining frequency depends on your business requirements and the recipe that you use. For most workloads, we recommend creating a new solution version weekly with training mode set to FULL. This creates a new solution version based on the entirety of the training data from the datasets in your dataset group.

For more information, see Maintaining recommendation relevance (p. 278).

Should I use AutoML?

No, instead we recommend that you match your use case to different Amazon Personalize recipes and choose a recipe. For information on matching use cases to recipes, see Choosing a recipe (p. 113).

Model deployment (custom campaigns)

How do I set a maximum transaction throughput for a campaign?

You can only set the minimum throughput for a campaign. When you create an Amazon Personalize campaign, you specify a dedicated transaction capacity for creating real-time recommendations for your application users. If your TPS increases beyond minProvisionedTPS, Amazon Personalize auto-scales the provisioned capacity up and down, but never below the minProvisionedTPS. For more information, see Minimum provisioned transactions per second and auto-scaling (p. 238).
How do I monitor the cost of my campaigns?

The Amazon Personalize Monitor project provides a CloudWatch dashboard, custom metrics, utilization alarms, and cost optimization functions for Amazon Personalize campaigns. See the Amazon Personalize Monitor in the Amazon Personalize samples repository.

Recommendations

How can I tell if my Amazon Personalize model is generating quality recommendations?

Evaluate the performance of your solution version with offline and online metrics (see Evaluating a solution version with metrics) and online testing (such as A/B testing). For more information about testing, see Using A/B testing to measure the efficacy of recommendations generated by Amazon Personalize.

How do I delete my batch inference job and why is its status “active”?

You can't delete batch inference jobs. When a batch inference job's status is active, the job is complete. You can access your recommendations in the output Amazon S3 bucket or folder. You won't incur additional cost from the batch inference job once the job is complete. However you may incur additional charges from other services such as Amazon S3 for input and output data storage.

Why does my SIMS-backed campaign recommend items that are not similar based on metadata?

SIMS uses your Interactions dataset to determine similarity; not item metadata such as color or price. SIMS identifies the co-occurrence of the item in user histories in your Interaction dataset to recommend similar items. For more information, see SIMS recipe.

Can I get more than 500 items from a single GetRecommendations API operation?

500 is the maximum number of items that you can retrieve in a single GetRecommendations (p. 608). This value cannot be increased.

Filtering recommendations

Why aren't my recommendations filtered as expected?

This can occur for a variety of reasons:

- There may be issue with the format or syntax of your filter expression. For examples of correctly formatted filter expressions, see Filter expression examples.
- Amazon Personalize considers up to 100 of the most recent interactions per user per event type. This is an adjustable quota. You can request a quota increase using the Service Quotas console.

For more information, see Filtering recommendations and user segments.

How can I remove already purchased items from recommendations?

For ECOMMERCE Domain dataset groups, if you create a recommender with the Recommended for you (p. 112) or Customers who viewed X also viewed (p. 112) use case, Amazon Personalize automatically filters items the user purchased based on the userid that you specify and Purchase events.

For other Domain dataset group use cases or custom resources, use a filter to remove purchased items. Add a Purchased event type attribute to your data, record Purchase events with the PutItems operation, and create a filter that removes purchased items from recommendations. For example:

EXCLUDE ItemID WHERE Interactions.EVENT_TYPE IN ("purchased")
Error messages

The following sections list and explain some of the messages that you might encounter when using Amazon Personalize.

Topics

- Data import and management (p. 380)
- Creating a solution and solution version (custom resources) (p. 381)
- Model deployment (custom campaigns) (p. 381)
- Recommenders (Domain dataset groups) (p. 381)
- Recommendations (p. 381)
- Filtering recommendations (p. 382)

Data import and management

Error message: Invalid Data location.

Make sure you used the correct syntax for your Amazon S3 bucket location. For dataset import jobs, use the following syntax for the location of your data in Amazon S3:

s3://<name of your S3 bucket>/<folder path>/<CSVfilename>

If your CSV files are in a folder and you want to upload multiple files with one dataset import job, use this syntax without the CSV file name.

Error message: An error occurred (LimitExceededException) when calling the CreateDatasetImportJob operation: More than 5 resources with PENDING or IN_PROGRESS status.

You can have a total of 5 pending or in progress dataset import jobs per region. This quota is not adjustable. For a complete list of quotas for Amazon Personalize, see Amazon Personalize endpoints and quotas (p. 415).

Error message: Failed to create a data import job for <dataset type> dataset...Insufficient privileges for accessing data in Amazon S3.

Give Amazon Personalize access to your Amazon S3 resources by attaching access policies to your Amazon S3 bucket and your Amazon Personalize service role. See Giving Amazon Personalize access to Amazon S3 resources (p. 16).

If you use AWS Key Management Service (AWS KMS) for encryption, you must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

Error message: Failed to create a data import job <dataset type> dataset...Input CSV is missing the following columns:[COLUMN_NAME, COLUMN_NAME].

The data that you import into Amazon Personalize, including attribute names and data types, must match the destination dataset's schema. For more information, see Schemas (p. 82).

Error message: Length cannot be more than <character limit> characters for <COLLumn_NAME>. If no values exceed the character limit, make sure your data follows the formatting guidelines listed in https://docs.aws.amazon.com/personalize/latest/dg/data-prep-formatting.html.
Check to make sure all values in this column don't exceed the character limit. If no values exceed the character limit, check any preceding textual fields for the following:

- Make sure any textual data is wrapped in double quotes. Use the \ character to escape any double quotes or \ characters in your data.
- Makes sure each record in your CSV file is on a single line.

Creating a solution and solution version (custom resources)

**Error message:** Create failed. Dataset has fewer than 25 users with at least 2 interactions each.

You must import more data before you can train the model. The minimum data requirements to train a model are:

- At minimum 1000 interactions records from users interacting with items in your catalog. These interactions can be from bulk imports, or streamed events, or both.
- At minimum 25 unique user IDs with at least two interactions for each.

For real-time recommendations, import more data with a dataset import job or record more interaction events for your users with an event tracker and the PutEvents (p. 598) operation. For more information on recording real-time events, see Recording events (p. 280).

For batch recommendations, import your data with a dataset import job when you have more data. For more information, about importing bulk data see Step 2: Preparing and importing data (p. 159).

Model deployment (custom campaigns)

**Error:** Cannot create a campaign. More than 5 resources in ACTIVE state. Please delete some and try again.

You can have a total of 5 active Amazon Personalize campaigns per dataset group. This quota is adjustable and you can request a quota increase using the Service Quotas console. For a complete list of limits and quotas for Amazon Personalize, see Amazon Personalize endpoints and quotas (p. 415).

Recommenders (Domain dataset groups)

**Error:** Dataset has fewer than 1000 interactions after filtering by event type: <event type>

Different use cases require different event types. Your data must have at minimum 1000 events with the required type for your use case. For more information, see Choosing a use case (p. 107)

Recommendations

**Batch inference job error message:** Invalid S3 input path or Invalid S3 output path

Make sure you use the correct syntax for your Amazon S3 input or output locations. Also make sure that your output location is different from your input data. It should be a folder in the same Amazon S3 bucket or a different bucket.

Use the following syntax for the input file location in Amazon S3: `s3://<name of your S3 bucket>/<folder name>/<input JSON file name>`

Use the following syntax for the output folder in Amazon S3: `s3://<name of your S3 bucket>/<output folder name>/`
Filtering recommendations

Error message: Could not create filter. Invalid input symbol: $parameterName. Placeholders are not allowed with NOT_IN operator.

You can't use placeholder parameters in a filter expression that uses the NOT_IN operator. Instead, use the IN operator and use the opposite Action: use Include instead of Exclude (or the reverse).

For example, if you want to use INCLUDE ItemID WHERE Items.GENRE NOT IN ($GENRE), you can use EXCLUDE ItemID WHERE Items.GENRE IN ($GENRE) and get the same results.

For more information about filters, see Filter expression elements (p. 313).

Error message: Could not create filter. Invalid Expression... when filtering on Boolean type fields

You can't create filter expressions that filter using values with a Boolean type in your schema. To filter based on Boolean values, use a schema with a field of type String and use the values True and False in your data. Or you can use type int or long and values 0 and 1.

For more information about filters, see Filter expression elements (p. 313).
Specifying resources with AWS CloudFormation

Amazon Personalize is integrated with AWS CloudFormation, a service that helps you to model and set up your AWS resources so that you can spend less time creating and managing your resources and infrastructure. You create a template that describes all the AWS resources that you can specify (such as Amazon Personalize dataset groups). AWS CloudFormation then provisions and configures those resources for you.

When you use AWS CloudFormation, you can reuse your template to set up your Amazon Personalize resources consistently and repeatedly. Describe your resources once, and then provision the same resources over and over in multiple AWS accounts and Regions.

Topics
- Amazon Personalize and AWS CloudFormation templates (p. 383)
- Example AWS CloudFormation templates for Amazon Personalize resources (p. 383)
- Learn more about AWS CloudFormation (p. 386)

Amazon Personalize and AWS CloudFormation templates

To provision and configure resources for Amazon Personalize and related services, you must understand AWS CloudFormation templates. Templates are formatted text files in JSON or YAML. These templates describe the resources that you want to provision in your AWS CloudFormation stacks. If you're unfamiliar with JSON or YAML, you can use AWS CloudFormation Designer to help you get started with AWS CloudFormation templates. For more information, see What is AWS CloudFormation Designer? in the AWS CloudFormation User Guide.

Amazon Personalize supports specifying datasets, dataset groups, dataset import jobs, schemas, and solutions in AWS CloudFormation. For more information, see the Amazon Personalize resource type reference in the AWS CloudFormation User Guide.

Example AWS CloudFormation templates for Amazon Personalize resources

The following AWS CloudFormation template examples show you how to specify different Amazon Personalize resources.

Topics
- CreateDatasetGroup (p. 384)
- CreateDataset (p. 384)
- CreateSchema (p. 385)
CreateDatasetGroup

**JSON**

```json
{
  "AWSTemplateFormatVersion": "2010-09-09",
  "Resources": {
    "MyDatasetGroup": {
      "Type": "AWS::Personalize::DatasetGroup",
      "Properties": {
        "Name": "my-dataset-group-name"
      }
    }
  }
}
```

**YAML**

```yaml
AWSTemplateFormatVersion: 2010-09-09
Resources:
  MyDatasetGroup:
    Type: 'AWS::Personalize::DatasetGroup'
    Properties:
      Name: my-dataset-group-name
```

CreateDataset

**JSON**

```json
{
  "AWSTemplateFormatVersion": "2010-09-09",
  "Resources": {
    "MyDataset": {
      "Type": "AWS::Personalize::Dataset",
      "Properties": {
        "Name": "my-dataset-name",
        "DatasetType": "Interactions",
        "DatasetImportJob": {
          "JobName": "my-import-job-name",
          "DataSource": {
            "DataLocation": "s3://bucket-name/file-name.csv"
          },
          "RoleArn": "arn:aws:iam:123456789012:role/personalize-role"
        }
      }
    }
  }
}
```

**YAML**

```yaml
AWSTemplateFormatVersion: 2010-09-09
Resources:
  MyDataset:
    Type: 'AWS::Personalize::Dataset'
    Properties:
      Name: my-dataset-name
      DatasetType: Interactions
      DatasetImportJob:
        JobName: my-import-job-name
        DataSource:
          DataLocation: s3://bucket-name/file-name.csv
        RoleArn: arn:aws:iam:123456789012:role/personalize-role
```
CreateSchema

JSON

```
{
  "AWSTemplateFormatVersion": "2010-09-09",
  "Resources": {
    "MySchema": {
      "Type": "AWS::Personalize::Schema",
      "Properties": {
        "Name": "my-schema-name",
        "Schema": "{"type": "record","name": "Interactions", "namespace": "com.amazonaws.personalize.schema", "fields": [ { "name": "USER_ID", "type": "string" }, { "name": "ITEM_ID", "type": "string" }, { "name": "TIMESTAMP", "type": "long"} ], "version": "1.0"}"
      }
    }
  }
}
```

YAML

```
AWSTemplateFormatVersion: 2010-09-09
Resources:
  MySchema:
    Type: AWS::Personalize::Schema
    Properties:
      Name: "my-schema-name"
      Schema: >-
        {"type": "record","name": "Interactions", "namespace": "com.amazonaws.personalize.schema", "fields": [ [ "name": "USER_ID", "type": "string" ], [ "name": "ITEM_ID", "type": "string" ], [ "name": "TIMESTAMP", "type": "long"], "version": "1.0"]}
```

CreateSolution

JSON

```
{
  "AWSTemplateFormatVersion": "2010-09-09",
  "Resources": {
    "MySolution": {
      "Type": "AWS::Personalize::Solution",
      "Properties": {
        "DatasetImportJob": {
          "JobName": "my-import-job-name",
          "DataSource": {
            "DataLocation": "s3://bucket-name/file-name.csv",
            "RoleArn": "arn:aws:iam::123456789012:role/personalize-role"
          }
        }
      }
    }
  }
}
```
Learn more about AWS CloudFormation

To learn more about AWS CloudFormation, see the following resources:

- AWS CloudFormation
- AWS CloudFormation user guide
- AWS CloudFormation API reference
- AWS CloudFormation Command Line Interface user guide
Security in Amazon Personalize

Cloud security at AWS is the highest priority. As an AWS customer, you benefit from a data center and network architecture that is built to meet the requirements of the most security-sensitive organizations.

Security is a shared responsibility between AWS and you. The shared responsibility model describes this as security of the cloud and security in the cloud:

• Security of the cloud – AWS is responsible for protecting the infrastructure that runs AWS services in the AWS Cloud. AWS also provides you with services that you can use securely. Amazon Personalize uses data encryption to protect your data. For more information see Data encryption (p. 388). Third-party auditors regularly test and verify the effectiveness of our security as part of the AWS Compliance Programs. To learn about the compliance programs that apply to Amazon Personalize, see AWS Services in Scope by Compliance Program.

• Security in the cloud – Your responsibility is determined by the AWS service that you use. You are also responsible for other factors including the sensitivity of your data, your company’s requirements, and applicable laws and regulations.

This documentation helps you understand how to apply the shared responsibility model when using Amazon Personalize. The following topics show you how to configure Amazon Personalize to meet your security and compliance objectives. You also learn how to use other AWS services that help you to monitor and secure your Amazon Personalize resources.

Topics

• Data protection in Amazon Personalize (p. 387)
• Identity and Access Management for Amazon Personalize (p. 389)
• Logging and monitoring in Amazon Personalize (p. 404)
• Compliance validation for Amazon Personalize (p. 411)
• Resilience in Amazon Personalize (p. 412)
• Infrastructure security in Amazon Personalize (p. 412)
• Amazon Personalize and interface VPC endpoints (AWS PrivateLink) (p. 413)

Data protection in Amazon Personalize

The AWS shared responsibility model applies to data protection in Amazon Personalize. As described in this model, AWS is responsible for protecting the global infrastructure that runs all of the AWS Cloud. You are responsible for maintaining control over your content that is hosted on this infrastructure. This content includes the security configuration and management tasks for the AWS services that you use. For more information about data privacy, see the Data Privacy FAQ. For information about data protection in Europe, see the AWS Shared Responsibility Model and GDPR blog post on the AWS Security Blog.

For data protection purposes, we recommend that you protect AWS account credentials and set up individual users with AWS IAM Identity Center or AWS Identity and Access Management (IAM). That way, each user is given only the permissions necessary to fulfill their job duties. We also recommend that you secure your data in the following ways:
• Use multi-factor authentication (MFA) with each account.
• Use SSL/TLS to communicate with AWS resources. We require TLS 1.2 and recommend TLS 1.3.
• Set up API and user activity logging with AWS CloudTrail.
• Use AWS encryption solutions, along with all default security controls within AWS services.
• Use advanced managed security services such as Amazon Macie, which assists in discovering and securing sensitive data that is stored in Amazon S3.
• If you require FIPS 140-2 validated cryptographic modules when accessing AWS through a command line interface or an API, use a FIPS endpoint. For more information about the available FIPS endpoints, see Federal Information Processing Standard (FIPS) 140-2.

We strongly recommend that you never put confidential or sensitive information, such as your customers’ email addresses, into tags or free-form text fields such as a Name field. This includes when you work with Amazon Personalize or other AWS services using the console, API, AWS CLI, or AWS SDKs. Any data that you enter into tags or free-form text fields used for names may be used for billing or diagnostic logs. If you provide a URL to an external server, we strongly recommend that you do not include credentials information in the URL to validate your request to that server.

Data encryption

The following information explains where Amazon Personalize uses data encryption to protect your data.

Encryption at rest

Any data stored within Amazon Personalize is always encrypted at rest with Amazon Personalize managed AWS Key Management Service (AWS KMS) keys. If you provide your own AWS KMS key during resource creation, Amazon Personalize uses the key to encrypt your data and store it. For example, if you provide a AWS KMS ARN in the CreateDatasetGroup (p. 444) operation, Amazon Personalize uses the key to encrypt and store data you import into any datasets that you create in that dataset group.

You must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).

For information about data encryption in Amazon S3 see Protecting data using encryption in the Amazon Simple Storage Service User Guide. For information about managing your own AWS KMS key, see Managing keys in the AWS Key Management Service Developer Guide.

Encryption in transit

Amazon Personalize uses TLS with AWS certificates to encrypt any data sent to other AWS services. Any communication with other AWS services happens over HTTPS, and Amazon Personalize endpoints support only secure connections over HTTPS.

Amazon Personalize copies data out of your account and processes it in an internal AWS system. When processing data, Amazon Personalize encrypts data with either a Amazon Personalize AWS KMS key or any AWS KMS key you provide.

Key management

AWS manages any default AWS KMS keys. It is your responsibility to manage any AWS KMS keys that you own. You must grant Amazon Personalize and your Amazon Personalize IAM service role permission to use your key. For more information, see Giving Amazon Personalize permission to use your AWS KMS key (p. 20).
Identity and Access Management for Amazon Personalize

AWS Identity and Access Management (IAM) is an AWS service that helps an administrator securely control access to AWS resources. IAM administrators control who can be authenticated (signed in) and authorized (have permissions) to use Amazon Personalize resources. IAM is an AWS service that you can use with no additional charge.

Topics

- Audience (p. 389)
- Authenticating with identities (p. 389)
- Managing access using policies (p. 392)
- How Amazon Personalize works with IAM (p. 393)
- Cross-service confused deputy prevention (p. 398)
- Identity-based policy examples for Amazon Personalize (p. 399)
- Troubleshooting Amazon Personalize identity and access (p. 403)

Audience

How you use AWS Identity and Access Management (IAM) differs, depending on the work that you do in Amazon Personalize.

Service user – If you use the Amazon Personalize service to do your job, then your administrator provides you with the credentials and permissions that you need. As you use more Amazon Personalize features to do your work, you might need additional permissions. Understanding how access is managed can help you request the right permissions from your administrator. If you cannot access a feature in Amazon Personalize, see Troubleshooting Amazon Personalize identity and access (p. 403).

Service administrator – If you’re in charge of Amazon Personalize resources at your company, you probably have full access to Amazon Personalize. It’s your job to determine which Amazon Personalize features and resources your service users should access. You must then submit requests to your IAM administrator to change the permissions of your service users. Review the information on this page to understand the basic concepts of IAM. To learn more about how your company can use IAM with Amazon Personalize, see How Amazon Personalize works with IAM (p. 393).

IAM administrator – If you’re an IAM administrator, you might want to learn details about how you can write policies to manage access to Amazon Personalize. To view example Amazon Personalize identity-based policies that you can use in IAM, see Identity-based policy examples for Amazon Personalize (p. 399).

Authenticating with identities

Authentication is how you sign in to AWS using your identity credentials. You must be authenticated (signed in to AWS) as the AWS account root user, as an IAM user, or by assuming an IAM role.
You can sign in to AWS as a federated identity by using credentials provided through an identity source. AWS IAM Identity Center (IAM Identity Center) users, your company's single sign-on authentication, and your Google or Facebook credentials are examples of federated identities. When you sign in as a federated identity, your administrator previously set up identity federation using IAM roles. When you access AWS by using federation, you are indirectly assuming a role.

Depending on the type of user you are, you can sign in to the AWS Management Console or the AWS access portal. For more information about signing in to AWS, see How to sign in to your AWS account in the AWS Sign-In User Guide.

If you access AWS programmatically, AWS provides a software development kit (SDK) and a command line interface (CLI) to cryptographically sign your requests by using your credentials. If you don't use AWS tools, you must sign requests yourself. For more information about using the recommended method to sign requests yourself, see Signing AWS API requests in the IAM User Guide.

Regardless of the authentication method that you use, you might be required to provide additional security information. For example, AWS recommends that you use multi-factor authentication (MFA) to increase the security of your account. To learn more, see Multi-factor authentication in the AWS IAM Identity Center User Guide and Using multi-factor authentication (MFA) in AWS in the IAM User Guide.

**AWS account root user**

When you create an AWS account, you begin with one sign-in identity that has complete access to all AWS services and resources in the account. This identity is called the AWS account root user and is accessed by signing in with the email address and password that you used to create the account. We strongly recommend that you don't use the root user for your everyday tasks. Safeguard your root user credentials and use them to perform the tasks that only the root user can perform. For the complete list of tasks that require you to sign in as the root user, see Tasks that require root user credentials in the IAM User Guide.

**Federated identity**

As a best practice, require human users, including users that require administrator access, to use federation with an identity provider to access AWS services by using temporary credentials.

A federated identity is a user from your enterprise user directory, a web identity provider, the AWS Directory Service, the Identity Center directory, or any user that accesses AWS services by using credentials provided through an identity source. When federated identities access AWS accounts, they assume roles, and the roles provide temporary credentials.

For centralized access management, we recommend that you use AWS IAM Identity Center. You can create users and groups in IAM Identity Center, or you can connect and synchronize to a set of users and groups in your own identity source for use across all your AWS accounts and applications. For information about IAM Identity Center, see What is IAM Identity Center? in the AWS IAM Identity Center User Guide.

**IAM users and groups**

An IAM user is an identity within your AWS account that has specific permissions for a single person or application. Where possible, we recommend relying on temporary credentials instead of creating IAM users who have long-term credentials such as passwords and access keys. However, if you have specific use cases that require long-term credentials with IAM users, we recommend that you rotate access keys. For more information, see Rotate access keys regularly for use cases that require long-term credentials in the IAM User Guide.

An IAM group is an identity that specifies a collection of IAM users. You can't sign in as a group. You can use groups to specify permissions for multiple users at a time. Groups make permissions easier to
manage for large sets of users. For example, you could have a group named IAMAdmins and give that group permissions to administer IAM resources.

Users are different from roles. A user is uniquely associated with one person or application, but a role is intended to be assumable by anyone who needs it. Users have permanent long-term credentials, but roles provide temporary credentials. To learn more, see When to create an IAM user (instead of a role) in the IAM User Guide.

IAM roles

An IAM role is an identity within your AWS account that has specific permissions. It is similar to an IAM user, but is not associated with a specific person. You can temporarily assume an IAM role in the AWS Management Console by switching roles. You can assume a role by calling an AWS CLI or AWS API operation or by using a custom URL. For more information about methods for using roles, see Using IAM roles in the IAM User Guide.

IAM roles with temporary credentials are useful in the following situations:

- **Federated user access** – To assign permissions to a federated identity, you create a role and define permissions for the role. When a federated identity authenticates, the identity is associated with the role and is granted the permissions that are defined by the role. For information about roles for federation, see Creating a role for a third-party Identity Provider in the IAM User Guide. If you use IAM Identity Center, you configure a permission set. To control what your identities can access after they authenticate, IAM Identity Center correlates the permission set to a role in IAM. For information about permissions sets, see Permission sets in the AWS IAM Identity Center User Guide.

- **Temporary IAM user permissions** – An IAM user or role can assume an IAM role to temporarily take on different permissions for a specific task.

- **Cross-account access** – You can use an IAM role to allow someone (a trusted principal) in a different account to access resources in your account. Roles are the primary way to grant cross-account access. However, with some AWS services, you can attach a policy directly to a resource (instead of using a role as a proxy). To learn the difference between roles and resource-based policies for cross-account access, see How IAM roles differ from resource-based policies in the IAM User Guide.

- **Cross-service access** – Some AWS services use features in other AWS services. For example, when you make a call in a service, it's common for that service to run applications in Amazon EC2 or store objects in Amazon S3. A service might do this using the calling principal's permissions, using a service role, or using a service-linked role.

- **Principal permissions** – When you use an IAM user or role to perform actions in AWS, you are considered a principal. Policies grant permissions to a principal. When you use some services, you might perform an action that then triggers another action in a different service. In this case, you must have permissions to perform both actions. To see whether an action requires additional dependent actions in a policy, see Actions, resources, and condition keys for Amazon Personalize in the Service Authorization Reference.

- **Service role** – A service role is an IAM role that a service assumes to perform actions on your behalf. An IAM administrator can create, modify, and delete a service role from within IAM. For more information, see Creating a role to delegate permissions to an AWS service in the IAM User Guide.

- **Service-linked role** – A service-linked role is a type of service role that is linked to an AWS service. The service can assume the role to perform an action on your behalf. Service-linked roles appear in your AWS account and are owned by the service. An IAM administrator can view, but not edit the permissions for service-linked roles.

- **Applications running on Amazon EC2** – You can use an IAM role to manage temporary credentials for applications that are running on an EC2 instance and making AWS CLI or AWS API requests. This is preferable to storing access keys within the EC2 instance. To assign an AWS role to an EC2 instance and make it available to all of its applications, you create an instance profile that is attached to the instance. An instance profile contains the role and enables programs that are running on the EC2 instance to get temporary credentials. For more information, see Using an IAM role to grant permissions to applications running on Amazon EC2 instances in the IAM User Guide.
Managing access using policies

You control access in AWS by creating policies and attaching them to AWS identities or resources. A policy is an object in AWS that, when associated with an identity or resource, defines their permissions. AWS evaluates these policies when a principal (user, root user, or role session) makes a request. Permissions in the policies determine whether the request is allowed or denied. Most policies are stored in AWS as JSON documents. For more information about the structure and contents of JSON policy documents, see Overview of JSON policies in the IAM User Guide.

Administrators can use AWS JSON policies to specify who has access to what. That is, which principal can perform actions on what resources, and under what conditions.

By default, users and roles have no permissions. To grant users permission to perform actions on the resources that they need, an IAM administrator can create IAM policies. The administrator can then add the IAM policies to roles, and users can assume the roles.

IAM policies define permissions for an action regardless of the method that you use to perform the operation. For example, suppose that you have a policy that allows the iam:GetRole action. A user with that policy can get role information from the AWS Management Console, the AWS CLI, or the AWS API.

Identity-based policies

Identity-based policies are JSON permissions policy documents that you can attach to an identity, such as an IAM user, group of users, or role. These policies control what actions users and roles can perform, on which resources, and under what conditions. To learn how to create an identity-based policy, see Creating IAM policies in the IAM User Guide.

Identity-based policies can be further categorized as inline policies or managed policies. Inline policies are embedded directly into a single user, group, or role. Managed policies are standalone policies that you can attach to multiple users, groups, and roles in your AWS account. Managed policies include AWS managed policies and customer managed policies. To learn how to choose between a managed policy or an inline policy, see Choosing between managed policies and inline policies in the IAM User Guide.

Resource-based policies

Resource-based policies are JSON policy documents that you attach to a resource. Examples of resource-based policies are IAM role trust policies and Amazon S3 bucket policies. In services that support resource-based policies, service administrators can use them to control access to a specific resource. For the resource where the policy is attached, the policy defines what actions a specified principal can perform on that resource and under what conditions. You must specify a principal in a resource-based policy. Principals can include accounts, users, roles, federated users, or AWS services.

Resource-based policies are inline policies that are located in that service. You can't use AWS managed policies from IAM in a resource-based policy.

Access control lists (ACLs)

Access control lists (ACLs) control which principals (account members, users, or roles) have permissions to access a resource. ACLs are similar to resource-based policies, although they do not use the JSON policy document format.

Amazon S3, AWS WAF, and Amazon VPC are examples of services that support ACLs. To learn more about ACLs, see Access control list (ACL) overview in the Amazon Simple Storage Service Developer Guide.
Other policy types

AWS supports additional, less-common policy types. These policy types can set the maximum permissions granted to you by the more common policy types.

- Permissions boundaries – A permissions boundary is an advanced feature in which you set the maximum permissions that an identity-based policy can grant to an IAM entity (IAM user or role). You can set a permissions boundary for an entity. The resulting permissions are the intersection of an entity's identity-based policies and its permissions boundaries. Resource-based policies that specify the user or role in the Principal field are not limited by the permissions boundary. An explicit deny in any of these policies overrides the allow. For more information about permissions boundaries, see Permissions boundaries for IAM entities in the IAM User Guide.

- Service control policies (SCPs) – SCPs are JSON policies that specify the maximum permissions for an organization or organizational unit (OU) in AWS Organizations. AWS Organizations is a service for grouping and centrally managing multiple AWS accounts that your business owns. If you enable all features in an organization, then you can apply service control policies (SCPs) to any or all of your accounts. The SCP limits permissions for entities in member accounts, including each AWS account root user. For more information about Organizations and SCPs, see How SCPs work in the AWS Organizations User Guide.

- Session policies – Session policies are advanced policies that you pass as a parameter when you programmatically create a temporary session for a role or federated user. The resulting session's permissions are the intersection of the user or role's identity-based policies and the session policies. Permissions can also come from a resource-based policy. An explicit deny in any of these policies overrides the allow. For more information, see Session policies in the IAM User Guide.

Multiple policy types

When multiple types of policies apply to a request, the resulting permissions are more complicated to understand. To learn how AWS determines whether to allow a request when multiple policy types are involved, see Policy evaluation logic in the IAM User Guide.

How Amazon Personalize works with IAM

Before you use IAM to manage access to Amazon Personalize, learn what IAM features are available to use with Amazon Personalize.

<table>
<thead>
<tr>
<th>IAM feature</th>
<th>Amazon Personalize support</th>
</tr>
</thead>
<tbody>
<tr>
<td>Identity-based policies (p. 394)</td>
<td>Yes</td>
</tr>
<tr>
<td>Resource-based policies (p. 394)</td>
<td>No</td>
</tr>
<tr>
<td>Policy actions (p. 395)</td>
<td>Yes</td>
</tr>
<tr>
<td>Policy resources (p. 395)</td>
<td>Yes</td>
</tr>
<tr>
<td>Policy condition keys (service-specific) (p. 396)</td>
<td>Yes</td>
</tr>
<tr>
<td>ACLs (p. 396)</td>
<td>No</td>
</tr>
<tr>
<td>ABAC (tags in policies) (p. 397)</td>
<td>Yes</td>
</tr>
</tbody>
</table>
IAM feature | Amazon Personalize support
--- | ---
Temporary credentials (p. 397) | Yes
Principal permissions (p. 397) | Yes
Service roles (p. 398) | Yes
Service-linked roles (p. 398) | No

To get a high-level view of how Amazon Personalize and other AWS services work with most IAM features, see [AWS services that work with IAM](#) in the IAM User Guide.

### Identity-based policies for Amazon Personalize

supports identity-based policies | Yes

Identity-based policies are JSON permissions policy documents that you can attach to an identity, such as an IAM user, group of users, or role. These policies control what actions users and roles can perform, on which resources, and under what conditions. To learn how to create an identity-based policy, see [Creating IAM policies](#) in the IAM User Guide.

With IAM identity-based policies, you can specify allowed or denied actions and resources as well as the conditions under which actions are allowed or denied. You can't specify the principal in an identity-based policy because it applies to the user or role to which it is attached. To learn about all of the elements that you can use in a JSON policy, see [IAM JSON policy elements reference](#) in the IAM User Guide.

### Identity-based policy examples for Amazon Personalize

To view examples of Amazon Personalize identity-based policies, see [Identity-based policy examples for Amazon Personalize](p. 399).

Resource-based policies within Amazon Personalize

supports resource-based policies | No

Resource-based policies are JSON policy documents that you attach to a resource. Examples of resource-based policies are IAM role trust policies and Amazon S3 bucket policies. In services that support resource-based policies, service administrators can use them to control access to a specific resource. For the resource where the policy is attached, the policy defines what actions a specified principal can perform on that resource and under what conditions. You must specify a principal in a resource-based policy. Principals can include accounts, users, roles, federated users, or AWS services.

To enable cross-account access, you can specify an entire account or IAM entities in another account as the principal in a resource-based policy. Adding a cross-account principal to a resource-based policy is only half of establishing the trust relationship. When the principal and the resource are in different AWS accounts, an IAM administrator in the trusted account must also grant the principal entity (user or role) permission to access the resource. They grant permission by attaching an identity-based policy to the entity. However, if a resource-based policy grants access to a principal in the same account, no additional identity-based policy is required. For more information, see [How IAM roles differ from resource-based policies](#) in the IAM User Guide.
Policy actions for Amazon Personalize

<table>
<thead>
<tr>
<th>Supports policy actions</th>
<th>Yes</th>
</tr>
</thead>
</table>

Administrators can use AWS JSON policies to specify who has access to what. That is, which principal can perform actions on what resources, and under what conditions.

The Action element of a JSON policy describes the actions that you can use to allow or deny access in a policy. Policy actions usually have the same name as the associated AWS API operation. There are some exceptions, such as permission-only actions that don't have a matching API operation. There are also some operations that require multiple actions in a policy. These additional actions are called dependent actions.

Include actions in a policy to grant permissions to perform the associated operation.

To see a list of Amazon Personalize actions, see Actions defined by Amazon Personalize in the Service Authorization Reference.

Policy actions in Amazon Personalize use the following prefix before the action:

personalize

To specify multiple actions in a single statement, separate them with commas.

"Action": [
  "personalize:action1",
  "personalize:action2"
]

You can specify multiple actions using wildcards (*). For example, to specify all actions that begin with the word Describe, include the following action:

"Action": "personalize:Describe*"

To view examples of Amazon Personalize identity-based policies, see Identity-based policy examples for Amazon Personalize (p. 399).

Policy resources for Amazon Personalize

<table>
<thead>
<tr>
<th>Supports policy resources</th>
<th>Yes</th>
</tr>
</thead>
</table>

Administrators can use AWS JSON policies to specify who has access to what. That is, which principal can perform actions on what resources, and under what conditions.

The Resource JSON policy element specifies the object or objects to which the action applies. Statements must include either a Resource or a NotResource element. As a best practice, specify a resource using its Amazon Resource Name (ARN). You can do this for actions that support a specific resource type, known as resource-level permissions.
For actions that don't support resource-level permissions, such as listing operations, use a wildcard (*) to indicate that the statement applies to all resources.

"Resource": "*

To see a list of Amazon Personalize resource types and their ARNs, see Resources defined by Amazon Personalize in the Service Authorization Reference. To learn with which actions you can specify the ARN of each resource, see Actions defined by Amazon Personalize.

To view examples of Amazon Personalize identity-based policies, see Identity-based policy examples for Amazon Personalize (p. 399).

Policy condition keys for Amazon Personalize

| Supports service-specific policy condition keys | Yes |

Administrators can use AWS JSON policies to specify who has access to what. That is, which principal can perform actions on what resources, and under what conditions.

The Condition element (or Condition block) lets you specify conditions in which a statement is in effect. The Condition element is optional. You can create conditional expressions that use condition operators, such as equals or less than, to match the condition in the policy with values in the request.

If you specify multiple Condition elements in a statement, or multiple keys in a single Condition element, AWS evaluates them using a logical AND operation. If you specify multiple values for a single condition key, AWS evaluates the condition using a logical OR operation. All of the conditions must be met before the statement's permissions are granted.

You can also use placeholder variables when you specify conditions. For example, you can grant an IAM user permission to access a resource only if it is tagged with their IAM user name. For more information, see IAM policy elements: variables and tags in the IAM User Guide.

AWS supports global condition keys and service-specific condition keys. To see all AWS global condition keys, see AWS global condition context keys in the IAM User Guide.

To see a list of Amazon Personalize condition keys, see Condition keys for Amazon Personalize in the Service Authorization Reference. To learn with which actions and resources you can use a condition key, see Actions defined by Amazon Personalize.

To view examples of Amazon Personalize identity-based policies, see Identity-based policy examples for Amazon Personalize (p. 399).

ACLs in Amazon Personalize

| Supports ACLs | No |

Access control lists (ACLs) control which principals (account members, users, or roles) have permissions to access a resource. ACLs are similar to resource-based policies, although they do not use the JSON policy document format.
ABAC with Amazon Personalize

| Supports ABAC (tags in policies) | Yes |

Attribute-based access control (ABAC) is an authorization strategy that defines permissions based on attributes. In AWS, these attributes are called tags. You can attach tags to IAM entities (users or roles) and to many AWS resources. Tagging entities and resources is the first step of ABAC. Then you design ABAC policies to allow operations when the principal's tag matches the tag on the resource that they are trying to access.

ABAC is helpful in environments that are growing rapidly and helps with situations where policy management becomes cumbersome.

To control access based on tags, you provide tag information in the condition element of a policy using the `aws:ResourceTag/key-name`, `aws:RequestTag/key-name`, or `aws:TagKeys` condition keys.

If a service supports all three condition keys for every resource type, then the value is Yes for the service. If a service supports all three condition keys for only some resource types, then the value is Partial.

For more information about ABAC, see What is ABAC? in the IAM User Guide. To view a tutorial with steps for setting up ABAC, see Use attribute-based access control (ABAC) in the IAM User Guide.

For more information about tagging Amazon Personalize resources, see Tagging Amazon Personalize resources (p. 370).

To view an example identity-based policy for limiting access to a resource based on the tags on that resource, see Using tags in IAM policies (p. 375).

Using temporary credentials with Amazon Personalize

| Supports temporary credentials | Yes |

Some AWS services don't work when you sign in using temporary credentials. For additional information, including which AWS services work with temporary credentials, see AWS services that work with IAM in the IAM User Guide.

You are using temporary credentials if you sign in to the AWS Management Console using any method except a user name and password. For example, when you access AWS using your company's single sign-on (SSO) link, that process automatically creates temporary credentials. You also automatically create temporary credentials when you sign in to the console as a user and then switch roles. For more information about switching roles, see Switching to a role (console) in the IAM User Guide.

You can manually create temporary credentials using the AWS CLI or AWS API. You can then use those temporary credentials to access AWS. AWS recommends that you dynamically generate temporary credentials instead of using long-term access keys. For more information, see Temporary security credentials in IAM.

Cross-service principal permissions for Amazon Personalize

| Supports principal permissions | Yes |
When you use an IAM user or role to perform actions in AWS, you are considered a principal. Policies grant permissions to a principal. When you use some services, you might perform an action that then triggers another action in a different service. In this case, you must have permissions to perform both actions. To see whether an action requires additional dependent actions in a policy, see Actions, resources, and condition keys for Amazon Personalize in the Service Authorization Reference.

### Service roles for Amazon Personalize

| Supports service roles | Yes |

A service role is an IAM role that a service assumes to perform actions on your behalf. An IAM administrator can create, modify, and delete a service role from within IAM. For more information, see Creating a role to delegate permissions to an AWS service in the IAM User Guide.

**Warning**

Changing the permissions for a service role might break Amazon Personalize functionality. Edit service roles only when Amazon Personalize provides guidance to do so.

### Service-linked roles for Amazon Personalize

| Supports service-linked roles | No |

A service-linked role is a type of service role that is linked to an AWS service. The service can assume the role to perform an action on your behalf. Service-linked roles appear in your AWS account and are owned by the service. An IAM administrator can view, but not edit the permissions for service-linked roles.

For details about creating or managing service-linked roles, see AWS services that work with IAM. Find a service in the table that includes a Yes in the Service-linked role column. Choose the Yes link to view the service-linked role documentation for that service.

### Cross-service confused deputy prevention

The confused deputy problem is a security issue where an entity that doesn't have permission to perform an action can coerce a more-privileged entity to perform the action. In AWS, cross-service impersonation can result in the confused deputy problem. Cross-service impersonation can occur when one service (the calling service) calls another service (the called service). The calling service can be manipulated to use its permissions to act on another customer's resources in a way it should not otherwise have permission to access. To prevent this, AWS provides tools that help you protect your data for all services with service principals that have been given access to resources in your account.

We recommend using the aws:SourceArn and aws:SourceAccount global condition context keys in resource policies to limit the permissions that Amazon Personalize gives another service to the resource.

To prevent the confused deputy problem in roles assumed by Amazon Personalize, in the role's trust policy set the value of aws:SourceArn to arn:aws:personalize:region:accountNumber:*.*. The wildcard (*) applies the condition for all Amazon Personalize resources.

The following trust relationship policy grants Amazon Personalize access to your resources and uses the aws:SourceArn and aws:SourceAccount global condition context keys to prevent the confused deputy problem. Use this policy when you create a role for Amazon Personalize (Creating an IAM role for Amazon Personalize (p. 15)).

```
Identity-based policy examples for Amazon Personalize

By default, users and roles don't have permission to create or modify Amazon Personalize resources. They also can't perform tasks by using the AWS Management Console, AWS Command Line Interface (AWS CLI), or AWS API. To grant users permission to perform actions on the resources that they need, an IAM administrator can create IAM policies. The administrator can then add the IAM policies to roles, and users can assume the roles.

To learn how to create an IAM identity-based policy by using these example JSON policy documents, see Creating IAM policies in the IAM User Guide.

For details about actions and resource types defined by Amazon Personalize, including the format of the ARNs for each of the resource types, see Actions, resources, and condition keys for Amazon Personalize in the Service Authorization Reference.

Topics

- Policy best practices (p. 399)
- AWS managed policies (p. 400)
- Using the Amazon Personalize console (p. 401)
- Allow users to view their own permissions (p. 401)
- Allowing full access to Amazon Personalize resources (p. 401)
- Allowing read-only access to Amazon Personalize resources (p. 402)

Policy best practices

Identity-based policies determine whether someone can create, access, or delete Amazon Personalize resources in your account. These actions can incur costs for your AWS account. When you create or edit identity-based policies, follow these guidelines and recommendations:

- Get started with AWS managed policies and move toward least-privilege permissions – To get started granting permissions to your users and workloads, use the AWS managed policies that grant

```json
"Version": "2012-10-17",
"Statement": [
  {
    "Sid": "",
    "Effect": "Allow",
    "Principal": {
      "Service": [
        "personalize.amazonaws.com"
      ]
    },
    "Action": "sts:AssumeRole",
    "Condition": {
      "StringEquals": {
        "aws:SourceAccount": "accountNumber"
      },
      "StringLike": {
        "aws:SourceArn": "arn:aws:personalize:region:accountNumber:*"
      }
    }
  }
]
```
permissions for many common use cases. They are available in your AWS account. We recommend that you reduce permissions further by defining AWS customer managed policies that are specific to your use cases. For more information, see AWS managed policies or AWS managed policies for job functions in the IAM User Guide.

- **Apply least-privilege permissions** – When you set permissions with IAM policies, grant only the permissions required to perform a task. You do this by defining the actions that can be taken on specific resources under specific conditions, also known as least-privilege permissions. For more information about using IAM to apply permissions, see Policies and permissions in IAM in the IAM User Guide.

- **Use conditions in IAM policies to further restrict access** – You can add a condition to your policies to limit access to actions and resources. For example, you can write a policy condition to specify that all requests must be sent using SSL. You can also use conditions to grant access to service actions if they are used through a specific AWS service, such as AWS CloudFormation. For more information, see IAM JSON policy elements: Condition in the IAM User Guide.

- **Use IAM Access Analyzer to validate your IAM policies to ensure secure and functional permissions** – IAM Access Analyzer validates new and existing policies so that the policies adhere to the IAM policy language (JSON) and IAM best practices. IAM Access Analyzer provides more than 100 policy checks and actionable recommendations to help you author secure and functional policies. For more information, see IAM Access Analyzer policy validation in the IAM User Guide.

- **Require multi-factor authentication (MFA)** – If you have a scenario that requires IAM users or a root user in your AWS account, turn on MFA for additional security. To require MFA when API operations are called, add MFA conditions to your policies. For more information, see Configuring MFA-protected API access in the IAM User Guide.

For more information about best practices in IAM, see Security best practices in IAM in the IAM User Guide.

### AWS managed policies

AWS managed policies are policies that are created and managed by AWS. The following are examples of AWS managed policies you might use when working with Amazon Personalize.

#### AmazonPersonalizeFullAccess Policy

You can use the AWS managed AmazonPersonalizeFullAccess policy to give users the following permissions:

- Access all Amazon Personalize resources
- Publish and list metrics on Amazon CloudWatch
- List, read, write, and delete all objects in an Amazon S3 bucket that contains Personalize or personalize in the bucket name
- Pass a role to Amazon Personalize

AmazonPersonalizeFullAccess provides more permissions than are necessary. We recommend creating a new IAM policy that only grants the necessary permissions (see Giving Amazon Personalize permission to access your resources (p. 14)).

#### CloudWatchFullAccess

To give your users permission to monitor Amazon Personalize with CloudWatch, attach the CloudWatchFullAccess policy to your role. For more information, see Monitoring Amazon Personalize (p. 404).

The CloudWatchFullAccess policy is optional and grants permission for the following actions:
• Publish and list Amazon Personalize metrics in CloudWatch
• View metrics and metric statistics.
• Set metric based alarms.

Using the Amazon Personalize console

To access the Amazon Personalize console, you must have a minimum set of permissions. These permissions must allow you to list and view details about the Amazon Personalize resources in your AWS account. If you create an identity-based policy that is more restrictive than the minimum required permissions, the console won't function as intended for entities (users or roles) with that policy.

You don't need to allow minimum console permissions for users that are making calls only to the AWS CLI or the AWS API. Instead, allow access to only the actions that match the API operation that they're trying to perform.

Allow users to view their own permissions

This example shows how you might create a policy that allows IAM users to view the inline and managed policies that are attached to their user identity. This policy includes permissions to complete this action on the console or programmatically using the AWS CLI or AWS API.

```
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Sid": "ViewOwnUserInfo",
            "Effect": "Allow",
            "Action": [
                "iam:GetUserPolicy",
                "iam:ListGroupsForUser",
                "iam:ListAttachedUserPolicies",
                "iam:ListUserPolicies",
                "iam:GetUser"
            ],
            "Resource": ["arn:aws:iam::*:user/${aws:username}"],
            "Resource": ["arn:aws:iam::*:user/${aws:username}"],
            "Resource": ["arn:aws:iam::*:user/${aws:username}"],
            "Resource": ["arn:aws:iam::*:user/${aws:username}"],
            "Resource": ["arn:aws:iam::*:user/${aws:username}"],
        },
        {
            "Sid": "NavigateInConsole",
            "Effect": "Allow",
            "Action": [
                "iam:GetGroupPolicy",
                "iam:GetPolicyVersion",
                "iam:GetPolicy",
                "iam:ListAttachedGroupPolicies",
                "iam:ListGroupPolicies",
                "iam:ListPolicyVersions",
                "iam:ListPolicies",
                "iam:ListUsers"
        },
        "Resource": ["*"]
    ]
}
```

Allowing full access to Amazon Personalize resources

The following example gives an IAM user in your AWS account full access to all Amazon Personalize resources and actions.
Identity-based policy examples

```json
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Action": [
                "personalize:*"
            ],
            "Resource": "*"
        }
    ]
}
```

Allowing read-only access to Amazon Personalize resources

In this example, you grant an IAM user in your AWS account read-only access to your Amazon Personalize resources, including Amazon Personalize datasets, dataset groups, solutions, and campaigns.

```json
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Action": [
                "personalize:DescribeAlgorithm",
                "personalize:DescribeBatchInferenceJob",
                "personalize:DescribeBatchSegmentJob",
                "personalize:DescribeCampaign",
                "personalize:DescribeDataset",
                "personalize:DescribeDatasetExportJob",
                "personalize:DescribeDatasetGroup",
                "personalize:DescribeDatasetImportJob",
                "personalize:DescribeEventTracker",
                "personalize:DescribeFeatureTransformation",
                "personalize:DescribeFilter",
                "personalize:DescribeRecipe",
                "personalize:DescribeRecommender",
                "personalize:DescribeSchema",
                "personalize:DescribeSolution",
                "personalize:DescribeSolutionVersion",
                "personalize:GetSolutionMetrics",
                "personalize:ListBatchInferenceJobs",
                "personalize:ListBatchSegmentJobs",
                "personalize:ListCampaigns",
                "personalize:ListDatasetExportJobs",
                "personalize:ListDatasetGroups",
                "personalize:ListDatasetImportJobs",
                "personalize:ListDatasets",
                "personalize:ListEventTrackers",
                "personalize:ListFilters",
                "personalize:ListRecipes",
                "personalize:ListRecommenders",
                "personalize:ListSchemas",
                "personalize:ListSolutions",
                "personalize:ListSolutionVersions"
            ],
            "Resource": "*"
        }
    ]
}
```
Troubleshooting Amazon Personalize identity and access

Use the following information to help you diagnose and fix common issues that you might encounter when working with Amazon Personalize and IAM.

Topics

- I am not authorized to perform an action in Amazon Personalize (p. 403)
- I am not authorized to perform iam:PassRole (p. 403)
- I want to allow people outside of my AWS account to access my Amazon Personalize resources (p. 403)

I am not authorized to perform an action in Amazon Personalize

If you receive an error that you're not authorized to perform an action, your policies must be updated to allow you to perform the action.

The following example error occurs when the mateojackson IAM user tries to use the console to view details about a fictional my-example-widget resource but doesn't have the fictional personalize:GetWidget permissions.

User: arn:aws:iam::123456789012:user/mateojackson is not authorized to perform: personalize:GetWidget on resource: my-example-widget

In this case, the policy for the mateojackson user must be updated to allow access to the my-example-widget resource by using the personalize:GetWidget action.

If you need help, contact your AWS administrator. Your administrator is the person who provided you with your sign-in credentials.

I am not authorized to perform iam:PassRole

If you receive an error that you're not authorized to perform the iam:PassRole action, your policies must be updated to allow you to pass a role to Amazon Personalize.

Some AWS services allow you to pass an existing role to that service instead of creating a new service role or service-linked role. To do this, you must have permissions to pass the role to the service.

The following example error occurs when an IAM user named marymajor tries to use the console to perform an action in Amazon Personalize. However, the action requires the service to have permissions that are granted by a service role. Mary does not have permissions to pass the role to the service.

User: arn:aws:iam::123456789012:user/marymajor is not authorized to perform: iam:PassRole

In this case, Mary's policies must be updated to allow her to perform the iam:PassRole action.

If you need help, contact your AWS administrator. Your administrator is the person who provided you with your sign-in credentials.

I want to allow people outside of my AWS account to access my Amazon Personalize resources

You can create a role that users in other accounts or people outside of your organization can use to access your resources. You can specify who is trusted to assume the role. For services that support
resource-based policies or access control lists (ACLs), you can use those policies to grant people access to your resources.

To learn more, consult the following:

- To learn whether Amazon Personalize supports these features, see How Amazon Personalize works with IAM (p. 393).
- To learn how to provide access to your resources across AWS accounts that you own, see Providing access to an IAM user in another AWS account that you own in the IAM User Guide.
- To learn how to provide access to your resources to third-party AWS accounts, see Providing access to AWS accounts owned by third parties in the IAM User Guide.
- To learn how to provide access through identity federation, see Providing access to externally authenticated users (identity federation) in the IAM User Guide.
- To learn the difference between using roles and resource-based policies for cross-account access, see How IAM roles differ from resource-based policies in the IAM User Guide.

Logging and monitoring in Amazon Personalize

This section provides information about monitoring and logging Amazon Personalize with Amazon CloudWatch and AWS CloudTrail.

Topics

- Monitoring Amazon Personalize (p. 404)
- CloudWatch metrics for Amazon Personalize (p. 407)
- Logging Amazon Personalize API calls with AWS CloudTrail (p. 410)

Monitoring Amazon Personalize

With Amazon CloudWatch, you can get metrics associated with Amazon Personalize. You can set up alarms to notify you when one or more of these metrics fall outside a defined threshold. To see metrics, you can use Amazon CloudWatch, Amazon AWS Command Line Interface, or the CloudWatch API.

Topics

- Using CloudWatch metrics for Amazon Personalize (p. 404)
- Accessing Amazon Personalize metrics (p. 405)
- Creating an alarm (p. 405)
- Amazon Personalize serverless monitoring app example (p. 407)

Using CloudWatch metrics for Amazon Personalize

To use metrics, you must specify the following information:

- The metric name.
- The metric dimension. A dimension is a name-value pair that helps you to uniquely identify a metric.

You can get monitoring data for Amazon Personalize using the AWS Management Console, the AWS CLI, or the CloudWatch API. You can also use the CloudWatch API through one of the AWS SDKs or the CloudWatch API tools. The console displays a series of graphs based on the raw data from the CloudWatch API. Depending on your needs, you might prefer to use either the graphs displayed in the console or retrieved from the API.
The following list shows some common uses for the metrics. These are suggestions to get you started, not a comprehensive list.

<table>
<thead>
<tr>
<th>How do I?</th>
<th>Relevant metric</th>
</tr>
</thead>
<tbody>
<tr>
<td>How do I track the number of events that have been recorded?</td>
<td>Monitor the PutEventsRequests metric.</td>
</tr>
<tr>
<td>How can I monitor the DatasetImportJob errors?</td>
<td>Use the DatasetImportJobError metric.</td>
</tr>
<tr>
<td>How can I monitor the latency of GetRecommendations calls?</td>
<td>Use the GetRecommendationsLatency metric.</td>
</tr>
</tbody>
</table>

You must have the appropriate CloudWatch permissions to monitor Amazon Personalize with CloudWatch. For more information, see [Authentication and access control for Amazon CloudWatch](https://docs.aws.amazon.com/AmazonCloudWatch/latest/monitoring/auth-access-control.html).

### Accessing Amazon Personalize metrics

The following examples show how to access Amazon Personalize metrics using the CloudWatch console, the AWS CLI, and the CloudWatch API.

**To view metrics (console)**

2. Choose **Metrics**, choose the **All metrics** tab, and then choose AWS/Personalize.
3. Choose the metric dimension.
4. Choose the desired metric from the list, and choose a time period for the graph.

**To view metrics for events received over a period of time (CLI)**

- Open the AWS CLI and enter the following command:

```bash
aws cloudwatch get-metric-statistics \
  --metric-name PutEventsRequests \
  --start-time 2019-03-15T00:00:20Z \
  --period 3600 \
  --end-time 2019-03-16T00:00:00Z \
  --namespace AWS/Personalize \
  --dimensions Name=EventTrackerArn,Value=EventTrackerArn \
  --statistics Sum
```

This example shows the events received for the given event tracker ARN over a period of time. For more information, see [get-metric-statistics](https://docs.aws.amazon.com/AmazonCloudWatch/latest/monitoring/changing-system-metrics.html).

**To access metrics (CloudWatch API)**


### Creating an alarm

You can create a CloudWatch alarm that sends an Amazon Simple Notification Service (Amazon SNS) message when the alarm changes state. An alarm watches a single metric over a time period you specify.
The alarm performs one or more actions based on the value of the metric relative to a given threshold over a number of time periods. The action is a notification sent to an Amazon SNS topic or an AWS Auto Scaling policy.

Alarms invoke actions for sustained state changes only. CloudWatch alarms do not invoke actions simply because they are in a particular state. The state must have changed and been maintained for a specified number of time periods.

**To set an alarm (console)**

2. In the navigation pane, Choose **Alarms**, and then choose **Create alarm**. This launches the *Create Alarm Wizard*.
3. Choose **Select metric**.
4. In the **All metrics** tab, choose **AWS/Personalize**.
5. Choose **EventTrackerArn**, and then choose **PutEventsRequests** metrics.
6. Choose the **Graphed metrics** tab.
7. For **Statistic** choose **Sum**.
8. Choose **Select metric**.
9. Fill in the **Name** and **Description**. For **Whenever** choose >, and then enter a maximum value of your choice.
10. If you want CloudWatch to send you email when the alarm state is reached, for **Whenever this alarm**: choose **State is ALARM**. To send alarms to an existing Amazon SNS topic, for **Send notification to**: choose an existing SNS topic. To set the name and email addresses for a new email subscription list, choose **New list**. CloudWatch saves the list and displays it in the field so you can use it to set future alarms.

   **Note**

   If you use **New list** to create a new Amazon SNS topic, the email addresses must be verified before the intended recipients receive notifications. Amazon SNS sends email only when the alarm enters an alarm state. If this alarm state change happens before the email addresses are verified, intended recipients do not receive a notification.

11. Choose **Create alarm**.

**To set an alarm (AWS CLI)**

- Open the AWS CLI, and then enter the following command. Change the value of the **alarm-actions** parameter to reference an Amazon SNS topic that you previously created.

```bash
aws cloudwatch put-metric-alarm \
  --alarm-name PersonalizeCLI \
  --alarm-description "Alarm when more than 10 events occur" \
  --metric-name PutEventsRequests \
  --namespace AWS/Personalize \
  --statistic Sum \
  --period 300 \
  --threshold 10 \
  --comparison-operator GreaterThanThreshold \
  --evaluation-periods 1 \
  --unit Count \
  --dimensions Name=EventTrackerArn,Value=EventTrackerArn \
  --alarm-actions SNSTopicArn
```

This example shows how to create an alarm for when more than 10 events occur for the given event tracker ARN within 5 minutes. For more information, see [put-metric-alarm](https://docs.aws.amazon.com/AmazonCloudWatch/latest/monitoring/alarm-console.html#cloudwatch-console-putmetricalarm-command).
To set an alarm (CloudWatch API)

- Call `PutMetricAlarm`. For more information, see Amazon CloudWatch API Reference.

Amazon Personalize serverless monitoring app example

For an example app that adds monitoring, alerting, and optimization capabilities for Amazon Personalize see Amazon Personalize monitor in the Amazon Personalize samples repository.

CloudWatch metrics for Amazon Personalize

This section contains information about the Amazon CloudWatch metrics available for Amazon Personalize. For more information, see Monitoring Amazon Personalize (p. 404).

The following table lists the Amazon Personalize metrics. All metrics except GetRecommendations and GetPersonalizedRanking support these statistics: Average, Minimum, Maximum, Sum. GetRecommendations and GetPersonalizedRanking support Sum only.

<table>
<thead>
<tr>
<th>Metric</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>DatasetImportJobRequests</td>
<td>The number of successful CreateDatasetImportJob API calls.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: DatasetGroupArn, DatasetArn, DatasetImportJobArn</td>
</tr>
<tr>
<td>DatasetImportJobError</td>
<td>The number of CreateDatasetImportJob API calls that resulted in an error.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: DatasetGroupArn, DatasetArn, DatasetImportJobArn</td>
</tr>
<tr>
<td>DatasetImportJobExecutionTime</td>
<td>The time between the CreateDatasetImportJob API call and the completion (or failure) of the operation.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: DatasetGroupArn, DatasetArn, DatasetImportJobArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Seconds</td>
</tr>
<tr>
<td>DatasetSize</td>
<td>The size of data imported by the dataset import job.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: DatasetGroupArn, DatasetArn, DatasetImportJobArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Bytes</td>
</tr>
<tr>
<td>SolutionTrainingJobRequests</td>
<td>The number of successful CreateSolutionVersion API calls.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: SolutionArn, SolutionVersionArn</td>
</tr>
<tr>
<td>SolutionTrainingJobError</td>
<td>The number of CreateSolutionVersion API calls that resulted in an error.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: SolutionArn, SolutionVersionArn</td>
</tr>
<tr>
<td>Metric</td>
<td>Description</td>
</tr>
<tr>
<td>--------------------------------------------</td>
<td>-------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>SolutionTrainingJobExecutionTime</td>
<td>The time between the CreateSolutionVersion API call and the completion (or failure) of the operation.</td>
</tr>
<tr>
<td></td>
<td>Dimensions: SolutionArn, SolutionVersionArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Seconds</td>
</tr>
<tr>
<td>GetPersonalizedRanking</td>
<td>Whether a GetPersonalizedRanking (p. 604) API call is successful. Use the sum statistic to view total count of successful GetPersonalizedRanking API calls. This metric doesn't support other statistics.</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td>GetPersonalizedRanking4xxErrors</td>
<td>The number of GetPersonalizedRanking API calls that returned a 4xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td>GetPersonalizedRanking5xxErrors</td>
<td>The number of GetPersonalizedRanking API calls that returned a 5xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td>GetPersonalizedRankingLatency</td>
<td>The time between receiving the GetPersonalizedRanking API call and the sending of recommendations (excludes 4xx and 5xx errors).</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Milliseconds</td>
</tr>
<tr>
<td>GetRecommendations</td>
<td>Whether a GetRecommendations (p. 608) API calls is successful. Use the sum statistic to view total count of successful GetRecommendations API calls. This metric doesn't support other statistics.</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td>GetRecommendations4xxErrors</td>
<td>The number of GetRecommendations API calls that returned a 4xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td>GetRecommendations5xxErrors</td>
<td>The number of GetRecommendations API calls that returned a 5xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td>GetRecommendationsLatency</td>
<td>The time between receiving the GetRecommendations API call and the sending of recommendations (excludes 4xx and 5xx errors).</td>
</tr>
<tr>
<td></td>
<td>Dimension: CampaignArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Milliseconds</td>
</tr>
<tr>
<td>Metric</td>
<td>Description</td>
</tr>
<tr>
<td>------------------------</td>
<td>-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>PutEventsRequests</td>
<td>The number of successful PutEvents API calls.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn, EventTrackerArn</td>
</tr>
<tr>
<td>PutEvents4xxErrors</td>
<td>The number of PutEvents API calls that returned a 4xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn, EventTrackerArn</td>
</tr>
<tr>
<td>PutEvents5xxErrors</td>
<td>The number of PutEvents API calls that returned a 5xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn, EventTrackerArn</td>
</tr>
<tr>
<td>PutEventLatency</td>
<td>The time taken for the completion of the PutEvents API call (excludes 4xx and 5xx errors).</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn, EventTrackerArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Milliseconds</td>
</tr>
<tr>
<td>PutItemsRequests</td>
<td>The number of successful PutItems API calls.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td>PutItems4xxErrors</td>
<td>The number of PutItems API calls that returned a 4xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td>PutItems5xxErrors</td>
<td>The number of PutItems API calls that returned a 5xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td>PutItemsLatency</td>
<td>The time taken for the completion of the PutItems API call (excludes 4xx and 5xx errors).</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Milliseconds</td>
</tr>
<tr>
<td>PutUsersRequests</td>
<td>The number of successful PutUsers API calls.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td>PutUsers4xxErrors</td>
<td>The number of PutUsers API calls that returned a 4xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
</tbody>
</table>
Logging Amazon Personalize API calls with AWS CloudTrail

Amazon Personalize is integrated with AWS CloudTrail, a service that provides a record of actions taken by a user, role, or an AWS service in Amazon Personalize. CloudTrail captures a subset of API calls for Amazon Personalize as events, including calls from the Amazon Personalize console and from code calls to the Amazon Personalize APIs. If you create a trail, you can enable continuous delivery of CloudTrail events to an Amazon S3 bucket, including events for Amazon Personalize. If you don't configure a trail, you can still view the most recent events in the CloudTrail console in Event history. Using the information collected by CloudTrail, you can determine the request that was made to Amazon Personalize, the IP address from which the request was made, who made the request, when it was made, and additional details.

To learn more about CloudTrail, including how to configure and enable it, see the AWS CloudTrail User Guide.

Amazon Personalize information in CloudTrail

CloudTrail is enabled on your AWS account when you create the account. When supported event activity occurs in Amazon Personalize, that activity is recorded in a CloudTrail event along with other AWS service events in Event history. You can view, search, and download recent events in your AWS account. For more information, see Viewing events with CloudTrail event history.

For an ongoing record of events in your AWS account, including events for Amazon Personalize, create a trail. A trail enables CloudTrail to deliver log files to an Amazon S3 bucket. By default, when you create a trail in the console, the trail applies to all regions. The trail logs events from all regions in the AWS partition and delivers the log files to the Amazon S3 bucket that you specify. Additionally, you can configure other AWS services to further analyze and act upon the event data collected in CloudTrail logs. For more information, see:

- Overview for creating a trail
- CloudTrail supported services and integrations
- Configuring Amazon SNS notifications for CloudTrail
- Receiving CloudTrail log files from multiple regions and Receiving CloudTrail log files from multiple accounts

Amazon Personalize supports logging every action (API operation) as an event in CloudTrail log files. For more information, see Actions (p. 420).

Every event or log entry contains information about who generated the request. The identity information helps you determine the following:

<table>
<thead>
<tr>
<th>Metric</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>PutUsers5xxErrors</td>
<td>The number of PutUsers API calls that returned a 5xx HTTP response code.</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td>PutUsersLatency</td>
<td>The time taken for the completion of the PutUsers API call (excludes 4xx and 5xx errors).</td>
</tr>
<tr>
<td></td>
<td>Dimension: DatasetGroupArn, DatasetArn</td>
</tr>
<tr>
<td></td>
<td>Unit: Milliseconds</td>
</tr>
</tbody>
</table>
• Whether the request was made with root or user credentials.
• Whether the request was made with temporary security credentials for a role or federated user.
• Whether the request was made by another AWS service.

For more information, see the CloudTrail userIdentity element.

Example: Amazon Personalize log file entries

A trail is a configuration that enables delivery of events as log files to an Amazon S3 bucket that you specify. CloudTrail log files contain one or more log entries. An event represents a single request from any source and includes information about the requested action, the date and time of the action, request parameters, and so on. CloudTrail log files are not an ordered stack trace of the public API calls, so they do not appear in any specific order.

The following example shows a CloudTrail log entry with actions for the ListDatasetGroups API operation. Note that because the ListDatasetGroups API operation is an action that doesn't change state, the responseElements response is null. For more information about the body of CloudTrail records, see CloudTrail record contents.

```json
{
  "eventVersion": "1.05",
  "userIdentity": {
    "type": "IAMUser",
    "principalId": "principal-id",
    "arn": "arn:aws:iam::user-arn",
    "accountId": "account-id",
    "accessKeyId": "access-key",
    "userName": "user-name"
  },
  "eventTime": "2018-11-22T02:18:03Z",
  "eventSource": "personalize.amazonaws.com",
  "eventName": "ListDatasetGroups",
  "awsRegion": "us-west-2",
  "sourceIPAddress": "source-ip-address",
  "userAgent": "aws-cli/1.11.16 Python/2.7.11 Darwin/15.6.0 botocore/1.4.73",
  "requestParameters": null,
  "responseElements": null,
  "requestID": "request-id",
  "eventType": "event-id",
  "eventType": "AwsApiCall",
  "recipientAccountId": "recipient-account-id"
}
```

Compliance validation for Amazon Personalize

Third-party auditors assess the security and compliance of Amazon Personalize as part of multiple AWS compliance programs. These include SOC, PCI, HIPAA, and others.

For a list of AWS services in scope of specific compliance programs, see AWS services in scope by compliance program. For general information, see AWS compliance programs.

You can download third-party audit reports using AWS Artifact. For more information, see Downloading reports in AWS Artifact.

Your compliance responsibility when using Amazon Personalize is determined by the sensitivity of your data, your company's compliance objectives, and applicable laws and regulations. AWS provides the following resources to help with compliance:
Resilience in Amazon Personalize

The AWS global infrastructure is built around AWS Regions and Availability Zones. AWS Regions provide multiple physically separated and isolated Availability Zones, which are connected with low-latency, high-throughput, and highly redundant networking. With Availability Zones, you can design and operate applications and databases that automatically fail over between zones without interruption. Availability Zones are more highly available, fault tolerant, and scalable than traditional single or multiple data center infrastructures.

Amazon Personalize leverages the AWS global infrastructure for data resiliency. When you create an Amazon Personalize resource in an AWS Region, Amazon Personalize manages the resilience and data redundancy of the resource across multiple Availability Zones. For a list of AWS regions where you can create Amazon Personalize resources, see AWS regions and endpoints in the Amazon Web Services General Reference. For more information about AWS Regions and Availability Zones, see AWS global infrastructure.

Infrastructure security in Amazon Personalize

As a managed service, Amazon Personalize is protected by AWS global network security. For information about AWS security services and how AWS protects infrastructure, see AWS Cloud Security. To design your AWS environment using the best practices for infrastructure security, see Infrastructure Protection in Security Pillar AWS Well-Architected Framework.

You use AWS published API calls to access Amazon Personalize through the network. Clients must support the following:

- Transport Layer Security (TLS). We require TLS 1.2 and recommend TLS 1.3.
- Cipher suites with perfect forward secrecy (PFS) such as DHE (Ephemeral Diffie-Hellman) or ECDHE (Elliptic Curve Ephemeral Diffie-Hellman). Most modern systems such as Java 7 and later support these modes.

Additionally, requests must be signed by using an access key ID and a secret access key that is associated with an IAM principal. Or you can use the AWS Security Token Service (AWS STS) to generate temporary security credentials to sign requests.
Amazon Personalize and interface VPC endpoints (AWS PrivateLink)

If you use Amazon Virtual Private Cloud (Amazon VPC) to host your AWS resources, you can establish a private connection between your VPC and Amazon Personalize. This connection allows Amazon Personalize to communicate with your resources on your VPC without going through the public internet.

Amazon VPC is an AWS service that you use to launch AWS resources in a virtual private cloud (VPC) or virtual network that you define. With a VPC, you have control over your network settings, such as the IP address range, subnets, route tables, and network gateways. With VPC endpoints, the AWS network handles the routing between your VPC and AWS services.

To connect your VPC to Amazon Personalize, you define an interface VPC endpoint for Amazon Personalize. An interface endpoint is an elastic network interface with a private IP address that serves as an entry point for traffic destined to a supported AWS service. The endpoint provides reliable, scalable connectivity to Amazon Personalize. It doesn't require an internet gateway, a network address translation (NAT) instance, or a VPN connection. For more information, see What is Amazon VPC in the Amazon VPC User Guide.

Interface VPC endpoints are enabled by AWS PrivateLink. This AWS technology enables private communication between AWS services by using an elastic network interface with private IP addresses.

Note
All Amazon Personalize Federal Information Processing Standard (FIPS) endpoints are supported by AWS PrivateLink.

Considerations for Amazon Personalize VPC endpoints

Before you set up an interface VPC endpoint for Amazon Personalize, ensure that you review Interface endpoint properties and limitations in the Amazon VPC User Guide.

Amazon Personalize supports making calls to all of its API actions from your VPC.

Creating an interface VPC endpoint for Amazon Personalize

You can create a VPC endpoint for the Amazon Personalize service with either the Amazon VPC console or the AWS Command Line Interface (AWS CLI). For more information, see Creating an interface endpoint in the Amazon VPC User Guide.

To create a VPC endpoint for Amazon Personalize, choose one of the following for the service:

- com.amazonaws.region.personalize
- com.amazonaws.region.personalize-events
- com.amazonaws.region.personalize-runtime

If you enable private DNS for the endpoint, you can make API requests to Amazon Personalize using its default DNS name for the Region, for example, personalize.us-east-1.amazonaws.com.

For more information, see Accessing a service through an interface endpoint in the Amazon VPC User Guide.
Creating a VPC endpoint policy for Amazon Personalize

You can attach an endpoint policy to your VPC endpoint that controls access to Amazon Personalize. The policy specifies the following information:

- The principal that can perform actions.
- The actions that can be performed.
- The resources on which actions can be performed.

For more information, see Controlling access to services with VPC endpoints in the Amazon VPC User Guide.

Example: VPC endpoint policy allowing all Amazon Personalize actions and passRole actions

When attached to an endpoint, this policy grants access to all Amazon Personalize actions and passRole actions.

```json
{
    "Statement": [
        {
            "Principal": "*",
            "Effect": "Allow",
            "Action": [
                "personalize:*",
                "iam:PassRole"
            ],
            "Resource": "*
        }
    ]
}
```

Example: VPC endpoint policy allowing Amazon Personalize ListDatasets actions

When attached to an endpoint, this policy grants access to the listed Amazon Personalize ListDatasets actions.

```json
{
    "Statement": [
        {
            "Principal": "*",
            "Effect": "Allow",
            "Action": [
                "personalize:ListDatasets"
            ],
            "Resource": "*
        }
    ]
}
```
Amazon Personalize endpoints and quotas

The following sections contain information about Amazon Personalize guidelines, quotas, and endpoints. For adjustable quotas, you can request a quota increase using the Service Quotas console. For more information see Requesting a quota increase (p. 419).

Topics
- Amazon Personalize endpoints and regions (p. 415)
- Compliance (p. 415)
- Service quotas (p. 415)
- Requesting a quota increase (p. 419)

Amazon Personalize endpoints and regions

For a list of Amazon Personalize endpoints by region, see AWS regions and endpoints in the Amazon Web Services General Reference.

Compliance

For information about Amazon Personalize compliance programs, see AWS compliance, AWS compliance programs, and AWS services in scope by compliance program.

Service quotas

Your AWS account has the following quotas for Amazon Personalize.

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Interactions</td>
<td></td>
</tr>
<tr>
<td>Minimum number of unique combined historical and event interactions (after filtering by eventType and eventValueThreshold, if provided) required to train a model (create a solution version).</td>
<td>1000</td>
</tr>
<tr>
<td>Maximum number of interactions that are considered by a model during training.</td>
<td>500 million (adjustable)</td>
</tr>
<tr>
<td>Maximum number of distinct event types combined with total number of optional metadata columns in Interactions datasets.</td>
<td>10</td>
</tr>
<tr>
<td>Maximum number of metadata columns, excluding reserved fields, in Interactions datasets.</td>
<td>5</td>
</tr>
<tr>
<td>Resource</td>
<td>Quota</td>
</tr>
<tr>
<td>----------------------------------------------</td>
<td>----------------------------------------------------------------------</td>
</tr>
<tr>
<td>Maximum number of characters for categorical data and impression values.</td>
<td>1000</td>
</tr>
<tr>
<td>Maximum amount of bulk interactions data per dataset import job with FULL import mode.</td>
<td>100 GB (increases to 1TB with any increase to Interactions considered by a model)</td>
</tr>
<tr>
<td>Maximum amount of bulk interactions data per dataset import job with INCREMENTAL import mode.</td>
<td>1 GB</td>
</tr>
<tr>
<td>Minimum number of interactions records per dataset import job with FULL or INCREMENTAL import mode.</td>
<td>1000</td>
</tr>
</tbody>
</table>

**Users**

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Minimum number of unique users, with at minimum 2 interactions each, required to create a domain recommender or custom solution version.</td>
<td>25</td>
</tr>
<tr>
<td>Minimum percentage of total users that must have at minimum 2 interactions or more before you can create a domain recommender or custom solution version.</td>
<td>1 percent</td>
</tr>
<tr>
<td>Maximum number of metadata fields for a Users dataset.</td>
<td>25</td>
</tr>
<tr>
<td>Maximum number of characters for USER_ID data values.</td>
<td>256</td>
</tr>
<tr>
<td>Maximum number of characters for categorical data values.</td>
<td>1000 characters</td>
</tr>
<tr>
<td>Maximum amount of bulk user data per dataset import job with FULL import mode.</td>
<td>100 GB</td>
</tr>
<tr>
<td>Maximum amount of bulk user data per dataset import job with INCREMENTAL import mode.</td>
<td>1 GB</td>
</tr>
</tbody>
</table>

**Items**

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Maximum number of items that are considered by a model during training and generating recommendations.</td>
<td>750,000</td>
</tr>
<tr>
<td>Maximum number of metadata fields for an Items dataset.</td>
<td>100</td>
</tr>
<tr>
<td>Maximum number of characters for ITEM_ID data values.</td>
<td>256</td>
</tr>
<tr>
<td>Maximum number of characters for categorical data values.</td>
<td>1000 characters</td>
</tr>
<tr>
<td>Maximum number of textual fields for an Items dataset.</td>
<td>1</td>
</tr>
<tr>
<td>Resource</td>
<td>Quota</td>
</tr>
<tr>
<td>-------------------------------------------------------------------------</td>
<td>---------------------------</td>
</tr>
<tr>
<td>Maximum number of characters for textual data values for Chinese and</td>
<td>7,000 characters</td>
</tr>
<tr>
<td>Japanese languages.</td>
<td></td>
</tr>
<tr>
<td>Maximum number of characters for textual data values for all other</td>
<td>20,000 characters</td>
</tr>
<tr>
<td>languages.</td>
<td></td>
</tr>
<tr>
<td>Maximum amount of bulk items data per dataset import job with BULK</td>
<td>100 GB</td>
</tr>
<tr>
<td>import mode.</td>
<td></td>
</tr>
<tr>
<td>Maximum amount of bulk item data per dataset import job with</td>
<td>1 GB</td>
</tr>
<tr>
<td>INCREMENTAL import mode.</td>
<td></td>
</tr>
</tbody>
</table>

**Individual record import APIs**

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Maximum rate of PutEvents requests per dataset group.</td>
<td>1000/second</td>
</tr>
<tr>
<td>Maximum number of events in a PutEvents call.</td>
<td>10</td>
</tr>
<tr>
<td>Maximum size of an event.</td>
<td>10 KB</td>
</tr>
<tr>
<td>Maximum rate of PutItems requests per dataset group.</td>
<td>10/second</td>
</tr>
<tr>
<td>Maximum number of items in a PutItems call.</td>
<td>10</td>
</tr>
<tr>
<td>Maximum rate of PutUsers requests per dataset group.</td>
<td>10/second</td>
</tr>
<tr>
<td>Maximum number of users in a PutUsers call.</td>
<td>10</td>
</tr>
</tbody>
</table>

**Legacy recipes**

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Maximum amount of combined data for Users and Items datasets for</td>
<td>5 GB</td>
</tr>
<tr>
<td>HRNN-metadata and HRNN-Coldstart recipes.</td>
<td></td>
</tr>
<tr>
<td>Maximum number of cold start items the HRNN-Coldstart recipe supports</td>
<td>80000</td>
</tr>
<tr>
<td>to train a model (create a solution version).</td>
<td></td>
</tr>
<tr>
<td>Minimum number of cold start items the HRNN-Coldstart recipe requires</td>
<td>100</td>
</tr>
<tr>
<td>to train a model (create a solution version).</td>
<td></td>
</tr>
</tbody>
</table>

**Filters**

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total number of filters per dataset group.</td>
<td>10</td>
</tr>
<tr>
<td>Maximum number of distinct dataset fields for a filter.</td>
<td>5</td>
</tr>
<tr>
<td>Total number of distinct dataset fields across all filters in a dataset</td>
<td>10</td>
</tr>
<tr>
<td>group.</td>
<td></td>
</tr>
<tr>
<td>Maximum number of interactions per user per event type considered by a</td>
<td>100 interactions (adjustable)</td>
</tr>
<tr>
<td>filter.</td>
<td></td>
</tr>
</tbody>
</table>

**GetRecommendations / GetPersonalizedRanking requests**
Resource | Quota
--- | ---
Maximum transaction rate (GetRecommendations and GetPersonalizedRanking requests) | 2500/sec
Maximum number of GetRecommendations requests per second per campaign | 500/sec
Maximum number of GetPersonalizedRanking requests per second per campaign | 500/sec.

**Metric attribution quotas**

Maximum number of metrics for a metric attribution | 10
Maximum number of unique event attribution sources | 100

**Batch inference jobs**

Maximum number of input files in a batch inference job | 1000
Maximum size of batch inference job input | 1 GB
Maximum number of records per input file in a batch inference job | 50 million

**Batch segment jobs**

Maximum number of queries per input file for Item-Affinity recipe | 500
Maximum number of queries per input file for Item-Attribute-Affinity recipe | 10
Maximum number of users per segment | 5 million

Your AWS account has the following quotas for each region.

<table>
<thead>
<tr>
<th>Resource</th>
<th>Quota</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total number of active schemas</td>
<td>500</td>
</tr>
<tr>
<td>Total number of active dataset groups</td>
<td>5 (adjustable)</td>
</tr>
<tr>
<td>Total number of pending or in progress dataset import jobs</td>
<td>5</td>
</tr>
<tr>
<td>Total number of pending or in progress batch inference jobs</td>
<td>5 (adjustable)</td>
</tr>
<tr>
<td>Total number of pending or in progress batch segment jobs</td>
<td>5</td>
</tr>
<tr>
<td>Total number of pending or in progress solution versions</td>
<td>20 (adjustable)</td>
</tr>
</tbody>
</table>

Each dataset group has the following quotas.
## Requesting a quota increase

For adjustable quotas, you can request a quota increase using the [Service Quotas console](https://console.aws.amazon.com/servicequotas/home). The following Amazon Personalize quotas are adjustable:

- Maximum number of interactions that are considered by a model during training.
- Active campaigns per dataset group
- Active dataset groups
- Active filters per dataset group
- Active solutions per dataset group
- Amount of data per incremental import
- Maximum number of interactions per user per event type considered by a filter
- Total number of pending or in progress batch inference jobs
- Total number of pending or in progress solution versions
- Maximum rate of `PutEvents` requests

To request a quota increase, use the [Service Quotas console](https://console.aws.amazon.com/servicequotas/home) and follow the steps in the [Requesting a quota increase](https://console.aws.amazon.com/servicequotas/home) section of the [Service Quotas User Guide](https://console.aws.amazon.com/servicequotas/home).
API reference

This section provides documentation for the Amazon Personalize API operations. For a list of Amazon Personalize endpoints by region, see AWS regions and endpoints in the AWS General Reference.

Topics

• Actions (p. 420)
• Data Types (p. 612)
• Common Errors (p. 734)
• Common Parameters (p. 735)

Actions

The following actions are supported by Amazon Personalize:

• CreateBatchInferenceJob (p. 424)
• CreateBatchSegmentJob (p. 428)
• CreateCampaign (p. 432)
• CreateDataset (p. 436)
• CreateDatasetExportJob (p. 440)
• CreateDatasetGroup (p. 444)
• CreateDatasetImportJob (p. 448)
• CreateEventTracker (p. 452)
• CreateFilter (p. 455)
• CreateMetricAttribution (p. 458)
• CreateRecommender (p. 461)
• CreateSchema (p. 465)
• CreateSolution (p. 468)
• CreateSolutionVersion (p. 473)
• DeleteCampaign (p. 477)
• DeleteDataset (p. 479)
• DeleteDatasetGroup (p. 481)
• DeleteEventTracker (p. 483)
• DeleteFilter (p. 485)
• DeleteMetricAttribution (p. 487)
• DeleteRecommender (p. 489)
• DeleteSchema (p. 491)
• DeleteSolution (p. 493)
• DescribeAlgorithm (p. 495)
• DescribeBatchInferenceJob (p. 497)
• DescribeBatchSegmentJob (p. 499)
• DescribeCampaign (p. 501)
• DescribeDataset (p. 503)
• DescribeDatasetExportJob (p. 505)
The following actions are supported by Amazon Personalize Events:

- PutEvents (p. 598)
- PutItems (p. 600)
- PutUsers (p. 602)

The following actions are supported by Amazon Personalize Runtime:
Amazon Personalize

The following actions are supported by Amazon Personalize:

- GetPersonalizedRanking (p. 604)
- GetRecommendations (p. 608)

CreateBatchInferenceJob (p. 424)
CreateBatchSegmentJob (p. 428)
CreateCampaign (p. 432)
CreateDataset (p. 436)
CreateDatasetExportJob (p. 440)
CreateDatasetGroup (p. 444)
CreateDatasetImportJob (p. 448)
CreateEventTracker (p. 452)
CreateFilter (p. 455)
CreateMetricAttribution (p. 458)
CreateRecommender (p. 461)
CreateSchema (p. 465)
CreateSolution (p. 468)
CreateSolutionVersion (p. 473)
DeleteCampaign (p. 477)
DeleteDataset (p. 479)
DeleteDatasetGroup (p. 481)
DeleteEventTracker (p. 483)
DeleteFilter (p. 485)
DeleteMetricAttribution (p. 487)
DeleteRecommender (p. 489)
DeleteSchema (p. 491)
DeleteSolution (p. 493)
DescribeAlgorithm (p. 495)
DescribeBatchInferenceJob (p. 497)
DescribeBatchSegmentJob (p. 499)
DescribeCampaign (p. 501)
DescribeDataset (p. 503)
DescribeDatasetExportJob (p. 505)
DescribeDatasetGroup (p. 507)
DescribeDatasetImportJob (p. 509)
DescribeEventTracker (p. 511)
DescribeFeatureTransformation (p. 513)
DescribeFilter (p. 515)
DescribeMetricAttribution (p. 517)
DescribeRecipe (p. 519)
DescribeRecommender (p. 521)
DescribeSchema (p. 524)
DescribeSolution (p. 526)
• DescribeSolutionVersion (p. 529)
• GetSolutionMetrics (p. 532)
• ListBatchInferenceJobs (p. 534)
• ListBatchSegmentJobs (p. 537)
• ListCampaigns (p. 540)
• ListDatasetExportJobs (p. 543)
• ListDatasetGroups (p. 546)
• ListDatasetImportJobs (p. 548)
• ListDatasets (p. 551)
• ListEventTrackers (p. 554)
• ListFilters (p. 556)
• ListMetricAttributionMetrics (p. 558)
• ListMetricAttributions (p. 560)
• ListRecipes (p. 562)
• ListRecommenders (p. 565)
• ListSchemas (p. 568)
• ListSolutions (p. 570)
• ListSolutionVersions (p. 573)
• ListTagsForResource (p. 576)
• StartRecommender (p. 578)
• StopRecommender (p. 580)
• StopSolutionVersionCreation (p. 582)
• TagResource (p. 584)
• UntagResource (p. 586)
• UpdateCampaign (p. 588)
• UpdateDataset (p. 591)
• UpdateMetricAttribution (p. 593)
• UpdateRecommender (p. 596)
CreateBatchInferenceJob
Service: Amazon Personalize

Creates a batch inference job. The operation can handle up to 50 million records and the input file must be in JSON format. For more information, see Creating a batch inference job.

Request Syntax

```
{
  "batchInferenceJobConfig": {
    "itemExplorationConfig": {
      "string": "string"
    }
  },
  "filterArn": "string",
  "jobInput": {
    "s3DataSource": {
      "kmsKeyArn": "string",
      "path": "string"
    }
  },
  "jobName": "string",
  "jobOutput": {
    "s3DataDestination": {
      "kmsKeyArn": "string",
      "path": "string"
    }
  },
  "numResults": number,
  "roleArn": "string",
  "solutionVersionArn": "string",
  "tags": [
    {
      "tagKey": "string",
      "tagValue": "string"
    }
  ]
}
```

Request Parameters

The request accepts the following data in JSON format.

**batchInferenceJobConfig** *(p. 424)*

The configuration details of a batch inference job.

Type: **BatchInferenceJobConfig** *(p. 624)* object

Required: No

**filterArn** *(p. 424)*

The ARN of the filter to apply to the batch inference job. For more information on using filters, see Filtering batch recommendations.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.*:*:*:*:`
The Amazon S3 path that leads to the input file to base your recommendations on. The input material must be in JSON format.

Type: \textit{BatchInferenceJobInput (p. 625)} object

Required: Yes

The name of the batch inference job to create.

Type: String


Pattern: \^[a-zA-Z0-9-9][a-zA-Z0-9-9\-_]*

The path to the Amazon S3 bucket where the job's output will be stored.

Type: \textit{BatchInferenceJobOutput (p. 626)} object

The number of recommendations to retrieve.

Type: Integer

The ARN of the Amazon Identity and Access Management role that has permissions to read and write to your input and output Amazon S3 buckets respectively.

Type: String

Length Constraints: Maximum length of 256.

Pattern: \texttt{arn\:([a-zA-Z0-9\-]+):iam::\d{12}:role/?([a-zA-Z0-9-9]+,.@\-/]+}

The Amazon Resource Name (ARN) of the solution version that will be used to generate the batch inference recommendations.

Type: String

Length Constraints: Maximum length of 256.

Pattern: \texttt{arn\:([a-zA-Z0-9\-]+):personalize:.+\:*:.+}

A list of \texttt{tags} to apply to the batch inference job.
Type: Array of Tag objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

Response Syntax

```
{
  "batchInferenceJobArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**batchInferenceJobArn (p. 426)**

The ARN of the batch inference job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: \arn:\([a-z\d\-]+:personalize:.+:.*\)+

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.
HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateBatchSegmentJob
Service: Amazon Personalize

Creates a batch segment job. The operation can handle up to 50 million records and the input file must be in JSON format. For more information, see Getting batch recommendations and user segments.

Request Syntax

```json
{
    "filterArn": "string",
    "jobInput": {
        "s3DataSource": {
            "kmsKeyArn": "string",
            "path": "string"
        }
    },
    "jobName": "string",
    "jobOutput": {
        "s3DataDestination": {
            "kmsKeyArn": "string",
            "path": "string"
        }
    },
    "numResults": number,
    "roleArn": "string",
    "solutionVersionArn": "string",
    "tags": [
        {
            "tagKey": "string",
            "tagValue": "string"
        }
    ]
}
```

Request Parameters

The request accepts the following data in JSON format.

**filterArn (p. 428)**

The ARN of the filter to apply to the batch segment job. For more information on using filters, see Filtering batch recommendations.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:.+.

Required: No

**jobInput (p. 428)**

The Amazon S3 path for the input data used to generate the batch segment job.

Type: BatchSegmentJobInput (p. 632) object

Required: Yes

**jobName (p. 428)**

The name of the batch segment job to create.
Type: String
Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_.]*$

Required: Yes

**jobOutput** *(p. 428)*

The Amazon S3 path for the bucket where the job's output will be stored.

Type: **BatchSegmentJobOutput** *(p. 633)* object

Required: Yes

**numResults** *(p. 428)*

The number of predicted users generated by the batch segment job for each line of input data. The maximum number of users per segment is 5 million.

Type: Integer

Required: No

**roleArn** *(p. 428)*

The ARN of the Amazon Identity and Access Management role that has permissions to read and write to your input and output Amazon S3 buckets respectively.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:(？a-z\d-+):iam::d{12}:role/?[a-zA-Z0-9\-=+.@\/-]+$

Required: Yes

**solutionVersionArn** *(p. 428)*

The Amazon Resource Name (ARN) of the solution version you want the batch segment job to use to generate batch segments.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:(？a-z\d-+):personalize:.+:.*:.+$

Required: Yes

**tags** *(p. 428)*

A list of tags to apply to the batch segment job.

Type: Array of **Tag** *(p. 723)* objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

**Response Syntax**

```json
{
```


```json
"batchSegmentJobArn": "string"
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**batchSegmentJobArn (p. 429)**

The ARN of the batch segment job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.+`

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](https://docs.aws.amazon.com/cli/latest/reference/personalize/)

- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateCampaign
Service: Amazon Personalize

Creates a campaign that deploys a solution version. When a client calls the GetRecommendations and GetPersonalizedRanking APIs, a campaign is specified in the request.

Minimum Provisioned TPS and Auto-Scaling

Important
A high \texttt{minProvisionedTPS} will increase your bill. We recommend starting with 1 for \texttt{minProvisionedTPS} (the default). Track your usage using Amazon CloudWatch metrics, and increase the \texttt{minProvisionedTPS} as necessary.

A transaction is a single GetRecommendations or GetPersonalizedRanking call. Transactions per second (TPS) is the throughput and unit of billing for Amazon Personalize. The minimum provisioned TPS (\texttt{minProvisionedTPS}) specifies the baseline throughput provisioned by Amazon Personalize, and thus, the minimum billing charge.

If your TPS increases beyond \texttt{minProvisionedTPS}, Amazon Personalize auto-scales the provisioned capacity up and down, but never below \texttt{minProvisionedTPS}. There's a short time delay while the capacity is increased that might cause loss of transactions.

The actual TPS used is calculated as the average requests/second within a 5-minute window. You pay for maximum of either the minimum provisioned TPS or the actual TPS. We recommend starting with a low \texttt{minProvisionedTPS}, track your usage using Amazon CloudWatch metrics, and then increase the \texttt{minProvisionedTPS} as necessary.

Status
A campaign can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

To get the campaign status, call \texttt{DescribeCampaign}.

Note
Wait until the status of the campaign is ACTIVE before asking the campaign for recommendations.

Related APIs

- \texttt{ListCampaigns}
- \texttt{DescribeCampaign}
- \texttt{UpdateCampaign}
- \texttt{DeleteCampaign}

Request Syntax

```json
{
  "campaignConfig": {
    "itemExplorationConfig": {
      "string": "string"
    }
  },
  "minProvisionedTPS": \texttt{number},
  "name": "\texttt{string}"
}
```
"solutionVersionArn": "string",
"tags": [
  {
    "tagKey": "string",
    "tagValue": "string"
  }
]

Request Parameters

The request accepts the following data in JSON format.

campaignConfig (p. 432)

The configuration details of a campaign.

Type: CampaignConfig (p. 638) object

Required: No

minProvisionedTPS (p. 432)

Specifies the requested minimum provisioned transactions (recommendations) per second that Amazon Personalize will support. A high minProvisionedTPS will increase your bill. We recommend starting with 1 for minProvisionedTPS (the default). Track your usage using Amazon CloudWatch metrics, and increase the minProvisionedTPS as necessary.

Type: Integer

Valid Range: Minimum value of 1.

Required: No

name (p. 432)

A name for the new campaign. The campaign name must be unique within your account.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9-\_]*

Required: Yes

solutionVersionArn (p. 432)

The Amazon Resource Name (ARN) of the solution version to deploy.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:.*

Required: Yes

tags (p. 432)

A list of tags to apply to the campaign.

Type: Array of Tag (p. 723) objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.
Required: No

Response Syntax

```
{
  "campaignArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response. The following data is returned in JSON format by the service.

**campaignArn (p. 434)**

The Amazon Resource Name (ARN) of the campaign.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:+

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateDataset
Service: Amazon Personalize

Creates an empty dataset and adds it to the specified dataset group. Use CreateDatasetImportJob to import your training data to a dataset.

There are three types of datasets:

- Interactions
- Items
- Users

Each dataset type has an associated schema with required field types. Only the Interactions dataset is required in order to train a model (also referred to as creating a solution).

A dataset can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

To get the status of the dataset, call DescribeDataset.

Related APIs

- CreateDatasetGroup
- ListDatasets
- DescribeDataset
- DeleteDataset

Request Syntax

```
{
  "datasetGroupArn": "string",
  "datasetType": "string",
  "name": "string",
  "schemaArn": "string",
  "tags": [{
    "tagKey": "string",
    "tagValue": "string"
  }]
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetGroupArn** *(p. 436)*

The Amazon Resource Name (ARN) of the dataset group to add the dataset to.

Type: String

Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+
Required: Yes

```datasetType (p. 436)```

The type of dataset.

One of the following (case insensitive) values:
- Interactions
- Items
- Users

Type: String
Length Constraints: Maximum length of 256.
Required: Yes

```name (p. 436)```

The name for the dataset.

Type: String
Pattern: ^[a-zA-Z0-9\-9][a-zA-Z0-9\-9\-\_]*
Required: Yes

```schemaArn (p. 436)```

The ARN of the schema to associate with the dataset. The schema defines the dataset fields.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+
Required: Yes

```tags (p. 436)```

A list of tags to apply to the dataset.

Type: Array of Tag (p. 723) objects
Array Members: Minimum number of 0 items. Maximum number of 200 items.
Required: No

Response Syntax

```
{
   "datasetArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**datasetArn (p. 437)**

The ARN of the dataset.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:\.*:`

**Errors**

*InvalidInputException*

Provide a valid value for the field or parameter.

HTTP Status Code: 400

*LimitExceededException*

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

*ResourceAlreadyExistsException*

The specified resource already exists.

HTTP Status Code: 400

*ResourceInUseException*

The specified resource is in use.

HTTP Status Code: 400

*ResourceNotFoundException*

Could not find the specified resource.

HTTP Status Code: 400

*TooManyTagsException*

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateDatasetExportJob

Service: Amazon Personalize

Creates a job that exports data from your dataset to an Amazon S3 bucket. To allow Amazon Personalize to export the training data, you must specify an service-linked IAM role that gives Amazon Personalize PutObject permissions for your Amazon S3 bucket. For information, see Exporting a dataset in the Amazon Personalize developer guide.

Status

A dataset export job can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

To get the status of the export job, call DescribeDatasetExportJob, and specify the Amazon Resource Name (ARN) of the dataset export job. The dataset export is complete when the status shows as ACTIVE. If the status shows as CREATE FAILED, the response includes a failureReason key, which describes why the job failed.

Request Syntax

```
{
  "datasetArn": "string",
  "ingestionMode": "string",
  "jobName": "string",
  "jobOutput": {
    "s3DataDestination": {
      "kmsKeyArn": "string",
      "path": "string"
    }
  },
  "roleArn": "string",
  "tags": [
    {
      "tagKey": "string",
      "tagValue": "string"
    }
  ]
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetArn (p. 440)**

The Amazon Resource Name (ARN) of the dataset that contains the data to export.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.+*

Required: Yes

**ingestionMode (p. 440)**

The data to export, based on how you imported the data. You can choose to export only BULK data that you imported using a dataset import job, only PUT data that you imported incrementally (using
the console, PutEvents, PutUsers and PutItems operations), or ALL for both types. The default value is PUT.

Type: String

Valid Values: BULK | PUT | ALL

Required: No

**jobName (p. 440)**

The name for the dataset export job.

Type: String


Pattern: `^[a-zA-Z0-9][a-zA-Z0-9\-_]*$`

Required: Yes

**jobOutput (p. 440)**

The path to the Amazon S3 bucket where the job's output is stored.

Type: `DatasetExportJobOutput (p. 650)` object

Required: Yes

**roleArn (p. 440)**

The Amazon Resource Name (ARN) of the IAM service role that has permissions to add data to your output Amazon S3 bucket.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):iam::\d{12}:role/[^a-zA-Z_0-9+=,.@_\-\//]`+

Required: Yes

**tags (p. 440)**

A list of tags to apply to the dataset export job.

Type: Array of `Tag (p. 723)` objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

**Response Syntax**

```json
{
   "datasetExportJobArn": "string"
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.
datasetExportJobArn (p. 441)

The Amazon Resource Name (ARN) of the dataset export job.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*:*:*:*

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

LimitExceededException

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

ResourceAlreadyExistsException

The specified resource already exists.

HTTP Status Code: 400

ResourceInUseException

The specified resource is in use.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

TooManyTagsException

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateDatasetGroup
Service: Amazon Personalize

Creates an empty dataset group. A dataset group is a container for Amazon Personalize resources. A dataset group can contain at most three datasets, one for each type of dataset:

- Interactions
- Items
- Users

A dataset group can be a Domain dataset group, where you specify a domain and use pre-configured resources like recommenders, or a Custom dataset group, where you use custom resources, such as a solution with a solution version, that you deploy with a campaign. If you start with a Domain dataset group, you can still add custom resources such as solutions and solution versions trained with recipes for custom use cases and deployed with campaigns.

A dataset group can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING

To get the status of the dataset group, call DescribeDatasetGroup. If the status shows as CREATE FAILED, the response includes a failureReason key, which describes why the creation failed.

**Note**
You must wait until the status of the dataset group is ACTIVE before adding a dataset to the group.

You can specify an AWS Key Management Service (KMS) key to encrypt the datasets in the group. If you specify a KMS key, you must also include an AWS Identity and Access Management (IAM) role that has permission to access the key.

**APIs that require a dataset group ARN in the request**

- CreateDataset
- CreateEventTracker
- CreateSolution

**Related APIs**

- ListDatasetGroups
- DescribeDatasetGroup
- DeleteDatasetGroup

**Request Syntax**

```json
{
    "domain": "string",
    "kmsKeyArn": "string",
    "name": "string",
    "roleArn": "string",
    "tags": [
        {
            "tagKey": "string",
            "tagValue": "string"
        }
    ]
}
```
```json
"tagValue": "string"
]
}

Request Parameters

The request accepts the following data in JSON format.

**domain (p. 444)**

The domain of the dataset group. Specify a domain to create a Domain dataset group. The domain you specify determines the default schemas for datasets and the use cases available for recommenders. If you don't specify a domain, you create a Custom dataset group with solution versions that you deploy with a campaign.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

Required: No

**kmsKeyArn (p. 444)**

The Amazon Resource Name (ARN) of a AWS Key Management Service (KMS) key used to encrypt the datasets.

Type: String

Length Constraints: Maximum length of 2048.

Pattern: `arn:aws.*:kms:.*:\[0-9\]\{12\}:key/.*`

Required: No

**name (p. 444)**

The name for the new dataset group.

Type: String


Pattern: `^[a-zA-Z0-9][a-zA-Z0-9-_.]*$`

Required: Yes

**roleArn (p. 444)**

The ARN of the AWS Identity and Access Management (IAM) role that has permissions to access the AWS Key Management Service (KMS) key. Supplying an IAM role is only valid when also specifying a KMS key.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:\([a-z\d-]+\)::iam::\d{12}::role/\?\[a-zA-Z_0-9+=,.@\-_\/]+`

Required: No

**tags (p. 444)**

A list of tags to apply to the dataset group.
Type: Array of `Tag (p. 723)` objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

**Response Syntax**

```
{
    "datasetGroupArn": "string",
    "domain": "string"
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**datasetGroupArn (p. 446)**

The Amazon Resource Name (ARN) of the new dataset group.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.*::*:*+.`

**domain (p. 446)**

The domain for the new Domain dataset group.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateDatasetImportJob
Service: Amazon Personalize

Creates a job that imports training data from your data source (an Amazon S3 bucket) to an Amazon Personalize dataset. To allow Amazon Personalize to import the training data, you must specify an IAM service role that has permission to read from the data source, as Amazon Personalize makes a copy of your data and processes it internally. For information on granting access to your Amazon S3 bucket, see Giving Amazon Personalize Access to Amazon S3 Resources.

If you already created a recommender or deployed a custom solution version with a campaign, how new bulk records influence recommendations depends on the domain use case or recipe that you use. For more information, see How new data influences real-time recommendations.

**Important**
By default, a dataset import job replaces any existing data in the dataset that you imported in bulk. To add new records without replacing existing data, specify INCREMENTAL for the import mode in the CreateDatasetImportJob operation.

**Status**
A dataset import job can be in one of the following states:

- CREATE PENDING > CREATE IN PROGRESS > ACTIVE -or- CREATE FAILED

To get the status of the import job, call DescribeDatasetImportJob, providing the Amazon Resource Name (ARN) of the dataset import job. The dataset import is complete when the status shows as ACTIVE. If the status shows as CREATE FAILED, the response includes a failureReason key, which describes why the job failed.

**Note**
Importing takes time. You must wait until the status shows as ACTIVE before training a model using the dataset.

**Related APIs**
- ListDatasetImportJobs
- DescribeDatasetImportJob

**Request Syntax**

```json
{
  "datasetArn": "string",
  "dataSource": {
    "dataLocation": "string"
  },
  "importMode": "string",
  "jobName": "string",
  "publishAttributionMetricsToS3": boolean,
  "roleArn": "string",
  "tags": [
    {
      "tagKey": "string",
      "tagValue": "string"
    }
  ]
}
```
**Request Parameters**

The request accepts the following data in JSON format.

**datasetArn (p. 448)**

The ARN of the dataset that receives the imported data.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.*:`

Required: Yes

**dataSource (p. 448)**

The Amazon S3 bucket that contains the training data to import.

Type: `DataSource (p. 670)` object

Required: Yes

**importMode (p. 448)**

Specify how to add the new records to an existing dataset. The default import mode is FULL. If you haven't imported bulk records into the dataset previously, you can only specify FULL.

- Specify FULL to overwrite all existing bulk data in your dataset. Data you imported individually is not replaced.
- Specify INCREMENTAL to append the new records to the existing data in your dataset. Amazon Personalize replaces any record with the same ID with the new one.

Type: String

Valid Values: FULL | INCREMENTAL

Required: No

**jobName (p. 448)**

The name for the dataset import job.

Type: String


Pattern: `^[a-zA-Z0-9][a-zA-Z0-9\-_.]*`

Required: Yes

**publishAttributionMetricsToS3 (p. 448)**

If you created a metric attribution, specify whether to publish metrics for this import job to Amazon S3

Type: Boolean

Required: No

**roleArn (p. 448)**

The ARN of the IAM role that has permissions to read from the Amazon S3 data source.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):iam::\d{12}:role/?[a-zA-Z0-9+=,.@\-_\+/]+
Required: Yes

tag(s) (p. 448)
A list of tags to apply to the dataset import job.
Type: Array of Tag (p. 723) objects
Array Members: Minimum number of 0 items. Maximum number of 200 items.
Required: No

Response Syntax

```json
{
   "datasetImportJobArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

datasetImportJobArn (p. 450)
The ARN of the dataset import job.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+

Errors

InvalidInputException
Provide a valid value for the field or parameter.
HTTP Status Code: 400

LimitExceededException
The limit on the number of requests per second has been exceeded.
HTTP Status Code: 400

ResourceAlreadyExistsException
The specified resource already exists.
HTTP Status Code: 400

ResourceInUseException
The specified resource is in use.
HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

TooManyTagsException

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateEventTracker
Service: Amazon Personalize

Creates an event tracker that you use when adding event data to a specified dataset group using the PutEvents API.

**Note**
Only one event tracker can be associated with a dataset group. You will get an error if you call CreateEventTracker using the same dataset group as an existing event tracker.

When you create an event tracker, the response includes a tracking ID, which you pass as a parameter when you use the PutEvents operation. Amazon Personalize then appends the event data to the Interactions dataset of the dataset group you specify in your event tracker.

The event tracker can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

To get the status of the event tracker, call DescribeEventTracker.

**Note**
The event tracker must be in the ACTIVE state before using the tracking ID.

**Related APIs**
- ListEventTrackers
- DescribeEventTracker
- DeleteEventTracker

**Request Syntax**

```json
{
  "datasetGroupArn": "string",
  "name": "string",
  "tags": [
    {
      "tagKey": "string",
      "tagValue": "string"
    }
  ]
}
```

**Request Parameters**

The request accepts the following data in JSON format.

**datasetGroupArn (p. 452)**

The Amazon Resource Name (ARN) of the dataset group that receives the event data.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize::*:*:`

Required: Yes
**name (p. 452)**

The name for the event tracker.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_]*

Required: Yes

**tags (p. 452)**

A list of tags to apply to the event tracker.

Type: Array of Tag (p. 723) objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

### Response Syntax

```
{
  "eventTrackerArn": "string",
  "trackingId": "string"
}
```

### Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**eventTrackerArn (p. 453)**

The ARN of the event tracker.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:.*

**trackingId (p. 453)**

The ID of the event tracker. Include this ID in requests to the PutEvents API.

Type: String

Length Constraints: Maximum length of 256.

### Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400
**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateFilter
Service: Amazon Personalize

Creates a recommendation filter. For more information, see Filtering recommendations and user segments.

Request Syntax

```
{
  "datasetGroupArn": "string",
  "filterExpression": "string",
  "name": "string",
  "tags": [
    {
      "tagKey": "string",
      "tagValue": "string"
    }
  ]
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 455)

The ARN of the dataset group that the filter will belong to.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*+:

Required: Yes

filterExpression (p. 455)

The filter expression defines which items are included or excluded from recommendations. Filter expression must follow specific format rules. For information about filter expression structure and syntax, see Filter expressions.

Type: String


Required: Yes

name (p. 455)

The name of the filter to create.

Type: String


Pattern: ^[a-zA-Z0-9-9][a-zA-Z0-9\-_.]*$

Required: Yes

tags (p. 455)

A list of tags to apply to the filter.
Type: Array of Tag (p. 723) objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

Response Syntax

```
{
  "filterArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**filterArn (p. 456)**

The ARN of the new filter.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:+

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
CreateMetricAttribution
Service: Amazon Personalize

Creates a metric attribution. A metric attribution creates reports on the data that you import into Amazon Personalize. Depending on how you imported the data, you can view reports in Amazon CloudWatch or Amazon S3. For more information, see Measuring impact of recommendations.

Request Syntax

```
{
  "datasetGroupArn": "string",
  "metrics": [
    {
      "eventType": "string",
      "expression": "string",
      "metricName": "string"
    }
  ],
  "metricsOutputConfig": {
    "roleArn": "string",
    "s3DataDestination": {
      "kmsKeyArn": "string",
      "path": "string"
    }
  },
  "name": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 458)

The Amazon Resource Name (ARN) of the destination dataset group for the metric attribution.

Type: String

Length Constraints: Maximum length of 256.

Pattern: \arn:\([a-z\d-]+\):personalize:.*:.*:.*

Required: Yes

metrics (p. 458)

A list of metric attributes for the metric attribution. Each metric attribute specifies an event type to track and a function. Available functions are SUM() or SAMPLECOUNT(). For SUM() functions, provide the dataset type (either Interactions or Items) and column to sum as a parameter. For example SUM(Items.PRICE).

Type: Array of MetricAttribute (p. 690) objects

Array Members: Maximum number of 10 items.

Required: Yes

metricsOutputConfig (p. 458)

The output configuration details for the metric attribution.

Type: MetricAttributionOutput (p. 693) object
Required: Yes

name (p. 458)

A name for the metric attribution.
Type: String
Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-\_]*$
Required: Yes

Response Syntax

```
{
  "metricAttributionArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

metricAttributionArn (p. 459)

The Amazon Resource Name (ARN) for the new metric attribution.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-zA-Z\d-]+):personalize:.+:*:.+.

Errors

InvalidInputException

Provide a valid value for the field or parameter.
HTTP Status Code: 400

LimitExceededException

The limit on the number of requests per second has been exceeded.
HTTP Status Code: 400

ResourceAlreadyExistsException

The specified resource already exists.
HTTP Status Code: 400

ResourceInUseException

The specified resource is in use.
HTTP Status Code: 400
ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
CreateRecommender
Service: Amazon Personalize

Creates a recommender with the recipe (a Domain dataset group use case) you specify. You create recommenders for a Domain dataset group and specify the recommender's Amazon Resource Name (ARN) when you make a GetRecommendations request.

Minimum recommendation requests per second

Important
A high minRecommendationRequestsPerSecond will increase your bill. We recommend starting with 1 for minRecommendationRequestsPerSecond (the default). Track your usage using Amazon CloudWatch metrics, and increase the minRecommendationRequestsPerSecond as necessary.

When you create a recommender, you can configure the recommender's minimum recommendation requests per second. The minimum recommendation requests per second (minRecommendationRequestsPerSecond) specifies the baseline recommendation request throughput provisioned by Amazon Personalize. The default minRecommendationRequestsPerSecond is 1. A recommendation request is a single GetRecommendations operation. Request throughput is measured in requests per second and Amazon Personalize uses your requests per second to derive your requests per hour and the price of your recommender usage.

If your requests per second increases beyond minRecommendationRequestsPerSecond, Amazon Personalize auto-scales the provisioned capacity up and down, but never below minRecommendationRequestsPerSecond. There's a short time delay while the capacity is increased that might cause loss of requests.

Your bill is the greater of either the minimum requests per hour (based on minRecommendationRequestsPerSecond) or the actual number of requests. The actual request throughput used is calculated as the average requests/second within a one-hour window. We recommend starting with the default minRecommendationRequestsPerSecond, track your usage using Amazon CloudWatch metrics, and then increase the minRecommendationRequestsPerSecond as necessary.

Status

A recommender can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- STOP PENDING > STOP IN_PROGRESS > INACTIVE > START PENDING > START IN_PROGRESS > ACTIVE
- DELETE PENDING > DELETE IN_PROGRESS

To get the recommender status, call DescribeRecommender.

Note
Wait until the status of the recommender is ACTIVE before asking the recommender for recommendations.

Related APIs

- ListRecommenders
- DescribeRecommender
- UpdateRecommender
- DeleteRecommender
## Request Syntax

```json
{
    "datasetGroupArn": "string",
    "name": "string",
    "recipeArn": "string",
    "recommenderConfig": {
        "itemExplorationConfig": {
            "string": "string"
        },
        "minRecommendationRequestsPerSecond": number,
        "trainingDataConfig": {
            "excludedDatasetColumns": {
                "string": [ "string"
            ]
        }
    }
},
"tags": [
    {
        "tagKey": "string",
        "tagValue": "string"
    }
]
}```

## Request Parameters

The request accepts the following data in JSON format.

**datasetGroupArn (p. 462)**

The Amazon Resource Name (ARN) of the destination domain dataset group for the recommender.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:*::*:+

Required: Yes

**name (p. 462)**

The name of the recommender.

Type: String


Pattern: ^[a-zA-Z0-9\-9][a-zA-Z0-9\-9\-_.]*

Required: Yes

**recipeArn (p. 462)**

The Amazon Resource Name (ARN) of the recipe that the recommender will use. For a recommender, a recipe is a Domain dataset group use case. Only Domain dataset group use cases can be used to create a recommender. For information about use cases see [Choosing recommender use cases](#).

Type: String

Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.+:.

Required: Yes

**recommenderConfig (p. 462)**

The configuration details of the recommender.

Type: **RecommenderConfig (p. 704)** object

Required: No

**tags (p. 462)**

A list of tags to apply to the recommender.

Type: Array of **Tag (p. 723)** objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: No

**Response Syntax**

```
{
  "recommenderArn": "string"
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**recommenderArn (p. 463)**

The Amazon Resource Name (ARN) of the recommender.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.+:.

**Errors**

* **InvalidInputException**
  
  Provide a valid value for the field or parameter.

  HTTP Status Code: 400

* **LimitExceededException**
  
  The limit on the number of requests per second has been exceeded.

  HTTP Status Code: 400

* **ResourceAlreadyExistsException**
  
  The specified resource already exists.
HTTP Status Code: 400
**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400
**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400
**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
**CreateSchema**
Service: Amazon Personalize

Creates an Amazon Personalize schema from the specified schema string. The schema you create must be in Avro JSON format.

Amazon Personalize recognizes three schema variants. Each schema is associated with a dataset type and has a set of required field and keywords. If you are creating a schema for a dataset in a Domain dataset group, you provide the domain of the Domain dataset group. You specify a schema when you call `CreateDataset`.

For more information on schemas, see [Datasets and schemas](#).

**Related APIs**
- ListSchemas
- DescribeSchema
- DeleteSchema

**Request Syntax**

```json
{
  "domain": "string",
  "name": "string",
  "schema": "string"
}
```

**Request Parameters**

The request accepts the following data in JSON format.

**domain (p. 465)**

The domain for the schema. If you are creating a schema for a dataset in a Domain dataset group, specify the domain you chose when you created the Domain dataset group.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

Required: No

**name (p. 465)**

The name for the schema.

Type: String


Pattern: `^[a-zA-Z0-9][a-zA-Z0-9-_.]*`

Required: Yes

**schema (p. 465)**

A schema in Avro JSON format.

Type: String
Length Constraints: Maximum length of 10000.
Required: Yes

Response Syntax

```json
{
  "schemaArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**schemaArn (p. 466)**

The Amazon Resource Name (ARN) of the created schema.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+`:

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
CreateSolution
Service: Amazon Personalize

Creates the configuration for training a model. A trained model is known as a solution version. After the configuration is created, you train the model (create a solution version) by calling the CreateSolutionVersion operation. Every time you call CreateSolutionVersion, a new version of the solution is created.

After creating a solution version, you check its accuracy by calling GetSolutionMetrics. When you are satisfied with the version, you deploy it using CreateCampaign. The campaign provides recommendations to a client through the GetRecommendations API.

To train a model, Amazon Personalize requires training data and a recipe. The training data comes from the dataset group that you provide in the request. A recipe specifies the training algorithm and a feature transformation. You can specify one of the predefined recipes provided by Amazon Personalize.

Note
Amazon Personalize doesn't support configuring the hpoObjective for solution hyperparameter optimization at this time.

Status
A solution can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

To get the status of the solution, call DescribeSolution. Wait until the status shows as ACTIVE before calling CreateSolutionVersion.

Related APIs
- ListSolutions
- CreateSolutionVersion
- DescribeSolution
- DeleteSolution
- ListSolutionVersions
- DescribeSolutionVersion

Request Syntax

```json
{
    "datasetGroupArn": "string",
    "eventType": "string",
    "name": "string",
    "performAutoML": boolean,
    "performHPO": boolean,
    "recipeArn": "string",
    "solutionConfig": {
        "algorithmHyperParameters": {
            "string": "string"
        },
        "autoMLConfig": {
            "metricName": "string",
            "recipeArn": "string"
        }
    }
}
```
"eventValueThreshold": "string",
"featureTransformationParameters": {
  "string" : "string"
},
"hpoConfig": {
  "algorithmHyperParameterRanges": {
    "categoricalHyperParameterRanges": [
      {
        "name": "string",
        "values": [ "string" ]
      }
    ],
    "continuousHyperParameterRanges": [
      {
        "maxValue": number,
        "minValue": number,
        "name": "string"
      }
    ],
    "integerHyperParameterRanges": [
      {
        "maxValue": number,
        "minValue": number,
        "name": "string"
      }
    ]
  },
  "hpoObjective": {
    "metricName": "string",
    "metricRegex": "string",
    "type": "string"
  },
  "hpoResourceConfig": {
    "maxNumberOfTrainingJobs": "string",
    "maxParallelTrainingJobs": "string"
  }
},
"optimizationObjective": {
  "itemAttribute": "string",
  "objectiveSensitivity": "string"
},
"trainingDataConfig": {
  "excludedDatasetColumns": {
    "string" : [ "string" ]
  }
},
"tags": [
  {
    "tagKey": "string",
    "tagValue": "string"
  }
]

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 468)

The Amazon Resource Name (ARN) of the dataset group that provides the training data.

Type: String
**eventType** *(p. 468)*

When you have multiple event types (using an EVENT_TYPE schema field), this parameter specifies which event type (for example, ‘click’ or ‘like’) is used for training the model.

If you do not provide an eventType, Amazon Personalize will use all interactions for training with equal weight regardless of type.

Type: String

Length Constraints: Maximum length of 256.

Required: Yes

**name** *(p. 468)*

The name for the solution.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_.]*$ 

Required: Yes

**performAutoML** *(p. 468)*

*Important*

We don't recommend enabling automated machine learning. Instead, match your use case to the available Amazon Personalize recipes. For more information, see Determining your use case.

Whether to perform automated machine learning (AutoML). The default is false. For this case, you must specify recipeArn.

When set to true, Amazon Personalize analyzes your training data and selects the optimal USER_PERSONALIZATION recipe and hyperparameters. In this case, you must omit recipeArn. Amazon Personalize determines the optimal recipe by running tests with different values for the hyperparameters. AutoML lengthens the training process as compared to selecting a specific recipe.

Type: Boolean

Required: No

**performHPO** *(p. 468)*

Whether to perform hyperparameter optimization (HPO) on the specified or selected recipe. The default is false.

When performing AutoML, this parameter is always true and you should not set it to false.

Type: Boolean

Required: No

**recipeArn** *(p. 468)*

The ARN of the recipe to use for model training. This is required when performAutoML is false.
Type: String  
Length Constraints: Maximum length of 256.  
Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+  
Required: No

**solutionConfig (p. 468)**

The configuration to use with the solution. When performAutoML is set to true, Amazon Personalize only evaluates the autoMLConfig section of the solution configuration.

**Note**  
Amazon Personalize doesn't support configuring the hpobojective at this time.

Type: [SolutionConfig](p. 713) object  
Required: No

**tags (p. 468)**

A list of tags to apply to the solution.

Type: Array of [Tag](p. 723) objects  
Array Members: Minimum number of 0 items. Maximum number of 200 items.  
Required: No

**Response Syntax**

```
{
  "solutionArn": "string"
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.  
The following data is returned in JSON format by the service.

**solutionArn (p. 471)**

The ARN of the solution.

Type: String  
Length Constraints: Maximum length of 256.  
Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.  
HTTP Status Code: 400
LimitExceededException

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

ResourceAlreadyExistsException

The specified resource already exists.

HTTP Status Code: 400

ResourceInUseException

The specified resource is in use.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

TooManyTagsException

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
CreateSolutionVersion
Service: Amazon Personalize

Trains or retrains an active solution in a Custom dataset group. A solution is created using the CreateSolution operation and must be in the ACTIVE state before calling CreateSolutionVersion. A new version of the solution is created every time you call this operation.

Status

A solution version can be in one of the following states:

- CREATE PENDING
- CREATE IN_PROGRESS
- ACTIVE
- CREATE FAILED
- CREATE STOPPING
- CREATE STOPPED

To get the status of the version, call DescribeSolutionVersion. Wait until the status shows as ACTIVE before calling CreateCampaign.

If the status shows as CREATE FAILED, the response includes a failureReason key, which describes why the job failed.

Related APIs

- ListSolutionVersions
- DescribeSolutionVersion
- ListSolutions
- CreateSolution
- DescribeSolution
- DeleteSolution

Request Syntax

```
{
    "name": "string",
    "solutionArn": "string",
    "tags": [
        {
            "tagKey": "string",
            "tagValue": "string"
        }
    ],
    "trainingMode": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**name (p. 473)**

The name of the solution version.
solutionArn (p. 473)

The Amazon Resource Name (ARN) of the solution containing the training configuration information.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+
Required: Yes

tags (p. 473)

A list of tags to apply to the solution version.

Type: Array of Tag (p. 723) objects
Array Members: Minimum number of 0 items. Maximum number of 200 items.

required: No

trainingMode (p. 473)

The scope of training to be performed when creating the solution version. The default is FULL. This creates a completely new model based on the entirety of the training data from the datasets in your dataset group.

If you use User-Personalization, you can specify a training mode of UPDATE. This updates the model to consider new items for recommendations. It is not a full retraining. You should still complete a full retraining weekly. If you specify UPDATE, Amazon Personalize will stop automatic updates for the solution version. To resume updates, create a new solution with training mode set to FULL and deploy it in a campaign. For more information about automatic updates, see Automatic updates.

The UPDATE option can only be used when you already have an active solution version created from the input solution using the FULL option and the input solution was trained with the User-Personalization recipe or the legacy HRNN-Coldstart recipe.

Type: String
Valid Values: FULL | UPDATE

required: No

Response Syntax

```
{
   "solutionVersionArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**solutionVersionArn (p. 474)**

The ARN of the new solution version.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.+:+`

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExceededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**TooManyTagsException**

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteCampaign
Service: Amazon Personalize

Removes a campaign by deleting the solution deployment. The solution that the campaign is based on is not deleted and can be redeployed when needed. A deleted campaign can no longer be specified in a GetRecommendations request. For information on creating campaigns, see CreateCampaign.

Request Syntax

```json
{
  "campaignArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**campaignArn (p. 477)**

The Amazon Resource Name (ARN) of the campaign to delete.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteDataset
Service: Amazon Personalize

Deletes a dataset. You can't delete a dataset if an associated DatasetImportJob or SolutionVersion is in the CREATE PENDING or IN PROGRESS state. For more information on datasets, see CreateDataset.

Request Syntax

```json
{
  "datasetArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetArn (p. 479)**

The Amazon Resource Name (ARN) of the dataset to delete.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:.+:.

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
DeleteDatasetGroup
Service: Amazon Personalize

Deletes a dataset group. Before you delete a dataset group, you must delete the following:

- All associated event trackers.
- All associated solutions.
- All datasets in the dataset group.

**Request Syntax**

```
{
    "datasetGroupArn": "string"
}
```

**Request Parameters**

The request accepts the following data in JSON format.

**datasetGroupArn (p. 481)**

- The ARN of the dataset group to delete.
- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: `arn:(\[a-z\d-]+):personalize:.*:.*:.*`:
- Required: Yes

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

**Errors**

- **InvalidInputException**
  - Provide a valid value for the field or parameter.
  - HTTP Status Code: 400
- **ResourceInUseException**
  - The specified resource is in use.
  - HTTP Status Code: 400
- **ResourceNotFoundException**
  - Could not find the specified resource.
  - HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:
• AWS Command Line Interface
• AWS SDK for .NET
• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
DeleteEventTracker
Service: Amazon Personalize

Deletes the event tracker. Does not delete the event-interactions dataset from the associated dataset group. For more information on event trackers, see CreateEventTracker.

Request Syntax

```
{
  "eventTrackerArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**eventTrackerArn (p. 483)**

The Amazon Resource Name (ARN) of the event tracker to delete.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: `arn:([a-z\d-]+):personalize:.+:.*:*`.
- Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteFilter
Service: Amazon Personalize
Deletes a filter.

Request Syntax
```
{
  "filterArn": "string"
}
```

Request Parameters
The request accepts the following data in JSON format.

filterArn (p. 485)
The ARN of the filter to delete.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:*:.+:
Required: Yes

Response Elements
If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

InvalidInputException
Provide a valid value for the field or parameter.
HTTP Status Code: 400

ResourceInUseException
The specified resource is in use.
HTTP Status Code: 400

ResourceNotFoundException
Could not find the specified resource.
HTTP Status Code: 400

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteMetricAttribution
Service: Amazon Personalize
Deletes a metric attribution.

Request Syntax

{
   "metricAttributionArn": "string"
}

Request Parameters

The request accepts the following data in JSON format.

**metricAttributionArn (p. 487)**

The metric attribution's Amazon Resource Name (ARN).
Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:*:.+` Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.
HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.
HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.
HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteRecommender
Service: Amazon Personalize

Deactivates and removes a recommender. A deleted recommender can no longer be specified in a GetRecommendations request.

Request Syntax

```json
{
   "recommenderArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

`recommenderArn (p. 489)`

The Amazon Resource Name (ARN) of the recommender to delete.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:\([a-z\d-]+\):personalize:.*:.*:.+

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceInUseException

The specified resource is in use.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteSchema
Service: Amazon Personalize

Deletes a schema. Before deleting a schema, you must delete all datasets referencing the schema. For more information on schemas, see CreateSchema.

Request Syntax

```
{
  "schemaArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**schemaArn (p. 491)**

  The Amazon Resource Name (ARN) of the schema to delete.
  
  Type: String
  
  Length Constraints: Maximum length of 256.
  
  Pattern: `arn:(\[a-z\d-]+):personalize:.*:.*:.+`
  
  Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

  Provide a valid value for the field or parameter.
  
  HTTP Status Code: 400

**ResourceInUseException**

  The specified resource is in use.
  
  HTTP Status Code: 400

**ResourceNotFoundException**

  Could not find the specified resource.
  
  HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DeleteSolution
Service: Amazon Personalize

Deletes all versions of a solution and the Solution object itself. Before deleting a solution, you must delete all campaigns based on the solution. To determine what campaigns are using the solution, call ListCampaigns and supply the Amazon Resource Name (ARN) of the solution. You can't delete a solution if an associated SolutionVersion is in the CREATE PENDING or IN PROGRESS state. For more information on solutions, see CreateSolution.

Request Syntax

```json
{
   "solutionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**solutionArn (p. 493)**

The ARN of the solution to delete.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:+

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:
- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeAlgorithm
Service: Amazon Personalize

Describes the given algorithm.

Request Syntax

```
{
"algorithmArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**algorithmArn (p. 495)**

The Amazon Resource Name (ARN) of the algorithm to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:*

Required: Yes

Response Syntax

```
{
"algorithm": {
  "algorithmArn": "string",
  "algorithmImage": {
    "dockerURI": "string",
    "name": "String"
  },
  "creationDateTime": number,
  "defaultHyperParameterRanges": {
    "categoricalHyperParameterRanges": [
      {
        "isTunable": boolean,
        "name": "String",
        "values": ["string"
      }
    ],
    "continuousHyperParameterRanges": [
      {
        "isTunable": boolean,
        "maxValue": number,
        "minValue": number,
        "name": "string"
      }
    ],
    "integerHyperParameterRanges": [
      {
        "isTunable": boolean,
        "maxValue": number,
        "minValue": number,
        "name": "string"
      }
    ]
  }
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**algorithm (p. 495)**

A listing of the properties of the algorithm.

Type: [Algorithm](p. 616) object

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
DescribeBatchInferenceJob
Service: Amazon Personalize

Gets the properties of a batch inference job including name, Amazon Resource Name (ARN), status, input and output configurations, and the ARN of the solution version used to generate the recommendations.

Request Syntax

```
{
  "batchInferenceJobArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**batchInferenceJobArn (p. 497)**

The ARN of the batch inference job to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:*.+:

Required: Yes

Response Syntax

```
{
  "batchInferenceJob": {
    "batchInferenceJobArn": "string",
    "batchInferenceJobConfig": {
      "itemExplorationConfig": {
        "string": "string"
      }
    },
    "creationDateTime": number,
    "failureReason": "string",
    "filterArn": "string",
    "jobInput": {
      "s3DataSource": {
        "kmsKeyArn": "string",
        "path": "string"
      }
    },
    "jobName": "string",
    "jobOutput": {
      "s3DataDestination": {
        "kmsKeyArn": "string",
        "path": "string"
      }
    },
    "lastUpdatedDateTime": number,
    "numResults": number,
    "roleArn": "string",
    "solutionVersionArn": "string",
    "status": "string"
  }
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**batchInferenceJob (p. 497)**

Information on the specified batch inference job.

Type: BatchInferenceJob (p. 621) object

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeBatchSegmentJob
Service: Amazon Personalize

Gets the properties of a batch segment job including name, Amazon Resource Name (ARN), status, input and output configurations, and the ARN of the solution version used to generate segments.

Request Syntax

```json
{
   "batchSegmentJobArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**batchSegmentJobArn (p. 499)**

The ARN of the batch segment job to describe.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: arn:([a-z\d-]+):personalize:.+:*+:.+
- Required: Yes

Response Syntax

```json
{
   "batchSegmentJob": {
      "batchSegmentJobArn": "string",
      "creationDateTime": number,
      "failureReason": "string",
      "filterArn": "string",
      "jobInput": {
         "s3DataSource": {
            "kmsKeyArn": "string",
            "path": "string"
         }
      },
      "jobName": "string",
      "jobOutput": {
         "s3DataDestination": {
            "kmsKeyArn": "string",
            "path": "string"
         }
      },
      "lastUpdatedDateTime": number,
      "numResults": number,
      "roleArn": "string",
      "solutionVersionArn": "string",
      "status": "string"
   }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**batchSegmentJob (p. 499)**

Information on the specified batch segment job.

Type: [BatchSegmentJob (p. 629)](#) object

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
DescribeCampaign
Service: Amazon Personalize

Describes the given campaign, including its status.

A campaign can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

When the status is CREATE FAILED, the response includes the failureReason key, which describes why.

For more information on campaigns, see CreateCampaign.

Request Syntax

```json
{
    "campaignArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**campaignArn (p. 501)**

The Amazon Resource Name (ARN) of the campaign.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+

Required: Yes

Response Syntax

```json
{
    "campaign": {
        "campaignArn": "string",
        "campaignConfig": {
            "itemExplorationConfig": { "string": "string" }
        },
        "creationDateTime": number,
        "failureReason": "string",
        "lastUpdatedDateTime": number,
        "latestCampaignUpdate": {
            "campaignConfig": {
                "itemExplorationConfig": { "string": "string" }
            },
            "creationDateTime": number,
            "failureReason": "string",
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

campaign (p. 501)

The properties of the campaign.

Type: Campaign (p. 636) object

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeDataset
Service: Amazon Personalize

Describes the given dataset. For more information on datasets, see CreateDataset.

Request Syntax

```
{
    "datasetArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetArn (p. 503)**

The Amazon Resource Name (ARN) of the dataset to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.*:.*:.*

Required: Yes

Response Syntax

```
{
    "dataset": {
        "creationDateTime": number,
        "datasetArn": "string",
        "datasetGroupArn": "string",
        "datasetType": "string",
        "lastUpdatedDateTime": number,
        "latestDatasetUpdate": {
            "creationDateTime": number,
            "failureReason": "string",
            "lastUpdatedDateTime": number,
            "schemaArn": "string",
            "status": "string"
        },
        "name": "string",
        "schemaArn": "string",
        "status": "string"
    }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**dataset (p. 503)**

A listing of the dataset's properties.
Type: Dataset (p. 645) object

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeDatasetExportJob
Service: Amazon Personalize

Describes the dataset export job created by CreateDatasetExportJob, including the export job status.

Request Syntax

```json
{
   "datasetExportJobArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetExportJobArn (p. 505)**

The Amazon Resource Name (ARN) of the dataset export job to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.+:.*

Required: Yes

Response Syntax

```json
{
   "datasetExportJob": {
      "creationDateTime": number,
      "datasetArn": "string",
      "datasetExportJobArn": "string",
      "failureReason": "string",
      "ingestionMode": "string",
      "jobName": "string",
      "jobOutput": {
         "s3DataDestination": {
            "kmsKeyArn": "string",
            "path": "string"
         }
      },
      "lastUpdatedDateTime": number,
      "roleArn": "string",
      "status": "string"
   }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**datasetExportJob (p. 505)**

Information about the dataset export job, including the status.
The status is one of the following values:

- CREATE PENDING
- CREATE IN_PROGRESS
- ACTIVE
- CREATE FAILED

Type: DatasetExportJob (p. 647) object

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeDatasetGroup
Service: Amazon Personalize

Describes the given dataset group. For more information on dataset groups, see CreateDatasetGroup.

Request Syntax

```
{
  "datasetGroupArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 507)

The Amazon Resource Name (ARN) of the dataset group to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:+

Required: Yes

Response Syntax

```
{
  "datasetGroup": {
    "creationDateTime": number,
    "datasetGroupArn": "string",
    "domain": "string",
    "failureReason": "string",
    "kmsKeyArn": "string",
    "lastUpdatedDateTime": number,
    "name": "string",
    "roleArn": "string",
    "status": "string"
  }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

datasetGroup (p. 507)

A listing of the dataset group's properties.

Type: DatasetGroup (p. 653) object
Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeDatasetImportJob
Service: Amazon Personalize

Describes the dataset import job created by `CreateDatasetImportJob`, including the import job status.

**Request Syntax**

```json
{
    "datasetImportJobArn": "string"
}
```

**Request Parameters**

The request accepts the following data in JSON format.

- **datasetImportJobArn**  
  The Amazon Resource Name (ARN) of the dataset import job to describe.
  
  Type: String
  
  Length Constraints: Maximum length of 256.
  
  Pattern: `arn:([a-z\d-]+):personalize:.+:.*:+`  
  
  Required: Yes

**Response Syntax**

```json
{
    "datasetImportJob": {
        "creationDateTime": number,
        "datasetArn": "string",
        "datasetImportJobArn": "string",
        "dataSource": {
            "dataLocation": "string"
        },
        "failureReason": "string",
        "importMode": "string",
        "jobName": "string",
        "lastUpdatedDateTime": number,
        "publishAttributionMetricsToS3": boolean,
        "roleArn": "string",
        "status": "string"
    }
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

- **datasetImportJob**  
  Information about the dataset import job, including the status.
  
  The status is one of the following values:
CREATE PENDING
CREATE IN_PROGRESS
ACTIVE
CREATE FAILED

Type: DatasetImportJob (p. 657) object

Errors

InvalidInputException
Provide a valid value for the field or parameter.
HTTP Status Code: 400

ResourceNotFoundException
Could not find the specified resource.
HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeEventTracker
Service: Amazon Personalize

Describes an event tracker. The response includes the trackingId and status of the event tracker. For more information on event trackers, see CreateEventTracker.

Request Syntax

```json
{
  "eventTrackerArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**eventTrackerArn (p. 511)**

The Amazon Resource Name (ARN) of the event tracker to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+

Required: Yes

Response Syntax

```json
{
  "eventTracker": {
    "accountId": "string",
    "creationDateTime": "number",
    "datasetGroupArn": "string",
    "eventTrackerArn": "string",
    "lastUpdatedDateTime": "number",
    "name": "string",
    "status": "string",
    "trackingId": "string"
  }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**eventTracker (p. 511)**

An object that describes the event tracker.

Type: EventTracker (p. 675) object
Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeFeatureTransformation
Service: Amazon Personalize
Describes the given feature transformation.

Request Syntax

```
{
    "featureTransformationArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**featureTransformationArn (p. 513)**

The Amazon Resource Name (ARN) of the feature transformation to describe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+.*+.+

Required: Yes

Response Syntax

```
{
    "featureTransformation": {
        "creationDateTime": number,
        "defaultParameters": {
            "string": "string"
        },
        "featureTransformationArn": "string",
        "lastUpdatedDateTime": number,
        "name": "string",
        "status": "string"
    }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**featureTransformation (p. 513)**

A listing of the FeatureTransformation properties.

Type: FeatureTransformation (p. 679) object

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.
HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeFilter
Service: Amazon Personalize

Describes a filter's properties.

Request Syntax

```json
{
  "filterArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

filterArn (p. 515)

The ARN of the filter to describe.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:
Required: Yes

Response Syntax

```json
{
  "filter": {
    "creationDateTime": number,
    "datasetGroupArn": "string",
    "failureReason": "string",
    "filterArn": "string",
    "filterExpression": "string",
    "lastUpdatedDateTime": number,
    "name": "string",
    "status": "string"
  }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

filter (p. 515)

The filter's details.
Type: Filter (p. 681) object

Errors

InvalidInputException

Provide a valid value for the field or parameter.
HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeMetricAttribution
Service: Amazon Personalize

Describes a metric attribution.

Request Syntax

```
{
  "metricAttributionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**metricAttributionArn (p. 517)**

The metric attribution's Amazon Resource Name (ARN).

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.*:.*:.*`  
- Required: Yes

Response Syntax

```
{
  "metricAttribution": {
    "creationDateTime": number,
    "datasetGroupArn": "string",
    "failureReason": "string",
    "lastUpdatedDateTime": number,
    "metricAttributionArn": "string",
    "metricsOutputConfig": {
      "roleArn": "string",
      "s3DataDestination": {
        "kmsKeyArn": "string",
        "path": "string"
      }
    },
    "name": "string",
    "status": "string"
  }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**metricAttribution (p. 517)**

The details of the metric attribution.

- Type: MetricAttribution (p. 691) object
Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeRecipe  
Service: Amazon Personalize  

Describes a recipe.  

A recipe contains three items:  

- An algorithm that trains a model.  
- Hyperparameters that govern the training.  
- Feature transformation information for modifying the input data before training.  

Amazon Personalize provides a set of predefined recipes. You specify a recipe when you create a solution with the CreateSolution API. CreateSolution trains a model by using the algorithm in the specified recipe and a training dataset. The solution, when deployed as a campaign, can provide recommendations using the GetRecommendations API.  

Request Syntax  

```
{
    "recipeArn": "string"
}
```

Request Parameters  

The request accepts the following data in JSON format.  

**recipeArn (p. 519)**  

The Amazon Resource Name (ARN) of the recipe to describe.  

Type: String  

Length Constraints: Maximum length of 256.  

Pattern: arn:([a-z\d-]+):personalize:.+:.*:.*:  

Required: Yes  

Response Syntax  

```
{
    "recipe": {
        "algorithmArn": "string",
        "creationDateTime": number,
        "description": "string",
        "featureTransformationArn": "string",
        "lastUpdatedDateTime": number,
        "name": "string",
        "recipeArn": "string",
        "recipeType": "string",
        "status": "string"
    }
}
```

Response Elements  

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**recipe (p. 519)**

An object that describes the recipe.

Type: *Recipe (p. 697)* object

**Errors**

*InvalidInputException*

Provide a valid value for the field or parameter.

HTTP Status Code: 400

*ResourceNotFoundException*

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
DescribeRecommender
Service: Amazon Personalize

Describes the given recommender, including its status.

A recommender can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- STOP PENDING > STOP IN_PROGRESS > INACTIVE > START PENDING > START IN_PROGRESS > ACTIVE
- DELETE PENDING > DELETE IN_PROGRESS

When the status is CREATE FAILED, the response includes the failureReason key, which describes why.

The modelMetrics key is null when the recommender is being created or deleted.

For more information on recommenders, see CreateRecommender.

Request Syntax

```
{
  "recommenderArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**recommenderArn (p. 521)**

  The Amazon Resource Name (ARN) of the recommender to describe.
  
  Type: String
  
  Length Constraints: Maximum length of 256.
  
  Pattern: arn:([a-z\d-]+):personalize:.+.*.*.
  
  Required: Yes

Response Syntax

```
[
  "recommender": {
    "creationDateTime": number,
    "datasetGroupArn": "string",
    "failureReason": "string",
    "lastUpdatedDateTime": number,
    "latestRecommenderUpdate": {
      "creationDateTime": number,
      "failureReason": "string",
      "lastUpdatedDateTime": number,
      "recommenderConfig": {
        "itemExplorationConfig": {
          "string": "string"
        },
        "minRecommendationRequestsPerSecond": number,
      }
    }
  }
]`

521
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**recommender (p. 521)**

The properties of the recommender.

Type: Recommender (p. 701) object

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
DescribeSchema
Service: Amazon Personalize

Describes a schema. For more information on schemas, see CreateSchema.

Request Syntax

```
{
    "schemaArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**schemaArn (p. 524)**

The Amazon Resource Name (ARN) of the schema to retrieve.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+

Required: Yes

Response Syntax

```
{
    "schema": {
        "creationDateTime": number,
        "domain": "string",
        "lastUpdatedDateTime": number,
        "name": "string",
        "schema": "string",
        "schemaArn": "string"
    }
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**schema (p. 524)**

The requested schema.

Type: DatasetSchema (p. 662) object

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.
HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeSolution
Service: Amazon Personalize

Describes a solution. For more information on solutions, see CreateSolution.

Request Syntax

```json
{
    "solutionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

solutionArn (p. 526)

The Amazon Resource Name (ARN) of the solution to describe.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.+

Required: Yes

Response Syntax

```json
{
    "solution": {
        "autoMLResult": {
            "bestRecipeArn": "string"
        },
        "creationDateTime": "number",
        "datasetGroupArn": "string",
        "eventType": "string",
        "lastUpdatedDateTime": "number",
        "latestSolutionVersion": {
            "creationDateTime": "number",
            "failureReason": "string",
            "lastUpdatedDateTime": "number",
            "solutionVersionArn": "string",
            "status": "string"
        },
        "name": "string",
        "performAutoML": boolean,
        "performHPO": boolean,
        "recipeArn": "string",
        "solutionArn": "string",
        "solutionConfig": {
            "algorithmHyperParameters": {
                "string": "string"
            },
            "autoMLConfig": {
                "metricName": "string",
                "recipeList": [ "string" ]
            },
            "eventValueThreshold": "string",
            "string": "string"
        }
    }
}
```
"featureTransformationParameters": {
  "string": "string"
},
"hpoConfig": {
  "algorithmHyperParameterRanges": {
    "categoricalHyperParameterRanges": [
      {
        "name": "string",
        "Values": [ "string" ]
      }
    ],
    "continuousHyperParameterRanges": [
      {
        "maxValue": number,
        "minValue": number,
        "name": "string"
      }
    ],
    "integerHyperParameterRanges": [
      {
        "maxValue": number,
        "minValue": number,
        "name": "string"
      }
    ]
  },
  "hpoObjective": {
    "metricName": "string",
    "metricRegex": "string",
    "type": "string"
  },
  "hpoResourceConfig": {
    "maxNumberOfTrainingJobs": "string",
    "maxParallelTrainingJobs": "string"
  }
},
"optimizationObjective": {
  "itemAttribute": "String",
  "objectiveSensitivity": "string"
},
"trainingDataConfig": {
  "excludedDatasetColumns": {
    "string": [ "string" ]
  }
}]
"status": "string"
}
Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
DescribeSolutionVersion
Service: Amazon Personalize

Describes a specific version of a solution. For more information on solutions, see CreateSolution

Request Syntax

```json
{
    "solutionVersionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**solutionVersionArn (p. 529)**

The Amazon Resource Name (ARN) of the solution version.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: \[a-z\d-]+:personalize:.+\]
- Required: Yes

Response Syntax

```json
{
    "solutionVersion": {
        "creationDateTime": number,
        "datasetGroupArn": "string",
        "eventType": "string",
        "failureReason": "string",
        "lastUpdatedDateTime": number,
        "name": "string",
        "performAutoML": boolean,
        "performHPO": boolean,
        "recipeArn": "string",
        "solutionArn": "string",
        "solutionConfig": {
            "algorithmHyperParameters": {},
            "autoMLConfig": {
                "metricName": "string",
                "recipeList": [ "string" ]
            },
            "eventValueThreshold": "string",
            "featureTransformationParameters": {
                "string": "string"
            },
            "hpoConfig": {
                "algorithmHyperParameterRanges": {},
                "categoricalHyperParameterRanges": [
                    {
                        "name": "string",
                        "values": [ "string" ]
                    }
                ],
                "continuousHyperParameterRanges": [  
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**solutionVersion (p. 529)**

The solution version.

Type: SolutionVersion (p. 717) object

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.
HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
GetSolutionMetrics
Service: Amazon Personalize

Gets the metrics for the specified solution version.

Request Syntax

```json
{
    "solutionVersionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**solutionVersionArn** *(p. 532)*

The Amazon Resource Name (ARN) of the solution version for which to get metrics.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([-a-z\d-]+):personalize:.*:.*:.+

Required: Yes

Response Syntax

```json
{
    "metrics": {
        "string" : number
    },
    "solutionVersionArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**metrics** *(p. 532)*

The metrics for the solution version. For more information, see [Evaluating a solution version with metrics](#).

Type: String to double map

Map Entries: Maximum number of 100 items.

Key Length Constraints: Maximum length of 256.

**solutionVersionArn** *(p. 532)*

The same solution version ARN as specified in the request.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:(\[a-z\d-\]+):personalize:.*:.*:.*

Errors

InvalidInputException
Provide a valid value for the field or parameter.
HTTP Status Code: 400

ResourceInUseException
The specified resource is in use.
HTTP Status Code: 400

ResourceNotFoundException
Could not find the specified resource.
HTTP Status Code: 400

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
ListBatchInferenceJobs
Service: Amazon Personalize

Gets a list of the batch inference jobs that have been performed off of a solution version.

Request Syntax

```json
{
  "maxResults": number,
  "nextToken": "string",
  "solutionVersionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**maxResults (p. 534)**

The maximum number of batch inference job results to return in each page. The default value is 100.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 534)**

The token to request the next page of results.

Type: String

Length Constraints: Maximum length of 1500.

Required: No

**solutionVersionArn (p. 534)**

The Amazon Resource Name (ARN) of the solution version from which the batch inference jobs were created.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.+:.+

Required: No

Response Syntax

```json
{
  "batchInferenceJobs": [
    {
      "batchInferenceJobArn": "string",
      "creationDateTime": number,
      "failureReason": "string",
      "jobName": "string",
      "lastUpdatedDateTime": number,
    }
  ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**batchInferenceJobs (p. 534)**

A list containing information on each job that is returned.

Type: Array of [BatchInferenceJobSummary (p. 627)] objects

Array Members: Maximum number of 100 items.

**nextToken (p. 534)**

The token to use to retrieve the next page of results. The value is `null` when there are no more results to return.

Type: String

Length Constraints: Maximum length of 1500.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
**ListBatchSegmentJobs**  
Service: Amazon Personalize

Gets a list of the batch segment jobs that have been performed off of a solution version that you specify.

**Request Syntax**

```json
{
    "maxResults": "number",
    "nextToken": "string",
    "solutionVersionArn": "string"
}
```

**Request Parameters**

The request accepts the following data in JSON format.

**maxResults (p. 537)**

- The maximum number of batch segment job results to return in each page. The default value is 100.
- Type: Integer
- Valid Range: Minimum value of 1. Maximum value of 100.
- Required: No

**nextToken (p. 537)**

- The token to request the next page of results.
- Type: String
- Length Constraints: Maximum length of 1500.
- Required: No

**solutionVersionArn (p. 537)**

- The Amazon Resource Name (ARN) of the solution version that the batch segment jobs used to generate batch segments.
- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: `arn:([a-z\d-]+):personalize:.+:.*+.+`
- Required: No

**Response Syntax**

```json
{
    "batchSegmentJobs": [
        {
            "batchSegmentJobArn": "string",
            "creationDateTime": "number",
            "failureReason": "string",
            "jobName": "string",
            "lastUpdatedDateTime": "number",
            ...
        },
        ...
    ]
}
```
"solutionVersionArn": "string",
"status": "string"
}
},
"nextToken": "string"
}

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**batchSegmentJobs (p. 537)**

A list containing information on each job that is returned.

Type: Array of [BatchSegmentJobSummary (p. 634)] objects

Array Members: Maximum number of 100 items.

**nextToken (p. 537)**

The token to use to retrieve the next page of results. The value is null when there are no more results to return.

Type: String

Length Constraints: Maximum length of 1500.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Cpp
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
ListCampaigns
Service: Amazon Personalize

Returns a list of campaigns that use the given solution. When a solution is not specified, all the campaigns associated with the account are listed. The response provides the properties for each campaign, including the Amazon Resource Name (ARN). For more information on campaigns, see CreateCampaign.

Request Syntax

```
{
  "maxResults": number,
  "nextToken": "string",
  "solutionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**maxResults (p. 540)**

The maximum number of campaigns to return.

- Type: Integer
- Valid Range: Minimum value of 1. Maximum value of 100.
- Required: No

**nextToken (p. 540)**

A token returned from the previous call to ListCampaigns for getting the next set of campaigns (if they exist).

- Type: String
- Length Constraints: Maximum length of 1500.
- Required: No

**solutionArn (p. 540)**

The Amazon Resource Name (ARN) of the solution to list the campaigns for. When a solution is not specified, all the campaigns associated with the account are listed.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: arn:([a-z\d-]+):personalize:.+:.+:.
- Required: No

Response Syntax

```
{
  "campaigns": [
    {
      "campaignArn": "string",
    }
  ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**campaigns (p. 540)**

A list of the campaigns.

Type: Array of [CampaignSummary (p. 639)] objects

Array Members: Maximum number of 100 items.

**nextToken (p. 540)**

A token for getting the next set of campaigns (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
• AWS SDK for Ruby V3
ListDatasetExportJobs
Service: Amazon Personalize

Returns a list of dataset export jobs that use the given dataset. When a dataset is not specified, all the dataset export jobs associated with the account are listed. The response provides the properties for each dataset export job, including the Amazon Resource Name (ARN). For more information on dataset export jobs, see `CreateDatasetExportJob`. For more information on datasets, see `CreateDataset`.

Request Syntax

```
{
   "datasetArn": "string",
   "maxResults": number,
   "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetArn (p. 543)

The Amazon Resource Name (ARN) of the dataset to list the dataset export jobs for.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:*:*:*:

Required: No

maxResults (p. 543)

The maximum number of dataset export jobs to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

nextToken (p. 543)

A token returned from the previous call to ListDatasetExportJobs for getting the next set of dataset export jobs (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
[
   "datasetExportJobs": [
      {
         "creationDateTime": number,
         "datasetExportJobArn": "string",
      }
   ]
]
```
"failureReason": "string",
"JobName": "string",
"lastUpdatedDateTime": number,
"status": "string"
}
]
"nextToken": "string"
}

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**datasetExportJobs (p. 543)**

The list of dataset export jobs.

Type: Array of DatasetExportJobSummary (p. 651) objects

Array Members: Maximum number of 100 items.

**nextToken (p. 543)**

A token for getting the next set of dataset export jobs (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
ListDatasetGroups
Service: Amazon Personalize

Returns a list of dataset groups. The response provides the properties for each dataset group, including the Amazon Resource Name (ARN). For more information on dataset groups, see CreateDatasetGroup.

Request Syntax

```
{
  "maxResults": number,
  "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**maxResults (p. 546)**

The maximum number of dataset groups to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 546)**

A token returned from the previous call to ListDatasetGroups for getting the next set of dataset groups (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
{
  "datasetGroups": [
    {
      "creationDateTime": number,
      "datasetGroupArn": "string",
      "domain": "string",
      "failureReason": "string",
      "lastUpdatedDateTime": number,
      "name": "string",
      "status": "string"
    }
  ],
  "nextToken": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**datasetGroups (p. 546)**

The list of your dataset groups.

Type: Array of [DatasetGroupSummary (p. 655)] objects

Array Members: Maximum number of 100 items.

**nextToken (p. 546)**

A token for getting the next set of dataset groups (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

**Errors**

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
ListDatasetImportJobs
Service: Amazon Personalize

Returns a list of dataset import jobs that use the given dataset. When a dataset is not specified, all the
dataset import jobs associated with the account are listed. The response provides the properties for
each dataset import job, including the Amazon Resource Name (ARN). For more information on dataset
import jobs, see CreateDatasetImportJob. For more information on datasets, see CreateDataset.

Request Syntax

```
{
    "datasetArn": "string",
    "maxResults": number,
    "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetArn (p. 548)
The Amazon Resource Name (ARN) of the dataset to list the dataset import jobs for.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:.*:.

Required: No

maxResults (p. 548)
The maximum number of dataset import jobs to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

nextToken (p. 548)
A token returned from the previous call to ListDatasetImportJobs for getting the next set of
dataset import jobs (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
{
    "datasetImportJobs": [
        {
            "creationDateTime": number,
            "datasetImportJobArn": "string",
        }
    ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

datasetImportJobs (p. 548)

The list of dataset import jobs.

Type: Array of DatasetImportJobSummary (p. 660) objects

Array Members: Maximum number of 100 items.

nextToken (p. 548)

A token for getting the next set of dataset import jobs (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
• AWS SDK for Ruby V3
ListDatasets
Service: Amazon Personalize

Returns the list of datasets contained in the given dataset group. The response provides the properties for each dataset, including the Amazon Resource Name (ARN). For more information on datasets, see CreateDataset.

Request Syntax

```json
{
   "datasetGroupArn": "string",
   "maxResults": number,
   "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetGroupArn (p. 551)**

The Amazon Resource Name (ARN) of the dataset group that contains the datasets to list.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.*:.*:.*`

Required: No

**maxResults (p. 551)**

The maximum number of datasets to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 551)**

A token returned from the previous call to ListDatasets for getting the next set of dataset import jobs (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```json
{
   "datasets": [
      {
         "creationDateTime": number,
         "datasetArn": "string",
         "datasetType": "string"
      }
   ]
}```
"lastUpdatedDateTime": number,
"name": "string",
"status": "string"
}
],
"nextToken": "string"
}

Response Elements

If the action is successful, the service sends back an HTTP 200 response. The following data is returned in JSON format by the service.

**datasets (p. 551)**

An array of Dataset objects. Each object provides metadata information.

Type: Array of [DatasetSummary (p. 666)] objects

Array Members: Maximum number of 100 items.

**nextToken (p. 551)**

A token for getting the next set of datasets (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
ListEventTrackers
Service: Amazon Personalize

Returns the list of event trackers associated with the account. The response provides the properties for each event tracker, including the Amazon Resource Name (ARN) and tracking ID. For more information on event trackers, see CreateEventTracker.

Request Syntax

```json
{
  "datasetGroupArn": "string",
  "maxResults": number,
  "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**datasetGroupArn (p. 554)**

The ARN of a dataset group used to filter the response.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:*:*:*+

Required: No

**maxResults (p. 554)**

The maximum number of event trackers to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 554)**

A token returned from the previous call to ListEventTrackers for getting the next set of event trackers (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```json
{
  "eventTrackers": [
    {
      "creationDateTime": number,
      "eventTrackerArn": "string",
      "lastUpdatedDateTime": number,
    }
  ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**eventTrackers (p. 554)**

A list of event trackers.

Type: Array of [EventTrackerSummary](p. 677) objects

Array Members: Maximum number of 100 items.

**nextToken (p. 554)**

A token for getting the next set of event trackers (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
ListFilters
Service: Amazon Personalize

Lists all filters that belong to a given dataset group.

Request Syntax

```
{
  "datasetGroupArn": "string",
  "maxResults": number,
  "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 556)

The ARN of the dataset group that contains the filters.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:*:*+

Required: No

maxResults (p. 556)

The maximum number of filters to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

nextToken (p. 556)

A token returned from the previous call to ListFilters for getting the next set of filters (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
{
  "Filters": [
    {
      "creationDateTime": number,
      "datasetGroupArn": "string",
      "failureReason": "string",
      "filterArn": "string",
      "lastUpdatedDateTime": number,
    }
  ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**Filters (p. 556)**

A list of returned filters.

Type: Array of FilterSummary (p. 683) objects

Array Members: Maximum number of 100 items.

**nextToken (p. 556)**

A token for getting the next set of filters (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

**Errors**

InvalidInputException

提供一个有效的字段或参数值。

HTTP Status Code: 400

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
**ListMetricAttributionMetrics**  
Service: Amazon Personalize  
Lists the metrics for the metric attribution.

**Request Syntax**

```
{
    "maxResults": number,
    "metricAttributionArn": "string",
    "nextToken": "string"
}
```

**Request Parameters**

The request accepts the following data in JSON format.

**maxResults (p. 558)**

The maximum number of metrics to return in one page of results.

- Type: Integer
- Valid Range: Minimum value of 1. Maximum value of 100.
- Required: No

**metricAttributionArn (p. 558)**

The Amazon Resource Name (ARN) of the metric attribution to retrieve attributes for.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+
- Required: No

**nextToken (p. 558)**

Specify the pagination token from a previous request to retrieve the next page of results.

- Type: String
- Length Constraints: Maximum length of 1500.
- Required: No

**Response Syntax**

```
{
    "metrics": [
        {
            "eventType": "string",
            "expression": "string",
            "metricName": "string"
        }
    ],
    "nextToken": "string"
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

metrics (p. 558)

The metrics for the specified metric attribution.

Type: Array of MetricAttribute (p. 690) objects

Array Members: Maximum number of 10 items.

nextToken (p. 558)

Specify the pagination token from a previous ListMetricAttributionMetricsResponse request to retrieve the next page of results.

Type: String

Length Constraints: Maximum length of 1500.

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
ListMetricAttributions
Service: Amazon Personalize
Lists metric attributions.

Request Syntax

```
{
    "datasetGroupArn": "string",
    "maxResults": number,
    "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 560)

The metric attributions' dataset group Amazon Resource Name (ARN).

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+.*.*.

Required: No

maxResults (p. 560)

The maximum number of metric attributions to return in one page of results.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

nextToken (p. 560)

Specify the pagination token from a previous request to retrieve the next page of results.

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
{
    "metricAttributions": [
        {
            "creationDateTime": number,
            "failureReason": "string",
            "lastUpdatedDateTime": number,
            "metricAttributionArn": "string",
            "name": "string",
            "status": "string"
        }
    ]
}
```

560
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**metricAttributions** *(p. 560)*

The list of metric attributions.

Type: Array of **MetricAttributionSummary** *(p. 694)* objects

Array Members: Maximum number of 100 items.

**nextToken** *(p. 560)*

Specify the pagination token from a previous request to retrieve the next page of results.

Type: String

Length Constraints: Maximum length of 1500.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
ListRecipes
Service: Amazon Personalize

Returns a list of available recipes. The response provides the properties for each recipe, including the recipe's Amazon Resource Name (ARN).

Request Syntax

```
{
  "domain": "string",
  "maxResults": number,
  "nextToken": "string",
  "recipeProvider": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**domain (p. 562)**

Filters returned recipes by domain for a Domain dataset group. Only recipes (Domain dataset group use cases) for this domain are included in the response. If you don't specify a domain, all recipes are returned.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

Required: No

**maxResults (p. 562)**

The maximum number of recipes to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 562)**

A token returned from the previous call to ListRecipes for getting the next set of recipes (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

**recipeProvider (p. 562)**

The default is SERVICE.

Type: String

Valid Values: SERVICE

Required: No
Response Syntax

```json
{
    "nextToken": "string",
    "recipes": [
        {
            "creationDateTime": "string",
            "domain": "string",
            "lastUpdatedDateTime": "string",
            "name": "string",
            "recipeArn": "string",
            "status": "string"
        }
    ]
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**nextToken (p. 563)**

A token for getting the next set of recipes.

Type: String

Length Constraints: Maximum length of 1500.

**recipes (p. 563)**

The list of available recipes.

Type: Array of RecipeSummary (p. 699) objects

Array Members: Maximum number of 100 items.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**InvalidNextTokenException**

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
ListRecommenders
Service: Amazon Personalize

Returns a list of recommenders in a given Domain dataset group. When a Domain dataset group is not specified, all the recommenders associated with the account are listed. The response provides the properties for each recommender, including the Amazon Resource Name (ARN). For more information on recommenders, see CreateRecommender.

Request Syntax

```
{
    "datasetGroupArn": "string",
    "maxResults": number,
    "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 565)

The Amazon Resource Name (ARN) of the Domain dataset group to list the recommenders for. When a Domain dataset group is not specified, all the recommenders associated with the account are listed.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.+:.

Required: No

maxResults (p. 565)

The maximum number of recommenders to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

nextToken (p. 565)

A token returned from the previous call to ListRecommenders for getting the next set of recommenders (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
{
    "nextToken": "string",
    "recommenders": [
```
{  
  "creationDateTime": number,
  "datasetGroupArn": "string",
  "lastUpdatedDateTime": number,
  "name": "string",
  "recipeArn": "string",
  "recommenderArn": "string",
  "recommenderConfig": {
    "itemExplorationConfig": {
      "string": "string"
    },
    "minRecommendationRequestsPerSecond": number,
    "trainingDataConfig": {
      "excludedDatasetColumns": {
        "string": [ "string" ]
      }
    }
  },
  "status": "string"
}

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

nextToken (p. 565)

A token for getting the next set of recommenders (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

recommenders (p. 565)

A list of the recommenders.

Type: Array of RecommenderSummary (p. 705) objects

Array Members: Maximum number of 100 items.

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:
• AWS Command Line Interface
• AWS SDK for .NET
• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
ListSchemas
Service: Amazon Personalize

Returns the list of schemas associated with the account. The response provides the properties for each schema, including the Amazon Resource Name (ARN). For more information on schemas, see CreateSchema.

Request Syntax

```
{
    "maxResults": number,
    "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**maxResults (p. 568)**

The maximum number of schemas to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 568)**

A token returned from the previous call to ListSchemas for getting the next set of schemas (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```
{
    "nextToken": "string",
    "schemas": [
        {
            "creationDateTime": number,
            "domain": "string",
            "lastUpdatedDateTime": number,
            "name": "string",
            "schemaArn": "string"
        }
    ]
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.
nextToken (p. 568)

A token used to get the next set of schemas (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

schemas (p. 568)

A list of schemas.

Type: Array of DatasetSchemaSummary (p. 664) objects

Array Members: Maximum number of 100 items.

Errors

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
ListSolutions
Service: Amazon Personalize

Returns a list of solutions that use the given dataset group. When a dataset group is not specified, all the solutions associated with the account are listed. The response provides the properties for each solution, including the Amazon Resource Name (ARN). For more information on solutions, see CreateSolution.

Request Syntax

```json
{
  "datasetGroupArn": "string",
  "maxResults": number,
  "nextToken": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetGroupArn (p. 570)

The Amazon Resource Name (ARN) of the dataset group.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-zA-z\d-]+):personalize:.+.*.*.*+

Required: No

maxResults (p. 570)

The maximum number of solutions to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

nextToken (p. 570)

A token returned from the previous call to ListSolutions for getting the next set of solutions (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

Response Syntax

```json
{
  "nextToken": "string",
  "solutions": [
    {
      "creationDateTime": number,
      "lastUpdatedDateTime": number,
    }
  ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

nextToken (p. 570)

A token for getting the next set of solutions (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

solutions (p. 570)

A list of the current solutions.

Type: Array of SolutionSummary (p. 715) objects

Array Members: Maximum number of 100 items.

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
ListSolutionVersions
Service: Amazon Personalize

Returns a list of solution versions for the given solution. When a solution is not specified, all the solution versions associated with the account are listed. The response provides the properties for each solution version, including the Amazon Resource Name (ARN).

Request Syntax

```json
{
    "maxResults": number,
    "nextToken": "string",
    "solutionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**maxResults (p. 573)**

The maximum number of solution versions to return.

Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

**nextToken (p. 573)**

A token returned from the previous call to ListSolutionVersions for getting the next set of solution versions (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

Required: No

**solutionArn (p. 573)**

The Amazon Resource Name (ARN) of the solution.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.+`

Required: No

Response Syntax

```json
{
    "nextToken": "string",
    "SolutionVersions": [
        {
            "creationDateTime": number,
            "failureReason": "string",
            ...
        }
    ]
}
```
Response Elements

If the action is successful, the service sends back an HTTP 200 response. The following data is returned in JSON format by the service.

nextToken (p. 573)

A token for getting the next set of solution versions (if they exist).

Type: String

Length Constraints: Maximum length of 1500.

solutionVersions (p. 573)

A list of solution versions describing the version properties.

Type: Array of SolutionVersionSummary (p. 721) objects

Array Members: Maximum number of 100 items.

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

InvalidNextTokenException

The token is not valid.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
ListTagsForResource
Service: Amazon Personalize

Get a list of tags attached to a resource.

Request Syntax

```json
{
   "resourceArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

resourceArn (p. 576)

The resource's Amazon Resource Name.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:*:.+:

Required: Yes

Response Syntax

```json
{
   "tags": [
      {
         "tagKey": "string",
         "tagValue": "string"
      }
   ]
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

tags (p. 576)

The resource's tags.

Type: Array of Tag (p. 723) objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Errors

InvalidInputException

Provide a valid value for the field or parameter.
HTTP Status Code: 400
**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400
**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
StartRecommender
Service: Amazon Personalize

Starts a recommender that is INACTIVE. Starting a recommender does not create any new models, but resumes billing and automatic retraining for the recommender.

Request Syntax

```json
{
   "recommenderArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**recommenderArn (p. 578)**

The Amazon Resource Name (ARN) of the recommender to start.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.*+`

Required: Yes

Response Syntax

```json
{
   "recommenderArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**recommenderArn (p. 578)**

The Amazon Resource Name (ARN) of the recommender you started.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.*+`

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400
ResourceInUseException

The specified resource is in use.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
StopRecommender
Service: Amazon Personalize

Stops a recommender that is ACTIVE. Stopping a recommender halts billing and automatic retraining for the recommender.

Request Syntax

```
{
  "recommenderArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

`recommenderArn (p. 580)`

The Amazon Resource Name (ARN) of the recommender to stop.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: `arn:(\[a-z\d-]+):personalize:.+:*:*:*+`
- Required: Yes

Response Syntax

```
{
  "recommenderArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

`recommenderArn (p. 580)`

The Amazon Resource Name (ARN) of the recommender you stopped.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: `arn:(\[a-z\d-]+):personalize:.+:*:*:*+

Errors

`InvalidInputException`

Provide a valid value for the field or parameter.

HTTP Status Code: 400
**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
StopSolutionVersionCreation

Service: Amazon Personalize

Stops creating a solution version that is in a state of CREATE_PENDING or CREATE_IN_PROGRESS.

Depending on the current state of the solution version, the solution version state changes as follows:

- CREATE_PENDING > CREATE_STOPPED
  
or
- CREATE_IN_PROGRESS > CREATE_STOPPING > CREATE_STOPPED

You are billed for all of the training completed up until you stop the solution version creation. You cannot resume creating a solution version once it has been stopped.

Request Syntax

```json
{
  "solutionVersionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**solutionVersionArn** *(p. 582)*

The Amazon Resource Name (ARN) of the solution version you want to stop creating.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:[a-z\d-]+:personalize:.+:*.+.

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.
HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
TagResource
Service: Amazon Personalize

Add a list of tags to a resource.

Request Syntax

```json
{
  "resourceArn": "string",
  "tags": [
    {
      "tagKey": "string",
      "tagValue": "string"
    }
  ]
}
```

Request Parameters

The request accepts the following data in JSON format.

**resourceArn (p. 584)**

The resource's Amazon Resource Name (ARN).

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:(\[a-z\d-]+):personalize:.*:.*:.+`

Required: Yes

**tags (p. 584)**

Tags to apply to the resource. For more information see [Tagging Amazon Personalize resources](#).

Type: Array of [Tag (p. 723)] objects

Array Members: Minimum number of 0 items. Maximum number of 200 items.

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**LimitExcededException**

The limit on the number of requests per second has been exceeded.

HTTP Status Code: 400
ResourceInUseException

The specified resource is in use.

HTTP Status Code: 400

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

TooManyTagsException

You have exceeded the maximum number of tags you can apply to this resource.

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
UntagResource
Service: Amazon Personalize

Remove tags that are attached to a resource.

Request Syntax

```json
{
   "resourceArn": "string",
   "tagKeys": [ "string" ]
}
```

Request Parameters

The request accepts the following data in JSON format.

**resourceArn (p. 586)**

The resource's Amazon Resource Name (ARN).

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+.*.*

Required: Yes

**tagKeys (p. 586)**

Keys to remove from the resource's tags.

Type: Array of strings

Array Members: Minimum number of 0 items. Maximum number of 200 items.


Pattern: ^([^\p{L}\p{Z}\p{N}\p{L}_,/-@=+\-]*)$

Required: Yes

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400
ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 400

TooManyTagKeysException

The request contains more tag keys than can be associated with a resource (50 tag keys per resource).

HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
UpdateCampaign
Service: Amazon Personalize

Updates a campaign to deploy a retrained solution version with an existing campaign, change your campaign's minProvisionedTPS, or modify your campaign's configuration, such as the exploration configuration.

To update a campaign, the campaign status must be ACTIVE or CREATE FAILED. Check the campaign status using the DescribeCampaign operation.

Note
You can still get recommendations from a campaign while an update is in progress. The campaign will use the previous solution version and campaign configuration to generate recommendations until the latest campaign update status is Active.

For more information about updating a campaign, including code samples, see Updating a campaign. For more information about campaigns, see Creating a campaign.

Request Syntax

```json
{
    "campaignArn": "string",
    "campaignConfig": {
        "itemExplorationConfig": {
            "string": "string"
        }
    },
    "minProvisionedTPS": number,
    "solutionVersionArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

**campaignArn (p. 588)**

The Amazon Resource Name (ARN) of the campaign.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*

Required: Yes

**campaignConfig (p. 588)**

The configuration details of a campaign.

Type: CampaignConfig (p. 638) object

Required: No

**minProvisionedTPS (p. 588)**

Specifies the requested minimum provisioned transactions (recommendations) per second that Amazon Personalize will support. A high minProvisionedTPS will increase your bill. We recommend starting with 1 for minProvisionedTPS (the default). Track your usage using Amazon CloudWatch metrics, and increase the minProvisionedTPS as necessary.
Type: Integer
Valid Range: Minimum value of 1.
Required: No

**solutionVersionArn (p. 588)**
The ARN of a new solution version to deploy.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+\*:\*:\+
Required: No

**Response Syntax**

```json
{
  "campaignArn": "string"
}
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**campaignArn (p. 589)**
The same campaign ARN as given in the request.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.*:*:*+  

**Errors**

**InvalidInputException**
Provide a valid value for the field or parameter.
HTTP Status Code: 400

**ResourceInUseException**
The specified resource is in use.
HTTP Status Code: 400

**ResourceNotFoundError**
Could not find the specified resource.
HTTP Status Code: 400
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3
UpdateDataset
Service: Amazon Personalize

Update a dataset to replace its schema with a new or existing one. For more information, see Replacing a dataset's schema.

Request Syntax

```
{
   "datasetArn": "string",
   "schemaArn": "string"
}
```

Request Parameters

The request accepts the following data in JSON format.

datasetArn (p. 591)
The Amazon Resource Name (ARN) of the dataset that you want to update.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+.*+.+.*
Required: Yes

schemaArn (p. 591)
The Amazon Resource Name (ARN) of the new schema you want use.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+.*+.+.*
Required: Yes

Response Syntax

```
{
   "datasetArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

datasetArn (p. 591)
The Amazon Resource Name (ARN) of the dataset you updated.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+

Errors

**InvalidInputException**
Provide a valid value for the field or parameter.
HTTP Status Code: 400

**ResourceInUseException**
The specified resource is in use.
HTTP Status Code: 400

**ResourceNotFoundException**
Could not find the specified resource.
HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
UpdateMetricAttribution
Service: Amazon Personalize
Updates a metric attribution.

Request Syntax

```json
{
    "addMetrics": [
        {
            "eventType": "string",
            "expression": "string",
            "metricName": "string"
        }
    ],
    "metricAttributionArn": "string",
    "metricsOutputConfig": {
        "roleArn": "string",
        "s3DataDestination": {
            "kmsKeyArn": "String",
            "path": "string"
        }
    },
    "removeMetrics": [ "string" ]
}
```

Request Parameters

The request accepts the following data in JSON format.

**addMetrics (p. 593)**

Add new metric attributes to the metric attribution.

- Type: Array of [MetricAttribute (p. 690)] objects
- Array Members: Maximum number of 10 items.
- Required: No

**metricAttributionArn (p. 593)**

The Amazon Resource Name (ARN) for the metric attribution to update.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: arn:\[a-zA-Z\d-]+:personalize:.+\*:\*:.*
- Required: No

**metricsOutputConfig (p. 593)**

An output config for the metric attribution.

- Type: [MetricAttributionOutput (p. 693)] object
- Required: No

**removeMetrics (p. 593)**

Remove metric attributes from the metric attribution.
Type: Array of strings
Array Members: Maximum number of 10 items.
Length Constraints: Maximum length of 256.
Required: No

Response Syntax

```
{
  "metricAttributionArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**metricAttributionArn (p. 594)**

The Amazon Resource Name (ARN) for the metric attribution that you updated.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*:*.*+

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.
HTTP Status Code: 400

**ResourceAlreadyExistsException**

The specified resource already exists.
HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.
HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.
HTTP Status Code: 400

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:
• AWS Command Line Interface
• AWS SDK for .NET
• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
UpdateRecommender
Service: Amazon Personalize

Updates the recommender to modify the recommender configuration. If you update the recommender
to modify the columns used in training, Amazon Personalize automatically starts a full retraining of the
models backing your recommender. While the update completes, you can still get recommendations
from the recommender. The recommender uses the previous configuration until the update
completes. To track the status of this update, use the latestRecommenderUpdate returned in the
DescribeRecommender operation.

Request Syntax

```
{
  "recommenderArn": "string",
  "recommenderConfig": {
    "itemExplorationConfig": {
      "string": "string"
    },
    "minRecommendationRequestsPerSecond": number,
    "trainingDataConfig": {
      "excludedDatasetColumns": {
        "string": ["string"]
      }
    }
  }
}
```

Request Parameters

The request accepts the following data in JSON format.

**recommenderArn (p. 596)**

The Amazon Resource Name (ARN) of the recommender to modify.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.+:.

Required: Yes

**recommenderConfig (p. 596)**

The configuration details of the recommender.

Type: RecommenderConfig (p. 704) object

Required: Yes

Response Syntax

```
{
  "recommenderArn": "string"
}
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response.
The following data is returned in JSON format by the service.

**recommenderArn (p. 596)**

The same recommender Amazon Resource Name (ARN) as given in the request.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+`.

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceInUseException**

The specified resource is in use.

HTTP Status Code: 400

**ResourceNotFoundException**

Could not find the specified resource.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)

**Amazon Personalize Events**

The following actions are supported by Amazon Personalize Events:

- [PutEvents (p. 598)](#)
- [PutItems (p. 600)](#)
- [PutUsers (p. 602)](#)


PutEvents
Service: Amazon Personalize Events

Records user interaction event data. For more information see Recording Events.

Request Syntax

POST /events HTTP/1.1
Content-type: application/json

{  
  
  "eventList": [  
    
    {        
      "eventId": "string",  
      "eventType": "string",  
      "eventValue": number,  
      "impression": [ "string" ],  
      "itemId": "string",  
      "metricAttribution": {  
        "eventAttributionSource": "string"  
      },  
      "properties": "string",  
      "recommendationId": "string",  
      "sentAt": number  
    }  
  ],  
  "sessionId": "string",  
  "trackingId": "string",  
  "userId": "string"  
}

URI Request Parameters

The request does not use any URI parameters.

Request Body

The request accepts the following data in JSON format.

**eventList (p. 598)**

A list of event data from the session.

Type: Array of [Event (p. 726)] objects

Array Members: Minimum number of 1 item. Maximum number of 10 items.

Required: Yes

**sessionId (p. 598)**

The session ID associated with the user's visit. Your application generates the sessionId when a user first visits your website or uses your application. Amazon Personalize uses the sessionId to associate events with the user before they log in. For more information, see Recording Events.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: Yes
**trackingId (p. 598)**

The tracking ID for the event. The ID is generated by a call to the `CreateEventTracker` API.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: Yes

**userId (p. 598)**

The user associated with the event.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: No

**Response Syntax**

```
HTTP/1.1 200
```

**Response Elements**

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.

**Errors**

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
PutItems
Service: Amazon Personalize Events

Adds one or more items to an Items dataset. For more information see Importing Items Incrementally.

Request Syntax

```json
POST /items HTTP/1.1
Content-type: application/json

{
  "datasetArn": "string",
  "items": [
    {
      "itemId": "string",
      "properties": "string"
    }
  ]
}
```

URI Request Parameters

The request does not use any URI parameters.

Request Body

The request accepts the following data in JSON format.

**datasetArn (p. 600)**

The Amazon Resource Name (ARN) of the Items dataset you are adding the item or items to.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:+

Required: Yes

**items (p. 600)**

A list of item data.

Type: Array of [item (p. 729)] objects

Array Members: Minimum number of 1 item. Maximum number of 10 items.

Required: Yes

Response Syntax

HTTP/1.1 200

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.
Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceInUseException

The specified resource is in use.

HTTP Status Code: 409

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 404

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS Command Line Interface](#)
- [AWS SDK for .NET](#)
- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for JavaScript](#)
- [AWS SDK for PHP V3](#)
- [AWS SDK for Python](#)
- [AWS SDK for Ruby V3](#)
PutUsers
Service: Amazon Personalize Events

Adds one or more users to a Users dataset. For more information see Importing Users Incrementally.

Request Syntax

```json
POST /users HTTP/1.1
Content-type: application/json

{
    "datasetArn": "string",
    "users": [
        {
            "properties": "string",
            "userId": "string"
        }
    ]
}
```

URI Request Parameters

The request does not use any URI parameters.

Request Body

The request accepts the following data in JSON format.

**datasetArn (p. 602)**

The Amazon Resource Name (ARN) of the Users dataset you are adding the user or users to.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:\[[a-z\d-]+:personalize:.+:.*:.+:

Required: Yes

**users (p. 602)**

A list of user data.

Type: Array of User (p. 731) objects

Array Members: Minimum number of 1 item. Maximum number of 10 items.

Required: Yes

Response Syntax

```
HTTP/1.1 200
```

Response Elements

If the action is successful, the service sends back an HTTP 200 response with an empty HTTP body.
Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceInUseException

The specified resource is in use.

HTTP Status Code: 409

ResourceNotFoundException

Could not find the specified resource.

HTTP Status Code: 404

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
- AWS SDK for Ruby V3

Amazon Personalize Runtime

The following actions are supported by Amazon Personalize Runtime:

- GetPersonalizedRanking (p. 604)
- GetRecommendations (p. 608)
GetPersonalizedRanking
Service: Amazon Personalize Runtime

Re-ranks a list of recommended items for the given user. The first item in the list is deemed the most likely item to be of interest to the user.

**Note**
The solution backing the campaign must have been created using a recipe of type PERSONALIZED_RANKING.

**Request Syntax**

```plaintext
POST /personalize-ranking HTTP/1.1
Content-type: application/json

{   "campaignArn": "string",
    "context": {
        "String": "string"
    },
    "filterArn": "string",
    "FilterValues": {
        "string": "string"
    },
    "inputList": [ "string" ],
    "userId": "string"
}
```

**URI Request Parameters**
The request does not use any URI parameters.

**Request Body**
The request accepts the following data in JSON format.

**campaignArn (p. 604)**
The Amazon Resource Name (ARN) of the campaign to use for generating the personalized ranking.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:+:+

Required: Yes

**context (p. 604)**
The contextual metadata to use when getting recommendations. Contextual metadata includes any interaction information that might be relevant when getting a user's recommendations, such as the user's current location or device type.

Type: String to string map

Map Entries: Maximum number of 150 items.

Key Length Constraints: Maximum length of 150.

Key Pattern: [A-Za-z\d\_]+
Value Length Constraints: Maximum length of 1000.

Required: No

**filterArn (p. 604)**

The Amazon Resource Name (ARN) of a filter you created to include items or exclude items from recommendations for a given user. For more information, see Filtering Recommendations.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([-a-z\d-]+):personalize:.+:.*:*.*`  

Required: No

**filterValues (p. 604)**

The values to use when filtering recommendations. For each placeholder parameter in your filter expression, provide the parameter name (in matching case) as a key and the filter value(s) as the corresponding value. Separate multiple values for one parameter with a comma.

For filter expressions that use an INCLUDE element to include items, you must provide values for all parameters that are defined in the expression. For filters with expressions that use an EXCLUDE element to exclude items, you can omit the filter-values. In this case, Amazon Personalize doesn't use that portion of the expression to filter recommendations.

For more information, see Filtering Recommendations.

Type: String to string map

Map Entries: Maximum number of 25 items.

Key Length Constraints: Maximum length of 50.

Key Pattern: `[A-Za-z0-9_]`+  

Value Length Constraints: Maximum length of 1000.

Required: No

**inputList (p. 604)**

A list of items (by `itemId`) to rank. If an item was not included in the training dataset, the item is appended to the end of the reranked list. The maximum is 500.

Type: Array of strings

Length Constraints: Maximum length of 256.

Required: Yes

**userId (p. 604)**

The user for which you want the campaign to provide a personalized ranking.

Type: String

Length Constraints: Maximum length of 256.

Required: Yes
Response Syntax

HTTP/1.1 200
Content-type: application/json

{  
  "personalizedRanking": [  
    {  
      "itemId": "string",  
      "promotionName": "string",  
      "score": number  
    }  
  ],  
  "recommendationId": "string"  
}

Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

personalizedRanking (p. 606)

A list of items in order of most likely interest to the user. The maximum is 500.

Type: Array of PredictedItem (p. 732) objects

recommendationId (p. 606)

The ID of the recommendation.

Type: String

Errors

InvalidInputException

Provide a valid value for the field or parameter.

HTTP Status Code: 400

ResourceNotFoundException

The specified resource does not exist.

HTTP Status Code: 404

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
• AWS SDK for PHP V3
• AWS SDK for Python
• AWS SDK for Ruby V3
GetRecommendations
Service: Amazon Personalize Runtime

Returns a list of recommended items. For campaigns, the campaign's Amazon Resource Name (ARN) is required and the required user and item input depends on the recipe type used to create the solution backing the campaign as follows:

- USER_PERSONALIZATION - userId required, itemId not used
- RELATED_ITEMS - itemId required, userId not used

**Note**
Campaigns that are backed by a solution created using a recipe of type PERSONALIZED_RANKING use the GetPersonalizedRanking (p. 604) API.

For recommenders, the recommender's ARN is required and the required item and user input depends on the use case (domain-based recipe) backing the recommender. For information on use case requirements see Choosing recommender use cases.

**Request Syntax**

POST /recommendations HTTP/1.1
Content-type: application/json

```
{
  "campaignArn": "string",
  "context": {
    "string": "string"
  },
  "filterArn": "string",
  "filterValues": {
    "string": "string"
  },
  "itemId": "string",
  "numResults": number,
  "promotions": [
    {
      "filterArn": "string",
      "filterValues": {
        "string": "string"
      },
      "name": "string",
      "percentPromotedItems": number
    }
  ],
  "recommenderArn": "string",
  "userId": "string"
}
```

**URI Request Parameters**

The request does not use any URI parameters.

**Request Body**

The request accepts the following data in JSON format.

**campaignArn (p. 608)**

The Amazon Resource Name (ARN) of the campaign to use for getting recommendations.
context (p. 608)

The contextual metadata to use when getting recommendations. Contextual metadata includes any interaction information that might be relevant when getting a user's recommendations, such as the user's current location or device type.

- Type: String to string map
- Map Entries: Maximum number of 150 items.
- Key Length Constraints: Maximum length of 150.
- Key Pattern: [A-Za-z\d_]+
- Value Length Constraints: Maximum length of 1000.
- Required: No

filterArn (p. 608)

The ARN of the filter to apply to the returned recommendations. For more information, see Filtering Recommendations.

When using this parameter, be sure the filter resource is ACTIVE.

- Type: String
- Length Constraints: Maximum length of 256.
- Pattern: arn:([a-z\d-]+):personalize:.+:.*:.*
- Required: No

filterValues (p. 608)

The values to use when filtering recommendations. For each placeholder parameter in your filter expression, provide the parameter name (in matching case) as a key and the filter value(s) as the corresponding value. Separate multiple values for one parameter with a comma.

For filter expressions that use an INCLUDE element to include items, you must provide values for all parameters that are defined in the expression. For filters with expressions that use an EXCLUDE element to exclude items, you can omit the filter-values. In this case, Amazon Personalize doesn't use that portion of the expression to filter recommendations.

For more information, see Filtering recommendations and user segments.

- Type: String to string map
- Map Entries: Maximum number of 25 items.
- Key Length Constraints: Maximum length of 50.
- Key Pattern: [A-Za-z0-9_]+
Required: No

**itemid (p. 608)**

The item ID to provide recommendations for.

Required for RELATED_ITEMS recipe type.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**numResults (p. 608)**

The number of results to return. The default is 25. The maximum is 500.

Type: Integer

Valid Range: Minimum value of 0.

Required: No

**promotions (p. 608)**

The promotions to apply to the recommendation request. A promotion defines additional business rules that apply to a configurable subset of recommended items.

Type: Array of Promotion (p. 733) objects

Array Members: Maximum number of 1 item.

Required: No

**recommenderArn (p. 608)**

The Amazon Resource Name (ARN) of the recommender to use to get recommendations. Provide a recommender ARN if you created a Domain dataset group with a recommender for a domain use case.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.*:.*:+

Required: No

**userId (p. 608)**

The user ID to provide recommendations for.

Required for USER_PERSONALIZATION recipe type.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**Response Syntax**

HTTP/1.1 200
Response Elements

If the action is successful, the service sends back an HTTP 200 response.

The following data is returned in JSON format by the service.

**itemList (p. 610)**

A list of recommendations sorted in descending order by prediction score. There can be a maximum of 500 items in the list.

Type: Array of [PredictedItem (p. 732)] objects

**recommendationId (p. 610)**

The ID of the recommendation.

Type: String

Errors

**InvalidInputException**

Provide a valid value for the field or parameter.

HTTP Status Code: 400

**ResourceNotFoundException**

The specified resource does not exist.

HTTP Status Code: 404

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS Command Line Interface
- AWS SDK for .NET
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for JavaScript
- AWS SDK for PHP V3
- AWS SDK for Python
Data Types

The following data types are supported by Amazon Personalize:

- Algorithm (p. 616)
- AlgorithmImage (p. 618)
- AutoMLConfig (p. 619)
- AutoMLResult (p. 620)
- BatchInferenceJob (p. 621)
- BatchInferenceJobConfig (p. 624)
- BatchInferenceJobInput (p. 625)
- BatchInferenceJobOutput (p. 626)
- BatchInferenceJobSummary (p. 627)
- BatchSegmentJob (p. 629)
- BatchSegmentJobInput (p. 632)
- BatchSegmentJobOutput (p. 633)
- BatchSegmentJobSummary (p. 634)
- Campaign (p. 636)
- CampaignConfig (p. 638)
- CampaignSummary (p. 639)
- CampaignUpdateSummary (p. 641)
- CategoricalHyperParameterRange (p. 643)
- ContinuousHyperParameterRange (p. 644)
- Dataset (p. 645)
- DatasetExportJob (p. 647)
- DatasetExportJobOutput (p. 650)
- DatasetExportJobSummary (p. 651)
- DatasetGroup (p. 653)
- DatasetGroupSummary (p. 655)
- DatasetImportJob (p. 657)
- DatasetImportJobSummary (p. 660)
- DatasetSchema (p. 662)
- DatasetSchemaSummary (p. 664)
- DatasetSummary (p. 666)
- DatasetUpdateSummary (p. 668)
- DataSource (p. 670)
- DefaultCategoricalHyperParameterRange (p. 671)
- DefaultContinuousHyperParameterRange (p. 672)
- DefaultHyperParameterRanges (p. 673)
- DefaultIntegerHyperParameterRange (p. 674)
- EventTracker (p. 675)
- EventTrackerSummary (p. 677)
- FeatureTransformation (p. 679)
The following data types are supported by Amazon Personalize Events:

- Event (p. 726)
- Item (p. 729)
- MetricAttribution (p. 730)
- User (p. 731)

The following data types are supported by Amazon Personalize Runtime:

- PredictedItem (p. 732)
- Promotion (p. 733)

Amazon Personalize

The following data types are supported by Amazon Personalize:

- Algorithm (p. 616)
- AlgorithmImage (p. 618)
- AutoMLConfig (p. 619)
• AutoMLResult (p. 620)
• BatchInferenceJob (p. 621)
• BatchInferenceJobConfig (p. 624)
• BatchInferenceJobInput (p. 625)
• BatchInferenceJobOutput (p. 626)
• BatchInferenceJobSummary (p. 627)
• BatchSegmentJob (p. 629)
• BatchSegmentJobInput (p. 632)
• BatchSegmentJobOutput (p. 633)
• BatchSegmentJobSummary (p. 634)
• Campaign (p. 636)
• CampaignConfig (p. 638)
• CampaignSummary (p. 639)
• CampaignUpdateSummary (p. 641)
• CategoricalHyperParameterRange (p. 643)
• ContinuousHyperParameterRange (p. 644)
• Dataset (p. 645)
• DatasetExportJob (p. 647)
• DatasetExportJobOutput (p. 650)
• DatasetExportJobSummary (p. 651)
• DatasetGroup (p. 653)
• DatasetGroupSummary (p. 655)
• DatasetImportJob (p. 657)
• DatasetImportJobSummary (p. 660)
• DatasetSchema (p. 662)
• DatasetSchemaSummary (p. 664)
• DatasetSummary (p. 666)
• DatasetUpdateSummary (p. 668)
• DataSource (p. 670)
• DefaultCategoricalHyperParameterRange (p. 671)
• DefaultContinuousHyperParameterRange (p. 672)
• DefaultHyperParameterRanges (p. 673)
• DefaultIntegerHyperParameterRange (p. 674)
• EventTracker (p. 675)
• EventTrackerSummary (p. 677)
• FeatureTransformation (p. 679)
• Filter (p. 681)
• FilterSummary (p. 683)
• HPOConfig (p. 685)
• HPOObjective (p. 686)
• HPOResourceConfig (p. 687)
• HyperParameterRanges (p. 688)
• IntegerHyperParameterRange (p. 689)
• MetricAttribute (p. 690)
• MetricAttribution (p. 691)
• MetricAttributionOutput (p. 693)
- MetricAttributionSummary (p. 694)
- OptimizationObjective (p. 696)
- Recipe (p. 697)
- RecipeSummary (p. 699)
- Recommender (p. 701)
- RecommenderConfig (p. 704)
- RecommenderSummary (p. 705)
- RecommenderUpdateSummary (p. 707)
- S3DataConfig (p. 709)
- Solution (p. 710)
- SolutionConfig (p. 713)
- SolutionSummary (p. 715)
- SolutionVersion (p. 717)
- SolutionVersionSummary (p. 721)
- Tag (p. 723)
- TrainingDataConfig (p. 724)
- TunedHPOParams (p. 725)
Algorithm
Service: Amazon Personalize

Describes a custom algorithm.

Contents

algorithmArn
The Amazon Resource Name (ARN) of the algorithm.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+
Required: No

algorithmImage
The URI of the Docker container for the algorithm image.
Type: AlgorithmImage (p. 618) object
Required: No

creationDateTime
The date and time (in Unix time) that the algorithm was created.
Type: Timestamp
Required: No

defaultHyperParameterRanges
Specifies the default hyperparameters, their ranges, and whether they are tunable. A tunable hyperparameter can have its value determined during hyperparameter optimization (HPO).
Type: DefaultHyperParameterRanges (p. 673) object
Required: No

defaultHyperParameters
Specifies the default hyperparameters.
Type: String to string map
Map Entries: Maximum number of 100 items.
Key Length Constraints: Maximum length of 256.
Value Length Constraints: Maximum length of 1000.
Required: No

defaultResourceConfig
Specifies the default maximum number of training jobs and parallel training jobs.
Type: String to string map
Map Entries: Maximum number of 100 items.
Key Length Constraints: Maximum length of 256.

Value Length Constraints: Maximum length of 1000.

Required: No

**lastUpdatedDateTime**

The date and time (in Unix time) that the algorithm was last updated.

Type: Timestamp

Required: No

**name**

The name of the algorithm.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_]*

Required: No

**roleArn**

The Amazon Resource Name (ARN) of the role.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:+

Required: No

**trainingInputMode**

The training input mode.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
AlgorithmImage
Service: Amazon Personalize

Describes an algorithm image.

Contents

dockerURI

The URI of the Docker container for the algorithm image.

Type: String

Length Constraints: Maximum length of 256.

Required: Yes

name

The name of the algorithm image.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_]*

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
AutoMLConfig
Service: Amazon Personalize

When the solution performs AutoML (performAutoML is true in CreateSolution), Amazon Personalize determines which recipe, from the specified list, optimizes the given metric. Amazon Personalize then uses that recipe for the solution.

Contents

metricName
The metric to optimize.
Type: String
Length Constraints: Maximum length of 256.
Required: No

recipeList
The list of candidate recipes.
Type: Array of strings
Array Members: Maximum number of 100 items.
Length Constraints: Maximum length of 256.
Pattern: arn:(\[a-z\d-]+):personalize:.*:.*:.+
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
AutoMLResult

Service: Amazon Personalize

When the solution performs AutoML (performAutoML is true in CreateSolution), specifies the recipe that best optimized the specified metric.

Contents

bestRecipeArn

The Amazon Resource Name (ARN) of the best recipe.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-zA-Z0-9\-]+):personalize:*:*:*:*:*:*:*:*

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
BatchInferenceJob
Service: Amazon Personalize

Contains information on a batch inference job.

Contents

batchInferenceJobArn
The Amazon Resource Name (ARN) of the batch inference job.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+*.":+.+
Required: No

batchInferenceJobConfig
A string to string map of the configuration details of a batch inference job.
Type: BatchInferenceJobConfig (p. 624) object
Required: No

creationDateTime
The time at which the batch inference job was created.
Type: Timestamp
Required: No

failureReason
If the batch inference job failed, the reason for the failure.
Type: String
Required: No

filterArn
The ARN of the filter used on the batch inference job.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+*.":+.+
Required: No

jobInput
The Amazon S3 path that leads to the input data used to generate the batch inference job.
Type: BatchInferenceJobInput (p. 625) object
Required: No

jobName
The name of the batch inference job.
Type: String
Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_.]*
Required: No

**jobOutput**

The Amazon S3 bucket that contains the output data generated by the batch inference job.

Type: [BatchInferenceJobOutput](p. 626) object

Required: No

**lastUpdatedDateTime**

The time at which the batch inference job was last updated.

Type: Timestamp

Required: No

**numResults**

The number of recommendations generated by the batch inference job. This number includes the error messages generated for failed input records.

Type: Integer

Required: No

**roleArn**

The ARN of the Amazon Identity and Access Management (IAM) role that requested the batch inference job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):iam::\d{12}:role/\?([a-zA-Z0-9-]+=.\@\-\/]+)

Required: No

**solutionVersionArn**

The Amazon Resource Name (ARN) of the solution version from which the batch inference job was created.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize::\+::\+.

Required: No

**status**

The status of the batch inference job. The status is one of the following values:

- PENDING
- IN PROGRESS
- ACTIVE
• CREATE FAILED
  
  Type: String
  
  Length Constraints: Maximum length of 256.
  
  Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for Ruby V3
BatchInferenceJobConfig
Service: Amazon Personalize

The configuration details of a batch inference job.

Contents

**itemExplorationConfig**

A string to string map specifying the exploration configuration hyperparameters, including `explorationWeight` and `explorationItemAgeCutOff`, you want to use to configure the amount of item exploration Amazon Personalize uses when recommending items. See [User-Personalization](#).

Type: String to string map

Map Entries: Maximum number of 100 items.

Key Length Constraints: Maximum length of 256.

Value Length Constraints: Maximum length of 1000.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
BatchInferenceJobInput
Service: Amazon Personalize

The input configuration of a batch inference job.

Contents

s3DataSource

The URI of the Amazon S3 location that contains your input data. The Amazon S3 bucket must be in the same region as the API endpoint you are calling.

Type: S3DataConfig (p. 709) object

Required: Yes

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
BatchInferenceJobOutput
Service: Amazon Personalize

The output configuration parameters of a batch inference job.

Contents

**s3DataDestination**

Information on the Amazon S3 bucket in which the batch inference job's output is stored.

Type: [S3DataConfig](#) object

Required: Yes

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
BatchInferenceJobSummary
Service: Amazon Personalize

A truncated version of the BatchInferenceJob. The ListBatchInferenceJobs operation returns a list of batch inference job summaries.

Contents

batchInferenceJobArn

The Amazon Resource Name (ARN) of the batch inference job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:\([a-z\d-]+\):personalize:.+:.+

Required: No

creationDateTime

The time at which the batch inference job was created.

Type: Timestamp

Required: No

failureReason

If the batch inference job failed, the reason for the failure.

Type: String

Required: No

jobName

The name of the batch inference job.

Type: String


Pattern: ^[a-zA-Z0-9-][a-zA-Z0-9\-\_]*$

Required: No

lastUpdatedDateTime

The time at which the batch inference job was last updated.

Type: Timestamp

Required: No

solutionVersionArn

The ARN of the solution version used by the batch inference job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:\([a-z\d-]+\):personalize:.+:.+


Required: No

**status**

The status of the batch inference job. The status is one of the following values:

- PENDING
- IN PROGRESS
- ACTIVE
- CREATE FAILED

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
BatchSegmentJob
Service: Amazon Personalize

Contains information on a batch segment job.

Contents

batchSegmentJobArn

The Amazon Resource Name (ARN) of the batch segment job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*

Required: No

creationDateTime

The time at which the batch segment job was created.

Type: Timestamp

Required: No

failureReason

If the batch segment job failed, the reason for the failure.

Type: String

Required: No

filterArn

The ARN of the filter used on the batch segment job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*

Required: No

jobInput

The Amazon S3 path that leads to the input data used to generate the batch segment job.

Type: BatchSegmentJobInput (p. 632) object

Required: No

jobName

The name of the batch segment job.

Type: String


Pattern: ^[a-zA-Z0-9\-0-9][a-zA-Z0-9\-0-9]*$
jobOutput

The Amazon S3 bucket that contains the output data generated by the batch segment job.

Type: BatchSegmentJobOutput (p. 633) object

lastUpdatedDateTime

The time at which the batch segment job last updated.

Type: Timestamp

numResults

The number of predicted users generated by the batch segment job for each line of input data. The maximum number of users per segment is 5 million.

Type: Integer

roleArn

The ARN of the Amazon Identity and Access Management (IAM) role that requested the batch segment job.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):iam::\d{12}:role/?[a-zA-Z\-_\d0-9+=,.@\-_/]+

solutionVersionArn

The Amazon Resource Name (ARN) of the solution version used by the batch segment job to generate batch segments.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.*:.*:.*

status

The status of the batch segment job. The status is one of the following values:

- PENDING
- IN PROGRESS
- ACTIVE
- CREATE FAILED

Type: String

Length Constraints: Maximum length of 256.
Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
BatchSegmentJobInput
Service: Amazon Personalize

The input configuration of a batch segment job.

Contents

s3DataSource

The configuration details of an Amazon S3 input or output bucket.

Type: S3DataConfig (p. 709) object

Required: Yes

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**BatchSegmentJobOutput**
Service: Amazon Personalize

The output configuration parameters of a batch segment job.

**Contents**

**s3DataDestination**

The configuration details of an Amazon S3 input or output bucket.

Type: [S3DataConfig](p. 709) object

Required: Yes

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
BatchSegmentJobSummary
Service: Amazon Personalize

A truncated version of the BatchSegmentJob datatype. ListBatchSegmentJobs operation returns a list of batch segment job summaries.

Contents

batchSegmentJobArn
The Amazon Resource Name (ARN) of the batch segment job.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:\([a-z\d-]+\):personalize:.+:.+:.
Required: No

creationDateTime
The time at which the batch segment job was created.
Type: Timestamp
Required: No

failureReason
If the batch segment job failed, the reason for the failure.
Type: String
Required: No

jobName
The name of the batch segment job.
Type: String
Pattern: ^[a-zA-Z0-9\-\_][a-zA-Z0-9\-\_]*
Required: No

lastUpdatedDateTime
The time at which the batch segment job was last updated.
Type: Timestamp
Required: No

solutionVersionArn
The Amazon Resource Name (ARN) of the solution version used by the batch segment job to generate batch segments.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.+

Required: No

status

The status of the batch segment job. The status is one of the following values:

- PENDING
- IN PROGRESS
- ACTIVE
- CREATE FAILED

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Campaign
Service: Amazon Personalize

An object that describes the deployment of a solution version. For more information on campaigns, see CreateCampaign.

Contents

campaignArn
  The Amazon Resource Name (ARN) of the campaign.
  Type: String
  Length Constraints: Maximum length of 256.
  Pattern: arn:([a-zA-Z0-9-]+):personalize:.+:+:.*
  Required: No

campaignConfig
  The configuration details of a campaign.
  Type: CampaignConfig (p. 638) object
  Required: No

creationDateTime
  The date and time (in Unix format) that the campaign was created.
  Type: Timestamp
  Required: No

failureReason
  If a campaign fails, the reason behind the failure.
  Type: String
  Required: No

lastUpdatedDateTime
  The date and time (in Unix format) that the campaign was last updated.
  Type: Timestamp
  Required: No

latestCampaignUpdate
  Provides a summary of the properties of a campaign update. For a complete listing, call the DescribeCampaign API.
  Type: CampaignUpdateSummary (p. 641) object
  Required: No

minProvisionedTPS
  Specifies the requested minimum provisioned transactions (recommendations) per second. A high minProvisionedTPS will increase your bill. We recommend starting with 1 for
minProvisionedTPS (the default). Track your usage using Amazon CloudWatch metrics, and increase the minProvisionedTPS as necessary.

Type: Integer

Valid Range: Minimum value of 1.

Required: No

**name**

The name of the campaign.

Type: String


Pattern: ^[a-zA-Z0-9-\[\]*\-\_]*$

Required: No

**solutionVersionArn**

The Amazon Resource Name (ARN) of a specific version of the solution.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-zA-Z0-9\-]+):personalize:.+:.+:.*

Required: No

**status**

The status of the campaign.

A campaign can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
CampaignConfig
Service: Amazon Personalize

The configuration details of a campaign.

Contents

itemExplorationConfig

Specifies the exploration configuration hyperparameters, including explorationWeight and explorationItemAgeCutOff, you want to use to configure the amount of item exploration Amazon Personalize uses when recommending items. Provide itemExplorationConfig data only if your solution uses the User-Personalization recipe.

Type: String to string map

Map Entries: Maximum number of 100 items.

Key Length Constraints: Maximum length of 256.

Value Length Constraints: Maximum length of 1000.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**CampaignSummary**
Service: Amazon Personalize

Provides a summary of the properties of a campaign. For a complete listing, call the [DescribeCampaign](https://docs.aws.amazon.com/personalize/latest/developerguide/lambda-campaigns.html) API.

**Contents**

- **campaignArn**
  - The Amazon Resource Name (ARN) of the campaign.
  - Type: String
  - Length Constraints: Maximum length of 256.
  - Pattern: `arn:([a-z\d-]+):personalize:.+:.*:`
  - Required: No

- **creationDateTime**
  - The date and time (in Unix time) that the campaign was created.
  - Type: Timestamp
  - Required: No

- **failureReason**
  - If a campaign fails, the reason behind the failure.
  - Type: String
  - Required: No

- **lastUpdatedDateTime**
  - The date and time (in Unix time) that the campaign was last updated.
  - Type: Timestamp
  - Required: No

- **name**
  - The name of the campaign.
  - Type: String
  - Pattern: `^[a-zA-Z0-9\-\_]*$`
  - Required: No

- **status**
  - The status of the campaign.
  - A campaign can be in one of the following states:
  - CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
  - DELETE PENDING > DELETE IN_PROGRESS
Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Amazon Personalize Developer Guide
Amazon Personalize

CampaignUpdateSummary
Service: Amazon Personalize

Provides a summary of the properties of a campaign update. For a complete listing, call the DescribeCampaign API.

Contents

campaignConfig

The configuration details of a campaign.

Type: CampaignConfig (p. 638) object

Required: No

creationDateTime

The date and time (in Unix time) that the campaign update was created.

Type: Timestamp

Required: No

failureReason

If a campaign update fails, the reason behind the failure.

Type: String

Required: No

lastUpdatedDateTime

The date and time (in Unix time) that the campaign update was last updated.

Type: Timestamp

Required: No

minProvisionedTPS

Specifies the requested minimum provisioned transactions (recommendations) per second that Amazon Personalize will support.

Type: Integer

Valid Range: Minimum value of 1.

Required: No

solutionVersionArn

The Amazon Resource Name (ARN) of the deployed solution version.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:(\[a-z\d-]+):personalize:.+:.*:+

Required: No

status

The status of the campaign update.
A campaign update can be in one of the following states:

- CREATE PENDING > CREATE IN PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
CategoricalHyperParameterRange
Service: Amazon Personalize

Provides the name and range of a categorical hyperparameter.

Contents

name

The name of the hyperparameter.

Type: String
Length Constraints: Maximum length of 256.
Required: No

values

A list of the categories for the hyperparameter.

Type: Array of strings
Array Members: Maximum number of 100 items.
Length Constraints: Maximum length of 1000.
Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
ContinuousHyperParameterRange
Service: Amazon Personalize

Provides the name and range of a continuous hyperparameter.

Contents

**maxValue**

The maximum allowable value for the hyperparameter.

Type: Double

Valid Range: Minimum value of -1000000.

Required: No

**minValue**

The minimum allowable value for the hyperparameter.

Type: Double

Valid Range: Minimum value of -1000000.

Required: No

**name**

The name of the hyperparameter.

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Dataset
Service: Amazon Personalize

Provides metadata for a dataset.

Contents

**creationDateTime**

The creation date and time (in Unix time) of the dataset.

Type: Timestamp

Required: No

**datasetArn**

The Amazon Resource Name (ARN) of the dataset that you want metadata for.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*+:.*

Required: No

**datasetGroupArn**

The Amazon Resource Name (ARN) of the dataset group.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*+:.*

Required: No

**datasetType**

One of the following values:
- Interactions
- Items
- Users

Type: String

Length Constraints: Maximum length of 256.

Required: No

**lastUpdatedDateTime**

A time stamp that shows when the dataset was updated.

Type: Timestamp

Required: No

**latestDatasetUpdate**

Describes the latest update to the dataset.
Type: **DatasetUpdateSummary** *(p. 668)* object

Required: No

**name**

The name of the dataset.

Type: String


Pattern: `^[a-zA-Z0-9][a-zA-Z0-9\-_]*$

Required: No

**schemaArn**

The ARN of the associated schema.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+`

Required: No

**status**

The status of the dataset.

A dataset can be in one of the following states:
- CREATE PENDING > CREATE IN PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
DatasetExportJob
Service: Amazon Personalize

Describes a job that exports a dataset to an Amazon S3 bucket. For more information, see CreateDatasetExportJob.

A dataset export job can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

Contents

creationDateTime
The creation date and time (in Unix time) of the dataset export job.
Type: Timestamp
Required: No

datasetArn
The Amazon Resource Name (ARN) of the dataset to export.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+.*.*
Required: No

datasetExportJobArn
The Amazon Resource Name (ARN) of the dataset export job.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+.*.*
Required: No

failureReason
If a dataset export job fails, provides the reason why.
Type: String
Required: No

ingestionMode
The data to export, based on how you imported the data. You can choose to export BULK data that you imported using a dataset import job, PUT data that you imported incrementally (using the console, PutEvents, PutUsers and PutItems operations), or ALL for both types. The default value is PUT.
Type: String
Valid Values: BULK | PUT | ALL
Required: No

**jobName**

The name of the export job.

Type: String


Pattern: ^[a-zA-Z0-9][-_.]*

Required: No

**jobOutput**

The path to the Amazon S3 bucket where the job's output is stored. For example:

`s3://bucket-name/folder-name/`

Type: DatasetExportJobOutput (p. 650) object

Required: No

**lastUpdatedDateTime**

The date and time (in Unix time) the status of the dataset export job was last updated.

Type: Timestamp

Required: No

**roleArn**

The Amazon Resource Name (ARN) of the IAM service role that has permissions to add data to your output Amazon S3 bucket.

Type: String

Length Constraints: Maximum length of 256.

Pattern: \barn:([a-z\d-]+):personalize:.*:.*:.*+

Required: No

**status**

The status of the dataset export job.

A dataset export job can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

Type: String

Length Constraints: Maximum length of 256.

Required: No

---

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for Ruby V3
DatasetExportJobOutput
Service: Amazon Personalize

The output configuration parameters of a dataset export job.

Contents

s3DataDestination

The configuration details of an Amazon S3 input or output bucket.

Type: S3DataConfig (p. 709) object

Required: Yes

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
DatasetExportJobSummary
Service: Amazon Personalize

Provides a summary of the properties of a dataset export job. For a complete listing, call the DescribeDatasetExportJob API.

Contents

creationDateTime

The date and time (in Unix time) that the dataset export job was created.
Type: Timestamp
Required: No

datasetExportJobArn

The Amazon Resource Name (ARN) of the dataset export job.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*:+
Required: No

failureReason

If a dataset export job fails, the reason behind the failure.
Type: String
Required: No

jobName

The name of the dataset export job.
Type: String
Pattern: ^[a-zA-Z0-9\-9][a-zA-Z0-9\-9\-_]*
Required: No

lastUpdatedDateTime

The date and time (in Unix time) that the dataset export job status was last updated.
Type: Timestamp
Required: No

status

The status of the dataset export job.
A dataset export job can be in one of the following states:
• CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
Type: String
Length Constraints: Maximum length of 256.
Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
DatasetGroup
Service: Amazon Personalize

A dataset group is a collection of related datasets (Interactions, User, and Item). You create a dataset group by calling `CreateDatasetGroup`. You then create a dataset and add it to a dataset group by calling `CreateDataset`. The dataset group is used to create and train a solution by calling `CreateSolution`. A dataset group can contain only one of each type of dataset.

You can specify an AWS Key Management Service (KMS) key to encrypt the datasets in the group.

**Contents**

`creationDateTime`

The creation date and time (in Unix time) of the dataset group.

Type: Timestamp

Required: No

`datasetGroupArn`

The Amazon Resource Name (ARN) of the dataset group.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d\-]+):personalize:.+:.*:*:*:*:`

Required: No

`domain`

The domain of a Domain dataset group.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

Required: No

`failureReason`

If creating a dataset group fails, provides the reason why.

Type: String

Required: No

`kmsKeyArn`

The Amazon Resource Name (ARN) of the AWS Key Management Service (KMS) key used to encrypt the datasets.

Type: String

Length Constraints: Maximum length of 2048.

Pattern: `arn:aws:*:kms:.+:[0-9]{12}:key/.*`

Required: No
lastUpdatedDateTime

The last update date and time (in Unix time) of the dataset group.

Type: Timestamp

Required: No

ame

The name of the dataset group.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_]*

Required: No

roleArn

The ARN of the AWS Identity and Access Management (IAM) role that has permissions to access the AWS Key Management Service (KMS) key. Supplying an IAM role is only valid when also specifying a KMS key.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):iam::\d{12}:role/?[a-zA-Z_0-9+=,.@\-_/]+

Required: No

status

The current status of the dataset group.

A dataset group can be in one of the following states:

• CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
• DELETE PENDING

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for Ruby V3
**DatasetGroupSummary**

Service: Amazon Personalize

Provides a summary of the properties of a dataset group. For a complete listing, call the \nDescribeDatasetGroup\n API.

**Contents**

**creationDateTime**

The date and time (in Unix time) that the dataset group was created.

Type: Timestamp

Required: No

**datasetGroupArn**

The Amazon Resource Name (ARN) of the dataset group.

Type: String

Length Constraints: Maximum length of 256.

Pattern: \arn:([a-z\d-]+):personalize:.+:..+

Required: No

**domain**

The domain of a Domain dataset group.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

Required: No

**failureReason**

If creating a dataset group fails, the reason behind the failure.

Type: String

Required: No

**lastUpdatedDateTime**

The date and time (in Unix time) that the dataset group was last updated.

Type: Timestamp

Required: No

**name**

The name of the dataset group.

Type: String


Pattern: ^\[a-zA-Z0-9\d-9][a-zA-Z0-9\d-9]*$

Required: No
status

The status of the dataset group.

A dataset group can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE - or - CREATE FAILED
- DELETE PENDING

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**DatasetImportJob**

Service: Amazon Personalize

Describes a job that imports training data from a data source (Amazon S3 bucket) to an Amazon Personalize dataset. For more information, see `CreateDatasetImportJob`.

A dataset import job can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

**Contents**

- **creationDateTime**
  
  The creation date and time (in Unix time) of the dataset import job.

  Type: Timestamp

  Required: No

- **datasetArn**
  
  The Amazon Resource Name (ARN) of the dataset that receives the imported data.

  Type: String

  Length Constraints: Maximum length of 256.

  Pattern: `arn:([a-z\d-]+):personalize:.+:.*.*.*

  Required: No

- **datasetImportJobArn**
  
  The ARN of the dataset import job.

  Type: String

  Length Constraints: Maximum length of 256.

  Pattern: `arn:([a-z\d-]+):personalize:.+:.*.*.*

  Required: No

- **dataSource**
  
  The Amazon S3 bucket that contains the training data to import.

  Type: `DataSource` (p. 670) object

  Required: No

- **failureReason**
  
  If a dataset import job fails, provides the reason why.

  Type: String

  Required: No

- **importMode**
  
  The import mode used by the dataset import job to import new records.
jobName
The name of the import job.
Type: String
Pattern: ^[a-zA-Z0-9-\_]*$
Required: No

lastUpdatedDateTime
The date and time (in Unix time) the dataset was last updated.
Type: Timestamp
Required: No

publishAttributionMetricsToS3
Whether the job publishes metrics to Amazon S3 for a metric attribution.
Type: Boolean
Required: No

roleArn
The ARN of the IAM role that has permissions to read from the Amazon S3 data source.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:(\[a-z\d-]+):personalize:.+:.*+.+
Required: No

status
The status of the dataset import job.
A dataset import job can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:
- AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for Ruby V3
DatasetImportJobSummary
Service: Amazon Personalize

Provides a summary of the properties of a dataset import job. For a complete listing, call the DescribeDatasetImportJob API.

Contents

creationDateTime
   The date and time (in Unix time) that the dataset import job was created.
   Type: Timestamp
   Required: No

datasetImportJobArn
   The Amazon Resource Name (ARN) of the dataset import job.
   Type: String
   Length Constraints: Maximum length of 256.
   Pattern: arn:([a-z\d-]+):personalize:.+:.*+.+
   Required: No

failureReason
   If a dataset import job fails, the reason behind the failure.
   Type: String
   Required: No

importMode
   The import mode the dataset import job used to update the data in the dataset. For more information see Updating existing bulk data.
   Type: String
   Valid Values: FULL | INCREMENTAL
   Required: No

jobName
   The name of the dataset import job.
   Type: String
   Pattern: ^[a-zA-Z0-9]+[a-zA-Z0-9\-_.]*
   Required: No

lastUpdatedDateTime
   The date and time (in Unix time) that the dataset import job status was last updated.
   Type: Timestamp
Required: No

**status**

The status of the dataset import job.

A dataset import job can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
DatasetSchema
Service: Amazon Personalize

Describes the schema for a dataset. For more information on schemas, see CreateSchema.

Contents

creationDateTime
The date and time (in Unix time) that the schema was created.
Type: Timestamp
Required: No
domain
The domain of a schema that you created for a dataset in a Domain dataset group.
Type: String
Valid Values: ECOMMERCE | VIDEO_ON_DEMAND
Required: No
lastUpdatedDateTime
The date and time (in Unix time) that the schema was last updated.
Type: Timestamp
Required: No
name
The name of the schema.
Type: String
Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_.]*
Required: No
schema
The schema.
Type: String
Length Constraints: Maximum length of 10000.
Required: No
schemaArn
The Amazon Resource Name (ARN) of the schema.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:\([a-z\d-]+\):personalize:::*:::*
Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**DatasetSchemaSummary**

Service: Amazon Personalize

Provides a summary of the properties of a dataset schema. For a complete listing, call the `DescribeSchema` API.

**Contents**

`creationDateTime`

The date and time (in Unix time) that the schema was created.

Type: Timestamp

Required: No

`domain`

The domain of a schema that you created for a dataset in a Domain dataset group.

Type: String

Valid Values: ECOMMERCE | VIDEO_ON_DEMAND

Required: No

`lastUpdatedDateTime`

The date and time (in Unix time) that the schema was last updated.

Type: Timestamp

Required: No

`name`

The name of the schema.

Type: String


Pattern: `^[a-zA-Z0-9][a-zA-Z0-9\-_]*`

Required: No

`schemArn`

The Amazon Resource Name (ARN) of the schema.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.+:.*+.+

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
DatasetSummary
Service: Amazon Personalize

Provides a summary of the properties of a dataset. For a complete listing, call the DescribeDataset API.

Contents

creationDateTime
The date and time (in Unix time) that the dataset was created.
Type: Timestamp
Required: No

datasetArn
The Amazon Resource Name (ARN) of the dataset.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.+:.*
Required: No

datasetType
The dataset type. One of the following values:
• Interactions
• Items
• Users
• Event-Interactions
Type: String
Length Constraints: Maximum length of 256.
Required: No

lastUpdatedDateTime
The date and time (in Unix time) that the dataset was last updated.
Type: Timestamp
Required: No

name
The name of the dataset.
Type: String
Pattern: ^[a-zA-Z0-9-Z0-9\-\_]*$
Required: No

status
The status of the dataset.
A dataset can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
DatasetUpdateSummary

Service: Amazon Personalize

Describes an update to a dataset.

Contents

creationDateTime

The creation date and time (in Unix time) of the dataset update.

Type: Timestamp

Required: No

defailureReason

If updating a dataset fails, provides the reason why.

Type: String

Required: No

lastUpdatedDateTime

The last update date and time (in Unix time) of the dataset.

Type: Timestamp

Required: No

schemaArn

The Amazon Resource Name (ARN) of the schema that replaced the previous schema of the dataset.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.\*:.*\:.+

Required: No

status

The status of the dataset update.

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Data Source
Service: Amazon Personalize

Describes the data source that contains the data to upload to a dataset.

Contents

dataLocation

The path to the Amazon S3 bucket where the data that you want to upload to your dataset is stored. For example:

s3://bucket-name/folder-name/

Type: String

Length Constraints: Maximum length of 256.

Pattern: (s3|http|https):/.+

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
DefaultCategoricalHyperParameterRange
Service: Amazon Personalize

Provides the name and default range of a categorical hyperparameter and whether the hyperparameter is tunable. A tunable hyperparameter can have its value determined during hyperparameter optimization (HPO).

Contents

isTunable
Whether the hyperparameter is tunable.
Type: Boolean
Required: No

name
The name of the hyperparameter.
Type: String
Length Constraints: Maximum length of 256.
Required: No

values
A list of the categories for the hyperparameter.
Type: Array of strings
Array Members: Maximum number of 100 items.
Length Constraints: Maximum length of 1000.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
DefaultContinuousHyperParameterRange
Service: Amazon Personalize

Provides the name and default range of a continuous hyperparameter and whether the hyperparameter is tunable. A tunable hyperparameter can have its value determined during hyperparameter optimization (HPO).

Contents

**isTunable**

Whether the hyperparameter is tunable.

Type: Boolean

Required: No

**maxValue**

The maximum allowable value for the hyperparameter.

Type: Double

Valid Range: Minimum value of -1000000.

Required: No

**minValue**

The minimum allowable value for the hyperparameter.

Type: Double

Valid Range: Minimum value of -1000000.

Required: No

**name**

The name of the hyperparameter.

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
**DefaultHyperParameterRanges**

Service: Amazon Personalize

Specifies the hyperparameters and their default ranges. Hyperparameters can be categorical, continuous, or integer-valued.

**Contents**

**categoricalHyperParameterRanges**

The categorical hyperparameters and their default ranges.

Type: Array of [DefaultCategoricalHyperParameterRange](#) objects

Array Members: Maximum number of 100 items.

Required: No

**continuousHyperParameterRanges**

The continuous hyperparameters and their default ranges.

Type: Array of [DefaultContinuousHyperParameterRange](#) objects

Array Members: Maximum number of 100 items.

Required: No

**integerHyperParameterRanges**

The integer-valued hyperparameters and their default ranges.

Type: Array of [DefaultIntegerHyperParameterRange](#) objects

Array Members: Maximum number of 100 items.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
DefaultIntegerHyperParameterRange
Service: Amazon Personalize

Provides the name and default range of a integer-valued hyperparameter and whether the hyperparameter is tunable. A tunable hyperparameter can have its value determined during hyperparameter optimization (HPO).

Contents

isTunable
Indicates whether the hyperparameter is tunable.
Type: Boolean
Required: No

maxValue
The maximum allowable value for the hyperparameter.
Type: Integer
Valid Range: Maximum value of 1000000.
Required: No

minValue
The minimum allowable value for the hyperparameter.
Type: Integer
Valid Range: Minimum value of -1000000.
Required: No

name
The name of the hyperparameter.
Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
EventTracker
Service: Amazon Personalize

Provides information about an event tracker.

Contents

**accountId**

The AWS account that owns the event tracker.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**creationDateTime**

The date and time (in Unix format) that the event tracker was created.

Type: Timestamp

Required: No

**datasetGroupArn**

The Amazon Resource Name (ARN) of the dataset group that receives the event data.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*:.*

Required: No

**eventTrackerArn**

The ARN of the event tracker.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*

Required: No

**lastUpdatedDateTime**

The date and time (in Unix time) that the event tracker was last updated.

Type: Timestamp

Required: No

**name**

The name of the event tracker.

Type: String

Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_]*$
Required: No

**status**

The status of the event tracker.

An event tracker can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

**trackingId**

The ID of the event tracker. Include this ID in requests to the `PutEvents` API.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
**EventTrackerSummary**  
Service: Amazon Personalize

Provides a summary of the properties of an event tracker. For a complete listing, call the [DescribeEventTracker](#) API.

**Contents**

- **creationDateTime**
  - The date and time (in Unix time) that the event tracker was created.
  - Type: Timestamp
  - Required: No

- **eventTrackerArn**
  - The Amazon Resource Name (ARN) of the event tracker.
  - Type: String
  - Length Constraints: Maximum length of 256.
  - Pattern: `arn:([a-z\d\-]+):personalize:.+:.*:*:*`  
  - Required: No

- **lastUpdatedDateTime**
  - The date and time (in Unix time) that the event tracker was last updated.
  - Type: Timestamp
  - Required: No

- **name**
  - The name of the event tracker.
  - Type: String
  - Pattern: `^[a-zA-Z0-9][a-zA-Z0-9-\_]*`  
  - Required: No

- **status**
  - The status of the event tracker.
  - An event tracker can be in one of the following states:
    - CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
    - DELETE PENDING > DELETE IN_PROGRESS
  - Type: String
  - Length Constraints: Maximum length of 256.
  - Required: No
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
FeatureTransformation
Service: Amazon Personalize

Provides feature transformation information. Feature transformation is the process of modifying raw input data into a form more suitable for model training.

Contents

creationDateTime
The creation date and time (in Unix time) of the feature transformation.
Type: Timestamp
Required: No
defaultParameters
Provides the default parameters for feature transformation.
Type: String to string map
Map Entries: Maximum number of 100 items.
Key Length Constraints: Maximum length of 256.
Value Length Constraints: Maximum length of 1000.
Required: No
featureTransformationArn
The Amazon Resource Name (ARN) of the FeatureTransformation object.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-zA-Z0-9-]+):personalize:.+*::.*
Required: No
lastUpdatedDateTime
The last update date and time (in Unix time) of the feature transformation.
Type: Timestamp
Required: No
name
The name of the feature transformation.
Type: String
Pattern: ^[a-zA-Z0-9-9][a-zA-Z0-9-9 \-_]*$
Required: No
status
The status of the feature transformation.
A feature transformation can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Filter
Service: Amazon Personalize

Contains information on a recommendation filter, including its ARN, status, and filter expression.

Contents

creationDateTime
The time at which the filter was created.
Type: Timestamp
Required: No

datasetGroupArn
The ARN of the dataset group to which the filter belongs.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:+
Required: No

failureReason
If the filter failed, the reason for its failure.
Type: String
Required: No

filterArn
The ARN of the filter.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:+
Required: No

filterExpression
Specifies the type of item interactions to filter out of recommendation results. The filter expression must follow specific format rules. For information about filter expression structure and syntax, see Filter expressions.
Type: String
Required: No

lastUpdatedDateTime
The time at which the filter was last updated.
Type: Timestamp
Required: No

**name**

The name of the filter.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9-\_]*$

Required: No

**status**

The status of the filter.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
FilterSummary
Service: Amazon Personalize

A short summary of a filter's attributes.

Contents

creationDateTime
  The time at which the filter was created.
  Type: Timestamp
  Required: No

datasetGroupArn
  The ARN of the dataset group to which the filter belongs.
  Type: String
  Length Constraints: Maximum length of 256.
  Pattern: arn:[a-z\d-]+:personalize::.*
  Required: No

failureReason
  If the filter failed, the reason for the failure.
  Type: String
  Required: No

filterArn
  The ARN of the filter.
  Type: String
  Length Constraints: Maximum length of 256.
  Pattern: arn:[a-z\d-]+:personalize::.*
  Required: No

lastUpdatedDateTime
  The time at which the filter was last updated.
  Type: Timestamp
  Required: No

name
  The name of the filter.
  Type: String
  Pattern: ^[a-zA-Z0-9-9][a-zA-Z0-9-9\-_]*
Required: No

**status**

The status of the filter.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
HPOConfig
Service: Amazon Personalize

Describes the properties for hyperparameter optimization (HPO).

Contents

algorithmHyperParameterRanges

The hyperparameters and their allowable ranges.

Type: HyperParameterRanges (p. 688) object

Required: No

hpoObjective

The metric to optimize during HPO.

Note
Amazon Personalize doesn't support configuring the hpoObjective at this time.

Type: HPOObjective (p. 686) object

Required: No

hpoResourceConfig

Describes the resource configuration for HPO.

Type: HPOResourceConfig (p. 687) object

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**HPOObjective**

Service: Amazon Personalize

The metric to optimize during hyperparameter optimization (HPO).

**Note**
Amazon Personalize doesn't support configuring the hpoObjective at this time.

**Contents**

**metricName**

The name of the metric.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**metricRegex**

A regular expression for finding the metric in the training job logs.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**type**

The type of the metric. Valid values are Maximize and Minimize.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
HPOResourceConfig
Service: Amazon Personalize

Describes the resource configuration for hyperparameter optimization (HPO).

Contents

**maxNumberOfTrainingJobs**

The maximum number of training jobs when you create a solution version. The maximum value for `maxNumberOfTrainingJobs` is 40.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**maxParallelTrainingJobs**

The maximum number of parallel training jobs when you create a solution version. The maximum value for `maxParallelTrainingJobs` is 10.

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
HyperParameterRanges
Service: Amazon Personalize

Specifies the hyperparameters and their ranges. Hyperparameters can be categorical, continuous, or integer-valued.

Contents

categoricalHyperParameterRanges
The categorical hyperparameters and their ranges.
Type: Array of CategoricalHyperParameterRange (p. 643) objects
Array Members: Maximum number of 100 items.
Required: No

continuousHyperParameterRanges
The continuous hyperparameters and their ranges.
Type: Array of ContinuousHyperParameterRange (p. 644) objects
Array Members: Maximum number of 100 items.
Required: No

integerHyperParameterRanges
The integer-valued hyperparameters and their ranges.
Type: Array of IntegerHyperParameterRange (p. 689) objects
Array Members: Maximum number of 100 items.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**IntegerHyperParameterRange**  
Service: Amazon Personalize  

Provides the name and range of an integer-valued hyperparameter.

**Contents**

**maxValue**

The maximum allowable value for the hyperparameter.

Type: Integer

Valid Range: Maximum value of 1000000.

Required: No

**minValue**

The minimum allowable value for the hyperparameter.

Type: Integer

Valid Range: Minimum value of -1000000.

Required: No

**name**

The name of the hyperparameter.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**MetricAttribute**
Service: Amazon Personalize

Contains information on a metric that a metric attribution reports on. For more information, see [Measuring impact of recommendations](#).

**Contents**

**eventType**

The metric's event type.

Type: String

Length Constraints: Maximum length of 256.

Required: Yes

**expression**

The attribute's expression. Available functions are `SUM()` or `SAMPLECOUNT()`. For `SUM()` functions, provide the dataset type (either Interactions or Items) and column to sum as a parameter. For example `SUM(Items.PRICE)`.

Type: String

Length Constraints: Maximum length of 256.

Required: Yes

**metricName**

The metric's name. The name helps you identify the metric in Amazon CloudWatch or Amazon S3.

Type: String

Length Constraints: Maximum length of 256.

Required: Yes

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
MetricAttribution
Service: Amazon Personalize

Contains information on a metric attribution. A metric attribution creates reports on the data that you import into Amazon Personalize. Depending on how you import the data, you can view reports in Amazon CloudWatch or Amazon S3. For more information, see Measuring impact of recommendations.

Contents

creationDateTime

The metric attribution's creation date time.
Type: Timestamp
Required: No

datasetGroupArn

The metric attribution's dataset group Amazon Resource Name (ARN).
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*+:+:
Required: No

failureReason

The metric attribution's failure reason.
Type: String
Required: No

lastUpdatedDateTime

The metric attribution's last updated date time.
Type: Timestamp
Required: No

metricAttributionArn

The metric attribution's Amazon Resource Name (ARN).
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*+:+:
Required: No

metricsOutputConfig

The metric attribution's output configuration.
Type: MetricAttributionOutput (p. 693) object
Required: No
name

The metric attribution's name.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_.]*

Required: No

status

The metric attribution's status.

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
MetricAttributionOutput
Service: Amazon Personalize

The output configuration details for a metric attribution.

Contents

roleArn

The Amazon Resource Name (ARN) of the IAM service role that has permissions to add data to your output Amazon S3 bucket and add metrics to Amazon CloudWatch. For more information, see Measuring impact of recommendations.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):iam::\d{12}:role/?[a-zA-Z_0-9+=,.@\-_/]+

Required: Yes

s3DataDestination

The configuration details of an Amazon S3 input or output bucket.

Type: S3DataConfig (p. 709) object

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
**MetricAttributionSummary**

Service: Amazon Personalize

Provides a summary of the properties of a metric attribution. For a complete listing, call the `DescribeMetricAttribution`.

**Contents**

- **creationDateTime**
  - The metric attribution's creation date time.
  - Type: Timestamp
  - Required: No

- **failureReason**
  - The metric attribution's failure reason.
  - Type: String
  - Required: No

- **lastUpdatedDateTime**
  - The metric attribution's last updated date time.
  - Type: Timestamp
  - Required: No

- **metricAttributionArn**
  - The metric attribution's Amazon Resource Name (ARN).
  - Type: String
  - Length Constraints: Maximum length of 256.
  - Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.+`
  - Required: No

- **name**
  - The name of the metric attribution.
  - Type: String
  - Pattern: `^[a-zA-Z0-9-\[\]_\-]*$`
  - Required: No

- **status**
  - The metric attribution's status.
  - Type: String
  - Length Constraints: Maximum length of 256.
  - Required: No
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
OptimizationObjective
Service: Amazon Personalize

Describes the additional objective for the solution, such as maximizing streaming minutes or increasing revenue. For more information see Optimizing a solution.

Contents

itemAttribute

The numerical metadata column in an Items dataset related to the optimization objective. For example, VIDEO_LENGTH (to maximize streaming minutes), or PRICE (to maximize revenue).

Type: String

Length Constraints: Minimum length of 1. Maximum length of 150.

Required: No

objectiveSensitivity

Specifies how Amazon Personalize balances the importance of your optimization objective versus relevance.

Type: String

Valid Values: LOW | MEDIUM | HIGH | OFF

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Recipe
Service: Amazon Personalize

Provides information about a recipe. Each recipe provides an algorithm that Amazon Personalize uses in model training when you use the CreateSolution operation.

Contents

algorithmArn
The Amazon Resource Name (ARN) of the algorithm that Amazon Personalize uses to train the model.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d\-]+):personalize:.+:*:+:+
Required: No

creationDateTime
The date and time (in Unix format) that the recipe was created.
Type: Timestamp
Required: No

description
The description of the recipe.
Type: String
Required: No

featureTransformationArn
The ARN of the FeatureTransformation object.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d\-]+):personalize:.+:*:+:+
Required: No

lastUpdatedDateTime
The date and time (in Unix format) that the recipe was last updated.
Type: Timestamp
Required: No

name
The name of the recipe.
Type: String
Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-\_]*$
Required: No

**recipeArn**

The Amazon Resource Name (ARN) of the recipe.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.+:.+
Required: No

**recipeType**

One of the following values:
- PERSONALIZED_RANKING
- RELATED_ITEMS
- USER_PERSONALIZATION

Type: String
Length Constraints: Maximum length of 256.
Required: No

**status**

The status of the recipe.

Type: String
Length Constraints: Maximum length of 256.
Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
RecipeSummary
Service: Amazon Personalize

Provides a summary of the properties of a recipe. For a complete listing, call the DescribeRecipe API.

Contents

creationDateTime
The date and time (in Unix time) that the recipe was created.
Type: Timestamp
Required: No

domain
The domain of the recipe (if the recipe is a Domain dataset group use case).
Type: String
Valid Values: ECOMMERCE | VIDEO_ON_DEMAND
Required: No

lastUpdatedDateTime
The date and time (in Unix time) that the recipe was last updated.
Type: Timestamp
Required: No

name
The name of the recipe.
Type: String
Pattern: ^[a-zA-Z0-9\-\_]\[a-zA-Z0-9\-\_]\*
Required: No

recipeArn
The Amazon Resource Name (ARN) of the recipe.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize::.*::.*
Required: No

status
The status of the recipe.
Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
Recommender

Service: Amazon Personalize

Describes a recommendation generator for a Domain dataset group. You create a recommender in a Domain dataset group for a specific domain use case (domain recipe), and specify the recommender in a `GetRecommendations` request.

Contents

creationDateTime

The date and time (in Unix format) that the recommender was created.

Type: Timestamp

Required: No

datasetGroupArn

The Amazon Resource Name (ARN) of the Domain dataset group that contains the recommender.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.*:.*:.+

Required: No

failureReason

If a recommender fails, the reason behind the failure.

Type: String

Required: No

lastUpdatedDateTime

The date and time (in Unix format) that the recommender was last updated.

Type: Timestamp

Required: No

latestRecommenderUpdate

Provides a summary of the latest updates to the recommender.

Type: `RecommenderUpdateSummary (p. 707)` object

Required: No

modelMetrics

Provides evaluation metrics that help you determine the performance of a recommender. For more information, see [Evaluating a recommender](#).

Type: String to double map

Map Entries: Maximum number of 100 items.

Key Length Constraints: Maximum length of 256.

Required: No
name
The name of the recommender.
Type: String
Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-\_]*
Required: No

recipeArn
The Amazon Resource Name (ARN) of the recipe (Domain dataset group use case) that the recommender was created for.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*+:.*+
Required: No

recommenderArn
The Amazon Resource Name (ARN) of the recommender.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*+:.*+
Required: No

recommenderConfig
The configuration details of the recommender.
Type: recommenderConfig (p. 704) object
Required: No

status
The status of the recommender.
A recommender can be in one of the following states:
• CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
• STOP PENDING > STOP IN_PROGRESS > INACTIVE > START PENDING > START IN_PROGRESS > ACTIVE
• DELETE PENDING > DELETE IN_PROGRESS
Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:
- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
RecommenderConfig
Service: Amazon Personalize

The configuration details of the recommender.

Contents

itemExplorationConfig

Specifies the exploration configuration hyperparameters, including explorationWeight and explorationItemAgeCutOff, you want to use to configure the amount of item exploration Amazon Personalize uses when recommending items. Provide itemExplorationConfig data only if your recommenders generate personalized recommendations for a user (not popular items or similar items).

Type: String to string map

Map Entries: Maximum number of 100 items.

Key Length Constraints: Maximum length of 256.

Value Length Constraints: Maximum length of 1000.

Required: No

minRecommendationRequestsPerSecond

Specifies the requested minimum provisioned recommendation requests per second that Amazon Personalize will support. A high minRecommendationRequestsPerSecond will increase your bill. We recommend starting with 1 for minRecommendationRequestsPerSecond (the default). Track your usage using Amazon CloudWatch metrics, and increase the minRecommendationRequestsPerSecond as necessary.

Type: Integer

Valid Range: Minimum value of 1.

Required: No

trainingDataConfig

Specifies the training data configuration to use when creating a domain recommender.

Type: TrainingDataConfig (p. 724) object

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
RecommenderSummary
Service: Amazon Personalize
Provides a summary of the properties of the recommender.

Contents

creationDateTime
The date and time (in Unix format) that the recommender was created.
Type: Timestamp
Required: No
datasetGroupArn
The Amazon Resource Name (ARN) of the Domain dataset group that contains the recommender.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d\-]+):personalize:.+:.*:.*
Required: No
lastUpdatedDateTime
The date and time (in Unix format) that the recommender was last updated.
Type: Timestamp
Required: No
name
The name of the recommender.
Type: String
Pattern: ^[a-zA-Z0-9-][a-zA-Z0-9\-\_]*$
Required: No
recipeArn
The Amazon Resource Name (ARN) of the recipe (Domain dataset group use case) that the recommender was created for.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d\-]+):personalize:.+:.*:.*
Required: No
recommenderArn
The Amazon Resource Name (ARN) of the recommender.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-zA-Z0-9-]+):personalize:.+:.*:+

Required: No

**recommenderConfig**

The configuration details of the recommender.

Type: [RecommenderConfig](#) object

Required: No

**status**

The status of the recommender. A recommender can be in one of the following states:

- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE - or - CREATE FAILED
- STOP PENDING > STOP IN_PROGRESS > INACTIVE > START PENDING > START IN_PROGRESS > ACTIVE
- DELETE PENDING > DELETE IN_PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
RecommenderUpdateSummary
Service: Amazon Personalize

Provides a summary of the properties of a recommender update. For a complete listing, call the DescribeRecommender API.

Contents

creationDateTime
The date and time (in Unix format) that the recommender update was created.
Type: Timestamp
Required: No

failureReason
If a recommender update fails, the reason behind the failure.
Type: String
Required: No

lastUpdatedDateTime
The date and time (in Unix time) that the recommender update was last updated.
Type: Timestamp
Required: No

recommenderConfig
The configuration details of the recommender update.
Type: RecommenderConfig (p. 704) object
Required: No

status
The status of the recommender update.
A recommender can be in one of the following states:
• CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
• STOP PENDING > STOP IN_PROGRESS > INACTIVE > START PENDING > START IN_PROGRESS > ACTIVE
• DELETE PENDING > DELETE IN_PROGRESS
Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:
• AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
S3DataConfig
Service: Amazon Personalize

The configuration details of an Amazon S3 input or output bucket.

Contents

path

The file path of the Amazon S3 bucket.

Type: String

Length Constraints: Maximum length of 256.

Pattern: (s3|http|https)://.+

Required: Yes

kmsKeyArn

The Amazon Resource Name (ARN) of the AWS Key Management Service (KMS) key that Amazon Personalize uses to encrypt or decrypt the input and output files.

Type: String

Length Constraints: Maximum length of 2048.

Pattern: arn:aws::*:kms::*:[0-9][12]:key/.*

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
An object that provides information about a solution. A solution is a trained model that can be deployed as a campaign.

Contents

**autoMLResult**
When `performAutoML` is true, specifies the best recipe found.

Type: `AutoMLResult` (p. 620) object

Required: No

**creationDateTime**
The creation date and time (in Unix time) of the solution.

Type: Timestamp

Required: No

**datasetGroupArn**
The Amazon Resource Name (ARN) of the dataset group that provides the training data.

Type: String

Length Constraints: Maximum length of 256.

Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.+`

Required: No

**eventType**
The event type (for example, 'click' or 'like') that is used for training the model. If no eventType is provided, Amazon Personalize uses all interactions for training with equal weight regardless of type.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**lastUpdatedDateTime**
The date and time (in Unix time) that the solution was last updated.

Type: Timestamp

Required: No

**latestSolutionVersion**
Describes the latest version of the solution, including the status and the ARN.

Type: `SolutionVersionSummary` (p. 721) object

Required: No

**name**
The name of the solution.
performAutoML

**Important**
We don't recommend enabling automated machine learning. Instead, match your use case to the available Amazon Personalize recipes. For more information, see Determining your use case.

When true, Amazon Personalize performs a search for the best USER_PERSONALIZATION recipe from the list specified in the solution configuration (recipeArn must not be specified). When false (the default), Amazon Personalize uses recipeArn for training.

Type: Boolean
Required: No

performHPO

Whether to perform hyperparameter optimization (HPO) on the chosen recipe. The default is false.

Type: Boolean
Required: No

recipeArn

The ARN of the recipe used to create the solution. This is required when performAutoML is false.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:.*+`
Required: No

solutionArn

The ARN of the solution.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:.*+`
Required: No

solutionConfig

Describes the configuration properties for the solution.

Type: SolutionConfig (p. 713) object
Required: No

status

The status of the solution.
A solution can be in one of the following states:
- CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
- DELETE PENDING > DELETE IN_PROGRESS

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
SolutionConfig
Service: Amazon Personalize

Describes the configuration properties for the solution.

Contents

algorithmHyperParameters

Lists the algorithm hyperparameters and their values.
Type: String to string map
Map Entries: Maximum number of 100 items.
Key Length Constraints: Maximum length of 256.
Value Length Constraints: Maximum length of 1000.
Required: No

autoMLConfig

The AutoMLConfig object containing a list of recipes to search when AutoML is performed.
Type: AutoMLConfig (p. 619) object
Required: No

eventValueThreshold

Only events with a value greater than or equal to this threshold are used for training a model.
Type: String
Length Constraints: Maximum length of 256.
Required: No

featureTransformationParameters

Lists the feature transformation parameters.
Type: String to string map
Map Entries: Maximum number of 100 items.
Key Length Constraints: Maximum length of 256.
Value Length Constraints: Maximum length of 1000.
Required: No

hpoConfig

Describes the properties for hyperparameter optimization (HPO).
Type: HPOConfig (p. 685) object
Required: No

optimizationObjective

Describes the additional objective for the solution, such as maximizing streaming minutes or increasing revenue. For more information see Optimizing a solution.
Type: OptimizationObjective (p. 696) object

Required: No

trainingDataConfig

Specifies the training data configuration to use when creating a custom solution version (trained model).

Type: TrainingDataConfig (p. 724) object

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
SolutionSummary
Service: Amazon Personalize

Provides a summary of the properties of a solution. For a complete listing, call the DescribeSolution API.

Contents

creationDateTime
The date and time (in Unix time) that the solution was created.
Type: Timestamp
Required: No

lastUpdatedDateTime
The date and time (in Unix time) that the solution was last updated.
Type: Timestamp
Required: No

name
The name of the solution.
Type: String
Pattern: ^[a-zA-Z0-9\-\_]*$
Required: No

recipeArn
The Amazon Resource Name (ARN) of the recipe used by the solution.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:[a-zA-Z0-9-]+:personalize:.+:.*:+
Required: No

solutionArn
The Amazon Resource Name (ARN) of the solution.
Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:[a-zA-Z0-9-]+:personalize:.+:.*:+
Required: No

status
The status of the solution.
A solution can be in one of the following states:
• CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED
• DELETE PENDING > DELETE IN_PROGRESS

Type: String
Length Constraints: Maximum length of 256.
Required: No

See Also
For more information about using this API in one of the language-specific AWS SDKs, see the following:

• AWS SDK for C++
• AWS SDK for Go
• AWS SDK for Java V2
• AWS SDK for Ruby V3
**SolutionVersion**  
Service: Amazon Personalize  

An object that provides information about a specific version of a Solution in a Custom dataset group.

**Contents**

creationDateTime  
The date and time (in Unix time) that this version of the solution was created.  
Type: Timestamp  
Required: No  
datasetGroupArn  
The Amazon Resource Name (ARN) of the dataset group providing the training data.  
Type: String  
Length Constraints: Maximum length of 256.  
Pattern: `arn:([a-z\d-]+):personalize:.*:.*:.+`  
Required: No  
eventType  
The event type (for example, 'click' or 'like') that is used for training the model.  
Type: String  
Length Constraints: Maximum length of 256.  
Required: No  
failureReason  
If training a solution version fails, the reason for the failure.  
Type: String  
Required: No  
lastUpdatedDateTime  
The date and time (in Unix time) that the solution was last updated.  
Type: Timestamp  
Required: No  
name  
The name of the solution version.  
Type: String  
Pattern: `^[a-zA-Z0-9][a-zA-Z0-9\-_.]*$`  
Required: No
**performAutoML**
When true, Amazon Personalize searches for the most optimal recipe according to the solution configuration. When false (the default), Amazon Personalize uses `recipeArn`.

Type: Boolean
Required: No

**performHPO**
Whether to perform hyperparameter optimization (HPO) on the chosen recipe. The default is false.

Type: Boolean
Required: No

**recipeArn**
The ARN of the recipe used in the solution.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:.*::*.*` 
Required: No

**solutionArn**
The ARN of the solution.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:.*::*.*` 
Required: No

**solutionConfig**
Describes the configuration properties for the solution.

Type: `SolutionConfig (p. 713)` object
Required: No

**solutionVersionArn**
The ARN of the solution version.

Type: String
Length Constraints: Maximum length of 256.
Pattern: `arn:([a-z\d-]+):personalize:.+:.*::*.*` 
Required: No

**status**
The status of the solution version.

A solution version can be in one of the following states:
CREATE PENDING
CREATE IN_PROGRESS
ACTIVE
CREATE FAILED
CREATE STOPPING
CREATE STOPPED

Type: String
Length Constraints: Maximum length of 256.
Required: No

trainingHours
The time used to train the model. You are billed for the time it takes to train a model. This field is visible only after Amazon Personalize successfully trains a model.

Type: Double
Valid Range: Minimum value of 0.
Required: No

trainingMode
The scope of training to be performed when creating the solution version. The FULL option trains the solution version based on the entirety of the input solution's training data, while the UPDATE option processes only the data that has changed in comparison to the input solution. Choose UPDATE when you want to incrementally update your solution version instead of creating an entirely new one.

Important
The UPDATE option can only be used when you already have an active solution version created from the input solution using the FULL option and the input solution was trained with the User-Personalization recipe or the HRNN-Coldstart recipe.

Type: String
Valid Values: FULL | UPDATE
Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
SolutionVersionSummary

Service: Amazon Personalize

Provides a summary of the properties of a solution version. For a complete listing, call the DescribeSolutionVersion API.

Contents

creationDateTime

The date and time (in Unix time) that this version of a solution was created.

Type: Timestamp

Required: No

failureReason

If a solution version fails, the reason behind the failure.

Type: String

Required: No

lastUpdatedDateTime

The date and time (in Unix time) that the solution version was last updated.

Type: Timestamp

Required: No

solutionVersionArn

The Amazon Resource Name (ARN) of the solution version.

Type: String

Length Constraints: Maximum length of 256.

Pattern: arn:([a-z\d-]+):personalize:.+:.*

Required: No

status

The status of the solution version.

A solution version can be in one of the following states:
  • CREATE PENDING > CREATE IN_PROGRESS > ACTIVE -or- CREATE FAILED

Type: String

Length Constraints: Maximum length of 256.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

• AWS SDK for C++
• **AWS SDK for Go**
• **AWS SDK for Java V2**
• **AWS SDK for Ruby V3**
Tag
Service: Amazon Personalize

The optional metadata that you apply to resources to help you categorize and organize them. Each tag consists of a key and an optional value, both of which you define. For more information see Tagging Amazon Personalize resources.

Contents

tagKey

One part of a key-value pair that makes up a tag. A key is a general label that acts like a category for more specific tag values.

Type: String


Pattern: ^([\p{L}\p{Z}\p{N}_.:/=+-@]*$)

Required: Yes

tagValue

The optional part of a key-value pair that makes up a tag. A value acts as a descriptor within a tag category (key).

Type: String

Length Constraints: Minimum length of 0. Maximum length of 256.

Pattern: ^([\p{L}\p{Z}\p{N}_.:/=+-@]*$)

Required: Yes

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
TrainingDataConfig
Service: Amazon Personalize

The training data configuration to use when creating a domain recommender or custom solution version (trained model).

Contents

excludedDatasetColumns

Specifies the columns to exclude from training. Each key is a dataset type, and each value is a list of columns. Exclude columns to control what data Amazon Personalize uses to generate recommendations. For example, you might have a column that you want to use only to filter recommendations. You can exclude this column from training and Amazon Personalize considers it only when filtering.

Type: String to array of strings map

Map Entries: Maximum number of 3 items.

Key Length Constraints: Maximum length of 256.

Array Members: Maximum number of 50 items.

Length Constraints: Maximum length of 150.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
TunedHPOParams
Service: Amazon Personalize

If hyperparameter optimization (HPO) was performed, contains the hyperparameter values of the best performing model.

Contents

algorithmHyperParameters

A list of the hyperparameter values of the best performing model.

Type: String to string map

Map Entries: Maximum number of 100 items.

Key Length Constraints: Maximum length of 256.

Value Length Constraints: Maximum length of 1000.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3

Amazon Personalize Events

The following data types are supported by Amazon Personalize Events:

- Event (p. 726)
- Item (p. 729)
- MetricAttribution (p. 730)
- User (p. 731)
**Event**
Service: Amazon Personalize Events

Represents user interaction event information sent using the PutEvents API.

**Contents**

**eventType**

The type of event, such as click or download. This property corresponds to the EVENT_TYPE field of your Interactions schema and depends on the types of events you are tracking.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: Yes

**sentAt**

The timestamp (in Unix time) on the client side when the event occurred.

Type: Timestamp

Required: Yes

**eventId**

An ID associated with the event. If an event ID is not provided, Amazon Personalize generates a unique ID for the event. An event ID is not used as an input to the model. Amazon Personalize uses the event ID to distinguish unique events. Any subsequent events after the first with the same event ID are not used in model training.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: No

**eventValue**

The event value that corresponds to the EVENT_VALUE field of the Interactions schema.

Type: Float

Required: No

**impression**

A list of item IDs that represents the sequence of items you have shown the user. For example, ["itemId1", "itemId2", "itemId3"]. Provide a list of items to manually record impressions data for an event. For more information on recording impressions data, see [Recording impressions data](#).

Type: Array of strings

Array Members: Minimum number of 1 item. Maximum number of 25 items.

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: No
itemId

The item ID key that corresponds to the ITEM_ID field of the Interactions schema.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: No

metricAttribution

Contains information about the metric attribution associated with an event. For more information about metric attributions, see Measuring impact of recommendations.

Type: MetricAttribution (p. 730) object

Required: No

properties

A string map of event-specific data that you might choose to record. For example, if a user rates a movie on your site, other than movie ID (itemId) and rating (eventValue), you might also send the number of movie ratings made by the user.

Each item in the map consists of a key-value pair. For example,

{"numberOfRatings": "12"}

The keys use camel case names that match the fields in the Interactions schema. In the above example, the numberOfRatings would match the 'NUMBER_OF_RATINGS' field defined in the Interactions schema.

The following can't be included as a keyword for properties (case insensitive).

- userId
- sessionId
- eventType
- timestamp
- recommendationId
- impression

Type: String


Required: No

recommendationId

The ID of the list of recommendations that contains the item the user interacted with. Provide a recommendationId to have Amazon Personalize implicitly record the recommendations you show your user as impressions data. Or provide a recommendationId if you use a metric attribution to measure the impact of recommendations.

For more information on recording impressions data, see Recording impressions data. For more information on creating a metric attribution see Measuring impact of recommendations.

Type: String


Required: No
See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
Item
Service: Amazon Personalize Events

Represents item metadata added to an Items dataset using the PutItems API. For more information see Importing Items Incrementally.

Contents

itemId

The ID associated with the item.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: Yes

properties

A string map of item-specific metadata. Each element in the map consists of a key-value pair. For example, {"numberOfRatings": "12"}.

The keys use camel case names that match the fields in the schema for the Items dataset. In the previous example, the numberOfRatings matches the 'NUMBER_OF_RATINGS' field defined in the Items schema. For categorical string data, to include multiple categories for a single item, separate each category with a pipe separator (|). For example, \"Horror|Action\".

Type: String


Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
MetricAttribution
Service: Amazon Personalize Events

Contains information about a metric attribution associated with an event. For more information about metric attributions, see Measuring impact of recommendations.

Contents

eventAttributionSource

The source of the event, such as a third party.

Type: String

Length Constraints: Maximum length of 1024.

Pattern: ^\[\x20-\x7E]*\[\x21-\x7E]+\[\x20-\x7E]+$ 

Required: Yes

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3
User
Service: Amazon Personalize Events

Represents user metadata added to a Users dataset using the PutUser API. For more information see Importing Users Incrementally.

Contents

**userId**

The ID associated with the user.

Type: String

Length Constraints: Minimum length of 1. Maximum length of 256.

Required: Yes

**properties**

A string map of user-specific metadata. Each element in the map consists of a key-value pair. For example, \{"numberOfVideosWatched": "45\}.

The keys use camel case names that match the fields in the schema for the Users dataset. In the previous example, the `numberOfVideosWatched` matches the `NUMBER_OF_VIDEOS_WATCHED` field defined in the Users schema. For categorical string data, to include multiple categories for a single user, separate each category with a pipe separator (`|`). For example, \"Member|Frequent shopper\".

Type: String


Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3

Amazon Personalize Runtime

The following data types are supported by Amazon Personalize Runtime:

- PredictedItem (p. 732)
- Promotion (p. 733)
**PredictedItem**  
Service: Amazon Personalize Runtime

An object that identifies an item. The GetRecommendations (p. 608) and GetPersonalizedRanking (p. 604) APIs return a list of PredictedItems.

**Contents**

**itemId**

The recommended item ID.

Type: String

Length Constraints: Maximum length of 256.

Required: No

**promotionName**

The name of the promotion that included the predicted item.

Type: String


Pattern: ^[a-zA-Z0-9][a-zA-Z0-9\-_]*

Required: No

**score**

A numeric representation of the model's certainty that the item will be the next user selection. For more information on scoring logic, see Recommendation scores (p. 244).

Type: Double

Required: No

**See Also**

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- [AWS SDK for C++](#)
- [AWS SDK for Go](#)
- [AWS SDK for Java V2](#)
- [AWS SDK for Ruby V3](#)
Promotion
Service: Amazon Personalize Runtime

Contains information on a promotion. A promotion defines additional business rules that apply to a configurable subset of recommended items.

Contents

filterArn

The Amazon Resource Name (ARN) of the filter used by the promotion. This filter defines the criteria for promoted items. For more information, see Promotion filters.

Type: String
Length Constraints: Maximum length of 256.
Pattern: arn:([a-z\d-]+):personalize:.+:.*:.+
Required: No

filterValues

The values to use when promoting items. For each placeholder parameter in your promotion's filter expression, provide the parameter name (in matching case) as a key and the filter value(s) as the corresponding value. Separate multiple values for one parameter with a comma.

For filter expressions that use an INCLUDE element to include items, you must provide values for all parameters that are defined in the expression. For filters with expressions that use an EXCLUDE element to exclude items, you can omit the filter-values. In this case, Amazon Personalize doesn't use that portion of the expression to filter recommendations.

For more information on creating filters, see Filtering recommendations and user segments.

Type: String to string map
Map Entries: Maximum number of 25 items.
Key Length Constraints: Maximum length of 50.
Key Pattern: [A-Za-z\-0-9_]+
Value Length Constraints: Maximum length of 1000.
Required: No

name

The name of the promotion.

Type: String
Pattern: ^[a-zA-Z\-0-9][a-zA-Z\-0-9\-_]*$*
Required: No

percentPromotedItems

The percentage of recommended items to apply the promotion to.
Type: Integer

Valid Range: Minimum value of 1. Maximum value of 100.

Required: No

See Also

For more information about using this API in one of the language-specific AWS SDKs, see the following:

- AWS SDK for C++
- AWS SDK for Go
- AWS SDK for Java V2
- AWS SDK for Ruby V3

Common Errors

This section lists the errors common to the API actions of all AWS services. For errors specific to an API action for this service, see the topic for that API action.

AccessDeniedException

You do not have sufficient access to perform this action.

HTTP Status Code: 400

IncompleteSignature

The request signature does not conform to AWS standards.

HTTP Status Code: 400

InternalFailure

The request processing has failed because of an unknown error, exception or failure.

HTTP Status Code: 500

InvalidAction

The action or operation requested is invalid. Verify that the action is typed correctly.

HTTP Status Code: 400

InvalidClientTokenId

The X.509 certificate or AWS access key ID provided does not exist in our records.

HTTP Status Code: 403

NotAuthorized

You do not have permission to perform this action.

HTTP Status Code: 400

OptInRequired

The AWS access key ID needs a subscription for the service.
HTTP Status Code: 403
**RequestExpired**

The request reached the service more than 15 minutes after the date stamp on the request or more than 15 minutes after the request expiration date (such as for pre-signed URLs), or the date stamp on the request is more than 15 minutes in the future.

HTTP Status Code: 400
**ServiceUnavailable**

The request has failed due to a temporary failure of the server.

HTTP Status Code: 503
**ThrottlingException**

The request was denied due to request throttling.

HTTP Status Code: 400
**ValidationError**

The input fails to satisfy the constraints specified by an AWS service.

HTTP Status Code: 400

---

**Common Parameters**

The following list contains the parameters that all actions use for signing Signature Version 4 requests with a query string. Any action-specific parameters are listed in the topic for that action. For more information about Signature Version 4, see [Signing AWS API requests](#) in the IAM User Guide.

**Action**

The action to be performed.

Type: string

Required: Yes

**Version**

The API version that the request is written for, expressed in the format YYYY-MM-DD.

Type: string

Required: Yes

**X-Amz-Algorithm**

The hash algorithm that you used to create the request signature.

Condition: Specify this parameter when you include authentication information in a query string instead of in the HTTP authorization header.

Type: string

Valid Values: AWS4-HMAC-SHA256

Required: Conditional
X-Amz-Credential

The credential scope value, which is a string that includes your access key, the date, the region you are targeting, the service you are requesting, and a termination string ("aws4_request"). The value is expressed in the following format: access_key/YYYYMMDD/region/service/aws4_request.

For more information, see Create a signed AWS API request in the IAM User Guide.

Condition: Specify this parameter when you include authentication information in a query string instead of in the HTTP authorization header.

Type: string

Required: Conditional

X-Amz-Date

The date that is used to create the signature. The format must be ISO 8601 basic format (YYYYMMDD'T'HMMSS'Z'). For example, the following date time is a valid X-Amz-Date value: 20120325T120000Z.

Condition: X-Amz-Date is optional for all requests; it can be used to override the date used for signing requests. If the Date header is specified in the ISO 8601 basic format, X-Amz-Date is not required. When X-Amz-Date is used, it always overrides the value of the Date header. For more information, see Elements of an AWS API request signature in the IAM User Guide.

Type: string

Required: Conditional

X-Amz-Security-Token

The temporary security token that was obtained through a call to AWS Security Token Service (AWS STS). For a list of services that support temporary security credentials from AWS STS, see AWS services that work with IAM in the IAM User Guide.

Condition: If you're using temporary security credentials from AWS STS, you must include the security token.

Type: string

Required: Conditional

X-Amz-Signature

Specifies the hex-encoded signature that was calculated from the string to sign and the derived signing key.

Condition: Specify this parameter when you include authentication information in a query string instead of in the HTTP authorization header.

Type: string

Required: Conditional

X-Amz-SignedHeaders

Specifies all the HTTP headers that were included as part of the canonical request. For more information about specifying signed headers, see Create a signed AWS API request in the IAM User Guide.

Condition: Specify this parameter when you include authentication information in a query string instead of in the HTTP authorization header.
Type: string

Required: Conditional
Document history for Amazon Personalize

The following table describes important changes in each release of the *Amazon Personalize Developer Guide*. For notification about updates to this documentation, you can subscribe to an RSS feed.

<table>
<thead>
<tr>
<th>Change</th>
<th>Description</th>
<th>Date</th>
</tr>
</thead>
<tbody>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to personalize search results from OpenSearch Service. For more information, see <a href="#">Personalizing search results from OpenSearch</a>.</td>
<td>October 16, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to import up to 100 metadata columns for Items datasets (up from 50) and 25 metadata columns for Users datasets (up from 5). For more information about Amazon Personalize limits, see <a href="#">Amazon Personalize endpoints and quotas</a>.</td>
<td>September 5, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to personalize search results from OpenSearch. For more information, see <a href="#">Personalizing search results from OpenSearch (self-managed)</a>.</td>
<td>July 25, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to replace a dataset's schema with a new or existing one. For more information, see <a href="#">Replacing a dataset's schema</a>.</td>
<td>July 13, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>If you use the User-Personalization or Personalized-Ranking recipes, Amazon Personalize batch inference jobs can now use data that you import incrementally without retraining. For information, see <a href="#">Getting batch recommendations</a>.</td>
<td>June 30, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to filter items based on the item you...</td>
<td>June 21, 2023</td>
</tr>
<tr>
<td>Date</td>
<td>Feature Description</td>
<td>Release Date</td>
</tr>
<tr>
<td>------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
<td>--------------</td>
</tr>
<tr>
<td>June 21, 2023</td>
<td>Amazon Personalize now supports the ability to filter items based on the item you specify in your request for related items recommendations. For information about filters, see Filtering recommendations and user segments.</td>
<td></td>
</tr>
<tr>
<td>June 12, 2023</td>
<td>Amazon Personalize now supports private connections between a virtual private cloud (VPC) and Amazon Personalize with an interface Amazon VPC endpoint. For more information, see Amazon Personalize and interface VPC endpoints (AWS PrivateLink).</td>
<td></td>
</tr>
<tr>
<td>May 30, 2023</td>
<td>Amazon Personalize now supports configuring the columns used when training when you create a recommender or custom solution. For information about configuring columns when creating a recommender, see Creating recommenders. For information about configuring columns when creating a solution, see Configuring columns used when training.</td>
<td></td>
</tr>
<tr>
<td>May 5, 2023</td>
<td>The Amazon Personalize developer guide now includes information about performing A/B testing with Amazon Personalize recommendations. For more information, see Measuring recommendation impact with A/B testing.</td>
<td></td>
</tr>
<tr>
<td>April 21, 2023</td>
<td>Amazon Personalize now supports configuring how popularity influences recommendations generated by the Similar-Items recipe. For more information, see Similar Items recipe.</td>
<td></td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports using Amazon SageMaker Data Wrangler to import data from 40+ sources into Amazon Personalize datasets. For more information, see Importing data using Amazon SageMaker Data Wrangler.</td>
<td>April 14, 2023</td>
</tr>
<tr>
<td>---------------------</td>
<td>----------------------------------------------------------------------------------------------------------------------------------------------------</td>
<td>----------------</td>
</tr>
<tr>
<td>New documentation feature (p. 738)</td>
<td>The Amazon Personalize developer guide now includes a new readiness checklist. This checklist helps you prepare to use Amazon Personalize with your own data. For more information, see Readiness checklist.</td>
<td>February 9, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports generating insights and statistics for data that you import into datasets. For more information, see Analyzing data in datasets.</td>
<td>January 25, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports a new Trending-Now recipe for Custom dataset groups. For more information, see Trending-Now recipe.</td>
<td>January 6, 2023</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports using tags in IAM policies. For more information, see Using tags in IAM policies.</td>
<td>December 28, 2022</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports making adjustments to the Maximum number of interactions that are considered by a model during training quota. Additionally, the Maximum number of users that are considered by a model during training limit quota no longer applies. For more information, see Amazon Personalize endpoints and quotas.</td>
<td>December 15, 2022</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports creating a metric attribution to measure the business impact of recommendations. For more information, see Measuring impact of recommendations.</td>
<td>November 17, 2022</td>
</tr>
<tr>
<td>Date</td>
<td>New feature (p. 738)</td>
<td>Details</td>
</tr>
<tr>
<td>---------------------</td>
<td>------------------------------------------------------------------------------------------------------------------</td>
<td>--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>September 7, 2022</td>
<td>Amazon Personalize quotas for total number of active solutions, active campaigns, recommenders, and filters have now increased. Each of these quotas now apply per dataset group rather than per account. For information about quotas, see <a href="#">Amazon Personalize endpoints and quotas</a>.</td>
<td></td>
</tr>
<tr>
<td>August 29, 2022</td>
<td>Amazon Personalize filters now consider up to 100 interactions per user per event type. For information about filters, see <a href="#">Filtering recommendations and user segments</a>.</td>
<td></td>
</tr>
<tr>
<td>August 17, 2022</td>
<td>Amazon Personalize now supports a new Trending now use case for the VIDEO_ON_DEMAND domain. For more information, see <a href="#">VIDEO_ON_DEMAND use cases</a>.</td>
<td></td>
</tr>
<tr>
<td>August 12, 2022</td>
<td>Amazon Personalize now supports promoting items in recommendations with a separate promotion filter. For information about promoting items see <a href="#">Promoting items in recommendations</a>.</td>
<td></td>
</tr>
<tr>
<td>August 12, 2022</td>
<td>Amazon Personalize now supports using comparison operators in filter expressions with placeholder parameters. For information about filter expressions, see <a href="#">Filter expressions</a>.</td>
<td></td>
</tr>
<tr>
<td>August 2, 2022</td>
<td>Amazon Personalize now supports incremental bulk updates to datasets. You can now use a dataset import job to update a dataset without replacing the existing data. For more information, see <a href="#">Updating existing bulk data</a>.</td>
<td></td>
</tr>
<tr>
<td>June 6, 2022</td>
<td>Amazon Personalize can now use unstructured text metadata in different languages. For more information, see <a href="#">Unstructured text metadata</a>.</td>
<td></td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize recommenders now generate offline metrics. You can use these metrics to evaluate the performance of your recommender. For more information see Evaluating a recommender.</td>
<td>May 24, 2022</td>
</tr>
<tr>
<td>---</td>
<td>---</td>
<td>---</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability stop a recommender and restart it later. This way, you can pause recommender billing and pay for it only when you use it. For more information, see Stopping and starting a recommender.</td>
<td>April 20, 2022</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports using tags to categorize and manage Amazon Personalize resources. For more information, see Tagging Amazon Personalize resources.</td>
<td>April 7, 2022</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports specifying resources with AWS CloudFormation. For more information, see Specifying resources with AWS CloudFormation.</td>
<td>March 11, 2022</td>
</tr>
<tr>
<td>New documentation feature (p. 738)</td>
<td>The Amazon Personalize developer guide now includes a new Troubleshooting topic that provides answers to common questions and troubleshooting advice for error messages that you might encounter with Amazon Personalize. For more information, see Troubleshooting.</td>
<td>February 15, 2022</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports creating a Domain dataset group with use case optimized resources for video on demand or e-commerce domains. For more information, see Domain dataset groups.</td>
<td>November 29, 2021</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports creating user segments with new USER_SEGMENTATION recipes. USER_SEGMENTATION recipes generate segments of users based on item input data. For more information, see USER_SEGMENTATION recipes.</td>
<td>November 29, 2021</td>
</tr>
<tr>
<td>----------------------</td>
<td>-------------------------------------------------------------------------------------------------</td>
<td>------------------</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports a new RELATED_ITEMS recommendation recipe Similar-Items. Use the Similar-Items recipe to generate recommendations for similar items based on both interactions data and item metadata. For more information, see Similar Items recipe.</td>
<td>October 5, 2021</td>
</tr>
<tr>
<td>New documentation feature (p. 738)</td>
<td>The Amazon Personalize developer guide now includes a getting started tutorial for using Amazon Personalize with the SDK for Java 2.x. For more information, see Getting started (SDK for Java 2.x).</td>
<td>August 25, 2021</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize can now extract meaningful information from unstructured text metadata in an Items dataset. For more information, see Items dataset.</td>
<td>June 9, 2021</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to stop creating a solution version (stop training a model). For more information, see Stopping the creation of a solution version.</td>
<td>May 20, 2021</td>
</tr>
<tr>
<td>New feature (preview release) (p. 738)</td>
<td>Amazon Personalize can now optimize a solution for an objective in addition to maximizing relevance, such as maximizing revenue. This feature is in preview release. For more information, see Optimizing a solution for an additional objective.</td>
<td>May 18, 2021</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize can now export the records in an Amazon Personalize dataset to an Amazon S3 bucket for analysis and tracking. For more information, see Exporting a dataset.</td>
<td>April 26, 2021</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now automatically updates the latest model (solution version) you trained with User-Personalization every two hours to include new data. For more information, see User-personalization recipe.</td>
<td>November 17, 2020</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize can now filter recommendations based on criteria you specify when you get recommendations. For more information, see Filtering recommendations.</td>
<td>November 10, 2020</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports the ability to incrementally import users and items. For more information, see Importing records incrementally.</td>
<td>October 2, 2020</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now supports a new USER_PERSONALIZATION recommendation recipe. USER_PERSONALIZATION features include modeling impression data, automatic item exploration, and automatic cold item selection. For more information, see User-personalization recipe.</td>
<td>August 5, 2020</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize can now filter recommendations based on item and user metadata using custom filter expressions. For more information, see Filtering recommendations.</td>
<td>July 31, 2020</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now allows you to filter results based on which items a user has interacted with. For more information, see Filtering recommendations.</td>
<td>June 3, 2020</td>
</tr>
<tr>
<td>New feature (p. 738)</td>
<td>Amazon Personalize now exposes scores for recommended items. Scores represent the Amazon Personalize model's certainty that a user will next choose a certain item. For more information, see Getting recommendations.</td>
<td>April 3, 2020</td>
</tr>
<tr>
<td>Category</td>
<td>Description</td>
<td>Date</td>
</tr>
<tr>
<td>--------------------------------</td>
<td>------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------</td>
<td>--------------------</td>
</tr>
<tr>
<td><strong>New Region (p. 738)</strong></td>
<td>Amazon Personalize adds support for the Asia Pacific (Seoul) Region. For a complete list of the AWS Regions supported by Amazon Personalize, see the <a href="https://aws.amazon.com/about-aws/whats-new/service-status/">AWS Region table</a> or <a href="https://aws.amazon.com/about-aws/whats-new/service-status/">AWS Regions and endpoints</a> in the Amazon Web Services General Reference.</td>
<td>January 21, 2020</td>
</tr>
<tr>
<td><strong>New feature (p. 738)</strong></td>
<td>Amazon Personalize can now get recommendations based on contextual metadata. For more information, see <a href="https://docs.aws.amazon.com/personalize/latest/dg/guide-contextual-metadata.html">Getting recommendations</a>.</td>
<td>December 19, 2019</td>
</tr>
<tr>
<td><strong>New regions (p. 738)</strong></td>
<td>Amazon Personalize adds support for the Asia Pacific (Mumbai), Asia Pacific (Sydney), and Canada (Central) Regions. For a complete list of the AWS Regions supported by Amazon Personalize, see the <a href="https://aws.amazon.com/about-aws/whats-new/service-status/">AWS Region table</a> or <a href="https://aws.amazon.com/about-aws/whats-new/service-status/">AWS Regions and endpoints</a> in the Amazon Web Services General Reference.</td>
<td>December 18, 2019</td>
</tr>
<tr>
<td><strong>New feature (p. 738)</strong></td>
<td>Amazon Personalize now supports batch recommendation workflows. For more information, see <a href="https://docs.aws.amazon.com/personalize/latest/dg/batch-processing.html">Get batch recommendations</a>.</td>
<td>November 14, 2019</td>
</tr>
<tr>
<td><strong>Amazon Personalize general availability (p. 738)</strong></td>
<td>Amazon Personalize is now available for general use.</td>
<td>June 10, 2019</td>
</tr>
<tr>
<td><strong>Amazon Personalize preview release (p. 738)</strong></td>
<td>This is the first preview release of the documentation for Amazon Personalize.</td>
<td>November 28, 2018</td>
</tr>
</tbody>
</table>
AWS glossary

For the latest AWS terminology, see the AWS glossary in the AWS Glossary Reference.