

Neptune Analytics User Guide

Neptune Analytics



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Neptune Analytics: Neptune Analytics User Guide

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What is Neptune Analytics?

Neptune Analytics is a memory-optimized graph database engine for analytics. With Neptune Analytics, you can get insights and find trends by processing large amounts of graph data in seconds. To analyze graph data quickly and easily, Neptune Analytics stores large graph datasets in memory. It supports a library of optimized graph analytic algorithms, low-latency graph queries, and vector search capabilities within graph traversals.

Neptune Analytics is an ideal choice for investigatory, exploratory, or data-science workloads that require fast iteration for data, analytical and algorithmic processing, or vector search on graph data. It complements <u>Amazon Neptune Database</u>, a popular managed graph database. To perform intensive analysis, you can load the data from a Neptune Database graph or snapshot into Neptune Analytics. You can also load graph data that's stored in Amazon S3.

Topics

- Features
- Pricing
- Getting started
- When to use Neptune Analytics and when to use Neptune Database

Features

Neptune Analytics operates in a managed environment that can load data extremely fast into memory and run graph algorithms natively. With Neptune Analytics, you can perform in-database analytics on large graphs.

This functionality lets you perform business intelligence and custom analytical queries, and use pre-built graph algorithms with the openCypher language. For example, with Neptune Analytics you can ingest text from cybersecurity reports to analyze relationships within security environments and calculate vulnerability mitigations using graph algorithms or openCypher queries.

Neptune Analytics offers a graph as a service experience by managing graphs instead of infrastructure, so you can focus on queries and workflows to solve problems. It automatically provisions the compute resources necessary to run analytics workloads based on the size of the graph.

You can load graph data into Neptune Analytics from Amazon S3 or from a Neptune Database endpoint. You can then run graph analytics queries using pre-built or custom graph queries.

Pricing

Neptune Analytics uses a simple, predictable pricing model based on the resources consumed. There are no additional costs or long-term contracts.

Getting started

You can get started by <u>creating a new Neptune Analytics graph</u> and loading data into it in a number of ways.

When to use Neptune Analytics and when to use Neptune Database

Amazon Neptune makes it easy to work with graph data in the AWS Cloud. Amazon Neptune includes both Neptune Database and Neptune Analytics.

<u>Neptune Database</u> is a serverless graph database designed for optimal scalability and availability. It provides a solution for graph database workloads that need to scale to 100,000 queries per second, Multi-AZ high availability, and multi-Region deployments. You can use Neptune Database for social networking, fraud alerting, and Customer 360 applications.

Neptune Analytics is an analytics database engine that can quickly analyze large amounts of graph data in memory to get insights and find trends. Neptune Analytics is a solution for quickly analyzing existing graph databases or graph datasets stored in a data lake. It uses popular graph analytic algorithms and low-latency analytic queries.

You can use Neptune Analytics to analyze and query graphs in data science workflows that build targeted content recommendations, assist with fraud investigations, and detect network threats.

By providing a simple API for loading, querying, and analyzing graph data, Neptune Analytics also removes the overhead of building and managing complex data-analytics pipelines.

Neptune Analytics makes it easy to apply powerful algorithms both to the data in your Neptune Database and to graph data that's stored externally. Because Neptune Analytics can load a large dataset very quickly into memory, it becomes possible to analyze graphs with tens of billions of relationships and to process thousands of analytic queries per second using popular graph analytics algorithms.

Changes and updates to Neptune Analytics

The following table lists important releases relating to Neptune Analytics.

Change	Description	Date
<u>StartImportTask available</u>	Neptune Analytics now allows you to efficiently import large datasets into an already provisioned graph database using the StartImportTask API. This API facilitates the direct loading of data from an Amazon S3 bucket into an empty Neptune Analytics graph. This is designed for loading data into existing empty clusters.	March 30, 2024
Neptune Analytics available in Europe (London) region	Neptune Analytics is now available in the eu-west-2 Europe (London) region.	March 14, 2024
<u>Column delimiter ; is now</u> <u>supported in values</u>	When a column delimiter presents in the values, e.g. for 'two;words', two values {'two','words'} will be inserted. However, if the column delimiter is escaped by a backslash '\', the value will be inserted as a whole with the escape character removed, e.g. 'one\;word' will be inserted as 'one;word'.	February 27, 2024

Graph provisioning time reduced to five minutes or less	Neptune Analytics graphs are now provisioned and ready to be used in five minutes or less.	February 19, 2024
<u>Query improvements - Data</u> <u>Plane SDK</u>	The Neptune Analytics data API provides support for data operations including query execution, query status checking, query cancellation, and graph summarizing via the HTTPS endpoint, the AWS CLI, and the SDK.	February 2, 2024
Initial release	Initial release of Neptune Analytics.	November 29, 2023

Getting started

To get started using Neptune Analytics, you need to create a graph using the AWS console, the AWS CLI, or AWS CloudFormation. You can load data into a graph from another Neptune database, Neptune database cluster snapshot, or from files located in Amazon S3.

Topics

- Create an empty Neptune graph
- Create a Neptune graph from existing sources

Create an empty Neptune graph

AWS console

- 1. Sign in to the AWS Management Console, and open the Amazon Neptune console at https://console.aws.amazon.com/neptune/.
- 2. In the upper right corner of the console, choose the AWS region in which you want to create the graph.
- 3. In the navigation pane, choose **Graphs** in the **Analytics** section.
- 4. Choose the **Create graph** button.
- 5. In **settings**, input the **graph name**, size, and replica configuration.
- 6. In the **data source** section, choose the **empty graph** option.

🚯 Note

Additional charges equivalent to the m-NCUs selected for the graph apply for each replica.

Neptune > Graphs > Create Graph
Create graph You can create a graph by specifying the graph name, the source of the data, network configuration, and the IAM roles to load he data into.
Settings
Graph name Specify a unique name for the graph. neptune-graph
The name must be unique cross all graphs owned by your AWS account in the current AWS Region, case-insensitive, 1 to 60 alphanumeric characters or hyphens, first character must be a letter, it can't contain two consecutive hyphens, it can't end with a hyphen.
Data source Graph import task O Create empty graph O Create Graph from existing source You can create a new graph by directly importing data into it from Neptune, an S3 bucket or a Neptune
Database snapshot. Memory-Optimized Neptune Credit Units (m-NCU) Info Image: State of the state of th
Availability settings
Replicas configuration A replica of your graph in another AZ provides warm failover capabilities.
• Default replicas One replica will be created in another AZ by default. O Use custom number of replicas You can specify the number of replicas for your graph.

7. You can connect to a Neptune graph from a public endpoint or a private endpoint. Select your network configuration accordingly.

Network and security
 Enable public connectivity Allows your graph to be reachable over the Internet. All access to the graph requires IAM authentication.
Your data is encrypted by default with a key that AWS owns and manages for you. To choose a different key, customize your encryption settings. Info
Customize encryption settings (advanced)
Private endpoint
 Set up Private Endpoint You can access a graph from within a VPC by creating a private graph endpoint, which ensures that the traffic is never exposed to the internet. You can further attach security groups, tags, or enable flow logs on the graph endpoints. Standard VPC interface endpoint charges apply. Virtual Private Cloud (VPC) Select the VPC in which the private graph endpoint should be created.
Default vpc-a7ff95c1
Subnets Select the subnets from which the endpoint will have access. Choose option C
subnet-9eaf8ab3 X us-east-1d VPC security groups Security groups have rules authorizing connections from all the EC2 instances and devices that need to access the graph.
Choose existing VPC security groups Create new VPC security group
Existing VPC security groups Select a list of EC2 VPC security groups to associate with the new graph endpoint.
Choose option C
default X VPC vpc-a7ff95c1 ID sg-49045b35

- 8. Additionally, you can select vector search configuration for the graph. For more information on vector search configuration, see <u>Vector indexing</u>.
- 9. Choose Create graph.

AWS CLI

Create a Neptune graph using the AWS CLI.

Create a public graph endpoint:

```
aws neptune-graph create-graph --graph-name 'test-neptune-graph' \
--region us-east-1 --provisioned-memory 128 --allow-from-public \
--replica-count 0 --vector-search '{"dimension": 384}'
```

Create a private graph endpoint:

```
aws neptune-p8 create-private-graph-endpoint -vpc-id vpc-0a9b7a5b15 \
--subnet-ids subnet-06a4b41a6221b subnet-0840a4b327ab77 subnet-0353627ab123 \
--vpc-security-group-ids sg-0ab7abab56ab \
--graph-identifier g-146a51b7a151ba -region us-east-1
```

AWS CloudFormation

Instead of using the console to create your Neptune graph, you can use AWS CloudFormation to provision AWS resources by treating infrastructure as code. To help you organize your AWS resources into smaller and more manageable units, you can use the AWS CloudFormation nested stack functionality. For more information, see <u>Creating a stack on the AWS</u> <u>CloudFormation console</u> and <u>working with nested stacks</u>.

AWS CloudFormation is free, but the resources that CloudFormation creates are live. You incur the standard usage fees for these resources until you terminate them. The total charges will be minimal. For information about how you might minimize any charges, see <u>AWS free tier</u>.

To create your resources using the AWS CloudFormation console, complete the following steps:

- 1. Create the CloudFormation template.
- 2. Configure your resources using CloudFormation.

The following sample template will create a Neptune graph with a public endpoint.

```
AWSTemplateFormatVersion: 2010-09-09
Description: NeptuneGraph Graph Create Demo using CloudFormation
Resources:
```

```
NeptuneGraph:
Type: AWS::NeptuneGraph::Graph
DeletionPolicy: Delete
Properties:
DeletionProtection: false
GraphName: neptune-graph-demo
ProvisionedMemory: 128
ReplicaCount: 1
PublicConnectivity: true
Tags:
- Key: stage
Value: test
```

The following sample template will create a Neptune graph with a public endpoint.

```
AWSTemplateFormatVersion: 2010-09-09
Description: NeptuneGraph Graph Create Demo using CloudFormation
Resources:
  NeptuneGraph:
    Type: AWS::NeptuneGraph::Graph
    DeletionPolicy: Delete
    Properties:
      DeletionProtection: false
      GraphName: neptune-graph-demo
      ProvisionedMemory: 128
      ReplicaCount: 1
      PublicConnectivity: false
      Tags:
        - Key: stage
          Value: test
  NeptuneGraphPrivateEndpoint:
    Type: AWS::NeptuneGraph::PrivateGraphEndpoint
    DeletionPolicy: Delete
    Properties:
      GraphIdentifier: NeptuneGraph
      VpcId: myVpc
```

▲ Important

You can't change the graph name, VPC, subnet ids and vector search configuration. After starting the graph creation, the graph has a status of **Creating** until the graph is ready to use. When the status of the graph changes to **Available**, you can connect to the DB cluster at that time. Depending on the configuration, it can take up to 20 minutes before the new graph is available.

Create a Neptune graph from existing sources

You can load data into a Neptune graph from another Neptune database, Neptune database cluster snapshot, or from Amazon S3 files. Select the data sources and an IAM role for the data import accordingly. For more information about loading data, see <u>Create a graph from Amazon S3, a</u> <u>Neptune cluster, or a snapshot</u>.

Data source	
Graph import task	
Create empty graph	 Create Graph from existing source You can create a new graph by directly importing data into it from Neptune, an S3 bucket or a Neptune Database snapshot.
Minimum Memory-Optimized Neptune Credit Units (m- NCU) Info 🖸 Minimum number of m-NCUs to be allocated to your graph	Maximum Memory-Optimized Neptune Credit Units (m- NCU) Info 🔀 Maximum number of m-NCUs to be allocated to your graph
Default 🔻	Default 🔹
Load role ARN Specify the ARN of role with permissions to load data from the selec GraphExecutionRole	ted source into the graph.
Select the type of source	
Neptune cluster When you import your data from an existing Neptune cluster, Ne exports the data from your cluster and stores it in the user specif a user specified KMS key.	
Neptune cluster snapshot When you import your data from an existing Neptune snapshot, exports the data from your snapshot and stores it in the user spe with a user specified KMS key.	
S 3	
Import task format	
CSV	▼
S3 Bucket location Resource URI	
Q s3://neptune-demo-data-us-east-1/demo-data-csv/	X View 🖸 Browse S3
Path must be in the form s3://bucket/prefix/. It must end with a slas	

AWS CLI

The following example creates a graph and loads data from Amazon S3.

```
aws neptune-graph create-graph-using-import-task \
--graph-name "neptune-graph-from-s3-source" \
--region "us-east-1" \
--format "CSV" \
--role-arn "arn:aws:iam::1234567890124:role/GraphExecutionRole" \
--source "s3://neptune-demo-test-us-east-1/test-data-csv/" \
--public-connectivity \
--min-provisioned-memory 256 \
--max-provisioned-memory 256
```

Using notebooks with Neptune Analytics

The Neptune managed open-source <u>graph-notebook project</u> provides a plethora of <u>Jupyter</u> extensions and sample notebooks that make it easy to interact with and learn to use a Neptune Analytics graph.

These graph notebooks support a suite of intuitive Jupyter line- and cell-magic commands. The magic commands abstract away much of the initial setup typically required for using Neptune Analytics, and take care of SigV4 signing of requests. They can create graph connections, load data, run openCypher queries, and interact with various Neptune Analytics APIs.

You can find a list of the full set of Neptune graph-notebook magics and their options <u>in the</u> <u>Neptune Userguide</u>. However, only the following magics are compatible with Neptune Analytics graphs:

- <u>%seed</u> (adds sample data to a graph).
- <u>%load</u> (uses the <u>neptune-load()</u> openCypher integration to let you batch-load data).
- <u>%status</u> (gets status information about the graph).
- <u>%%opencypher or %%oc</u> (issues an openCypher query).
- <u>%opencypher_status</u>, or <u>%oc_status</u> (retrieves query status for an opencypher query).
- <u>%%graph_notebook_config</u> (displays a JSON object containing the configuration that the notebook is using).
- <u>%graph_notebook_host</u> (sets the line input as the notebook's host).
- <u>%graph_notebook_version</u> (returns the Neptune workbench notebook release number).
- <u>%%graph_notebook_vis_options</u> (lets you set visualization options for the notebook).
- <u>%summary</u> (retrieves graph summary information).
- <u>%statistics</u> (retrieves or manages graph data statistics; the refresh mode is disabled).

You can use a Neptune graph notebook to generate an interactive visualization of the results returned from an openCypher query, and use options to customize the appearance of the visualized graph (see <u>Graph visualization in the Neptune workbench</u>).

Take advantage of all the sample notebooks

A wide variety of sample Jupyter notebooks are available in the Neptune <u>graph-notebook project</u>. Some of these are purpose-built for learning how to get the most of a Neptune Analytics graph and its powerful built-in algorithms in the context of common real-world applications.

After installing the graph-notebook project either locally or on SageMaker, you should be able to find sample notebooks under the notebook directory, .../Neptune/02-Neptune-Analytics.

Creating a new Neptune Analytics notebook using an AWS CloudFormation template

<u>Amazon SageMaker Notebook instances</u> provide a fully managed Jupyter environment for running graph notebooks that are connected to a Neptune Analytics graph. SageMaker Notebooks run natively on Amazon Linux 2, and support use of the Jupyter Classic Notebook or JupyterLab 3 interface on the same instance.

You can use one of the following AWS CloudFormation templates to set up a new Neptune Analytics notebook to use with your Neptune Analytics graph:

To use an AWS CloudFormation stack to create a new Neptune Analytics notebook

1. Choose one of the **Launch Stack** buttons in the following table to launch the AWS CloudFormation stack on the AWS CloudFormation console.

Region	View	View in Designer	Launch
US East (N. Virginia)	View	View in Designer	Launch Stack 🕖
US East (Ohio)	View	View in Designer	Launch Stack 🕠
US West (Oregon)	View	View in Designer	Launch Stack 🕖
Europe (Ireland)	View	View in Designer	Launch Stack 🕖

Region	View	View in Designer	Launch
Europe (Frankfurt)	View	View in Designer	Launch Stack 🕖
Asia Pacific (Tokyo)	View	View in Designer	Launch Stack 🕖
Asia Pacific (Singapore)	View	View in Designer	Launch Stack 🕖

- 2. On the Select Template page, choose Next.
- 3. In the **Stack Details** page, under **GraphEndpoint**, enter the public or private endpoint of your Neptune Analytics graph.
- 4. Under **Notebook Name** enter a name for the new notebook that is unique for your account and region in SageMaker.
- 5. On the **Options** page, choose **Next**.
- 6. If you're using a private endpoint for your Neptune Analytics graph, enter the following under **Network Options**:
 - a. Under **GraphVPC** enter the ID of a VPC associated with the private graph endpoint.
 - b. Under **GraphSubnetId** enter the ID of any subnet associated with your private graph endpoint.
 - c. Under **GraphSecurityGroup** enter the ID of a security group associated with the VPC. This is optional; if not provided, a new security group is automatically created for this purpose.
- 7. Click through the rest of the stack creation steps, leaving everything as default, and submit for creation.

In around 5 minutes, you should see the new Neptune Analytics notebook appear in the SageMaker and Neptune consoles.

Creating a new Neptune Analytics notebook using the AWS Management Console

You can create a new notebook manually using the AWS Management Console if you aren't able to use AWS CloudFormation. The first thing you need is an IAM role to use for the notebook. If you already have one, you can skip the following section.

Create an IAM role for a Neptune Analytics notebook

To create an IAM role for a Neptune Analytics notebook

- 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, expand **Access management**, then choose **Roles**.
- 3. Select Create role.
- 4. Under **Trusted entity type**, select **Custom trust policy** and copy in the following trust policy:

```
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Principal": {
               "Service": "sagemaker.amazonaws.com"
        },
        "Action": "sts:AssumeRole"
        }
   ]
}
```

- 5. Choose **Next**, and then **Next** again.
- 6. Enter a name and description for the role, and select **Create role**.
- 7. Go back to the **Roles** page, search for the name of the role you just created, and open it.
- 8. On the **Permissions** tab Under **Permissions policies**, select **Add permissions** and choose **Create inline policy**.
- 9. In the **Policy editor**, switch to the **JSON** option, and copy in the following policy:

{

```
Create a notebook on the console
```

```
"Version": "2012-10-17",
  "Statement": [
    {
      "Effect": "Allow",
      "Action": [
        "s3:GetObject",
        "s3:ListBucket"
      ],
      "Resource": [
        "arn:aws:s3:::aws-neptune-notebook-(AWS region)",
        "arn:aws:s3:::aws-neptune-notebook-(AWS region)/*"
     ]
    },
    {
      "Effect": "Allow",
      "Action": "neptune-graph:*",
      "Resource": [
        "arn:aws:neptune-graph:(AWS region):(AWS account ID):graph/(Neptune Graph
resource ID)"
      ]
    },
    {
      "Effect": "Allow",
      "Action": [
        "logs:CreateLogGroup",
        "logs:CreateLogStream",
        "logs:PutLogEvents"
     ],
      "Resource": [
        "arn:aws:logs:*:*:log-group:/aws/sagemaker/*"
     ]
    },
    {
      "Effect": "Allow",
      "Action": "sagemaker:DescribeNotebookInstance",
      "Resource": [
        arn:aws:sagemaker:(AWS region):(AWS account ID):notebook-instance/*"
      ]
    }
 ]
}
```

10. Choose Next.

- 11. Give a name to the inline policy.
- 12. Select **Create policy**. Make note of the name of the policy you just created.

Next, create the Neptune Analytics notebook in SageMaker

- 1. Open the Amazon SageMaker console at https://console.aws.amazon.com/sagemaker/.
- 2. In the navigation pane, expand Notebook, then choose Notebook instances.
- 3. Choose Create notebook instance.
- 4. In **Notebook instance settings**, under **Notebook instance name**, give the notebook a name prefixed by aws-neptune- (for example, aws-neptune-my-test-notebook).
- 5. Under Platform identifier, select Amazon Linux 2, JupyterLab 3.
- 6. Select Additional configuration.
- 7. Under Lifecycle configuration, choose Create a new lifecycle configuration.
- 8. In **Configuration**, under **Name** enter the notebook instance name from step 4.
- 9. In **Scripts**, under **Start notebook**, replace the existing script with this:

```
#!/bin/bash
sudo -u ec2-user -i <<'EOF'</pre>
echo "export GRAPH_NOTEBOOK_AUTH_MODE=IAM" >> ~/.bashrc
echo "export GRAPH_NOTEBOOK_SSL=True" >> ~/.bashrc
echo "export GRAPH_NOTEBOOK_SERVICE=neptune-graph" >> ~/.bashrc
echo "export GRAPH_NOTEBOOK_HOST=(Neptune Analytics graph endpoint, public or
private)" >> ~/.bashrc
echo "export GRAPH_NOTEBOOK_PORT=8182" >> ~/.bashrc
echo "export NEPTUNE_LOAD_FROM_S3_ROLE_ARN=" >> ~/.bashrc
echo "export AWS_REGION=(AWS region)" >> ~/.bashrc
aws s3 cp s3://aws-neptune-notebook-(AWS region)/graph_notebook.tar.gz /tmp/
graph_notebook.tar.gz
rm -rf /tmp/graph_notebook
tar -zxvf /tmp/graph_notebook.tar.gz -C /tmp
/tmp/graph_notebook/install.sh
EOF
```

- 10. Select Create configuration.
- 11. In **Permissions and encryption**, under **IAM Role**, select the role you created above.
- 12. In **Network**, if you are using a private graph endpoint:
 - a. Under **VPC**, select the VPC where the Neptune Analytics graph resides.
 - b. Under **Subnet**, select a subnet associated with the Neptune Analytics graph.
 - c. Under **Security Group(s)**, select all the security groups associated with the Neptune Analytics graph.
- 13. Choose **Create notebook instance**.
- 14. After 5 or 10 minutes, when your new notebook reaches Ready status, select it. Choose **Open Jupyter** or **Open JupyterLab**.

Hosting a Neptune Analytics graph-notebook on your local machine

It is also possible to install and run a Neptune Analytics graph notebook on your local machine. You can find instructions in the <u>GitHub graph-notebook repository</u>:

- Prerequisites
- Jupyter Classic Notebook or <u>https://github.com/aws/graph-notebook/#jupyterlab-3x</u> installation
- <u>Connecting to Neptune</u>

When setting up for Neptune Analytics:

- When setting the connection using <u>%%graph_notebook_config</u>, make sure to set the neptune_service field to the value neptune-graph.
- If you're connecting to a private graph endpoint, you need to enable access to the VPC where the Neptune Analytics instance resides. The easiest way to set this is up is using an SSH tunnel to a proxy EC2 instance in the VPC. For more information, see <u>Connecting graph notebook locally to</u> <u>Amazon Neptune in GitHub</u>.
- If you're using a public graph endpoint, no additional connectivity setup is required.

Creating a new Neptune Analytics graph using the AWS Management Console

You can use the Neptune console to create a new Neptune Analytics graph.

i Note

If you are working with a large dataset (on the order of 50 GiB or larger) that you intend to load at the same time the new graph is created, be sure to <u>create an IAM role</u> that grants permissions to load the dataset from the location where it resides.

To use the Neptune console to create a graph

- Sign in to the AWS Management Console, and open the Amazon Neptune console at https://console.aws.amazon.com/neptune/.
- 2. In the navigation pane, select **Graphs** under **Analytics**.
- 3. Select Create graph.
- 4. Enter a name for the new graph.
- 5. The next steps depend on whether you are creating an empty graph or one preloaded with data.
 - If you choose **Create empty graph**, choose the number of memory-optimized Neptune Capacity Units (m-NCUs) to allocate to the new Neptune Analytics graph, between 128 and 1024. Each m-NCU has around one GiB of memory capacity and corresponding compute and networking.
 - If you choose **Create Graph from existing source**, Neptune Analytics will bulk-load data for you when the graph is created. Choose this option if you want to import a large dataset, on the order of 50 GiB or larger. See <u>Bulk import</u> for details.
 - a. Set values for the minimum and maximum m-NCUs, or just leave them at their default values (128 m-NCUs). The units are memory-optimized Neptune Capacity Units (m-NCUs), each of which is roughly equivalent to 1 GiB of memory and corresponding compute and networking. Neptune Analytics evaluates the data that you want to load and estimates the resources needed to handle it, within the range of m-NCUs that you specify.

- b. Under Load role ARN, select an IAM role that you have created to provide the necessary permissions for the data import. See <u>Create an IAM role with permissions to export from Neptune to Neptune Analytics</u> for instructions about how to create the role.
- c. The next steps depend on what source you're loading data from:
 - If you choose **Neptune cluster** as the type of source, select the DB cluster that you want to load from under **Neptune cluster**.
 - If you choose **Neptune cluster snapshot** as the type of source, select one of your manual DB snapshots that you want to load from under **Neptune DB snapshot**.
 - If you choose **S3** as the type of source, enter the URL of the Amazon S3 location where the data file(s) to be loaded are located, under **Resource URI**. The path to the folder location must end in a slash rather than specify to a particular file.
- 6. Under **Availability settings**, choose how many failover replicas you want to create for the new graph. The default is one, but if you select Use custom number of replicas you can choose from zero to two failover replicas.

🔥 Important

Additional charges equivalent to the m-NCUs selected for the graph apply for each replica.

- 7. Under **Network and security**, check **Allow from public** to create a public endpoint for your new Neptune Analytics graph to make it accessible over the internet. If you want to use your own KMS key to encrypt your data, check **Customize encryption settings** a specify a KMS key of your choosing.
- 8. Under **Vector search settings**, if you want to set up a vector index for the graph, choose **Use vector dimension** and then specify the number of dimensions for the vectors in the index.
- 9. Under Advanced settings, you can make it easier to delete your new graph by selecting **Turn** off deletion protection. Deletion protection is turned on by default.
- 10. Finally, under **Tags**, you can associate tags with your new Neptune Analytics graph.
- 11. When everything is configured as you want it to be, choose **Create Graph**.

Loading data into a Neptune Analytics graph

Neptune Analytics provides several options for loading data into a graph.

- Bulk import Create a graph loaded with data from files in <u>Amazon S3</u> in a CSV-like <u>data</u> format, or data in a <u>Neptune Database cluster</u>. You can also load data from files in Amazon S3 into an <u>empty graph</u>. This could be the fastest way to load large volumes of initial data.
- **Batch load** Add more data to an existing <u>non-empty graph</u> from files in Amazon S3 in a CSVlike <u>data format</u>. This could be the fastest way to add more data or update single cardinality property values in a non-empty graph. The volume of data that can be ingested in a single request is lower than what bulk import can support, and multiple requests with smaller batches of files could be a workaround.
- openCypher queries Add more data through <u>queries</u>, if data is not available from files in Amazon S3 or the data volume is small. This is also a more generic approach for conditional inserts based on data already in the graph, and updating contents of the graph.

Topics

- Data format for loading from Amazon S3 into Neptune Analytics
- Batch load
- Bulk import data into a graph

Data format for loading from Amazon S3 into Neptune Analytics

Neptune Analytics, just like <u>Neptune Database</u>, supports two formats for loading property graph data: <u>csv</u> and <u>opencypher</u>. Both are csv-based formats with schema restrictions. A csv file must contain a header row and the column values. The remainder of the files are interpreted based on the corresponding header column. The header could contain predefined system column names and user-defined column names annotated with predefined datatypes and cardinality.

Differences with Neptune csv (opencypher) format

Edge files:

• The ~id (:ID) column in edge (relationship) files in CSV (opencypher) format are not supported. They are ignored if provided in any of the edge (relationship) files.

Vertex files:

- Only explicitly provided labels are associated with the vertices. If the label provided is empty, the vertex would be added without a label. If a row contains just the vertex id without any labels or properties then the row is ignored, and no vertex is added. For more information about vertices, see <u>vertices</u>.
- A new column type Vector is supported for associating embeddings with vertices. Since Neptune Analytics only supports one index type at this moment, the property name for embeddings is currently fixed to Embeddings. If the element type of the embedding are not floating point (FP32), there will be typecasted to FP32. The embeddings in the csv files are optional when the vector index is enabled. This means that not every node needs to be associated with an embedding. If you want to set up a vector index for the graph, choose use vector dimension and then specify the number of dimensions for the vectors in the index. Note that the dimension must match the dimension of the embeddings in the vertex files. For more details of loading embeddings, refer to vector-index.

Edge or vertex files:

- Unlike Neptune Database, a vertex identifier could appear just in edge files. Neptune Analytics allows loading just the edge data from files in Amazon S3, and running an algorithm over the data without needing to provide any additional vertex information. The edges are created between vertices with the given identifiers, and the vertices have no labels or properties unless any are provided in the vertex files. For more information on vertices and what they are, see <u>vertices</u>.
- Date column type is not supported. The datetime column type is supported and could be used instead. Pure date values in the datetime column are implicitly assigned 00:00:00z time.
- The datetime values can either be provided in the XSD format or one of the following formats:
 - yyyy-MM-dd
 - yyyy-MM-ddTHH:mm
 - yyyy-MM-ddTHH:mm:ss
 - yyyy-MM-ddTHH:mm:ssZ
 - yyyy-MM-ddTHH:mm:ss.SSSZ

- yyyy-MM-ddTHH:mm:ss[+|-]hhmm
- yyyy-MM-ddTHH:mm:ss.SSS[+|-]hhmm
- Float and double values in scientific notation are currently not supported. Also, Infinity, INF, -Infinity, -INF, and NaN (Not-a-number) values are not recognized.
- Gzip files are not supported.
- The maximum length of the strings supported is smaller, and limited to 1,048,062 bytes. The limit would be lower for strings with unicode characters since some unicode characters are represented using multiple bytes.
- Multi-line string values are not supported. Imports behavior is undefined if the dataset contains multi-line string values.
- Quoted string values must not have leading space between the delimiter and quotes. For example, if a line is abc, "def" then that is interpreted as line with two fields with string values abc and "def". "def" is a non-quoted string field and quotes are stored as is in the value, with a size of 6 characters. If the line is abc, "def" then it is interpreted as a line with two fields with string values abc and def.

Batch load

Neptune Analytics supports a CALL procedure neptune.load to load data from Amazon S3, to insert new vertices, edges, and properties, or to update single cardinality vertex property values. It executes as a mutation query and does atomic writes. It uses the IAM credentials of the caller to access the data in Amazon S3. See <u>Create your IAM role for Amazon S3 access</u> to set up the permissions.

Request syntax

The signature of the CALL procedure is shown below:

```
CALL neptune.load(
    {
        source: "string",
        region: "us-east-1",
        format: "csv",
        failOnError: "true",
        concurrency: 1
    }
```

)

- source (required) An Amazon S3 URI prefix. All object names with matching prefixes are loaded. See <u>Neptune Database loader reference</u> for Amazon S3 URI prefix examples. The IAM user who signs the openCypher request must have permissions to list and download these objects, and must be authorized for WriteDataViaQuery and DeleteDataViaQuery actions. See <u>IAM role</u> <u>mapping</u> for more IAM authentication related details.
- **region** (required) The AWS region where the Amazon S3 bucket is hosted. Currently, cross-region loads are not supported.
- format (required) The data format of the Amazon S3 data to be loaded, either csv or opencypher. For more information, see <u>Data format for loading from Amazon S3 into Neptune</u> <u>Analytics</u>.
- failOnError (optional) default: true If set to true (the default), the load process halts whenever there is an error parsing or inserting data. If set to false, the load process continues and commits whatever data was successfully inserted.

Inserting edge or relationship data when failOnError is set to true is not recommended, since a subsequent load would lead to duplication of partially committed edges or relationships.

 concurrency (optional) default: 1 – This value controls the number of threads used to run the load process, up to the maximum available.

🚯 Note

Unlike bulk import, there is no need to pass the role-arn for batch load since the IAM credentials of the signer of the openCypher query are used to download data from Amazon S3. The signer must have permissions to download data from Amazon S3 with the trust relationship set up to assume the role, so that Neptune Analytics can assume the role to load the data into the graph from files in Amazon S3.

Response syntax

A sample response is shown below.

```
"results": [
```

{

```
{
    "totalRecords": 108070,
    "totalDuplicates": 46521,
    "totalTimeSpentMillis": 558,
    "numThreads": 16,
    "insertErrors": 0,
    "throughputRecordsPerSec": 193673,
    "loadId": "13a60c3b-754d-c49b-4c23-06b9dd5b346b"
    }
]
```

- totalRecords: The number of graph elements vertex labels, edges, and properties attempted for insertion.
- totalDuplicates: The count of duplicate graph elements vertex labels or properties encountered. These elements may have pre-existed before the load request or were duplicates within the input CSV files. Each edge is treated as new, so edges are excluded from this count.
- totalTimeSpentMillis: The total time taken for downloading, parsing, and inserting data from CSV files, excluding the request queue time.
- numThreads: The number of threads utilized for downloading and inserting data. This correlates with the provided concurrency parameter input, reflecting any caps applied.
- insertErrors: Errors faced during insertions, including parsing errors and Amazon S3 access issues. Error details are available in the CloudWatch logs. Refer to the <u>Troubleshooting</u> section of this document to understand troubleshooting insertErrors. Concurrent modification errors may also cause insert errors in batch loads attempting to modify a vertex property value being concurrently changed by another request.
- throughputRecordsPerSec: The total throughput in records per second.
- loadId: The loadId for searching errors and load summary. All batch information is published to CloudWatch logs under /aws/neptune/import-task-logs/<graph-id>/<load-id>.

i Note

Around 2.5Gb of Amazon S3 files can be loaded in a single request on 128 m-NCU. Larger sized datasets could run into out of memory errors. To workaround that, the Amazon S3 files can be split across multiple serial batch load requests. The source argument takes a prefix, so files can be partitioned across requests by including prefixes of file names. The limit scales linearly based on m-NCUs, so for example 5Gb of Amazon S3 files can be

loaded in a single request on 256 m-NCU. Also, if the dataset contains larger string values for example, then larger volumes of data can also be ingested in a single request, since they would generate fewer number of graph elements per byte of dataset. It is recommended to run tests with your data to determine the exact details for this process.

<u> Important</u>

Duplicate edges get created if the same edge file content is loaded more than once. This could happen if, for example:

- 1. The same Amazon S3 source or file is accidentally included for load in more than one request that succeeded.
- 2. The edge data is first loaded with failOnError set to false and runs into partial errors, and the errors are fixed and the entire dataset is reloaded. All of the edges that were successfully inserted on the first request would get duplicated after the second request.

Bulk import data into a graph

The task system in Neptune Analytics provides a powerful and flexible way to bulk import data into your graph. The import task is specifically designed to handle large-scale data ingestion from various data sources, such as CSV files.

To initiate a bulk data import, you would first create an import task by specifying the data source, the target graph, and any necessary configuration options. This can be done through the AWS console or programmatically via the API.

Throughout the import process, you can monitor the progress of the import task through the user interface or via API calls. Progress reports, and any potential errors or warnings will be accessible in your CloudWatch account, allowing for close monitoring and <u>troubleshooting</u> if needed.

Importing of data through Import Task is supported in two ways:

- During graph creation: Create a graph from Amazon S3, a Neptune cluster, or a snapshot
- On an existing empty graph: Bulk import data into an existing Neptune Analytics graph

Create a graph from Amazon S3, a Neptune cluster, or a snapshot

You can create a Neptune Analytics graph directly from Amazon S3 or from Neptune using the CreateGraphUsingImportTask API. This is recommended for importing large graphs from files in Amazon S3 (>50GB of data), importing from existing Neptune clusters, or importing from existing Neptune snapshots. This API automatically analyzes the data, provisions a new graph based on the analysis, and imports data as one atomic operation using maximum available resources.

1 Note

The graph is made available for querying only after the data loading is completed successfully.

If errors are encountered during the import process, Neptune Analytics will automatically roll back the provisioned resources, and perform the cleanup. No manual cleanup actions are needed. Error details are available in the CloudWatch logs. See <u>troubleshooting</u> for more details.

Topics

- <u>Creating a Neptune Analytics graph from Amazon S3</u>
- Creating a Neptune Analytics graph from Neptune cluster or snapshot

Creating a Neptune Analytics graph from Amazon S3

Neptune Analytics supports bulk importing of CSV data directly from Amazon S3 into a Neptune Analytics graph using the CreateGraphUsingImportTask API. The data format is the same as the CSV format used for <u>Batch load</u>. Refer to <u>Data format for loading from Amazon S3 into</u> <u>Neptune Analytics</u> for the details of the format. It is recommended that you try the batch load process with a subset of your data first to validate that it is correctly formatted. Once you have validated that your data files are fully compatible with Neptune Analytics, you can prepare your full dataset and perform the bulk import using the steps below.

A quick summary of steps needed to import a graph from Amazon S3:

 <u>Copy the data files to an Amazon S3 bucket</u>: Copy the data files to an Amazon Simple Storage Service bucket in the same region where you want the Neptune Analytics graph to be created. See <u>Data format for loading from Amazon S3 into Neptune Analytics</u> for the details of the format when loading data from Amazon S3 into Neptune Analytics.

- <u>Create your IAM role for Amazon S3 access</u>: Create an IAM role with read and list access to the bucket and a trust relationship that allows Neptune Analytics graphs to use your IAM role for importing.
- Use the CreateGraphUsingImportTask API to import from Amazon S3: Create a graph using the CreateGraphUsingImportTask API. This will generate a taskId for the operation.
- Use the GetImportTask API to get the details of the import task. The response will indicate the status of the task (ie. INITIALIZING, ANALYZING_DATA, IMPORTING etc.).
- Once the task has completed successfully, you will see a COMPLETED status for the import task and also the graphId for the newly created graph.
- Use the GetGraphs API to fetch all the details about your new graph, including the ARN, endpoint, etc.

Copy the data files to an Amazon S3 bucket

The Amazon S3 bucket must be in the same AWS region as the cluster that loads the data. You can use the following AWS CLI command to copy the files to the bucket.

```
aws s3 cp data-file-name s3://bucket-name/object-key-name
```

Note

In Amazon S3, an object key name is the entire path of a file, including the file name. In the command

aws s3 cp datafile.txt s3://examplebucket/mydirectory/datafile.txt

the object key name is mydirectory/datafile.txt

You can also use the AWS management console to upload files to the Amazon S3 bucket. Open the Amazon S3 <u>console</u>, and choose a bucket. In the upper-left corner, choose **Upload** to upload files.

Create your IAM role for Amazon S3 access

Create an IAM role with permissions to read and list the contents of your bucket. Add a trust relationship that allows Neptune Analytics to assume this role for doing the import task. You could do this using the AWS console, or through the CLI/SDK.

- Open the IAM console at <u>https://console.aws.amazon.com/iam/</u>. Choose Roles, and then choose Create Role.
- 2. Provie a role name.
- 3. Choose Amazon S3 as the AWS service.
- 4. In the **permissions** section, choose AmazonS3ReadOnlyAccess.

Note

This policy grants s3:Get* and s3:List* permissions to all buckets. Later steps restrict access to the role using the trust policy. The loader only requires s3:Get* and s3:List* permissions to the bucket you are loading from, so you can also restrict these permissions by the Amazon S3 resource. If your Amazon S3 bucket is encrypted, you need to add kms:Decrypt permissions as well. kms:Decrypt permission is needed for the exported data from Neptune Database

5. On the **Trust Relationships** tab, choose **Edit trust relationship**, and paste the following trust policy. Choose **Save** to save the trust relationship.

Your IAM role is now ready for import.

Use CreateGraphUsingImportTask API to import from Amazon S3

You can perform this operation from the Neptune console as well as from AWS CLI/SDK. For more information on different parameters, see <u>https://docs.aws.amazon.com/neptune-analytics/latest/</u>apiref/API_CreateGraphUsingImportTask.html

Via CLI/SDK

```
aws neptune-graph create-graph-using-import-task \
    --graph-name <name> \
    --format <format> \
    --source <s3 path> \
    --role-arn <role arn> \
    [--fail-on-error | --no-fail-on-error] \
    [--deletion-protection | --no-deletion-protection]
    [--public-connectivity | --no-public-connectivity]
    [--min-provisioned-memory]
    [--max-provisioned-memory]
    [--vector-search-configuration]
```

- Different Minimum and Maximum Provisioned Memory: When the --min-provisionedmemory and --max-provisioned-memory values are specified differently, the graph is created with the maximum provisioned memory specified by --max-provisioned-memory.
- **Single Provisioned Memory Value**: When only one of --min-provisioned-memory or -max-provisioned-memory is provided, the graph is created with the specified memory value.
- No Provisioned Memory Values: If neither --min-provisioned-memory nor --maxprovisioned-memory is provided, the graph is created with a default provisioned memory of 128 m-NCU (memory optimized Neptune Compute Units).

Example 1: Create a graph from Amazon S3, with no min/max provisioned memory.

```
aws neptune-graph create-graph-using-import-task \
    --graph-name 'graph-1' \
    --source "s3://bucket-name/gremlin-format-dataset/" \
    --role-arn "arn:aws:iam::<account-id>:role/<role-name>" \
    --format CSV
```

Example 2: Create a graph from Amazon S3, with min & max provisioned memory. A graph with m-NCU of 1024 is created.

```
aws neptune-graph create-graph-using-import-task \
    --graph-name 'graph-1' \
    --source "s3://bucket-name/gremlin-format-dataset/" \
    --role-arn "arn:aws:iam::<account-id>:role/<role-name>" \
    --format CSV
    --min-provisioned-memory 128 \
    --max-provisioned-memory 1024
```

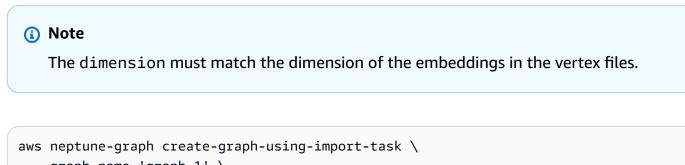
Example 3: Create a graph from Amazon S3, and not fail on parsing errors.

```
aws neptune-graph create-graph-using-import-task \
    --graph-name 'graph-1' \
    --source "s3://bucket-name/gremlin-format-dataset/" \
    --role-arn "arn:aws:iam::<account-id>:role/<role-name>" \
    --format CSV
    --no-fail-on-error
```

Example 4: Create a graph from Amazon S3, with 2 replicas.

```
aws neptune-graph create-graph-using-import-task \
    --graph-name 'graph-1' \
    --source "s3://bucket-name/gremlin-format-dataset/" \
    --role-arn "arn:aws:iam::<account-id>:role/<role-name>" \
    --format CSV
    --replica-count 2
```

Example 5: Create a graph from Amazon S3 with vector search index.



```
--graph-name 'graph-1' \
--source "s3://bucket-name/gremlin-format-dataset/" \
```

```
--role-arn "arn:aws:iam::<account-id>:role/<role-name>" \
```

```
--format CSV
--replica-count 2 \
--vector-search-configuration "{\"dimension\":768}"
```

Via Neptune console

1. Start the Create Graph wizard and choose **Create graph from existing source**.

Create graph ou can create a graph by specific the identifier, th ata into.	he source of the data, network configuration and the IAM roles to load the
Settings	
Graph Name The graph name.	
myGraph	
Data source Graph import task	
 Create empty graph 	Create Graph from existing source You can import data into a graph from an S3 bucket, Neptune cluster or a previous Analytics snapshot by creating a Graph Import task.

2. Choose type of source as Amazon S3, minimum and maximum provisioned memory, Amazon S3 path, and load role ARN.

Default C
▼ C
• C
▼ C
View 🖸 Browse S3

3. Choose the Network Settings and Replica counts.

Availability settings		
Replicas configuration		
Default replicas	 Use custom number of replicas You can specify the number of extra replicas you want 	
Network and security		
Public access		
Enable public access		
Your data is encrypted by default with a key that AWS owns your encryption settings. Info	and manages for you. To choose a different key, customize	
Customize encryption settings (advanced)		
Private endpoint		
Set up Private Endpoint You can access a graph from within a VPC by creating a private gra the traffic is never exposed to the internet. You can further attach flow logs on the graph endpoints. Standard VPC interface endpoint	security groups, tags or enable	

4. Create graph.

Creating a Neptune Analytics graph from Neptune cluster or snapshot

Neptune Analytics provides an easy way to bulk import data from an existing Neptune Database cluster or snapshot into a new Neptune Analytics graph, using the CreateGraphUsingImportTask API. Data from your source cluster or snapshot is bulk exported into an Amazon S3 bucket that you configure, analyzed to find the right memory configuration, and bulk imported into a new Neptune Analytics graph. You can check the progress of your bulk import at any time using the GetImportTask API as well.

A few things to consider while using this feature:

• You can only import from Neptune Database clusters and snapshots running on a version newer than or equal to 1.3.0.

- The exported data from your source Neptune Database cluster or snapshot will reside in your buckets only, and will be encrypted using a KMS key that you provide. The exported data is not directly consumable in any other way into Neptune outside of the CreateGraphUsingImportTask API. The exported data is not used after the lifetime of the request, and can be deleted by the user.
- You need to provide permissions to perform the export task on the Neptune Database cluster or snapshot, write to your Amazon S3 bucket, and use your KMS key while writing data.
- If your source is a Neptune Database cluster, a clone is taken from it and used for export. The original Neptune Database cluster will not be impacted. The cloned cluster is internally managed by the service and is deleted upon completion.
- If your source is a Neptune snapshot, a restored DBCluster is created from it, and used for export. The restored cluster is internally managed by the service and is deleted upon completion.
- This process is not recommended for small sized graphs. The export process is async, and works best for medium/large sized graphs with a size greater than 25GB. For smaller graphs, a better alternative is to use the <u>Neptune export</u> feature to generate CSV data directly from your source, upload that to Amazon S3 and then use the <u>Batch load</u> API instead.

A quick summary of steps to import from a Neptune cluster or a Neptune snapshot:

- 1. <u>Obtain the ARN of your Neptune cluster or snapshot</u>: This can be done from the AWS console or using the Neptune CLI.
- Create an IAM role with permissions to export from Neptune to Neptune Analytics: Create an IAM role that has permissions to perform an export of your Neptune graph, write to Amazon S3 and use your KMS key for writing data in Amazon S3.
- 3. Use the CreateGraphUsingImportTask API with source = NEPTUNE, and provide the ARN of your source, Amazon S3 path to export the data, KMS key to use for exporting data and additional arguments for your Neptune Analytics graph. This should return a task-id.
- 4. Use GetImportTask API to get the details of your task.

Obtain the ARN of your Neptune cluster or snapshot

Via the CLI:

```
# Obtaining the ARN of an existing DB Cluster
aws neptune describe-db-clusters \
    --db-cluster-identifier *<name> \
```

```
--query 'DBClusters[0].DBClusterArn'
```

```
# Obtaining the ARN of an existing DB Cluster Snapshot
aws neptune describe-db-cluster-snapshots \
        --db-cluster-snapshot-identifier <snapshot name> \
```

- --query 'DBClusterSnapshots[0].DBClusterSnapshotArn'

Via the AWS console. The ARN can be found on the cluster details page.

٩	Filter databases							
	☐ Identifier ▲	Status 🔻	Role 🛡	Engine version	▼ Region/AZ ▼	Size	▼ Cpu	
0	neptune-cluster	🕑 Available	Cluster	1.1.1.0	us-east-1	-	-	
0	<u>neptune-instance</u>	🔗 Available	Writer	1.1.1.0	us-east-1a	db.r5.xlarg	ge 🔲	5.37%
Сог	nectivity & Security Mon	nitoring Lo	ogs & Events	Configuration	Maintenance	Tags		
					ine version	1495		
DB	nfiguration Cluster				ine version	1495		
DB Reso	nfiguration Cluster ource ARN			Eng 1.1.	ine version			
DB Reso	nfiguration Cluster ource ARN arn:aws:rds:us-east-1:1234567			Eng 1.1. Crea	ine version 1.0			
DB Reso D Stat	nfiguration Cluster ource ARN arn:aws:rds:us-east-1:1234567			Eng 1.1. Crea Oct	ine version 1.0 ated time ober 9, 2023, 15:06			
DB Reso D Stat	nfiguration Cluster ource ARN arn:aws:rds:us-east-1:1234567			Eng 1.1. Oct Clus	ine version 1.0 ated time ober 9, 2023, 15:06 ster parameter group			
DB Reso D Stat	nfiguration Cluster ource ARN arn:aws:rds:us-east-1:1234567			Eng 1.1. Oct Clus	ine version 1.0 ated time ober 9, 2023, 15:06			

<u>Neptune</u> > <u>Snapshot</u> > neptune-cluster-snapshot

neptune-cluster-snapshot

napshot ARN	Status
arn:aws:rds:us-east-1:123456789000:cluster-snapshot:neptune-cluster	🕗 Available
napshot	
	Progress
napshot name	🕗 Completed
eptune-cluster-snapshot	
	License model
napshot type	neptune
anual	
	KSM Key ID
B cluster identifier	

Create an IAM role with permissions to export from Neptune to Neptune Analytics

- Open the IAM console at <u>https://console.aws.amazon.com/iam/</u>. Choose Roles, and then choose Create Role.
- 2. Provie a role name.
- 3. Choose Amazon S3 as the AWS service.
- 4. In the **permissions** section, choose AmazonS3FullAccess, NeptuneFullAccess, AmazonRDSFullAccess, AWSKeyManagementServicePowerUser.
- 5. On the **Trust Relationships** tab, choose **Edit trust relationship**, and paste the following trust policy. Choose **Save** to save the trust relationship.

```
{
    "Version": "2012-10-17",
    "Statement": [
        {
            "Effect": "Allow",
            "Principal": {
              "Service": [
              "service": [
              "export.rds.amazonaws.com",
              "neptune-graph.amazonaws.com"
]
```

```
},
"Action": "sts:AssumeRole"
}
]
}
```

Your IAM role is now ready for import.

Via CLI/SDK

For importing data via Neptune , the API expects additional import-options as defined here <u>NeptuneImportOptions</u>.

Example 1: Create a graph from a Neptune cluster.

```
aws neptune-graph create-graph-using-import-task \
    --graph-name <graph-name>
    --source arn:aws:rds:<region>:451235071234:cluster:nodelete-p8-test-snb-neptune-
gremlin-data-100 \
    --min-provisioned-memory 1024 \
    --max-provisioned-memory 1024 \
    --role-arn <role-arn> \
    --import-options '{"neptune": {
        "s3ExportKmsKeyId":"arn:aws:kms:<region>:<account>:key/<key>",
        "s3ExportPath": :"<s3 path for exported data>"
    }}'
```

Example 2: Create a graph from a Neptune cluster with the default vertex preserved.

```
aws neptune-graph create-graph-using-import-task \
    --graph-name <graph-name>
    --source arn:aws:rds:<region>:451235071234:cluster:nodelete-p8-test-snb-neptune-
gremlin-data-100 \
    --min-provisioned-memory 1024 \
    --max-provisioned-memory 1024 \
    --role-arn <role-arn> \
    --import-options '{"neptune": {
        "s3ExportKmsKeyId":"arn:aws:kms:<region>:<account>:key/<key>",
        "s3ExportPath": :"<s3 path for exported data>",
        "preserveDefaultVertexLabels" : true
    }}'
```

Example 3: Create a graph from Neptune cluster with the default edge Id preserved

```
aws neptune-graph create-graph-using-import-task \
    --graph-name <graph-name>
    --source arn:aws:rds:<region>:451235071234:cluster:nodelete-p8-test-snb-neptune-
gremlin-data-100 \
    --min-provisioned-memory 1024 \
    --max-provisioned-memory 1024 \
    --role-arn <role-arn> \
    --import-options '{"neptune": {
        "s3ExportKmsKeyId":"arn:aws:kms:<region>:<account>:key/<key>",
        "s3ExportPath": :"<s3 path for exported data>",
        "preserveEdgeIds" : true
    }}'
```

Bulk import data into an existing Neptune Analytics graph

Neptune Analytics now allows you to efficiently import large datasets into an already provisioned graph database using the StartImportTask API. This API facilitates the direct loading of data from an Amazon S3 bucket into an **empty** Neptune Analytics graph. This is designed for loading data into existing empty clusters.

Two common use cases for using this feature:

- 1. Bulk importing data multiple times without provisioning a new graph for each dataset. This helps during the development phase of a project where datasets are being converted into Neptune Analytics compatible load formats.
- 2. Use cases where graph provisioning privileges need to be separated from data operation privileges. For example, scenarios where graph provisioning needs to be done by only by the infrastructure team, and data loading and querying is done by the data engineering team.

For use cases where you want to create a new graph loaded with data, use the CreateGraphUsingImportTask API instead.

For incrementally loading data from Amazon S3 you can use the loader integration with the openCypher CALL clause. For more information see <u>Use Neptune load()</u>.

Prerequisites

• An empty Amazon Neptune Analytics graph.

- Data stored in an Amazon Amazon S3 bucket in the same region as the graph.
- An IAM role with permissions to access the Amazon S3 bucket. For more information, see <u>Create</u> your IAM role for Amazon S3 access.

Important considerations

- Data integrity: The StartImportTask API is designed to work with graphs that are empty. If your graph contains data, you can first reset the graph using the <u>reset-graph</u> API. If the Import task finds that the graph is not empty the operation will fail. This operation will delete all data from the graph, so ensure you have backups if necessary. You can use the <u>create-graph-snapshot</u> API to create snapshot of your existing graph.
- Atomic Operation: The data import is atomic, meaning it either completes fully or does not apply at all. If the import fails we would reset the state back to an empty graph.
- Format Support: Loading data supports the same data format as supported by create-graphusing-import-task and neptune.load() This API doesn't support importing data from Neptune.
- **Queries**: Queries will stop working while the import is in progress. You will get a Cannot execute any query until bulk import is complete error until the import finishes.

Steps for bulk importing data

1. Resetting the graph (if necessary):

If your graph is not empty, reset it using the following command:

```
aws neptune-graph reset-graph --graph-identifier <graph-id>
```

Note

This command will completely remove all existing data from your graph. It is recommended that you take a graph snapshot before performing this action.

2. Start the import task:

To load data into your Neptune graph, use the start-import-task command as follows:

```
aws neptune-graph start-import-task \setminus
```

```
--graph-identifier <graph-id> \
--source <s3-path-to-data> \
--format <data-format> \
--role-arn <IAM-role-ARN> \
[--fail-on-error | --no-fail-on-error]
```

- graph-identifier: The unique identifier of your Neptune graph.
- source: An Amazon S3 URI prefix. All object names with matching prefixes are loaded. See Neptune loader request parameters for Amazon S3 URI prefix examples.
- format: The data format of the Amazon S3 data to be loaded, either csv or openCypher. For more information, see Data formats.
- role-arn: The ARN of the IAM role that Neptune Analytics can assume to access your Amazon S3 data.
- (--no-)fail-on-error: (Optional) Stops the import process early if an error occurs. By default, the system attempts to stop at the first error.

Checking the details and progress of an import task

You can use the <u>GetImportTask</u> API to track the progress and the status of your import task.

```
aws neptune-graph get-import-task --task-id <task-id>
```

An Import task can be in the following state:

- **INITIALIZING**: The task is preparing for import, including provisioning a graph when using the CreateGraphUsingImportTask API.
- **ANALYZING_DATA**: The task is taking an initial pass through the dataset to determine the optimal configuration for the graph.
- IMPORTING: The data is being loaded into the graph.
- **EXPORTING**: Data is being exported from the Neptune cluster or snapshot. This is only applicable when performing an import task with a source of Neptune and through the CreateGraphUsingImportTask API.
- **ROLLING_BACK**: The import task encountered an error. Refer to the <u>troubleshooting</u> section to investigate the errors. The import task will be rolled back and eventually marked as FAILED.
- **SUCCEEDED**: Graph creation and data loading have succeeded. Use the get-graph API to view details of the final graph.

- REPROVISIONING: A temporary state while the graph is being reconfigured during the import task.
- FAILED: Graph creation or data loading has failed. Refer to the <u>troubleshooting</u> section to understand the reason for the failure.
- **CANCELLING**: The user has cancelled the import task, and cancellation is in progress.
- **CANCELLED**: The import task has been cancelled, and all resources have been released.

Additionally, import task can be used to track the progress of the load, error count and graph summary.

Canceling an import task

You can cancel a running import task by using the <u>CancelImportTask</u> API.

```
aws neptune-graph cancel-import-task \
--task-id <task-id>
```

The import task will will be canceled and all changes rolled back. The state of the import task will switch to CANCELING after cancel-import-task API is called and eventually the state will be CANCELED when rollback finishes. You can check the current state of your import task using the <u>GetImportTask</u> API.

```
aws neptune-graph get-import-task \
--task-id <task-id>
```

Troubleshooting

For both bulk load and batch load, all the errors and summary of the load is sent to the CloudWatch log group in your account. To view the logs go to CloudWatch, click log groups from the left column, then search for and click /aws/neptune/import-task-logs/.

- 1. Batch Load: The logs for each load is saved under /aws/neptune/import-task-logs/ <graph-id>/<load-id> CloudWatch log stream.
- 2. Bulk Load using Import Task: The logs are saved under /aws/neptune/import-task-logs/ <graph-id>/<task-id> CloudWatch log stream.

- S3_ACCESS_DENIED: The server does not have permissions to list or download the given file. Fix the permissions and retry. See <u>Create your IAM role for Amazon S3 access</u> for help setting up the Amazon S3 permissions.
- LARGE_STRING_ERROR: One or more strings exceeded the limit on the size of strings. This data cannot be inserted as is. Update the strings exceeding the limit and retry.
- **PARSING_ERROR**: Error parsing the given value(s). Correct the value(s) and retry. More information on different parsing errors is provided in this section.
- **OUT_OF_MEMORY**: No more data can be loaded in the current m-NCU. If encountered during import task, set a higher m-NCU and retry. If encountered during batch load, scale the number of m-NCU and retry the batch load.
- **PARTITION_FULL_ERROR**: No more data can be loaded in the internal server configuration. If encountered during import task, the import workflow would change the server configuration and retry. If encountered during batch load, reach out to the AWS service team to unblock loading of new data.

Error template	Solution
<pre>Invalid data type encountered for header val:badtype when parsing line [:ID,firs tName:String,val:badtype,:L ABEL] .</pre>	Incorrect Datatype provided. Check the documentation for supported data types. See <u>Data formats</u> for more information.
Multi-valued columns are not supported firstName:String[] when parsing line [:ID, firstName:String[], val :String,:LABEL] .	The opencypher format does not support multivalued user defined properties. Try using the csv format to insert multivalued vertex properties, or remove multivalued properties.
Bad header for a file in 'OPEN_CYPHER ' format, could not determine node or relations hip file, found system columns from 'csv' format when parsing line [~id, firs tName:String, val:int,:LABEL] .	Both the opencypher and csv format expect certain header columns to be present. Make sure you have entered them correctly. Check the <u>Data formats</u> documentation for required fields by format.

Common parsing errors and solutions

Error template	Solution
Bad header for a file in 'OPEN_CYPHER ' format, could not determine node or relations hip file.	The header of the files does not have the required system columns. Check the <u>Data</u> <u>formats</u> for required fields by format.
Relationship file in 'OPEN_CYPHER ' format should contain both :START_ID and :END_ID columns when parsing line [:START_ID,firstName:String] .	The header of the edge files does not have all the required system columns. Check the <u>Data</u> <u>formats</u> for required fields by format.
<pre>Invalid data type. Found system columns from 'OPEN_CYPHER ' format : ID when parsing line [:ID,firstName:String,val:I nt,~label] .</pre>	The opencypher and csv formats have different system column names, and they begin with : and ~ respectively. User defined properties cannot begin with those reserved prefixes in the respective formats. Confirm the format name and system column names, or update user defined properties to not use reserved prefixes.
Named column name is not present for header field :BLAH when parsing line [:ID,:BLA H,firstName:String] .	The opencypher and csv formats have different system column names, and they begin with : and ~ respectively. User defined properties cannot begin with those reserved prefixes in the respective formats. Confirm the format name and system column names, or update user defined properties to not use reserved prefixes.
System column other than ID cannot be stored as a property: <columnheader>.</columnheader>	The opencypher and csv formats have different system column names, and they begin with : and ~ respectively. User defined properties cannot begin with those reserved prefixes in the respective formats. Confirm the format name and system column names, or update user defined properties to not use reserved prefixes.

Error template	Solution
<pre>Duplicate user column firstName when parsing line [:ID,:LABEL, firstName :String, firstName:String] .</pre>	The file contains duplicate user defined property column names in the header. Remove all of the duplicate columns.
<pre>Duplicate system column :ID found when parsing line [:ID,:ID,firstName :String,:LABEL] .</pre>	The file contains duplicate system column names in the header. Remove all of the duplicate columns.
<pre>Invalid column name provided for loading embeddings: [abcd] for filename: someFilen ame. Embedding column name must be the same as their corresponding vector index name when parsing line [:ID,firs tName:String,abcd:Vector,:L ABEL] in [filename] .</pre>	An incorrect name is used for the vector embeddings.
"date" type is curretly not supported. "datetime" may be an alternative type.	Use datetime as the field type as date type suppoorted yet in Neptune Analytics.
Headers must be non-empty.	Headers need to be non empty. If the file has an empty line in the beginning, remove the empty line.
Failure encounted while parsing the csv file.	Likely reason is the number of columns in the row doesn't match the number of columns provided in the header. If you dont have a value for a column, provide an empty value. For example: 123, vertex, , , .
Could not process value of type:http:// www.w3.org/2001/XMLSchema#int for value: a when parsing line [v1,v1968 3,con,a] in [file].	There is a mismatch between the type of the value provided for that column in the row and the type specified in the header. In this specific case the column header is annotated with integer type but a is not parseable as an integer.

Error template	Solution
Could not load vector embedding: [a,bc]. Check the dimensionality for this vector.	The size of the vector does not match the dimension defined in the vector search configuration for the graph.
Could not load vector embedding: [a, NaN]. Check the value for this vector.	Float and double values in scientific notation are currently not supported. Also Infinity, -Infinity, INF, -INF, and NaN are not recognized.

Security in Neptune Analytics

Cloud security at AWS is the highest priority. As an AWS customer, you benefit from data centers and network architectures that are built to meet the requirements of the most security-sensitive organizations.

Security is a shared responsibility between AWS and you. The <u>shared responsibility model</u> 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. Third-party auditors regularly test and verify the effectiveness of our security as part of the <u>AWS</u>
 <u>Compliance Programs</u>. To learn about the compliance programs that apply to Neptune Analytics, see <u>AWS Services in Scope by Compliance Program</u>.
- 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 Neptune Analytics. The following topics show you how to configure Neptune Analytics to meet your security and compliance objectives. You also learn how to use other AWS services that help you to monitor and secure your Neptune Analytics resources.

Topics

- Data protection in Neptune Analytics
- Identity and access management for Neptune Analytics
- <u>Compliance validation for Neptune Analytics</u>
- <u>Resilience in Neptune Analytics</u>
- Infrastructure Security in Neptune Analytics
- <u>Cross-service confused deputy prevention</u>
- Using service-linked roles (SLRs) in Neptune Analytics

Data protection in Neptune Analytics

The AWS <u>shared responsibility model</u> applies to data protection in Neptune Analytics. 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. You are also responsible for the security configuration and management tasks for the AWS services that you use. For more information about data privacy, see the <u>Data Privacy FAQ</u>. For information about data protection in Europe, see the <u>AWS Shared Responsibility Model and</u> 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 <u>Federal Information Processing Standard (FIPS) 140-2</u>.

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 Neptune Analytics 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.

Identity and access management for Neptune Analytics

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 Neptune Analytics resources. IAM is an AWS service that you can use with no additional charge.

Topics

- Audience
- Authenticating with identities
- Managing access using policies
- How Neptune Analytics works with IAM
- Identity-based policy examples for Neptune Analytics
- <u>Troubleshooting Neptune Analytics identity and access</u>

Audience

How you use AWS Identity and Access Management (IAM) differs, depending on the work that you do in Neptune Analytics.

Service user – If you use the Neptune Analytics service to do your job, then your administrator provides you with the credentials and permissions that you need. As you use more Neptune Analytics 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 Neptune Analytics, see <u>Troubleshooting Neptune Analytics identity and access</u>.

Service administrator – If you're in charge of Neptune Analytics resources at your company, you probably have full access to Neptune Analytics. It's your job to determine which Neptune Analytics 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 Neptune Analytics, see How Neptune Analytics works with IAM.

IAM administrator – If you're an IAM administrator, you might want to learn details about how you can write policies to manage access to Neptune Analytics. To view example Neptune Analytics identity-based policies that you can use in IAM, see <u>Identity-based policy examples for Neptune Analytics</u>.

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 <u>How to sign in to your AWS</u> <u>account</u> 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 <u>Signing AWS API requests</u> 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 <u>Multi-factor authentication</u> in the *AWS IAM Identity Center User Guide* and <u>Using multi-factor authentication (MFA) in AWS</u> 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 <u>Tasks that require root</u> <u>user credentials</u> 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 <u>What is IAM Identity Center?</u> in the AWS IAM Identity Center User Guide.

IAM users and groups

An <u>IAM user</u> 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 <u>Rotate access keys regularly for use cases that require long-</u> term credentials in the *IAM User Guide*.

An <u>IAM group</u> 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 <u>When to create an IAM user</u> (instead of a role) in the *IAM User Guide*.

IAM roles

An <u>IAM role</u> 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 <u>switching roles</u>. 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 <u>Creating a role for a third-party Identity Provider</u> 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 <u>Permission sets</u> 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 <u>How IAM roles differ from resource-based policies</u> 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.
 - Forward access sessions (FAS) When you use an IAM user or role to perform actions in AWS, you are considered a principal. When you use some services, you might perform an action that then initiates another action in a different service. FAS uses the permissions of the principal calling an AWS service, combined with the requesting AWS service to make requests to downstream services. FAS requests are only made when a service receives a request that requires interactions with other AWS services or resources to complete. In this case, you must have permissions to perform both actions. For policy details when making FAS requests, see Forward access sessions.
 - Service role A service role is an <u>IAM role</u> 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 <u>Creating a role to delegate permissions to an AWS service</u> 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*.

To learn whether to use IAM roles or IAM users, see <u>When to create an IAM role (instead of a user)</u> 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 <u>Overview of JSON policies</u> 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 <u>Choosing between managed policies and inline policies</u> 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 <u>specify a principal</u> 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 <u>Access control list (ACL) overview</u> 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 <u>Permissions boundaries for IAM entities</u> 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 <u>How SCPs</u> 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 <u>Session policies</u> 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 <u>Policy evaluation logic</u> in the *IAM User Guide*.

How Neptune Analytics works with IAM

Before you use IAM to manage access to Neptune Analytics, learn what IAM features are available to use with Neptune Analytics.

IAM features you can use with Neptune Analytics

IAM feature	Neptune Analytics support
Identity-based policies	Yes
Resource-based policies	No
Policy actions	Yes
Policy resources	Yes
Policy condition keys	Yes
ACLs	No
ABAC (tags in policies)	Partial
Temporary credentials	Yes
Principal permissions	Yes
Service roles	Yes
Service-linked roles	Yes

To get a high-level view of how Neptune Analytics 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 Neptune Analytics

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 <u>Creating IAM policies</u> 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 <u>IAM JSON policy elements reference</u> in the *IAM User Guide*.

Identity-based policy examples for Neptune Analytics

To view examples of Neptune Analytics identity-based policies, see <u>Identity-based policy examples</u> for Neptune Analytics.

Resource-based policies within Neptune Analytics

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 <u>specify a principal</u> 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 Neptune Analytics

Supports policy actions

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 Neptune Analytics actions, see <u>Actions Defined by Neptune Analytics</u> in the Service Authorization Reference.

Policy actions in Neptune Analytics use the following prefix before the action:

To specify multiple actions in a single statement, separate them with commas.

"Action": [":action1", ":action2"]

To view examples of Neptune Analytics identity-based policies, see <u>Identity-based policy examples</u> <u>for Neptune Analytics</u>.

Policy resources for Neptune Analytics

Supports policy resources 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 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 <u>Amazon Resource Name (ARN)</u>. 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 Neptune Analytics resource types and their ARNs, see <u>Resources Defined by Neptune</u> <u>Analytics</u> in the *Service Authorization Reference*. To learn with which actions you can specify the ARN of each resource, see Actions Defined by Neptune Analytics.

To view examples of Neptune Analytics identity-based policies, see <u>Identity-based policy examples</u> for Neptune Analytics.

Policy condition keys for Neptune Analytics

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 <u>condition operators</u>, 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 Neptune Analytics condition keys, see <u>Condition Keys for Neptune Analytics</u> in the *Service Authorization Reference*. To learn with which actions and resources you can use a condition key, see Actions Defined by Neptune Analytics.

To view examples of Neptune Analytics identity-based policies, see <u>Identity-based policy examples</u> for Neptune Analytics.

ACLs in Neptune Analytics

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 Neptune Analytics

Supports ABAC (tags in policies)

Partial

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 <u>condition element</u> 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 <u>What is ABAC?</u> in the *IAM User Guide*. To view a tutorial with steps for setting up ABAC, see <u>Use attribute-based access control (ABAC)</u> in the *IAM User Guide*.

Using temporary credentials with Neptune Analytics

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 <u>AWS services that</u> 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 <u>Switching to a role (console)</u> 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 Neptune Analytics

Supports forward access sessions (FAS) Yes

When you use an IAM user or role to perform actions in AWS, you are considered a principal. When you use some services, you might perform an action that then initiates another action in a different service. FAS uses the permissions of the principal calling an AWS service, combined with the requesting AWS service to make requests to downstream services. FAS requests are only made when a service receives a request that requires interactions with other AWS services or resources to complete. In this case, you must have permissions to perform both actions. For policy details when making FAS requests, see <u>Forward access sessions</u>.

Service roles for Neptune Analytics

Supports service roles

A service role is an <u>IAM role</u> 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 <u>Creating a role to delegate permissions to an AWS service in the IAM User Guide</u>.

<u> M</u>arning

Changing the permissions for a service role might break Neptune Analytics functionality. Edit service roles only when Neptune Analytics provides guidance to do so.

Service-linked roles for Neptune Analytics

Supports service-linked roles Yes

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 Neptune Analytics service-linked roles, see <u>Using service-</u> linked roles (SLRs) in Neptune Analytics.

For details about creating or managing service-linked roles for other services, see <u>AWS services that</u> <u>work with IAM</u>. 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.

Identity-based policy examples for Neptune Analytics

By default, users and roles don't have permission to create or modify Neptune Analytics 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 Neptune Analytics, including the format of the ARNs for each of the resource types, see <u>Actions, Resources, and Condition Keys for Neptune</u> Analytics in the *Service Authorization Reference*.

Topics

- Policy best practices
- Using the Neptune Analytics console
- Allow users to view their own permissions

Policy best practices

Identity-based policies determine whether someone can create, access, or delete Neptune Analytics 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 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 <u>AWS managed policies</u> or <u>AWS</u> <u>managed policies for job functions</u> 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 <u>Policies and permissions in IAM</u> 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 <u>IAM JSON policy elements: Condition</u> 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 <u>IAM Access Analyzer policy validation</u> 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 <u>Configuring MFA-protected API access</u> in the IAM User Guide.

For more information about best practices in IAM, see <u>Security best practices in IAM</u> in the *IAM User Guide*.

Using the Neptune Analytics console

To access the Neptune Analytics console, you must have a minimum set of permissions. These permissions must allow you to list and view details about the Neptune Analytics 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.

To ensure that users and roles can still use the Neptune Analytics console, also attach the Neptune Analytics *ConsoleAccess* or *ReadOnly* AWS managed policy to the entities. For more information, see Adding permissions to a user in the *IAM User Guide*.

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.



Troubleshooting Neptune Analytics identity and access

Use the following information to help you diagnose and fix common issues that you might encounter when working with Neptune Analytics and IAM.

Topics

- I am not authorized to perform an action in Neptune Analytics
- I am not authorized to perform iam:PassRole
- I want to allow people outside of my AWS account to access my Neptune Analytics resources

I am not authorized to perform an action in Neptune Analytics

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 :*GetWidget* permissions.

```
User: arn:aws:iam::123456789012:user/mateojackson is not authorized to
  perform: :GetWidget on resource: my-example-widget
```

In this case, the policy for the mateojackson user must be updated to allow access to the *myexample-widget* resource by using the :*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 Neptune Analytics.

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 Neptune Analytics. 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 Neptune Analytics 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 Neptune Analytics supports these features, see <u>How Neptune Analytics works</u> with IAM.
- 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 <u>Providing</u> access to AWS accounts owned by third parties in the *IAM User Guide*.
- To learn how to provide access through identity federation, see <u>Providing access to externally</u> 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.

Compliance validation for Neptune Analytics

To learn whether an AWS service is within the scope of specific compliance programs, see <u>AWS</u> <u>services in Scope by Compliance Program</u> and choose the compliance program that you are interested in. For general information, see AWS Compliance Programs.

You can download third-party audit reports using AWS Artifact. For more information, see <u>Downloading Reports in AWS Artifact</u>.

Your compliance responsibility when using AWS services 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:

- <u>Security and Compliance Quick Start Guides</u> These deployment guides discuss architectural considerations and provide steps for deploying baseline environments on AWS that are security and compliance focused.
- <u>Architecting for HIPAA Security and Compliance on Amazon Web Services</u> This whitepaper describes how companies can use AWS to create HIPAA-eligible applications.

í) Note

Not all AWS services are HIPAA eligible. For more information, see the <u>HIPAA Eligible</u> <u>Services Reference</u>.

- <u>AWS Compliance Resources</u> This collection of workbooks and guides might apply to your industry and location.
- <u>AWS Customer Compliance Guides</u> Understand the shared responsibility model through the lens of compliance. The guides summarize the best practices for securing AWS services and map the guidance to security controls across multiple frameworks (including National Institute of Standards and Technology (NIST), Payment Card Industry Security Standards Council (PCI), and International Organization for Standardization (ISO)).
- <u>Evaluating Resources with Rules</u> in the *AWS Config Developer Guide* The AWS Config service assesses how well your resource configurations comply with internal practices, industry guidelines, and regulations.
- <u>AWS Security Hub</u> This AWS service provides a comprehensive view of your security state within AWS. Security Hub uses security controls to evaluate your AWS resources and to check your compliance against security industry standards and best practices. For a list of supported services and controls, see <u>Security Hub controls reference</u>.
- <u>AWS Audit Manager</u> This AWS service helps you continuously audit your AWS usage to simplify how you manage risk and compliance with regulations and industry standards.

Resilience in Neptune Analytics

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.

For more information about AWS Regions and Availability Zones, see AWS Global Infrastructure.

In addition to the AWS global infrastructure, Neptune Analytics offers several features to help support your data resiliency and backup needs.

Infrastructure Security in Neptune Analytics

As a managed service, Neptune Analytics is protected by the AWS global network security procedures that are described in the <u>Amazon Web Services: Overview of Security Processes</u> whitepaper.

You use AWS published API calls to access Neptune Analytics through the network. Clients must support Transport Layer Security (TLS) 1.0 or later. We recommend TLS 1.2 or later. Clients must also support 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 <u>AWS Security Token Service</u> (AWS STS) to generate temporary security credentials to sign requests.

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 <u>aws:SourceArn</u> and <u>aws:SourceAccount</u> global condition context keys in resource policies to limit the permissions that ServiceNameLongEntity gives another service to the resource. Use aws:SourceArn if you want only one resource to be associated with the cross-service access. Use aws:SourceAccount if you want to allow any resource in that account to be associated with the cross-service use.

The most effective way to protect against the confused deputy problem is to use the aws:SourceArn global condition context key with the full ARN of the resource. If you don't know the full ARN of the resource or if you are specifying multiple resources, use the aws:SourceArn global context condition key with wildcard characters (*) for the unknown portions of the ARN. For example, arn:aws:servicename:*:123456789012:*.

If the aws: SourceArn value does not contain the account ID, such as an Amazon S3 bucket ARN, you must use both global condition context keys to limit permissions.

The value of aws:SourceArn must be ResourceDescription.

The following example shows how you can use the aws:SourceArn and aws:SourceAccount global condition context keys in ServiceNameEntity to prevent the confused deputy problem.

```
{
  "Version": "2012-10-17",
  "Statement": {
    "Sid": "ConfusedDeputyPreventionExamplePolicy",
    "Effect": "Allow",
    "Principal": {
      "Service": "servicename.amazonaws.com"
    },
    "Action": "servicename:ActionName",
    "Resource": [
      "arn:aws:servicename:::ResourceName/*"
    ],
    "Condition": {
      "ArnLike": {
        "aws:SourceArn": "arn:aws:servicename:*:123456789012:*"
      },
      "StringEquals": {
        "aws:SourceAccount": "123456789012"
      }
    }
  }
}
```

Using service-linked roles (SLRs) in Neptune Analytics

Neptune Analytics graphs use AWS Identity and Access Management (IAM) <u>service-linked roles</u>. A service-linked role is a unique type of IAM role that is linked directly to Neptune Analytics graphs. Service-linked roles are predefined by Neptune Analytics graphs and include all the permissions that the service requires to call other AWS services on your behalf.

A service-linked role makes using Neptune Analytics graphs easier because you don't have to add the necessary permissions manually. Neptune Analytics defines the permissions in its servicelinked roles, and unless defined otherwise, only Neptune Analytics graphs can assume its roles. The defined permissions include the trust policy and the permissions policy, and that permissions policy cannot be attached to any other IAM entity. You can delete the roles only after first deleting their related resources. This protects your Neptune Analytics graph resources because you can't inadvertently remove the permissions to access the resources.

For information about other services that support service-linked roles, see <u>AWS services that work</u> <u>with IAM</u> and look for the services that are marked with **Yes** in the **Service-Linked Role** column. Choose a **Yes** with a link to view the service-linked role documentation for that service.

Service-linked role permissions for Neptune Analytics Graphs

Neptune Analytics graphs uses the service-linked role named AWSServiceRoleForNeptuneGraph to allow them to call AWS services on behalf of your DB clusters.

This service-linked role has an IAM managed permissions policy attached to it named <u>AWSServiceRoleForNeptuneGraphPolicy</u> that grants it permissions to operate in your account. See <u>AWS managed policies for Amazon Neptune</u>. This policy provides read-only access to all Amazon Neptune Analytics resources along with read-only permissions for dependent services, as follows:

```
{
  "Version": "2012-10-17",
  "Statement": [
    {
      "Sid": "GraphMetrics",
      "Effect": "Allow",
      "Action": [
        "cloudwatch:PutMetricData"
      ],
      "Resource": "*",
      "Condition": {
        "StringEquals": {
          "cloudwatch:namespace": [
            "AWS/Neptune",
            "AWS/Usage"
          ]
        }
      }
    },
    {
      "Sid": "GraphLogGroup",
      "Effect": "Allow",
```

```
"Action": [
        "logs:CreateLogGroup"
      ],
      "Resource": [
        "arn:aws:logs:*:*:log-group:/aws/neptune/*"
      ],
      "Condition": {
        "StringEquals": {
          "aws:ResourceAccount": "${aws:PrincipalAccount}"
        }
      }
    },
    {
      "Sid": "GraphLogEvents",
      "Effect": "Allow",
      "Action": [
        "logs:CreateLogStream",
        "logs:PutLogEvents",
        "logs:DescribeLogStreams"
      ],
      "Resource": [
        "arn:aws:logs:*:*:log-group:/aws/neptune/*:log-stream:*"
      ],
      "Condition": {
        "StringEquals": {
          "aws:ResourceAccount": "${aws:PrincipalAccount}"
        }
      }
    }
  ]
}
```

i Note

To allow an IAM entity such as a user, group, or role to be able to create, edit, or delete a service-linked role, you must set the appropriate permissions, like this:

```
{
    "Action": "iam:CreateServiceLinkedRole",
    "Effect": "Allow",
    "Resource": "arn:aws:iam::*:role/aws-service-role/neptune-graph.amazonaws.com/
AWSServiceRoleForNeptuneGraph",
    "Condition": {
```

}

```
"StringLike": {
"iam:AWSServiceName":"neptune-graph.amazonaws.com"
}
```

If those permissions have not been set, or have not yet propagated, you may receive the following error message when you try to create a service-linked role:

```
Unable to create the resource. Verify that you have permission
to create service linked role. Otherwise wait and try again later.
```

For more information, see <u>Service-linked role permissions</u> in the IAM User Guide.

Creating a service-linked role for Neptune Analytics

You don't have to create a service-linked role manually for Neptune Analytics. When you create a graph, Neptune Analytics automatically creates the service-linked role for you.

Editing a service-linked role for Neptune Analytics

Neptune Analytics doesn't allow you to edit the AWSServiceRoleForNeptuneGraph servicelinked role. After you create a service-linked role, you cannot change the name of the role because various entities might reference it. However, you can edit the description of the role using IAM. For more information, see Editing a service-linked role in the IAM User Guide.

Deleting a service-linked role

If you no longer need to use a feature or service that requires a service-linked role, it's best to delete that role so you don't have an unused entity that is not actively monitored or maintained.

However, before you can delete the service-linked role, you must first confirm that the role has no active sessions, and remove any resources that it uses.

To check whether a service-linked role has an active session in the IAM console

- Sign in to the AWS Management Console and open the IAM console at <u>https://</u> <u>console.aws.amazon.com/iam/</u>.
- 2. In the navigation pane of the IAM console, choose *Roles*. Then choose the name (not the check box) of the AWSServiceRoleForNeptuneGraph role.

3. On the **Summary** page for the chosen role, choose the **Access Advisor** tab.

🚯 Note

If you are unsure whether Neptune Analytics is using the AWSServiceRoleForNeptuneGraph role, you can try to delete the role. If the service is using the role, then the deletion fails and you can view the AWS Regions where the role is being used. If the role is being used, then you must wait for the session to end before you can delete the role. You cannot revoke the session for a service-linked role.

To delete your clusters so that you can delete AWSServiceRoleForNeptuneGraph

- 1. Open the Neptune console at https://console.aws.amazon.com/neptune/.
- 2. In the navigation pane, choose **Graphs**.
- 3. Choose a cluster that you want to delete.
- 4. For Actions, choose Delete.
- 5. If you are prompted to **Create final Snapshot?**, choose **Yes** or **No**. If you choose **Yes** enter the name of your final snapshot for **Final snapshot name**.
- 6. Choose Delete.

You can use the IAM console, the IAM CLI, or the IAM API to delete the AWSServiceRoleForNeptuneGraph service-linked role. For more information, see <u>Deleting a</u> <u>service-linked role</u> in the <u>IAM User Guide</u>.

Querying Neptune Analytics

Neptune Analytics currently supports only the openCypher query language to access a graph. openCypher is a declarative query language for property graphs that was originally developed by Neo4j, then open-sourced in 2015, and contributed to the <u>openCypher</u> project under an Apache 2 open-source license. Its syntax is documented in the <u>openCypher</u> spec.

Topics

- Query APIs
- Query plan cache
- Query explain
- Statistics
- Exceptions
- Neptune Analytics openCypher data model
- Neptune Analytics openCypher specification compliance
- Transaction isolation levels in Neptune Analytics

Query APIs

The Neptune Analytics data API provides support for data operations including query execution, query status checking, query cancellation, and graph summarizing via the HTTPS endpoint, the AWS CLI, and the SDK.

Topics

- ExecuteQuery
- ListQueries
- GetQuery
- CancelQuery
- GraphSummary
- IAM role mappings

ExecuteQuery

ExecuteQuery runs queries against a Neptune Analytics graph. Supported language: openCypher.

ExecuteQuery inputs

• graph-identifier (required)

Type: String

The identifier representing a graph.

• region (required)

Type: String

The region where the graph is present.

• query-string (required)

Type: String

Default: none

- A string representing a query.
- language (required)

Type: Enum

Default: none

The query language the query is written in. Currently, only OPEN_CYPHER is supported.

• parameters (optional)

Туре: Мар

A map from String to String where the key is the parameter name and the value is the parameter value.

• plan-cache (optional)

Type: Enum

Query plan cache is a feature that saves the query plan and reuses it on successive executions of the same query, reducing query latency. Query plan cache works for both read-only and mutation queries. The plan cache is an LRU cache with a five minute TTL and a capacity of 1000. It supports the following values:

- AUTO: The engine will automatically decide to cache the query plan. If the query is parameterized and the runtime is shorter than 100ms, the query plan is automatically cached.
- ENABLED: The query plan is cached regardless of the query runtime. The plan cache uses the query string as the key, this means that if a query is slightly different (i.e. different constants), it will not be able to reuse the plan cache of similar queries.
- DISABLED: The query plan cache is not used.

For more information on the query plan cache, see <u>Query plan cache</u>.

• explain-mode (optional)

Type: Enum

The explain mode parameters allow getting a query explain instead of the actual query results. A query explain can be used to gather insights about the query execution such as planning decisions, time spent on each operator, number of records flowing etc. If this parameter is not set the query is executed normally and the result is returned. The acceptable values for query explain are:

- STATIC: Returns a query explain without executing the query. This can give an estimate on what the query plan looks like without actually executing the query. The static query plan may differ from the actual query plan. Actual queries may make planning decisions based on runtime statistics, which may not be considered when fetching a static query plan. A static query plan is useful when it is necessary to observe a plan for a query that either does not complete or runs for too long.
- DETAILS: Returns a detailed query plan that shows what the running query did. This includes
 information such as operators runtime, number of records flowing through the plan, runtime
 planning decisions and more. If a query does not succeed in NONE mode, it will not succeed in
 DETAILS mode either. In this instance, you would want to use STATIC mode.

For more information on query explain and its output, see <u>Query explain</u>.

• query-timeout-milliseconds (optional)

Type: Enum

If specified, provides an upper bound to the query run time. This parameter will override the graph default timeout (30 minutes). Neptune Analytics graph have a maximum query runtime of 60 minutes. If the specified timeout is greater than the maximum query runtime, the query will only run for the maximum query runtime.

 Using the default settings, any CLI or SDK request will timeout in 60 seconds and attempt a retry. For the cases where you are running queries that can take longer than 60 seconds, it is recommended to set the CLI/SDK timeout to 0 (no timeout), or a much larger value to avoid unnecessary retries.

It is also recommended to set MAX_ATTEMPTS for CLI/SDK to 1 for execute_query to avoid any retries by CLI/SDK.

For the Boto client, set the read_timeout to None, and the total_max_attempts to 1.

For the CLI, set the --cli-read-timeout parameter to 0 for no timeout, and set the environment variable AWS_MAX_ATTEMPTS to 1 to prevent retries.

export AWS_MAX_ATTEMPTS=1

```
aws neptune-graph execute-query \
--graph-identifier <graph-id> \
--region <region> \
--query-string "MATCH (p:Person)-[r:KNOWS]->(p1) RETURN *;" \
--cli-read-timeout 0
--language open_cypher /tmp/out.txt
```

ExecuteQuery examples

AWS CLI

Sample query

```
aws neptune-graph execute-query \setminus
--graph-identifier <graph-id> \
--region <region> \
--query-string "MATCH (p:Person)-[r:KNOWS]->(p1) RETURN *;" \
--language open_cypher \
/tmp/out.txt
# Sample query that prints directly to the console.
aws neptune-graph execute-query \setminus
--graph-identifier <graph-id> \
--region <region> \
--query-string "MATCH (p:Person)-[r:KNOWS]->(p1) RETURN *;" \
--language open_cypher \
/dev/stdout
# parameters supported
query-string [REQUIRED] : String
language [REQUIRED] : open_cypher
explain-mode [OPTIONAL] : static | details
query-timeout-milliseconds [OPTIONAL] : Integer
plan-cache [OPTIONAL] : enabled | disabled | auto
parameters [OPTIONAL] : Map
```

AWSCURL

```
# Sample query
awscurl -X POST "https://<graph-id>.data.us-east-1-alpha.p8.neptune.aws.dev/queries"
   \
   -H "Content-Type: application/x-www-form-urlencoded" \
   --region <region> \
   --service neptune-graph \
   -d "query=MATCH (p:Person)-[r:KNOWS]->(p1) RETURN *;"
```

ExecuteQuery output

```
{
    "results": [{
        "p": {
            "~id": "fa1ef9b0-fa32-4b37-8051-78f2bf0e0d63",
            "~entityType": "node",
            "~labels": ["Person"],
            "~properties": {
            "
```

```
"name": "Simone"
        }
      },
      "p1": {
        "~id": "edaded10-b22b-4818-a22e-ddebfcf37acb",
        "~entityType": "node",
        "~labels": ["Person"],
        "~properties": {
          "name": "Mirro"
        }
      },
      "r": {
        "~id": "neptune_reserved_1_1154145192329347075",
        "~entityType": "relationship",
        "~start": "fa1ef9b0-fa32-4b37-8051-78f2bf0e0d63",
        "~end": "edaded10-b22b-4818-a22e-ddebfcf37acb",
        "~type": "KNOWS",
        "~properties": {}
      }
    }]
}
```

Parameterized queries

Neptune Analytics supports parameterized openCypher queries. This allows you to use the same query structure multiple times with different arguments. Since the query structure doesn't change, Neptune Analytics tries to cache the plan for these parameterized queries that run in less than 100 milliseconds.

The following is an example of using a parameterized query with the Neptune openCypher HTTPS endpoint. The query is:

```
MATCH (n {name: $name, age: $age})
RETURN n
```

The parameters are definied as follows:

```
parameters={"name": "john", "age": 20}
```

AWS CLI

Sample query

```
aws neptune-graph execute-query \
--graph-identifier <graph-id> \
--region <region> \
--query-string "MATCH (n {name: \$name, age: \$age}) RETURN n" \
--parameters "{\"name\": \"john\", \"age\": 20}"
--language open_cypher /tmp/out.txt
```

AWSCURL

```
# Sample query
awscurl -X POST "https://[graph-id].<endpoint>/queries" \
-H "Content-Type: application/x-www-form-urlencoded" \
--region <region> \
--service neptune-graph \
-d "query=MATCH (n {name: \$name, age: \$age}) RETURN n;&parameters={\"name\":
    \"john\", \"age\": 20}"
```

ListQueries

ListQueries API fetches the list of running/waiting/cancelling queries on the graph.

ListQueries syntax

```
aws neptune-graph list-queries \
    --graph-identifier <graph-id> \
    --region <region> \
    --max-results <result_count>
    --state [all | running | waiting | cancelling]
```

ListQueries inputs

• graph-identifier (required)

Type: String

Identifier representing your graph.

region (required)

Type: String

Region where the graph is present.

max-results (required)

Type: Integer

The maximum number of results to be fetched by the API.

state (optional)

Type: String

Supported values: all | running | waiting | cancelling

If state parameter is not specified, the API fetches all types.

ListQueries outputs

```
# Sample Response
{
    "queries": [
        {
            "id": "130ab841-8b4b-46c3-afbe-af00274c7fd9",
                "queryString": "MATCH p=(n)-[*]-(m) RETURN p;",
                "waited": 0,
                "elapsed": 1686,
                "state": "RUNNING"
        }
    ]
}
```

The output contains a list of query objects, each containing:

- id: String representing the unique identifier of the query.
- queryString: String The actual query text. The queryString may be truncated if the actual query string is too long.
- waited: Integer The time in milliseconds for which the query has waited in the waiting queue before being picked up by a worker thread.
- elapsed: Integer The time in milliseconds representing the running time of the query.
- state: Current state of the query (running | waiting | cancelling).

The default list order is queries that are running, followed by waiting and cancelling.

ListQueries Examples

AWS CLI

```
aws neptune-graph list-queries \
    --graph-identifier <graph-id> \
    --region us-east-1 \
    --max-results 200
    --state waiting
```

AWSCURL

GetQuery

The GetQuery API can be used to get the status of a specific query request.

GetQuery inputs

• graph-identifier (required)

Type: String

The identifier representing a graph.

• region (required)

Type: String

The region where the graph is present.

• query-id (required)

Type: String

The id of the query request for which you want to get information.

GetQuery outputs

- id: The same id used in this request.
- queryString: Non-truncated query string associated to this query-id.
- waited: Time in milliseconds this query request had to wait to be executed.
- elapsed: Time in milliseconds the query spent while in execution.
- state: Current state of the query running | waiting | cancelling.

```
{
    "id" : "d6873456-40a7-44d7-be5c-46b4acfdc171",
    "queryString" : "UNWIND range(1,100000) AS i MATCH (n) RETURN i, n",
    "waited" : 1,
    "elapsed" : 8645,
    "state" : "RUNNING"
}
```

GetQuery examples

AWS CLI

```
aws neptune-graph get-query \
    --graph-identifier <graph-id> \
    --region <region> \
    --query-id <query-id>
```

AWSCURL

```
awscurl -X GET "https://<graph-id>.<endpoint>/queries/<query-id>" \
    -H "Content-Type: application/x-www-form-urlencoded" \
    --region us-east-1 \
    --service neptune-graph
```

CancelQuery

CancelQuery cancels a specific query request.

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CancelQuery inputs

• graph-identifier (required)

Type: String

The identifier representing a graph.

• region (required)

Type: String

The region where the graph is present.

• query-id (required)

```
Type: String
```

The id of the query request for which you want to cancel.

CancelQuery outputs

CancelQuery does not have any output.

CancelQuery examples

AWS CLI

```
aws neptune-graph cancel-query \
    --graph-identifier <graph-id> \
    --region <region> \
    --query-id <query-id>
```

AWSCURL

```
awscurl -X DELETE "https://<graph-id>.<endpoint>/queries/<query-id>" --region us-
east-1 --service neptune-graph
```

GraphSummary

You can use the GetGraphSummary API to quickly gain a high-level understanding of your graph data, size and content. In a graph application, this API can be used to improve the search results by providing discovered node or edge labels as part of the search.

The GetGraphSummary API retrieves a read-only list of node and edge labels and property keys, along with counts of nodes, edges, and properties. The API also accepts an optional parameter named mode, which can take one of two values, namely basic (the default) and detailed. The detailed graph summary response contains two additional fields, nodeStructures and edgeStructures.

GetGraphSummary inputs

GetGraphSummary accepts two inputs:

- graph-identifier (required) The unique identifier of the graph.
- mode (optional) Can be basic or detailed.

GetGraphSummary outputs

The response contains the following fields:

- version The version of this graph summary response.
- lastStatisticsComputationTime The timestamp, in ISO 8601 format, of the time at which Neptune Analytics last computed statistics.
- graphSummary
 - numNodes The number of nodes in the graph.
 - numEdges The number of edges in the graph.
 - numNodeLabels The number of distinct node labels in the graph.
 - numEdgeLabels The number of disctinct edge labels in the graph.
 - nodeLabels List of distinct node labels in the graph.
 - edgeLabels List of distinct edge labels in the graph.
 - numNodeProperties The number of distinct node properties in the graph.
 - numEdgeProperites The number of distinct edge properties in the graph.

- nodeProperties List of distinct node properties in the graph along with the count of nodes where each property is used.
- edgeProperties List of distinct edge properties in the graph along with the count of edges where each property is used.
- totalNodePropertyValues Total number of usages of all node properties.
- totalEdgePropertyValues Total number of usages of all edge properties.
- nodeStructures (only present for mode=detailed) Contains a list of node structures, each containing the following fields:
 - count Number of nodes that have this specific structure.
 - nodeProperties List of node properties present in this specific structure.
 - distinctOutgoingEdgeLabels List of distinct outgoing edge labels present in this specific structure.
- edgeStructures (only present for mode=detailed) Contains a list of edge structures each containing the following fields:
 - count Number of edges that have this specific structure.
 - edgeProperties List of edge properties present in this specific structure.

GetGraphSummary examples

AWS CLI

```
# Sample query
aws neptune-graph get-graph-summary \
--graph-identifier <graph-id> \
--region <region>
--mode detailed
# parmeters supported
mode [Optional] : basic | detailed
```

AWSCURL

```
# Sample query
awscurl "https://<graph-id>.<endpoint>/summary" \
-H "Content-Type: application/x-www-form-urlencoded" \
--region <region> \
```

--service neptune-graph

Sample output payload:

```
# this is the graph summary with "mode=detailed"
{
    "version": "v1",
    "lastStatisticsComputationTime": "2024-01-25T19:50:42+00:00",
    "graphSummary": {
        "numNodes": 3749,
        "numEdges": 57645,
        "numNodeLabels": 4,
        "numEdgeLabels": 2,
        "nodeLabels": [
            "continent",
            "country",
            "version",
            "airport"
        ],
        "edgeLabels": [
            "contains",
            "route"
        ],
        "numNodeProperties": 14,
        "numEdgeProperties": 1,
        "nodeProperties": [
            {
                 "code": 3749
            },
            {
                 "desc": 3749
            },
            {
                 "type": 3749
            },
            {
                 "city": 3504
            },
            {
                 "country": 3504
            },
             ſ
```

```
"elev": 3504
   },
    {
        "icao": 3504
    },
    {
        "lat": 3504
    },
    {
        "lon": 3504
    },
    {
        "longest": 3504
    },
    {
        "region": 3504
    },
    {
        "runways": 3504
    },
    {
        "author": 1
    },
    {
        "date": 1
    }
],
"edgeProperties": [
   {
        "dist": 50637
    }
],
"totalNodePropertyValues": 42785,
"totalEdgePropertyValues": 50637,
"nodeStructures": [
                       // will not be present with mode=basic
    {
        "count": 3475,
        "nodeProperties": [
            "city",
            "code",
            "country",
            "desc",
            "elev",
            "icao",
```

```
"lat",
        "lon",
        "longest",
        "region",
        "runways",
        "type"
    ],
    "distinctOutgoingEdgeLabels": [
        "route"
    ]
},
{
    "count": 238,
    "nodeProperties": [
        "code",
        "desc",
        "type"
    ],
    "distinctOutgoingEdgeLabels": [
        "contains"
    ]
},
{
    "count": 29,
    "nodeProperties": [
        "city",
        "code",
        "country",
        "desc",
        "elev",
        "icao",
        "lat",
        "lon",
        "longest",
        "region",
        "runways",
        "type"
    ],
    "distinctOutgoingEdgeLabels": []
},
{
    "count": 6,
    "nodeProperties": [
        "code",
```

```
"desc",
                     "type"
                 ],
                 "distinctOutgoingEdgeLabels": []
             },
             {
                 "count": 1,
                 "nodeProperties": [
                     "author",
                     "code",
                     "date",
                     "desc",
                     "type"
                 ],
                 "distinctOutgoingEdgeLabels": []
             }
        ],
        "edgeStructures": [
                                       //will not be present with mode=basic
             {
                 "count": 50637,
                 "edgeProperties": [
                     "dist"
                 ]
             }
        ]
    }
}
```

IAM role mappings

When you're calling Neptune Analytics API methods on a cluster, you require an IAM policy attached to the user or role making the calls that provides permissions for the actions you want to make. You set those permissions in the policy using corresponding IAM actions. You can also restrict the actions that can be taken using <u>IAM condition keys</u>.

Most IAM actions have the same name as the API methods that they correspond to, but some methods in the data API have different names, because some are shared by more than one method. The table below lists data methods and their corresponding IAM actions.

Data API operation name	IAM correspondences
ListQueries	Action: ListQueries
GetQuery	Action: GetQueryStatus
Cancel Query	Action: CancelQuery
GetGraphSummary	Action: GetGraphSummary
ExecuteQuery	Action: ReadDataViaQuery
	Action: WriteDataViaQuery
	Action: DeleteDataViaQuery

For more information, see Actions, resources and condition keys for Neptune Analytics.

Query plan cache

When a query is submitted to Neptune , the query string is parsed and translated into a query plan, which then gets optimized and executed by the engine. Often, the applications are backed by common query patterns that are instantiated with different values, and query plan cache would be optimal to reduce latency of those common query patterns. The query plan cache does this by storing a parameterized version of frequently used query plans (at most 1000 at any point), which gets reused and instantiated properly based on new parameter values provided, if any.

Why use the query plan cache?

Reusing the query plan can reduce the latency, as the later executions skip parsing and optimization steps.

Where can it be used?

Query plan cache can be used for all type of queries. By default, it automatically caches plan for low-latency parameterized queries, whose execution time is less than 100ms.

How to force enable/disable the query plan cache?

For read-only queries, query plan cache is enabled by default for low-latency queries. A plan is cached only when latency is lower than the threshold of 100ms. This behavior can be overridden on a per-query basis by HTTP parameter. HTTP parameter --plan-cache can take enabled or disabled as a value.

```
# Forcing plan to be cached or reused
% aws neptune-graph execute-query \
    --graph-identifier <graph-id> \
    --query-string "MATCH (n) RETURN n LIMIT 1"
    --region <region> \
    --plan-cache "enabled"
    --language open_cypher /tmp/out.txt
% aws neptune-graph execute-query \
    --graph-identifier <graph-id> \
    --query-string "RETURN \$arg"
    --region <region> \
    --plan-cache "enabled" \
    --parameters "{\"arg\": 123}"
    --language open_cypher /tmp/out.txt
```

How to check if a plan is cached?

To check if a plan is cached, use explain. For read-only queries, if the query was submitted and the plan was cached, explain would show explain details relevant to the query plan cache.

```
% aws neptune-graph execute-query \
    --graph-identifier <graph-id> \
    --query-string "MATCH (n) RETURN n LIMIT 1"
    --region <region> \
    --plan-cache "enabled" \
    --explain-mode "static" \
    --language open_cypher /tmp/out.txt
```

Query: <QUERY STRING> Plan cached by request: <REQUEST ID OF FIRST TIME EXECUTION> Plan cached at: <TIMESTAMP OF FIRST TIME EXECUTION> Parameters: <PARAMETERS IF QUERY IS PARAMETERIZED QUERY> Plan cache hits: <NUMBER OF CACHE HITS FOR CACHED PLAN> First query evaluation time: <LATENCY OF FIRST TIME EXECUTION> The query has been executed based on a cached query plan. Detailed explain with operator runtime statistics can be obtained by running the query with plan cache disabled (using HTTP parameter planCache=disabled).

🚺 Note

For a mutation query, explain is not yet supported.

Eviction

A query plan is evicted by cache TTL or maximum number of cached query plans reached. When the query plan is hit, the TTL is refreshed. The defaults are:

- The maximum number of plans cached per instance is 1000.
- TTL: 300_000 milliseconds or 5 minutes. Note that cache hit refreshes the TTL back to 5 min.

Conditions when a query plan is not cached

The following list demonstrates conditions for when a query plan would not be cached.

- If submitted with query-specific parameter --plan-cache "disabled".
 - If a cache is wanted, you can rerun the query without --plan-cache "disabled".
- If the query evaluation time is larger than latency threshold, it's not cached since it's a longrunning query and is considered to not benefit from query plan cache.
- If the query contains pattern that does not return any results.
 - i.e. MATCH (n:nonexistentLabel) return n when there are zero nodes with specified label.
 - i.e. MATCH (n {name: \$param}) return n with parameters={"param": "abcde"} when there are zero nodes with name=abcde.
- If the query parameter is composite type (list, map).

```
aws neptune-graph execute-query \
    --graph-identifier <graph-id> \
    --query-string "RETURN \$arg"
    --region <region> \
    --plan-cache "enabled" \
```

```
--parameters "{\"arg\": [1, 2, 3]}"
--language open_cypher /tmp/out.txt
aws neptune-graph execute-query \
    --graph-identifier <graph-id> \
    --query-string "RETURN \$arg"
    --region <region> \
    --plan-cache "enabled" \
    --parameters "{\"arg\": {\"a\": 1}}"
    --language open_cypher /tmp/out.txt
```

- If the query parameter is a string that has not been part of data load or data insertion.
 - If CREATE (n {name: "X"}), is done to insert "X".
 - RETURN "X" is cached, while RETURN "Y" isn't, as "Y" has not been inserted and does not exist in the database.

Query explain

The openCypher explain feature is a feature that helps users to understand how the query is executed. Usually this is used in the context of query performance analysis.

Explain inputs

To invoke explain, you can pass the explain-mode parameter to an ExecuteQuery request specifying the desired explain mode (i.e., level of detail), where this explain mode value can be one of the following:

- static In static mode, explain doesn't run the query, but instead prints only the static structure of the query plan.
- details In details mode, explain runs the query, and includes dynamic aspects of the query plan. These may include the number of intermediate bindings flowing through the operators, the ratio of incoming bindings to outgoing bindings, and the total time taken by each operator. Additional details, such as the actual openCypher query string and the estimated range count for the pattern underlying a join operator, are also shown.

The following code examples provide the explain-mode when using either the AWS CLI or AWSCURL.

AWS CLI

```
aws neptune-graph execute-query \
--region <region> \
--graph-identifier <graph-id> \
--query-string <query-string> \
--explain-mode <explain-mode> \
--language open_cypher /tmp/out.txt
```

AWSCURL

```
awscurl -X POST "https://<graph-id>.<endpoint>/queries" \
-H "Content-Type: application/x-www-form-urlencoded" \
--region <region> \
--service neptune-graph \
-d "query=<query>&explain=<mode>"
```

Explain outputs

DFE operators in openCypher explain output

To use the information that the openCypher explain feature provides, you need to understand some details about how the DFE query engine works (DFE being the engine that Neptune uses to process openCypher queries).

The DFE engine translates every query into a pipeline of operators. Starting from the first operator, intermediate solutions flow from one operator to the next through this operator pipeline. Each row in the explain table represents a result, up to the point of evaluation. The operators that can appear in a DFE query plan are as follows:

- DFEApply Executes the function specified by functor in the arguments section, on the value stored in the specified variable
- DFEAlgoWriteProperty Explain operator for the property-writing portion of mutate algorithm invocations.
- DFEBFSAlgo Explain operator for invocations of the Breadth First Search algorithm, which searches for nodes from a starting vertex (or starting vertices, also called multi-source BFS) in a graph in breadth-first order.
- DFEBindRelation Binds together variables with the specified names.

- DFEChunkLocalSubQuery This is a non-blocking operation that acts as a wrapper around subqueries being performed.
- DFEClosenessCentralityAlgo Explain operator for invocations of the Closeness Centrality
 algorithm, which computes a metric that can be used as a positive measure of how close a given
 node is to all other nodes or how central it is in the graph.
- DFECommonNeighborsAlgo Explain operator for invocations of the Common Neighbors algorithm, which counts the number of common neighbors of two input nodes.
- DFECreateConstant Extends the given input relation with new columns containing constant values.
- DFEDegreeAlgo Explain operator for invocations of the Degree algorithm, which calculates the number of edges that are incident to a vertex.
- DFEDistinctColumn Returns the distinct subset of the input values based on the variable specified.
- DFEDistinctRelation Returns the distinct subset of the input solutions based on the variable specified.
- DFEDrain Appears at the end of a subquery to act as a termination step for that subquery. The number of solutions is recorded as Units In. Units Out is always zero.
- DFEForwardValue Copies all input chunks directly as output chunks to be passed to its downstream operator.
- DFEGroupByHashIndex This is a blocking operation that organizes the rows of a relation according to a set of variables, outputting a single group identifier column that is one-to-one with the rows of the input relation. Groups here are defined by the join variables used to build the hash index (See DFEHashIndexBuild for where this hash index might be built.)
- DFEHashIndexBuild Builds a hash index over a set of variables as a side-effect. This hash index is typically reused in later operations. (See DFEHashIndexJoin for where this hash index might be used.)
- DFEHashIndexJoin Performs a join over the incoming solutions against a previously built hash index. (See DFEHashIndexBuild for where this hash index might be built.)
- DFEJaccardSimilarityAlgo Explain operator for invocations of the Jaccard similarity algorithm, which measures the similarity between two sets of nodes.
- DFEJoinExists Takes a left and right hand input relation, and retains values from the left relation that have a corresponding value in the right relation as defined by the given join variables.

- DFELabelPropagationAlgo Explain operator for invocations of the Label Propagation algorithm, which is used for community detection.
- DFELoopSubQuery This is a non-blocking operation that acts as a wrapper for a subquery, allowing it to be run repeatedly for use in loops.
- DFEMergeChunks This is a blocking operation that combines chunks from its upstream operator into a single chunk of solutions to pass to its downstream operator (inverse of DFESplitChunks).
- DFEMinus Takes a left and right hand input relation, and retains values from the left relation that do not have a corresponding value in the right relation as defined by the given join variables. If there is no overlap in join variables across both relations, then this operator simply returns the left hand input relation as is.
- DFENotExists Takes a left and right hand input relation, and retains values from the left relation that do not have a corresponding value in the right relation as defined by the given join variables. If there is no overlap in join variables, then this operator will return an empty relation.
- DFEOptionalJoin Performs the optional join A OPTIONAL B ≡ (A JOIN B) UNION (A MINUS_NE B). This is a blocking operation.
- DFEOverlapSimilarityAlgo Explain operator for invocations of the Overlap Similarity algorithm, which measures the overlap between the neighbors of two nodes.
- DFEPageRankAlgo Explain operator for invocations of the Page Rank algorithm, which calculates a score for a given node based on the number, quality, and importance of the edges pointing to that node.
- DFEPipelineJoin Joins the input against the tuple pattern defined by the pattern argument.
- DFEPipelineRangeCount Counts the number of solutions matching a given pattern, and returns a single solution containing the count value.
- DFEPipelineScan Scans the database for the given pattern argument, with or without a given filter on column(s).
- DFEProject Takes multiple input columns and projects only the desired columns.
- DFEReduce Performs the specified aggregation function on specified variables.
- DFERelational Join Joins the input of the previous operator based on the specified pattern keys using a merge join. This is a blocking operation.
- DFERouteChunks Takes input chunks from its singular incoming edge and routes those chunks along its multiple outgoing edges.

- DFESCCAlgo Explain operator for invocations of the Strongly Connected Components algorithm, which calculates the maximally connected subgraphs of a directed graph where every node is reachable from every other node.
- DFESelectRows This operator selectively takes rows from its left input relation solutions to forward to its downstream operator. The rows selected based on the row identifiers supplied in the operator's right input relation.
- DFESerialize Serializes a query's final results into a JSON string serialization, mapping each input solution to the appropriate variable name. For node and edge results, these results are serialized into a map of entity properties and metadata.
- DFESort Takes an input relation and produces a sorted relation based on the provided sort key.
- DFESplitByGroup Splits each single input chunk from one incoming edge into smaller output chunks corresponding to row groups identified by row ids from the corresponding input chunk from the other incoming edge.
- DFESplitChunks Splits each single input chunk into smaller output chunks (inverse of DFEMergeChunks).
- DFESSSPAlgo Explain operator for invocations of the single source shortest path (SSSP) algorithms (Delta-stepping and Bellman-ford).
- DFEStreamingHashIndexBuild Streaming version of DFEHashIndexBuild.
- DFEStreamingGroupByHashIndex Streaming version of DFEGroupByHashIndex.
- DFESubquery This operator appears at the beginning of all plans and encapsulates the portions of the plan that are run on the DFE engine, which is the entire plan for openCypher.
- DFESymmetricHashJoin Joins the input of the previous operator based on the specified pattern keys using a hash join. This is a non-blocking operation.
- DFESync This operator is a synchronization operator supporting non-blocking plans. It takes solutions from two incoming edges and forwards these solutions to the appropriate downstream edges. For synchronization purposes, the inputs along one of these edges may be buffered internally.
- DFETee This is a branching operator that sends the same set of solutions to multiple operators.
- DFETermResolution Performs a localize or globalize operation on its inputs, resulting in columns of either localized or globalized identifiers respectively.
- DFETopKSSSPAlgo Explain operator for invocations of the TopK hop-limited single source (weighted) shortest path algorithm algorithm, which finds the single-source weighted shortest paths from a source node to its neighbors out to the distance specified by maxDepth.

- DFETotalNeighborsAlgo Explain operator for invocations of the Total Neighbors algorithm, which counts the total number of unique neighbors of two input vertices.
- DFEUnfold Unfolds lists of values from an input column into the output column as individual elements.
- DFEUnion Takes two or more input relations and produces a union of those relations using the desired output schema.
- DFEVSSAlgo Explain operator for invocations of the Vector similarity search algorithms, which find similar vectors based on the distance to each other.
- DFEWCCAlgo Explain operator for invocations of the Weakly Connected Components algorithm, which finds the weakly-connected components in a directed graph.
- SolutionInjection Appears before everything else in the explain output, with a value of one in the Units Out column. However, it serves a no-op, and doesn't actually inject any solutions into the DFE engine.
- TermResolution Appears at the end of plans and translates of objects from the Neptune engine into openCypher objects.

Columns in openCypher explain output

The query plan information generated as openCypher explain output contains tables with one operator per row. The table has the following columns:

- ID The numeric ID of this operator in the plan.
- Out #1 (and Out #2) The ID(s) of operator(s) that are downstream from this operator. There can be at most two downstream operators.
- Name The name of this operator.
- Arguments Any relevant details for the operator. This includes things like input schema, output schema, pattern (for PipelineScan and PipelineJoin), and so on.
- Mode A label describing fundamental operator behavior. This column is mostly blank (-). One exception is TermResolution, where mode can be id2value_opencypher, indicating a resolution from ID to openCypher value.
- Units In The number of solutions passed as input to this operator. Operators without upstream
 operators, such as DFEPipelineScan, SolutionInjections, and a DFESubquery with no
 static value injected, would have zero value.

- Units Out The number of solutions produced as output of this operator. DFEDrain is a special case, where the number of solutions being drained is recorded in Units In and Units Out is always zero.
- Ratio The ratio of Units Out to Units In.
- Time (ms) The CPU time consumed by this operator, in milliseconds.

🚯 Note

Depending on the level of detail selected via the explain mode parameter, some of these columns may not appear in the output.

Explain examples

The following is a basic example of openCypher explain output. The query is a single-node lookup in the air routes dataset for a node with the airport code ATL that invokes explain using the details mode:

```
## sample query
aws neptune-graph execute-query \
--region <region> \
--graph-identifier <graph-id> \
--query-string "MATCH (n {code: 'ATL'}) RETURN n" \
--explain-mode details \
--language open_cypher /tmp/out.txt
## output
Query:
MATCH (n {code: 'ATL'}) RETURN n
# ID # Out #1 # Out #2 # Name
                           # Arguments
                                       # Mode # Units In #
Units Out # Ratio # Time (ms) #
# 0 # 1
        # -
                                                  #
             # SolutionInjection
                           # solutions=[{}]
                                       # -
                                           # 0
1
      # 0.00 # 0
                  #
# 0
# 1 # -
        # -
             # DFESubquery
                           # subQuery=subQuery1 # -
                                                  #
0
      # 0.00 # 8.00
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```

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At the top-level, SolutionInjection appears before everything else, with 1 unit out. Note that it doesn't actually inject any solutions. You can see that the next operator, DFESubquery, has 0 units in.

After SolutionInjection at the top-level is the DFESubquery operator. DFESubquery encapsulates the parts of the query execution plan that are pushed to the DFE engine (for openCypher queries, the entire query plan is executed by the DFE). All the operators in the query plan are nested inside subQuery1 that is referenced by DFESubquery.

All the operators that are pushed down to the DFE engine have names that start with a DFE prefix. As mentioned above, the whole openCypher query plan is executed by the DFE, so as a result, all of the operators start with DFE.

Inside subQuery1, there can be zero (as in this case) or more DFEChunkLocalSubQuery or DFELoopSubQuery operators that encapsulate a part of the pushed execution plan that is executed in a memory-bounded mechanism. A DFEChunkLocalSubQuery contains one SolutionInjection that is used as an input to the subquery. To find the table for that subquery in the output, search for the subQuery=graph URI specified in the Arguments column for the DFEChunkLocalSubQuery or DFELoopSubQuery operator.

In subQuery1, DFEPipelineScan with ID 0 scans the database for a specified pattern. The pattern scans for vertices ?n with property code saved as a variable ?n_code2. The inlineFilters argument shows the filtering for the code property equaling ATL.

Next, the DFEProject operator propagates forward only the ?n variable we're interested in. Finally, the DFESerialize operator performs result serialization, transforming the input solutions into a readable format.

Statistics

Neptune Analytics uses similar statistics for planning query execution as in <u>Neptune Database</u>. Computing these statistics is performed as an integrated part of the Neptune Analytics storage system. There are a number of differences between the features and usage of statistics between Neptune Analytics and Neptune Database:

- 1. Initial statistics generation is performed as part of either the initial import task or an initial data load occurring before any query-driven updates. Subsequently, statistics re-computation is triggered automatically based on the amount of update operations performed by the database.
- 2. Like with Neptune Database, Neptune Analytics has a size limit for statistics data, beyond which statistics will be disabled. The number of predicate statistics, may not exceed one million (the same as Neptune Database). There is no hard limit on the number of characteristic sets present in the underlying data. However, beyond 10,000 characteristic sets, the system will begin to merge statistics data in order to limit the overall size of data being managed.
- 3. Statistics generation is fully managed by the storage system. There are no APIs to disable or recompute statistics.
- 4. There are no CloudWatch metrics relating to statistics generation.

Exceptions

The following table lists query-side exceptions that could be encountered while using a query.

Neptune Analytics error code	HTTP status	Retriable	Description
Validation Exception	400	No	Something is wrong with the required information - Eg. a malformed query.
AccessDeniedExcept ion	403	No	User is not authorize d to perform the requested operation.
ResourceNotFoundEx ception	404	No	Requested resource is not available.
ThrottlingException	429	Yes	The server has received too many concurrent requests.
InternalServerErro rException	500	Yes	The server failed to process the request for an unknown reason.
UnprocessableExcep tion	422	No	Request cannot be processed due to known reasons - Eg. The query timed out.
ConflictException	409	Yes	Concurrently running queries attempted to modify resources or data records concurrently and the

Neptune Analytics error code	HTTP status	Retriable	Description
			conflict could not be resolved automatic ally. Please retry with an exponential back- off strategy.

Neptune Analytics openCypher data model

For details on the openCypher data model, please refer to the Neptune Database <u>documentation</u>. There are some differences in modeling of vertices without labels. Neptune Database adds vertices with a default label if one is not explicitly provided. All but the last label of a vertex can be deleted.

What is a vertex?

As well as loading both vertices and edges, unlike Neptune Database, Neptune Analytics also allows loading just edges and is still able to run algorithms and queries from that starting point. This is useful if your main interest is, for example, loading a file of edge data from a CSV file and running an algorithm over the data without needing to provide any additional vertex information. This has some implications on how vertices are managed. For the Neptune Analytics query engine, a vertex implicitly exists if it either has an explicit label, a property, or an edge. Likewise, a vertex gets implicitly deleted if all its labels, properties, and edges get removed. Unlike Neptune Database, Neptune Analytics stores a label for a vertex only if one is explicitly provided by the user, and all labels of a vertex can be deleted.

This affects some common openCypher queries. An attempt to create a vertex that has neither a label nor properties or edges has no effect. That is, queries such as CREATE (n) or CREATE (n {`~id`: "xyz"}) do not add any vertices to the graph. CREATE (n {key:value}), where key is different from `~id`, creates a vertex with the property key, and CREATE (n)-[knows]->(m) creates two vertices with the one shared edge.

CREATE (n {key:value}), where key is different from `~id`, creates a vertex with the property key, and a subsequent MATCH (n) will discover that vertex. A query such as MATCH (n {key:value}) REMOVE n.key will remove the only property for the (edge- and label-less) vertex, which implicitly deletes the vertex. A subsequent MATCH (n) query will not return that vertex. Likewise, CREATE (n:Label) adds a vertex with the label Label (and no other

properties or edges). Now, MATCH (n) REMOVE n:Label deletes the only label of the vertex, which implicitly deletes the vertex.

Similarly, CREATE (n) - [knows] - >(m) creates two nodes and one edge. MATCH (n) will discover those two vertices. Now, MATCH (n) - [r:knows] - >(m) DELETE r will delete that edge, and implicitly deletes the two vertices. Those two vertices are no longer returned when running a MATCH (n) query.

Merge on empty vertices, MERGE (n) or MERGE (n {`~id`: "xyz"}), are not permitted and will throw an exception. MERGE (n {key:value}) creates a vertex with property key if a matching vertex does not exist.

The following table illustrates the differences between Neptune Database and Neptune Analytics.

Query (run on empty graph)	Neptune Database	Neptune Analytics
CREATE (n)	Adds a vertex with label "vertex" to the graph. Each repeat request adds a new vertex to the graph.	No change to the graph, query returns without exception. Repeat requests similarly do not change the graph, and query returns without exception.
CREATE (n {`~id`: "xyz"})	Adds a vertex with id "xyz" and label "vertex" to the graph. Repeat request fails with exception.	No change to the graph, query returns without exception. Repeat requests similarly do not change the graph, and query returns without exception.
CREATE (n {key:valu e})	Adds a vertex with label "vertex" and property "key" to the graph.	Adds a vertex with property "key" to the graph. This vertex has no label.
CREATE (n {key:valu e})	The REMOVE query removes the "key" property on the	The remove query removes the property on the vertex,

Query (run on empty graph)	Neptune Database	Neptune Analytics
MATCH (n {key:value}) REMOVE n.key	vertex. The graph contains a vertex with label "vertex" but no property. MATCH (n) returns the vertex.	and as a side effect the vertex gets deleted from the graph. MATCH (n) does not return the vertex.
CREATE (n:Label {`~id`: "xyz", key:value}) MATCH (n {`~id`: "xyz"}) REMOVE n:Label	The REMOVE query errors out, the last label on a vertex cannot be deleted.	The REMOVE query removes the label. The graph contains a graph with id "xyz" and property "key".
CREATE (n)-[knows]- >(m)	Adds two vertices with label "vertex" and an edge with label "knows" to the graph. MATCH (n) returns both those vertices.	Adds an edge between two new vertices to the graph. MATCH (n) returns both those vertices.
CREATE (n)-[knows]- >(m) MATCH (n)-[r:knows]- >(n) DELETE r	Deletes the edge. The graph contains two isolated vertices. MATCH (n) returns both those vertices.	Deletes the edge, and as a side effect the two vertices also get deleted from the graph. The graph is now empty. MATCH (n) does not return the two vertices.
MERGE (n)	Adds a vertex with label "vertex" if graph is empty. Matches all vertices in a non- empty graph.	Throws an exception.

Query (run on empty graph)	Neptune Database	Neptune Analytics
MERGE (n {`~id`: "xyz"})	Adds a vertex with label "vertex" and id "xyz" if one does not exist in the graph. Matches vertex with id "xyz".	Throws an exception.
<pre>MERGE (n {key:value})</pre>	Adds a vertex with label "vertex" and property "key" to the graph, if such a vertex does not already exists.	Adds a vertex with property "key" to the graph, if such a vertex does not already exist. This vertex has no label.
MERGE (n)-[knows]- >(m)	Adds two vertices with label "vertex" and an edge with label "knows" to the graph, if an edge with label knows does not exist. MATCH (n) returns both those vertices.	Adds an edge between two new vertices to the graph, if an edge with label "knows" does not exist. The two vertices have no label. MATCH (n) returns both those vertices.

Note

A workaround to implicit deletion of a vertex when all of its labels, properties, and edges get removed is to assign immutable labels to all vertices. This way, the deletion of all properties, edges, or mutable labels of a vertex will not lead to an implicit deletion. A vertex would not get deleted until explicitly deleted.

Likewise a workaround to no-op vertex create queries is to always create a vertex with a label or a property. To combine it with the previous point, always create a vertex with an immutable label. Extending this to bulk or batch loads, include all vertices in some vertex files and assign a property or an immutable label to all vertices.

Neptune Analytics openCypher specification compliance

Please refer to the Neptune Database documentation found <u>here</u> for openCypher specification compliance, with the exception that Neptune Analytics does not support custom edge IDs.

Vertex and edge IDs

Custom IDs for vertices

Neptune Analytics supports both querying and creating vertices with custom IDs. See <u>openCypher</u> <u>custom IDs</u> for more details.

Custom IDs for edges

Neptune Analytics does not support edge creation with custom edge IDs. Custom IDs are not permitted in CREATE or MERGE clauses. Edges are assigned IDs by Neptune , using a reserved prefix neptune_reserved_. Edges can be queried by the server assigned ids, just as in <u>Neptune</u> <u>Database</u>.

```
# Supported
MATCH (n)-[r:knows {`~id`: 'neptune_reserved_1_123456789'}]->(m)
RETURN r
# Unsupported
CREATE (n:Person {name: 'John'})-[:knows {`~id`: 'john-knows->jim'}]->(m:Person {name:
'Jim'})
# Unsupported
MERGE (n)-[r:knows {`~id`: 'neptune_reserved_1_123456789'}]->(m)
RETURN r
```

Server assigned IDs are recycled. After an edge is deleted, a new edge created could get assigned the same ID.

🚺 Note

The edges could get assigned new IDs if the graph gets restructured and the older IDs would then become invalid. If the edges are reassigned IDs, older IDs would match no other edges. It is not recommended to store these IDs externally for long-term querying purposes.

Transaction isolation levels in Neptune Analytics

Neptune Analytics has some differences with isolation level supported by Neptune Database.

Read-only query isolation in Neptune Analytics: Neptune Analytics evaluates read-only queries under snapshot isolation, just like Neptune Database.

Mutation query isolation in Neptune Analytics: Reads for mutation queries are normally executed under snapshot isolation, unlike Neptune Database. This is less stricter isolation than Neptune Database as the conditions in the query for proceeding to a write satisfied in a snapshot could have changed concurrently before the query commits.

For some specific steps, such as node/relationship deletion or conditional creation of new data using the MERGE step, reads also look at the concurrent writes, to avoid inconsistencies. Below are some examples where concurrent execution of queries one and two always lead to a consistent state. At most, one vertex gets created in example #1. The age is set to 10 or 11 in example #2, not both. And in example #3, either the vertex is fully deleted or the age is set to 11 without any deletion or removal of other properties.

```
# EXAMPLE 1
Query 1: MERGE (m:Person {ssn: '123456789'})
Query 2: MERGE (n:Person {ssn: '123456789'})
```

```
# EXAMPLE 2
Query 1: MATCH (n {ssn : '123456789'}) SET n.age=10
Query 2: MATCH (n {ssn : '123456789'}) SET n.age=11
```

```
# EXAMPLE 3
Query 1: MATCH (n {ssn : '123456789'}) DETACH DELETE n
Query 2: MATCH (n {ssn : '123456789'}) SET n.age = 11
```

Conflict detection: Different from Neptune Database, conflicts are evaluated more precisely over individual graph elements (properties or edges) rather than over a range of data. Queries one and two in example #4 would not conflict when run concurrently because they search and merge on different property values ('lname1' and 'lname2'). However, queries one and two in example #5 merge on different property-value sets, but they could still confict when run concurrently because they share a property-value (firstName: 'fname').

[#] EXAMPLE 4

```
Query 1: MERGE (n {lastName: 'lname1'})
Query 2: MERGE (n {lastName: 'lname2'})
```

EXAMPLE 5
Query 1: MERGE (n {firstName: 'fname', lastName: 'lname1'})
Query 2: MERGE (n {firstName: 'fname', lastName: 'lname2'})

Neptune Analytics algorithms

Graph algorithms are powerful tools for gaining insights into data. Neptune Analytics provides a set of optimized in-database implementations of common graph algorithms that are exposed as openCypher procedures. These algorithms analyze inherent aspects of the underlying graph structure, such as connectedness (path finding), relative importance (centrality), and community membership (community detection).

Neptune Analytics natively supports over 25 optimized graph algorithms and variants in the 5 most popular categories that help customers extract insights from their graphs, which are listed in the following table.

Category	Action	Algorithms	Common Uses
<u>Pathfinding</u>	Find the existence, quality, or availabil ity of a path between nodes.	 Breadth-First Search Single-Source Shortest Path Top-K Source Shortest Path 	 Logistics optimizat ion Social network recommendations Route optimization
<u>Centrality</u>	Determines the absolute or relative importance of a node in the graph.	 Degree PageRank Closeness Centralit y 	 Fraud ring/Coll usion detection Social network influencer identific ation Supply chain risk analysis
<u>Similarity</u>	Compare the similarit ies between different graph structures.	 Common Neighbors Total Neighbors Jaccard Similarity Overlap Similarity 	 Biological structura l analysis Social network cluster comparison Link prediction

Category	Action	Algorithms	Common Uses
<u>Clustering and</u> <u>Community</u> <u>Detection</u>	Identify meaningfu l groups or clusters within graph structures.	 Weakly Connected Components (WCC) Strongly Connected Components (SCC) Label Propagation 	 Social network clusters Fraud ring identific ation Householding Biological interacti on
<u>Vector Similarity</u> <u>Search</u>	Identify approxima te nearest neighbor (ANN) nodes by comparing vector embeddings using the Hierarchical Navigable Small World (HNSW) algorithm.	DistanceTop-K	 RAG applications Knowledge graph backed chat bots Approximate nearest neighbors

Many of these algorithms require interacting with most or all the nodes and edges in a graph, often in an iterative fashion. As a result, they are too computationally expensive to process using normal analytic technologies. Neptune Analytics has built highly optimized implementations that allow them to run over graphs of any size.

Algorithms in Neptune Analytics are integrated naturally into openCypher through the CALL clause, as illustrated below. This lets you combine algorithms naturally with openCypher clauses, functions, and semantics to build very complex queries. For example, you could look for the top 10 most important airports in the US-AK region like this:

```
MATCH (n:airport {region: 'US-AK'})
CALL neptune.algo.pageRank(n, {edgeLabels: ['route'], numOfIterations: 10})
YIELD rank
RETURN n.code, rank
ORDER BY rank DESC LIMIT 10
```

You can run algorithms in the SDKs using the ExecuteOpenCypherQuery operation or in boto3 and the AWS CLI using the execute-query command. If you don't want to use the SDK or CLI, you can use you can use <u>awscurl</u> to sign your Neptune Analytics requests using <u>signed using Signature</u> Version 4 (Sig4). For example, you can run a simple breadth-first search like this:

```
awscurl -X POST -H "Content-Type: application/x-www-form-urlencoded" \
    https://(graphIdentifier).(region).neptune-graph.amazonaws.com/opencypher \
    --service neptune-graph \
    --region (region) \
    -d "query=CALL neptune.algo.bfs([\"101\", \"102\"], {edgeLabels: [\"route\"]})"
```

You could run the same query using the AWS CLI like this:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string 'CALL neptune.algo.bfs(["101", "102"], {edgeLabels: ["route"]})' \
    --language open_cypher \
    /tmp/out.txt
```

Algorithms having signatures with different kinds of input in Neptune Analytics are exposed as separate algorithms. Unless otherwise indicated, the examples here are using the <u>Air Routes</u> <u>dataset</u>.

Neptune Analytics currently supports five main categories of algorithm:

• <u>Path finding algorithms</u> – These find the existence, quality, or availability of a path or paths between two or more nodes in the graph. A path in this sense is a set of nodes and connecting edges.

By efficiently determining the optimal route between two nodes, path-finding algorithms enable you to model real-world systems like roads or social networks as interconnected nodes and edges. Finding the shortest paths between various points is crucial in applications like route planning for GPS systems, logistics optimization, and even in solving complex problems in fields like biology or engineering.

• <u>Centrality algorithms</u> – These are used to determine the absolute or relative importance or influence of a node or nodes in the graph.

By identifying the most influential or important nodes within a network, centrality algorithms can provide insights about key players or critical points of interaction. This is valuable in social

network analysis, where it helps pinpoint influential individuals, and in transportation networks, where it aids in identifying crucial hubs for efficient routing and resource allocation.

- <u>Similarity algorithms</u> Graph similarity algorithms allow you to compare and analyze the similarities and dissimilarities between different graph structures, which can provide insight into relationships, patterns, and commonalities across diverse datasets. This is invaluable in various fields, such as biology, for comparing molecular structures, such as social networks, for identifying similar communities, and such as recommendation systems, for suggesting similar items based on user preferences.
- <u>Clustering or community-detection algorithms</u> Community-detection algorithms can identify meaningful groups or clusters of nodes in a network, revealing hidden patterns and structures that can provide insights into the organization and dynamics of complex systems. This is valuable in social network analysis, and in biology, for identifying functional modules in protein-protein interaction networks, and more generally for understanding information flow and influence propagation in many different domains.
- Vector Similarity Search Vector similarity algorithms work by using vector based representations of data, a.k.a. embeddings, to answer questions about the data's context and its similarity and connection to other data. This is valuable in applications such as Retrieval Augmented Generation (RAG) applications, knowledge graph backed chatbots, and recommendation engines.

Custom Algorithms

Property graph information

Property Graph Information (graph.pg_info) summarizes some of the basic metrics of the graph, such as the number of vertices, the number of edges, the number of edge properties, the number of vertex properties, the number of edge labels, and the number of vertex labels.

Inputs for graph.pg_info

There are no inputs for graph.pg_info.

Outputs for graph.pg_info

There are two columns in the output relation: the first column is the metric name and the second column is the count.

metric: the metrics that graph.pg_info will return, which include:

- numVertices: the number of vertices in the graph.
- numEdges: the number of edges in the graph.
- numVertexProperties: the number of node properties in the graph.
- numEdgeProperties: the number of edge properties in the graph.
- numVertexLabels: the number of unique vertex labels in the graph.
- numEdgeLabels: the number of unique edge labels in the graph.

count

• count: the value of the above metrics.

graph.pg_info query example

```
## Syntax
CALL neptune.graph.pg_info()
YIELD metric, count
RETURN metric, count
```

graph.pg_info query integration

```
# sample query integration
CALL neptune.graph.pg_info()
YIELD metric, count
WHERE metric = 'numVertices'
RETURN count
```

Sample graph.pg_info output

```
# sample output of graph.pg_info
aws neptune-graph execute-query \
         --graph-identifier ${graphIdentifier} \
         --query-string "CALL neptune.graph.pg_info()
        YIELD metric, count
        RETURN metric, count " \
         --language open_cypher \
```

```
/tmp/out.txt
cat /tmp/out.txt
{
  "results": [{
      "metric": "numVertices",
      "count": 3748
    }, {
      "metric": "numEdges",
      "count": 57538
    }, {
      "metric": "numVertexProperties",
      "count": 42773
    }, {
      "metric": "numEdgeProperties",
      "count": 50532
    }, {
      "metric": "numVertexLabels",
      "count": 4
    }, {
      "metric": "numEdgeLabels",
      "count": 2
    }]
}
```

Path-finding algorithms in Neptune Analytics

Path-finding algorithms are a category of graph algorithms that focus on finding a path, a connected set of nodes and edges, between two or more sets of nodes within a graph. They are often used to find available or optimized paths based on the existence, quantity, or quality of the paths and the values of properties along those paths.

By efficiently determining the best route between two nodes, path-finding algorithms enable you to model real-world systems like roads or social networks as interconnected nodes and edges. Finding the shortest paths between various points is crucial in applications like route planning for GPS systems, logistics optimization, and even in solving complex problems in fields like biology or engineering.

Breadth-first search (BFS) path finding algorithms

Breadth-first search (BFS) path-finding algorithms search for nodes in breadth-first order, starting from a single vertex. They can also, in the multi-source case, start from more than one vertex.

They can systematically explore and evaluates all neighboring nodes from a starting point before moving on to the neighbors of those nodes, which ensures that the algorithm searches the shallowest levels of the graph first.

Breadth-first-search is used in computer networking to find the shortest path between two devices, and in social networks to understand how information spreads through connections, and in games to explore possible moves and strategies.

Time complexity – The time complexity of breadth-first search algorithms is O(|V|+|E|), where |V| is the number of vertices in the graph and |E| is the number of edges in the graph.

A breadth-first algorithm can be invoked as a *standalone* operation whose inputs are explicitly defined, or as a *query-algorithm integration* which takes as its input the output of an immediately preceding MATCH clause.

Neptune Analytics supports these BFS algorithms:

- <u>.bfs</u> This standard breadth-first search algorithm starts from the source vertex of the graph and returns a column of visited vertices.
- <u>.bfs.parents</u> This variant of BFS starts from a source vertex or vertices and finds the parent of each vertex during the search. It returns a key column of the vertices and a value column of the parents of the key vertices.

<u>.bfs.levels</u> – This variant of BFS starts from a source vertex or vertices and finds the levels of each vertex during the search. It returns a key column of the vertices and a value column of integers that are the level values of the key vertices.

Note that the level of a source vertex is 0.

Single-source shortest-path algorithms

A single-source-shortest-path algorithm finds the shortest paths (or the distance of the shortest paths) between a given vertex and all reachable vertices in the graph (including itself).

By determining the most efficient routes from a single starting node to all other nodes in the graph, single-source-shortest-path can be used calculate the shortest distances or lowest cost required to reach each destination. This is applicable in GPS systems to find the fastest routes between a starting point and differeent destinations, and in logistics to optimize delivery routes, and in transportation planning for efficient navigation through road networks.

Neptune Analytics supports the following single-source-shortest-path (SSSP) algorithms:

- <u>.sssp.bellmanFord</u> Computes the shortest path distances from a source vertex to all other vertices in the graph using the <u>Bellman-Ford</u> algorithm. Positive edge weights must be provided using the edgeWeightProperty, and the traversal direction must not be set to both.
- <u>.sssp.bellmanFord.parents</u> Identifies the parent vertices along the shortest paths from the source vertex to all other vertices in the graph using the Bellman-Ford algorithm. Positive edge weights must be provided using the edgeWeightProperty, and the traversal direction must not be set to both.
- <u>.sssp.deltaStepping</u> Computes the shortest path distances from a source vertex to all other vertices in the graph using a <u>delta-stepping</u> algorithm. Positive edge weights must be provided using the edgeWeightProperty, and the traversal direction must not be set to both.
- <u>.sssp.deltaStepping.parents</u> Identifies the parent vertices along the shortest paths from the source vertex to all other vertices in the graph using a delta-stepping algorithm.
 Positive edge weights must be provided using the edgeWeightProperty, and the traversal direction must not be set to both.
- <u>topksssp</u> The TopK hop-limited single source shortest path algorithm finds the single-source weighted shortest paths starting from a source vertex to all its maxDepth neighbors. The distance or cost from the source vertex to each target vertex is accumulated on the edge weights of the path. The topK distances of the paths are sorted in descending or ascending order.

The algorithm can be run unweighted as well as weighted. When you run it unweighted, it's equivalent to <u>.bfs.levels</u>.

Standard breadth-first search (BFS) algorithm

Standard breadth-first search (BFS) is an algorithm for finding nodes from a starting node or nodes in a graph in breadth-first order.

It returns the source node or nodes that it started from, and all of the nodes visited by each search.

i Note

Because every source node passed in leads to its own execution of the algorithm, your queries should limit the number of source nodes as much as possible.

.bfs syntax

```
CALL neptune.algo.bfs(
  [source-node list (required)],
  {
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    maxDepth: maximum number of hops to traverse from a source node (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
    YIELD the outputs to generate (source and/or node)
RETURN the outputs to return
```

.bfs inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The source-node list contains the node or nodes used as the starting locations for the algorithm.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) – type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only nodes matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• maxDepth (optional) – type: positive integer or 0 or -1; default: -1.

The maximum number of hops to traverse from a source node. If set at -1 then there's no maximum depth limit. If set to 0, only the nodes in the source node list are returned.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.bfs outputs

The .bfs algorithm returns:

• source - type: Node[].

The source nodes.

• **node** – *type:* Node[].

The nodes that the algorithm traversed from each source node.

.bfs query examples

This is a standalone example, where the query provides an explicit source node list.

```
CALL neptune.algo.bfs(
  ["101", "102"],
  {
    edgeLabels: ["route"],
    vertexLabel: "airport",
    maxDepth: 11,
    traversalDirection: "both",
    concurrency: 2
  }
)
YIELD node
```

You can run that query using the execute-query operation in the AWS CLI like this:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string 'CALL neptune.algo.bfs(["101", "102"],
        {edgeLabels: ["route"], vertexLabel: "airport", maxDepth: 11,
        traversalDirection: "both", concurrency: 2})' \
    --language open_cypher \
    /tmp/out.txt
```

A query like this one would return an empty result because the source list is empty:

```
CALL neptune.algo.bfs([], {edgeLabels: ["route"]})
```

By default, both the source nodes ("source" output) and the visited nodes ("node" output) are returned. You can use YIELD to specify which of those outputs you would like to see. For example, to see only the "node" outputs:

```
CALL neptune.algo.bfs(["101"], {edgeLabels: ["route"]}) YIELD node
```

The examples below are query integration examples, where .bfs follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (n) WITH n LIMIT 5
CALL neptune.algo.bfs(n, {edgeLabels: ["route"]})
```

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YIELD node RETURN node

The MATCH clause can also explitly specify a starting node list using the id() function, like this:

```
MATCH (n) where id(n)="101"
CALL neptune.algo.bfs(n, {edgeLabels: ["route"]})
YIELD node
RETURN node
```

Also:

```
MATCH (n) where id(n) IN ["101", "102"]
CALL neptune.algo.bfs(n, {edgeLabels: ["route"]})
YIELD node
RETURN COUNT(node)
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .bfs output

Here is an example of the output returned by .bfs when run against the <u>sample air-routes dataset</u> using the following query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.bfs(['101'], {maxDepth: 1}) yield source, node
  return source, node limit 2" \
    --language open_cypher \
    /tmp/out.txt

cat /tmp/out.txt
{
    "results": [{
        "source": {
    }
}
```

```
"~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.681099891662599,
      "elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "lon": 100.74700164794901,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "1483",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 39.490000000000000,
      "elev": 4557,
      "longest": 9186,
      "city": "Ordos",
      "type": "airport",
      "region": "CN-15",
      "desc": "Ordos Ejin Horo Airport",
      "code": "DSN",
      "lon": 109.861388889,
      "country": "CN",
      "icao": "ZBDS",
      "runways": 1
    }
  }
}, {
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.681099891662599,
```

```
"elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "lon": 100.74700164794901,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "103",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 55.972599029541001,
      "elev": 622,
      "longest": 12139,
      "city": "Moscow",
      "type": "airport",
      "region": "RU-MOS",
      "desc": "Moscow, Sheremetyevo International Airport",
      "code": "SVO",
      "lon": 37.414600372314503,
      "country": "RU",
      "icao": "UUEE",
      "runways": 2
    }
  }
}]
```

}

Parents breadth-first search (BFS) algorithm

The parents variant of breadth-first search is an algorithm for finding nodes from a starting node or vertices in breadth-first order and then performing a breadth-first search for the parent of each node.

It returns a key column of vertices, and a value column of the vertices that are the parents of the key vertices. The parent of a source node is itself.

🚯 Note

Because every source node passed in initiates its own execution of the algorithm, your queries should limit the number of source nodes as much as possible.

.bfs.parents syntax

```
CALL neptune.algo.bfs.parents(
  [source-node list (required)],
  {
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    maxDepth: maximum number of hops to traverse from a source node (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
    YIELD the outputs to generate (source and/or node and/or parent)
RETURN the outputs to return
```

.bfs.parents inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The source-node list contains the node or nodes used as the starting locations for the algorithm.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only vertices matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• **maxDepth** (optional) – type: positive integer or 0 or -1; default: -1.

The maximum number of hops to traverse from a source node. If set at -1 then there's no maximum depth limit. If set to 0, only the vertices in the source node list are returned.

• traversalDirection (optional) - type: string; default: outbound.

The direction of edge to follow. Must be one of: inbound, oubound, or both.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.bfs.parents outputs

The .bfs.parents algorithm returns:

• **source** – *type:* Node[].

The source nodes.

node – type: Node[].

The vertices that the algorithm traversed from each source node.

• parent – *type:* Node[].

.bfs.parents query examples

Thus is a standalone examples, where the source node list is explicitly provided in the query:

```
CALL neptune.algo.bfs.parents(
    ["105", "113"],
    {
      edgeLabels: ["route"],
      vertexLabel: "airport",
      maxDepth: 2,
      traversalDirection: "both",
      concurrency: 2
    }
)
YIELD node, parent
```

A query like this one would return an empty result because the source list is empty:

CALL neptune.algo.bfs.parents([], {edgeLabels: ["route"]})

This is a query integration example, where .bfs.parents follows a MATCH clause that provides the source node list for .bfs.parents:

```
Match (n) with n LIMIT 5
CALL neptune.algo.bfs.parents(n, {edgeLabels: ["route"]})
YIELD node
RETURN n, node
```

This query is an example of aliasing the algorithm output:

```
MATCH (n {code: "AUS"})
CALL neptune.algo.bfs.parents(n, { edgeLabels: ["route"], maxDepth: 2})
YIELD node AS ReachedNode
RETURN ReachedNode
```

This query searches for routes to BFS from BKK, returning the starting node (BKK), 5 visited vertices, and their parents:

```
MATCH (n) where n.code CONTAINS "BKK"
CALL neptune.algo.bfs.parents(n, {edgeLabels: ["route"]})
YIELD node, parent
```

RETURN n, node, parent LIMIT 5

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .bfs.parents output

Here is an example of the output returned by .bfs when run against the <u>sample air-routes dataset</u> using the following query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.bfs.parents(['101'], {maxDepth: 1})
                       YIELD source, node, parent
                        RETURN source, node, parent
                        LIMIT 2"
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
          "desc": "Suvarnabhumi Bangkok International Airport",
          "code": "BKK",
```

```
"lon": 100.747001647949,
    "country": "TH",
    "icao": "VTBS",
    "runways": 2
  }
},
"node": {
  "~id": "1483",
  "~entityType": "node",
  "~labels": ["airport"],
  "~properties": {
    "lat": 39.49,
    "elev": 4557,
    "longest": 9186,
    "city": "Ordos",
    "type": "airport",
    "region": "CN-15",
    "desc": "Ordos Ejin Horo Airport",
    "code": "DSN",
    "lon": 109.861388889,
    "country": "CN",
    "icao": "ZBDS",
    "runways": 1
  }
},
"parent": {
  "~id": "101",
  "~entityType": "node",
  "~labels": ["airport"],
  "~properties": {
    "lat": 13.6810998916626,
    "elev": 5,
    "longest": 13123,
    "city": "Bangkok",
    "type": "airport",
    "region": "TH-10",
    "desc": "Suvarnabhumi Bangkok International Airport",
    "code": "BKK",
    "lon": 100.747001647949,
    "country": "TH",
    "icao": "VTBS",
    "runways": 2
  }
}
```

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```
},
{
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.6810998916626,
      "elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "lon": 100.747001647949,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "103",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 55.972599029541,
      "elev": 622,
      "longest": 12139,
      "city": "Moscow",
      "type": "airport",
      "region": "RU-MOS",
      "desc": "Moscow, Sheremetyevo International Airport",
      "code": "SVO",
      "lon": 37.4146003723145,
      "country": "RU",
      "icao": "UUEE",
      "runways": 2
    }
  },
  "parent": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
```

```
"~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
          "desc": "Suvarnabhumi Bangkok International Airport",
          "code": "BKK",
          "lon": 100.747001647949,
          "country": "TH",
          "icao": "VTBS",
          "runways": 2
        }
      }
    }
  ]
}
```

Levels breadth-first search (BFS) algorithm

The levels variant of breadth-first search is an algorithm for searching nodes from a starting node or nodes in breadth-first order. From there it performs a breadth-first search and records the hop level from the starting node of each node that it finds.

It returns a key column of nodes, and a value column containing the level values of those key nodes.

The level of a source node is 0. Note that because every source node passed into breadth-first search levels initiates its own execution of the algorithm, your queries should filter to a subset of the graph before executing BFS levels whenever possible.

.bfs.levels syntax

```
CALL neptune.algo.bfs.levels(
  [source-node list (required)],
  {
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    maxDepth: maximum number of hops to traverse from a source node (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
)
YIELD the outputs to generate (source and/or node)
RETURN the outputs to return
```

.bfs.levels inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The source-node list contains the node or nodes used as the starting locations for the algorithm.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only nodes matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• maxDepth (optional) – type: positive integer or 0 or -1; default: -1.

The maximum number of hops to traverse from a source node. If set at -1 then there's no maximum depth limit. If set to 0, only the nodes in the source node list are returned.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.bfs.levels outputs

The .bfs.levels algorithm returns:

• **source** – *type:* Node[].

The source nodes.

node – type: Node[].

The nodes that the algorithm traversed from each source node.

• level - type: integer[].

.bfs.levels standalone query examples

The examples below are standalone examples, where the query provides an explicit source node list.

A query like this one would return an empty result because the source list is empty:

```
CALL neptune.algo.bfs.levels(
  ["101", "102"],
  {
    edgeLabels: ["route"],
    vertexLabel: "airport",
    maxDepth: 6,
    traversalDirection: "both",
    concurrency: 2
   }
)
YIELD node
```

You can run the algorithm using the execute-query operation in the AWS CLI like this:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string 'CALL neptune.algo.bfs.levels(["101", "102"], {edgeLabels:
    ["route"]})' \
    --language open_cypher \
    /tmp/out.txt
```

By default, all the outputs are generated. You can use YIELD to specify which of those outputs to generate. For example, to generate only the "node" and level outputs:

CALL neptune.algo.bfs.levels(["101"], {edgeLabels: ["route"]}) YIELD node, level

.bfs.levels query integration examples

The examples below are query integration examples, where .bfs.levels follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (n) WITH n LIMIT 5
CALL neptune.algo.bfs.levels(n, {edgeLabels: ["route"]})
YIELD node, level
```

RETURN n, node, level

This query illustrates various ways to constrain the input and output:

```
MATCH (n) where id(n)="101"
CALL neptune.algo.bfs.levels(n, { edgeLabel: "route", maxDepth: 2})
YIELD node, level WHERE node.city CONTAINS "New"
RETURN n.city, node.city, level
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .bfs.levels output

Here is an example of the output returned by .bfs.levels when run against the <u>sample air</u>routes dataset using the following query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.bfs.levels(['101'], {maxDepth: 1}) yield source,
 node, level return source, node, level limit 2" \setminus
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
```

```
"city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "lon": 100.747001647949,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "1483",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 39.49,
      "elev": 4557,
      "longest": 9186,
      "city": "Ordos",
      "type": "airport",
      "region": "CN-15",
      "desc": "Ordos Ejin Horo Airport",
      "code": "DSN",
      "lon": 109.861388889,
      "country": "CN",
      "icao": "ZBDS",
      "runways": 1
    }
  },
  "level": 1
},
{
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.6810998916626,
      "elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
```

```
"desc": "Suvarnabhumi Bangkok International Airport",
        "code": "BKK",
        "lon": 100.747001647949,
        "country": "TH",
        "icao": "VTBS",
        "runways": 2
      }
    },
    "node": {
      "~id": "103",
      "~entityType": "node",
      "~labels": ["airport"],
      "~properties": {
        "lat": 55.972599029541,
        "elev": 622,
        "longest": 12139,
        "city": "Moscow",
        "type": "airport",
        "region": "RU-MOS",
        "desc": "Moscow, Sheremetyevo International Airport",
        "code": "SVO",
        "lon": 37.4146003723145,
        "country": "RU",
        "icao": "UUEE",
        "runways": 2
      }
    },
    "level": 1
  }
]
```

}

Bellman-Ford single source shortest path (SSSP) algorithm

The .sssp.bellmanFord algorithm computes the shortest path distances from a single source vertex to all other vertices in the graph using the <u>Bellman-Ford</u> algorithm.

Neptune Analytics implements the algorithm such that:

- Positive edge weights must be provided using the edgeWeightProperty field
- Negative edge weights are not supported.
- The traversal direction cannot be set to both.

.sssp.bellmanFord syntax

```
CALL neptune.algo.sssp.bellmanFord(
  [source-node list (required)],
  {
    edgeWeightProperty: edge weight predicate for traversal (required)
    edgeWeightType: numeric type of the edge weight property (required)
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
)
YIELD the outputs to generate (source and/or node)
RETURN the outputs to return
```

.sssp.bellmanFord inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.
- a configuration object that contains:
 - edgeWeightProperty (required) type: string; default: none.

The edge weight predicate for traversal.

 edgeWeightType (required) – type: string; valid values: "int", "long", "float", "double".

The numeric data type of the values in the property specified by edgeWeightProperty.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only nodes matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound" or "oubound".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the .sssp.bellmanFord algorithm

For every node that can be reached from the specified source list, the algorithm returns:

- **source** The source node.
- **node** A node found traversing from the source.
- distance The distance between the source node and the found node.

.sssp.bellmanFord query examples

This is a standalone query, where a source node (or nodes) is explicitly provided:

```
CALL neptune.algo.sssp.bellmanFord(
  [101],
  {
    edgeLabels: ["route"],
    edgeWeightProperty: "dist",
    edgeWeightType: "int"
  }
)
```

This is a query integration example, where .sssp.bellmanFord follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (source:airport {code: 'ANC'})
CALL neptune.algo.sssp.bellmanFord(
   "101",
   {
    edgeLabels: ["route"],
    edgeWeightProperty: "dist",
    edgeWeightType: "int",
    vertexLabel: "airport",
    traversalDirection: "outbound",
    concurrency: 1
   }
)
YIELD node, parent, distance
RETURN source, node, parent, distance
```

<u> M</u>arning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .sssp.bellmanFord output

Here is an example of the output returned by .sssp.bellmanFord when run against the <u>sample</u> <u>air-routes dataset</u> using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.sssp.bellmanFord(['101'],
       {edgeWeightProperty: 'dist', edgeWeightType: 'int'})
     yield source, node, distance
     return source, node, distance
     limit 2" \setminus
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
          "desc": "Suvarnabhumi Bangkok International Airport",
          "code": "BKK",
          "prscore": 0.002498496090993285,
          "degree": 308,
          "lon": 100.747001647949,
          "wccid": 2357352929951779,
          "country": "TH",
          "icao": "VTBS",
          "runways": 2
        }
      },
      "node": {
        "~id": "2709",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 65.4809036254883,
          "elev": 49,
```

```
"longest": 8711,
      "city": "Nadym",
      "type": "airport",
      "region": "RU-YAN",
      "desc": "Nadym Airport",
      "code": "NYM",
      "prscore": 0.00016044313088059425,
      "degree": 18,
      "lon": 72.6988983154297,
      "wccid": 2357352929951779,
      "country": "RU",
      "icao": "USMM",
      "runways": 1
    }
  },
  "distance": 3812
},
{
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.6810998916626,
      "elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "prscore": 0.002498496090993285,
      "degree": 308,
      "lon": 100.747001647949,
      "wccid": 2357352929951779,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "2667",
    "~entityType": "node",
    "~labels": ["airport"],
```

```
"~properties": {
          "lat": 56.8567008972168,
          "elev": 2188,
          "longest": 6562,
          "city": "Ust-Kut",
          "type": "airport",
          "region": "RU-IRK",
          "desc": "Ust-Kut Airport",
          "code": "UKX",
          "prscore": 0.000058275499999999997,
          "degree": 4,
          "lon": 105.730003356934,
          "wccid": 2357352929951779,
          "country": "RU",
          "icao": "UITT",
          "runways": 1
        }
      },
      "distance": 2993
    }
  ]
}
```

Bellman-Ford single source shortest path (SSSP) parents algorithm

The .sssp.bellmanFord.parents algorithm uses the <u>Bellman-Ford</u> algorithm to find the parent nodes along with the shortest path distances from the source node to all other nodes in the graph.

Neptune Analytics implements the algorithm such that:

- Positive edge weights must be provided using the edgeWeightProperty field
- Negative edge weights are not supported.
- The traversal direction cannot be set to both.

.sssp.bellmanFord.parents syntax

```
CALL neptune.algo.sssp.bellmanFord.parents(
  [source-node list (required)],
  {
    edgeWeightProperty: edge weight predicate for traversal (required)
    edgeWeightType: numeric type of the edge weight property (required)
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
    YIELD the outputs to generate (source and/or node)
    RETURN the outputs to return
```

.sssp.bellmanFord.parents inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.

- a configuration object that contains:
 - edgeWeightProperty (required) type: string; example: "distnce"; default: none.

The edge weight predicate for traversal.

 edgeWeightType (required) – type: string; valid values: "int", "long", "float", "double".

The numeric data type of the values in the property specified by edgeWeightProperty.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only nodes matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound" or "oubound".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the .sssp.bellmanFord.parents algorithm

For every node that can be reached from the specified source list, the algorithm returns:

- **source** The source node.
- **node** A node found traversing from the source.
- distance The distance between the source node and the found node.
- parent The parent of the found node. Note that the parent of the source vertex is itself.

.sssp.bellmanFord.parents query examples

This is a standalone query, where a source node (or nodes) is explicitly provided:

```
CALL neptune.algo.sssp.bellmanFord.parents(
  [101],
  {
    edgeLabels: ["route"],
    edgeWeightProperty: "dist",
    edgeWeightType: "int"
  }
)
```

This is a query integration example, where where .sssp.bellmanFord.parents follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (source:airport {code: 'ANC'})
CALL neptune.algo.sssp.bellmanFord.parents(
    "101",
    {
      edgeLabels: ["route"],
      edgeWeightProperty: "dist",
      edgeWeightType: "int",
      vertexLabel: "airport",
      traversalDirection: "outbound",
      concurrency: 1
    }
)
YIELD node, parent, distance
RETURN source, node, parent, distance
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .sssp.bellmanFord.parents output

Here is an example of the output returned by .sssp.bellmanFord.parents when run against the sample air-routes dataset using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.sssp.bellmanFord.parents(['101'],
       {edgeWeightProperty: 'dist', edgeWeightType: 'int'})
     yield source, node, parent
     return source, node, parent
     limit 2" \setminus
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
          "desc": "Suvarnabhumi Bangkok International Airport",
          "code": "BKK",
          "prscore": 0.002498496090993285,
          "degree": 308,
          "lon": 100.747001647949,
          "wccid": 2357352929951779,
          "country": "TH",
          "icao": "VTBS",
          "runways": 2
        }
      },
      "node": {
        "~id": "2709",
```

```
"~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 65.4809036254883,
      "elev": 49,
      "longest": 8711,
      "city": "Nadym",
      "type": "airport",
      "region": "RU-YAN",
      "desc": "Nadym Airport",
      "code": "NYM",
      "prscore": 0.00016044313088059425,
      "degree": 18,
      "lon": 72.6988983154297,
      "wccid": 2357352929951779,
      "country": "RU",
      "icao": "USMM",
      "runways": 1
    }
  },
  "parent": {
    "~id": "810",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 55.0125999450684,
      "elev": 365,
      "longest": 11818,
      "city": "Novosibirsk",
      "type": "airport",
      "region": "RU-NVS",
      "desc": "Tolmachevo Airport",
      "code": "OVB",
      "prscore": 0.0012910010991618038,
      "degree": 162,
      "lon": 82.6507034301758,
      "wccid": 2357352929951779,
      "country": "RU",
      "icao": "UNNT",
      "runways": 2
    }
  }
},
{
```

```
"source": {
  "~id": "101",
  "~entityType": "node",
  "~labels": ["airport"],
  "~properties": {
    "lat": 13.6810998916626,
    "elev": 5,
    "longest": 13123,
    "city": "Bangkok",
    "type": "airport",
    "region": "TH-10",
    "desc": "Suvarnabhumi Bangkok International Airport",
    "code": "BKK",
    "prscore": 0.002498496090993285,
    "degree": 308,
    "lon": 100.747001647949,
    "wccid": 2357352929951779,
    "country": "TH",
    "icao": "VTBS",
    "runways": 2
  }
},
"node": {
  "~id": "2667",
  "~entityType": "node",
  "~labels": ["airport"],
  "~properties": {
    "lat": 56.8567008972168,
    "elev": 2188,
    "longest": 6562,
    "city": "Ust-Kut",
    "type": "airport",
    "region": "RU-IRK",
    "desc": "Ust-Kut Airport",
    "code": "UKX",
    "prscore": 0.000058275499999999997,
    "degree": 4,
    "lon": 105.730003356934,
    "wccid": 2357352929951779,
    "country": "RU",
    "icao": "UITT",
    "runways": 1
  }
},
```

```
"parent": {
        "~id": "1038",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 52.2680015563965,
          "elev": 1675,
          "longest": 10384,
          "city": "Irkutsk",
          "type": "airport",
          "region": "RU-IRK",
          "desc": "Irkutsk Airport",
          "code": "IKT",
          "prscore": 0.0008466026629321277,
          "degree": 84,
          "lon": 104.388999938965,
          "wccid": 2357352929951779,
          "country": "RU",
          "icao": "UIII",
          "runways": 1
        }
      }
    }
  ]
}
```

Delta-stepping single source shortest path (SSSP) algorithm

The .sssp.deltaStepping algorithm computes the shortest path distances from a single source vertex to all other vertices in the graph using a delta-stepping algorithm.

Neptune Analytics implements the algorithm such that:

- Positive edge weights must be provided using the edgeWeightProperty field
- Negative edge weights are not supported.
- The traversal direction cannot be set to both.

.sssp.deltaStepping syntax

```
CALL neptune.algo.sssp.deltaStepping(
  [source-node list (required)],
  {
    edgeWeightProperty: edge weight predicate for traversal (required)
    edgeWeightType: numeric type of the edge weight property (required)
    delta: the stepping delta (optional)
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
    YIELD the outputs to generate (source and/or node)
RETURN the outputs to return
```

.sssp.deltaStepping inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.

- a configuration object that contains:
 - edgeWeightProperty (required) type: string; default: none.

The edge weight predicate for traversal.

 edgeWeightType (required) - type: string; valid values: "int", "long", "float", "double".

The numeric data type of the values in the property specified by edgeWeightProperty.

• **delta** (optional) – type: float; example: 3.0; default: 2.0.

The delta stepping value.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only nodes matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound" or "oubound".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the sssp.deltaStepping algorithm

For every node that can be reached from the specified source list, the algorithm returns:

• **source** – The source node.

• distance – The distance between the source node and the found node.

.sssp.deltaStepping query examples

This is a standalone query, where a source node (or nodes) is explicitly provided:

```
CALL neptune.algo.sssp.deltaStepping(
  [101],
  {
    edgeLabels: ["route"],
    edgeWeightProperty: "dist",
    edgeWeightType: "int"
  }
)
```

This is a query integration example, where where .sssp.deltaStepping follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (source:airport {code: 'ANC'})
CALL neptune.algo.sssp.deltaStepping(
   "101",
   {
    edgeLabels: ["route"],
    edgeWeightProperty: "dist",
    edgeWeightType: "int",
    vertexLabel: "airport",
    traversalDirection: "outbound",
    concurrency: 1
   }
)
YIELD node, parent, distance
RETURN source, node, parent, distance
```

<u> M</u>arning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .sssp.deltaStepping output

Here is an example of the output returned by .sssp.deltaStepping when run against the sample air-routes dataset using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.sssp.deltaStepping(['101'],
       {edgeWeightProperty: 'dist', edgeWeightType: 'int'})
     yield source, node, distance
     return source, node, distance
     limit 2" \setminus
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
          "desc": "Suvarnabhumi Bangkok International Airport",
          "code": "BKK",
          "prscore": 0.002498496090993285,
          "degree": 308,
          "lon": 100.747001647949,
          "wccid": 2357352929951779,
          "country": "TH",
          "icao": "VTBS",
          "runways": 2
        }
      },
      "node": {
        "~id": "2709",
```

```
"~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 65.4809036254883,
      "elev": 49,
      "longest": 8711,
      "city": "Nadym",
      "type": "airport",
      "region": "RU-YAN",
      "desc": "Nadym Airport",
      "code": "NYM",
      "prscore": 0.00016044313088059425,
      "degree": 18,
      "lon": 72.6988983154297,
      "wccid": 2357352929951779,
      "country": "RU",
      "icao": "USMM",
      "runways": 1
    }
  },
  "distance": 3812
},
{
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.6810998916626,
      "elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "prscore": 0.002498496090993285,
      "degree": 308,
      "lon": 100.747001647949,
      "wccid": 2357352929951779,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
```

```
},
      "node": {
        "~id": "2667",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 56.8567008972168,
          "elev": 2188,
          "longest": 6562,
          "city": "Ust-Kut",
          "type": "airport",
          "region": "RU-IRK",
          "desc": "Ust-Kut Airport",
          "code": "UKX",
          "prscore": 0.000058275499999999997,
          "degree": 4,
          "lon": 105.730003356934,
          "wccid": 2357352929951779,
          "country": "RU",
          "icao": "UITT",
          "runways": 1
        }
      },
      "distance": 2993
    }
  ]
}
```

Delta-stepping aingle source shortest path (SSSP) parents algorithm

The .sssp.deltaStepping.parents algorithm computes the shortest path distances from a single source vertex to all other vertices in the graph using a delta-stepping algorithm.

Neptune Analytics implements the algorithm such that:

- Positive edge weights must be provided using the edgeWeightProperty field
- Negative edge weights are not supported.
- The traversal direction cannot be set to both.

.sssp.deltaStepping.parents syntax

```
CALL neptune.algo.sssp.deltaStepping.parents(
  [source-node list (required)],
  {
    edgeWeightProperty: edge weight predicate for traversal (required)
    edgeWeightType: numeric type of the edge weight property (required)
    delta: the stepping delta (optional)
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    traversalDirection: traversal direction (optional),
    concurrency: number of threads to use (optional)
    }
)
YIELD the outputs to generate (source and/or node)
RETURN the outputs to return
```

.sssp.deltaStepping.parents inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.

- a configuration object that contains:
 - edgeWeightProperty (required) type: string; default: none.

The edge weight predicate for traversal.

 edgeWeightType (required) - type: string; valid values: "int", "long", "float", "double".

The numeric data type of the values in the property specified by edgeWeightProperty.

• **delta** (optional) – type: float; example: 3.0; default: 2.0.

The delta stepping value.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; example: "airport"; default: no node filtering.

If you provide a node label to filter on then only nodes matching that label will be traversed. This does not, however, filter out any nodes in the source node list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound" or "oubound".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the sssp.deltaStepping.parents algorithm

For every node that can be reached from the specified source list, the algorithm returns:

• **source** – The source node.

- distance The distance between the source node and the found node.
- **parent** The parent of the found node. Note that the parent of the source vertex is itself.

.sssp.deltaStepping.parents query examples

This is a standalone query, where a source node (or nodes) is explicitly provided:

```
CALL neptune.algo.sssp.deltaStepping.parents(
  [101],
  {
    edgeLabels: ["route"],
    edgeWeightProperty: "dist",
    edgeWeightType: "int"
  }
)
```

This is a query integration example, where where .sssp.deltaStepping.parents follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (source:airport {code: 'ANC'})
CALL neptune.algo.sssp.deltaStepping.parents(
    "101",
    {
        edgeLabels: ["route"],
        edgeWeightProperty: "dist",
        edgeWeightType: "int",
        vertexLabel: "airport",
        traversalDirection: "outbound",
        concurrency: 1
    }
)
YIELD node, parent, distance
RETURN source, node, parent, distance
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which

can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .sssp.deltaStepping.parents output

Here is an example of the output returned by .sssp.deltaStepping.parents when run against the sample air-routes dataset using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.sssp.deltaStepping.parents(['101'],
       {edgeWeightProperty: 'dist', edgeWeightType: 'int'})
     yield source, node, distance
     return source, node, distance
     limit 2" \setminus
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
          "desc": "Suvarnabhumi Bangkok International Airport",
          "code": "BKK",
          "prscore": 0.002498496090993285,
          "degree": 308,
          "lon": 100.747001647949,
          "wccid": 2357352929951779,
          "country": "TH",
          "icao": "VTBS",
          "runways": 2
```

```
}
},
"node": {
  "~id": "2709",
  "~entityType": "node",
  "~labels": ["airport"],
  "~properties": {
    "lat": 65.4809036254883,
    "elev": 49,
    "longest": 8711,
    "city": "Nadym",
    "type": "airport",
    "region": "RU-YAN",
    "desc": "Nadym Airport",
    "code": "NYM",
    "prscore": 0.00016044313088059425,
    "degree": 18,
    "lon": 72.6988983154297,
    "wccid": 2357352929951779,
    "country": "RU",
    "icao": "USMM",
    "runways": 1
  }
},
"parent": {
  "~id": "810",
  "~entityType": "node",
  "~labels": ["airport"],
  "~properties": {
    "lat": 55.0125999450684,
    "elev": 365,
    "longest": 11818,
    "city": "Novosibirsk",
    "type": "airport",
    "region": "RU-NVS",
    "desc": "Tolmachevo Airport",
    "code": "OVB",
    "prscore": 0.0012910010991618038,
    "degree": 162,
    "lon": 82.6507034301758,
    "wccid": 2357352929951779,
    "country": "RU",
    "icao": "UNNT",
    "runways": 2
```

```
}
  }
},
{
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.6810998916626,
      "elev": 5,
      "longest": 13123,
      "city": "Bangkok",
      "type": "airport",
      "region": "TH-10",
      "desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "prscore": 0.002498496090993285,
      "degree": 308,
      "lon": 100.747001647949,
      "wccid": 2357352929951779,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "2667",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 56.8567008972168,
      "elev": 2188,
      "longest": 6562,
      "city": "Ust-Kut",
      "type": "airport",
      "region": "RU-IRK",
      "desc": "Ust-Kut Airport",
      "code": "UKX",
      "prscore": 0.000058275499999999997,
      "degree": 4,
      "lon": 105.730003356934,
      "wccid": 2357352929951779,
      "country": "RU",
```

```
"icao": "UITT",
          "runways": 1
        }
      },
      "parent": {
        "~id": "1038",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 52.2680015563965,
          "elev": 1675,
          "longest": 10384,
          "city": "Irkutsk",
          "type": "airport",
          "region": "RU-IRK",
          "desc": "Irkutsk Airport",
          "code": "IKT",
          "prscore": 0.0008466026629321277,
          "degree": 84,
          "lon": 104.388999938965,
          "wccid": 2357352929951779,
          "country": "RU",
          "icao": "UIII",
          "runways": 1
        }
      }
    }
  ]
}
```

TopK hop-limited single source (weighted) shortest path algorithm

The .topkssspalgorithm finds the single-source weighted shortest paths from a source node to its neighbors out to the distance specified by maxDepth. It accumulates the path lengths using the edge weights along the paths and then returns a sorted list of the shortest paths.

.topksssp syntax

```
CALL neptune.algo.topksssp(
  [source-node list (required)],
  {
    hopCount: maximum hops on the shortest path (required),
    perHopLimits: [a list of the maximum number of nodes to carry forward at each hop
 (required)],
    edgeLabels: [list of edge labels for filtering (optional)],
    edgeWeightProperty: a numeric edge property to weight the traversal (optional),
    edgeWeightType: numeric type of the specified edgeWeightProperty (optional),
    vertexLabel: a node label for filtering (optional),
    traversalDirection: traversal direction (optional, default: outbound),
    costFunction: determines whether the topK distances are in ascending or descending
 order (optional),
    concurrency: number of threads to use (optional)
  }
)
YIELD source, node, distance
RETURN source, node, distance
```

Inputs for the topksssp algorithm

• a source node list (required) - type: Node[] or NodeId[]; default: none.

- Each starting node triggers its own execution of the algorithm.
- If the source-node list is empty then the query result is also empty.
- If the algorithm is called following a MATCH clause (this is known as query-algorithm integration), the output of the MATCH clause is used as the source-node list for the algorithm.
- a configuration object that contains:
 - **hopCount** (*required*) *type:* positive integer; *default:* none.

Restricts the number of hops on a shortest path, which restricts the number of iterations of the SSSP algorithm to be used.

 perHopLimits (required) – type: a list of integers; valid values: positive integers, or -1 meaning unlimited; default: none.

Each integer represents the maximum number of candidate vertices to carry to the next hop.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• edgeWeightProperty (optional) - type: string; default: none.

The edge weight predicate to for traversal. If no property is specified then the algorithm runs unweighted. If multiple properties exist on an edge having the specified name, then one of them is selected at random for the weight value.

 edgeWeightType (optional) - type: string; valid values: "int", "long", "float", "double".

The numeric data type of the values in the property specified by edgeWeightProperty. If the edgeWeightProperty is not present, edgeWeightType is ignored and the algorithm runs unweighted. If an edge contains a property specified by edgeWeightProperty that has a numeric type different from what is specified in edgeWeightType, the property value is typecast to the type specified by edgeWeightType.

• **vertexLabel** (optional) – type: string; default: none.

A node label for node filtering. If a node label is provided, vertices matching the label are the only vertices that are included, including vertices in the input list.

• costFunction (optional) - type: string; valid values: "min", "max"; default: "min".

Specifies the ordering of the topK distances returned. A "min" value indicates that the topK distances between the source and target vertices should be returned in descending order, whereas a "max" value indicates that they should be returned in ascending order.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the topksssp algorithm

For every node that can be reached from the specified source list, the algorithm returns:

- **source** The source node.
- **node** A node found traversing from the source.
- **distance** The distance between the source node and the found node.

.topksssp query examples

This ia a standalone query, where the source node list is explicitly provided in the query:

```
CALL neptune.algo.topksssp(
  ["101"],
  {
    edgeLabels: ["route"],
    hopCount: 3,
    perHopLimits: [10, 100, 1000],
    edgeWeightProperty: "dist",
    edgeWeightType: "int"
  }
)
```

This is a query integration example, where .topksssp follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (n) WHERE id(n) IN ["108","109"]
CALL neptune.algo.topksssp(
    n,
    {
      edgeLabels: ["route"],
      edgeWeightProperty: "dist",
      edgeWeightType: "int",
      hopCount: 5,
      perHopLimits: [5,10,15,20,25]
```

```
}
)
YIELD distance
RETURN n, collect(distance) AS distances'
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .topksssp output

Here is an example of the output returned by .topksssp when run against the <u>sample air-routes</u> <u>dataset</u> using this query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier}
  --query-string "CALL neptune.algo.topksssp(['101'], {hopCount: 2, perHopLimits: [3,
 5]})
      YIELD source, node, distance
      RETURN source, node, distance limit 2" \setminus
  --language open_cypher
  /tmp/out.txt
  cat /tmp/out.txt
  {
  "results": [
    {
      "source": {
        "~id": "101",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 13.6810998916626,
          "elev": 5,
          "longest": 13123,
          "city": "Bangkok",
          "type": "airport",
          "region": "TH-10",
```

```
"desc": "Suvarnabhumi Bangkok International Airport",
      "code": "BKK",
      "prscore": 0.002498496090993285,
      "degree": 308,
      "lon": 100.747001647949,
      "wccid": 2357352929951779,
      "country": "TH",
      "icao": "VTBS",
      "runways": 2
    }
  },
  "node": {
    "~id": "170",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 40.8860015869,
      "elev": 294,
      "longest": 8622,
      "city": "Naples",
      "type": "airport",
      "region": "IT-72",
      "desc": "Naples International Airport",
      "code": "NAP",
      "prscore": 0.001119577675126493,
      "degree": 222,
      "lon": 14.2908000946,
      "wccid": 2357352929951779,
      "country": "IT",
      "icao": "LIRN",
      "runways": 1
    }
  },
  "distance": 2.0
},
{
  "source": {
    "~id": "101",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 13.6810998916626,
      "elev": 5,
      "longest": 13123,
```

```
"city": "Bangkok",
        "type": "airport",
        "region": "TH-10",
        "desc": "Suvarnabhumi Bangkok International Airport",
        "code": "BKK",
        "prscore": 0.002498496090993285,
        "degree": 308,
        "lon": 100.747001647949,
        "wccid": 2357352929951779,
        "country": "TH",
        "icao": "VTBS",
        "runways": 2
      }
    },
    "node": {
      "~id": "12",
      "~entityType": "node",
      "~labels": ["airport"],
      "~properties": {
        "lat": 40.63980103,
        "elev": 12,
        "longest": 14511,
        "city": "New York",
        "type": "airport",
        "region": "US-NY",
        "desc": "New York John F. Kennedy International Airport",
        "code": "JFK",
        "prscore": 0.002885053399950266,
        "degree": 403,
        "lon": -73.77890015,
        "wccid": 2357352929951779,
        "country": "US",
        "icao": "KJFK",
        "runways": 4
      }
    },
    "distance": 2.0
  }
]
```

}

Centrality algorithms in Neptune Analytics

Centrality algorithms utilize the topology of a network to determine the relative importance or influence of a specific node within the graph. By measuring the relative importance of a node or edge within a network, centrality values can indicate which elements in a graph play a critical role in that network.

By identifying the most influential or important nodes within a network, centrality algorithms can provide insights about key players or critical points of interaction. This is valuable in social network analysis, where it helps pinpoint influential individuals, and in transportation networks, where it aids in identifying crucial hubs for efficient routing and resource allocation.

Different types of centrality algorithms use different techniques to measure the importance of a node. Understanding how an algorithm calculates centrality is important to understanding the meaning of its outputs.

In addition to returning centrality data to the client, Neptune Analytics provides mutate variations of the centrality algorithms which store the calculated centrality values as vertex properties in the graph.

Neptune Analytics supports three centrality algorithms along with their mutate variants:

- <u>degree</u> This measures a nodes's centrality by the number of edges connected to it, and can therefore be used to find the most connected nodes in a network.
- <u>degree.mutate</u> The degree centrality mutate algorithm measures the number of incident edges of each vertex it traverses and writes that calculated degree value as a property of the vertex.
- <u>pageRank</u> This is an iterative algorithm that measures a nodes's centrality by the number and quality of incident edges and adjacent vertices. The centrality of a node connected to a few important nodes may therefore be higher than that of a node connected to many less important nodes. The output of this algorithm is a value that indicates the importance of a given node relative to the other nodes in the graph.
- <u>pageRank.mutate</u> This algorithm stores the calculated PageRank of a given node as a property of the node.
- <u>closenessCentrality</u> This algorithm computes the closeness centrality (CC) metric of nodes in a graph. The closeness centrality metric of a vertex is a positive measure of how close it is to all other vertices, or how central it is in the graph. Because it indicates how quickly all other

nodes in a network can be reached from a given node, it can be used in transportation networks to identify key hub locations, and in disease-spread modeling to pinpoint central locations for targeted intervention efforts.

<u>closenessCentrality.mutate</u> – This algorithm computes the closeness centrality (CC) metric of vertices in a graph and writes them as a property of each vertex.

Degree centrality algorithm

The .degree centrality algorithm counts the number of incident edges at each node that it traverses. This measure of how connected the node is can in turn indicate the node's importance and level of influence in the network.

The .degree algorithm is used in social networks to identify popular individuals with many connections, in transportation networks to locate central hubs with numerous roads leading to and from them, and in web analysis to find influential web pages with many incoming links.

The time complexity of .degree is O(|E|), where |E| is the number of edges in the graph. The space complexity is O(|V|), where |V| is the number of vertices in the graph.

.degree syntax

```
CALL neptune.algo.degree(
   [node list (required)],
   {
    edgeLabels: [a list of edge labels for filtering (optional)],
    vertexLabel: "a node label for filtering (optional)",
    traversalDirection: traversal direction (optional),
    concurrency: the number of cores to be used to run the algorithm (optional)
    }
   )
YIELD node, degree
RETURN node, degree
```

Inputs for the .degree algorithm

• a node list (required) - type: Node[] or NodeId[]; default: none.

The node or nodes for which to return the edge count (degree). If an empty list is provided, the query result is also empty.

If the algorithm is called following a MATCH clause (query integration), the result returned by the MATCH clause is taken as the node list.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

A vertex label for vertex filtering. If a vertex label is provided, vertices matching the label are the only vertices that are included, including vertices in the input list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.degree outputs

- node A list of the requested nodes. If vertexLabel is present, only the requested nodes that match the vertexLabel value are included.
- degree A list of corresponding degree values for the nodes with respect to edges with a label specified in edgeLabels.

If the input vertex list is empty, the output is empty.

Query examples for .degree

This is a standalone example, where the source node list is explicitly specified in the query:

```
CALL neptune.algo.degree(["101"], {edgeLabel: "route"})
```

This is a more complicated standalone query submitted using the AWS CLI:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string 'CALL neptune.algo.degree(
```

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```
["101", "102", "103"],
    {
        edgeLabels: ["route"],
        vertexLabel: "airport",
        traversalDirection: "inbound",
        concurrency: 2
     }
    )
    YIELD node, degree
    RETURN node, degree' \
--language open_cypher \
/tmp/out.txt
```

This is a query integration example with frontier injection, where .degree follows a MATCH clause and finds the degree value for all vertices returned by MATCH(n:airport):

```
MATCH(n:airport)
CALL neptune.algo.degree(n, {edgeLabels: ["route"]})
YIELD degree
RETURN n, degree'
```

This is an example of multiple . degree invocations chained together, where the output of one invocation serves as the input of another:

```
CALL neptune.algo.degree(
  ["108"],
  {
    edgeLabels: ["route"],
    vertexLabel: "airport"
  }
)
YIELD node
CALL neptune.algo.degree(
  node,
  {
    edgeLabels: ["route"],
    vertexLabel: "airport"
  }
)
YIELD node AS node2 WITH id(node2) AS id
RETURN id
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .degree output

Here is an example of the output returned by .degree when run against the <u>sample air-routes</u> <u>dataset</u> using this query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string 'MATCH (n)
      CALL neptune.algo.degree(n)
      YIELD node, degree
      RETURN node, degree
      LIMIT 2' \
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "10",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 38.94449997,
          "elev": 313,
          "longest": 11500,
          "city": "Washington D.C.",
          "type": "airport",
          "region": "US-VA",
          "desc": "Washington Dulles International Airport",
          "code": "IAD",
          "prscore": 0.002264724113047123,
          "lon": -77.45580292,
```

```
"wccid": 2357352929951779,
          "country": "US",
          "icao": "KIAD",
          "runways": 4
        }
      },
      "degree": 312
    },
    {
      "node": {
        "~id": "12",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 40.63980103,
          "elev": 12,
          "longest": 14511,
          "city": "New York",
          "type": "airport",
          "region": "US-NY",
          "desc": "New York John F. Kennedy International Airport",
          "code": "JFK",
          "prscore": 0.002885053399950266,
          "lon": -73.77890015,
          "wccid": 2357352929951779,
          "country": "US",
          "icao": "KJFK",
          "runways": 4
        }
      },
      "degree": 403
    }
  ]
}
```

Degree mutate centrality algorithm

The .degree.mutate centrality algorithm counts the number of incident edges of every node in the graph. This measure of how connected the node is can in turn indicate the node's importance and level of influence in the network. The .degree.mutate algorithm then stores each node's calculated degree value as a property of the node.

The algorithm returns a single success flag (true or false), which indicates whether the writes succeeded or failed.

.degree.mutate syntax

```
CALL neptune.algo.degree.mutate(
    [node list (required)],
    {
        writeProperty: A name for the new node property where the degree values will be
    written,
        edgeLabels: [a list of edge labels for filtering (optional)],
        vertexLabel: "a node label for filtering (optional)",
        traversalDirection: traversal direction (optional),
        concurrency: the number of cores to be used to run the algorithm (optional)
    }
    YIELD success
RETURN success
```

.degree.mutate inputs

a configuration object that contains:

• writeProperty (required) - type: string; default: none.

A name for the new node property that will contain the computed degree values.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

A node label for node filtering. If vertexLabel is provided, vertices matching the label are the only vertices that are processed, including vertices in the input list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Output of the .degree.mutate algorithm

The computed degree values are written to a new vertex property using the property name specified by the writeProperty input parameter.

A single Boolean success value (true or false) is returned, which indicates whether or not the writes succeeded.

.degree.mutate query examples

The example below is a standalone example, where the source vertex list is explicitly provided in the query.

This query writes the degree values of all nodes in the graph to a new vertex property called DEGREE:

```
CALL neptune.algo.degree.mutate({writeProperty: "DEGREE", edgeLabels: ["route]})
```

After using the mutate algorithm, the newly written properties can then be accessed in subsequent queries. For example, after the mutate algorithm call above, you could use the following query to retrieve the .degree property of specific nodes:

```
MATCH (n) WHERE id(n) IN ["101", "102", "103"]
RETURN n.DEGREE'
```

Sample output from .degree.mutate

Here is an example of the output returned by .degree.mutate when run against the <u>sample air</u>routes dataset using this query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.degree.mutate({writeProperty: 'degree'}) YIELD
success RETURN success" \
    --language open_cypher \
    /tmp/out.txt
cat /tmp/out.txt
{
    "results": [
        { "success": true }
    ]
}
```

PageRank centrality algorithm

PageRank is an algorithm originally developed by Larry Page and Sergey Brin, co-founders of Google. It was originally developed to rank web pages in search engine results. The PageRank score for a given node is calculated based on the number and quality of the edges pointing to that node, as well as the importance of the nodes that are connected to it. The PageRank algorithm assigns a higher score to nodes that are linked to other high-scoring nodes, and a lower score to nodes that are linked to low-scoring nodes.

The output of PageRank can be visualized as a ranking metric for the importance of a node within a given graph, with the most important nodes having the highest score, and the least important node having the lowest score. PageRank is used in search engines to rank web pages based on their importance and influence, in citation networks to identify highly cited scientific papers, and in recommendation systems to suggest popular and relevant content to users.

The space complexity of the .pageRank algorithm is:

```
O(|V|) + [storage of data O(|V|+|E|)]
```

In more detail, this is:

```
|V| * (3 * sizeof(local id type) +
    2 * sizeof(double) +
    sizeof(global id type)) +
    0(|1DData Parts|)
```

.pageRank syntax

```
CALL neptune.algo.pageRank(
  [list of source nodes (required)],
  {
    numOfIterations: a small positive integer like 20 (optional),
    dampingFactor: a positive float less than 1.0, like 0.85 (optional)
    edgeLabels: [a list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    concurrency: the number of threads to use to run the algorithm (optional)
    }
    YIELD node, rank
RETURN node, rank
```

.pageRank inputs

• a source vertex list (required) – type: Node[] or NodeId[]; default: none.

The node or nodes to use as the starting locations for the algorithm. Each node in the list triggers an execution of the algorithm. If an empty list is provided, the query result is also empty.

If the algorithm is called following a MATCH clause (query algo integration), the source query list is the result returned by the MATCH clause.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; default: none.

A vertex label for vertex filtering. If a vertex label is provided, vertices matching the label are the only vertices that are included, including vertices in the input list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "oubound" or "inbound".

• **numOfIterations** (optional) – type: a positive integer greater than zero; default: 20.

The number of iterations to perform to reach convergence. A number between 10 and 20 is recommended.

dampingFactor (optional) – type: a positive floating-point number less than 1.0; default:
 0.85.

A positive floating-point damping factor between 0.0 and 1.0 that expresses the probability, at any step, that the node will continue.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be

Outputs for the .pageRank algorithm

- **node** A key column of the input nodes.
- rank A key column of the corresponding page-rank scores for those nodes.

If the input nodes list is empty, the output is empty.

Query examples for .pageRank

This is a standalone example, where the source vertex list is explicitly specified in the query.

```
CALL neptune.algo.pageRank(
  ["101"],
  {
    numOfIterations: 1,
    dampingFactor: 0.85,
    edgeLabels: ["route"]
  }
)
```

This is a query integration examples, where .pageRank follows a MATCH clause and uses frontier injection to take the output of the MATCH clause as its list of source nodes:

```
MATCH (n)
CALL neptune.algo.pageRank(
    n,
    {
        dampingFactor: 0.85,
        numOfIterations: 1,
        edgeLabels: ["route"]
    }
)
YIELD rank
RETURN n, rank
```

This query is an example of constraining the results of .pageRank based on the PageRank values, and returning them in ascending order:

```
MATCH (n)
CALL neptune.algo.pageRank(
```

```
n,
{
    numOfIterations: 10,
    dampingFactor: 0.85,
    vertexLabel: "airport",
    edgeLabels: ["route"]
  }
)
YIELD rank WHERE rank > 0.004
RETURN n, rank ORDER BY rank
```

Sample .pageRank output

Here is an example of the output returned by .pageRank when run against the <u>sample air-routes</u> <u>dataset</u> using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.pageRank(n) YIELD node, rank RETURN node, rank
 LIMIT" \
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "2709",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 65.4809036254883,
          "elev": 49,
          "longest": 8711,
          "city": "Nadym",
          "type": "airport",
          "region": "RU-YAN",
          "desc": "Nadym Airport",
          "code": "NYM",
          "lon": 72.6988983154297,
          "country": "RU",
          "icao": "USMM",
```

```
"runways": 1
        }
      },
      "rank": 0.00016044313088059425
    },
    {
      "node": {
        "~id": "3747",
        "~entityType": "node",
        "~labels": ["continent"],
        "~properties": {
          "code": "AN",
          "type": "continent",
          "desc": "Antarctica"
        }
      },
      "rank": 0.0000404242
    }
  ]
}
```

PageRank mutate centrality algorithm

The ranking metric computed by .pageRank.mutate can indicate the importance of a node within a given graph, with the most important nodes having the highest score, and the least important node having the lowest score. PageRank is used in search engines to rank web pages based on their importance and influence, in citation networks to identify highly cited scientific papers, and in recommendation systems to suggest popular and relevant content to users.

The mutate variant of the PageRank algorithm performs the PageRank calculation over the entire graph unless the configuration parameters establish a filter, and each traversed node's calculated PageRank value is stored on that node as a property.

pageRank.mutate inputs

Inputs for the pageRank.mutate algorithm are passed in a configuration object parameter that contains:

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• writeProperty (required) – type: string; default: none.

A name for the new vertex property that will contain the computed PageRank values. If a property of that name already exists, it is overwritten.

• **vertexLabel** (optional) – type: string; default: none.

A vertex label for vertex filtering. If a vertex label is provided, vertices matching the label are the only vertices that are included, including vertices in the input list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "oubound" or "inbound".

• **numOfIterations** (optional) – type: a positive integer greater than zero; default: 20.

The number of iterations to perform to reach convergence. A number between 10 and 20 is recommended.

dampingFactor (optional) – type: a positive floating-point number less than 1.0; default:
 0.85.

A positive floating-point damping factor between 0.0 and 1.0 that expresses the probability, at any step, that the node will continue.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the pageRank.mutate algorithm

The computed PageRank values are written to a new vertex property on each node using the property name specified by the writeProperty input parameter.

A single Boolean success value (true or false) is returned, which indicates whether or not the writes succeeded.

Query example for pageRank.mutate

The example below computes the PageRank score of every vertex in the graph, and writes that score to a new vertex property named P_RANK:

```
CALL neptune.algo.pageRank.mutate(
    {
        writeProperty:"P_RANK",
        dampingFactor: 0.85,
        numOfIterations: 1,
        edgeLabels: ["route"]
    }
)
```

This query illustrates how you could then access the PageRank values in the P_RANK vertex property. It counts how many nodes have a P_RANK property value greater than the "SEA" node's P_RANK property value:

```
MATCH (n) WHERE n.code = "SEA" WITH n.P_RANK AS lowerBound
MATCH (m) WHERE m.P_RANK > lowerBound
RETURN count(m)
```

Sample .pageRank.mutate output

Here is an example of the output returned by .pageRank.mutate when run against the <u>sample</u> air-routes dataset using this query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.pageRank.mutate({writeProperty: 'prscore'}) YIELD
success RETURN success" \
    --language open_cypher \
    /tmp/out.txt
{
    results": [
        { "success": true }
    ]
}
```

Closeness centrality algorithm

The closeness centrality algorithm computes a Closeness Centrality (CC) metric for specified nodes in a graph. The CC metric of a node can be used as a positive measure of how close it is to all other nodes or how central it is in the graph.

The CC metric can be interpreted to show how quickly all other nodes in a network can be reached from a given node, and how important it is as a central hub for rapid information flow. It can be used in transportation networks to identify key hub locations, and in disease-spread modeling to pinpoint central points for targeted intervention efforts.

The closeness centrality (CC) score of a node is calculated based on the sum of its distances to all other nodes. The CC score itself is the inverse of that number; in other words, one divided by that sum. In practice, the calculation is commonly normalized to use the average length of the shortest paths rather than the actual sum of their lengths.

.closenessCentrality syntax

```
CALL neptune.algo.closenessCentrality(
   [node list (required)],
   {
     numSources: the number of BFS sources to use for computing the CC (required)
     edgeLabels: [a list of edge labels for filtering (optional)],
     vertexLabel: a node label for filtering (optional),
     traversalDirection: traversal direction (optional),
     normalize: Boolean, set to false to prevent normalization (optional)
     concurrency: the number of cores to be used to run the algorithm (optional)
     }
     YIELD node, score
RETURN node, score
```

Inputs for the .closenessCentrality algorithm

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The node or nodes to use as the starting locations for the algorithm. Each node in the list triggers an execution of the algorithm. If an empty list is provided, the query result is also empty.

If the algorithm is called following a MATCH clause (query algo integration), the source query list is the result returned by the MATCH clause.

- a configuration object that contains:
 - **numSources** (required) type: unsigned long; default: none.

The number of BFS sources for computing approximate Closeness Centrality (CC). To compute exact closeness centrality, set numSources to a number larger than number of nodes, such as maxInt.

Because of the computational complexity of the algorithm for large graphs, it's generally best to specify a number in the order of thousands to ten thousands, such as 8,192.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• normalize (optional) – type: Boolean; default: true.

You can use this field to turn off normalization, which is on by default. Without normalization, only centrality scores of nodes within the same component can be meaningfully compared. Normalized scores can be compared across different connected components.

The CC is normalized using the Wasserman-Faust normalization formula for unconnected graphs. If there are n vertices reachable from vertex u (including vertex u itself), the Wasserman-Faust normalized closeness centrality score of vertex u is calculated as follows:

 $(n-1)^2 / (|V| - 1) * sum(distance from u to these n vertices)$

Without normalization, the centrality score of vertex u is calculated as:

(|V| - 1) / sum(distance from u to all other vertices in the graph)

• vertexLabel (optional) - type: string; default: none.

A node label for node filtering. If a node label is provided, nodes matching the label are the only nodes that are included, including nodes in the input list.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

<u>concurrency</u> (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the .closenessCentrality algorithm

- **node** A key column of the input nodes.
- score A key column of the corresponding closeness-centrality (CC) scores for those nodes.

If the input node list is empty, the output is empty.

.closenessCentrality query examples

This is a standalone example, where the source node list is explicitly provided in the query:

```
CALL neptune.algo.closenessCentrality(
  ["101"],
  {
    numSources: 10,
    edgeLabels: ["route"],
    vertexLabel: "airport",
    traversalDirection: "outbound",
    normalize: true,
    concurrency: 1
    }
)
YIELD node, score
RETURN node, score
```

This is a query integration example, where .closenessCentrality.mutate follows a MATCH clause and uses the output of the MATCH clause as its list of source nodes:

```
Match (n)
CALL neptune.algo.closenessCentrality(
    n,
    {
        numSources: 10,
        edgeLabels: ["route"],
```

```
vertexLabel: "airport",
    traversalDirection: "outbound",
    normalize: true,
    concurrency: 1
  }
)
YIELD score
RETURN n, score
```

This is a query integration examples that returns the nodes with the 10 highest CC scores:

```
CALL neptune.algo.closenessCentrality(
   n,
   {
     edgeLabels: ["route"],
     numSources: 10
   }
)
YIELD score
RETURN n, score
ORDER BY score DESC
LIMIT 10"
```

A Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .closenessCentrality output

Here is an example of the output returned by .closenessCentrality when run against the sample air-routes dataset using this query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.closenessCentrality(n, {numSources: 10}) YIELD
node, score RETURN node, score limit 2" \
    --language open_cypher \
```

```
/tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "10",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 38.94449997,
          "elev": 313,
          "longest": 11500,
          "city": "Washington D.C.",
          "type": "airport",
          "region": "US-VA",
          "desc": "Washington Dulles International Airport",
          "code": "IAD",
          "prscore": 0.002264724113047123,
          "degree": 312,
          "lon": -77.45580292,
          "wccid": 2357352929951779,
          "country": "US",
          "icao": "KIAD",
          "runways": 4
        }
      },
      "score": 0.20877772569656373
    },
    {
      "node": {
        "~id": "12",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 40.63980103,
          "elev": 12,
          "longest": 14511,
          "city": "New York",
          "type": "airport",
          "region": "US-NY",
          "desc": "New York John F. Kennedy International Airport",
          "code": "JFK",
```

```
"prscore": 0.002885053399950266,
    "degree": 403,
    "lon": -73.77890015,
    "wccid": 2357352929951779,
    "country": "US",
    "icao": "KJFK",
    "runways": 4
    }
    },
    "score": 0.2199712097644806
    }
]
```

Closeness centrality mutatealgorithm

The closeness centrality mutate algorithm computes a Closeness Centrality (CC) metric for specified nodes in a graph. The CC metric of a node can be used as a positive measure of how close it is to all other nodes or how central it is in the graph.

The CC metric can be interpreted to show how quickly all other nodes in a network can be reached from a given node, and how important it is as a central hub for rapid information flow. It can be used in transportation networks to identify key hub locations, and in disease-spread modeling to pinpoint central points for targeted intervention efforts.

The closeness centrality (CC) score of a node is calculated based on the sum of its distances to all other vertices. The CC score itself is the inverse of that number; in other words, one divided by that sum. In practice, the calculation is commonly normalized to use the average length of the shortest paths rather than the actual sum of their lengths.

.closenessCentrality.mutate syntax

```
CALL neptune.algo.closenessCentrality.mutate(
    [node list (required)],
    {
        numSources: the number of BFS sources to use for computing the CC (required)
        writeProperty: name of the node property to write the CC score to (required)
        edgeLabels: [a list of edge labels for filtering (optional)],
        vertexLabel: "a node label for filtering (optional)",
        traversalDirection: traversal direction (optional),
        normalize: Boolean, set to false to prevent normalization (optional)
        concurrency: the number of cores to be used to run the algorithm (optional)
    }
    YIELD success
RETURN success
```

.closenessCentrality.mutate inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The node or nodes to use as the starting locations for the algorithm. Each node in the list triggers an execution of the algorithm. If an empty list is provided, the query result is also empty.

If the algorithm is called following a MATCH clause (query algo integration), the source node list is the result returned by the MATCH clause.

- a configuration object that contains:
 - **numSources** (required) type: uint64_t; default: none.

The number of BFS sources for computing approximate Closeness Centrality (CC). To compute exact closeness centrality, set numSources to a number larger than number of vertices, such as maxInt.

Because of the computational complexity of the algorithm for large graphs, it's generally best to specify a number in the order of thousands to ten thousands, such as 8,192.

• writeProperty (required) – type: string; default: none.

A name for the new node property that will contain the computed CC score of each node.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **normalize** (optional) – type: Boolean; default: true.

You can use this field to turn off normalization, which is on by default. Without normalization, only centrality scores of nodes within the same component can be meaningfully compared. Normalized scores can be compared across different connected components.

The CC is normalized using the Wasserman-Faust normalization formula for unconnected graphs. If there are n vertices reachable from vertex u (including vertex u itself), the Wasserman-Faust normalized closeness centrality score of vertex u is calculated as follows:

 $(n-1)^2 / (|V| - 1) * sum(distance from u to these n vertices)$

Without normalization, the centrality score of vertex u is calculated as:

(|V| - 1) / sum(distance from u to all other vertices in the graph)

• vertexLabel (optional) - type: string; default: none.

A node label for node filtering. If a node label is provided, vertices matching the label are the only vertices that are included, including vertices in the input list.

• traversalDirection (optional) – type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.closenessCentrality.mutate outputs

The closeness centrality score of each source node in the input list is written as a new node property using the property name specified in writeProperty.

If the algorithm is invoked as a standalone query, there is no other output.

If the algorithm is invoked following a MATCH clause that provides its source node list (query integration), the algorithm outputs a key column of the source vertices from the MATCH clause and a value column of Booleans (true or false) that indicate whether the CC value was successfully written to the node in question.

Query examples for .closenessCentrality.mutate

This example computes closeness centrality scores and writes them as a new node property called ccScore:

```
CALL neptune.algo.closenessCentrality.mutate(
    {
        numSources: 10,
        writeProperty: "ccScore",
        edgeLabels: ["route"],
        vertexLabel: "airport",
        traversalDirection: "outbound",
        normalize: true,
        concurrency: 1
```

)

}

Then you can query the ccScore property in a subsequent query:

```
MATCH (n) RETURN id(n), n.ccScore limit 5
```

Sample .closenessCentrality output

Here is an example of the output returned by .closenessCentrality when run against the sample air-routes dataset using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.closenessCentrality.mutate(
       {
         writeProperty: 'ccscore',
         numSources: 10
       }
     )
     YIELD success
     RETURN success"
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "success": true
    }
  ]
}
```

Similarity algorithms in Neptune Analytics

Graph similarity algorithms allow you to compare and analyze the similarities and dissimilarities between different graph structures, which can provide insight into relationships, patterns, and commonalities across diverse datasets. This is invaluable in various fields, such as biology, for comparing molecular structures, such as social networks, for identifying similar communities, and such as recommendation systems, for suggesting similar items based on user preferences.

Neptune Analytics supports the following similarity algorithms:

• <u>neighbors.common</u> – This algorithm counts the number of common neighbors of two input vertices, which is the intersection of the neighborhoods of those vertices.

By counting how many neighboring nodes are shared by two nodes, it provides a measure of their potential interaction or similarity within the network. It's used in social network analysis to identify individuals with mutual connections, in citation networks to find influential papers referenced by multiple sources, and in transportation networks to locate critical hubs with many direct connections to other nodes.

- <u>neighbors.total</u> This algorithm counts the number of total unique neighbors among two input vertices, which is the union of the neighborhoods of those vertices.
- <u>jaccardSimilarity</u> This algorithm measures the similarity between two sets by dividing the size of their intersection by the size of their union.

By measuring the proportion of shared neighbors relative to the total number of unique neighbors, it provides a metric for understanding the degree of overlap or commonality between different parts of a network. Jaccard similarity is applied in recommendation systems to suggest products or content to users based on their shared preferences and in biology to compare genetic sequences for identifying similarities in DNA fragments.

<u>overlapSimilarity</u> – This algorithm measures the overlap between the neighbors of two vertices.

It quantifies the similarity between nodes by calculating the ratio of common neighbors they share to the total number of neighbors they collectively have, providing a measure of their closeness or similarity within the network. Overlap similarity is applied in social network analysis to identify communities of individuals with shared interests or interactions, and in biological networks to detect common functionalities among proteins in molecular pathways.

Common neighbors algorithm

Common neighbors is an algorithm that counts the number of common neighbors of two input nodes, which is the intersection of their neighborhoods. This provides a measure of their potential interaction or similarity within the network. The common neighbors algorithm is used in social network analysis to identify individuals with mutual connections, in citation networks to find influential papers referenced by multiple sources, and in transportation networks to locate critical hubs with many direct connections to other nodes.

.neighbors.common syntax

```
CALL neptune.algo.neighbors.common(
  [first node(s)],
  [second node(s)],
  {
    edgeLabels: [a list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
  }
)
YIELD common
RETURN firstNodes, secondNodes, common
```

.neighbors.common inputs

• first node(s) (required) - type: Node[] or NodeId[]; default: none.

One or more nodes of which to find the common neighbors with the corresponding second node(s).

• **second node(s)** (required) – type: Node[] or NodeId[]; default: none.

One or more nodes of which to find the common neighbors with the corresponding first node(s).

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

A node label for node filtering. If a node label is provided, nodes matching the label are the only nodes that are considered neighbors. This does not filter the nodes in the first or second node lists.

.neighbors.common outputs

common: A row for each node in the first node list and corresponding node in the second node list, and the number of neighboring nodes they have in common.

If either input node list is empty, the output is empty.

.neighbors.common query examples

This example specifies only two nodes:

```
MATCH (sydairport:airport {code: 'SYD'})
MATCH (jfkairport:airport {code: 'JFK'})
CALL neptune.algo.neighbors.common( sydairport, jfkairport, { edgeLabels: ['route'] })
YIELD common
RETURN sydairport, jfkairport, common
```

This example specifies multiple nodes. It returns a row for each combination of a US airport and a UK airport, and the number of destinations we could reach from both of those two airports:

```
MATCH (usairports:airport {country: 'US'})
MATCH (ukairports:airport {country: 'UK'})
CALL neptune.algo.neighbors.common(usairports, ukairports, {edgeLabels: ['route']})
YIELD common
RETURN usairports, ukairports, common
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .neighbors.common output

Here is an example of the output returned by .neighbors.common when run against the <u>sample</u> air-routes dataset using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH (sydairport:airport {code: 'SYD'})
                        MATCH (jfkairport:airport {code: 'JFK'})
                        CALL neptune.algo.neighbors.common(sydairport, jfkairport,
 {edgeLabels: ['route']})
                       YIELD common
                        RETURN sydairport, jfkairport, common" \
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "sydairport": {
        "~id": "55",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": -33.9460983276367,
          "elev": 21,
          "type": "airport",
          "code": "SYD",
          "lon": 151.177001953125,
          "runways": 3,
          "longest": 12999,
          "communityId": 2357352929951971,
          "city": "Sydney",
          "region": "AU-NSW",
          "desc": "Sydney Kingsford Smith",
          "prscore": 0.0028037719894200565,
          "degree": 206,
          "wccid": 2357352929951779,
          "ccscore": 0.19631840288639069,
          "country": "AU",
          "icao": "YSSY"
        }
```

```
},
      "jfkairport": {
        "~id": "12",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 40.63980103,
          "elev": 12,
          "type": "airport",
          "code": "JFK",
          "lon": -73.77890015,
          "runways": 4,
          "longest": 14511,
          "communityId": 2357352929951971,
          "city": "New York",
          "region": "US-NY",
          "desc": "New York John F. Kennedy International Airport",
          "prscore": 0.002885053399950266,
          "degree": 403,
          "wccid": 2357352929951779,
          "ccscore": 0.2199712097644806,
          "country": "US",
          "icao": "KJFK"
        }
      },
      "common": 24
    }
  ]
}
```

Total neighbors algorithm

Total neighbors is an algoithm that counts the total number of unique neighbors of two input vertices, which is the union of the neighborhoods of those vertices.

.neighbors.total syntax

```
CALL neptune.algo.neighbors.total(
  [first node(s)],
  [second node(s)],
  {
    edgeLabels: [a list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
  }
)
YIELD total
RETURN firstNodes, secondNodes, total
```

Inputs for the .neighbors.total algorithm

• first node(s) (required) - type: Node[] or NodeId[]; default: none.

One or more nodes of which to find the common neighbors with the corresponding second nodes.

• second node(s) (required) - type: Node[] or NodeId[]; default: none.

One or more nodes of which to find the common neighbors with the corresponding first nodes.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

A node label for node filtering. If a node label is provided, nodes matching the label are the only nodes that are considered neighbors. This does not filter the nodes in the first or second node lists.

.neighbors.total outputs

total: A row for each node in the first node list and corresponding node in the second node list, and the total number of neighboring nodes they have.

If either input node list is empty, the output is empty.

.neighbors.total query examples

This example returns a row for each combination of a US airport and a UK airport, and the total number of destinations we could reach if we could fly out of either of the two airports.

```
MATCH (usairports:airport {country: 'US'})
MATCH (ukairports:airport {country: 'UK'})
CALL neptune.algo.neighbors.total(usairports, ukairports, {edgeLabels: ['route']})
YIELD total
RETURN usairports, ukairports, total"
```



It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .neighbors.total output

Here is an example of the output returned by .neighbors.total when run against the <u>sample</u> air-routes dataset using this query:

```
cat /tmp/out.txt
{
  "results": [
    {
      "sydairport": {
        "~id": "55",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": -33.9460983276367,
          "elev": 21,
          "type": "airport",
          "code": "SYD",
          "lon": 151.177001953125,
          "runways": 3,
          "longest": 12999,
          "communityId": 2357352929951971,
          "city": "Sydney",
          "region": "AU-NSW",
          "desc": "Sydney Kingsford Smith",
          "prscore": 0.0028037719894200565,
          "degree": 206,
          "wccid": 2357352929951779,
          "ccscore": 0.19631840288639069,
          "country": "AU",
          "icao": "YSSY"
        }
      },
      "jfkairport": {
        "~id": "12",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 40.63980103,
          "elev": 12,
          "type": "airport",
          "code": "JFK",
          "lon": -73.77890015,
          "runways": 4,
          "longest": 14511,
          "communityId": 2357352929951971,
          "city": "New York",
          "region": "US-NY",
```

```
"desc": "New York John F. Kennedy International Airport",
    "prscore": 0.002885053399950266,
    "degree": 403,
    "wccid": 2357352929951779,
    "ccscore": 0.2199712097644806,
    "country": "US",
    "icao": "KJFK"
    }
    },
    "total": 279
  }
}
```

Jaccard similarity algorithm

The Jaccard similarity algorithm measures the similarity between two sets. It is calculated by dividing the size of the intersection of the two sets by the size of their union.

By measuring the proportion of shared neighbors relative to the total number of unique neighbors, this algorithm provides a metric for the degree of overlap or commonality between different parts of a network. It can be useful in recommendation systems to suggest products or content to users based on their shared preferences and in biology to compare genetic sequences for identifying similarities in DNA fragments.

.jaccardSimilarity syntax

```
CALL neptune.algo.jaccardSimilarity(
  [first node(s)],
  [second node(s)],
  {
    edgeLabels: [a list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    }
)
YIELD score
RETURN firstNodes, secondNodes, score
```

.jaccardSimilarity inputs

• first node(s) (required) - type: Node[] or NodeId[]; default: none.

One or more nodes for which to find the Jaccard similarity score with respect to the corresponding second node(s).

• second node(s) (required) - type: Node[] or NodeId[]; default: none.

One or more nodes for which to find the Jaccard similarity score with respect to the corresponding first node(s).

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

A node label for node filtering. If a node label is provided, nodes matching the label are the only nodes that are considered neighbors. This does not filter the nodes in the first or second node lists.

Outputs for the .jaccardSimilarity algorithm

score: A row for each node in the first node list and corresponding node in the second node list, and the Jaccard similarity score for the two.

If either input node list is empty, the output is empty.

.jaccardSimilarity query examples

The example below is a query integration examples, where the node list inputs for .jaccardSimilarity come from a preceding MATCH clause:

```
MATCH (n1:Person {name: "Alice"}), (n2:Person {name: "Bob"})
CALL neptune.algo.jaccardSimilarity(n1, n2, {edgeLabels: ['knows']})
YIELD score
RETURN n1, n2, score
```

Another example:

```
MATCH (n {code: "AUS"})
MATCH (m {code: "FLL"})
CALL neptune.algo.jaccardSimilarity(
    n,
    m,
    {
       edgeLabels: ["route"],
       vertexLabel: "airport"
    }
)
YIELD score
RETURN n, m, score
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .jaccardSimilarity output

Here is an example of the output returned by .jaccardSimilarity when run against the <u>sample</u> <u>air-routes dataset</u> using this query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH (n {code: 'AUS'})
                       MATCH (m {code: "FLL"})
                       CALL neptune.algo.jaccardSimilarity(n, m,
                            {edgeLabels: [\"route\"], vertexLabel: \"airport\"})
                       YIELD score
                        RETURN n, m, score"
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "n": {
        "~id": "3",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 30.1944999694824,
          "elev": 542,
          "type": "airport",
          "code": "AUS",
          "lon": -97.6698989868164,
          "runways": 2,
          "longest": 12250,
          "communityId": 2357352929951971,
          "city": "Austin",
```

```
"region": "US-TX",
        "desc": "Austin Bergstrom International Airport",
        "prscore": 0.0012390684569254518,
        "degree": 188,
        "wccid": 2357352929951779,
        "ccscore": 0.1833982616662979,
        "country": "US",
        "icao": "KAUS"
      }
    },
    "m": {
      "~id": "9",
      "~entityType": "node",
      "~labels": ["airport"],
      "~properties": {
        "lat": 26.0725994110107,
        "elev": 64,
        "type": "airport",
        "code": "FLL",
        "lon": -80.152702331543,
        "runways": 2,
        "longest": 9000,
        "communityId": 2357352929951971,
        "city": "Fort Lauderdale",
        "region": "US-FL",
        "desc": "Fort Lauderdale/Hollywood International Airport",
        "prscore": 0.0024497462436556818,
        "degree": 316,
        "wccid": 2357352929951779,
        "ccscore": 0.19741515815258027,
        "country": "US",
        "icao": "KFLL"
      }
    },
    "score": 0.2953367829322815
  }
]
```

}

Overlap similarity algorithm

Overlap Similarity is an algorithm that measures the overlap between the neighbors of two nodes. It does this by dividing the intersection of the two neighborhoods by the neighbor with minimum degree.

By calculating the ratio of common neighbors shared by two nodes to the total number of neighbors they collectively have, it provides a measure of their closeness or similarity within the network. Overlap similarity is applied in social network analysis to identify communities of individuals with shared interests or interactions, and in biological networks to detect common functionalities among proteins in molecular pathways.

.overlapSimilarity syntax

```
CALL neptune.algo.overlapSimilarity(
  [first node(s)],
  [second node(s)],
  {
    edgeLabels: [a list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
  }
)
YIELD score
RETURN firstNodes, secondNodes, score
```

.overlapSimilarity inputs

• first node(s) (required) - type: Node[] or NodeId[]; default: none.

One or more nodes for which to find the overlap similarity score with respect to the corresponding second node(s).

• **second node(s)** (required) – type: Node[] or NodeId[]; default: none.

One or more nodes for which to find the overlap similarity score with respect to the corresponding first node(s).

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

A node label for node filtering. If a node label is provided, nodes matching the label are the only nodes that are considered neighbors. This does not filter the nodes in the first or second node lists.

.overlapSimilarity outputs

score: A row for each node in the first node list and corresponding node in the second node list, and the overlap similarity score for the two.

If either input node list is empty, the output is empty.

.overlapSimilarity query examples

This is a query integration examples, where .overlapSimilarity takes its input node lists from the output of a MATCH clause:

```
MATCH (n1:Person {name: "Alice"}), (n2:Person {name: "Bob"})
CALL neptune.algo.overlapSimilarity(n1, n2, {edgeLabel: 'knows'})
YIELD score
RETURN n1, n2, score
```

Another example:

```
MATCH (n {code: "AUS"})
MATCH (m {code: "FLL"})
CALL neptune.algo.overlapSimilarity(
    n,
    m,
    {
       edgeLabels: ["route"],
       vertexLabel: "airport"
    }
)
YIELD score
RETURN n, m, score'
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .overlapSimilarity output

Here is an example of the output returned by .overlapSimilarity when run against the <u>sample</u> <u>air-routes dataset</u> using this query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string 'MATCH (n {code: "AUS"})
                        MATCH (m {code: "FLL"})
                        CALL neptune.algo.overlapSimilarity(
                          n,
                          m,
                          {
                            edgeLabels: ["route"],
                            vertexLabel: "airport"
                          }
                        )
                        YIELD score
                        RETURN n, m, score' ∖
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "n": {
        "~id": "3",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 30.1944999694824,
          "elev": 542,
          "type": "airport",
```

```
"code": "AUS",
        "lon": -97.6698989868164,
        "runways": 2,
        "longest": 12250,
        "communityId": 2357352929951971,
        "city": "Austin",
        "region": "US-TX",
        "desc": "Austin Bergstrom International Airport",
        "prscore": 0.0012390684569254518,
        "degree": 188,
        "wccid": 2357352929951779,
        "ccscore": 0.1833982616662979,
        "country": "US",
        "icao": "KAUS"
      }
    },
    "m": {
      "~id": "9",
      "~entityType": "node",
      "~labels": ["airport"],
      "~properties": {
        "lat": 26.0725994110107,
        "elev": 64,
        "type": "airport",
        "code": "FLL",
        "lon": -80.152702331543,
        "runways": 2,
        "longest": 9000,
        "communityId": 2357352929951971,
        "city": "Fort Lauderdale",
        "region": "US-FL",
        "desc": "Fort Lauderdale/Hollywood International Airport",
        "prscore": 0.0024497462436556818,
        "degree": 316,
        "wccid": 2357352929951779,
        "ccscore": 0.19741515815258027,
        "country": "US",
        "icao": "KFLL"
      }
    },
    "score": 0.6129032373428345
  }
1
```

}

Clustering and community detection algorithms in Neptune Analytics

Clustering algorithms evaluate how nodes are clustered in communities, in closely-knit sets, or in highly or loosely interconnected groups.

These algorithms can identify meaningful groups or clusters of nodes in a network, revealing hidden patterns and structures that can provide insights into the organization and dynamics of complex systems. This is valuable in social network analysis and in biology, for identifying functional modules in protein-protein interaction networks, and more generally for understanding information flow and influence propagation in many different domains.

Neptune Analytics supports these community detection algorithms:

 wcc – The Weakly Connected Components (WCC) algorithm finds weakly-connected components in a directed graph. A weakly-connected component is a group of nodes where every node in the group is reachable from every other mode in the group if edge direction is ignored.

Identifying weakly-conected components helps in understanding the overall connectivity and structure of the graph. Weakly-connected components can be used in transportation networks to identify disconnected regions that may require improved connectivity, and in social networks to find isolated groups of users with limited interactions, and in webpage analysis to pinpoint sections with low accessibility.

- wcc.mutate This algorithm stores the calculated component value of each given node as a property of the node.
- <u>labelPropagation</u> Label Propagation Algorithm (LPA) is an algorithm for community detection that is also used in semi-supervised machine learning for data classification.
- <u>labelPropagation.mutate</u> Label Propagation Algorithm (LPA) is an algorithm tha assigns labels to nodes based on the consensus of their neighboring nodes, making it useful for identifying groups. Label propagation can be applied in social networks to find groups, and in identity management to identify households, and in recommendation systems to group similar products for personalized suggestions. It can also be used in semi-supervised machine learning for data classification.
- <u>scc</u> The Strongly Connected Components (SCC) algorithm identifies maximally connected subgraphs of a directed graph, where every node is reachable from every other node. This can provide insights into the tightly interconnected portions of a graph and highlight key structures

within it. Strongly connected components are valuable in computer programming for detecting loops or cycles in code, in social networks to find tightly connected groups of users who interact frequently, and in web crawling to identify clusters of interlinked pages for efficient indexing.

• <u>scc.mutate</u> – This algorithm finds the maximally connected subgraphs of a directed graph and writes their component IDs as a new property of each subgraph node.

Weakly connected components algorithm

The Weakly Connected Components (WCC) algorithm finds the weakly-connected components in a directed graph. A weakly-connected component is a group of nodes in which every node is reachable from every other node when edge directions are ignored. Weakly connected components are the maximal connected subgraphs of an undirected graph.

Identifying weakly-conected components helps in understanding the overall connectivity and structure of the graph. Weakly-connected components can be used in transportation networks to identify disconnected regions that may require improved connectivity, and in social networks to find isolated groups of users with limited interactions, and in webpage analysis to pinpoint sections with low accessibility.

The time complexity of the WCC algorithm is 0(|E|logD), where |E| is the number of edges in the graph, and D is the diameter (the length of the longest path from one node to any other node) of the graph.

The memory used by the WCC algorithm is approximately |V| * 20 bytes.

.wcc syntax

```
CALL neptune.algo.wcc(
  [source-node list (required)],
  {
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    concurrency: number of threads to use (optional)
    }
)
YIELD node, component
RETURN node, component
```

.wcc inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The node or nodes to use as the starting locations for the algorithm. Each node in the list triggers an execution of the algorithm. If an empty list is provided, the query result is also empty.

If the algorithm is called following a MATCH clause (query algo integration), the source query list is the result returned by the MATCH clause.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; default: none.

A node label for node filtering. If a node label is provided, only nodes matching the label are considered. This includes the nodes in the source node lists.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.wcc outputs

For each source node:

- **node** The source node.
- component The component ID associated with the source node.

If the input node list is empty, the output is empty.

.wcc query examples

This is a standalone example, where the source node list is explicitly provided in the query:

```
CALL neptune.algo.wcc(
 ["101"],
 {
    edgeLabels: ["route"],
    vertexLabel: "airport",
    concurrency: 2
 }
)
```

YIELD node, component RETURN node, component

This is a query integration examples, where .wcc follows a MATCH clause and uses the output of the MATCH clause as its source node list:

```
MATCH (n) WHERE n.region = 'US-WA'
CALL neptune.algo.wcc(
    n,
    {
       edgeLabels: ["route"],
       vertexLabel: "airport"
    }
)
YIELD component
RETURN n, component
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .wcc output

Here is an example of the output returned by .wcc when run against the <u>sample air-routes dataset</u> using the following query:

Neptune Analytics

```
{
  "results": [
    {
      "node": {
        "~id": "10",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 38.94449997,
          "elev": 313,
          "longest": 11500,
          "city": "Washington D.C.",
          "type": "airport",
          "region": "US-VA",
          "desc": "Washington Dulles International Airport",
          "code": "IAD",
          "prscore": 0.002264724113047123,
          "lon": -77.45580292,
          "wccid": 2357352929951779,
          "country": "US",
          "icao": "KIAD",
          "runways": 4
        }
      },
      "component": 2357352929951779
    },
    {
      "node": {
        "~id": "12",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 40.63980103,
          "elev": 12,
          "longest": 14511,
          "city": "New York",
          "type": "airport",
          "region": "US-NY",
          "desc": "New York John F. Kennedy International Airport",
          "code": "JFK",
          "prscore": 0.002885053399950266,
          "lon": -73.77890015,
          "wccid": 2357352929951779,
          "country": "US",
```

```
"icao": "KJFK",
"runways": 4
}
},
"component": 2357352929951779
}
]
}
```

Weakly connected components mutate algorithm

The mutate variant of the weakly connected components (WCC) algorithm performs the weakly connected components calculation over the entire graph unless the configuration parameters establish a filter, and each traversed node's calculated WCC value is stored as a property on the node.

.wcc.mutate syntax

```
CALL neptune.algo.wcc.mutate(
    {
        writeProperty: the name for the node property to which to write component IDs
        edgeLabels: [list of edge labels for filtering (optional)],
        vertexLabel: a node label for filtering (optional),
        concurrency: number of threads to use (optional)
     }
     //
YIELD success
RETURN success
```

.wcc.mutate inputs

Inputs for .wcc.mutate are passed in a configuration object that contains:

• writeProperty (required) - type: string; default: none.

A name for the new node property where the component IDs will be written.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; default: none.

The node label to filter on for traversing. Only nodes matching this label will be traversed. For example: "airport".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.wcc.mutate outputs

success: The computed component IDs are written as a new property on each node using the property name specified by writeProperty, and a single success flag (true or false) is returned to indicate whether or not the writes succeeded.

.wcc.mutate query examples

This query writes the calculated component ID of each vertex in the graph to a new property of the vertex named CCID:

```
CALL neptune.algo.wcc.mutate(
    {
        writeProperty: "CCID",
        edgeLabels: ["route"],
        vertexLabel: "airport",
        concurrency: 2
    }
)
```

After the mutate algorithm call above, the following query can retrieve the CCID property of a specific node:

```
MATCH (n: airport {code: "SEA"})
RETURN n.CCID
```

Sample .wcc.mutate output

Here is an example of the output returned by .wcc.mutate when run against the <u>sample air-</u> routes dataset using the following query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.wcc.mutate({writeProperty: 'wccid'}) YIELD success
RETURN success"
```

```
--language open_cypher \
   /tmp/out.txt
cat /tmp/out.txt
{
    "results": [
      {
        "success": true
      }]
}
```

Label propagation algorithm (LPA)

Label Propagation Algorithm (LPA) is an algorithm for community detection that is also used in semi-supervised machine learning for data classification.

A community structure is loosely defined as a tightly knit group of entities in social networks. LPA can be enhanced by providing a set of seed nodes, the quality of which can dramatically influence the solution quality of the found communities. If the seeds are well-selected, the quality of the solution can be good, but if not, the quality of the solution can be very bad.

See Xu T. Liu *et al*, <u>Direction-optimizing label propagation and its application to community</u> <u>detection</u>, and Xu T. Liu *et al*, <u>Direction-optimizing Label Propagation Framework for Structure</u> <u>Detection in Graphs: Design, Implementation, and Experimental Analysis</u>, and the <u>Neo4j Label</u> <u>Propagation API</u>.

The time complexity of the algorithm is O(k|E|), where |E| is the number of edges in the graph, and k is the number of iterations for the algorithm to converge. Its space complexity is O(|V|), where |V| is the number of nodes in the graph.

.labelPropagation syntax

```
CALL neptune.algo.labelPropagation(
  [source-node list (required)],
  {
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    vertexWeightProperty: a numeric node property used to weight the community ID
 (optional),
    vertexWeightType: numeric type of the specified vertexWeightProperty (optional),
    edgeWeightProperty: a numeric edge property used to weight the community ID
 (optional),
    edgeWeightType: numeric type of the specified edgeWeightProperty (optional),
    maxIterations: the maximum number of iterations to run (optional, default: 10),
    traversalDirection: traversal direction (optional, default: outbound),
    concurrency: number of threads to use (optional)
  }
)
Yield node, community
Return node, community
```

.labelPropagation inputs

• a source node list (required) – type: Node[] or NodeId[]; default: none.

The node or nodes to use as the starting locations for the algorithm. Each node in the list triggers an execution of the algorithm. If an empty list is provided, the query result is also empty.

If the algorithm is called following a MATCH clause (query algo integration), the source query list is the result returned by the MATCH clause.

- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) - type: string; default: none.

A node label for node filtering. If a node label is provided, nodes matching the label are the only nodes that are included in the calculation, including nodes in the input list.

• vertexWeightProperty (optional) - type: string; default: none.

The node weight used in Label Propagation. When vertexWeightProperty is not specified, each node's communityId is treated equally, as if the node weight were 1.0. When the vertexWeightProperty is specified without an edgeWeightProperty, the weight of the communityId for each node is the value of the node weight property. When both vertexWeightProperty and edgeWeightProperty are specified, the weight of the communityId is the product of the node property value and edge property value.

Note that if multiple properties exist on the node with the name specified by vertexWeightProperty, one of those property values will be sampled at random.

vertexWeightType (required if vertexWeightProperty is presnt) – type: string; valid values: "int", "long", "float", "double"; default: empty.

The type of the numeric values in the node property specified by vertexWeightProperty.

If vertexWeightProperty is not provided, vertexWeightType is ignored. If a node contains a numeric property with the name specified by vertexWeightProperty but

its value is a different numeric type than is specified by vertexWeightType, the value is .labelPropagation

typecast to the type specified by vertexWeightType. If both vertexWeightType and edgeWeightType are given, the type specified by edgeWeightType is used for both node and edge properties.

• edgeWeightProperty (optional) - type: string; default: none.

The numeric edge property used as a weight in Label Propagation. When vertexWeightProperty is not specified, the default edge weight is 1.0, so each edge is treated equally. When only edgeWeightProperty is provided, the weight of the communityId is the value of that edge property. When both vertexWeightProperty and edgeWeightProperty are present, the weight of a communityId is the product of the edge property value and the node property value.

Note that if multiple properties exist on the edge with the name specified by edgeWeightProperty, one of those property values will be sampled at random.

 edgeWeightType (required if edgeWeightProperty is presnt) – type: string; valid values: "int", "unsignedInt", "long", "unsignedLong", "float", "double"; default: none.

The type of the numeric values in the edge property specified by edgeWeightProperty.

If edgeWeightProperty is not provided, edgeWeightType is ignored. If a node contains a numeric property with the name specified by edgeWeightProperty but its value is a different numeric type than is specified by edgeWeightType, the value is typecast to the type specified by edgeWeightType. If both vertexWeightType and edgeWeightType are given, the type specified by edgeWeightType is used for both node and edge properties.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• maxIterations (optional) – type: integer; default: 10.

The maximum number of iterations to run.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.labelPropagation outputs

- **node** A key column of the input nodes.
- community A key column of the corresponding communityId values for those nodes. All the nodes with the same communityId are in the same weakly-connected component.

If the input node list is empty, the output is empty.

.labelPropagation query examples

This is a standalone example, where the source node list is explicitly provided in the query. It runs the algorithm over the whole graph, but only queries the component ID of one node:

```
CALL neptune.algo.labelPropagation(
  ["101"],
  {
    edgeLabels: ["route"],
    maxIterations: 10,
    vertexLabel: "airport",
    vertexWeightProperty: "runways",
    vertexWeightType: "int",
    edgeWeightProperty: "dist",
    edgeWeightType: "int",
    traversalDirection: "both",
    concurrency: 2
  }
)
YIELD node, community
RETURN node, community
```

This is a query integration example, where .labelPropagation uses the output of a preceding MATCH clause as its source node list:

```
Match (n)
CALL neptune.algo.labelPropagation(
    n,
    {
        edgeLabels: ["route"],
        maxIterations: 10,
        vertexLabel: "airport",
        vertexWeightProperty: "runways",
```

```
vertexWeightType: "int",
edgeWeightProperty: "dist",
edgeWeightType: "int",
traversalDirection: "both",
concurrency: 2
}
)
YIELD community
RETURN n, community
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .pageRank output

Here is an example of the output returned by .pageRank when run against the <u>sample air-routes</u> dataset using this query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH (n) CALL neptune.algo.labelPropagation(n) YIELD node, community
 RETURN node, community LIMIT 2" \
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "10",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 38.94449997,
          "elev": 313,
          "type": "airport",
```

```
"code": "IAD",
      "lon": -77.45580292,
      "runways": 4,
      "longest": 11500,
      "communityId": 2357352929951971,
      "city": "Washington D.C.",
      "region": "US-VA",
      "desc": "Washington Dulles International Airport",
      "prscore": 0.002264724113047123,
      "degree": 312,
      "wccid": 2357352929951779,
      "ccscore": 0.20877772569656373,
      "country": "US",
      "icao": "KIAD"
    }
  },
  "community": 2357352929951971
},
{
  "node": {
    "~id": "12",
    "~entityType": "node",
    "~labels": ["airport"],
    "~properties": {
      "lat": 40.63980103,
      "elev": 12,
      "type": "airport",
      "code": "JFK",
      "lon": -73.77890015,
      "runways": 4,
      "longest": 14511,
      "communityId": 2357352929951971,
      "city": "New York",
      "region": "US-NY",
      "desc": "New York John F. Kennedy International Airport",
      "prscore": 0.002885053399950266,
      "degree": 403,
      "wccid": 2357352929951779,
      "ccscore": 0.2199712097644806,
      "country": "US",
      "icao": "KJFK"
    }
  },
  "community": 2357352929951971
```

}] }

Label propagation mutate algorithm

Label Propagation Algorithm (LPA) is an algorithm for community detection that is also used in semi-supervised machine learning for data classification.

The .labelPropagation.mutate variant of the algorithm writes the derived community component ID of each node in the source list to a new property of that node.

.labelPropagation.mutate syntax

```
CALL neptune.algo.labelPropagation(
  {
   writeProperty: the name for the node property to which to write component IDs
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    vertexWeightProperty: a numeric node property used to weight the community ID
 (optional),
    vertexWeightType: numeric type of the specified vertexWeightProperty (optional),
    edgeWeightProperty: a numeric edge property used to weight the community ID
 (optional),
    edgeWeightType: numeric type of the specified edgeWeightProperty (optional),
    maxIterations: the maximum number of iterations to run (optional, default: 10),
    traversalDirection: traversal direction (optional, default: outbound),
    concurrency: number of threads to use (optional)
  }
)
```

.labelPropagation.mutate inputs

Inputs for .wcc.mutate are passed in a configuration object that contains:

• writeProperty (required) - type: string; default: none.

A name for the new node property that will contain the computed community component ID of the node.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexWeightProperty (optional) – type: string; default: none.

The node weight used in Label Propagation. When vertexWeightProperty is not specified, each node's communityId is treated equally, as if the node weight were 1.0. When the vertexWeightProperty is specified without an edgeWeightProperty, the weight of the communityId for each node is the value of the node weight property. When both vertexWeightProperty and edgeWeightProperty are specified, the weight of the communityId is the product of the node property value and edge property value.

Note that if multiple properties exist on the node with the name specified by vertexWeightProperty, one of those property values will be sampled at random.

vertexWeightType (required if vertexWeightProperty is presnt) – type: string; valid values: "int", "long", "float", "double"; default: empty.

The type of the numeric values in the node property specified by vertexWeightProperty.

If vertexWeightProperty is not provided, vertexWeightType is ignored. If a node contains a numeric property with the name specified by vertexWeightProperty but its value is a different numeric type than is specified by vertexWeightType, the value is typecast to the type specified by vertexWeightType. If both vertexWeightType and edgeWeightType are given, the type specified by edgeWeightType is used for both node and edge properties.

• edgeWeightProperty (optional) – type: string; default: none.

The numeric edge property used as a weight in Label Propagation. When vertexWeightProperty is not specified, the default edge weight is 1.0, so each edge is treated equally. When only edgeWeightProperty is provided, the weight of the communityId is the value of that edge property. When both vertexWeightProperty and edgeWeightProperty are present, the weight of a communityId is the product of the edge property value and the node property value.

Note that if multiple properties exist on the edge with the name specified by edgeWeightProperty, one of those property values will be sampled at random.

 edgeWeightType (required if edgeWeightProperty is presnt) – type: string; valid values: "int", "unsignedInt", "long", "unsignedLong", "float", "double"; default: none.

The type of the numeric values in the edge property specified by edgeWeightProperty.

If edgeWeightProperty is not provided, edgeWeightType is ignored. If a node contains a numeric property with the name specified by edgeWeightProperty but its value is a different

numeric type than is specified by edgeWeightType, the value is typecast to the type specified by edgeWeightType. If both vertexWeightType and edgeWeightType are given, the type specified by edgeWeightType is used for both node and edge properties.

• traversalDirection (optional) - type: string; default: "outbound".

The direction of edge to follow. Must be one of: "inbound", "oubound", or "both".

• maxIterations (optional) – type: integer; default: 10.

The maximum number of iterations to run.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the .labelPropagation.mutate algorithm

The community component IDs are written as a new node property of each source node using the property name specified by writeProperty.

If the algorithm is invoked as a standalone query, there is no other output.

If the algorithm is invoked immediately after a MATCH clause that supplies its source node list, the algorithm outputs a key column of the source nodes from the MATCH clause and a value column of success flags (true or false) to indicate whether or not the write to the new node property of that node succeeded.

.labelPropagation.mutate query example

```
CALL neptune.algo.labelPropagation.mutate(
    {
        writeProperty: "COMM_ID",
        edgeLabels: ["route"],
        maxIterations: 10,
        vertexLabel: "airport",
        vertexWeightProperty: "runways",
        vertexWeightType: "int",
        edgeWeightProperty: "dist",
```

```
edgeWeightType: "int",
   traversalDirection: "both",
   concurrency: 2
 }
)
```

Sample .labelPropagation.mutate output

Here is an example of the output returned by .labelPropagation.mutate when run against the sample air-routes dataset using the following query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.labelPropagation.mutate({writeProperty:
    'communityId'}) YIELD success RETURN success" \
    --language open_cypher \
    /tmp/out.txt
{
    results": [
        {
            "results": [
            {
            "success": true
        }
      ]
    }
```

Strongly connected components algorithm

Strongly connected components (SCC) are the maximally connected subgraphs of a directed graph where every node is reachable from every other node (in other words, there exists a path between every node in the subgraph).

Neptune Analytics implements this algorithm using a modified multi-step approach (see <u>BFS and</u> <u>Coloring-based Parallel Algorithms for Strongly Connected Components and Related Problems</u>, by George M. Slota, Sivasankaran Rajamanickam, and Kamesh Madduri, IPDPS 2014).

The time complexity of the .scc algorithm in the worst case is O(|V|+|E|*D), where |V| is the number of nodes in the graph, |E| is the number of edges in the graph, and D is the diameter, defined as the length of the longest path from one node to another in the graph.

```
The algorith's space complexity is O(|V|) + [storage of data O(|V|+|E|)]. This can be expressed in more detail as: |V| * (7 * sizeof(local id type) + 4 * sizeof(global id type)) + O(|1DData Parts|).
```

.scc syntax

```
CALL neptune.algo.scc(
  [source-node list (required)],
  {
    edgeLabels: [list of edge labels for filtering (optional)],
    vertexLabel: a node label for filtering (optional),
    concurrency: number of threads to use (optional)
    }
)
YIELD node, component
RETURN node, component
```

.scc inputs

- a source node list (required) type: Node[] or NodeId[]; default: none.
- a configuration object that contains:
 - edgeLabels (optional) type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• vertexLabel (optional) – type: string; default: none.

A node label to filter on.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.scc outputs

For each source node:

- node The source node.
- **component** The component ID associated with the source node.

If the input node list is empty, the output is empty.

.scc query examples

This openCypher query has an empty input list, and so will have no output:

```
Match (n)
CALL neptune.algo.scc(n, {edgeLabels: ["route", "contains"]})
YIELD component
RETURN n, component
```

This is a query integration example, where .scc follows a MATCH clause that generates its input node list:

```
Match (n)
CALL neptune.algo.scc(n, {})
Yield component
Return n, component
```

🔥 Warning

It is not good practice to use MATCH(n) without restriction in query integrations. Keep in mind that every node returned by the MATCH(n) clause invokes the algorithm once, which can result a very long-running query if a large number of nodes is returned. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .scc output

Here is an example of the output returned by .wcc.mutate when run against the <u>sample air</u>routes dataset using the following query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string "CALL neptune.algo.wcc.mutate({writeProperty: 'wccid'}) YIELD success
 RETURN success" ∖
  --language open_cypher \
  /tmp/out.txt
cat /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "10",
        "~entityType": "node",
        "~labels": ["airport"],
        "~properties": {
          "lat": 38.94449997,
          "elev": 313,
          "type": "airport",
          "code": "IAD",
          "lon": -77.45580292,
          "runways": 4,
          "longest": 11500,
          "communityId": 2357352929951971,
          "city": "Washington D.C.",
          "region": "US-VA",
          "desc": "Washington Dulles International Airport",
          "prscore": 0.002264724113047123,
          "degree": 312,
```

```
"wccid": 2357352929951779,
        "ccscore": 0.20877772569656373,
        "country": "US",
        "icao": "KIAD"
      }
    },
    "component": 2357352929966149
  },
  {
    "node": {
      "~id": "12",
      "~entityType": "node",
      "~labels": ["airport"],
      "~properties": {
        "lat": 40.63980103,
        "elev": 12,
        "type": "airport",
        "code": "JFK",
        "lon": -73.77890015,
        "runways": 4,
        "longest": 14511,
        "communityId": 2357352929951971,
        "city": "New York",
        "region": "US-NY",
        "desc": "New York John F. Kennedy International Airport",
        "prscore": 0.002885053399950266,
        "degree": 403,
        "wccid": 2357352929951779,
        "ccscore": 0.2199712097644806,
        "country": "US",
        "icao": "KJFK"
      }
    },
    "component": 2357352929966149
  }
]
```

}

Strongly connected components mutate algorithm

<u>Strongly connected components (SCC)</u> are the maximally connected subgraphs of a directed graph, where every node is reachable from every other node (in other words, there exists a path between every node in the subgraph).

The time complexity of the .scc-mutate algorithm in the worst case is O(|V|+|E|*D), where |V| is the number of nodes in the graph, |E| is the number of edges in the graph, and D is the diameter, the length of the longest path from one node to another in the graph.

The algorith's space complexity is O(|V|) + [storage of data O(|V|+|E|)]. This can be expressed in more detail as: |V| * (7 * sizeof(local id type) + 4 * sizeof(global id type)) + O(|1DData Parts|).

.scc.mutate syntax

```
CALL neptune.algo.scc.mutate(
    {
        writeProperty: the name for the node property to which to write component IDs
        edgeLabels: [list of edge labels for filtering (optional)],
        vertexLabel: a node label for filtering (optional),
        concurrency: number of threads to use (optional)
     }
     )
YIELD success
RETURN success
```

Inputs for the .scc.mutate algorithm

Inputs for .scc.mutate are passed in a configuration object that contains:

• writeProperty (required) – type: string; default: none.

A name for the new node property where the component IDs will be written.

 edgeLabels (optional) – type: a list of edge label strings; example: ["route", ...]; default: no edge filtering.

To filter on one more edge labels, provide a list of the ones to filter on. If no edgeLabels field is provided then all edge labels are processed during traversal.

• **vertexLabel** (optional) – type: string; default: none.

The node label to filter on for traversing. Only nodes matching this label will be traversed. For example: "airport".

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

Outputs for the .scc.mutate algorithm

The computed strongly connected component IDs are written as a new node property using the specified property name. A single success flag (true or false) is returned to indicate whether the computation and writes succeeded or failed.

.scc.mutate query example

```
CALL neptune.algo.scc.mutate(
    {
        writeProperty: "SCOMM_ID",
        edgeLabels: ["route", ..],
        vertexLabel: "airport",
        concurrency: 2
    }
)
```

Sample .scc.mutate output

Here is an example of the output returned by .scc.mutate when run against the <u>sample air</u>routes dataset using the following query:

```
aws neptune-graph execute-query \
    --graph-identifier ${graphIdentifier} \
    --query-string "CALL neptune.algo.scc.mutate({writeProperty: 'sccid'}) YIELD success
RETURN success" \
    --language open_cypher \
    /tmp/out.txt
```

```
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```

```
cat /tmp/out.txt
{
    "results": [
      {
        "success": true
      }
    ]
}
```

Working with vector similarity in Neptune Analytics

You can answer complex questions about your data by transforming data shapes into embeddings (that is, vectors). Using a vector search index lets you answer questions about the your data's context and its similarity and connection to other data.

For example, a support agent could translate a question that they receive into a vector and use it to search the support knowledge base for articles that are similar to the words in the question (implicit similarity). For the most applicable articles, they could then collect metadata about the author, previous cases, runbooks, and so on so as to provide additional context when answering the question (explicit data).

Vector similarity search in Neptune Analytics makes it easy for you to build machine learning (ML) augmented search experiences and generative artificial intelligence (GenAI) applications. It also gives you an overall lower total cost of ownership and simpler management overhead because you no longer need to manage separate data stores, build pipelines, or worry about keep the data stores in sync. You can use vector similarity search in Neptune Analytics to augment your LLMs by integrating graph queries for domain-specific context with the results from low-latency, nearest-neighbor similarity search on embeddings imported from LLMs hosted in Amazon Bedrock, Graph Neural Networks (GNNs) in GraphStorm, or other sources.

As an example, Bioinformatics researchers who are interested in re-purposing existing blood pressure drugs for other treatable diseases, want to use vector similarity search over in-house knowledge graphs to find patterns in protein interaction networks.

For another example, a large online book retailer may need to use known pirated material to quickly identify similar media in conjunction with a knowledge graph to identify patterns of deceptive listing behaviours and find malicious sellers.

In both cases, vector search over a knowledge graph increases accuracy and speed when building the solution. It reduces the operational overhead and complexity using the tools available today.

You can create a vector index for your graph to try out this feature. Neptune Analytics supports associating embeddings generated from LLMs with the nodes of your graphs.

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Vector indexing in Neptune Analytics

You can only create a vector search index for a Neptune Analytics graph at the time the graph is created. Neptune Analytics lets you create only one vector index for a graph, with a fixed dimension between 1 and 65,535 inclusive.

When you create a Neptune Analytics graph in the console, you specify the index dimension under **Vector search settings** near the end of the process.

Loading vectors into a Neptune Analytics graph vector index

Note that the nodes in your graph must have at least one user property or label in order to associate them with embeddings. Also, Neptune Analytics does not support the special positive and negative infinity (INF, -INF) and not-a-number (NaN) floating-point values.

Neptune Analytics supports optional embeddings in the CSV file when the vector index is enabled. This means that not every node needs to be associated with an embedding.

Neptune Analytics does not currently support loading vectors from Neptune Database or a snapshot.

There are two ways you can load vectors associated with nodes in your graph:

Load the vectors from graph data files Amazon S3

When you're loading graph data from files in Amazon S3 using the console or the <u>neptune.load</u> openCypher integration, you can add a column to your CSV data with an embedding:vector header. This column should contain a list of integer or floating-point values separated by semicolons (;) that forms a vector of the required dimension and is the embedding for the node in question.

For example, associating a 4-dimensional vector with nodes in your graph in the openCypher CSV format would look like this:

```
:ID, name:String, embedding:Vector, :LABEL
v1,"ABC",0.1;0.5;0.8;-1.32,person
v2,"DEF",8.1;-0.2;0.432;-1.02,person
v3,"GHI",12323343;24324;2433554;-4343434,person
v4,"JKL",121.12213;3223.212;265;-1.32,person
```

In the Gremlin CSV format, the same thing would look like this:

```
~id, name, embedding:vector, ~label
v1,"ABC",0.1;0.5;0.8;-1.32,person
v2,"DEF",8.1;-0.2;0.432;-1.02,person
v3,"GHI",12323343;24324;2433554;-4343434,person
v4,"JKL",121.12213;3223.212;265;-1.32,person
```

Using the vectors.upsert algorithm to load vectors for your graph

You can also use the <u>vectors.upsert</u> algorithm to insert or update embeddings in a Neptune Analytics graph that has a vector search index. For example, in openCypher you can call the algorithm like this:

```
CALL neptune.algo.vectors.upsert(
   "person933",
   [0.1, 0.2, 0.3, ..]
)
YIELD node, embedding, success
RETURN node, embedding, success
```

Another example is:

```
UNWIND [
    {id: "933", embedding: [1,2,3,4]},
    {id: "934", embedding: [-1,-2,-3,-4]}
] as entry
MATCH (n:person) WHERE id(n)=entry.id WITH n, entry.embedding as embedding
CALL neptune.algo.vectors.upsert(n, embedding)
YIELD success
RETURN n, embedding, success
```

Common errors you may encounter when loading embeddings

• If the embeddings you are trying to load have a different dimension than is expected by the vector index, the load fails with parsing exception and a message like the following:

```
An error occurred (ParsingException) when calling the
ExecuteOpenCypherQuery operation: Could not load vector embedding: (the
embedding in question). Please check the dimensionality for this vector
```

```
when parsing line [(line number)] in [(file
name)]
```

 If the embeddings in a file are not properly formatted, Neptune Analytics reports a Parsing Exception before starting the load. For example, if the column header for the embedding column is not embedding:vector, Neptune Analytics would report an error like this:

• If embeddings are present in a file to be loaded but no vector index is present, Neptune Analytics simply ignores the embeddings and loads the graph data without them.

Vector-search algorithms in Neptune Analytics

Neptune Analytics supports a variety of vector-search algorithms that are listed in the <u>VSS</u> <u>algorithms</u> section.

Vector-similarity search (VSS) algorithms in Neptune Analytics

Vector simlarity search algorithms identify similar vectors based on the vector distance between them.

Neptune Analytics supports the following vector-similarity search algorithms:

1 Note

The following special floating-point values are not supported in Neptune Analytics vectorsimilarity search algorithms:

- INF (infinity)
- -INF (negative infinity)
- NaN (not-a-number)

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The .vectors.distance algorithm

The .vectors.distance algorithm computes the distance between two vectors based on their embeddings. The distance is the L2 norm of the vectors.

.vectors.distance syntax

```
MATCH( n {`~id`: "the ID of the source node(s)"} )
MATCH( m {`~id`: "the ID of the target node(s)" })
CALL neptune.algo.vectors.distance(n, m)
YIELD distance
RETURN n, m, distance
```

.vectors.distance inputs

• a source node list (required) - type: Node[] or NodeId[]; default: none.

The result of a MATCH statement from which you want to source distance computations.

• a target node (required) - type: Node[] or NodeId[]; default: none.

The result of a MATCH statement from which you want to source distance computations.

<u> M</u>arning

Be careful to limit MATCH(n) and MATCH(m) so that they don't return a large number of nodes. Keep in mind that every pair of n and m in the join result invokes .vectors.distance once. Too many inputs can therefore result in very long runtimes. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

.vectors.distance outputs

For every pair of source node and target node:

- source The starting point for the distance call.
- target The target for the distance computation.
- distance The distance between source and target.

.vectors.distance query example

```
MATCH ( n {`~id`: "106"} )
MATCH ( m {`~id`: "110" } )
CALL neptune.algo.vectors.distance( n, m )
YIELD distance
RETURN n, m, distance
```

Sample .vectors.distance output

Here is an example of the output returned by .vectors.distance when run against a sample Wikipedia dataset using the following query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH (n{`~id`: '0'})
                       MATCH (m{`~id`: '1'})
                       CALL neptune.algo.vectors.distance(n, m)
                       YIELD distance
                       RETURN n, m, distance" \
  --language open_cypher \setminus
  /tmp/out.txt
{
  "results": [
    {
      "n": {
        "~id": "0",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 0,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "The 24-hour clock is a way of telling the time in which the day
 runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to
 23. It does not use a.m. or p.m. This system is also referred to (only in the US and
 the English speaking parts of Canada) as military time or (only in the United Kingdom
```

and now very rarely) as continental time. In some parts of the world\\, it is called

```
railway time. Also\\, the international standard notation of time (ISO 8601) is based
on this format."
       }
     },
     "m": {
       "~id": "1",
       "~entityType": "node",
       "~labels": [],
       "~properties": {
         "title": "24-hour clock",
         "views": 2450.62548828125,
         "wiki_id": 9985,
         "paragraph_id": 1,
         "url": "https://simple.wikipedia.org/wiki?curid=9985",
         "langs": 30,
         "text": "A time in the 24-hour clock is written in the form hours:minutes
(for example\\, 01:23)\\, or hours:minutes:seconds (01:23:45). Numbers under 10 have
a zero in front (called a leading zero); e.g. 09:07. Under the 24-hour clock system\\,
the day begins at midnight\\, 00:00\\, and the last minute of the day begins at 23:59
and ends at 24:00\\, which is identical to 00:00 of the following day. 12:00 can only
be mid-day. Midnight is called 24:00 and is used to mean the end of the day and 00:00
is used to mean the beginning of the day. For example \lambda, you would say \forall Tuesday at
```

```
24:00\" and \"Wednesday at 00:00\" to mean exactly the same time."

},

"distance": 27.762847900390626

}

]
```

The .vectors.get algorithm

The .vectors.get algorithm retrieves the embedding for a node.

.vectors.get syntax

```
MATCH( n {`~id`: "the ID of the node"} )
CALL neptune.algo.vectors.get(n)
YIELD embedding
RETURN n, embedding
```

.vectors.get input

• a source node or nodes (required) - type: Node[] or NodeId[]; default: none.

The result of a MATCH statement that produces the node(s) for which you want to retrieve the embedding.

🔥 Warning

Be careful to limit MATCH(n) so that it doesn't return a large number of nodes. Keep in mind that every source node in the n result invokes .vectors.get once. Too many inputs can therefore result in very long runtimes. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

.vectors.get outputs

For each source node provided:

- **node** The source node.
- embedding The embedding of that source node.

.vectors.get query example

```
MATCH ( n {`~id`: "0"} )
CALL neptune.algo.vectors.get(n)
YIELD embedding
```

RETURN n, embedding

Sample .vectors.get output

Here is an example of the output returned by .vectors.get when run against the <u>sample</u> Wikipedia dataset using the following query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH ( n {`~id`: '0'} )
                       CALL neptune.algo.vectors.get(n)
                       YIELD embedding
                        RETURN n, embedding" ∖
  --language open_cypher \
  /tmp/out.txt
{
  "results": [
    {
      "n": {
        "~id": "0",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 0,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "The 24-hour clock is a way of telling the time in which the day
 runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to
 23. It does not use a.m. or p.m. This system is also referred to (only in the US and
 the English speaking parts of Canada) as military time or (only in the United Kingdom
 and now very rarely) as continental time. In some parts of the world\backslash, it is called
 railway time. Also\\, the international standard notation of time (ISO 8601) is based
 on this format."
        }
      },
      "embedding": [
        0.07711287587881088,
        0.3197174072265625,
       -0.2051590085029602,
        0.6302579045295715,
```

0.032093219459056857, 0.200703963637352, 0.16665680706501008, -0.31295087933540347, 0.17575109004974366, 0.5308129191398621, -0.37528499960899355, 0.3338659405708313, -0.046272162348032, 0.07841536402702332, -0.3490406274795532, 0.27182886004447939, 0.3073517680168152, -0.08306130766868592, 0.5035958886146545, 0.254621684551239, -0.40407684445381167, 0.28878292441368105, -0.22588828206062318, -0.13185778260231019, -0.21559733152389527, 0.4900434613227844, 0.03866531699895859, 0.507415771484375, -0.3067346513271332, 0.10740984976291657, 0.08998646587133408, -0.2652775049209595, -0.28492602705955508, 0.33600345253944399, -0.27227747440338137, 0.3691731095314026, -0.2815995514392853, 0.0856710895895958, -0.13187488913536073, 0.4753035008907318, -0.2241700142621994, 0.20263174176216126, 0.4390721619129181, 0.06424559652805329, 0.2463042289018631, -0.39631763100624087, 0.2971232533454895, 0.2415716052055359,

-0.02803819440305233, 0.32105034589767458, -0.02222033031284809, -0.008510420098900795, -0.00032598740654066205, 0.031057516112923623, -0.5332233309745789, 0.45022767782211306, -0.6829474568367004, 1.3313145637512208, 0.19445496797561646, -0.15697629749774934, -0.09996363520622254, -0.2786232829093933, -0.09833164513111115, -0.17644722759723664, 0.11717787384986878, 0.2820119559764862, 0.029635537415742875, 0.5247654914855957, 0.5323811173439026, -0.06254086643457413, -0.05274389684200287, 0.3877565860748291, 0.43260684609413149, 0.5207982063293457, -0.27160540223121645, -0.06000519543886185, -0.032806672155857089, -0.3594319522380829, 0.4218965470790863, -0.3766363263130188, 0.44727250933647158, -0.04586323723196983, 0.06902860850095749, 0.3030509352684021, 0.18945887684822083, 0.21681705117225648, -0.014492596499621868, -0.38649576902389529, -0.1129651814699173, 0.050081491470336917, -0.01697717048227787, 0.1415158063173294,

-0.3284287750720978, -0.02309800498187542, -0.2051207274198532, -0.017861712723970414, -0.07372242212295532, -0.12263767421245575, 0.21828559041023255, -0.36898064613342287, 0.3558262288570404, -0.16924124956130982, -0.31757786870002749, 0.5452765226364136, 0.24666202068328858, -0.08289600908756256, -0.14674079418182374, -0.18049933016300202, 0.3646247982978821, 0.42489132285118105, 0.0909421369433403, -0.1764664500951767, 0.22471413016319276, 0.049531541764736179, -0.022898104041814805, 0.08607156574726105, 0.14532636106014253, -0.205774188041687, -0.3457978069782257, -1.2771626710891724, 0.2826114892959595, 0.2066900134086609, -0.3884444832801819, -0.3564482629299164, -0.25118574500083926, -0.728326141834259, 0.5217206478118897, -0.43305152654647829, 0.3510914444923401, 0.5106240510940552, -0.11594267934560776, 0.43993058800697329, 0.25412991642951968, 0.4275965392589569, 0.1463870108127594,

0.3510439395904541,

0.1619710624217987, 0.11160195618867874, -0.22760489583015443, -0.23652249574661256, 0.05374380201101303, 0.7251803278923035, -0.13991153240203858, 0.9363659024238586, -0.05858418717980385, 0.5233941674232483, 0.12388131022453308, 0.6248424649238586, -0.11751417070627213, 0.09689709544181824, 0.7467237710952759, 0.2247271090745926, -0.6747357845306397, -0.16039365530014039, -0.41555172204971316, -0.04566565155982971, 0.21260707080364228, 0.2549103796482086, 0.24795542657375337, 0.5625612735748291, 0.8036459684371948, 0.15800043940544129, 0.04797195643186569, -0.15839435160160066, -0.06506697088479996, -0.2577322721481323, 0.3262946903705597, 0.5458049178123474, 0.616370439529419, -0.35092639923095705, 0.048758912831544879, 0.11522434651851654, 0.04175107553601265, -0.12269306182861328, 0.1227836161851883, 0.4020257890224457, 0.07093577086925507, -0.1880340874195099, 0.5334663391113281, 0.46888044476509097,

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0.15788795053958894, -0.3212168216705322, -0.20272433757781983, -0.8973743319511414, 0.060059018433094028, -0.014103145338594914, -0.3387225568294525, -0.49839726090431216, -0.011007139459252358, -0.16101065278053285, -0.20850643515586854, 0.4891682267189026, 0.33551496267318728, -0.23595896363258363, -0.4257577359676361, -0.48884832859039309, 0.48760101199150088, 0.34031161665916445, 0.1722799688577652, -0.35575979948043826, 0.629051923751831, -0.8014369010925293, 0.575096607208252, 0.421142578125, -0.2668846547603607, -0.046029768884181979, 0.2791147530078888, -0.22112232446670533, 0.02008579671382904, 0.22087614238262177, -0.17961964011192323, 0.4235396981239319, 0.295818567276001, -0.18260923027992249, 0.3227207660675049, 0.11412205547094345, 0.04591478034853935, 0.5127033591270447, 0.428005576133728, 0.20718106627464295, 0.18405631184577943, -0.22416146099567414, 0.4277373254299164, 0.5384698510169983,

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.vectors.get

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.vectors.topKByEmbedding algorithm

The .vectors.topKByEmbedding algorithm finds the topK nearest neighbors of an embedding based on the distance of their vector embeddings.

.vectors.topKByEmbedding syntax

```
CALL neptune.algo.vectors.topKByEmbedding(
  [an embedding (required)],
  {
   topK: the number of result nodes to return (optional, default: 10),
   concurrency: the number of cores to use to run the algorithm (optional)
  }
)
YIELD embedding, node, score
RETURN embedding, node, score
```

.vectors.topKByEmbedding input

• an embedding (required) type: a list of floating-point values.

The source input embedding to use to compute the distance to the embeddings of the candidate target nodes. The dimension of the embedding must match the declared dimension of the associated vector index.

The embedding may or may not exist in the database. If not, it can be any vector of the same dimension as is declared in the associated vector index.

• **topK** (optional) type: a positive integer; default: 10.

The number of result nodes to return.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.vectors.topKByEmbedding outputs

For each node returned:

- embedding The input embedding.
- node A node whose embedding is at one of the topK nearest distances from the input embedding.
- score The distance between the input embedding and the embedding of this node.

.vectors.topKByEmbedding query example

You can provide the embedding explicitly in the query, although embeddings tend to be very large:

```
CALL neptune.algo.vectors.topKByEmbedding(
  [0.1, 0.2, 0.3, ...],
  {
    topK: 7,
    concurrency: 1
   }
)
YIELD embedding, node, score
RETURN embedding, node, score
```

Most often, you will by generating embeddings to pass to the algorithm. For example:

```
MATCH ( n:airport {code: 'ANC'} )
CALL neptune.algo.vectors.get(n) YIELD embedding AS vector WITH vector
CALL neptune.algo.vectors.topKByEmbedding(
   vector,
   {
    topK: 10,
    concurrency: 1
   }
)
YIELD node, score
RETURN vector, node, score
```

🔥 Warning

In queries like the one above, be careful to limit MATCH(n) so that it doesn't return a large number of nodes. Keep in mind that every node in n invokes a separate run of both .vectors.get and .vectors.topKByEmbedding. Too many inputs can therefore result in very long runtimes. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .vectors.topKByEmbedding output

Here is an example of the output returned by .vectors.topKByEmbedding when run against the sample Wikipedia dataset using the following query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH ( n {`~id`: '0'} )
                        CALL neptune.algo.vectors.get(n) YIELD embedding AS vector
                        CALL neptune.algo.vectors.topKByEmbedding( vector, { topK: 3 })
                       YIELD node, score
                        RETURN node, score"
  --language open_cypher \setminus
  /tmp/out.txt
{
  "results": [
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        "~id": "0",
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          "langs": 30,
          "text": "The 24-hour clock is a way of telling the time in which the day
 runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to
 23. It does not use a.m. or p.m. This system is also referred to (only in the US and
 the English speaking parts of Canada) as military time or (only in the United Kingdom
 and now very rarely) as continental time. In some parts of the world\\, it is called
```

```
railway time. Also\\, the international standard notation of time (ISO 8601) is based
 on this format."
        }
      },
      "score": 0.0
    },
    {
      "node": {
        "~id": "2",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 2,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "However\\, the US military prefers not to say 24:00 - they do not
 like to have two names for the same thing\\, so they always say \"23:59\"\\, which is
 one minute before midnight."
        }
      },
      "score": 24.000200271606447
    },
    {
      "node": {
        "~id": "3",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 3,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "24-hour clock time is used in computers\\, military\\, public safety
\backslash, and transport. In many Asian\backslash, European and Latin American countries people use it
to write the time. Many European people use it in speaking."
        }
      },
      "score": 25.013729095458986
    }
```

ι
ſ

]

.vectors.topKByNode algorithm

The .vectors.topKByNode algorithm finds the topK nearest neighbors of a node based on the distance of their vector embeddings from the node.

.vectors.topKByNode syntax

```
CALL neptune.algo.vectors.topKByNode(
   [a list of one or more nodes (required)],
   {
    topK: the number of result nodes to return (optional, default: 10),
    concurrency: the number of cores to use to run the algorithm (optional)
   }
   YIELD node, score
RETURN node, score
```

.vectors.topKByNode input

• a list of one or more source nodes (required) - type: Node[] or NodeId[].

If the source-node list is empty then the query result is also empty.

• **topK** (optional) type: a positive integer; default: 10.

The number of result nodes to return.

• **concurrency** (optional) – type: positive integer or 0; default: 0.

Controls the number of concurrent threads used to run the algorithm.

If concurrency is set to 0 or is not specified, all the available compute cores will be used to run algorithms concurrently. If the number is greater than the machine limit, it will be automatically capped to the limit.

.vectors.topKByNode outputs

For each source node:

- **source** The source node.
- node A node whose embedding is at one of the topK nearest distances from the source node's embedding.

 score – The distance between the source node's embedding and the embedding of the close node.

.vectors.topKByNode query example

```
MATCH ( n:airport {code: 'ANC'} )
CALL neptune.algo.vectors.topKByNode(
    n,
    {
      topK: 10,
      concurrency: 1
    }
)
YIELD node, score
RETURN n, node, score
```

🔥 Warning

In queries like the one above, be careful to limit MATCH(n) so that it doesn't return a large number of nodes. Keep in mind that every node in n invokes a separate run of .vectors.topKByNode. Too many inputs can therefore result in very long runtimes. Use LIMIT or put conditions on the MATCH clause to restrict its output appropriately.

Sample .vectors.topKByNode output

Here is an example of the output returned by .vectors.topKByNode when run against the sample Wikipedia dataset using the following query:

```
"n": {
    "~id": "0",
    "~entityType": "node",
    "~labels": [],
    "~properties": {
        "title": "24-hour clock",
        "views": 2450.62548828125,
        "wiki_id": 9985,
        "paragraph_id": 0,
        "url": "https://simple.wikipedia.org/wiki?curid=9985",
        "langs": 30,
```

"text": "The 24-hour clock is a way of telling the time in which the day runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to 23. It does not use a.m. or p.m. This system is also referred to (only in the US and the English speaking parts of Canada) as military time or (only in the United Kingdom and now very rarely) as continental time. In some parts of the world\\, it is called railway time. Also\\, the international standard notation of time (ISO 8601) is based on this format."

```
}
},
"node": {
    "~id": "0",
    "~entityType": "node",
    "~labels": [],
    "~properties": {
        "title": "24-hour clock",
        "views": 2450.62548828125,
        "wiki_id": 9985,
        "paragraph_id": 0,
        "url": "https://simple.wikipedia.org/wiki?curid=9985",
        "langs": 30,
```

"text": "The 24-hour clock is a way of telling the time in which the day runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to 23. It does not use a.m. or p.m. This system is also referred to (only in the US and the English speaking parts of Canada) as military time or (only in the United Kingdom and now very rarely) as continental time. In some parts of the world\\, it is called railway time. Also\\, the international standard notation of time (ISO 8601) is based on this format."

```
}
},
"score": 0.0
},
{
    "n": {
```

```
"~id": "0",
       "~entityType": "node",
       "~labels": [],
       "~properties": {
         "title": "24-hour clock",
         "views": 2450.62548828125,
         "wiki_id": 9985,
         "paragraph_id": 0,
         "url": "https://simple.wikipedia.org/wiki?curid=9985",
         "langs": 30,
         "text": "The 24-hour clock is a way of telling the time in which the day
runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to
23. It does not use a.m. or p.m. This system is also referred to (only in the US and
the English speaking parts of Canada) as military time or (only in the United Kingdom
and now very rarely) as continental time. In some parts of the world\\, it is called
railway time. Also\\, the international standard notation of time (ISO 8601) is based
on this format."
       }
     },
     "node": {
       "~id": "2",
       "~entityType": "node",
       "~labels": [],
       "~properties": {
         "title": "24-hour clock",
         "views": 2450.62548828125,
         "wiki_id": 9985,
         "paragraph_id": 2,
         "url": "https://simple.wikipedia.org/wiki?curid=9985",
         "langs": 30,
         "text": "However\\, the US military prefers not to say 24:00 - they do not
like to have two names for the same thing\\, so they always say \"23:59\"\\, which is
one minute before midnight."
       }
     },
     "score": 24.000200271606447
   },
   {
     "n": {
       "~id": "0",
       "~entityType": "node",
       "~labels": [],
       "~properties": {
         "title": "24-hour clock",
```

```
"views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 0,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "The 24-hour clock is a way of telling the time in which the day
 runs from midnight to midnight and is divided into 24 hours\\, numbered from 0 to
 23. It does not use a.m. or p.m. This system is also referred to (only in the US and
 the English speaking parts of Canada) as military time or (only in the United Kingdom
 and now very rarely) as continental time. In some parts of the world\backslash, it is called
 railway time. Also\\, the international standard notation of time (ISO 8601) is based
 on this format."
        }
      },
      "node": {
        "~id": "3",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 3,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "24-hour clock time is used in computers\\, military\\, public safety
\backslash, and transport. In many Asian\backslash, European and Latin American countries people use it
to write the time. Many European people use it in speaking."
        }
      },
      "score": 25.013729095458986
    }
  ]
}
```

.vectors.upsert algorithm

The .vectors.upsert algorithm is used to add a new embedding or update an existing one for a node.

.vectors.upsert syntax

```
CALL neptune.algo.vectors.upsert(
    "a target node (required)",
    [the embedding to upsert for the target node (required)]
)
YIELD node, embedding, success
RETURN node, embedding, success
```

.vectors.upsert input

• a target node (required) - type: Node or NodeId.

The node for which you want to upsert an embedding.

• an embedding (required) – type: a list of floating-point values.

The embedding that you want to upsert for the target node.

If the node has an existing embedding, this must match the dimension of the existing one or an exception is thrown.

.vectors.upsert outputs

If the target node already has an existing embedding then .vectors.upsert replaces it with the one supplied. Otherwise .vectors.upsert adds the supplied embedding for the target node.

- node The target node.
- embedding The embedding that was supplied to be upserted.
- success A Boolean value: true indicates that the upsert succeded, and false that it failed.

.vectors.upsert query examples

```
CALL neptune.algo.vectors.upsert(
    "person933",
```

```
[0.1, 0.2, 0.3, ..]
)
YIELD node, embedding, success
RETURN node, embedding, success
```

UNWIND [
 {id: "933", embedding: [1,2,3,4]},
 {id: "934", embedding: [-1,-2,-3,-4]}
] as entry
MATCH (n:person) WHERE id(n)=entry.id WITH n, entry.embedding as embedding
CALL neptune.algo.vectors.upsert(n, embedding)
YIELD success
RETURN n, embedding, success

Sample .vectors.upsert output

Here is an example of the output returned by .vectors.upsert when run against the <u>sample</u> Wikipedia dataset using the following query:

```
aws neptune-graph execute-query \
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH (n{\`~id\`:\"0\"})
                       CALL neptune.algo.vectors.get(n)
                       YIELD embedding AS vector
                       MATCH (m{`~id`: '1'})
                       CALL neptune.algo.vectors.upsert(m, vector)
                       YIELD node, embedding, success
                       RETURN node, embedding, success"
  --language open_cypher \
  /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "1",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 1,
```

```
"url": "https://simple.wikipedia.org/wiki?curid=9985",
         "langs": 30,
         "text": "A time in the 24-hour clock is written in the form hours:minutes
(for example\\, 01:23)\\, or hours:minutes:seconds (01:23:45). Numbers under 10 have
a zero in front (called a leading zero); e.g. 09:07. Under the 24-hour clock system\\,
the day begins at midnight \langle 00:00\rangle, and the last minute of the day begins at 23:59
and ends at 24:00\\, which is identical to 00:00 of the following day. 12:00 can only
be mid-day. Midnight is called 24:00 and is used to mean the end of the day and 00:00
is used to mean the beginning of the day. For example \lambda, you would say \"Tuesday at
24:00\" and \"Wednesday at 00:00\" to mean exactly the same time."
       }
     },
     "embedding": [
       0.07711287587881088,
       0.3197174072265625,
      -0.2051590085029602,
       0.6302579045295715,
       0.032093219459056857,
       0.200703963637352,
       0.16665680706501008,
      -0.31295087933540347,
       0.17575109004974366,
       0.5308129191398621,
      -0.37528499960899355,
       0.3338659405708313,
      -0.046272162348032,
       0.07841536402702332,
      -0.3490406274795532,
       0.27182886004447939,
       0.3073517680168152,
      -0.08306130766868592,
       0.5035958886146545,
       0.254621684551239,
      -0.40407684445381167,
       0.28878292441368105,
      -0.22588828206062318,
      -0.13185778260231019,
      -0.21559733152389527,
       0.4900434613227844,
       0.03866531699895859,
       0.507415771484375,
      -0.3067346513271332,
       0.10740984976291657,
       0.08998646587133408,
```

-0.2652775049209595, -0.28492602705955508, 0.33600345253944399, -0.27227747440338137, 0.3691731095314026, -0.2815995514392853, 0.0856710895895958, -0.13187488913536073, 0.4753035008907318, -0.2241700142621994, 0.20263174176216126, 0.4390721619129181, 0.06424559652805329, 0.2463042289018631, -0.39631763100624087, 0.2971232533454895, 0.2415716052055359, -0.02803819440305233, 0.32105034589767458, -0.02222033031284809, -0.008510420098900795, -0.00032598740654066205, 0.031057516112923623, -0.5332233309745789, 0.45022767782211306, -0.6829474568367004, 1.3313145637512208, 0.19445496797561646, -0.15697629749774934, -0.09996363520622254, -0.2786232829093933, -0.09833164513111115, -0.17644722759723664, 0.11717787384986878, 0.2820119559764862, 0.029635537415742875, 0.5247654914855957, 0.5323811173439026, -0.06254086643457413, -0.05274389684200287, 0.3877565860748291, 0.43260684609413149, 0.5207982063293457, -0.27160540223121645,

-0.06000519543886185, -0.032806672155857089, -0.3594319522380829, 0.4218965470790863, -0.3766363263130188, 0.44727250933647158, -0.04586323723196983, 0.06902860850095749, 0.3030509352684021, 0.18945887684822083, 0.21681705117225648, -0.014492596499621868, -0.38649576902389529, -0.1129651814699173, 0.050081491470336917, -0.01697717048227787, 0.1415158063173294, -0.3284287750720978, -0.02309800498187542, -0.2051207274198532, -0.017861712723970414, -0.07372242212295532, -0.12263767421245575, 0.21828559041023255, -0.36898064613342287, 0.3558262288570404, -0.16924124956130982, -0.31757786870002749, 0.5452765226364136, 0.24666202068328858, -0.08289600908756256, -0.14674079418182374, -0.18049933016300202, 0.3646247982978821, 0.42489132285118105, 0.0909421369433403, -0.1764664500951767, 0.22471413016319276, 0.049531541764736179, -0.022898104041814805, 0.08607156574726105, 0.14532636106014253, -0.205774188041687,

-1.2771626710891724, 0.2826114892959595, 0.2066900134086609, -0.3884444832801819, -0.3564482629299164, -0.25118574500083926, -0.728326141834259, 0.5217206478118897, -0.43305152654647829, 0.3510914444923401, 0.5106240510940552, -0.11594267934560776, 0.43993058800697329, 0.25412991642951968, 0.4275965392589569, 0.1463870108127594, 0.3510439395904541, 0.1619710624217987, 0.11160195618867874, -0.22760489583015443, -0.23652249574661256, 0.05374380201101303, 0.7251803278923035, -0.13991153240203858, 0.9363659024238586, -0.05858418717980385, 0.5233941674232483, 0.12388131022453308, 0.6248424649238586, -0.11751417070627213, 0.09689709544181824, 0.7467237710952759, 0.2247271090745926, -0.6747357845306397, -0.16039365530014039, -0.41555172204971316, -0.04566565155982971, 0.21260707080364228, 0.2549103796482086, 0.24795542657375337, 0.5625612735748291, 0.8036459684371948, 0.15800043940544129, 0.04797195643186569, -0.15839435160160066, -0.06506697088479996, -0.2577322721481323, 0.3262946903705597, 0.5458049178123474, 0.616370439529419, -0.35092639923095705, 0.048758912831544879, 0.11522434651851654, 0.04175107553601265, -0.12269306182861328, 0.1227836161851883, 0.4020257890224457, 0.07093577086925507, -0.1880340874195099, 0.5334663391113281, 0.46888044476509097, 0.18104688823223115, 0.30756646394729617, 0.29316428303718569, -0.10604366660118103, 0.44999250769615176, 0.18227706849575044, 0.5962150692939758, 0.38278165459632876, -0.40461188554763796, 0.17775404453277589, -0.16349074244499207, 0.06950787454843521, 0.7547341585159302, -0.4842711389064789, 0.4062837064266205, 0.09000574052333832, 0.03859427571296692, 0.24143263697624207, -0.3383118510246277, 0.3363209366798401, 0.10778547078371048, 0.3429640233516693, -0.20395530760288239, 0.011477324180305004, 0.6145590543746948, -0.5488739609718323,

-0.09723474085330963, -0.19020821154117585, -0.18068274855613709, 0.1601778119802475, 0.038950759917497638, 0.6372026205062866, -0.12897184491157533, 0.10720998793840409, 0.13482464849948884, -0.07540713250637055, -0.0881727784872055, 0.5626690983772278, -0.31975486874580386, -0.029084375128149987, 0.43618619441986086, 0.32975345849990847, -0.4053913652896881,0.15788795053958894, -0.3212168216705322, -0.20272433757781983, -0.8973743319511414, 0.060059018433094028, -0.014103145338594914, -0.3387225568294525, -0.49839726090431216, -0.011007139459252358, -0.16101065278053285, -0.20850643515586854, 0.4891682267189026, 0.33551496267318728, -0.23595896363258363, -0.4257577359676361, -0.48884832859039309, 0.48760101199150088, 0.34031161665916445, 0.1722799688577652, -0.35575979948043826, 0.629051923751831, -0.8014369010925293, 0.575096607208252, 0.421142578125, -0.2668846547603607, -0.046029768884181979,

0.2791147530078888,

-0.22112232446670533, 0.02008579671382904, 0.22087614238262177, -0.17961964011192323, 0.4235396981239319, 0.295818567276001, -0.18260923027992249, 0.3227207660675049, 0.11412205547094345, 0.04591478034853935, 0.5127033591270447, 0.428005576133728, 0.20718106627464295, 0.18405631184577943, -0.22416146099567414, 0.4277373254299164, 0.5384698510169983, 0.04109276458621025, 0.5105301141738892, 0.473961740732193, -0.6853302717208862, -0.16557902097702027, -0.12704522907733918, 0.0026600745040923359, 0.5272349715232849, 0.12121742218732834, 0.427141010761261, -0.3047095239162445, 0.5948843359947205, 0.335798442363739, 0.35749775171279909, -0.18497343361377717, 0.26501506567001345, 0.1564970314502716, 0.4210122525691986, -0.1915784478187561, 0.057152874767780307, -0.28498271107673647, 0.04969947412610054, 0.7697478532791138, 0.5546697974205017, 0.0958070456981659, -0.3533228933811188, 0.4784282147884369,

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] }

.vectors.remove algorithm

The .vectors.remove algorithm is used to remove the embedding from a node.

.vectors.remove syntax

```
CALL neptune.algo.vectors.remove(
   [a list of one or more nodes]
)
YIELD node, success
RETURN node, success
```

.vectors.remove input

• a target node list (required) - type: Node[] or NodeId[].

The node(s) from which you want to remove the embedding. If an empty list is supplied, the result will be empty.

.vectors.remove outputs

The following outputs are returned for each target node, and if the node has an embedding, the embedding is removed:

- node The target node.
- success A Boolean value: true indicates that the removal succeded for the node, and false indicates that it failed.

.vectors.remove query examples

```
CALL neptune.algo.vectors.remove( ["person933"] )
YIELD node, success
RETURN node, success
```

```
MATCH (n: Student)
CALL neptune.algo.vectors.remove(n)
YIELD status
RETURN n, success
```

Sample .vectors.remove output

Here is an example of the output returned by .vectors.remove when run against the <u>sample</u> Wikipedia dataset using the following query:

```
aws neptune-graph execute-query \setminus
  --graph-identifier ${graphIdentifier} \
  --query-string "MATCH (n {`~id`: '1'})
                       CALL neptune.algo.vectors.remove(n)
                       YIELD node, success
                       RETURN node, success" ∖
  --language open_cypher \
  /tmp/out.txt
{
  "results": [
    {
      "node": {
        "~id": "1",
        "~entityType": "node",
        "~labels": [],
        "~properties": {
          "title": "24-hour clock",
          "views": 2450.62548828125,
          "wiki_id": 9985,
          "paragraph_id": 1,
          "url": "https://simple.wikipedia.org/wiki?curid=9985",
          "langs": 30,
          "text": "A time in the 24-hour clock is written in the form hours:minutes
 (for example\\, 01:23)\\, or hours:minutes:seconds (01:23:45). Numbers under 10 have
 a zero in front (called a leading zero); e.g. 09:07. Under the 24-hour clock system\\,
 the day begins at midnight\\, 00:00\\, and the last minute of the day begins at 23:59
 and ends at 24:00\\, which is identical to 00:00 of the following day. 12:00 can only
 be mid-day. Midnight is called 24:00 and is used to mean the end of the day and 00:00
 is used to mean the beginning of the day. For example \lambda, you would say \"Tuesday at
 24:00\" and \"Wednesday at 00:00\" to mean exactly the same time."
        }
      },
      "success": true
```

] } }

Best practices

The following are some general recommendations for working with Neptune Analytics. Use this information as a reference to quickly find recommendations for maximizing performance while using Neptune Analytics.

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- openCypher query best practices
 - Use the SET clause to remove multiple properties at once
 - Use parameterized queries
 - Use flattened maps instead of nested maps in UNWIND clause
 - Place more restrictive nodes on the left side in Variable-Length Path (VLP) expressions
 - Avoid redundant node label checks by using granular relationship names
 - Specify edge labels where possible
 - Avoid using the WITH clause when possible
 - Place restrictive filters as early in the query as possible
 - Explicitly check whether properties exist
 - Do not use named path (unless it is required)
 - Avoid COLLECT(DISTINCT())
 - Prefer the properties function over individual property lookup when retrieving all property values
 - Perform static computations outside of the query
 - Batch inputs using UNWIND instead of individual statements
 - Prefer using custom IDs for node
 - Avoid doing ~id computations in the query

openCypher query best practices

Use the SET clause to remove multiple properties at once

When using the openCypher language, REMOVE is used to remove properties from an entity. In Neptyme Analytics, teach property being removed requires a separate operation, adding query 3 latency. You can instead use SET with a map to set all property values to null, which in Neptune Analytics is equivalent to removing properties. Neptune Analytics will have increased performance when multiple properties on a single entity are required to be removed.

Use:

```
WITH {prop1: null, prop2: null, prop3: null} as propertiesToRemove
MATCH (n)
SET n += propertiesToRemove
```

Instead of:

```
MATCH (n)
REMOVE n.prop1, n.prop2, n.prop3
```

Use parameterized queries

It is recommended to always use parameterized queries when querying using openCypher. The query engine can leverage repeated parameterized queries for features like query plan cache, where repeated invocation of the same parameterized structure with different parameters can leverage the cached plans. The query plan generated for parameterized queries is cached and reused only when it completes within 100ms and the parameter types are either NUMBER, BOOLEAN or STRING.

Use:

```
MATCH (n:foo) WHERE id(n) = $id RETURN n
```

With parameters:

```
parameters={"id": "first"}
parameters={"id": "second"}
parameters={"id": "third"}
```

Instead of:

```
MATCH (n:foo) WHERE id(n) = "first" RETURN n
MATCH (n:foo) WHERE id(n) = "second" RETURN n
MATCH (n:foo) WHERE id(n) = "third" RETURN n
```

You can determine if the query is using a cached plan by observing the plan cache hits: value in the output of the openCypher explain endpoint.

Use flattened maps instead of nested maps in UNWIND clause

Deep nested structure can restrict the ability of the query engine to generate an optimal query plan. To partially alleviate this issue, the following defined patterns will create optimal plans for the following scenarios:

- Scenario 1: UNWIND with a list of cypher literals, which includes NUMBER, STRING and BOOLEAN.
- Scenario 2: UNWIND with a list of flattened maps, which includes only cypher literals (NUMBER, STRING, BOOLEAN) as values.

When writing a query containing UNWIND clause, use the above recommendation to improve performance.

Scenario 1 example:

```
UNWIND $ids as x
MATCH(t:ticket {`~id`: x})
```

With parameters:

```
parameters={
    "ids": [1, 2, 3]
}
```

An example for Scenario 2 is to generate a list of nodes to CREATE or MERGE. Instead of issuing multiple statements, use the following pattern to define the properties as a set of flattened maps:

```
UNWIND $props as p
CREATE(t:ticket {title: p.title, severity:p.severity})
```

With parameters:

parameters={
 "props": [

```
{"title": "food poisoning", "severity": "2"},
   {"title": "Simone is in office", "severity": "3"}
]
}
```

Instead of nested node objects like:

```
UNWIND $nodes as n
CREATE(t:ticket n.properties)
```

With parameters:

```
parameters={
    "nodes": [
        {"id": "ticket1", "properties": {"title": "food poisoning", "severity": "2"}},
        {"id": "ticket2", "properties": {"title": "Simone is in office", "severity": "3"}}
]
```

Place more restrictive nodes on the left side in Variable-Length Path (VLP) expressions

In Variable-Length Path (VLP) queries, the query engine optimizes the evaluation by choosing to start the traversal on the left or right side of the expression. The decision is based on the cardinality of the patterns on the left and right side. Cardinality is the number of nodes matching the specified pattern.

- If the right pattern has a cardinality of one, then the right side will be the starting point.
- If the left and the right side have cardinality of one, the expansion is checked on both sides and starts on the side with the smaller expansion. Expansion is the number of outgoing or incoming edges for the node on the left and the node on the right side of the VLP expression. This part of the optimization is only used if the VLP relationship is unidirectional and the relationship type is provided.
- Otherwise, the left side will be the starting point.

For a chain of VLP expressions, this optimization can only be applied to the first expression. The other VLPs are evaluated starting with the left side. As an example, let the cardinality of (a), (b) be one, and the cardinality of (c) be greater than one.

- (a)-[*1..]->(c): Evaluation starts with (a).
- (c)-[*1..]->(a): Evaluation starts with (a).
- (a)-[*1..]-(c): Evaluation starts with (a).
- (c)-[*1..]-(a): Evaluation starts with (a).

Now let the incoming edges of (a) be two, and the outgoing edges of (a) be three, the incoming edges of (b) be four, and the outgoing edges of (b) be five.

- (a)-[*1..]->(b): Evaluation starts with (a) as the outgoing edges of (a) are less than the incoming edges of (b).
- (a)<-[*1..]-(b): Evaluation starts with (a) as the incoming edges of (a) are less than the outgoing edges of (b).

As a general rule, place the more restrictive pattern on the left side of a VLP expression.

Avoid redundant node label checks by using granular relationship names

When optimizing for performance, using relationship labels that are exclusive to node patterns allows the removal of label filtering on nodes. Consider a graph model where the relationship likes is only used to define a relationship between two person nodes. We could write the following query to find this pattern:

```
MATCH (n:person)-[:likes]->(m:person)
RETURN n, m
```

The person label check on n and m is redundant, as we defined the relationship to only appear when both are of the type person. To optimize on performance, we can write the query as follows:

```
MATCH (n)-[:likes]->(m)
RETURN n, m
```

This pattern can also apply when properties are exclusive to a single node label. Assume that only person nodes have the property email, therefore verifying the node label matches person is redundant. Writing this query as:

```
MATCH (n:person)
WHERE n.email = 'xxx@gmail.com'
RETURN n
```

Is less efficient than writing this query as:

```
MATCH (n)
WHERE n.email = 'xxx@gmail.com'
RETURN n
```

You should only adopt this pattern when performance is important and you have checks in your modeling process to ensure these edge labels are not reused for patterns involving other node labels. If you later introduce an email property on another node label such as company, then the results will differ between these two versions of the query.

Specify edge labels where possible

It is recommended to provide an edge label where possible when specifying an edge in a pattern. Consider the following example query, which is used to link all of the people living in a city with all of the people who visited that city.

```
MATCH (person)-->(city {country: "US"})-->(anotherPerson)
RETURN person, anotherPerson
```

If your graph model links people to nodes other than just cities using multiple edge labels, by not specifying the end label, Neptune will need to evaluate additional paths that will later be discarded. In the above query, as an edge label was not given, the engine does more work first and then filters out values to obtain the correct result. A better version of above query might be:

```
MATCH (person)-[:livesIn]->(city {country: "US"})-[:visitedBy]->(anotherPerson)
RETURN person, anotherPerson
```

This not only helps in evaluation, but enables the query planner to create better plans. You could even combine this best practice with redundant node label checks to remove the city label check and write the query as:

```
MATCH (person)-[:livesIn]->({country: "US"})-[:visitedBy]->(anotherPerson)
```

RETURN person, anotherPerson

Avoid using the WITH clause when possible

The WITH clause in openCypher acts as a boundary where everything before it executes, and then the resulting values are passed to the remaining portions of the query. The WITH clause is needed when you require interim aggregation or want to limit the number of results, but aside from that you should try to avoid using the WITH clause. The general guidance is to remove these simple WITH clauses (without aggregation, order by or limit) to enable the query planner to work on the entire query to create a globally optimal plan. As an example, assume you wrote a query to return all people living in India:

```
MATCH (person)-[:lives_in]->(city)
WITH person, city
MATCH (city)-[:part_of]->(country {name: 'India'})
RETURN collect(person) AS result
```

In the above version, the WITH clause restricts the placement of the pattern (city)-[:part_of]->(country {name: 'India'}) (which is more restrictive) before (person)-[:lives_in]->(city). This makes the plan sub-optimal. An optimization on this query would be to remove the WITH clause and let the planner compute the best plan.

```
MATCH (person)-[:lives_in]->(city)
MATCH (city)-[:part_of]->(country {name: 'India'})
RETURN collect(person) AS result
```

Place restrictive filters as early in the query as possible

In all scenarios, early placement of filters in the query helps in reducing the intermediate solutions a query plan must consider. This means less memory and fewer compute resources are needed to execute the query.

The following example helps you understand these impacts. Suppose you write a query to return all of the people who live in India. One version of the query could be:

```
MATCH (n)-[:lives_in]->(city)-[:part_of]->(country)
WITH country, collect(n.firstName + " " + n.lastName) AS result
WHERE country.name = 'India'
```

RETURN result

The above version of the query is not the most optimal way to achieve this use case. The filter country.name = 'India' appears later in the query pattern. It will first collect all persons and where they live, and group them by country, then filter for only the group for country.name = India. The optimal way to query for only people living in India and then perform the collect aggregation.

```
MATCH (n)-[:lives_in]->(city)-[:part_of]->(country)
WHERE country.name = 'India'
RETURN collect(n.firstName + " " + n.lastName) AS result
```

A general rule is to place a filter as soon as possible after the variable is introduced.

Explicitly check whether properties exist

Based on openCypher semantics, when a property is accessed it is equivalent to an optional join and must retain all rows even if the property does not exist. If you know based on your graph schema that a particular property will always exist for that entity, explicitly checking that property for existence allows the query engine to create optimal plans and improve performance.

Consider a graph model where nodes of type person always have a property name. Instead of doing this:

```
MATCH (n:person)
RETURN n.name
```

Explicitly verify the property existence in the query with an IS NOT NULL check:

```
MATCH (n:person)
WHERE n.name IS NOT NULL
RETURN n.name
```

Do not use named path (unless it is required)

Named path in a query always comes at an additional cost, which can add penalties in terms of higher latency and memory usage. Consider the following query:

```
MATCH p = (n) - [:commentedOn] - >(m)
```

```
WITH p, m, n, n.score + m.score as total
WHERE total > 100
MATCH (m)-[:commentedON]->(o)
WITH p, m, n, distinct(o) as o1
RETURN p, m.name, n.name, o1.name
```

In the above query, assuming we only want to know the properties of the nodes, the use of path "p" is unnecessary. By specifying the named path as a variable, the aggregation operation using DISTINCT will get expensive both in terms of time and memory usage. A more optimized version of above query could be:

```
MATCH (n)-[:commentedOn]->(m)
WITH m, n, n.score + m.score as total
WHERE total > 100
MATCH (m)-[:commentedON]->(o)
WITH m, n, distinct(o) as o1
RETURN m.name, n.name, o1.name
```

Avoid COLLECT(DISTINCT())

COLLECT(DISTINCT()) is used whenever a list is to be formed containing distinct values. COLLECT is an aggregation function, and grouping is done based on additional keys being projected in the same statement. When distinct is used, the input is split in multiple chunks where each chunk denotes one group for reduction. Performance will be impacted as the number of groups increases. In Neptune Analytics, it is much more efficient to perform DISTINCT before actually collecting/ forming the list. This allows grouping to be done directly on the grouping keys for the whole chunk.

Consider the following query:

```
MATCH (n:Person)-[:commented_on]->(p:Post)
WITH n, collect(distinct(p.post_id)) as post_list
RETURN n, post_list
```

A more optimal way of writing this query is:

```
MATCH (n:Person)-[:commented_on]->(p:Post)
WITH DISTINCT n, p.post_id as postId
WITH n, collect(postId) as post_list
RETURN n, post_list
```

Prefer the properties function over individual property lookup when retrieving all property values

The properties() function is used to return a map containing all properties for an entity, and is much more efficient than returning properties individually.

Assuming your Person nodes contain 5 properties, firstName, lastName, age, dept, and company, the following query would be preferred:

```
MATCH (n:Person)
WHERE n.dept = 'AWS'
RETURN properties(n) as personDetails
```

Rather than using:

```
MATCH (n:Person)
WHERE n.dept = 'AWS'
RETURN n.firstName, n.lastName, n.age, n.dept, n.company
=== OR ===
MATCH (n:Person)
WHERE n.dept = 'AWS'
RETURN {firstName: n.firstName, lastName: n.lastName, age: n.age,
department: n.dept, company: n.company} as personDetails
```

Perform static computations outside of the query

It is recommended to resolve static computations (simple mathematical/string operations) on the client-side. Consider this example where you want to find all people one year older or less than the author:

```
MATCH (m:Message)-[:HAS_CREATOR]->(p:person)
WHERE p.age <= ($age + 1)
RETURN m</pre>
```

Here, \$age is injected into the query via parameters, and is then added to a fixed value. This value is then compared with p.age. Instead, a better approach would be doing the addition on the client-side and passing the calculated value as a parameter \$ageplusone. This helps the query

engine to create optimized plans, and avoids static computation for each incoming row. Following these guidelines, a more efficient verson of the query would be:

```
MATCH (m:Message)-[:HAS_CREATOR]->(p:person)
WHERE p.age <= $ageplusone
RETURN m</pre>
```

Batch inputs using UNWIND instead of individual statements

Whenever the same query needs to be executed for different inputs, instead of executing one query per input, it would be much more performant to run a query for a batch of inputs.

If you want to merge on a set of nodes, one option is to run a merge query per input:

```
MERGE (n:Person {`~id`: $id})
SET n.name = $name, n.age = $age, n.employer = $employer
```

With parameters:

```
params = {id: '1', name: 'john', age: 25, employer: 'Amazon'}
```

The above query needs to be executed for every input. While this approach works, it may require many queries to be executed for a large set of input. In this scenario, batching may help reduce the number of queries executed on the server, as well as improve the overall throughput.

Use the following pattern:

```
UNWIND $persons as person
MERGE (n:Person {`~id`: person.id})
SET n += person
```

With parameters:

```
params = {persons: [{id: '1', name: 'john', age: 25, employer: 'Amazon'},
{id: '2', name: 'jack', age: 28, employer: 'Amazon'},
{id: '3', name: 'alice', age: 24, employer: 'Amazon'}...]}
```

Experimentation with different batch sizes is recommended to determine what works best for your workload.

Batch inputs using UNWIND instead of individual statements

Prefer using custom IDs for node

Neptune Analytics allows users to explicitly assign IDs on nodes. The ID must be globally unique in the dataset and deterministic to be useful. A deterministic ID can be used as a lookup or a filtering mechanism just like properties; however, using an ID is much more optimized from query execution perspective than using properties. There are several benefits to using custom IDs -

- Properties can be null for an existing entity, but the ID must exist. This allows the query engine to use an optimized join during execution.
- When concurrent mutation queries are executed, the chances of <u>concurrent modification</u> <u>exceptions</u> (CMEs) are reduced significantly when IDs are used to access nodes because fewer locks are taking on IDs than properties due to their enforced uniqueness.
- Using IDs avoids the chance of creating duplicate data as Neptune enforces uniqueness on IDs, unlike properties.

The following query example uses a custom ID:

i Note

The property ~id is used to specify the ID, whereas id is just stored as any other property.

```
CREATE (n:Person {`~id`: '1', name: 'alice'})
```

Without using a custom ID:

```
CREATE (n:Person {id: '1', name: 'alice'})
```

If using the latter mechanism, there is no uniqueness enforcement and you could later execute the query:

```
CREATE (n:Person {id: '1', name: 'john'})
```

This creates a second node with id=1 named john. In this scenario, you would now have two nodes with id=1, each having a different name - (alice and john).

Avoid doing ~id computations in the query

When using custom IDs in the queries, always perform static computations outside the queries and provide these values in the parameters. When static values are provided, the engine is better able to optimize lookups and avoid scanning and filtering these values.

If you want to create edges between nodes that are existing in the database, one option could be:

```
UNWIND $sections as section
MATCH (s:Section {`~id`: 'Sec-' + section.id})
MERGE (s)-[:IS_PART_OF]->(g:Group {`~id`: 'g1'})
```

With parameters:

```
parameters={sections: [{id: '1'}, {id: '2'}]}
```

In the query above, the id of the section is being computed in the query. Since the computation is dynamic, the engine cannot statically inline ids and ends up scanning all section nodes. The engine then performs post-filtering for required nodes. This can be costly if there are many section nodes in the database.

A better way to achieve this is to have Sec - prepended in the ids being passed into the database:

```
UNWIND $sections as section
MATCH (s:Section {`~id`: section.id})
MERGE (s)-[:IS_PART_OF]->(g:Group {`~id`: 'g1'})
```

With parameters:

```
parameters={sections: [{id: 'Sec-1'}, {id: 'Sec-2'}]}
```

Limits for Neptune Analytics

Regions

Neptune Analytics is available in the following AWS Regions:

- US East (N. Virginia): us-east-1
- US East (Ohio): us-east-2
- US West (Oregon): us-west-2
- Asia Pacific (Singapore): ap-southeast-1
- Asia Pacific (Tokyo): ap-northeast-1
- Europe (Ireland): eu-west-1
- Europe (London): eu-west-2
- Europe (Frankfurt): eu-central-1

Quotas

Your AWS account has default quotas, formerly referred to as limits, for each AWS service. Unless otherwise noted, each quota is Region-specific. You can request increases for some quotas, and other quotas cannot be increased.

To view the quotas for Neptune Analytics, open the <u>Service Quotas console</u>. In the navigation pane, choose **AWS services** and select **Neptune Analytics**.

To request a quota increase, see <u>Requesting a Quota Increase</u> in the *Service Quotas User Guide*. If the quota is not yet available in Service Quotas, use the <u>limit increase form</u>.

Vertex enumeration is not memory bounded

The following quotas and limits apply to Neptune Analytics:

The current implementation of vertex enumeration and counting is not memory bounded. As a consequence, queries such as MATCH (n) RETURN count(n) will require a significant amount of memory and, depending on the chosen capacity and dataset shape, may run into out-of-memory exceptions.

Where possible, we recommend replacing such queries with queries that operate on a perlabel basis. For instance, a query such as MATCH (n : Person) RETURN count(n) will be significantly more efficient, both in terms of memory consumption and memory utilization.

Parameterized openCypher queries not supported for algorithms

Neptune Analytics supports <u>parameterized openCypher queries</u> with the limitation that parameters are not allowed inside algorithms.

For instance, a query such as CALL neptune.algo.degree(\$id) where \$id is passed in as a parameter is currently not supported.

Size limits on properties, labels and strings

The maximum length of the strings supported is 1,048,062 bytes. The limit would be lower for strings with unicode characters since some unicode characters are represented using multiple bytes.

API reference

The <u>Neptune Analytics API reference</u> is available for more information.