



Maturity model for adopting generative AI on AWS

AWS Prescriptive Guidance



AWS Prescriptive Guidance: Maturity model for adopting generative AI on AWS

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Table of Contents

Introduction	1
Intended audience	1
Target business objectives	2
Model overview	3
Maturity levels	3
Maturity aspects	8
Pillars of adoption	9
Focus areas	9
Key activities	9
Transformation strategy	10
Level 1: Envision	11
Focus and criteria	11
Key activities	11
Transformation strategy	16
Level 2: Experiment	18
Focus and criteria	18
Key activities	19
Transformation strategy	22
Level 3: Launch	24
Focus and criteria	24
Key activities	25
Transformation strategy	27
Level 4: Scale	29
Focus and criteria	29
Key activities	30
Continuing the journey	33
Next steps	35
Resources	35
AWS service documentation	35
AWS Prescriptive Guidance	36
Other resources	36
Contributors	37
Authoring	37
Reviewing	37

Technical writing	37
Document history	38
Glossary	39
#	39
A	40
B	43
C	45
D	48
E	52
F	54
G	56
H	57
I	58
L	60
M	62
O	66
P	68
Q	71
R	71
S	74
T	78
U	79
V	80
W	80
Z	81

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[Generative AI](#) is a subset of AI models that have been trained on large amounts of data and can generate new content, including text, images, music, and video. The models can use pretrained [foundation models](#), custom models, and augmented or proprietary datasets. The impact of generative AI spans industries. It can enhance creativity, improve productivity, and enable new business models. If your organization wants generative AI to enhance operations, drive innovation, and deliver business growth, a structured, phased approach is crucial for navigating the adoption journey.

According to a [CIO article](#), 88% of AI pilots fail to reach production. This leads to what is termed *pilot fatigue*. The article says that "Companies are simply weary of spending more time, money, and energy to support pilots that do not progress into production quickly or at all." This fatigue can stifle innovation and discourage further experimentation with generative AI. Additionally, according to a [McKinsey report](#), organizations are grappling with significant data quality and integration challenges in their AI implementations.

This strategy document provides a structured framework to help organizations implement generative AI solutions. This framework is designed to help you navigate the complexities of technology adoption and make sure that you do not overlook critical steps or best practices. Use the recommendations in this guide to comprehensively understand your generative AI maturity. By assessing the maturity level, you can identify focus areas for each level and launch an end-to-end generative AI adoption journey. This framework explores four maturity levels, from initial awareness to full-scale transformation. It outlines key activities and essential practices for each level.

Intended audience

This article is intended for executives, directors of technology, business leaders, data scientists, generative AI and AI/ML specialists, IT professionals, and decision-makers who want to create value by adopting generative AI in their organizations.

Target business objectives

Through systematic progression through the generative AI maturity levels, organizations can achieve the following key business outcomes:

- Strategic business process innovation through validated generative AI use cases
- Operational excellence through robust, production-ready AI solutions
- Enterprise-wide efficiency through standardized, reusable AI components
- Competitive advantage through strategic transformation and scalable AI capabilities

Overview of the generative AI maturity model

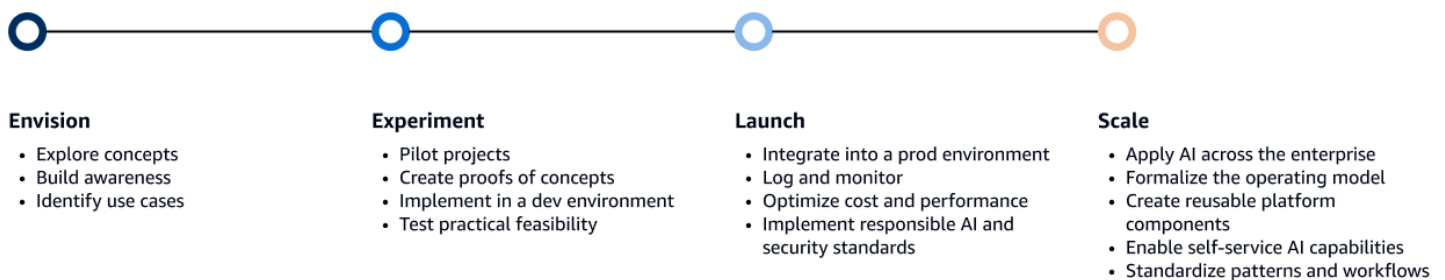
Organizations can use the framework of the maturity model to effectively integrate generative AI capabilities and avoid common implementation pitfalls and implementation gaps. Through a detailed maturity assessment, you can gain clarity on where your organization stands in its AI journey and pinpoint specific areas that require attention. The progression spans four distinct levels, beginning with basic understanding and culminating in complete organizational transformation. Each level contains focused action items and strategic guidelines to drive success.

This section contains the following topics:

- [Levels in the generative AI maturity model](#)
- [Aspects of generative AI maturity](#)

Levels in the generative AI maturity model

The generative AI maturity model is structured across four primary levels. Each level represents an organization's progress toward using generative AI capabilities. This model can help organizations understand where they currently stand and guide them toward the next steps in their generative AI journey. The following diagram shows the four levels of the generative AI maturity model and key activities for each level.



The following are the four levels in the generative AI maturity model:

- [Level 1: Envision](#)
- [Level 2: Experiment](#)
- [Level 3: Launch](#)
- [Level 4: Scale](#)

The labels for each maturity level reflect the impact of generative AI adoption within the organization. As you identify your organization's position at a given level, you can gain insights into the opportunities in the next level of maturity. Lower levels generally encompass more tactical generative AI use cases, and the higher levels tend to be more strategic and transformative in nature.

Many organizations will find that characteristics of multiple maturity levels apply across their teams and use cases. This is because no single level is inherently superior or inferior - the appropriate maturity level is contextual to the organization's goals and readiness.

Note

This generative AI maturity model is not intended to classify an organization or its generative AI capabilities as solely *beginner* or *transformative*. Rather, each aspect of generative AI adoption should be considered independently. The characteristics of each maturity level represent a continuum within that specific aspect, but are not necessarily correlated to the same level across other aspects.

The following table provides an overview of the four levels.

Category	Level 1: Envision	Level 2: Experiment	Level 3: Launch	Level 4: Scaling
Description	Organizations explore generative AI concepts, build awareness, and identify potential use cases.	Organizations validate generative AI's potential through structured pilot projects and proofs of concepts, while building core technical capabilities and foundational	Organizations systematically deploy production-ready generative AI solutions with robust governance, monitoring, and support mechanisms to deliver consistent value and operation	Organizations establish enterprise-wide generative AI capabilities through reusable components, standardized patterns, and self-service platforms to accelerate adoption while

Category	Level 1: Envision	Level 2: Experiment	Level 3: Launch	Level 4: Scaling
		frameworks for implementation.	al excellence while maintaining security and compliance standards.	maintaining automated governance and fostering innovation.
Focus	Build awareness and understanding of generative AI technologies, explore potential applications, and identify areas where AI can add value to the business	Validate business values through structured pilot programs and build core competencies	Deploy production-ready solutions that deliver measurable business value through robust launch processes , comprehensive governance frameworks, and performance monitoring	Create reusable components and patterns that accelerate generative AI adoption across the enterprise

Category	Level 1: Envision	Level 2: Experiment	Level 3: Launch	Level 4: Scaling
Criteria	<ul style="list-style-type: none"> Gain a basic understanding of generative AI concepts No formal projects or resource allocation Gain awareness of industry trends and value opportunities 	<ul style="list-style-type: none"> Run pilot projects and proofs of concepts Form small teams to explore generative AI capabilities Establish foundational and governance frameworks 	<ul style="list-style-type: none"> Release a few generative AI applications into production Implement risk, governance, and responsible AI policies for generative AI applications Establish operational and support teams 	<ul style="list-style-type: none"> Broadly adopt generative AI across various departments in the organization Release many generative AI applications into production Prioritize investments in generative AI infrastructure and tools Formalize the operating model and responsible, accountable, consulted, informed (RACI) matrix

Category	Level 1: Envision	Level 2: Experiment	Level 3: Launch	Level 4: Scaling
Key activities	<ul style="list-style-type: none"> Attend AI awareness training, workshops, and conferences Engage with AI subject matter experts and consultants Explore potential use cases and business benefits Evaluate cultural readiness Evaluate generative AI governance Build knowledge 	<ul style="list-style-type: none"> Define and refine business use cases for pilot projects Develop proofs of concepts Evaluate and select appropriate generative AI models and tooling Measure business benefits realization Build internal capabilities and technical expertise 	<ul style="list-style-type: none"> Initialize an operating model Create solution architecture governance Create a production-ready implementation strategy Establish monitoring and performance tracking mechanisms Implement risk and governance management Integrate an IT Infrastructure Library (ITIL) framework Set up the operation and support structure 	<ul style="list-style-type: none"> Formalize the generative AI operating model and RACI matrix Create reusable generative AI capabilities and components Standardize generative AI use case patterns Establish an organization-wide collaborative development framework Evolve AI capabilities into an internal development platform (IDP) or software as a service (SaaS)

Category	Level 1: Envision	Level 2: Experiment	Level 3: Launch	Level 4: Scaling
				<ul style="list-style-type: none">• Share and democratize knowledge

To further explain and understand the maturity model, it's important to understand how organizations typically progress in their generative AI adoption journey. This progression reflects not only how organizations use generative AI capabilities, but also what motivates them to advance their adoption. In the early levels, many users might not have formalized AI processes at all. Rather, they see their tools as an improved collection of capabilities from various internal sources. As organizations mature, these capabilities become more consistently managed and standardized. Eventually, as the capabilities become more refined and discoverable and as users naturally opt into using AI capabilities, organizations typically shift away from external motivations such as mandates or incentives. Ideally, they even start to invest their own efforts into wider AI innovation and development.

Aspects of generative AI maturity

The successful adoption of generative AI requires a holistic understanding of multiple organizational dimensions. This section examines four key aspects that organizations must consider and develop throughout their maturity journey: the fundamental pillars that support AI adoption, the focus areas that guide strategic priorities, the key activities that drive implementation, and the transformation strategy that guides the organization's maturity advancement. Together, these aspects provide a comprehensive framework for assessing and advancing generative AI capabilities. Organizations can use this framework to identify gaps, prioritize investments, and create actionable plans for progression through the maturity levels. Each aspect has been chosen based on extensive field experience with enterprise AI adoption. They reflect the critical elements that distinguish successful implementations from unsuccessful ones.

This section contains the following topics:

- [Pillars of adoption](#)
- [Focus areas](#)
- [Key activities](#)
- [Transformation strategy](#)

Pillars of adoption

Each maturity level is evaluated across the following pillars of adoption:

- **Business** – Strategic alignment and measurable impact on business goals
- **People** – Talent development, skill-building, and cross-functional collaboration
- **Governance** – Establishment of risk management, compliance, and ethical guidelines
- **Platform** – Investment in scalable infrastructure and platforms for generative AI capabilities
- **Security** – Protecting data, privacy, and the deployment of generative AI models
- **Operations** – Managing generative AI solution lifecycles, optimizing deployments, implementing feedback mechanisms, and monitoring performance

These pillars align with and extend the [AWS Cloud Adoption Framework \(AWS CAF\)](#) to address generative AI needs. The recommendations in this strategy document add AI-specific elements to each pillar, such as ethical AI implementation, model lifecycle management, and AI infrastructure requirements. This alignment helps organizations use existing AWS CAF best practices while addressing unique AI adoption challenges.

Focus areas

The focus areas for each maturity level help organizations prioritize activities and investments. The following are the four focus areas:

- **Innovation and feasibility** – Exploring and validating innovative generative AI use cases and the availability and quality of required datasets
- **Integration and efficiency** – Integrating generative AI into existing business processes
- **Scalability and optimization** – Scaling generative AI applications and continuously improving performance
- **Transformation and leadership** – Using generative AI to drive strategic shifts and gain a competitive edge

Key activities

Organizations can use the key activities in the generative AI maturity model to navigate their journey and successfully define and implement their generative AI strategy. The activities progress

from initial exploration and understanding of generative AI technologies, to experimenting with prototypes, integrating AI solutions into business processes, scaling them across the organization, and then establishing governance for continuous improvement and strategic transformation. Key activities fall into one of the following categories:

- **Exploration and awareness** – Develop foundational knowledge of generative AI technologies and identify strategic opportunities for adoption
- **Experimentation and validation** – Facilitate and conduct pilot projects and prototypes to assess technical feasibility and business value
- **Integration and implementation** – Embed generative AI capabilities into existing business processes and deploy solutions into production environments
- **Scaling and optimization** – Integrate generative AI applications across the organization and continuously improve their performance and efficiency
- **Governance and leadership** – Establish frameworks and best practices for managing generative AI initiatives and using them for strategic transformation

Transformation strategy

The transformation strategy at each level focuses on guiding organizations through incremental improvements. This includes developing a generative AI roadmap and a data strategy, aligning with business goals, investing in talent and tools, and implementing governance frameworks.

Generative AI maturity model level 1: Envision

This foundational level serves as a critical starting point where organizations explore generative AI concepts, build organizational awareness, and identify potential use cases that align with their business objectives. By establishing this essential groundwork, companies can develop a clear vision for their AI journey while addressing key considerations across business, people, governance, platform, security, and operational dimensions.

This section includes the following topics:

- [Focus and criteria](#)
- [Key activities](#)
- [Transformation strategy to reach the next level](#)

Focus and criteria

The goal at this level is to build a foundational understanding and awareness of generative AI technologies and emerging industry trends related to this technology. This includes assessing potential applications and identifying areas where generative AI could benefit the business. This level focuses on educating stakeholders about generative AI and beginning to explore use cases and conduct risk and cultural readiness assessment.

The following are the criteria for being at this level:

- The organization has demonstrated basic knowledge of generative AI fundamentals.
- The organization has documented awareness of industry generative AI applications and opportunities.
- The organization has an emerging understanding of its cultural readiness for AI.
- The organization has performed an initial exploration of potential use cases and benefits.
- The organization has given preliminary consideration to governance and security requirements.

Key activities

The following table shows the key activities for each pillar of adoption.

Pillar of adoption	Activities	
Business	<ul style="list-style-type: none">• Understand how generative AI can solve specific business problems.• Map initial generative AI use cases to business objectives, such as improving customer engagement or automating content creation.• Identify high-value data sources in relation to selected use cases.	
People	<ul style="list-style-type: none">• Conduct internal training sessions and knowledge-sharing workshops.• Identify AI champions within the organization to lead the exploration of generative AI opportunities.• Evaluate your organization's culture and change management readiness for generative AI adoption.• Assess the current technological skill gaps in your organization, and determine the required investments for generative AI adoption.• Design educational initiatives to help senior executives understand AI's strategic	

Pillar of adoption	Activities	
	<p>potential, technological capabilities, transformative business impact, and the importance of data in generative AI projects.</p> <ul style="list-style-type: none">• Attend industry forums and conferences to learn from the AI adoption experiences of other companies.• Organize internal hackathons to encourage experimentation and foster innovation.	
Governance	<ul style="list-style-type: none">• Explore ethical and regulatory considerations for generative AI adoption, such as privacy and data sovereignty.• Develop an initial set of guidelines for responsible AI use in the organization.	

Pillar of adoption	Activities	
Platform	<ul style="list-style-type: none">• Explore the requirements for adopting generative AI to align with your organization's standards.• Explore AI/ML models and tooling, such as Amazon Bedrock for accessing foundation models and Amazon SageMaker AI, for quick experimentation.• Assess and catalog existing internal and external data sources. Evaluate the data infrastructure and quality to determine generative AI feasibility and potential implementation requirements.	

Pillar of adoption	Activities	
Security	<ul style="list-style-type: none">• Understand the security implications and tasks associated with adopting generative AI in the organization, such as:<ul style="list-style-type: none">• Data privacy and protection risks, which includes potential exposure of sensitive information through training data, prompts, and model outputs• Access control and authentication challenges, which encompasses the complexities of user verification and role-based permissions in AI systems• Model security vulnerabilities, which includes susceptibility to prompt injection attacks and the potential for generating unsafe or inappropriate content	

Pillar of adoption	Activities	
Operations	<ul style="list-style-type: none"> Understand the operational challenges associated with adopting generative AI in the organization, such as: <ul style="list-style-type: none"> Plan for performance monitoring needs for your AI solutions. Consider governance and versioning requirements. Understand what is required for incident response procedures. 	

Transformation strategy to reach the next level

To progress to the next maturity level, consider the following aspects:

- **Establish cross-functional generative AI squads** – Form cross-functional generative AI squads that have clear roles and responsibilities. Squads should include IT representatives, business representatives, security and governance stakeholders, and generative AI SMEs who can lead experimentation efforts. This group will form the foundation for a more formally defined center of excellence (CoE) later, as you scale your generative AI efforts.
- **Identify and prioritize use cases** – Develop a use case matrix that helps you prioritize projects based on feasibility, business impact, and alignment with strategic goals. For proofs of concepts (PoCs), create a short list of the top use cases.
- **Allocate resources for pilot projects** – Secure budget and personnel for running small-scale PoCs.
- **Develop generative AI skills** – Upskill staff on specific tools and technologies, such as [Amazon Bedrock](#), [SageMaker AI](#), [Amazon Q Business](#), [Amazon Q Developer](#), [prompt engineering](#), [Retrieval Augmented Generation \(RAG\)](#), and agentic AI and workflows.
- **Complete preliminary governance** – Establish preliminary governance that guides the use of generative AI. It should cover compliance, risk management, and ethical considerations.

- **Cultural readiness** – Begin planning organizational change management for company-wide generative AI adoption.
- **Identify success metrics** – For each PoC, define the success criteria and the business and technical metrics.

By taking these actions, organizations can expect to:

- Gain practical experience with generative AI technologies.
- Validate the feasibility and potential impact of specific use cases.
- Build internal capabilities and expertise in generative AI.
- Identify potential challenges and risks associated with generative AI adoption.
- Improve the readiness of generative AI adoption in order advance to the next level.

Generative AI maturity model level 2: Experiment

Building upon the foundational awareness established in the previous level, the Experiment level marks a crucial transition from theoretical exploration to practical implementation of generative AI technologies. At this level, organizations move beyond conceptual understanding to engage in hands-on PoC projects and pilot programs. These PoC and pilot projects are designed to validate business value and build core competencies. This level is characterized by structured experimentation, where organizations form dedicated teams, establish governance frameworks, and begin developing internal technical expertise. Through carefully controlled pilot projects, organizations can test their hypotheses about generative AI's potential while minimizing risks and maximizing learning opportunities. This sets the stage for broader implementation and scaling of successful initiatives.

This section includes the following topics:

- [Focus and criteria](#)
- [Key activities](#)
- [Transformation strategy to reach the next level](#)

Focus and criteria

At this level, organizations transition from exploration to hands-on PoC experimentation and pilot projects with generative AI technologies. The focus is on validating business value through structured pilot programs and building core competencies. This level emphasizes practical learning, building internal capabilities and technical expertise, and establishing foundational and governance frameworks.

The following are the criteria for being at this level:

- The organization has active pilot projects and proofs of concept in progress.
- Dedicated, cross-functional teams are assigned to generative AI initiatives.
- A structured internal training program is established.
- The organizations has selected and validated AI models and tools.
- The organization has defined its initial governance and data frameworks.

Key activities

The following table shows the key activities for each pillar of adoption.

Pillar of adoption	Activities
Business	<ul style="list-style-type: none">• Define and prioritize strategic use cases based on business value and feasibility.• For PoCs, establish success metrics and frameworks for measuring the return on investment (ROI).• Create value assessment scorecards for each PoC.• Limit the scope of PoCs to a manageable scale with clear success metrics.• For each PoC, measure the ROI and evaluate whether it achieved the success criteria.
People	<ul style="list-style-type: none">• Implement structured training programs in prompt engineering, RAG, and model fine-tuning.• Create generative AI certification paths and career progression frameworks.• Hire generative AI and data science experts.• Partner with external specialists, such as the AWS Generative AI Innovation Center or AWSProfessional Services, to co-build a PoC, provide support, and transfer knowledge.• Establish AI certification paths and career-pr ogression frameworks.
Governance	<ul style="list-style-type: none">• Develop preliminary frameworks that encompass data governance for generativ e AI, such as the quality of content used for vector search.

Pillar of adoption	Activities
	<ul style="list-style-type: none">• Establish model evaluation criteria and quality controls.• Set up risk assessment protocols for generative AI projects.• Set up guidelines for the ethical and responsible use of generative AI. Train developers, data scientists, and generative AI specialists to comply with these guidelines.

Pillar of adoption	Activities
Platform	<ul style="list-style-type: none">• Set up the foundation infrastructure for the PoC, such as an AWS landing zone and the permissions that developers need.• Set up an environment for generative AI experimentation and PoC development, such as an Amazon Bedrock playground or an Amazon SageMaker AI JupyterLab space or notebook instance.• Implement a RAG approach or an agentic workflow that developers can easily use. For a RAG approach, consider Amazon Bedrock Knowledge Bases, and for an agentic workflow, consider Amazon Bedrock Agents.• Set up frameworks or pipelines that manage prompts, models, and prompt evaluations. These resources should help developers quickly evaluate the results and performance of the PoC application.• Implement early-stage data-integration efforts, including structured and unstructured data pipelines. Set up vector databases for RAG experiments.• Evaluate foundation models based on cost, performance, and use case suitability. You can use Amazon Bedrock, Amazon SageMaker AI, and Amazon SageMaker AI JumpStart.

Pillar of adoption	Activities
Security	<ul style="list-style-type: none">• Implement data access controls for training generative AI models, and make sure that they adhere to compliance requirements. Amazon Q Business can simplify the implementation of RAG by enabling fine-grained controls that allow generative AI workloads to retrieve only the data that the user is authorized to access.• Develop a strategy for protecting personally identifiable information (PII) in datasets that are used to train models.
Operations	<ul style="list-style-type: none">• Create documentation and support processes for the following:<ul style="list-style-type: none">• PoC implementations and learnings• Basic platform configurations and security controls• Testing and evaluation procedures• Hand-over processes for successful PoCs that are moving to production

Transformation strategy to reach the next level

Organizations can transition to next maturity level by doing the following:

- **Create production-grade infrastructure to support generative AI** – Use AWS services to implement CI/CD pipelines, standardized deployment patterns, and proper scaling mechanisms for production deployments.
- **Implement governance** – Establish production-grade governance frameworks to manage ongoing generative AI usage and model updates.
- **Implement observability** – Implement observability, monitoring, and logging practices that are specifically adapted for generative AI workloads. This includes model performance metrics, usage patterns, and response quality assessment.

- **Focus on compliance** – Make sure that you comply with industry standards and regulations for data privacy and security.
- **Build dedicated AI teams** – Set up a team that creates and maintains standardized paths to production for generative AI solutions.
- **Implement operational excellence** – Create an incident response and escalation process. Establish service-level agreements (SLAs) and performance metrics. Implement cost optimization strategies.

By taking these actions, organizations can:

- Validate that generative AI applications are stable, reliable, and continuously deliver value to the organization.
- Support the growth of generative AI solutions as demand and usage increase across various departments.
- Manage risks, maintain oversight, and align AI initiatives with regulatory standards as they become an integral part of business operations.
- Provide continuous monitoring, improvement, and support for generative AI solutions. This reduces the reliance on ad-hoc or temporary project teams.
- Prepare the organization to move from isolated projects to a strategic and cohesive approach, where AI becomes a core enabler of business processes. The organization is ready for further scale and broader adoption.

Generative AI maturity model level 3: Launch

At this level, organizations transition from proof-of-concept initiatives to the methodical deployment of select, proven generative AI solutions into production environments. This level represents a pivotal shift away from experimentation to focus on robust governance protocols, real-time monitoring systems, and dedicated support infrastructures. Companies focus on launching a few production-grade applications that demonstrate clear business impact. This level emphasizes operational rigor - implementing comprehensive launch frameworks, establishing clear governance guidelines, and maintaining strong security standards. Releasing reliable generative AI solutions that deliver quantifiable results prepares the organization for broader adoption.

This section includes the following topics:

- [Focus and criteria](#)
- [Key activities](#)
- [Transformation strategy to reach the next level](#)

Focus and criteria

At this level, organizations systematically deploy generative AI solutions into production environments and implement robust governance, monitoring, and support mechanisms. These mechanisms deliver consistent value and operational excellence while maintaining security and compliance standards. The focus shifts from experimental generative AI applications to deploying production-ready solutions that deliver measurable business value through robust launch processes, comprehensive governance frameworks, and systematic performance monitoring. This level focuses on deploying a select number of production-ready generative AI solutions that serve as foundational implementations for launch frameworks and governance mechanisms.

The following are the criteria for being at this level:

- Production-ready generative AI solutions are delivering measurable business outcomes.
- The organization has implemented baseline security, governance, and responsible AI frameworks.
- Operational controls are established and include automated monitoring and alerting systems.
- The organization has defined a human-in-the-loop process for AI decisions.
- For cross-functional AI teams, preliminary roles and operational responsibilities have been defined.

Key activities

The following table shows the key activities for each pillar of adoption.

Pillar of adoption	Activities
Business	<ul style="list-style-type: none">• Sign off on a first version of a RACI matrix for generative AI operations.• Identify key roles that are needed for platform architecture, development, and support.• Measure operational efficiency and business value through comprehensive dashboards.• Track and optimize operational costs and resource utilization.
People	<ul style="list-style-type: none">• Create generative AI platform teams or squads for architecture, development, and maintenance.• Implement an always available, tiered support structure and training programs.
Governance	<ul style="list-style-type: none">• Obtain formal architecture endorsements from an enterprise architecture review board.• Establish a responsible AI policy framework and secure stakeholder approvals.• Create a cross-functional oversight committee for AI implementation reviews.• For generative AI solutions, maintain documentation for governance approvals , risk assessments, standardized design patterns, and technical specifications.
Platform	<ul style="list-style-type: none">• Implement automated CI/CD pipelines for generative AI solutions.

Pillar of adoption	Activities
	<ul style="list-style-type: none">• Deploy infrastructure as code (IaC) to manage AWS resources.• Document design patterns and technical specifications for generative AI solutions.• Maintain CMDB records for generative AI platform components.
Security	<ul style="list-style-type: none">• Implement robust security controls for generative AI solutions and their data pipelines.• Implement a preliminary policy for responsible AI.• Optimize scalable infrastructure to support real-time data ingestion, vector search, and fine-tuning.• Conduct regular security assessments and audits.• Deploy Amazon Bedrock Guardrails to standardize safety and privacy controls across generative AI applications.

Pillar of adoption	Activities
Operations	<ul style="list-style-type: none">• Establish SLA frameworks and performance metrics.• Monitor model performance and guardrail violations. Set up alerts.• Create operational dashboards that have automated alerting systems.• Follow ITIL processes for change management and asset management.• Established a centralized knowledge repository that contains operational runbooks, playbooks, FAQs, and troubleshooting guides.• Establish data observability practices. Track data lineage, provenance, and quality metrics to identify gaps before scaling.• Establish tiered support levels that have clear escalation paths.• Implement regular performance reviews and analyze customer feedback.

Transformation strategy to reach the next level

To scale generative AI initiatives, organizations should:

- **Formalize the generative AI operating model** – Formalize the RACI matrix across the organization.
- **Enhance the generative AI platform** – Conduct assessment of existing generative AI implementations to identify reusable patterns and components. Evaluate whether the technology stack is ready to scale. Start to envision and design modular architecture that has centralized prompt management, automated evaluation frameworks, and standardized patterns for efficient scaling of generative AI solutions.

- **Expand use cases** – Integrate AI capabilities across multiple departments and explore new applications.
- **Improve the developer experience** – Transform the existing platform into a self-service internal platform. This platform is a comprehensive environment that provides standardized tools, workflows, and governance for AI development across the enterprise.
- **Share knowledge** – Establish inner-source practices and create a component marketplace for sharing reusable AI assets across teams. *Inner-source practices* is the strategy of applying an open source development approach within an organization.
- **Set up operational scaling** – Enhance your support infrastructure with automated incident response and capacity planning. This prepares the infrastructure to scale for enterprise-wide adoption of generative AI.
- **Invest in advanced analytics** – Use advanced analytics tools in the cloud, such as [Amazon CloudWatch](#) for metrics and [Amazon QuickSight](#) for visualization, to use data analytics for continuous improvement.
- **Review the data governance model** – Assess whether your data governance model currently supports self-service capabilities while maintaining standardized policies and access controls. An overly restrictive or centralized approach might hinder your ability to scale data initiatives beyond the core team, especially across diverse business units.

By taking these actions, organizations can:

- Scale generative AI initiatives across the organization for broad impact.
- Continue to enhance the platform while identifying opportunities to improve productivity and reusability.
- Improve the developer experience and reduce cognitive loads.
- Foster a data-driven culture.
- Attract top talent by positioning the organization as a generative AI leader.

Generative AI maturity model level 4: Scale

Level 4 of the generative AI maturity model, the Scale level, transitions from operational excellence to scalable innovation. Organizations begin to move beyond individual production deployments to create a robust ecosystem of reusable components, standardized patterns, and automated workflows. This ecosystem helps organizations to accelerate generative AI adoption across multiple departments while maintaining robust governance and cost optimization. By establishing scalable architectures and self-service capabilities, this maturity level empowers enterprises to efficiently deploy numerous generative AI applications, which ultimately drive organization-wide transformation and sustainable innovation.

This section includes the following topics:

- [Focus and criteria](#)
- [Key activities](#)

Focus and criteria

At this level, organizations transition from operational excellence to scalable innovation, focusing on creating reusable components and patterns that accelerate generative AI adoption across the enterprise. The emphasis shifts from individual production deployments to building capabilities that enables self-service capabilities, standardized patterns, and automated workflows while optimizing costs and maintaining governance at scale. Unlike Level 3 which focuses on select production workloads, Level 4 enables rapid deployment of a large number of generative AI applications through standardized and reusable components, achieving enterprise-wide efficiency and productivity gains.

The following are the criteria for being at this level:

- Multiple departments have adopted widespread use of generative AI.
- The organization has established an enterprise-wide generative AI infrastructure and tooling ecosystem.
- An operating model and RACI matrix are defined and implemented.
- An available library includes standardized, reusable AI components, patterns, and applications. Self-service capabilities make the library accessible across the organization.
- Automated governance mechanisms operate at an enterprise-wide scale.

- The organization has evidence of sustained innovation practices and outcomes.

Key activities

The following table shows the key activities for each pillar of adoption.

Pillar of adoption	Activities
Business	<ul style="list-style-type: none">• Align generative AI projects with long-term business goals. Focus on revenue growth, cost reduction, and customer satisfaction.• Drive enterprise-wide generative AI adoption through reusable components and standardized patterns that deliver value.• Finalize the generative AI operating model and RACI matrix for scaled operations.• Establish specialized squads for platform architecture, development, and maintenance.• Create standardized governance and approval workflows.• Implement advanced analytics and monitoring for continuous improvement.• Establish a proactive approach to identify the next innovative and high value use cases for AI. Consider internal use cases that improve productivity and external use cases that focus on products.• Evaluate complex decision-making automation opportunities• Assess personalization and product enhancement possibilities

Pillar of adoption	Activities
People	<ul style="list-style-type: none">• Cross-train staff to use generative AI tools and foster a culture of continuous learning and innovation.• Within the center of excellence, develop mentorship programs that transfer knowledge from generative AI experts to other team members.• Use an inner-source or crowd-source model to help accelerate the development of the generative AI reusable components.• Run AI certification programs through a center of excellence.
Governance	<ul style="list-style-type: none">• Establish enterprise-wide AI governance and ethics frameworks that cover data usage, model fairness, and transparency.• Scale responsible AI practices through standardized frameworks and automated guardrails.• Establish contribution guidelines and quality standards.

Pillar of adoption	Activities
Platform	<ul style="list-style-type: none">• Develop reusable AI components, such as microservices architectures and automated pipelines for evaluating solutions with human oversight.• Create standardized solution templates such as RAG implementations and agentic workflows.• Establish a standardized blueprint to integrate with third-party tools, using industry standards such as Model Context Protocol (MCP).• Implement self-service capabilities through an internal portal, such as an API-first integration architecture and a component marketplace.
Security	<ul style="list-style-type: none">• Implement enterprise-grade security controls and automated compliance verification.
Operations	<ul style="list-style-type: none">• Build process and guidelines to support an inner-source or crowd-source development model.• Deploy comprehensive observability frameworks.• Create dashboards that help you monitor performance.• Implement automated systems to collect feedback.

Continuing the maturity journey

For organizations that have successfully achieved Level 4 in the generative AI maturity model, you can continue to advance to even higher levels of sophistication. Doing so requires a comprehensive strategy that extends beyond technical implementation. This progression demands strategic initiatives that embed generative AI deeply into the organization's DNA, combining organizational vision, cultural transformation, and technical excellence. To transcend four maturity levels, organizations must strengthen their internal capabilities, forge strategic partnerships, and invest in cutting-edge research and development. This comprehensive advancement strategy, coupled with a strong emphasis on talent development, enables enterprises to move beyond scaled operations toward transformative AI leadership. This unlocks greater operational efficiency and sustainable competitive advantages.

Consider the following actions to progress beyond the maturity model:

- **Embed generative AI in the organization's strategic vision** – Position generative AI as a core component of the company's mission and vision. Make sure that you use its capabilities to drive strategic initiatives and maintain a competitive advantage.
- **Foster a culture of continuous innovation** – Encourage employees to explore new applications of generative AI, and reward experimentation that aligns with business goals.
- **Collaborate with industry partners and academia** – Engage in research partnerships, and collaborate with external experts to stay at the forefront of AI innovation.
- **Invest in cutting-edge generative AI research and development** – Dedicate resources to exploring new methodologies, such as multi-modal AI and advanced reinforcement learning, that can push the boundaries of generative AI.
- **Attract and retain top generative AI talent** – Focus on building a strong talent pipeline by offering attractive incentives, professional development opportunities, and a collaborative environment.

By continuing to scale generative AI solutions across the organization, enterprises can achieve the following benefits:

- **Broad impact across business units** – Generative AI solutions become embedded in daily operations across multiple departments, which enhances productivity and drives efficiency.
- **Enhanced decision-making** – With real-time insights and predictive capabilities from generative AI, organizations can make faster, data-driven decisions.

- **Strategic competitive advantage** – By using generative AI for innovation and optimization, organizations can differentiate themselves from competitors and open new revenue streams.
- **Mature generative AI platform/blueprints and optimized resource management** – By automating processes and improving management of generative solutions, you can reduce operational costs and improve scalability.

Next steps

The generative AI maturity model provides a structured approach for organizations to navigate their generative AI adoption journey on AWS. Understanding the different maturity levels and activities helps organizations assess their readiness and take informed steps toward realizing the full potential of generative AI. This framework helps organizations develop tailored strategies that align with their unique business objectives so that generative AI becomes a key driver of growth and innovation.

It's important to recognize that the adoption of generative AI is not a one-size-fits-all process. Each organization's journey is unique, and it's influenced by factors such as industry, business objectives, and existing technological capabilities. However, this strategy document serves as a valuable guide. It offers a framework for organizations to evaluate their readiness, identify gaps, and implement the necessary measures to successfully use the transformative potential of generative AI.

As organizations embark on their generative AI adoption journey, they should remain agile and adaptable. Continuously reassess your maturity level and adjust your strategies accordingly. The rapid pace of innovation in the field of AI necessitates a commitment to continuous learning, skill development, and the adoption of best practices.

By following this guidance and using AWS AI/ML services, organizations can unlock new opportunities, drive efficiency, and achieve sustained competitive advantage in an increasingly AI-driven world.

Resources

The following resources can help you learn more about adopting generative AI.

AWS service documentation

- [Amazon Bedrock](#)
- [Amazon Bedrock Guardrails](#)
- [Amazon Q Business](#)
- [Amazon Q Developer](#)
- [Amazon SageMaker AI](#)

AWS Prescriptive Guidance

- [Accelerating software development lifecycles on AWS with generative AI](#)
- [Generative AI workload assessment](#)
- [Retrieval Augmented Generation options and architectures on AWS](#)
- [Transforming application development and maintenance operating models on AWS with generative AI](#)

Other resources

- [The state of AI: How organizations are rewiring to capture value](#) (McKinsey report)
- [88% of AI pilots fail to reach production — but that's not all on IT](#) (CIO article)

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Document history

The following table describes significant changes to this guide. If you want to be notified about future updates, you can subscribe to an [RSS feed](#).

Change	Description	Date
Initial publication	—	June 4, 2025

AWS Prescriptive Guidance glossary

The following are commonly used terms in strategies, guides, and patterns provided by AWS Prescriptive Guidance. To suggest entries, please use the **Provide feedback** link at the end of the glossary.

Numbers

7 Rs

Seven common migration strategies for moving applications to the cloud. These strategies build upon the 5 Rs that Gartner identified in 2011 and consist of the following:

- Refactor/re-architect – Move an application and modify its architecture by taking full advantage of cloud-native features to improve agility, performance, and scalability. This typically involves porting the operating system and database. Example: Migrate your on-premises Oracle database to the Amazon Aurora PostgreSQL-Compatible Edition.
- Replatform (lift and reshape) – Move an application to the cloud, and introduce some level of optimization to take advantage of cloud capabilities. Example: Migrate your on-premises Oracle database to Amazon Relational Database Service (Amazon RDS) for Oracle in the AWS Cloud.
- Repurchase (drop and shop) – Switch to a different product, typically by moving from a traditional license to a SaaS model. Example: Migrate your customer relationship management (CRM) system to Salesforce.com.
- Rehost (lift and shift) – Move an application to the cloud without making any changes to take advantage of cloud capabilities. Example: Migrate your on-premises Oracle database to Oracle on an EC2 instance in the AWS Cloud.
- Relocate (hypervisor-level lift and shift) – Move infrastructure to the cloud without purchasing new hardware, rewriting applications, or modifying your existing operations. You migrate servers from an on-premises platform to a cloud service for the same platform. Example: Migrate a Microsoft Hyper-V application to AWS.
- Retain (revisit) – Keep applications in your source environment. These might include applications that require major refactoring, and you want to postpone that work until a later time, and legacy applications that you want to retain, because there's no business justification for migrating them.

- Retire – Decommission or remove applications that are no longer needed in your source environment.

A

ABAC

See [attribute-based access control](#).

abstracted services

See [managed services](#).

ACID

See [atomicity, consistency, isolation, durability](#).

active-active migration

A database migration method in which the source and target databases are kept in sync (by using a bidirectional replication tool or dual write operations), and both databases handle transactions from connecting applications during migration. This method supports migration in small, controlled batches instead of requiring a one-time cutover. It's more flexible but requires more work than [active-passive migration](#).

active-passive migration

A database migration method in which the source and target databases are kept in sync, but only the source database handles transactions from connecting applications while data is replicated to the target database. The target database doesn't accept any transactions during migration.

aggregate function

A SQL function that operates on a group of rows and calculates a single return value for the group. Examples of aggregate functions include SUM and MAX.

AI

See [artificial intelligence](#).

AIOps

See [artificial intelligence operations](#).

anonymization

The process of permanently deleting personal information in a dataset. Anonymization can help protect personal privacy. Anonymized data is no longer considered to be personal data.

anti-pattern

A frequently used solution for a recurring issue where the solution is counter-productive, ineffective, or less effective than an alternative.

application control

A security approach that allows the use of only approved applications in order to help protect a system from malware.

application portfolio

A collection of detailed information about each application used by an organization, including the cost to build and maintain the application, and its business value. This information is key to [the portfolio discovery and analysis process](#) and helps identify and prioritize the applications to be migrated, modernized, and optimized.

artificial intelligence (AI)

The field of computer science that is dedicated to using computing technologies to perform cognitive functions that are typically associated with humans, such as learning, solving problems, and recognizing patterns. For more information, see [What is Artificial Intelligence?](#)

artificial intelligence operations (AIOps)

The process of using machine learning techniques to solve operational problems, reduce operational incidents and human intervention, and increase service quality. For more information about how AIOps is used in the AWS migration strategy, see the [operations integration guide](#).

asymmetric encryption

An encryption algorithm that uses a pair of keys, a public key for encryption and a private key for decryption. You can share the public key because it isn't used for decryption, but access to the private key should be highly restricted.

atomicity, consistency, isolation, durability (ACID)

A set of software properties that guarantee the data validity and operational reliability of a database, even in the case of errors, power failures, or other problems.

attribute-based access control (ABAC)

The practice of creating fine-grained permissions based on user attributes, such as department, job role, and team name. For more information, see [ABAC for AWS](#) in the AWS Identity and Access Management (IAM) documentation.

authoritative data source

A location where you store the primary version of data, which is considered to be the most reliable source of information. You can copy data from the authoritative data source to other locations for the purposes of processing or modifying the data, such as anonymizing, redacting, or pseudonymizing it.

Availability Zone

A distinct location within an AWS Region that is insulated from failures in other Availability Zones and provides inexpensive, low-latency network connectivity to other Availability Zones in the same Region.

AWS Cloud Adoption Framework (AWS CAF)

A framework of guidelines and best practices from AWS to help organizations develop an efficient and effective plan to move successfully to the cloud. AWS CAF organizes guidance into six focus areas called perspectives: business, people, governance, platform, security, and operations. The business, people, and governance perspectives focus on business skills and processes; the platform, security, and operations perspectives focus on technical skills and processes. For example, the people perspective targets stakeholders who handle human resources (HR), staffing functions, and people management. For this perspective, AWS CAF provides guidance for people development, training, and communications to help ready the organization for successful cloud adoption. For more information, see the [AWS CAF website](#) and the [AWS CAF whitepaper](#).

AWS Workload Qualification Framework (AWS WQF)

A tool that evaluates database migration workloads, recommends migration strategies, and provides work estimates. AWS WQF is included with AWS Schema Conversion Tool (AWS SCT). It analyzes database schemas and code objects, application code, dependencies, and performance characteristics, and provides assessment reports.

B

bad bot

A [bot](#) that is intended to disrupt or cause harm to individuals or organizations.

BCP

See [business continuity planning](#).

behavior graph

A unified, interactive view of resource behavior and interactions over time. You can use a behavior graph with Amazon Detective to examine failed logon attempts, suspicious API calls, and similar actions. For more information, see [Data in a behavior graph](#) in the Detective documentation.

big-endian system

A system that stores the most significant byte first. See also [endianness](#).

binary classification

A process that predicts a binary outcome (one of two possible classes). For example, your ML model might need to predict problems such as "Is this email spam or not spam?" or "Is this product a book or a car?"

bloom filter

A probabilistic, memory-efficient data structure that is used to test whether an element is a member of a set.

blue/green deployment

A deployment strategy where you create two separate but identical environments. You run the current application version in one environment (blue) and the new application version in the other environment (green). This strategy helps you quickly roll back with minimal impact.

bot

A software application that runs automated tasks over the internet and simulates human activity or interaction. Some bots are useful or beneficial, such as web crawlers that index information on the internet. Some other bots, known as *bad bots*, are intended to disrupt or cause harm to individuals or organizations.

botnet

Networks of [bots](#) that are infected by [malware](#) and are under the control of a single party, known as a *bot herder* or *bot operator*. Botnets are the best-known mechanism to scale bots and their impact.

branch

A contained area of a code repository. The first branch created in a repository is the *main branch*. You can create a new branch from an existing branch, and you can then develop features or fix bugs in the new branch. A branch you create to build a feature is commonly referred to as a *feature branch*. When the feature is ready for release, you merge the feature branch back into the main branch. For more information, see [About branches](#) (GitHub documentation).

break-glass access

In exceptional circumstances and through an approved process, a quick means for a user to gain access to an AWS account that they don't typically have permissions to access. For more information, see the [Implement break-glass procedures](#) indicator in the AWS Well-Architected guidance.

brownfield strategy

The existing infrastructure in your environment. When adopting a brownfield strategy for a system architecture, you design the architecture around the constraints of the current systems and infrastructure. If you are expanding the existing infrastructure, you might blend brownfield and [greenfield](#) strategies.

buffer cache

The memory area where the most frequently accessed data is stored.

business capability

What a business does to generate value (for example, sales, customer service, or marketing). Microservices architectures and development decisions can be driven by business capabilities. For more information, see the [Organized around business capabilities](#) section of the [Running containerized microservices on AWS](#) whitepaper.

business continuity planning (BCP)

A plan that addresses the potential impact of a disruptive event, such as a large-scale migration, on operations and enables a business to resume operations quickly.

C

CAF

See [AWS Cloud Adoption Framework](#).

canary deployment

The slow and incremental release of a version to end users. When you are confident, you deploy the new version and replace the current version in its entirety.

CCoE

See [Cloud Center of Excellence](#).

CDC

See [change data capture](#).

change data capture (CDC)

The process of tracking changes to a data source, such as a database table, and recording metadata about the change. You can use CDC for various purposes, such as auditing or replicating changes in a target system to maintain synchronization.

chaos engineering

Intentionally introducing failures or disruptive events to test a system's resilience. You can use [AWS Fault Injection Service \(AWS FIS\)](#) to perform experiments that stress your AWS workloads and evaluate their response.

CI/CD

See [continuous integration and continuous delivery](#).

classification

A categorization process that helps generate predictions. ML models for classification problems predict a discrete value. Discrete values are always distinct from one another. For example, a model might need to evaluate whether or not there is a car in an image.

client-side encryption

Encryption of data locally, before the target AWS service receives it.

Cloud Center of Excellence (CCoE)

A multi-disciplinary team that drives cloud adoption efforts across an organization, including developing cloud best practices, mobilizing resources, establishing migration timelines, and leading the organization through large-scale transformations. For more information, see the [CCoE posts](#) on the AWS Cloud Enterprise Strategy Blog.

cloud computing

The cloud technology that is typically used for remote data storage and IoT device management. Cloud computing is commonly connected to [edge computing](#) technology.

cloud operating model

In an IT organization, the operating model that is used to build, mature, and optimize one or more cloud environments. For more information, see [Building your Cloud Operating Model](#).

cloud stages of adoption

The four phases that organizations typically go through when they migrate to the AWS Cloud:

- Project – Running a few cloud-related projects for proof of concept and learning purposes
- Foundation – Making foundational investments to scale your cloud adoption (e.g., creating a landing zone, defining a CCoE, establishing an operations model)
- Migration – Migrating individual applications
- Re-invention – Optimizing products and services, and innovating in the cloud

These stages were defined by Stephen Orban in the blog post [The Journey Toward Cloud-First & the Stages of Adoption](#) on the AWS Cloud Enterprise Strategy blog. For information about how they relate to the AWS migration strategy, see the [migration readiness guide](#).

CMDB

See [configuration management database](#).

code repository

A location where source code and other assets, such as documentation, samples, and scripts, are stored and updated through version control processes. Common cloud repositories include GitHub or Bitbucket Cloud. Each version of the code is called a *branch*. In a microservice structure, each repository is devoted to a single piece of functionality. A single CI/CD pipeline can use multiple repositories.

cold cache

A buffer cache that is empty, not well populated, or contains stale or irrelevant data. This affects performance because the database instance must read from the main memory or disk, which is slower than reading from the buffer cache.

cold data

Data that is rarely accessed and is typically historical. When querying this kind of data, slow queries are typically acceptable. Moving this data to lower-performing and less expensive storage tiers or classes can reduce costs.

computer vision (CV)

A field of [AI](#) that uses machine learning to analyze and extract information from visual formats such as digital images and videos. For example, Amazon SageMaker AI provides image processing algorithms for CV.

configuration drift

For a workload, a configuration change from the expected state. It might cause the workload to become noncompliant, and it's typically gradual and unintentional.

configuration management database (CMDB)

A repository that stores and manages information about a database and its IT environment, including both hardware and software components and their configurations. You typically use data from a CMDB in the portfolio discovery and analysis stage of migration.

conformance pack

A collection of AWS Config rules and remediation actions that you can assemble to customize your compliance and security checks. You can deploy a conformance pack as a single entity in an AWS account and Region, or across an organization, by using a YAML template. For more information, see [Conformance packs](#) in the AWS Config documentation.

continuous integration and continuous delivery (CI/CD)

The process of automating the source, build, test, staging, and production stages of the software release process. CI/CD is commonly described as a pipeline. CI/CD can help you automate processes, improve productivity, improve code quality, and deliver faster. For more information, see [Benefits of continuous delivery](#). CD can also stand for *continuous deployment*. For more information, see [Continuous Delivery vs. Continuous Deployment](#).

CV

See [computer vision](#).

D

data at rest

Data that is stationary in your network, such as data that is in storage.

data classification

A process for identifying and categorizing the data in your network based on its criticality and sensitivity. It is a critical component of any cybersecurity risk management strategy because it helps you determine the appropriate protection and retention controls for the data. Data classification is a component of the security pillar in the AWS Well-Architected Framework. For more information, see [Data classification](#).

data drift

A meaningful variation between the production data and the data that was used to train an ML model, or a meaningful change in the input data over time. Data drift can reduce the overall quality, accuracy, and fairness in ML model predictions.

data in transit

Data that is actively moving through your network, such as between network resources.

data mesh

An architectural framework that provides distributed, decentralized data ownership with centralized management and governance.

data minimization

The principle of collecting and processing only the data that is strictly necessary. Practicing data minimization in the AWS Cloud can reduce privacy risks, costs, and your analytics carbon footprint.

data perimeter

A set of preventive guardrails in your AWS environment that help make sure that only trusted identities are accessing trusted resources from expected networks. For more information, see [Building a data perimeter on AWS](#).

data preprocessing

To transform raw data into a format that is easily parsed by your ML model. Preprocessing data can mean removing certain columns or rows and addressing missing, inconsistent, or duplicate values.

data provenance

The process of tracking the origin and history of data throughout its lifecycle, such as how the data was generated, transmitted, and stored.

data subject

An individual whose data is being collected and processed.

data warehouse

A data management system that supports business intelligence, such as analytics. Data warehouses commonly contain large amounts of historical data, and they are typically used for queries and analysis.

database definition language (DDL)

Statements or commands for creating or modifying the structure of tables and objects in a database.

database manipulation language (DML)

Statements or commands for modifying (inserting, updating, and deleting) information in a database.

DDL

See [database definition language](#).

deep ensemble

To combine multiple deep learning models for prediction. You can use deep ensembles to obtain a more accurate prediction or for estimating uncertainty in predictions.

deep learning

An ML subfield that uses multiple layers of artificial neural networks to identify mapping between input data and target variables of interest.

defense-in-depth

An information security approach in which a series of security mechanisms and controls are thoughtfully layered throughout a computer network to protect the confidentiality, integrity, and availability of the network and the data within. When you adopt this strategy on AWS, you add multiple controls at different layers of the AWS Organizations structure to help secure resources. For example, a defense-in-depth approach might combine multi-factor authentication, network segmentation, and encryption.

delegated administrator

In AWS Organizations, a compatible service can register an AWS member account to administer the organization's accounts and manage permissions for that service. This account is called the *delegated administrator* for that service. For more information and a list of compatible services, see [Services that work with AWS Organizations](#) in the AWS Organizations documentation.

deployment

The process of making an application, new features, or code fixes available in the target environment. Deployment involves implementing changes in a code base and then building and running that code base in the application's environments.

development environment

See [environment](#).

detective control

A security control that is designed to detect, log, and alert after an event has occurred. These controls are a second line of defense, alerting you to security events that bypassed the preventative controls in place. For more information, see [Detective controls](#) in *Implementing security controls on AWS*.

development value stream mapping (DVSM)

A process used to identify and prioritize constraints that adversely affect speed and quality in a software development lifecycle. DVSM extends the value stream mapping process originally designed for lean manufacturing practices. It focuses on the steps and teams required to create and move value through the software development process.

digital twin

A virtual representation of a real-world system, such as a building, factory, industrial equipment, or production line. Digital twins support predictive maintenance, remote monitoring, and production optimization.

dimension table

In a [star schema](#), a smaller table that contains data attributes about quantitative data in a fact table. Dimension table attributes are typically text fields or discrete numbers that behave like text. These attributes are commonly used for query constraining, filtering, and result set labeling.

disaster

An event that prevents a workload or system from fulfilling its business objectives in its primary deployed location. These events can be natural disasters, technical failures, or the result of human actions, such as unintentional misconfiguration or a malware attack.

disaster recovery (DR)

The strategy and process you use to minimize downtime and data loss caused by a [disaster](#). For more information, see [Disaster Recovery of Workloads on AWS: Recovery in the Cloud](#) in the AWS Well-Architected Framework.

DML

See [database manipulation language](#).

domain-driven design

An approach to developing a complex software system by connecting its components to evolving domains, or core business goals, that each component serves. This concept was introduced by Eric Evans in his book, *Domain-Driven Design: Tackling Complexity in the Heart of Software* (Boston: Addison-Wesley Professional, 2003). For information about how you can use domain-driven design with the strangler fig pattern, see [Modernizing legacy Microsoft ASP.NET \(ASMX\) web services incrementally by using containers and Amazon API Gateway](#).

DR

See [disaster recovery](#).

drift detection

Tracking deviations from a baselined configuration. For example, you can use AWS CloudFormation to [detect drift in system resources](#), or you can use AWS Control Tower to [detect changes in your landing zone](#) that might affect compliance with governance requirements.

DVSM

See [development value stream mapping](#).

E

EDA

See [exploratory data analysis](#).

EDI

See [electronic data interchange](#).

edge computing

The technology that increases the computing power for smart devices at the edges of an IoT network. When compared with [cloud computing](#), edge computing can reduce communication latency and improve response time.

electronic data interchange (EDI)

The automated exchange of business documents between organizations. For more information, see [What is Electronic Data Interchange](#).

encryption

A computing process that transforms plaintext data, which is human-readable, into ciphertext.

encryption key

A cryptographic string of randomized bits that is generated by an encryption algorithm. Keys can vary in length, and each key is designed to be unpredictable and unique.

endianness

The order in which bytes are stored in computer memory. Big-endian systems store the most significant byte first. Little-endian systems store the least significant byte first.

endpoint

See [service endpoint](#).

endpoint service

A service that you can host in a virtual private cloud (VPC) to share with other users. You can create an endpoint service with AWS PrivateLink and grant permissions to other AWS accounts or to AWS Identity and Access Management (IAM) principals. These accounts or principals can connect to your endpoint service privately by creating interface VPC endpoints. For more

information, see [Create an endpoint service](#) in the Amazon Virtual Private Cloud (Amazon VPC) documentation.

enterprise resource planning (ERP)

A system that automates and manages key business processes (such as accounting, [MES](#), and project management) for an enterprise.

envelope encryption

The process of encrypting an encryption key with another encryption key. For more information, see [Envelope encryption](#) in the AWS Key Management Service (AWS KMS) documentation.

environment

An instance of a running application. The following are common types of environments in cloud computing:

- development environment – An instance of a running application that is available only to the core team responsible for maintaining the application. Development environments are used to test changes before promoting them to upper environments. This type of environment is sometimes referred to as a *test environment*.
- lower environments – All development environments for an application, such as those used for initial builds and tests.
- production environment – An instance of a running application that end users can access. In a CI/CD pipeline, the production environment is the last deployment environment.
- upper environments – All environments that can be accessed by users other than the core development team. This can include a production environment, preproduction environments, and environments for user acceptance testing.

epic

In agile methodologies, functional categories that help organize and prioritize your work. Epics provide a high-level description of requirements and implementation tasks. For example, AWS CAF security epics include identity and access management, detective controls, infrastructure security, data protection, and incident response. For more information about epics in the AWS migration strategy, see the [program implementation guide](#).

ERP

See [enterprise resource planning](#).

exploratory data analysis (EDA)

The process of analyzing a dataset to understand its main characteristics. You collect or aggregate data and then perform initial investigations to find patterns, detect anomalies, and check assumptions. EDA is performed by calculating summary statistics and creating data visualizations.

F

fact table

The central table in a [star schema](#). It stores quantitative data about business operations. Typically, a fact table contains two types of columns: those that contain measures and those that contain a foreign key to a dimension table.

fail fast

A philosophy that uses frequent and incremental testing to reduce the development lifecycle. It is a critical part of an agile approach.

fault isolation boundary

In the AWS Cloud, a boundary such as an Availability Zone, AWS Region, control plane, or data plane that limits the effect of a failure and helps improve the resilience of workloads. For more information, see [AWS Fault Isolation Boundaries](#).

feature branch

See [branch](#).

features

The input data that you use to make a prediction. For example, in a manufacturing context, features could be images that are periodically captured from the manufacturing line.

feature importance

How significant a feature is for a model's predictions. This is usually expressed as a numerical score that can be calculated through various techniques, such as Shapley Additive Explanations (SHAP) and integrated gradients. For more information, see [Machine learning model interpretability with AWS](#).

feature transformation

To optimize data for the ML process, including enriching data with additional sources, scaling values, or extracting multiple sets of information from a single data field. This enables the ML model to benefit from the data. For example, if you break down the "2021-05-27 00:15:37" date into "2021", "May", "Thu", and "15", you can help the learning algorithm learn nuanced patterns associated with different data components.

few-shot prompting

Providing an [LLM](#) with a small number of examples that demonstrate the task and desired output before asking it to perform a similar task. This technique is an application of in-context learning, where models learn from examples (*shots*) that are embedded in prompts. Few-shot prompting can be effective for tasks that require specific formatting, reasoning, or domain knowledge. See also [zero-shot prompting](#).

FGAC

See [fine-grained access control](#).

fine-grained access control (FGAC)

The use of multiple conditions to allow or deny an access request.

flash-cut migration

A database migration method that uses continuous data replication through [change data capture](#) to migrate data in the shortest time possible, instead of using a phased approach. The objective is to keep downtime to a minimum.

FM

See [foundation model](#).

foundation model (FM)

A large deep-learning neural network that has been training on massive datasets of generalized and unlabeled data. FMs are capable of performing a wide variety of general tasks, such as understanding language, generating text and images, and conversing in natural language. For more information, see [What are Foundation Models](#).

G

generative AI

A subset of [AI](#) models that have been trained on large amounts of data and that can use a simple text prompt to create new content and artifacts, such as images, videos, text, and audio. For more information, see [What is Generative AI](#).

geo blocking

See [geographic restrictions](#).

geographic restrictions (geo blocking)

In Amazon CloudFront, an option to prevent users in specific countries from accessing content distributions. You can use an allow list or block list to specify approved and banned countries. For more information, see [Restricting the geographic distribution of your content](#) in the CloudFront documentation.

Gitflow workflow

An approach in which lower and upper environments use different branches in a source code repository. The Gitflow workflow is considered legacy, and the [trunk-based workflow](#) is the modern, preferred approach.

golden image

A snapshot of a system or software that is used as a template to deploy new instances of that system or software. For example, in manufacturing, a golden image can be used to provision software on multiple devices and helps improve speed, scalability, and productivity in device manufacturing operations.

greenfield strategy

The absence of existing infrastructure in a new environment. When adopting a greenfield strategy for a system architecture, you can select all new technologies without the restriction of compatibility with existing infrastructure, also known as [brownfield](#). If you are expanding the existing infrastructure, you might blend brownfield and greenfield strategies.

guardrail

A high-level rule that helps govern resources, policies, and compliance across organizational units (OUs). *Preventive guardrails* enforce policies to ensure alignment to compliance standards. They are implemented by using service control policies and IAM permissions boundaries.

Detective guardrails detect policy violations and compliance issues, and generate alerts for remediation. They are implemented by using AWS Config, AWS Security Hub, Amazon GuardDuty, AWS Trusted Advisor, Amazon Inspector, and custom AWS Lambda checks.

H

HA

See [high availability](#).

heterogeneous database migration

Migrating your source database to a target database that uses a different database engine (for example, Oracle to Amazon Aurora). Heterogeneous migration is typically part of a re-architecting effort, and converting the schema can be a complex task. [AWS provides AWS SCT](#) that helps with schema conversions.

high availability (HA)

The ability of a workload to operate continuously, without intervention, in the event of challenges or disasters. HA systems are designed to automatically fail over, consistently deliver high-quality performance, and handle different loads and failures with minimal performance impact.

historian modernization

An approach used to modernize and upgrade operational technology (OT) systems to better serve the needs of the manufacturing industry. A *historian* is a type of database that is used to collect and store data from various sources in a factory.

holdout data

A portion of historical, labeled data that is withheld from a dataset that is used to train a [machine learning](#) model. You can use holdout data to evaluate the model performance by comparing the model predictions against the holdout data.

homogeneous database migration

Migrating your source database to a target database that shares the same database engine (for example, Microsoft SQL Server to Amazon RDS for SQL Server). Homogeneous migration is typically part of a rehosting or replatforming effort. You can use native database utilities to migrate the schema.

hot data

Data that is frequently accessed, such as real-time data or recent translational data. This data typically requires a high-performance storage tier or class to provide fast query responses.

hotfix

An urgent fix for a critical issue in a production environment. Due to its urgency, a hotfix is usually made outside of the typical DevOps release workflow.

hypercare period

Immediately following cutover, the period of time when a migration team manages and monitors the migrated applications in the cloud in order to address any issues. Typically, this period is 1–4 days in length. At the end of the hypercare period, the migration team typically transfers responsibility for the applications to the cloud operations team.

I

laC

See [infrastructure as code](#).

identity-based policy

A policy attached to one or more IAM principals that defines their permissions within the AWS Cloud environment.

idle application

An application that has an average CPU and memory usage between 5 and 20 percent over a period of 90 days. In a migration project, it is common to retire these applications or retain them on premises.

IIoT

See [Industrial Internet of Things](#).

immutable infrastructure

A model that deploys new infrastructure for production workloads instead of updating, patching, or modifying the existing infrastructure. Immutable infrastructures are inherently more consistent, reliable, and predictable than [mutable infrastructure](#). For more information, see the [Deploy using immutable infrastructure](#) best practice in the AWS Well-Architected Framework.

inbound (ingress) VPC

In an AWS multi-account architecture, a VPC that accepts, inspects, and routes network connections from outside an application. The [AWS Security Reference Architecture](#) recommends setting up your Network account with inbound, outbound, and inspection VPCs to protect the two-way interface between your application and the broader internet.

incremental migration

A cutover strategy in which you migrate your application in small parts instead of performing a single, full cutover. For example, you might move only a few microservices or users to the new system initially. After you verify that everything is working properly, you can incrementally move additional microservices or users until you can decommission your legacy system. This strategy reduces the risks associated with large migrations.

Industry 4.0

A term that was introduced by [Klaus Schwab](#) in 2016 to refer to the modernization of manufacturing processes through advances in connectivity, real-time data, automation, analytics, and AI/ML.

infrastructure

All of the resources and assets contained within an application's environment.

infrastructure as code (IaC)

The process of provisioning and managing an application's infrastructure through a set of configuration files. IaC is designed to help you centralize infrastructure management, standardize resources, and scale quickly so that new environments are repeatable, reliable, and consistent.

industrial Internet of Things (IIoT)

The use of internet-connected sensors and devices in the industrial sectors, such as manufacturing, energy, automotive, healthcare, life sciences, and agriculture. For more information, see [Building an industrial Internet of Things \(IIoT\) digital transformation strategy](#).

inspection VPC

In an AWS multi-account architecture, a centralized VPC that manages inspections of network traffic between VPCs (in the same or different AWS Regions), the internet, and on-premises networks. The [AWS Security Reference Architecture](#) recommends setting up your Network account with inbound, outbound, and inspection VPCs to protect the two-way interface between your application and the broader internet.

Internet of Things (IoT)

The network of connected physical objects with embedded sensors or processors that communicate with other devices and systems through the internet or over a local communication network. For more information, see [What is IoT?](#)

interpretability

A characteristic of a machine learning model that describes the degree to which a human can understand how the model's predictions depend on its inputs. For more information, see [Machine learning model interpretability with AWS](#).

IoT

See [Internet of Things](#).

IT information library (ITIL)

A set of best practices for delivering IT services and aligning these services with business requirements. ITIL provides the foundation for ITSM.

IT service management (ITSM)

Activities associated with designing, implementing, managing, and supporting IT services for an organization. For information about integrating cloud operations with ITSM tools, see the [operations integration guide](#).

ITIL

See [IT information library](#).

ITSM

See [IT service management](#).

L

label-based access control (LBAC)

An implementation of mandatory access control (MAC) where the users and the data itself are each explicitly assigned a security label value. The intersection between the user security label and data security label determines which rows and columns can be seen by the user.

landing zone

A landing zone is a well-architected, multi-account AWS environment that is scalable and secure. This is a starting point from which your organizations can quickly launch and deploy workloads and applications with confidence in their security and infrastructure environment. For more information about landing zones, see [Setting up a secure and scalable multi-account AWS environment](#).

large language model (LLM)

A deep learning [AI](#) model that is pretrained on a vast amount of data. An LLM can perform multiple tasks, such as answering questions, summarizing documents, translating text into other languages, and completing sentences. For more information, see [What are LLMs](#).

large migration

A migration of 300 or more servers.

LBAC

See [label-based access control](#).

least privilege

The security best practice of granting the minimum permissions required to perform a task. For more information, see [Apply least-privilege permissions](#) in the IAM documentation.

lift and shift

See [7 Rs](#).

little-endian system

A system that stores the least significant byte first. See also [endianness](#).

LLM

See [large language model](#).

lower environments

See [environment](#).

M

machine learning (ML)

A type of artificial intelligence that uses algorithms and techniques for pattern recognition and learning. ML analyzes and learns from recorded data, such as Internet of Things (IoT) data, to generate a statistical model based on patterns. For more information, see [Machine Learning](#).

main branch

See [branch](#).

malware

Software that is designed to compromise computer security or privacy. Malware might disrupt computer systems, leak sensitive information, or gain unauthorized access. Examples of malware include viruses, worms, ransomware, Trojan horses, spyware, and keyloggers.

managed services

AWS services for which AWS operates the infrastructure layer, the operating system, and platforms, and you access the endpoints to store and retrieve data. Amazon Simple Storage Service (Amazon S3) and Amazon DynamoDB are examples of managed services. These are also known as *abstracted services*.

manufacturing execution system (MES)

A software system for tracking, monitoring, documenting, and controlling production processes that convert raw materials to finished products on the shop floor.

MAP

See [Migration Acceleration Program](#).

mechanism

A complete process in which you create a tool, drive adoption of the tool, and then inspect the results in order to make adjustments. A mechanism is a cycle that reinforces and improves itself as it operates. For more information, see [Building mechanisms](#) in the AWS Well-Architected Framework.

member account

All AWS accounts other than the management account that are part of an organization in AWS Organizations. An account can be a member of only one organization at a time.

MES

See [manufacturing execution system](#).

Message Queuing Telemetry Transport (MQTT)

A lightweight, machine-to-machine (M2M) communication protocol, based on the [publish/subscribe](#) pattern, for resource-constrained [IoT](#) devices.

microservice

A small, independent service that communicates over well-defined APIs and is typically owned by small, self-contained teams. For example, an insurance system might include microservices that map to business capabilities, such as sales or marketing, or subdomains, such as purchasing, claims, or analytics. The benefits of microservices include agility, flexible scaling, easy deployment, reusable code, and resilience. For more information, see [Integrating microservices by using AWS serverless services](#).

microservices architecture

An approach to building an application with independent components that run each application process as a microservice. These microservices communicate through a well-defined interface by using lightweight APIs. Each microservice in this architecture can be updated, deployed, and scaled to meet demand for specific functions of an application. For more information, see [Implementing microservices on AWS](#).

Migration Acceleration Program (MAP)

An AWS program that provides consulting support, training, and services to help organizations build a strong operational foundation for moving to the cloud, and to help offset the initial cost of migrations. MAP includes a migration methodology for executing legacy migrations in a methodical way and a set of tools to automate and accelerate common migration scenarios.

migration at scale

The process of moving the majority of the application portfolio to the cloud in waves, with more applications moved at a faster rate in each wave. This phase uses the best practices and lessons learned from the earlier phases to implement a *migration factory* of teams, tools, and processes to streamline the migration of workloads through automation and agile delivery. This is the third phase of the [AWS migration strategy](#).

migration factory

Cross-functional teams that streamline the migration of workloads through automated, agile approaches. Migration factory teams typically include operations, business analysts and owners,

migration engineers, developers, and DevOps professionals working in sprints. Between 20 and 50 percent of an enterprise application portfolio consists of repeated patterns that can be optimized by a factory approach. For more information, see the [discussion of migration factories](#) and the [Cloud Migration Factory guide](#) in this content set.

migration metadata

The information about the application and server that is needed to complete the migration. Each migration pattern requires a different set of migration metadata. Examples of migration metadata include the target subnet, security group, and AWS account.

migration pattern

A repeatable migration task that details the migration strategy, the migration destination, and the migration application or service used. Example: Rehost migration to Amazon EC2 with AWS Application Migration Service.

Migration Portfolio Assessment (MPA)

An online tool that provides information for validating the business case for migrating to the AWS Cloud. MPA provides detailed portfolio assessment (server right-sizing, pricing, TCO comparisons, migration cost analysis) as well as migration planning (application data analysis and data collection, application grouping, migration prioritization, and wave planning). The [MPA tool](#) (requires login) is available free of charge to all AWS consultants and APN Partner consultants.

Migration Readiness Assessment (MRA)

The process of gaining insights about an organization's cloud readiness status, identifying strengths and weaknesses, and building an action plan to close identified gaps, using the AWS CAF. For more information, see the [migration readiness guide](#). MRA is the first phase of the [AWS migration strategy](#).

migration strategy

The approach used to migrate a workload to the AWS Cloud. For more information, see the [7 Rs](#) entry in this glossary and see [Mobilize your organization to accelerate large-scale migrations](#).

ML

See [machine learning](#).

modernization

Transforming an outdated (legacy or monolithic) application and its infrastructure into an agile, elastic, and highly available system in the cloud to reduce costs, gain efficiencies, and take advantage of innovations. For more information, see [Strategy for modernizing applications in the AWS Cloud](#).

modernization readiness assessment

An evaluation that helps determine the modernization readiness of an organization's applications; identifies benefits, risks, and dependencies; and determines how well the organization can support the future state of those applications. The outcome of the assessment is a blueprint of the target architecture, a roadmap that details development phases and milestones for the modernization process, and an action plan for addressing identified gaps. For more information, see [Evaluating modernization readiness for applications in the AWS Cloud](#).

monolithic applications (monoliths)

Applications that run as a single service with tightly coupled processes. Monolithic applications have several drawbacks. If one application feature experiences a spike in demand, the entire architecture must be scaled. Adding or improving a monolithic application's features also becomes more complex when the code base grows. To address these issues, you can use a microservices architecture. For more information, see [Decomposing monoliths into microservices](#).

MPA

See [Migration Portfolio Assessment](#).

MQTT

See [Message Queuing Telemetry Transport](#).

multiclass classification

A process that helps generate predictions for multiple classes (predicting one of more than two outcomes). For example, an ML model might ask "Is this product a book, car, or phone?" or "Which product category is most interesting to this customer?"

mutable infrastructure

A model that updates and modifies the existing infrastructure for production workloads. For improved consistency, reliability, and predictability, the AWS Well-Architected Framework recommends the use of [immutable infrastructure](#) as a best practice.

O

OAC

See [origin access control](#).

OAI

See [origin access identity](#).

OCM

See [organizational change management](#).

offline migration

A migration method in which the source workload is taken down during the migration process. This method involves extended downtime and is typically used for small, non-critical workloads.

OI

See [operations integration](#).

OLA

See [operational-level agreement](#).

online migration

A migration method in which the source workload is copied to the target system without being taken offline. Applications that are connected to the workload can continue to function during the migration. This method involves zero to minimal downtime and is typically used for critical production workloads.

OPC-UA

See [Open Process Communications - Unified Architecture](#).

Open Process Communications - Unified Architecture (OPC-UA)

A machine-to-machine (M2M) communication protocol for industrial automation. OPC-UA provides an interoperability standard with data encryption, authentication, and authorization schemes.

operational-level agreement (OLA)

An agreement that clarifies what functional IT groups promise to deliver to each other, to support a service-level agreement (SLA).

operational readiness review (ORR)

A checklist of questions and associated best practices that help you understand, evaluate, prevent, or reduce the scope of incidents and possible failures. For more information, see [Operational Readiness Reviews \(ORR\)](#) in the AWS Well-Architected Framework.

operational technology (OT)

Hardware and software systems that work with the physical environment to control industrial operations, equipment, and infrastructure. In manufacturing, the integration of OT and information technology (IT) systems is a key focus for [Industry 4.0](#) transformations.

operations integration (OI)

The process of modernizing operations in the cloud, which involves readiness planning, automation, and integration. For more information, see the [operations integration guide](#).

organization trail

A trail that's created by AWS CloudTrail that logs all events for all AWS accounts in an organization in AWS Organizations. This trail is created in each AWS account that's part of the organization and tracks the activity in each account. For more information, see [Creating a trail for an organization](#) in the CloudTrail documentation.

organizational change management (OCM)

A framework for managing major, disruptive business transformations from a people, culture, and leadership perspective. OCM helps organizations prepare for, and transition to, new systems and strategies by accelerating change adoption, addressing transitional issues, and driving cultural and organizational changes. In the AWS migration strategy, this framework is called *people acceleration*, because of the speed of change required in cloud adoption projects. For more information, see the [OCM guide](#).

origin access control (OAC)

In CloudFront, an enhanced option for restricting access to secure your Amazon Simple Storage Service (Amazon S3) content. OAC supports all S3 buckets in all AWS Regions, server-side encryption with AWS KMS (SSE-KMS), and dynamic PUT and DELETE requests to the S3 bucket.

origin access identity (OAI)

In CloudFront, an option for restricting access to secure your Amazon S3 content. When you use OAI, CloudFront creates a principal that Amazon S3 can authenticate with. Authenticated principals can access content in an S3 bucket only through a specific CloudFront distribution. See also [OAC](#), which provides more granular and enhanced access control.

ORR

See [operational readiness review](#).

OT

See [operational technology](#).

outbound (egress) VPC

In an AWS multi-account architecture, a VPC that handles network connections that are initiated from within an application. The [AWS Security Reference Architecture](#) recommends setting up your Network account with inbound, outbound, and inspection VPCs to protect the two-way interface between your application and the broader internet.

P

permissions boundary

An IAM management policy that is attached to IAM principals to set the maximum permissions that the user or role can have. For more information, see [Permissions boundaries](#) in the IAM documentation.

personally identifiable information (PII)

Information that, when viewed directly or paired with other related data, can be used to reasonably infer the identity of an individual. Examples of PII include names, addresses, and contact information.

PII

See [personally identifiable information](#).

playbook

A set of predefined steps that capture the work associated with migrations, such as delivering core operations functions in the cloud. A playbook can take the form of scripts, automated runbooks, or a summary of processes or steps required to operate your modernized environment.

PLC

See [programmable logic controller](#).

PLM

See [product lifecycle management](#).

policy

An object that can define permissions (see [identity-based policy](#)), specify access conditions (see [resource-based policy](#)), or define the maximum permissions for all accounts in an organization in AWS Organizations (see [service control policy](#)).

polyglot persistence

Independently choosing a microservice's data storage technology based on data access patterns and other requirements. If your microservices have the same data storage technology, they can encounter implementation challenges or experience poor performance. Microservices are more easily implemented and achieve better performance and scalability if they use the data store best adapted to their requirements. For more information, see [Enabling data persistence in microservices](#).

portfolio assessment

A process of discovering, analyzing, and prioritizing the application portfolio in order to plan the migration. For more information, see [Evaluating migration readiness](#).

predicate

A query condition that returns true or false, commonly located in a WHERE clause.

predicate pushdown

A database query optimization technique that filters the data in the query before transfer. This reduces the amount of data that must be retrieved and processed from the relational database, and it improves query performance.

preventative control

A security control that is designed to prevent an event from occurring. These controls are a first line of defense to help prevent unauthorized access or unwanted changes to your network. For more information, see [Preventative controls](#) in *Implementing security controls on AWS*.

principal

An entity in AWS that can perform actions and access resources. This entity is typically a root user for an AWS account, an IAM role, or a user. For more information, see *Principal* in [Roles terms and concepts](#) in the IAM documentation.

privacy by design

A system engineering approach that takes privacy into account through the whole development process.

private hosted zones

A container that holds information about how you want Amazon Route 53 to respond to DNS queries for a domain and its subdomains within one or more VPCs. For more information, see [Working with private hosted zones](#) in the Route 53 documentation.

proactive control

A [security control](#) designed to prevent the deployment of noncompliant resources. These controls scan resources before they are provisioned. If the resource is not compliant with the control, then it isn't provisioned. For more information, see the [Controls reference guide](#) in the AWS Control Tower documentation and see [Proactive controls](#) in *Implementing security controls on AWS*.

product lifecycle management (PLM)

The management of data and processes for a product throughout its entire lifecycle, from design, development, and launch, through growth and maturity, to decline and removal.

production environment

See [environment](#).

programmable logic controller (PLC)

In manufacturing, a highly reliable, adaptable computer that monitors machines and automates manufacturing processes.

prompt chaining

Using the output of one [LLM](#) prompt as the input for the next prompt to generate better responses. This technique is used to break down a complex task into subtasks, or to iteratively refine or expand a preliminary response. It helps improve the accuracy and relevance of a model's responses and allows for more granular, personalized results.

pseudonymization

The process of replacing personal identifiers in a dataset with placeholder values. Pseudonymization can help protect personal privacy. Pseudonymized data is still considered to be personal data.

publish/subscribe (pub/sub)

A pattern that enables asynchronous communications among microservices to improve scalability and responsiveness. For example, in a microservices-based [MES](#), a microservice can publish event messages to a channel that other microservices can subscribe to. The system can add new microservices without changing the publishing service.

Q

query plan

A series of steps, like instructions, that are used to access the data in a SQL relational database system.

query plan regression

When a database service optimizer chooses a less optimal plan than it did before a given change to the database environment. This can be caused by changes to statistics, constraints, environment settings, query parameter bindings, and updates to the database engine.

R

RACI matrix

See [responsible, accountable, consulted, informed \(RACI\)](#).

RAG

See [Retrieval Augmented Generation](#).

ransomware

A malicious software that is designed to block access to a computer system or data until a payment is made.

RASCI matrix

See [responsible, accountable, consulted, informed \(RACI\)](#).

RCAC

See [row and column access control](#).

read replica

A copy of a database that's used for read-only purposes. You can route queries to the read replica to reduce the load on your primary database.

re-architect

See [7 Rs](#).

recovery point objective (RPO)

The maximum acceptable amount of time since the last data recovery point. This determines what is considered an acceptable loss of data between the last recovery point and the interruption of service.

recovery time objective (RTO)

The maximum acceptable delay between the interruption of service and restoration of service.

refactor

See [7 Rs](#).

Region

A collection of AWS resources in a geographic area. Each AWS Region is isolated and independent of the others to provide fault tolerance, stability, and resilience. For more information, see [Specify which AWS Regions your account can use](#).

regression

An ML technique that predicts a numeric value. For example, to solve the problem of "What price will this house sell for?" an ML model could use a linear regression model to predict a house's sale price based on known facts about the house (for example, the square footage).

rehost

See [7 Rs](#).

release

In a deployment process, the act of promoting changes to a production environment.

relocate

See [7 Rs](#).

replatform

See [7 Rs](#).

repurchase

See [7 Rs](#).

resiliency

An application's ability to resist or recover from disruptions. [High availability](#) and [disaster recovery](#) are common considerations when planning for resiliency in the AWS Cloud. For more information, see [AWS Cloud Resilience](#).

resource-based policy

A policy attached to a resource, such as an Amazon S3 bucket, an endpoint, or an encryption key. This type of policy specifies which principals are allowed access, supported actions, and any other conditions that must be met.

responsible, accountable, consulted, informed (RACI) matrix

A matrix that defines the roles and responsibilities for all parties involved in migration activities and cloud operations. The matrix name is derived from the responsibility types defined in the matrix: responsible (R), accountable (A), consulted (C), and informed (I). The support (S) type is optional. If you include support, the matrix is called a *RASCI matrix*, and if you exclude it, it's called a *RACI matrix*.

responsive control

A security control that is designed to drive remediation of adverse events or deviations from your security baseline. For more information, see [Responsive controls](#) in *Implementing security controls on AWS*.

retain

See [7 Rs](#).

retire

See [7 Rs](#).

Retrieval Augmented Generation (RAG)

A [generative AI](#) technology in which an [LLM](#) references an authoritative data source that is outside of its training data sources before generating a response. For example, a RAG model might perform a semantic search of an organization's knowledge base or custom data. For more information, see [What is RAG](#).

rotation

The process of periodically updating a [secret](#) to make it more difficult for an attacker to access the credentials.

row and column access control (RCAC)

The use of basic, flexible SQL expressions that have defined access rules. RCAC consists of row permissions and column masks.

RPO

See [recovery point objective](#).

RTO

See [recovery time objective](#).

runbook

A set of manual or automated procedures required to perform a specific task. These are typically built to streamline repetitive operations or procedures with high error rates.

S

SAML 2.0

An open standard that many identity providers (IdPs) use. This feature enables federated single sign-on (SSO), so users can log into the AWS Management Console or call the AWS API operations without you having to create user in IAM for everyone in your organization. For more information about SAML 2.0-based federation, see [About SAML 2.0-based federation](#) in the IAM documentation.

SCADA

See [supervisory control and data acquisition](#).

SCP

See [service control policy](#).

secret

In AWS Secrets Manager, confidential or restricted information, such as a password or user credentials, that you store in encrypted form. It consists of the secret value and its metadata.

The secret value can be binary, a single string, or multiple strings. For more information, see [What's in a Secrets Manager secret?](#) in the Secrets Manager documentation.

security by design

A system engineering approach that takes security into account through the whole development process.

security control

A technical or administrative guardrail that prevents, detects, or reduces the ability of a threat actor to exploit a security vulnerability. There are four primary types of security controls: [preventative](#), [detective](#), [responsive](#), and [proactive](#).

security hardening

The process of reducing the attack surface to make it more resistant to attacks. This can include actions such as removing resources that are no longer needed, implementing the security best practice of granting least privilege, or deactivating unnecessary features in configuration files.

security information and event management (SIEM) system

Tools and services that combine security information management (SIM) and security event management (SEM) systems. A SIEM system collects, monitors, and analyzes data from servers, networks, devices, and other sources to detect threats and security breaches, and to generate alerts.

security response automation

A predefined and programmed action that is designed to automatically respond to or remediate a security event. These automations serve as [detective](#) or [responsive](#) security controls that help you implement AWS security best practices. Examples of automated response actions include modifying a VPC security group, patching an Amazon EC2 instance, or rotating credentials.

server-side encryption

Encryption of data at its destination, by the AWS service that receives it.

service control policy (SCP)

A policy that provides centralized control over permissions for all accounts in an organization in AWS Organizations. SCPs define guardrails or set limits on actions that an administrator can delegate to users or roles. You can use SCPs as allow lists or deny lists, to specify which services or actions are permitted or prohibited. For more information, see [Service control policies](#) in the AWS Organizations documentation.

service endpoint

The URL of the entry point for an AWS service. You can use the endpoint to connect programmatically to the target service. For more information, see [AWS service endpoints](#) in *AWS General Reference*.

service-level agreement (SLA)

An agreement that clarifies what an IT team promises to deliver to their customers, such as service uptime and performance.

service-level indicator (SLI)

A measurement of a performance aspect of a service, such as its error rate, availability, or throughput.

service-level objective (SLO)

A target metric that represents the health of a service, as measured by a [service-level indicator](#).

shared responsibility model

A model describing the responsibility you share with AWS for cloud security and compliance. AWS is responsible for security *of* the cloud, whereas you are responsible for security *in* the cloud. For more information, see [Shared responsibility model](#).

SIEM

See [security information and event management system](#).

single point of failure (SPOF)

A failure in a single, critical component of an application that can disrupt the system.

SLA

See [service-level agreement](#).

SLI

See [service-level indicator](#).

SLO

See [service-level objective](#).

split-and-seed model

A pattern for scaling and accelerating modernization projects. As new features and product releases are defined, the core team splits up to create new product teams. This helps scale your

organization's capabilities and services, improves developer productivity, and supports rapid innovation. For more information, see [Phased approach to modernizing applications in the AWS Cloud](#).

SPOF

See [single point of failure](#).

star schema

A database organizational structure that uses one large fact table to store transactional or measured data and uses one or more smaller dimensional tables to store data attributes. This structure is designed for use in a [data warehouse](#) or for business intelligence purposes.

strangler fig pattern

An approach to modernizing monolithic systems by incrementally rewriting and replacing system functionality until the legacy system can be decommissioned. This pattern uses the analogy of a fig vine that grows into an established tree and eventually overcomes and replaces its host. The pattern was [introduced by Martin Fowler](#) as a way to manage risk when rewriting monolithic systems. For an example of how to apply this pattern, see [Modernizing legacy Microsoft ASP.NET \(ASMX\) web services incrementally by using containers and Amazon API Gateway](#).

subnet

A range of IP addresses in your VPC. A subnet must reside in a single Availability Zone.

supervisory control and data acquisition (SCADA)

In manufacturing, a system that uses hardware and software to monitor physical assets and production operations.

symmetric encryption

An encryption algorithm that uses the same key to encrypt and decrypt the data.

synthetic testing

Testing a system in a way that simulates user interactions to detect potential issues or to monitor performance. You can use [Amazon CloudWatch Synthetics](#) to create these tests.

system prompt

A technique for providing context, instructions, or guidelines to an [LLM](#) to direct its behavior. System prompts help set context and establish rules for interactions with users.

T

tags

Key-value pairs that act as metadata for organizing your AWS resources. Tags can help you manage, identify, organize, search for, and filter resources. For more information, see [Tagging your AWS resources](#).

target variable

The value that you are trying to predict in supervised ML. This is also referred to as an *outcome variable*. For example, in a manufacturing setting the target variable could be a product defect.

task list

A tool that is used to track progress through a runbook. A task list contains an overview of the runbook and a list of general tasks to be completed. For each general task, it includes the estimated amount of time required, the owner, and the progress.

test environment

See [environment](#).

training

To provide data for your ML model to learn from. The training data must contain the correct answer. The learning algorithm finds patterns in the training data that map the input data attributes to the target (the answer that you want to predict). It outputs an ML model that captures these patterns. You can then use the ML model to make predictions on new data for which you don't know the target.

transit gateway

A network transit hub that you can use to interconnect your VPCs and on-premises networks. For more information, see [What is a transit gateway](#) in the AWS Transit Gateway documentation.

trunk-based workflow

An approach in which developers build and test features locally in a feature branch and then merge those changes into the main branch. The main branch is then built to the development, preproduction, and production environments, sequentially.

trusted access

Granting permissions to a service that you specify to perform tasks in your organization in AWS Organizations and in its accounts on your behalf. The trusted service creates a service-linked role in each account, when that role is needed, to perform management tasks for you. For more information, see [Using AWS Organizations with other AWS services](#) in the AWS Organizations documentation.

tuning

To change aspects of your training process to improve the ML model's accuracy. For example, you can train the ML model by generating a labeling set, adding labels, and then repeating these steps several times under different settings to optimize the model.

two-pizza team

A small DevOps team that you can feed with two pizzas. A two-pizza team size ensures the best possible opportunity for collaboration in software development.

U

uncertainty

A concept that refers to imprecise, incomplete, or unknown information that can undermine the reliability of predictive ML models. There are two types of uncertainty: *Epistemic uncertainty* is caused by limited, incomplete data, whereas *aleatoric uncertainty* is caused by the noise and randomness inherent in the data. For more information, see the [Quantifying uncertainty in deep learning systems](#) guide.

undifferentiated tasks

Also known as *heavy lifting*, work that is necessary to create and operate an application but that doesn't provide direct value to the end user or provide competitive advantage. Examples of undifferentiated tasks include procurement, maintenance, and capacity planning.

upper environments

See [environment](#).

V

vacuuming

A database maintenance operation that involves cleaning up after incremental updates to reclaim storage and improve performance.

version control

Processes and tools that track changes, such as changes to source code in a repository.

VPC peering

A connection between two VPCs that allows you to route traffic by using private IP addresses. For more information, see [What is VPC peering](#) in the Amazon VPC documentation.

vulnerability

A software or hardware flaw that compromises the security of the system.

W

warm cache

A buffer cache that contains current, relevant data that is frequently accessed. The database instance can read from the buffer cache, which is faster than reading from the main memory or disk.

warm data

Data that is infrequently accessed. When querying this kind of data, moderately slow queries are typically acceptable.

window function

A SQL function that performs a calculation on a group of rows that relate in some way to the current record. Window functions are useful for processing tasks, such as calculating a moving average or accessing the value of rows based on the relative position of the current row.

workload

A collection of resources and code that delivers business value, such as a customer-facing application or backend process.

workstream

Functional groups in a migration project that are responsible for a specific set of tasks. Each workstream is independent but supports the other workstreams in the project. For example, the portfolio workstream is responsible for prioritizing applications, wave planning, and collecting migration metadata. The portfolio workstream delivers these assets to the migration workstream, which then migrates the servers and applications.

WORM

See [write once, read many](#).

WQF

See [AWS Workload Qualification Framework](#).

write once, read many (WORM)

A storage model that writes data a single time and prevents the data from being deleted or modified. Authorized users can read the data as many times as needed, but they cannot change it. This data storage infrastructure is considered [immutable](#).

Z

zero-day exploit

An attack, typically malware, that takes advantage of a [zero-day vulnerability](#).

zero-day vulnerability

An unmitigated flaw or vulnerability in a production system. Threat actors can use this type of vulnerability to attack the system. Developers frequently become aware of the vulnerability as a result of the attack.

zero-shot prompting

Providing an [LLM](#) with instructions for performing a task but no examples (*shots*) that can help guide it. The LLM must use its pre-trained knowledge to handle the task. The effectiveness of zero-shot prompting depends on the complexity of the task and the quality of the prompt. See also [few-shot prompting](#).

zombie application

An application that has an average CPU and memory usage below 5 percent. In a migration project, it is common to retire these applications.