



AI & Machine Learning

Implementing MLOps practices
with Amazon SageMaker

Shelbee Eigenbrode

AI/ML Specialist Solutions Architect

Agenda

300 Level



- MLOps: Setting the stage
- SageMaker MLOps-ready capabilities
- SageMaker Pipelines
- Wrap-up

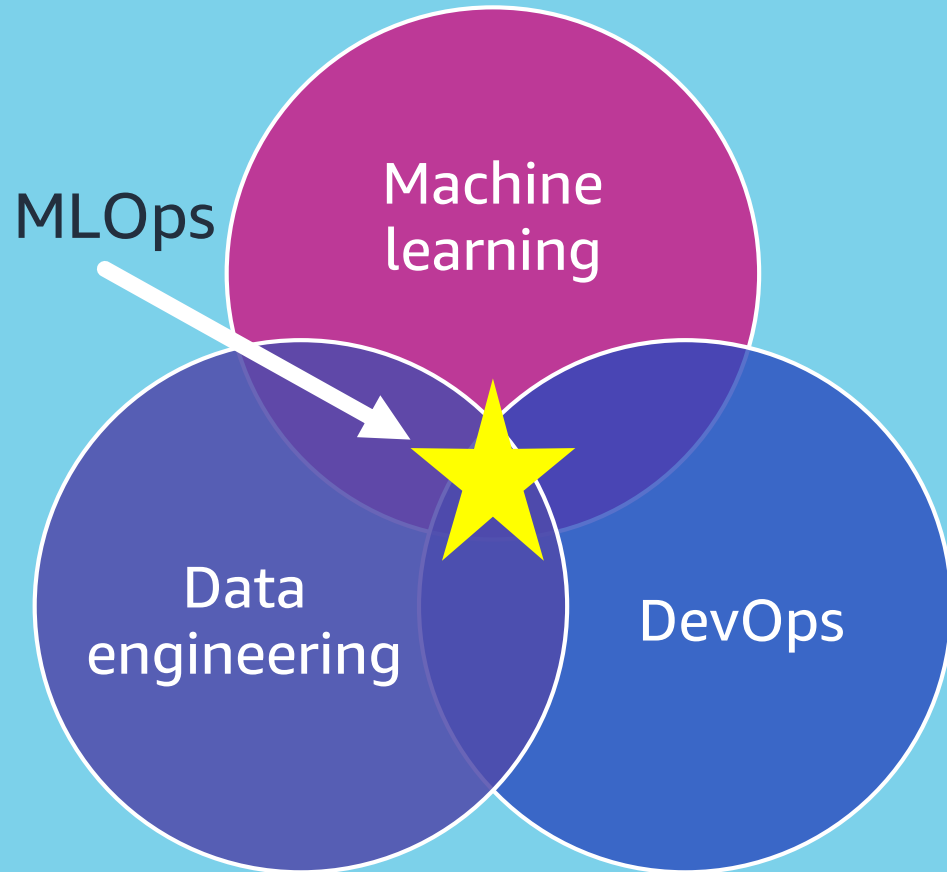
MLOps: Setting the stage



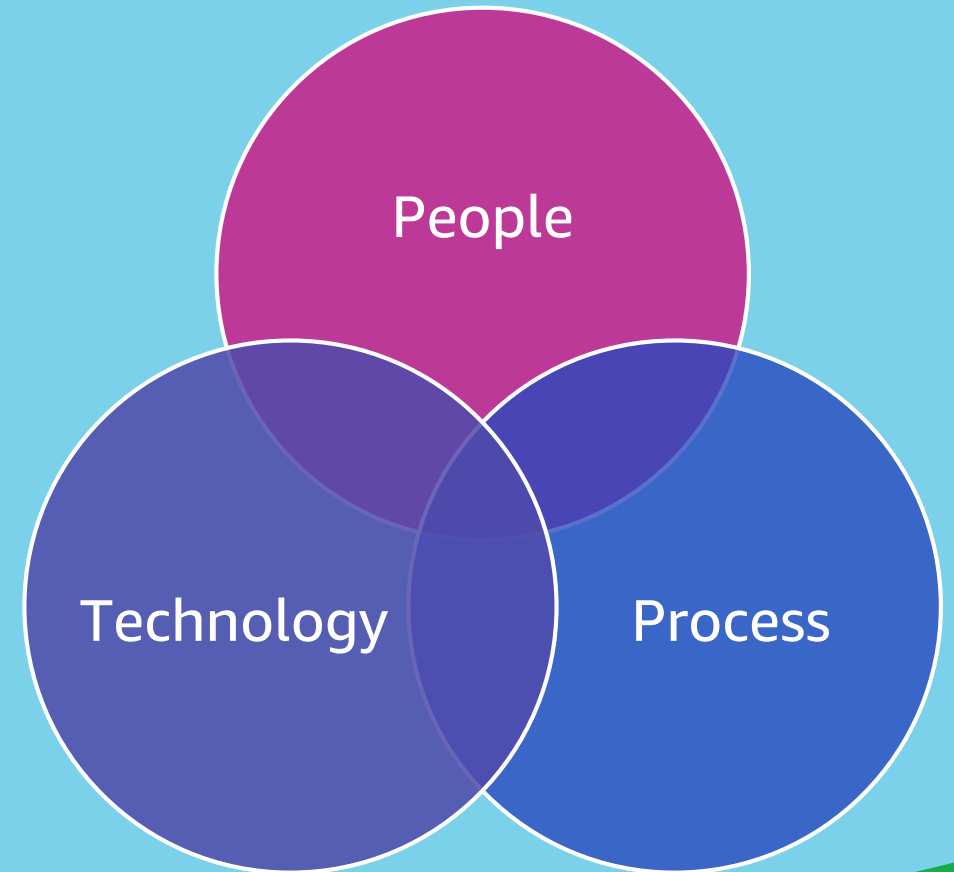
MLOps: High-level view

Intersection of Domains with Enabling Practices

Domains

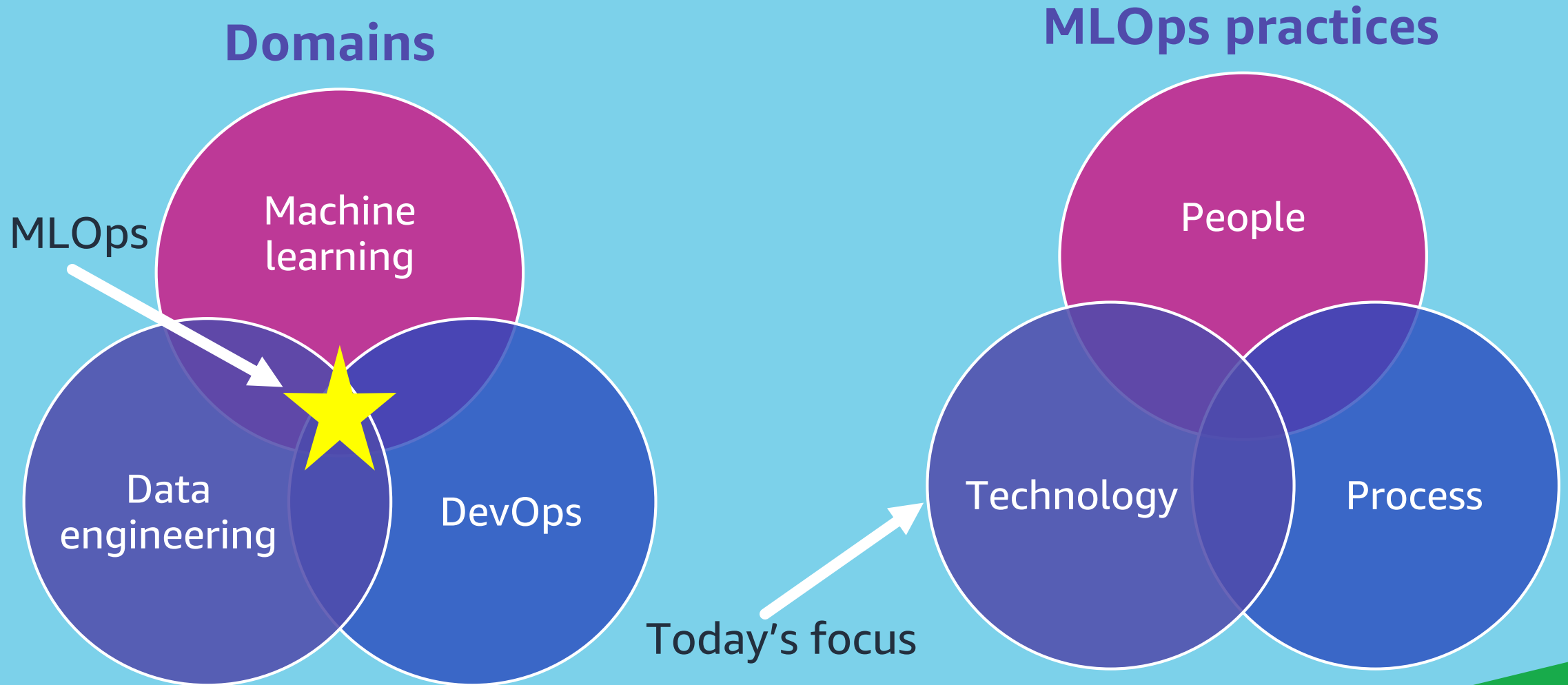


MLOps practices



MLOps: High-level view

Intersection of Domains with Enabling Practices



MLOps: The golden pipeline

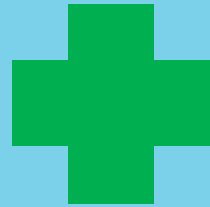
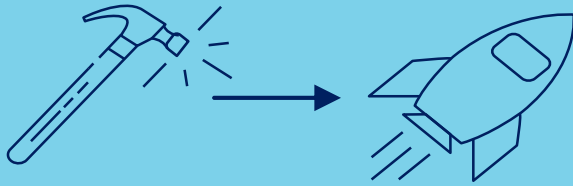


While the technical implementation can vary, the core practices remain the same.

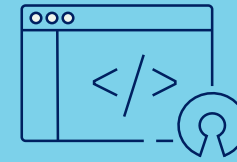
Machine learning pipelines

Technical Implementations: Services & Tooling

Purpose-built services for
automating machine learning tasks



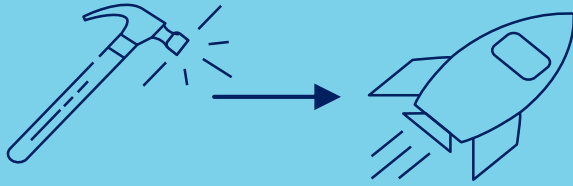
Purpose-built services for incorporating
everything-as-code and **CI/CD practices**



Machine learning pipelines

Technical Implementations: Services & Tooling

Purpose-built services for
automating machine learning tasks



AWS-native examples:



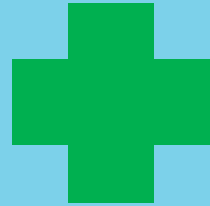
Amazon SageMaker
Pipelines



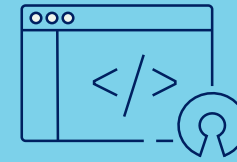
Amazon Managed
Workflows
for Apache Airflow
(Amazon MWAA)



AWS
Step Functions



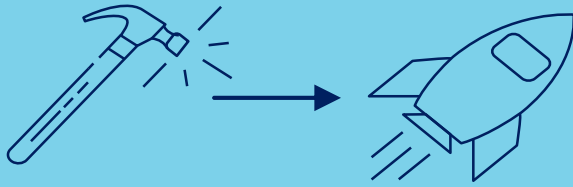
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Machine learning pipelines

Technical Implementations: Services & Tooling

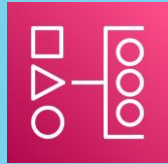
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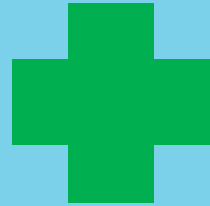
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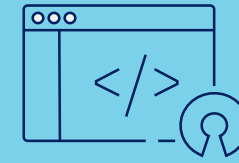
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AWS
Step Functions



Purpose-built services for incorporating
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AWS-native examples:



Amazon
SageMaker
Projects



AWS CodeBuild



AWS CodePipeline



AWS CodeCommit

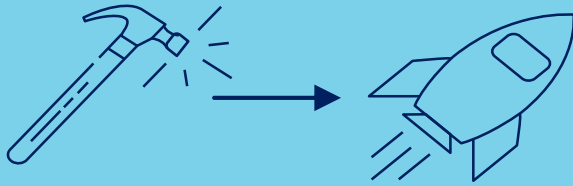


AWS CloudFormation

Machine learning pipelines

Technical Implementations: Services & Tooling

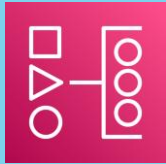
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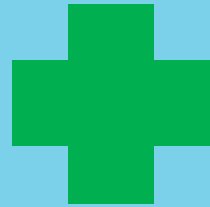
Amazon SageMaker
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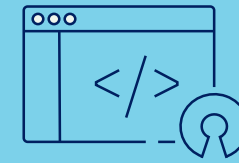
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AWS
Step Functions



Purpose-built services for incorporating
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AWS-native examples:



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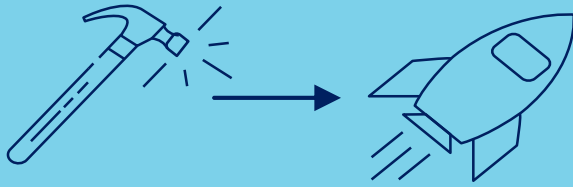
AWS CloudFormation

Do-it-yourself, 3rd Party & Open Source

Machine learning pipelines

Technical Implementations: Services & Tooling

Purpose-built services for **automating machine learning tasks**



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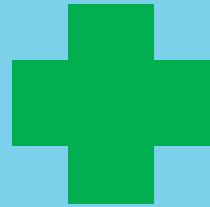
Amazon SageMaker Pipelines



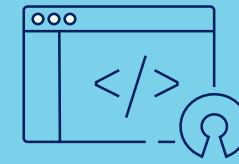
Amazon Managed Workflows for Apache Airflow (Amazon MWAA)



AWS Step Functions



Purpose-built services for incorporating **everything-as-code** and **CI/CD practices**



AWS-native examples:



Amazon SageMaker Projects



AWS CodeBuild



AWS CodeCommit



AWS CodePipeline



AWS CloudFormation

Do-it-yourself, 3rd Party & Open Source

SageMaker MLOps-Ready Capabilities



MLOps: Core practices & benefits

Common challenges

Enabling innovation
while balancing
governance needs

MLOps practice

Everything as Code
(DevOps)

Benefit

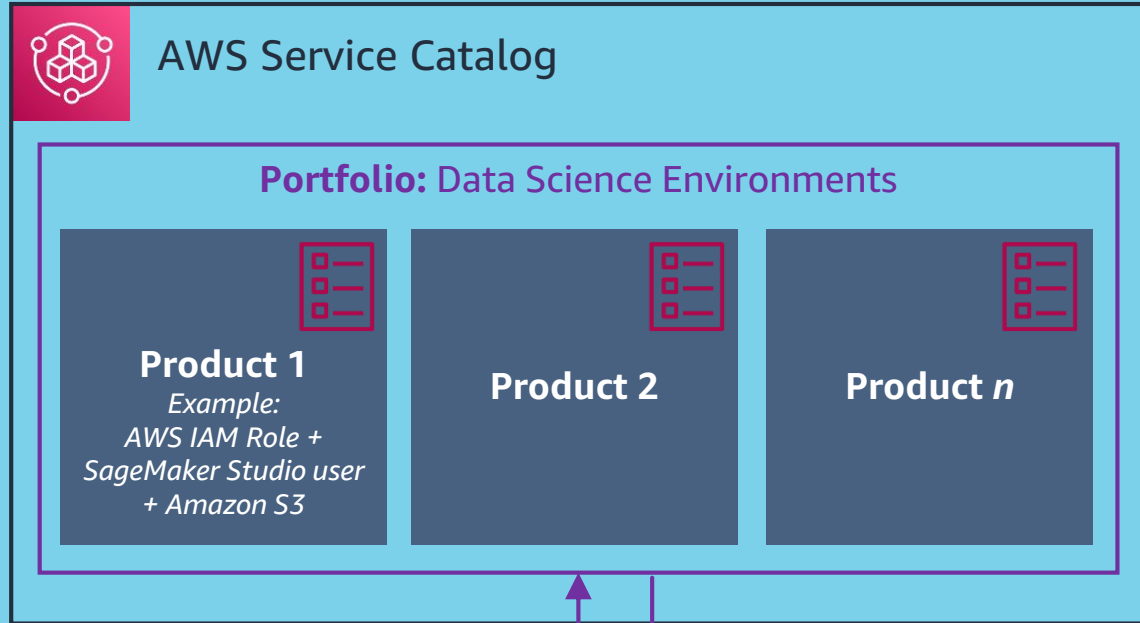
Self-service governed data
science environments

Solution

AWS
Service Catalog

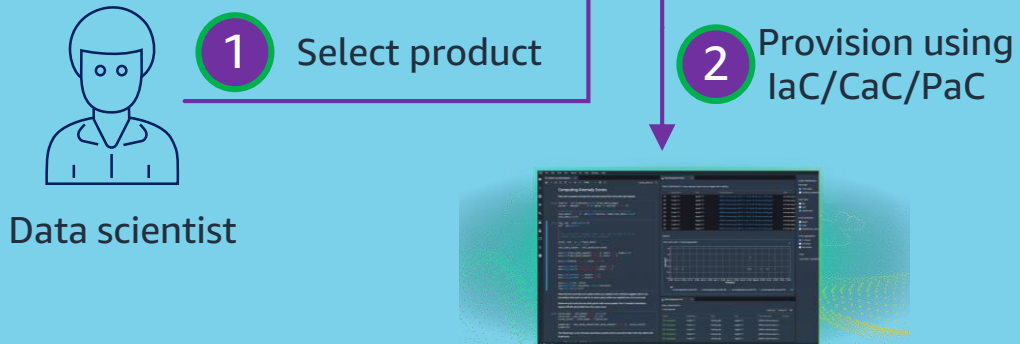
Data science environments

Challenge: Enabling innovation while balancing governance needs



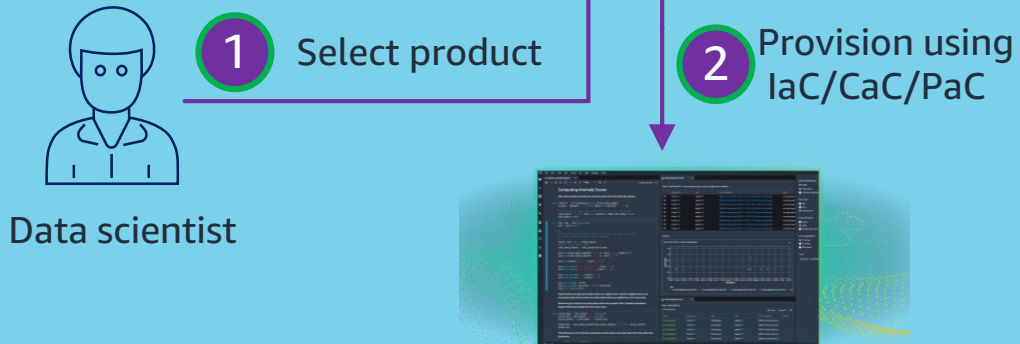
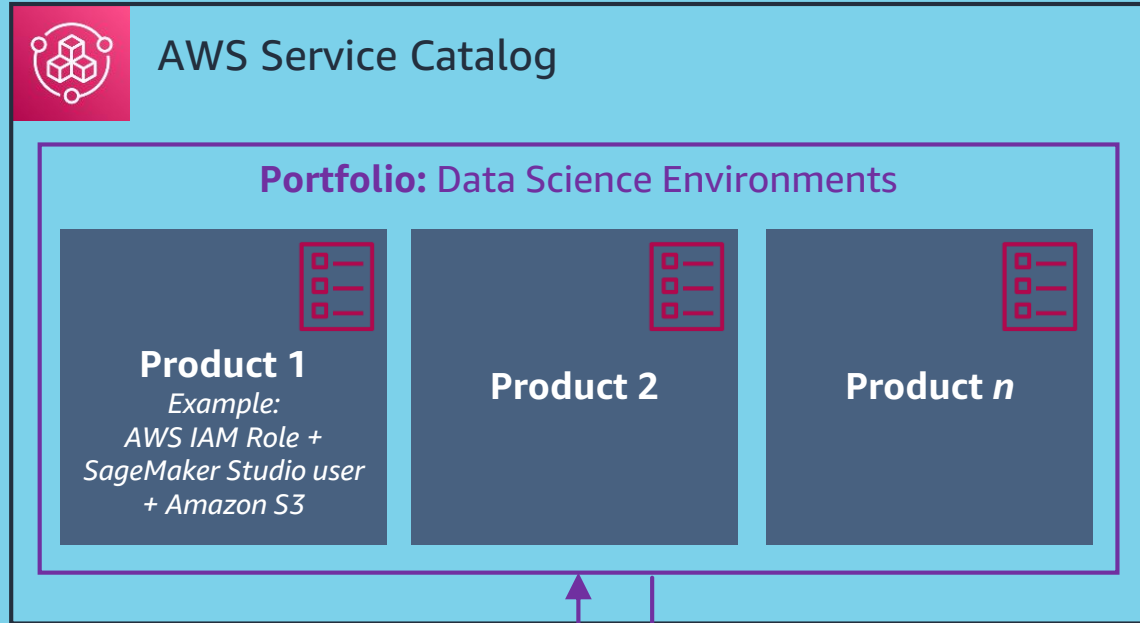
Solution

Utilize AWS Service Catalog to provision environments using infrastructure, configuration, or policy as code



Data science environments

Challenge: Enabling innovation while balancing governance needs



Solution

Utilize AWS Service Catalog to provision environments using infrastructure, configuration, or policy as code

Benefits:

- Ensure repeatability and consistency
- Standardize data science environments
- Implement guardrails via code

MLOps: Core practices & benefits

Common challenges

Enabling innovation while balancing governance needs

Bridging the gap between model build and model deploy tasks

MLOps practice

Everything as Code (DevOps)

Lineage tracking and configuration management

Benefit

Self-service governed data science environments

Repeatable process to create workflows across personas

Solution

AWS Service Catalog

Amazon SageMaker Model Registry

MLOps: Core practices & benefits

Common challenges

Enabling innovation while balancing governance needs

Bridging the gap between model build and model deploy tasks

Providing end-to-end traceability

MLOps practice

Everything as Code (DevOps)

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CI/CD

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Self-service governed data science environments

Repeatable process to create workflows across personas

Auditable ML pipeline

Solution

AWS Service Catalog

Amazon SageMaker Model Registry

Amazon SageMaker projects

MLOps: Core practices & benefits

Common challenges

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Bridging the gap between model build and model deploy tasks

Providing end-to-end traceability

Building a repeatable and consistent process for deploying models

MLOps practice

Everything as Code (DevOps)

Lineage tracking and configuration management

CI/CD

Continuous delivery and deployment

Benefit

Self-service governed data science environments

Repeatable process to create workflows across personas

Auditable ML pipeline

Improve time-to-market

Solution

AWS Service Catalog

Amazon SageMaker Model Registry

Amazon SageMaker projects

Amazon SageMaker projects and Pipelines

MLOps: Core practices & benefits

Common challenges				
Enabling innovation while balancing governance needs	Bridging the gap between model build and model deploy tasks	Providing end-to-end traceability	Building a repeatable and consistent process for deploying models	Monitoring models
MLOps practice				
Everything as Code (DevOps)	Lineage tracking and configuration management	CI/CD	Continuous delivery and deployment	Continuous monitoring
Benefit				
Self-service governed data science environments	Repeatable process to create workflows across personas	Auditable ML pipeline	Improve time-to-market	Maintain model performance over time
Solution				
AWS Service Catalog	Amazon SageMaker Model Registry	Amazon SageMaker projects	Amazon SageMaker projects and Pipelines	Amazon SageMaker Model Monitor

MLOps-ready features and capabilities

Data preparation

Prepare data

SageMaker
Data Wrangler

SageMaker
Processing Job

MLOps-ready features and capabilities

Data preparation

Prepare data

SageMaker
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Processing Job

Curate features

SageMaker
Feature Store

MLOps-ready features and capabilities

Data preparation

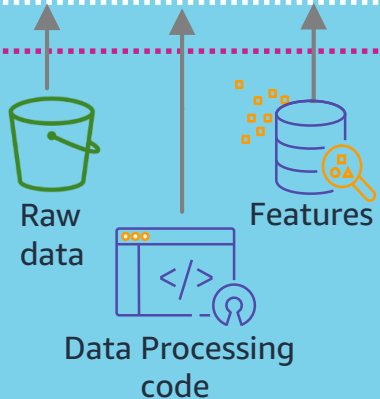
Prepare data

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MLOps-ready features and capabilities

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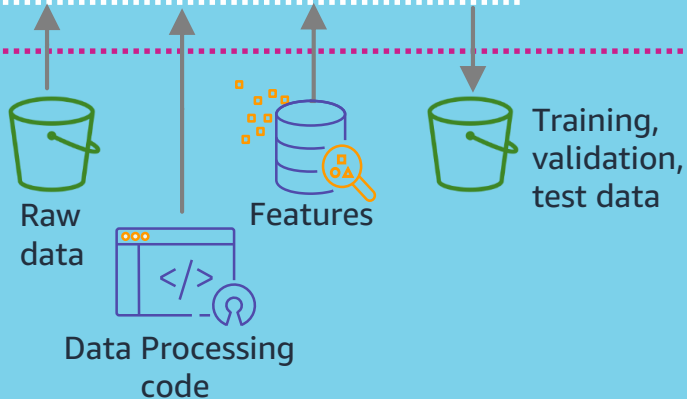
Prepare data

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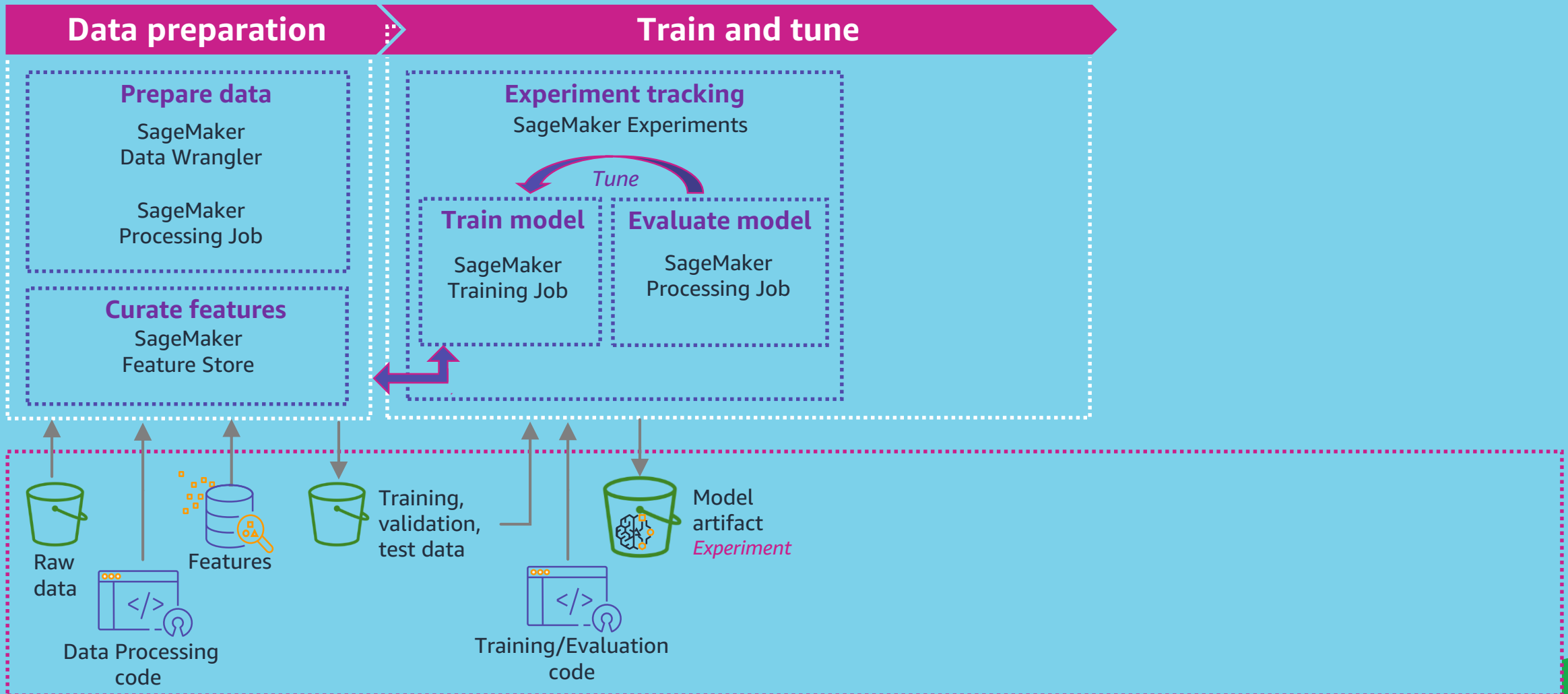
SageMaker
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Curate features

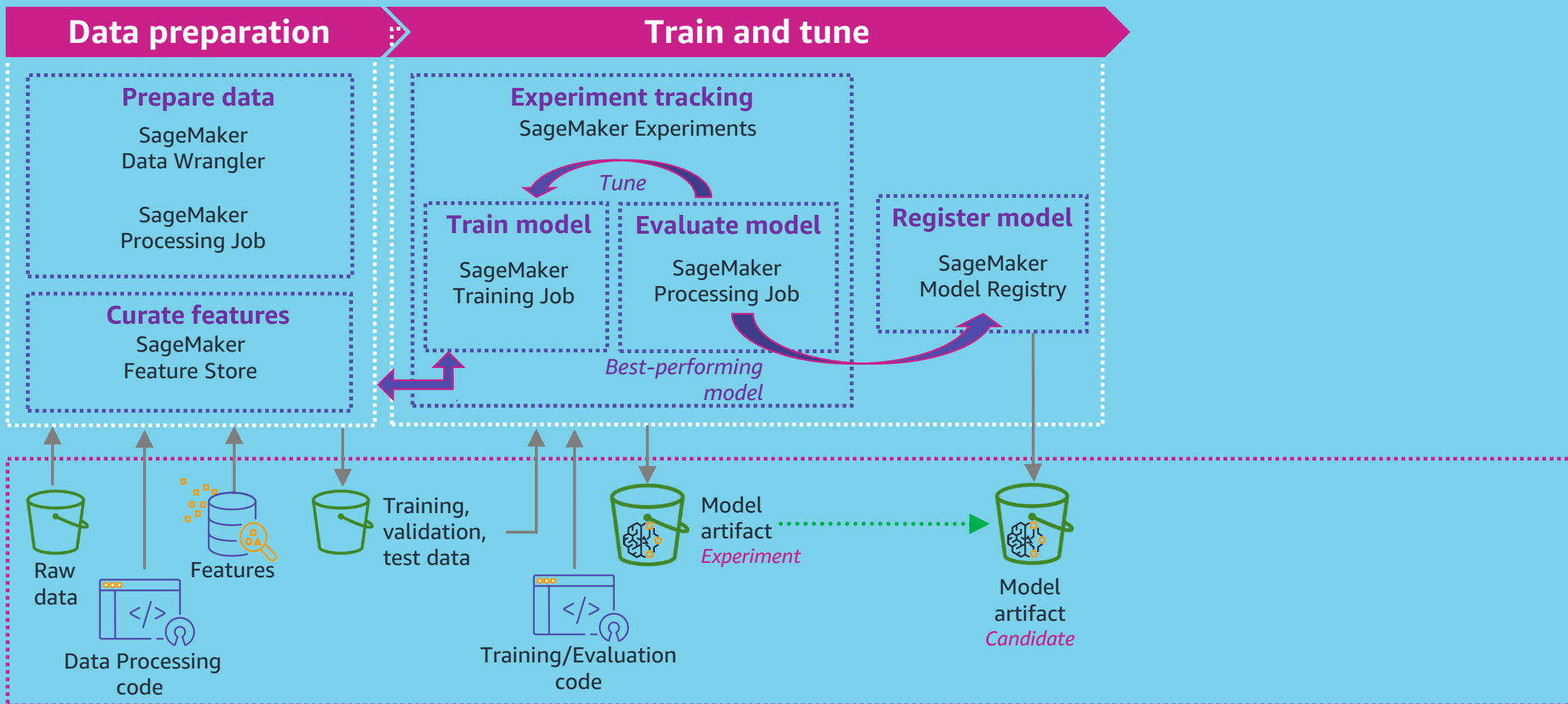
SageMaker
Feature Store



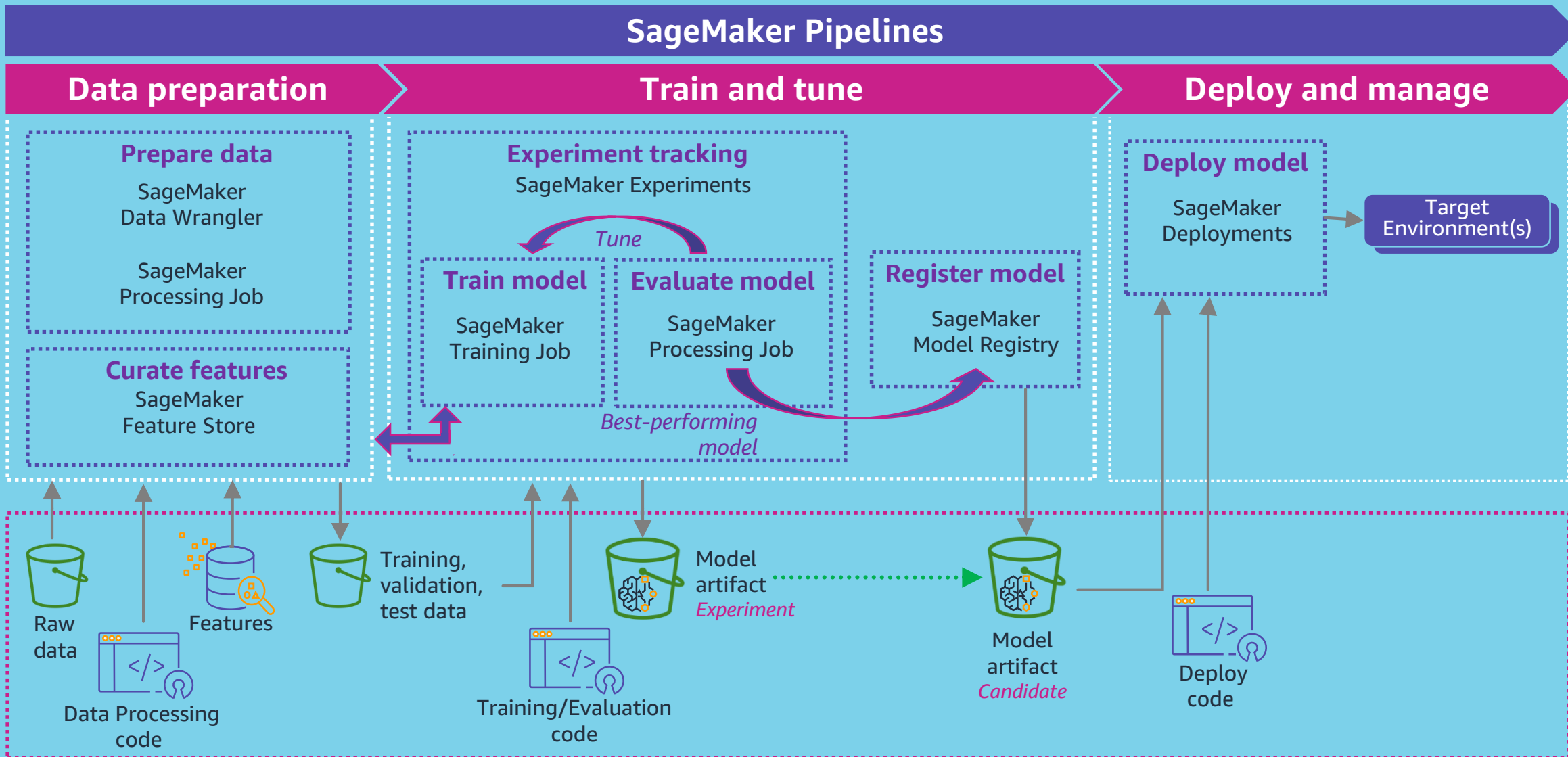
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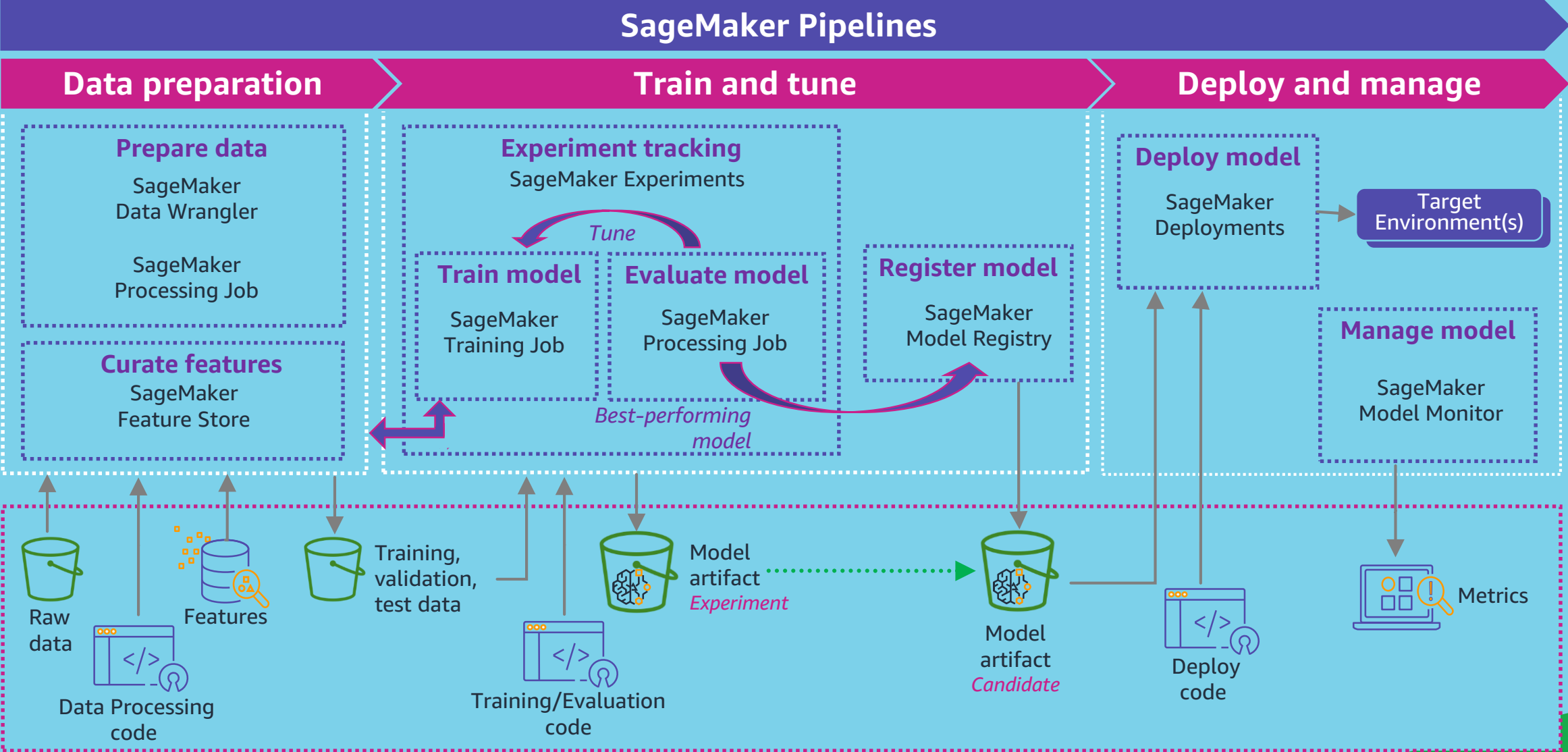
MLOps-ready features and capabilities



MLOps-ready features and capabilities



MLOps-ready features and capabilities



SageMaker Pipelines



SageMaker Pipelines



Python SDK for quickly and easily **building ML workflows**

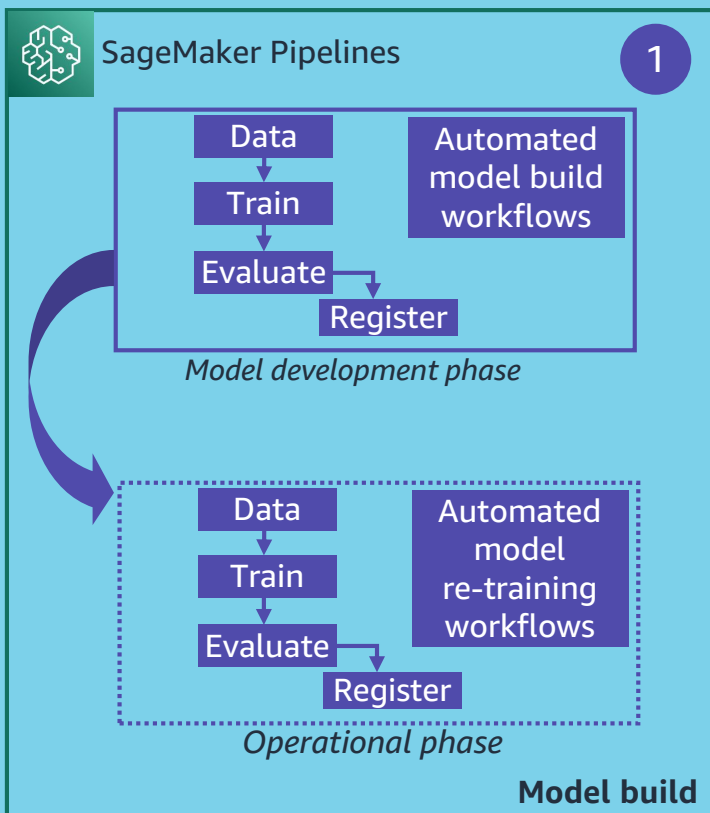


Catalog models to manage models at scale and trigger automated deployment workflows

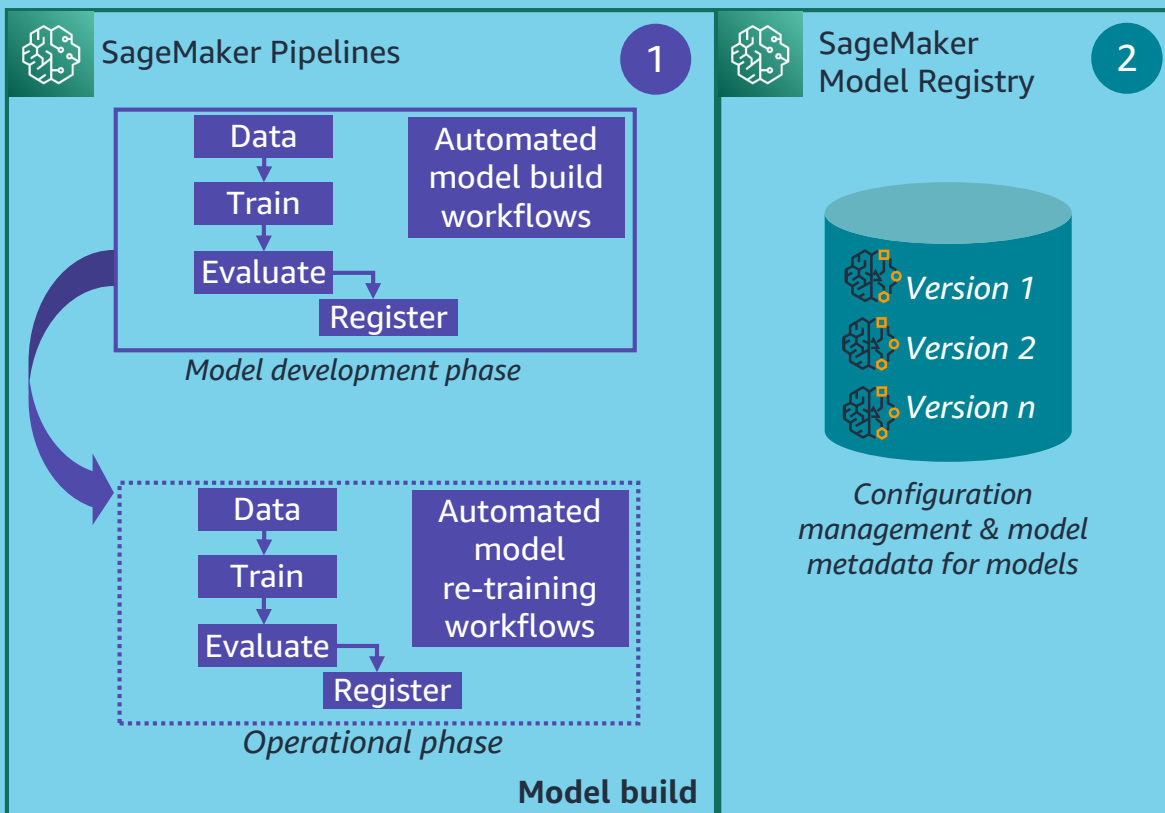


Built-in support for **CI/CD** and end-to-end **lineage tracking**

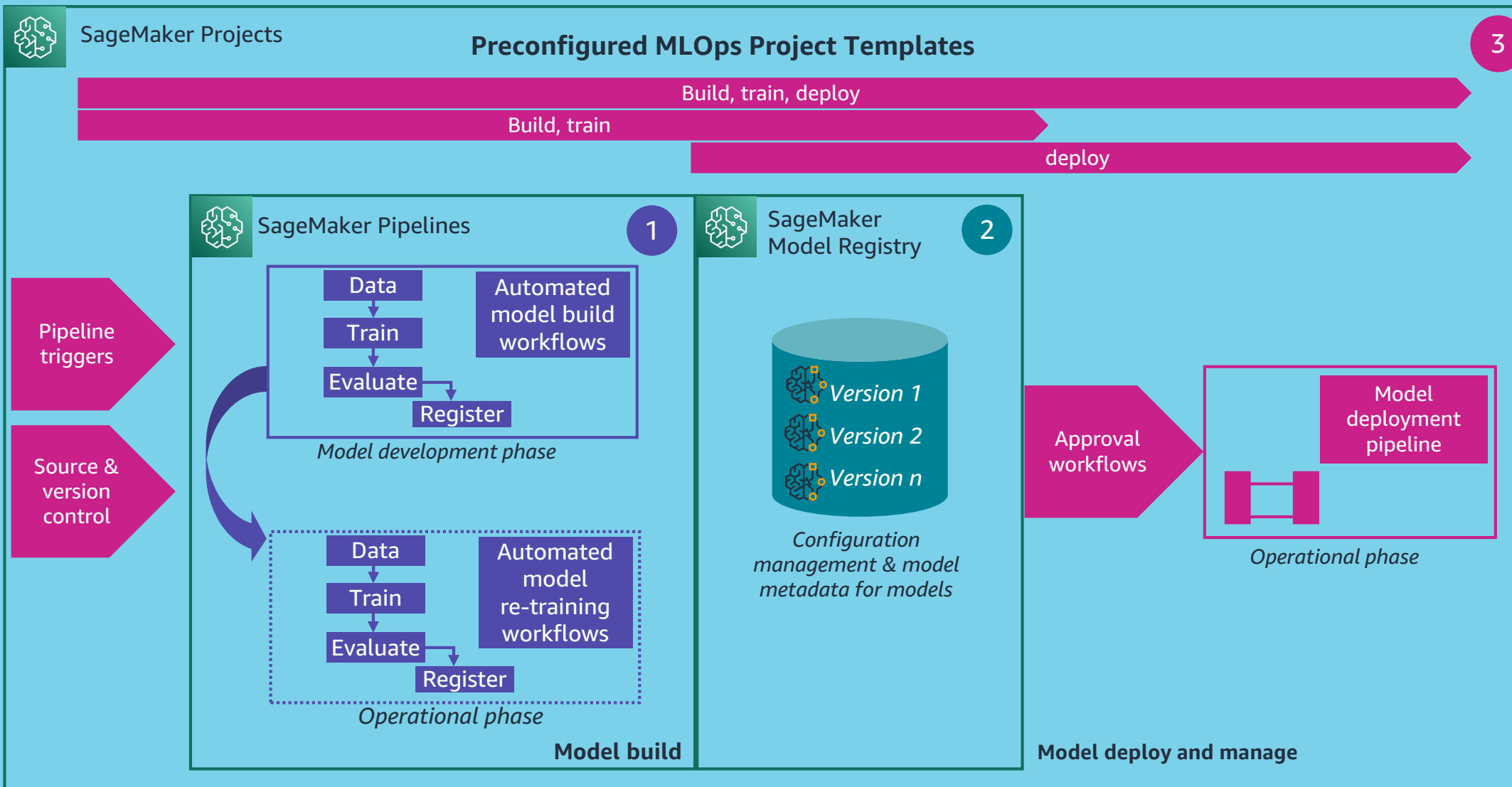
SageMaker Pipelines: components



SageMaker Pipelines: components



SageMaker Pipelines: components



SageMaker Pipelines - Demo

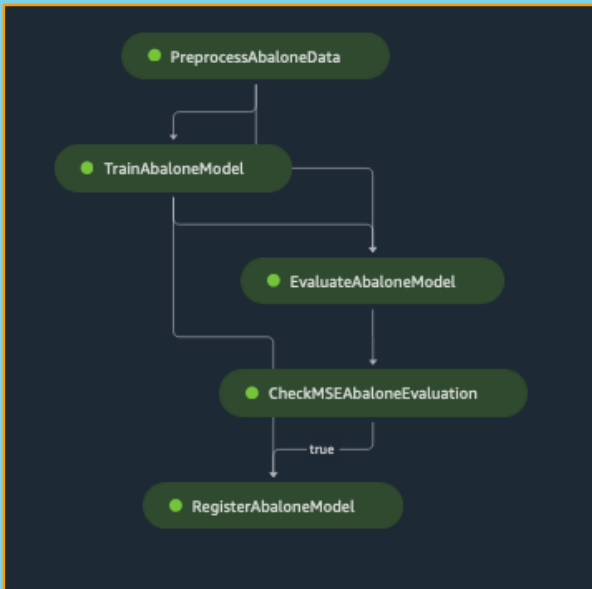
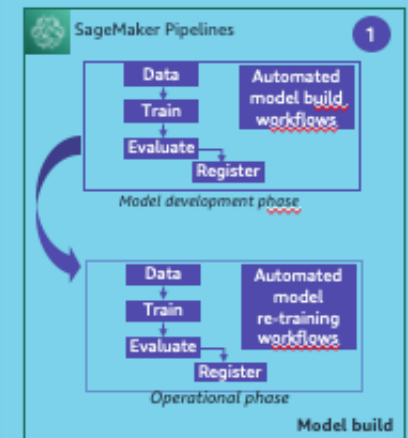


<<Insert Recorded Demo>>

SageMaker Pipelines: Model build pipelines

High Level Features

Built-in steps to support model build activities



- **SageMaker native:** Processing, Training, Tuning, Create Model, Register Model, Batch Transform, Clarify Check, Quality Check
- **AWS services:** Lambda, EMR
- **Extend to other AWS services & tasks using Callback step**
- **Conditional steps to assess next pipeline steps**

SageMaker Pipelines: Model build pipelines

High Level Features



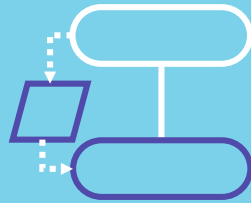
Integration with SageMaker Experiments

Automatically track pipeline executions – no setup required!



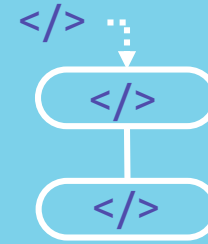
Pipeline step retry policies

Retry failed steps in the pipeline without re-running the full pipeline



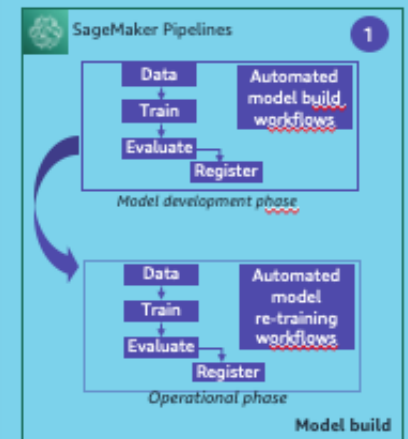
Step caching

Enable step caching to avoid unnecessary runs of pipeline steps



Pipeline parameterization

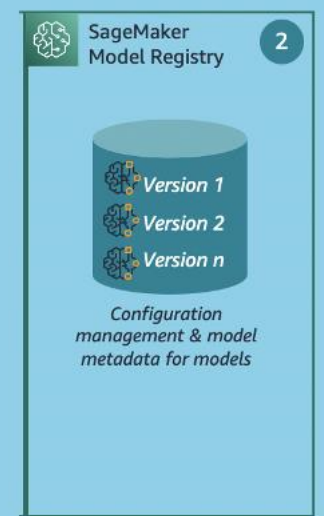
Use pipeline parameters to define configurations specific to a pipeline execution



SageMaker Pipelines: Model Registry

High Level Features

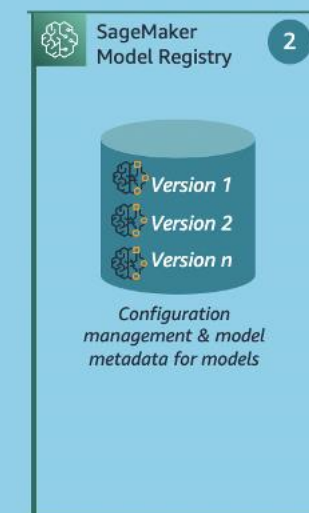
Versions		Settings				
Search column name to start						
Version	Stage	Status	Short description	Modified by	Last modified	
3	None	Pending				
2	prod	Approved		workshop-user	1 day ago	
1	staging	Approved		shelbee-iggy	2 months ago	



- Manage model versions & associated metadata
- Manage the approval status of a model
- Deploy models to production (with Projects)
- Support for cross-account model registration & deployments

SageMaker Pipelines: Model Registry

High Level Features



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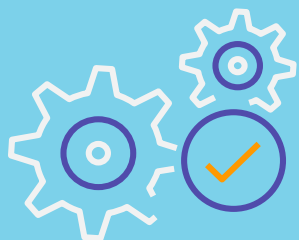
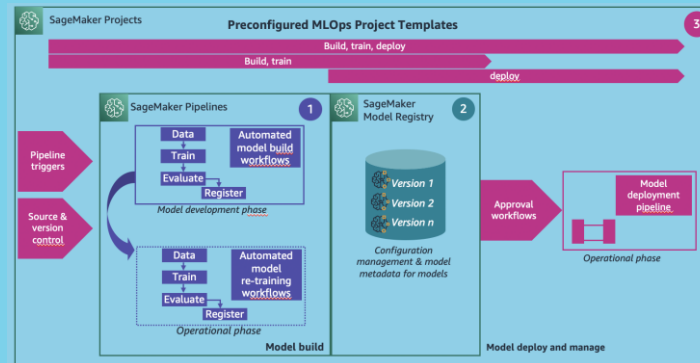
Version 2		Status	Pipeline	Execution	Last Stage	Model group	Update status
		Approved	shelbee-demoagain-p-...	workflow-2	prod	shelbee-demoagain-p-...	

Model metrics		
Model metric	Metric value	Standard deviation
mse	4.823176167079859	2.1960237865043672

- Track model performance metrics

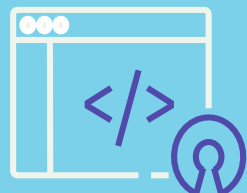
SageMaker Pipelines: Projects

High Level Features



Enable CI/CD practices

source/version control, pipeline traceability, and automated triggers for training & deployment



Support for third-party tools

Including: GitHub, GitHub Enterprise, BitBucket Cloud, and Jenkins



Customizable MLOps Project templates

Utilize built-in templates or build custom templates to meet your requirements



Automatic rollback to previous model version

Rollback deployment to previous version of model

Wrap-up

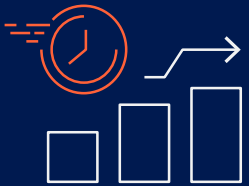


SageMaker MLOps summary

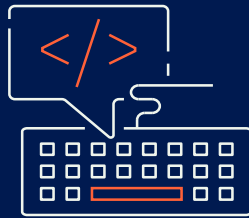


Start small
and iterate

SageMaker MLOps summary

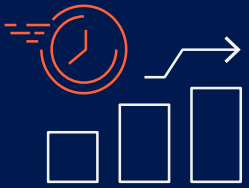


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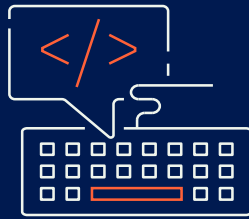


Practice
Everything-as-Code
for repeatability,
consistency, and
governance

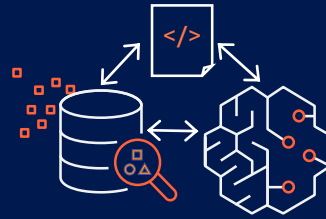
SageMaker MLOps summary



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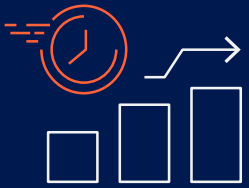


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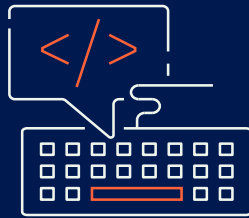


Consider ML
specific inputs,
outputs, and artifacts
for model and pipeline
lineage

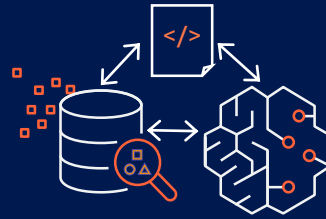
SageMaker MLOps summary



Start small
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Practice
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Use integrated
SageMaker services
when possible to
reduce management
and setup time

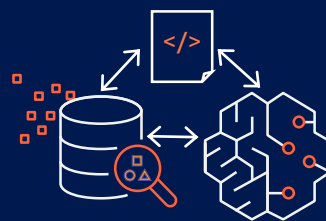
SageMaker MLOps summary



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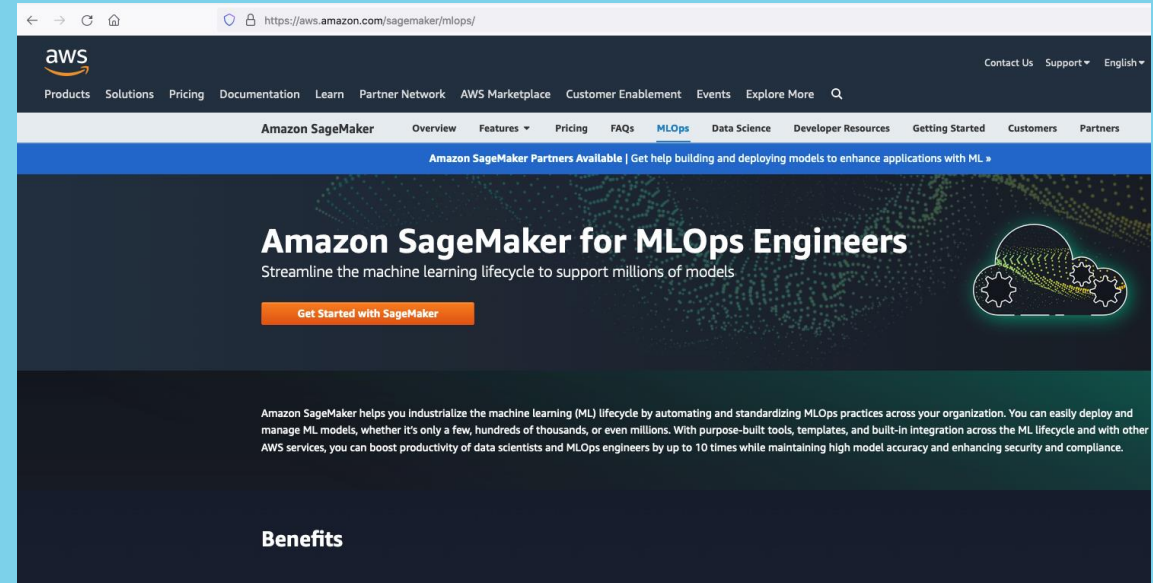


Implement a Model
Registry as part of
your ML pipeline for
managing models

SageMaker MLOps: Resources & Next Steps

Check out the new MLOps Page

<https://aws.amazon.com/sagemaker/mlops/>



Additional Resources

1. SageMaker Documentation: <https://docs.aws.amazon.com/sagemaker/latest/dg/pipelines.html>
2. Custom Project Template Examples: <https://github.com/aws-samples/sagemaker-custom-project-templates>



Thank you!