#### AIM321

# Productionize ML workloads using Amazon SageMaker MLOps, feat. NatWest

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# AIM321 🔼

Productionize ML workloads using Amazon Sagemaker MLOps featuring NatWest

#### Speakers

- Shelbee Eigenbrode, Principal Al/ML Specialist, WW MLOps Lead, AWS
- Greig Cowan, Head of Data Science, NatWest Group
- Usman Anwer, Principal Product Manager, AWSRelevant

#### **Announcements**

n/a – although new improvements are shown.

#### Takeaways

NatWest Case Study is best section of this session. How to scale your MLOps w/Vision, Metrics and Outcomes.

#### Azure

Azure ML standardizes on the Mlflow open-source framework for MLOps.

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## What is MLOps

- The process of continuously delivering high-performance ML Models at scale. DevOps for Machine Learning.
- People, Process, Tools
- Overall, a weak definition of MLOps. A better overview can be found on Nvidia's site here.

## How does Sagemaker Support MLOps

- This session focuses on the Model Lifecycle
  - Project Environment Creation
  - Model Training
  - Model Registry
  - Model CI/CD Pipelines
  - Model Deployment
  - Model Monitoring

## What's new in SageMaker for MLOps

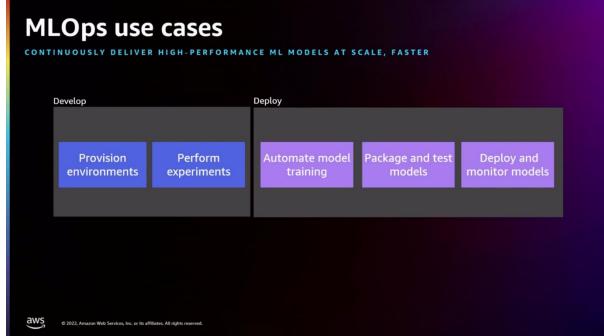
- Pipeline (CI/CD) Improvements
  - Local Mode: Helps to run, test pipelines locally, to reduce cycle time and costs during pipeline creation. Deploy to cloud when ready.
  - AutoML Support: Allows the pipelining of AutoML jobs in workflow.
  - Cross-Account Sharing: Coordinate pipelines across enterprise accounts
  - SDK Simplifications: Improved Python SDK Capabilities
- Model Monitor
  - Batch Inference: New capability to Monitor Batch Inference Models
  - <u>Demo</u> of Batch Inference Model Monitoring

## Scaling MLOps

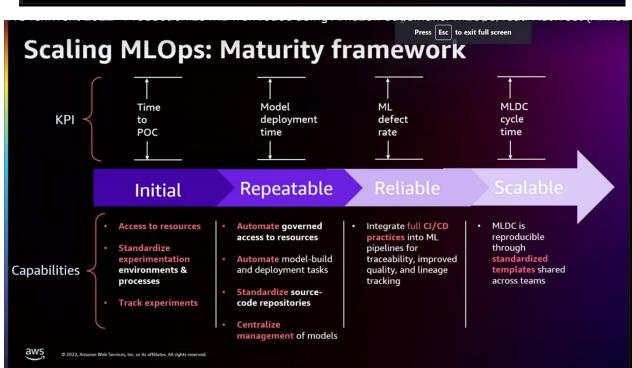
■ Initial, Repeatable, Reliable, Scalable

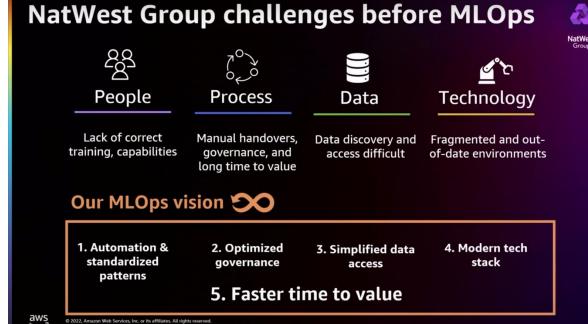
## NatWest Case Study

- Challenges around People, Process, Data, Technology.
- Established Metrics to solve these for Faster Time to Value
  - Faster delivery of end-to-end solutions (> 12 mos) => (< 3 mos)
  - Simplified discovery and access to data (> 5 Days) => (< 1 day)</li>
  - Simplified ML Model Route to Production (> 6 mos) => (< 2 weeks)</li>
  - End User self-service environment creation (> 50 days) => (< 2 days)</li>









# **Adoption and engagement**





**Training** 

Trained 100s of

data scientists.

engineers and

leaders across >700

AWS courses and

**Immersion Days** 

**Dedicated** support

Established wellresourced (12 FTE) and capable enablement team...

...gathering continuous feedback



Visible metrics

Continually track and report on KPIs related to speed of setup, training completion, use-case success and

data access



User champions

Early adopters identified and given priority support to validate and test the capability



Reach far and wide

Project teams and use cases from across the organization, e.g.,

> Fraud Financial crime **Finance** Audit Credit Climate Pricing

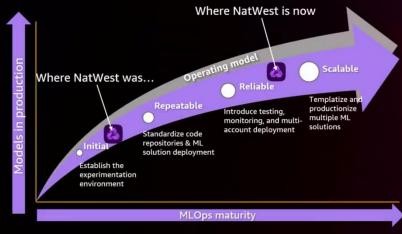


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# A metric-based approach focused on business value



NatWest Group



NatWest was	NatWest is now
~12 months	< 3 months
~5 days	< 1 day
3-6 months	< 2 weeks
> 30 days	< 2 hours
	vas ~12 months ~5 days  3-6 months

\* Other metrics also used during project delivery



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# **Key takeaways**



Hearts & minds

Create a

compelling and

business-led vision

for MLOps

Adoption &

engagement

community

**Build for** complexity

Data scientists & engineers are a smart bunch!

Consider orchestration beyond SageMaker support for the user

Bring your own containers & templates



**Build for** flexibility

MLOps is a journey that you won't take in one step

Deliver value quickly to users and be able to respond promptly to new requests



Operating model

Consider carefully how your users will build ML-driven products

Centralized vs. federated

Standardization frees devs to focus on the value-add



Legacy tech

Build your system to work with legacy estate

Working between cloud and onpremises environments