



Designing Observable Microservice Architectures for End-to-End ML Pipelines on AWS



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AWS Container Hero



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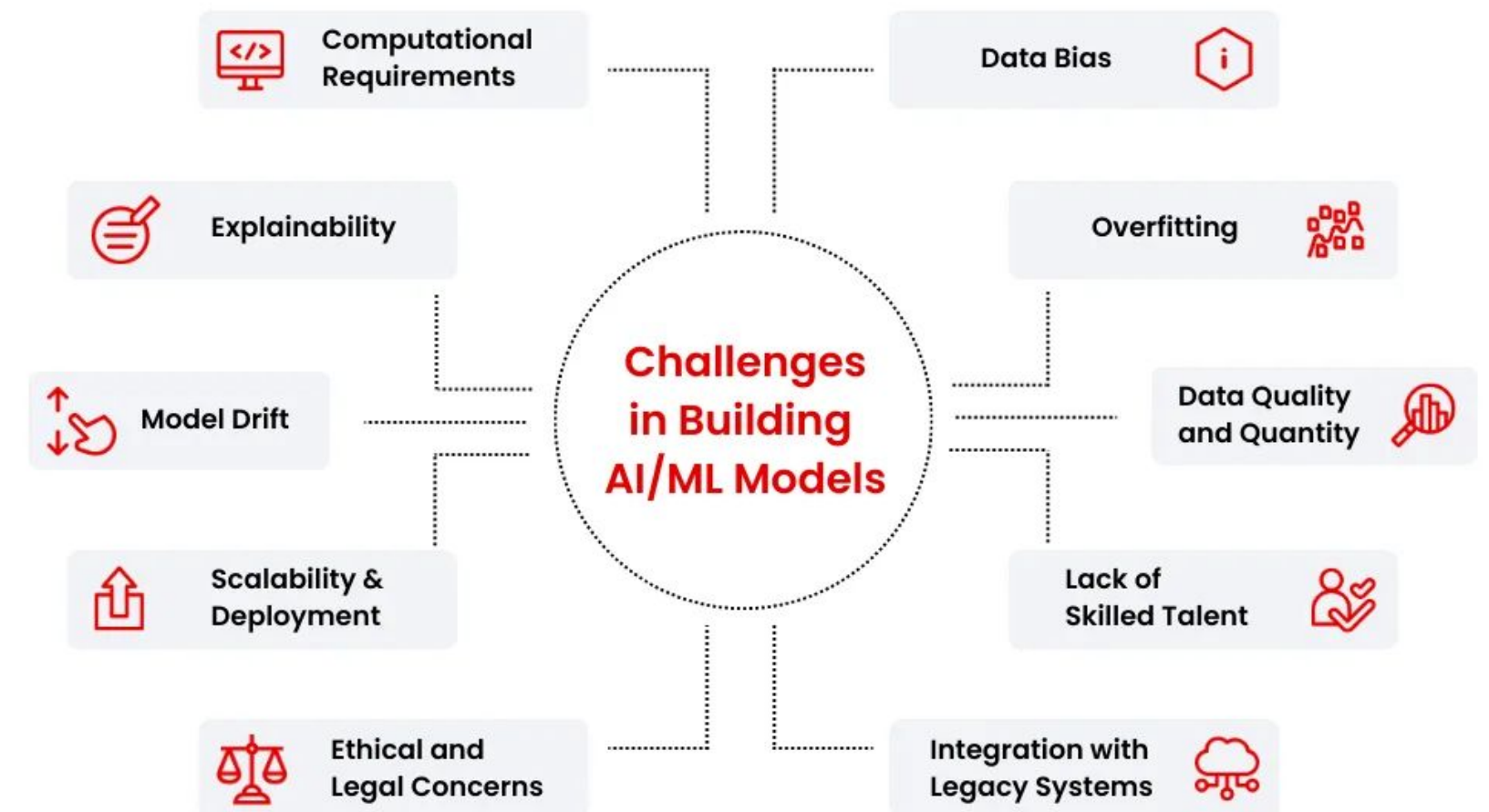
- Cloud & DevOps Engineer, Singapore
- AWS Container Hero
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- <https://linktr.ee/misskecupbung>

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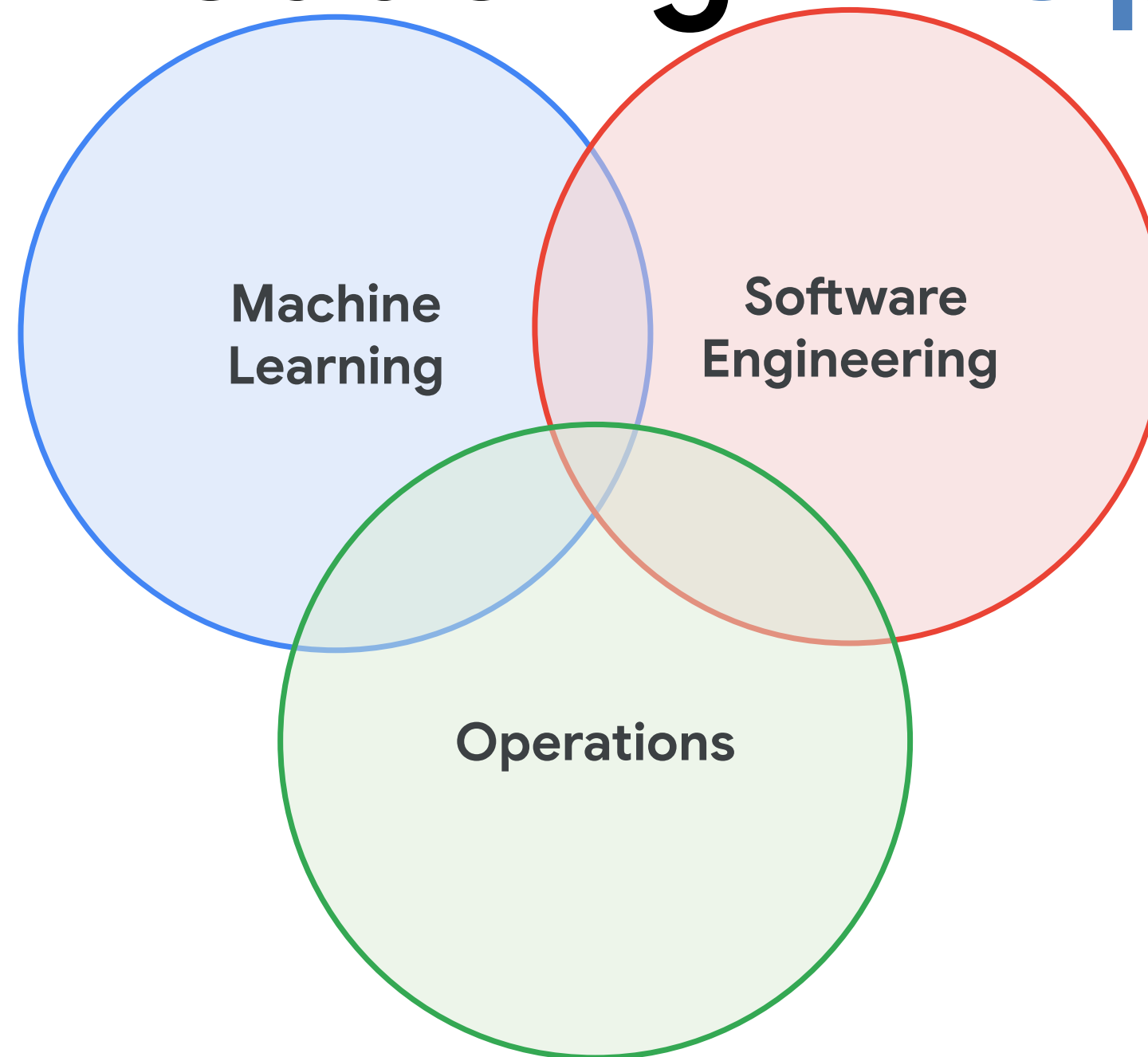
Operating ML models, presents several challenges:

- **Model drift:** As real-world data changes, models become less accurate, requiring frequent retraining.
- **Resource management:** ML workloads have varying demands, making efficient allocation crucial.
- **Data quality:** Consistent, reliable input data is essential for model performance.
- **Compliance:** Meeting governance and regulatory requirements is challenging.
- **Versioning:** Tracking models, datasets, and experiments is difficult at scale.



Introducing MLOps

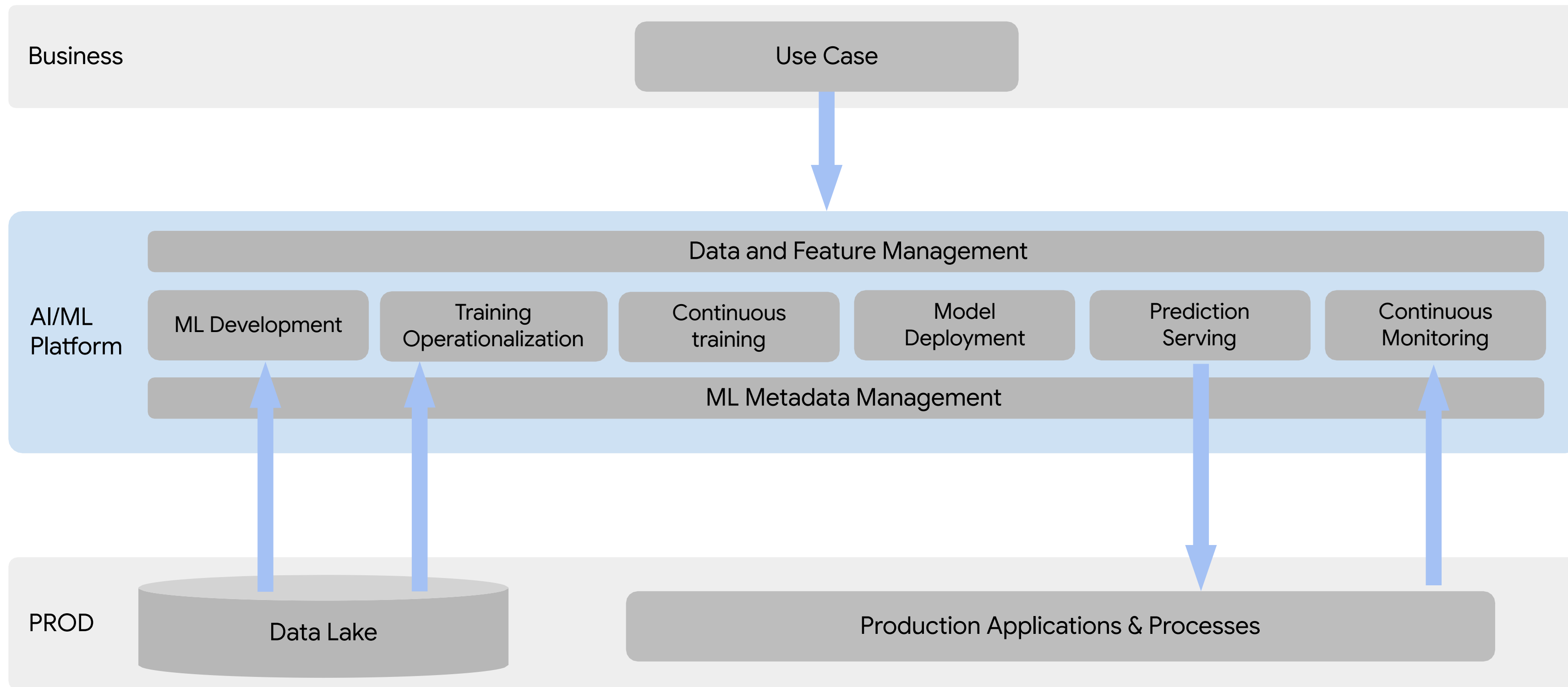
- Model development
- Model evaluation
- Parameter tuning



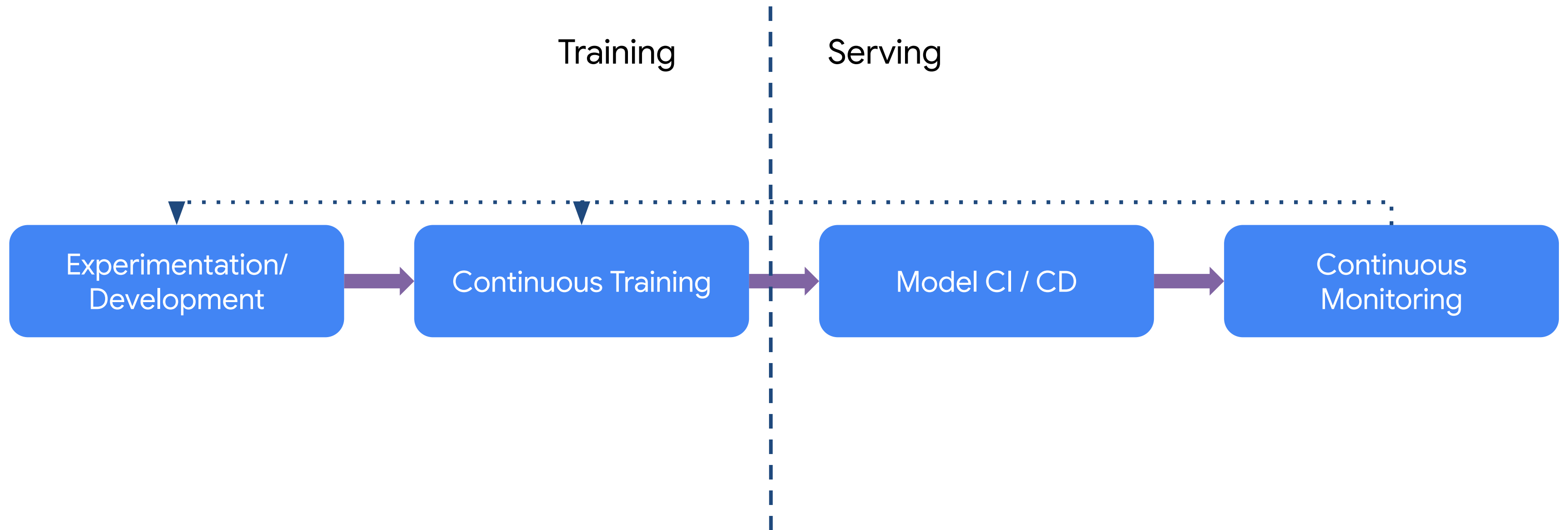
- Data engineering
- Pipeline development
- Integration of model into business application

- Model deployment
- Metadata management
- Logging and monitoring

MLOps: quick recap

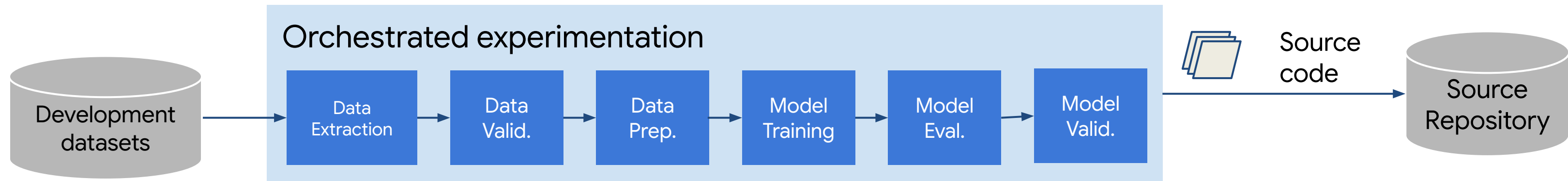


ML Solution Lifecycle



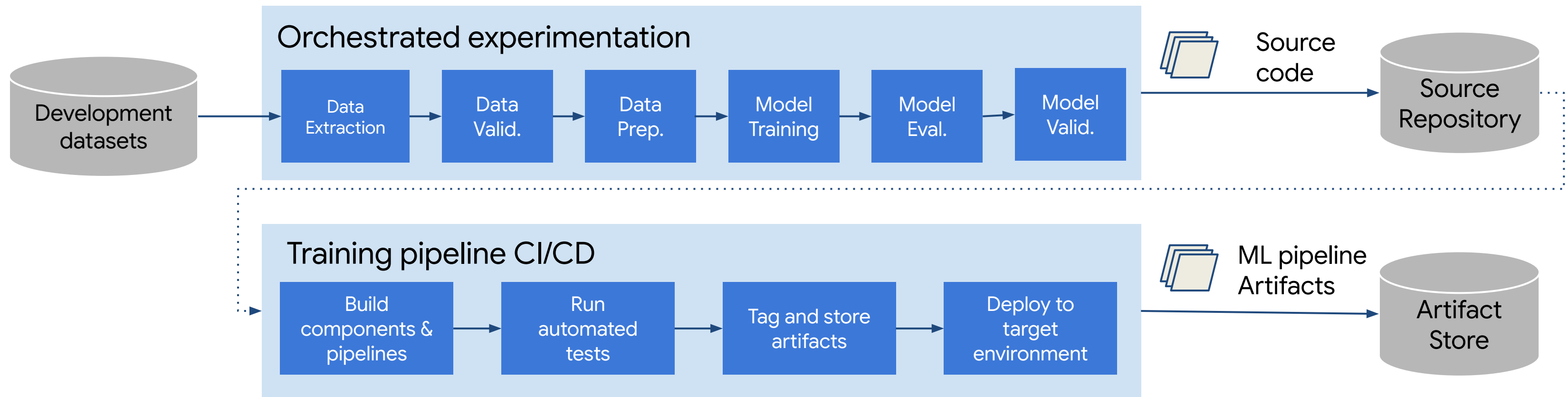
Reliable and repeatable training

Automated E2E Pipelines



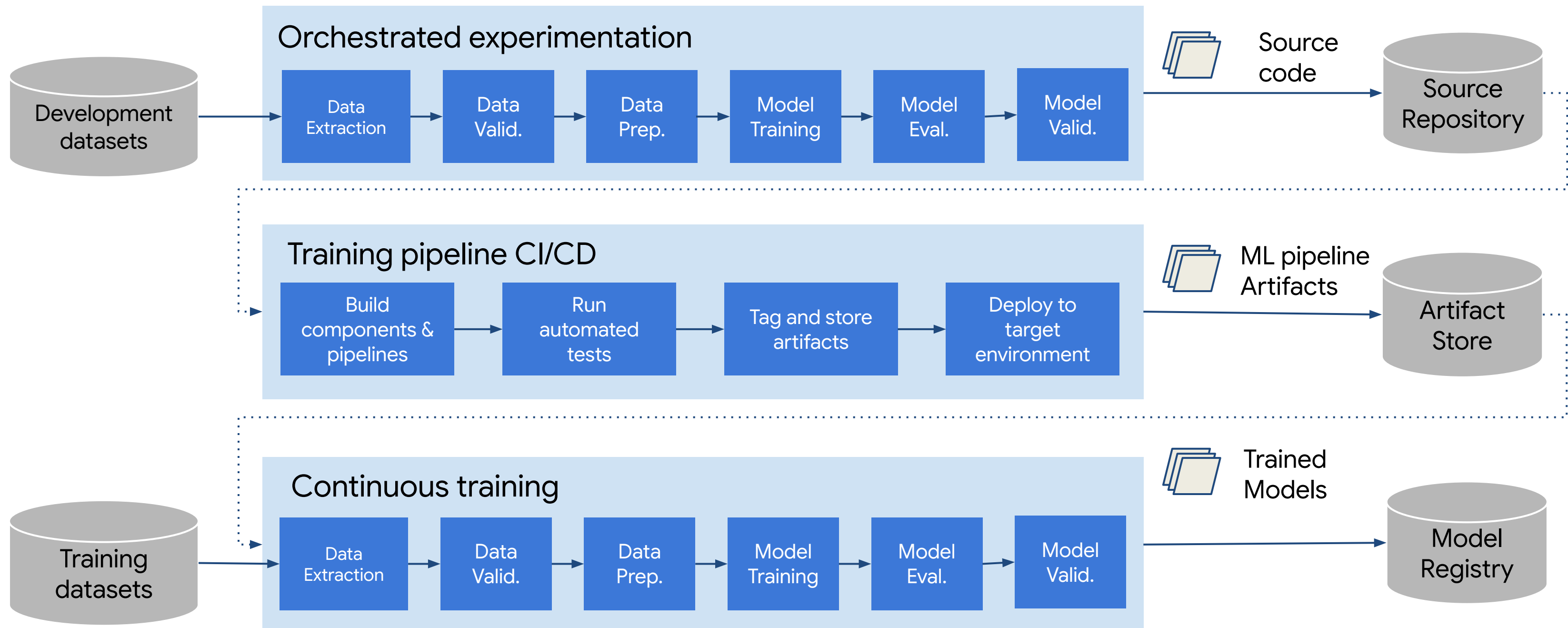
Reliable and repeatable training

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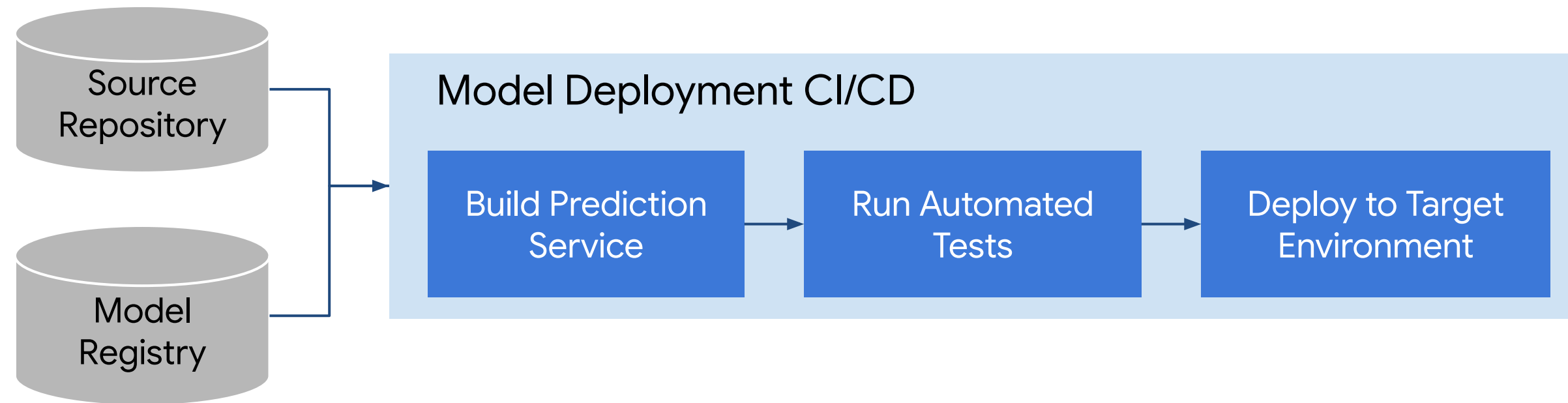
Reliable and repeatable training

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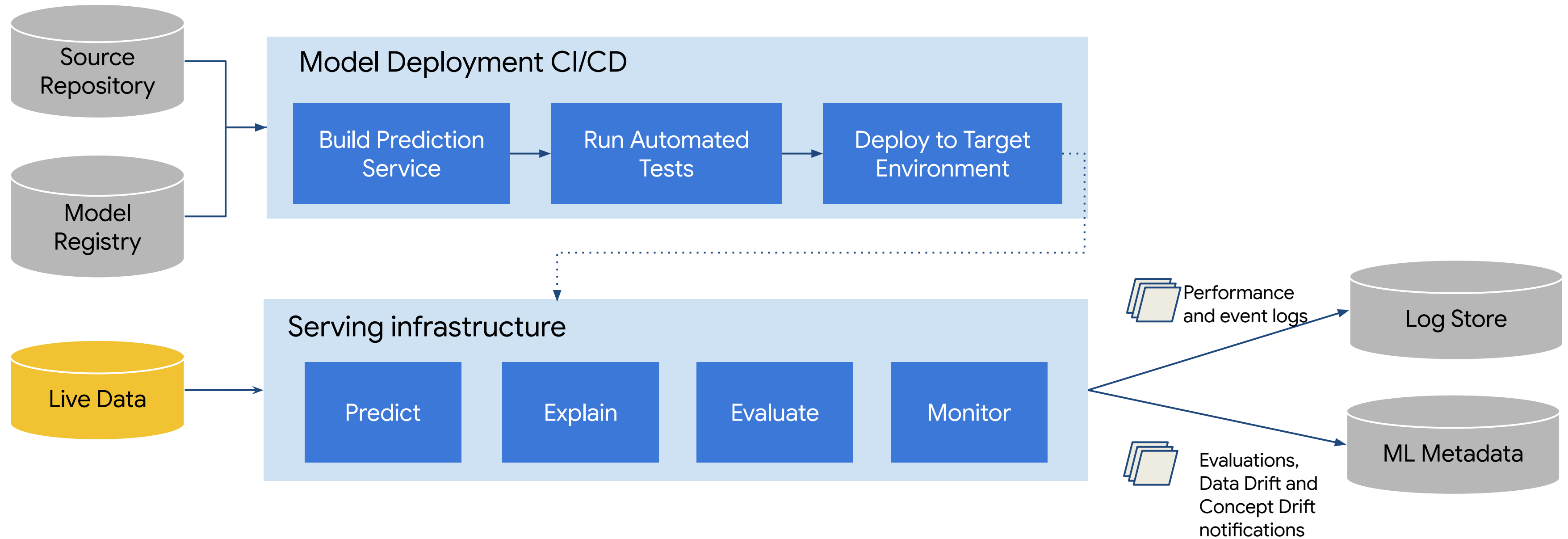
Reliable and monitored serving

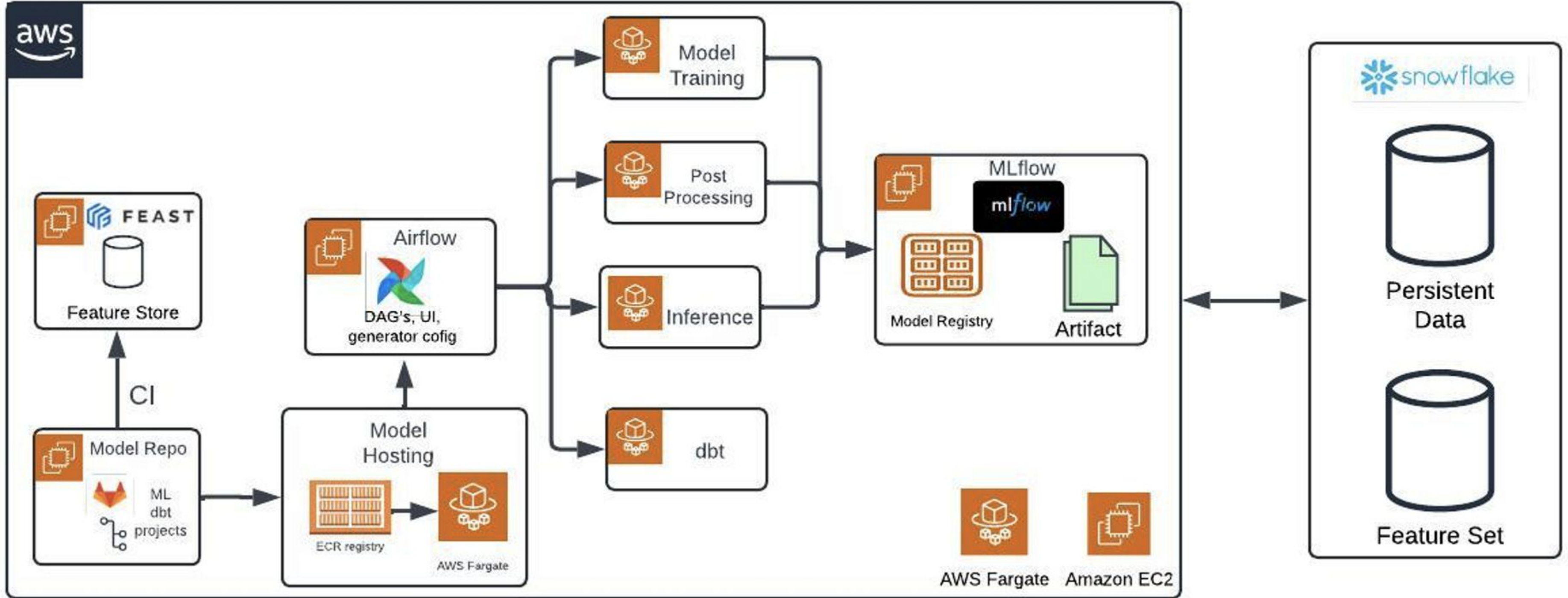
Automated E2E Pipelines



Reliable and monitored serving

Automated E2E Pipelines





Architecture overview:

- Airflow for workflow orchestration Feast for feature
- management dbt for accelerated data transformation
- MLflow for experiment tracking and model management

Source:

<https://aws.amazon.com/blogs/machine-learning/building-an-efficient-mlops-platform-with-oss-tools-on-amazon-ecs-with-aws-fargate/>

“In control theory, **observability** is a measure of **how well** internal states of a system can be inferred from knowledge of its external outputs.”

“Monitoring tells you whether a system is working;
Observability lets you understand why isn't working.”

Unique ML Characteristics

Resource Patterns:

- Sustained high GPU usage during training vs consistent CPU usage in traditional apps
- Specialized GPU node scheduling vs typical short-lived batch jobs
- Variable computational demands requiring dynamic resource allocation

Monitoring Focus:

- **Model-specific metrics:** Accuracy, F1 scores (irrelevant for standard applications)
- **Data drift monitoring:** Track shifts in user preferences and data patterns
- **Continuous feedback loops:** Analyze interactions for targeted improvements
- **Granular observations:** Sometimes per-prediction monitoring vs standard application metrics

Observability

Observability = **gaining insights into ML model behavior & infrastructure.**

Enables Teams to:

- Quickly identify and diagnose issues
- Optimize resource usage
- Ensure compliance
- Monitor model performance and detect drift
- Track data quality and integrity

Feedback Loop:

- Continuous monitoring and retraining using real-world data
- Helps models adapt to user behavior, new data patterns, and emerging trends
- Drives better decision-making, user experience, and business value

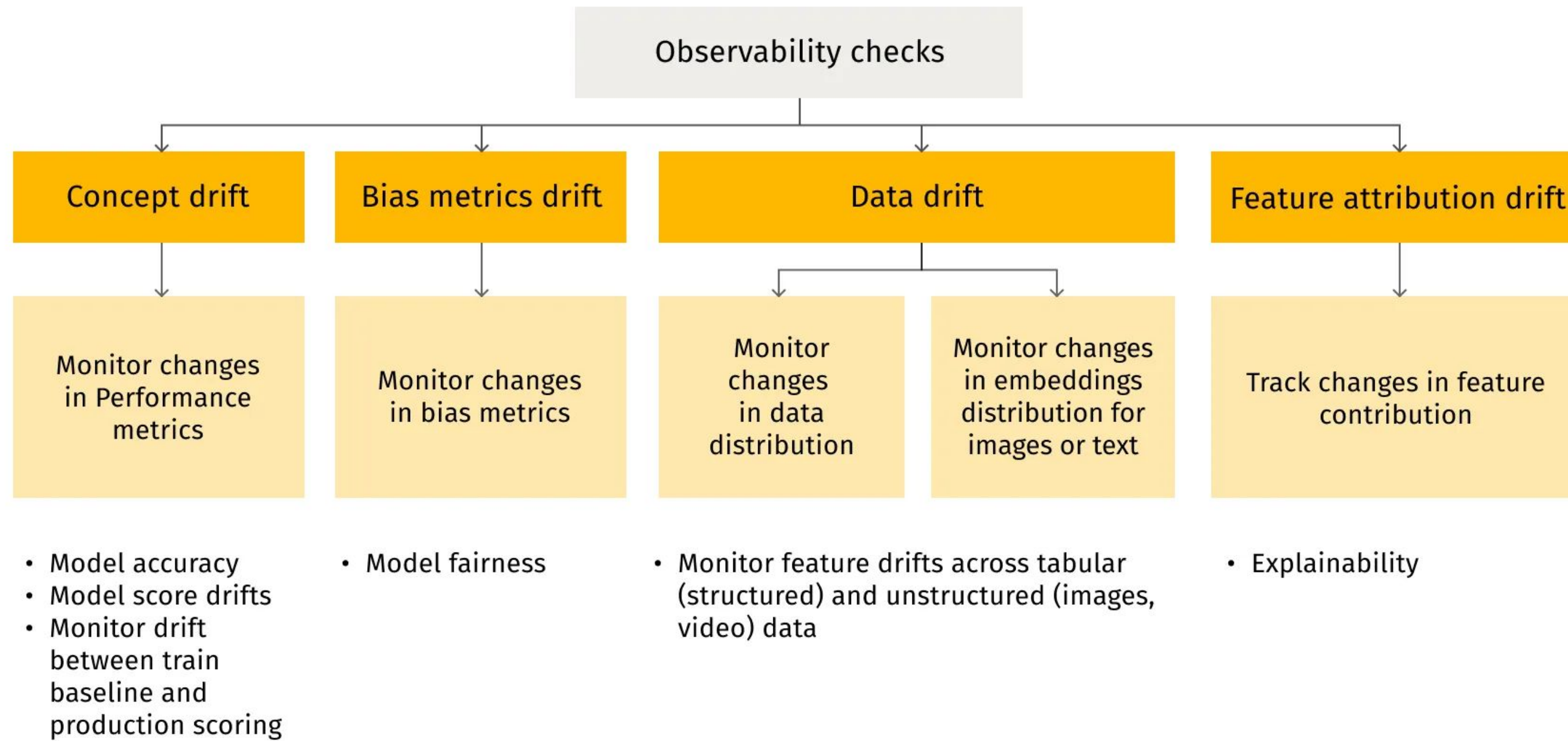
Your Pods Are
Probably Fine...
Probably.



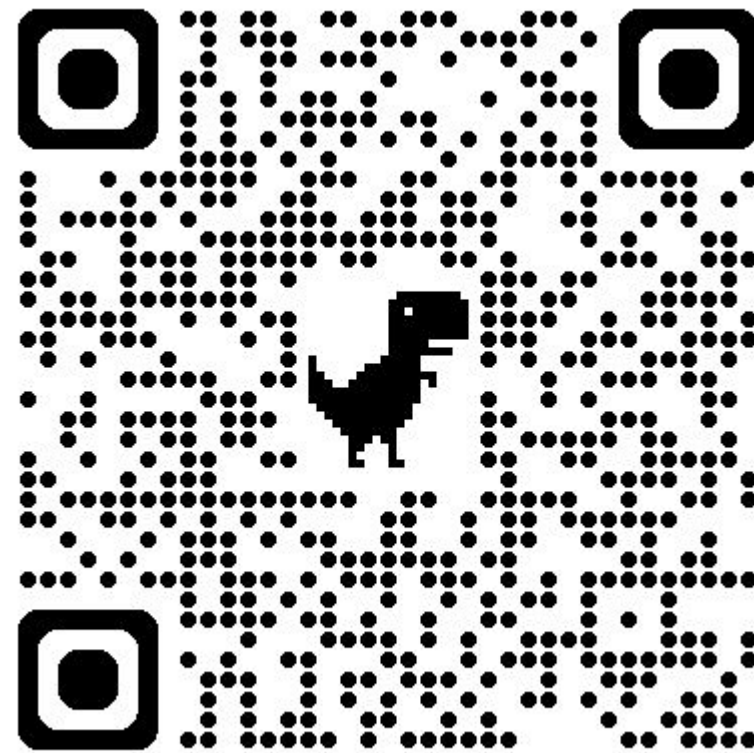
Source: <https://www.cloudlaunchpad.app/blog/observability-in-aws-eks>

The Need for ML Observability in MLOps

Pillar	What It Covers	AWS Services / Methods
Data Quality	Detect schema mismatches, cardinality shifts, out-of-range values; track distribution drift in features. (Grid Dynamics)	Use AWS SageMaker + Model Monitor; batch & streaming ingestion; baseline jobs. (Grid Dynamics)
Fairness / Bias	Pre- and post-training bias detection; monitoring predictions' distribution across sensitive groups. (Grid Dynamics)	AWS Clarify for bias metrics; set facets; integrate into model evaluation pipelines. (Grid Dynamics)
Explainability	Understanding which features drive predictions (global & local); detecting unjustified dependencies. (Grid Dynamics)	Methods like SHAP, LIME; use feature attribution drift jobs; visualize top features & heatmaps. (Grid Dynamics)
Model Performance / Drift	Monitoring model accuracy, recall, F1, etc.; detecting concept drift; comparing predictions vs ground truth. (Grid Dynamics)	AWS Model Monitor, performance / drift jobs; optionally use open-source libs like <i>nannyml</i> . (Grid Dynamics)



Let's Demo



Clusters state summary (3)

📅 As of May 13, 2024, 11:37 AM (UTC+01:00)

Clusters with alarms

⚠️ 0 In alarm ✅ 0 OK

Clusters without alarms

🔵 2 High utilization 🔵 1 Low utilization



Performance and status summary

📅 Last 1 min

Clusters CPU (avg)

Utilization 13%
Reserved 14%

Clusters Memory (avg)

Utilization 30%
Reserved 2%

Clusters GPU (avg)

Utilization 0%
Memory 0%

Clusters NeuronCore (avg)

Utilization 72%

Pods (sum)

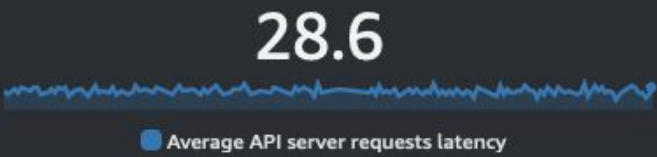
Desired 66
Ready 65

Nodes (sum)

Unavailable 0
Available 6

Control plane summary

📅 Last 3 hours



Top 10

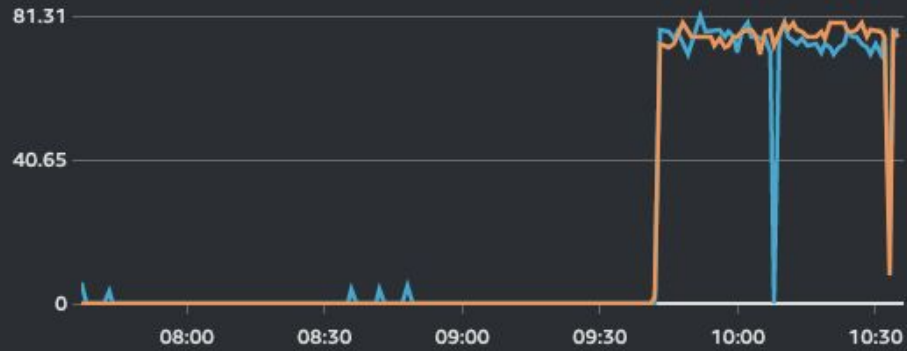
Nodes ▾

per metric

NeuronCore Utilization ▾

📅 Last 3 hours

Percent



1 - ip-192-168-78-145.us-west-2.comput...
2 - ip-192-168-70-170.us-west-2.comput...

Top 10

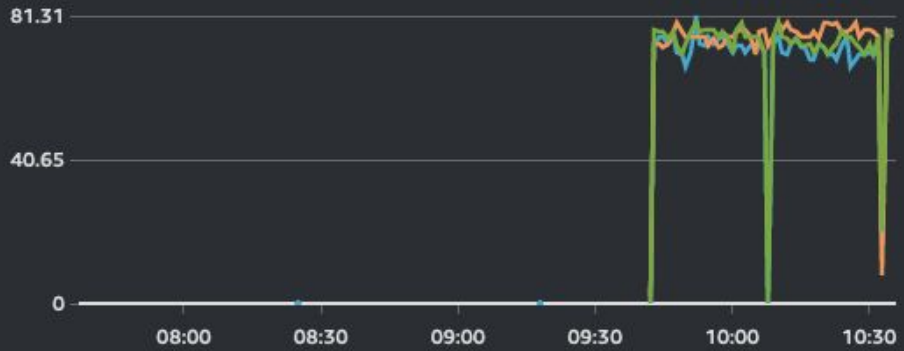
Containers ▾

per metric

NeuronCore Utilization ▾

📅 Last 3 hours

Percent



1 - neuron-sample neuron-sample-f48b5...
2 - role1-0 berttrain-d9gfxh055hxjsd-role...
3 - role1-1 berttrain-d9gfxh055hxjsd-role...

▼ Performance dashboard views Info

Clusters

No alarms detected

Namespaces

No alarms detected

Nodes

No alarms detected

Services

No alarms detected

Workloads

No alarms detected

Pods

No alarms detected

Containers

No alarms detected

More dashboard views

▼ Metrics aggregation

Neuron metrics aggregated by:

Container

NeuronCore

▼ Filters

By default only top 10 metrics are shown in the widgets on this dashboard. Use the filters to view a specific resource.

Cluster

my-trn1-cluster

Namespace

Select a namespace

Workload

berttrain-d9gfxh055hxjsd

Pod

Select a pod

Container

Select a container

Clear Applied Filters

Container EFA RDMA RX

Bytes/Second

1 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1

2 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0

Container EFA RDMA TX

Bytes/Second

1 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0

2 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1

Container EFA network RX dropped packets

Count/Second

1 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0

2 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1

Container EFA RDMA write RX bytes

Bytes/Second

1 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0

2 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1

NeuronCore Utilization (per NeuronCore)

Percent

7 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1 device15 core31

8 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0 device1 core3

9 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0 device10 core21

10 - berttrain-d9gfxh055hxjsd-role1-0-0 role1-0 device7 core14

NeuronCore Memory Usage (per NeuronCore)

MiB

1 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1 device7 core15

2 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1 device6 core13

3 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1 device4 core9

4 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1 device0 core0

5 - berttrain-d9gfxh055hxjsd-role1-1-0 role1-1 device2 core4

Pod status metrics

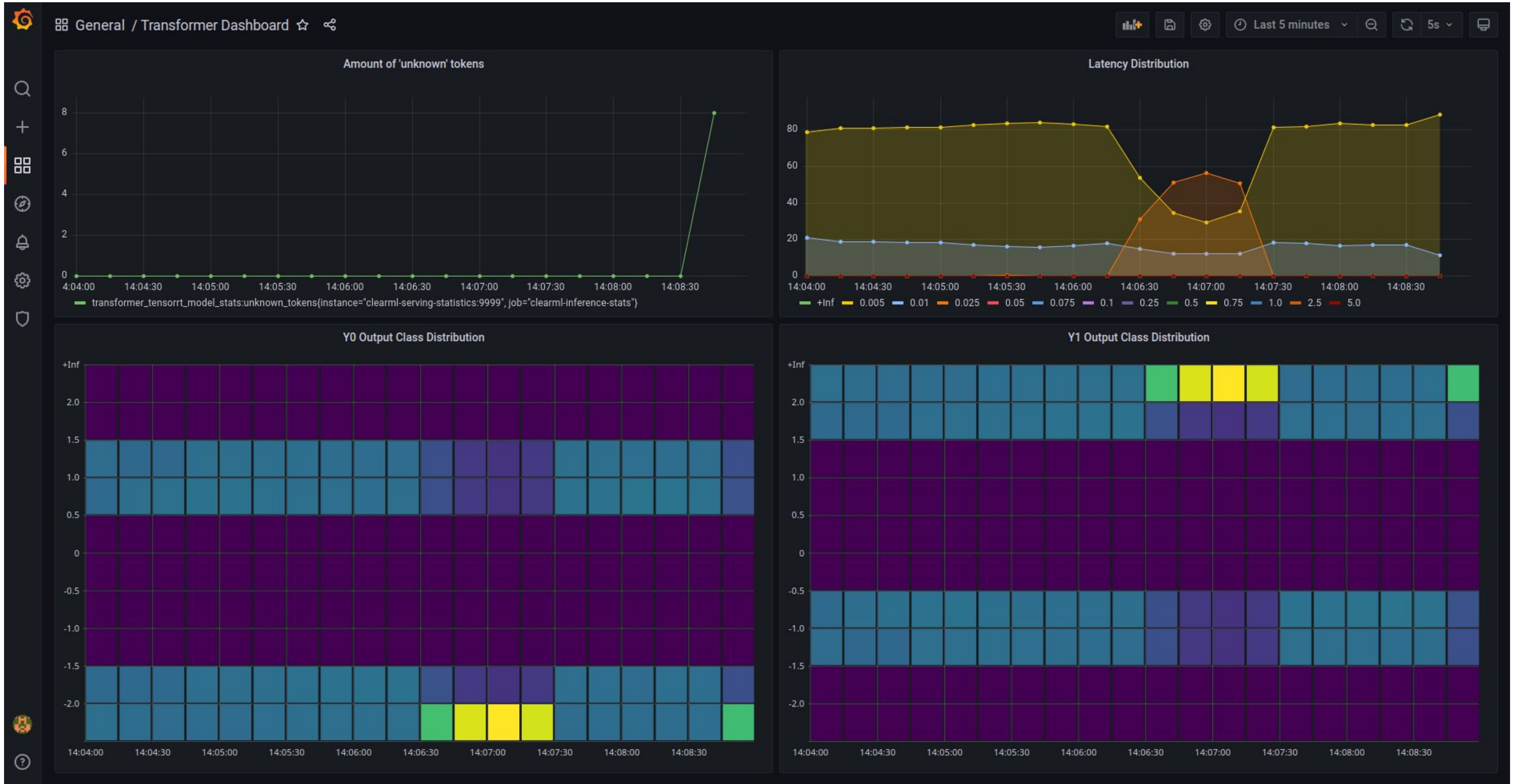
Containers running

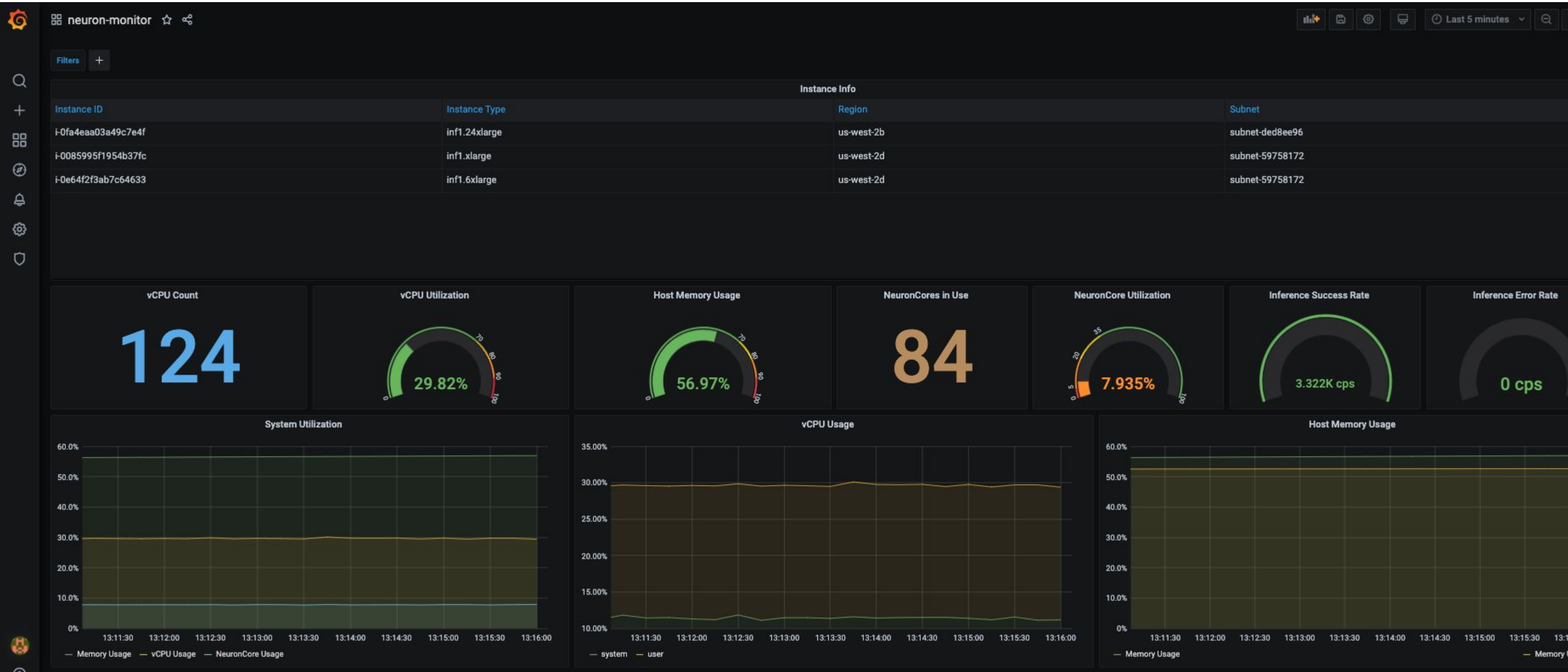
Count

Containers terminated

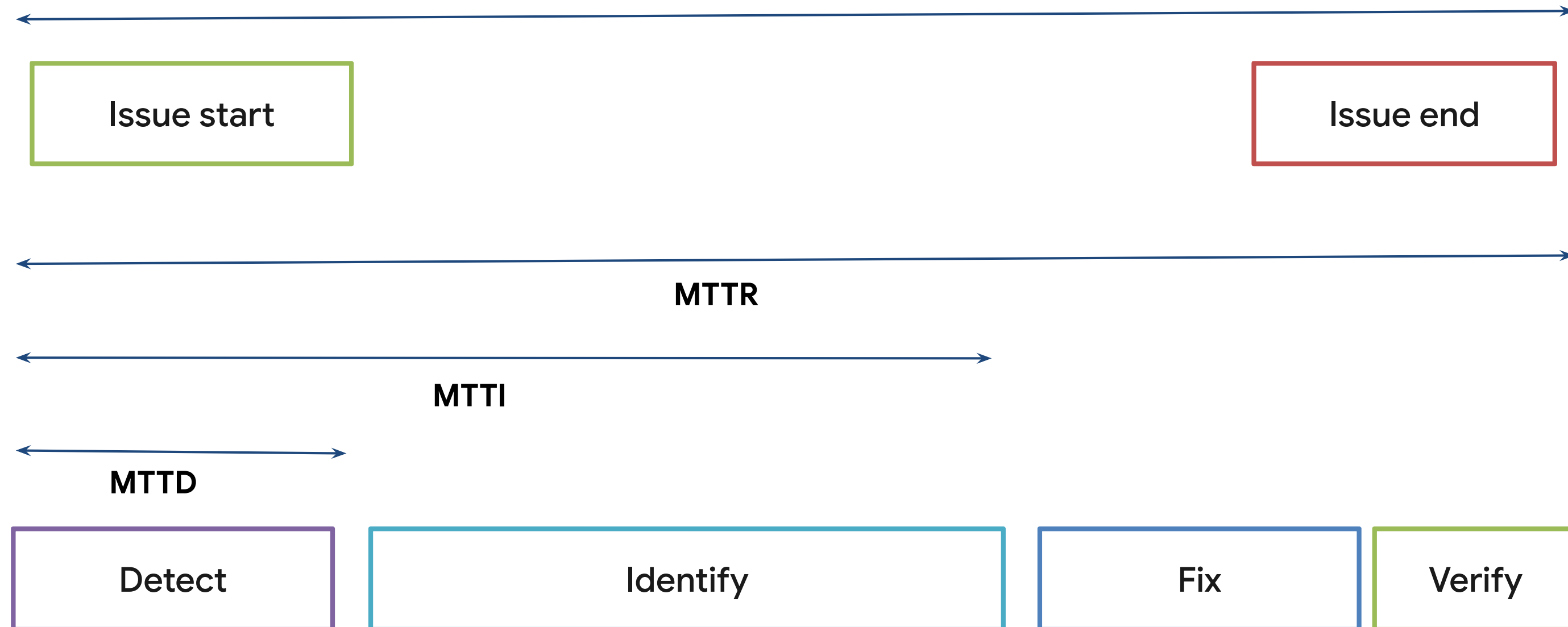
Count







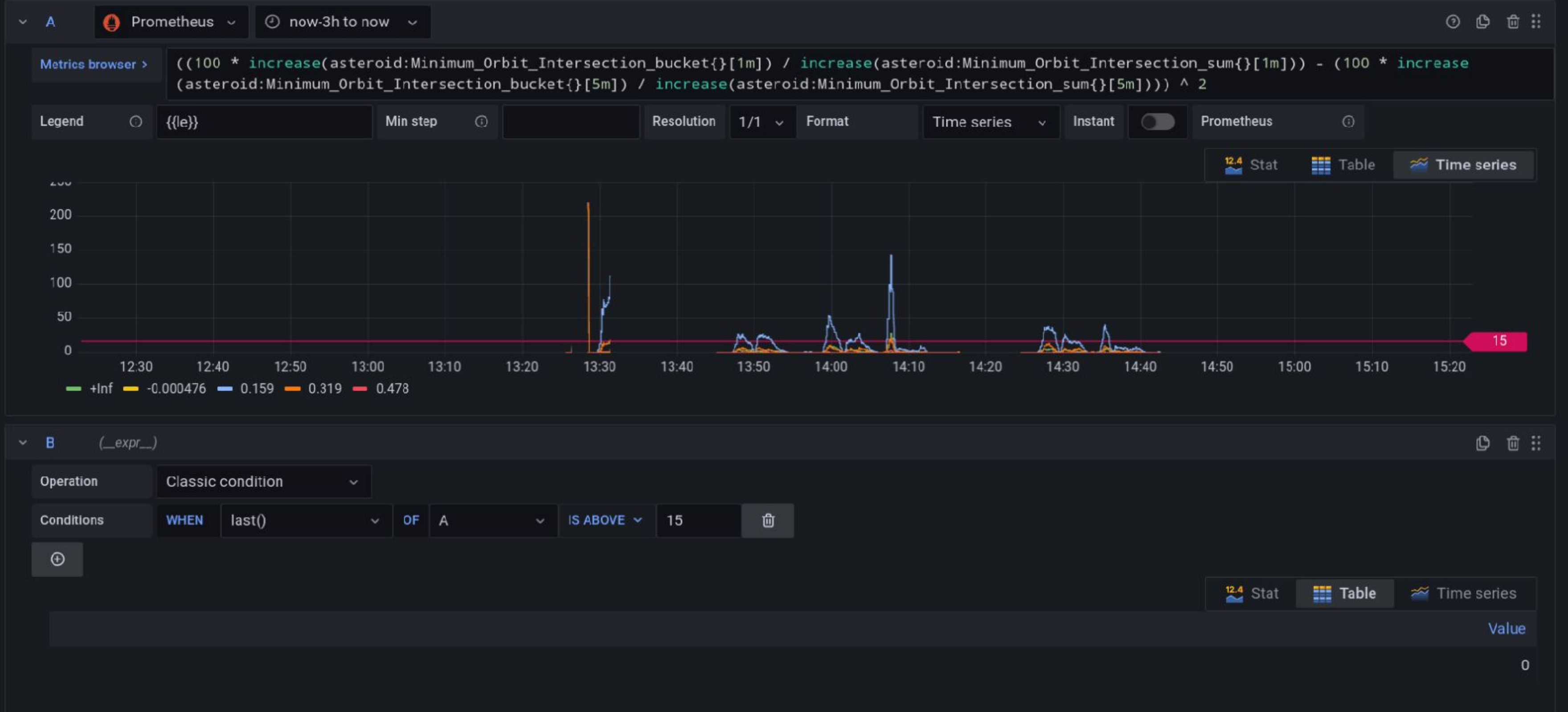
Issue Timeline



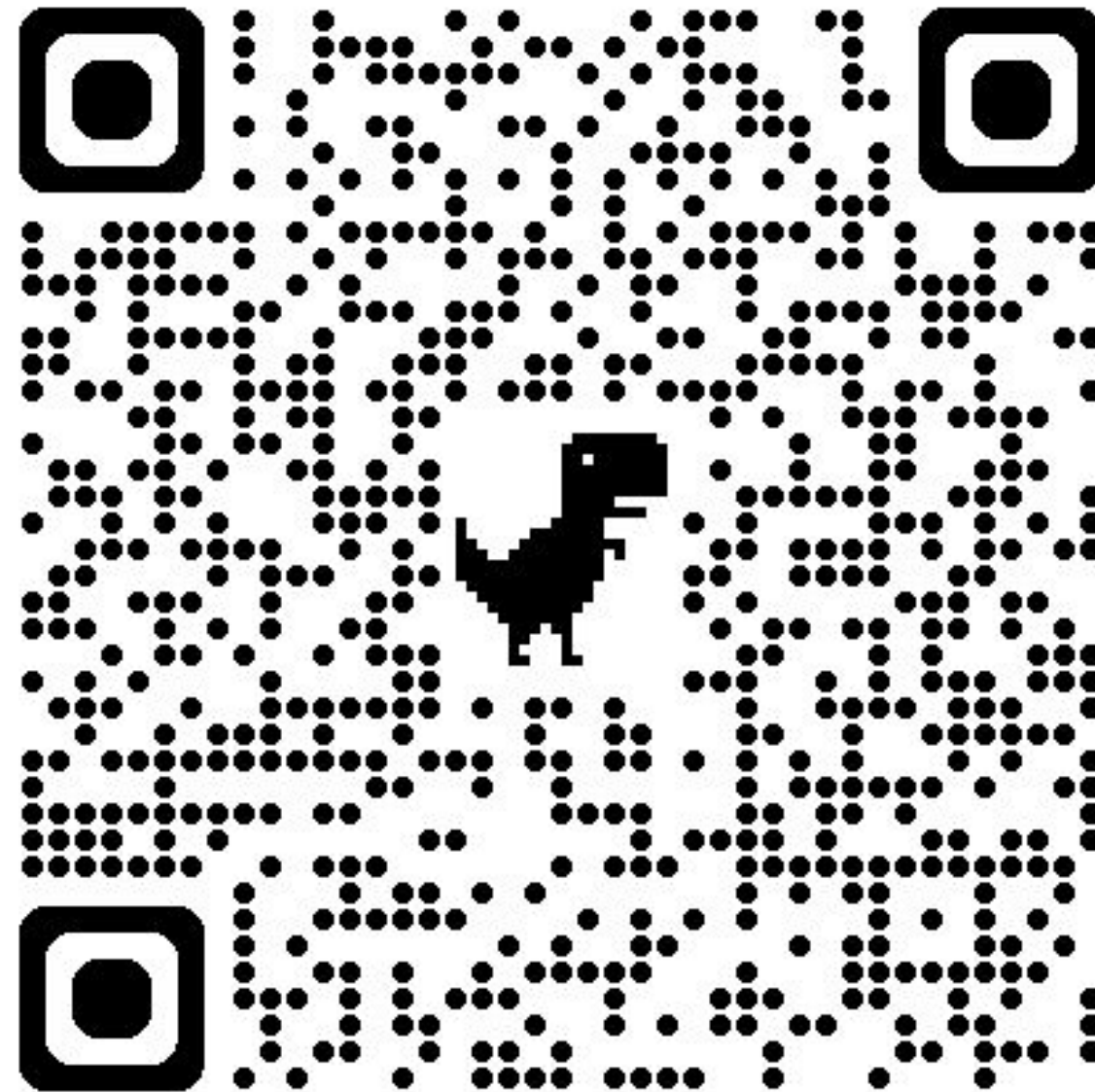
How to Measuring

- Changes made to monitoring configuration
- "Out of hours" alerts
- Team alerting balance
- False positives & negatives
- Alert creation
- Alert acknowledgement
- Alert silencing and silence duration
- Unactionable alerts
- Usability: alerts, runbooks, dashboards
- MTTD, MTTR, impact

2 Create a query to be alerted on



Feedback



Thank you

