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#KubeCon #CloudNativeCon

# Feast as a Cloud Native Data System

Bridging Batch Processing and Online Serving on Kubernetes

Nikhil Kathole,  
Red Hat



FEAST



# The Problem



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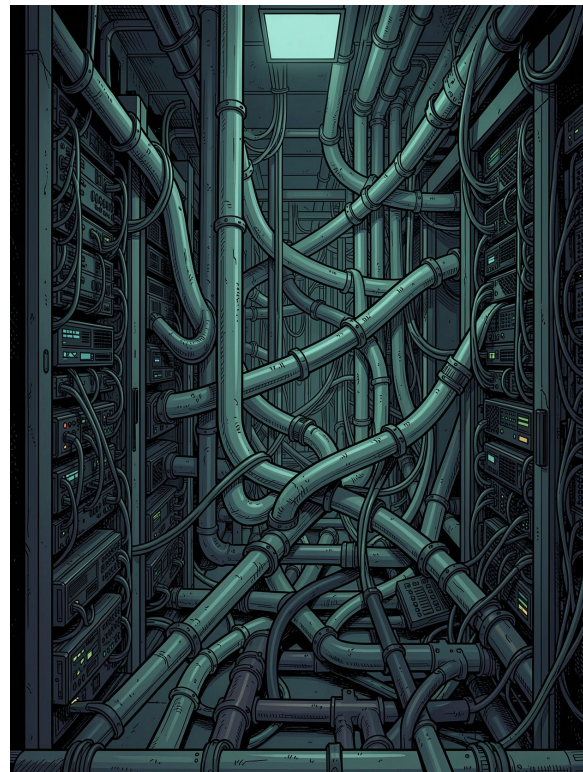
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**Your Lake Has the Data.**

**Your Model Needs It in Milliseconds.**

**Who owns the path between?**



# Why Production ML breaks



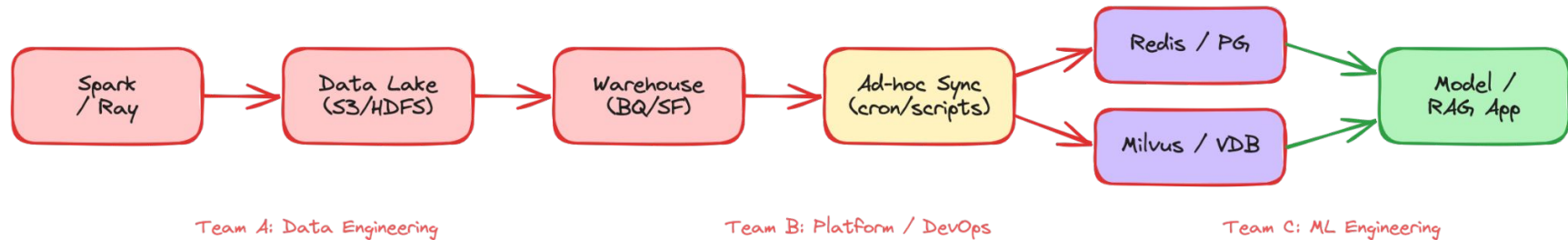
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## The Problem: Fragmented ML Data Infrastructure



Every arrow = a team, a runbook, a potential consistency bug

### Training-Serving Skew

Offline transforms  $\neq$  online transforms. Silent correctness failures in production.

### No Feature Reuse

Teams rebuild the same features for every new model. Zero discoverability.

### Duct-Tape Infrastructure

Data processing + lake + warehouse + cache + vector DB - stitched together by convention.

# What is Feature Store ?

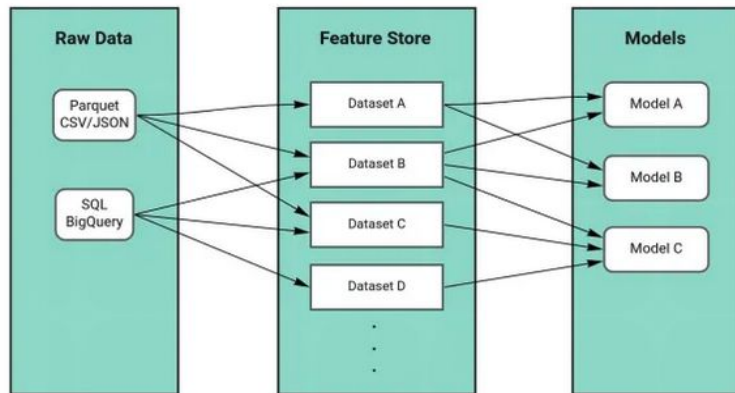


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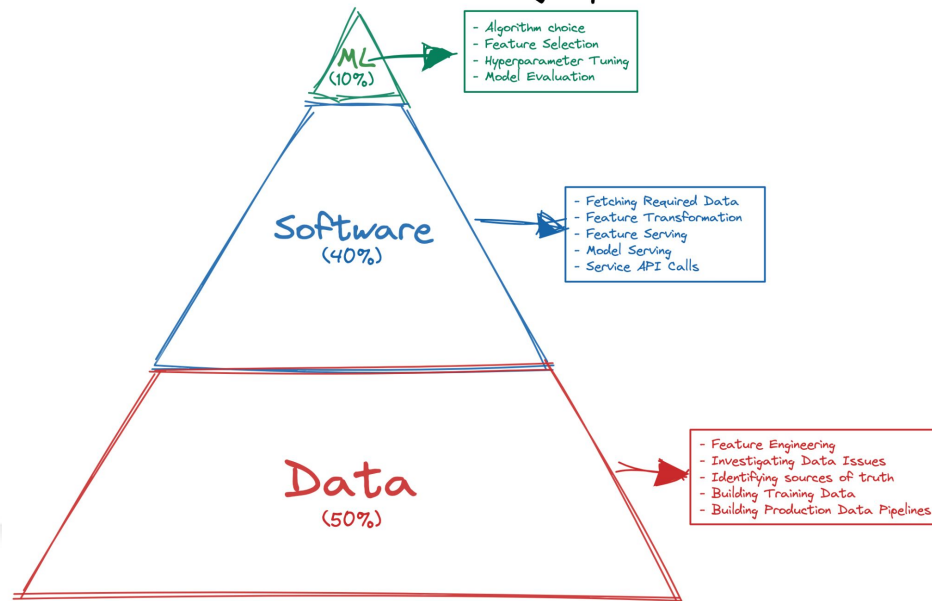
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Feature Store as a platform layer serving features to the machine learning models

## The Production Machine Learning Pyramid



# Define once. Train offline. Serve online.



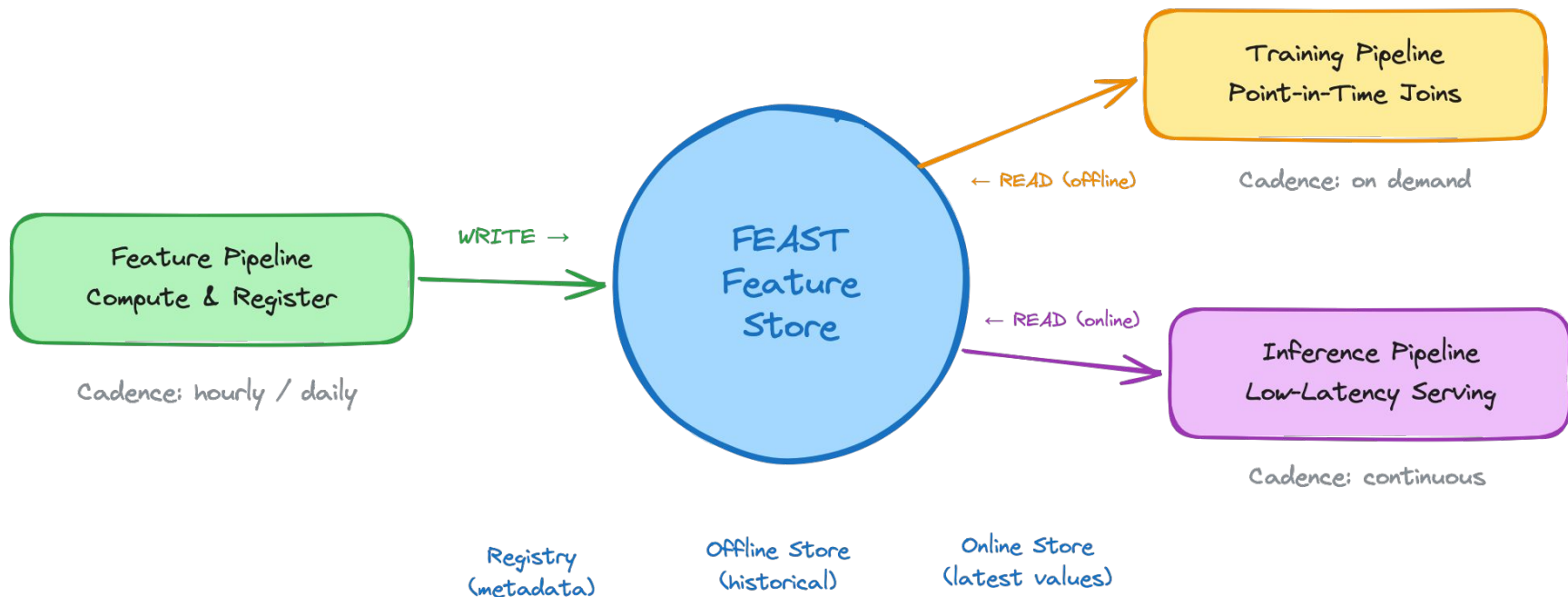
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## Feature · Training · Inference



# Feast: The Missing Data Layer

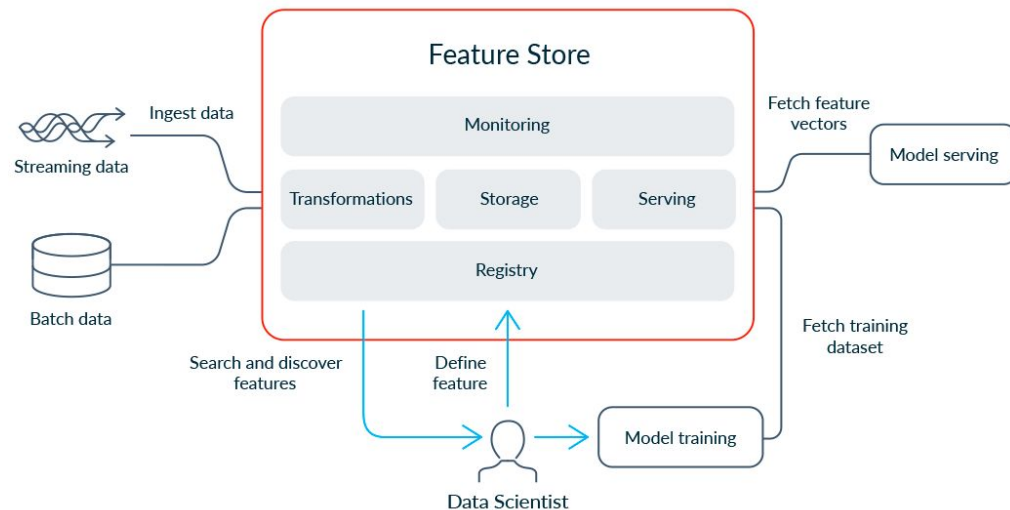


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- **Central Registry** - discover, version, and govern features
- **Dual storage** - offline for training, online for inference
- **Point-in-time correct joins** - no data leakage
- Materialization syncs batch compute to **low-latency serve**
- Build once, **reuse** across models and teams



FEAST

# Lakehouse/Warehouse Native

Do I need to migrate my data into Feast?

Feast integrates with what you already run.



# Feast Architecture

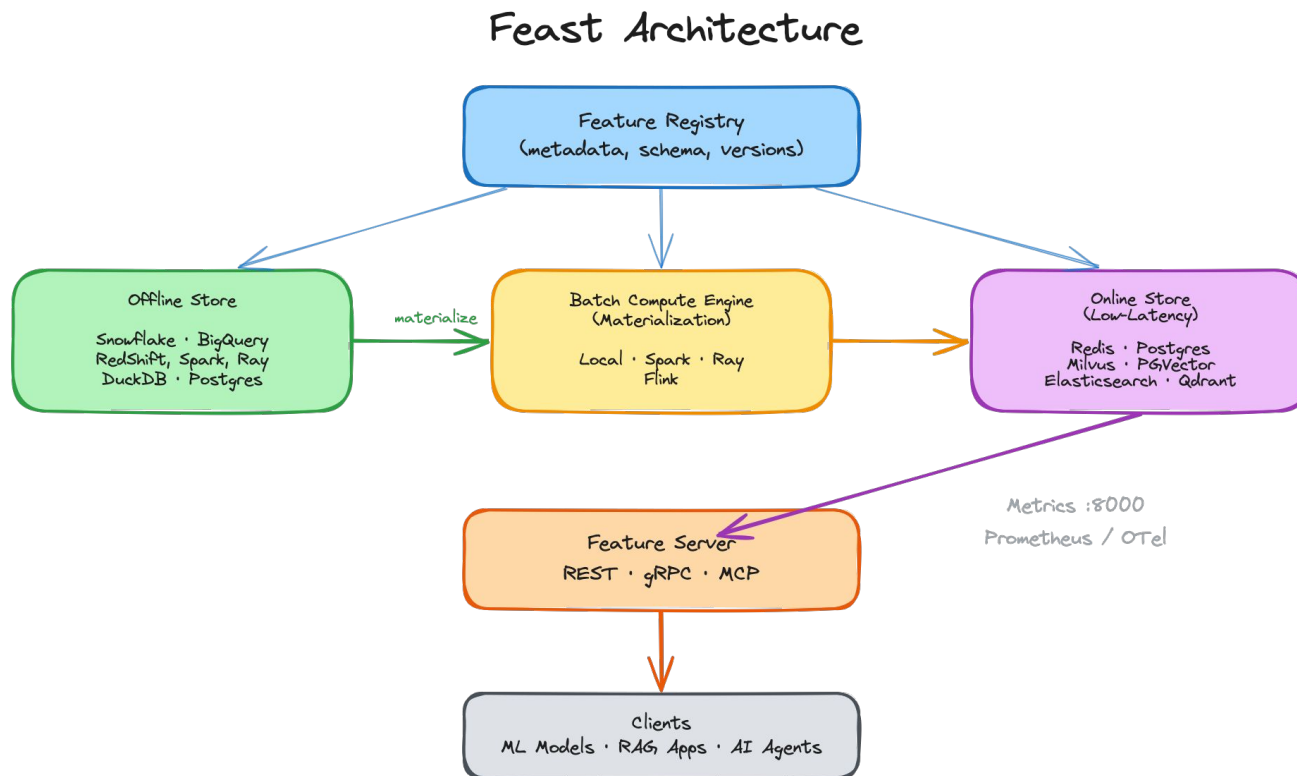


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# Feast in AI/ML Lifecycle

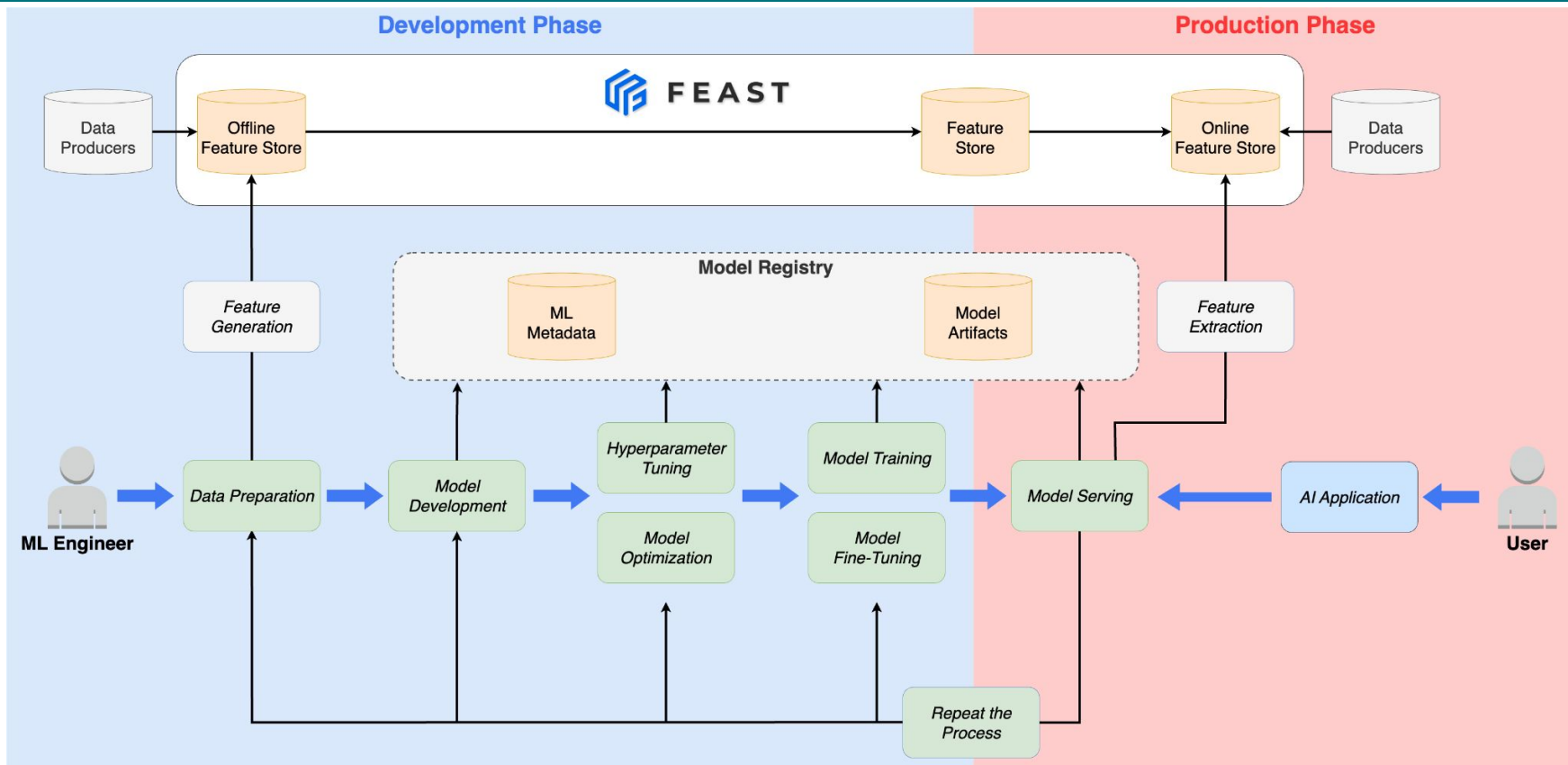


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```

from feast import FeatureStore

# Initialize the feature store
store = FeatureStore(repo_path="feature_repo")

# Get features for training
training_df = store.get_historical_features(
    entity_df=training_entities,
    features=[
        "customer_stats:daily_transactions",
        "customer_stats:lifetime_value",
        "product_features:price"
    ]
).to_df()

# Get online features for inference
features = store.get_online_features(
    features=[
        "customer_stats:daily_transactions",
        "customer_stats:lifetime_value",
        "product_features:price"
    ],
    entity_rows=[{"customer_id": "C123", "product_id": "P456"}]
).to_dict()

# Retrieve your documents using vector similarity search for RAG
features = store.retrieve_online_documents(
    features=[
        "corpus:document_id",
        "corpus:chunk_id",
        "corpus:chunk_text",
        "corpus:chunk_embedding",
    ],
    query="What is the biggest city in the USA?"
).to_dict()

```



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# Run it on Kubernetes

One CR | Full Stack | Day-2 Ready



# Batch to Online on Kubernetes



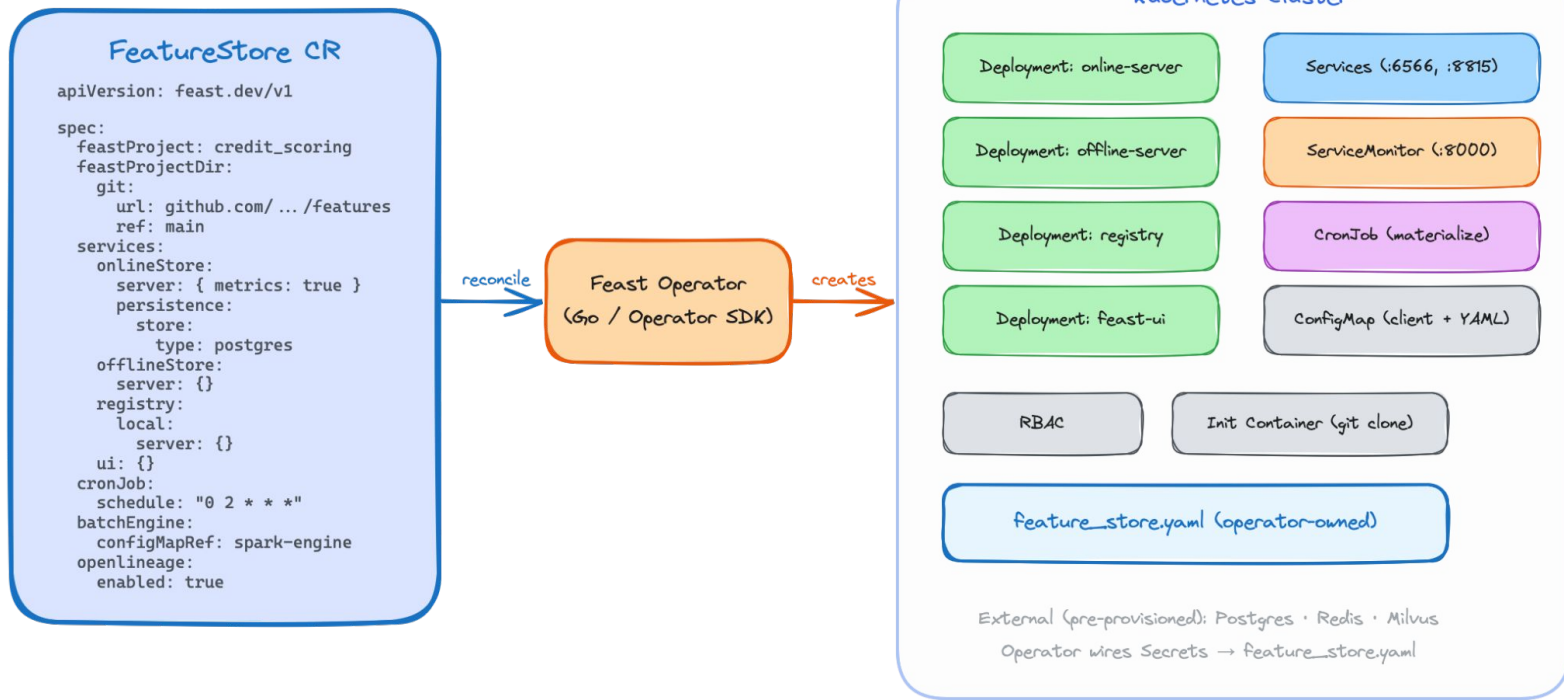
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## Feast Operator: One CR → Full Kubernetes Stack





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# Observe Everything

Metrics | Traces | Lineage



# If You Cannot See It, You Cannot Scale It



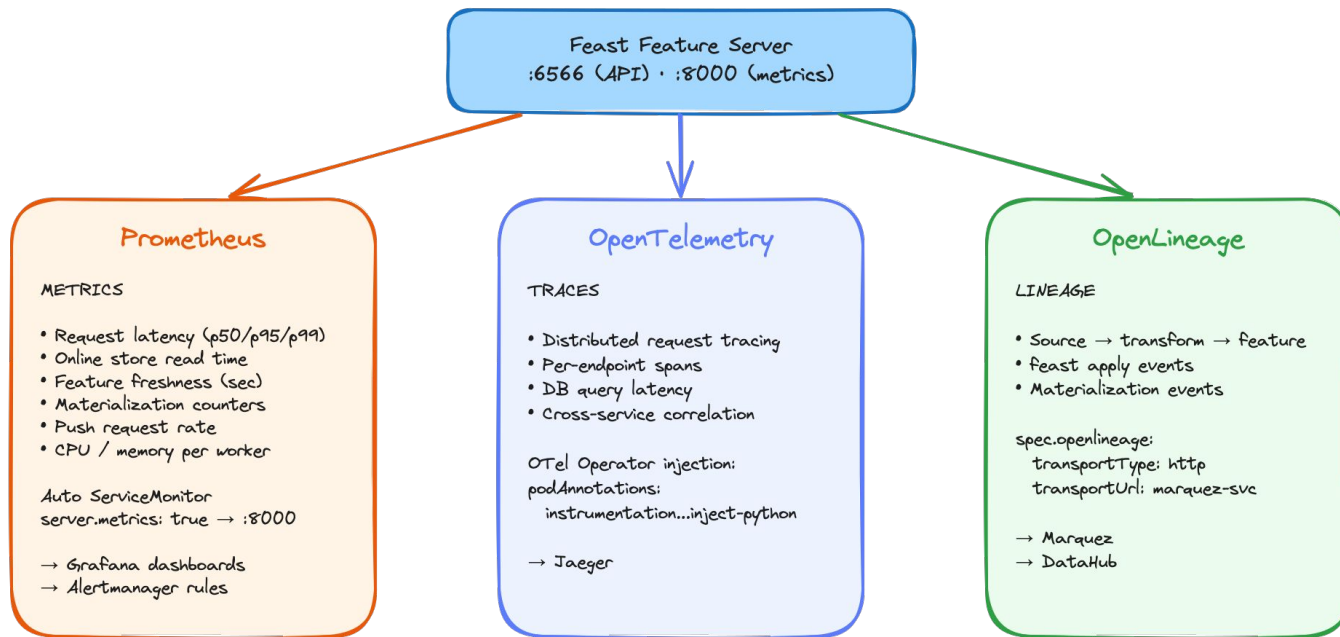
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Observability: Metrics · Traces · Lineage



All configured from the same FeatureStore CR

# Metrics

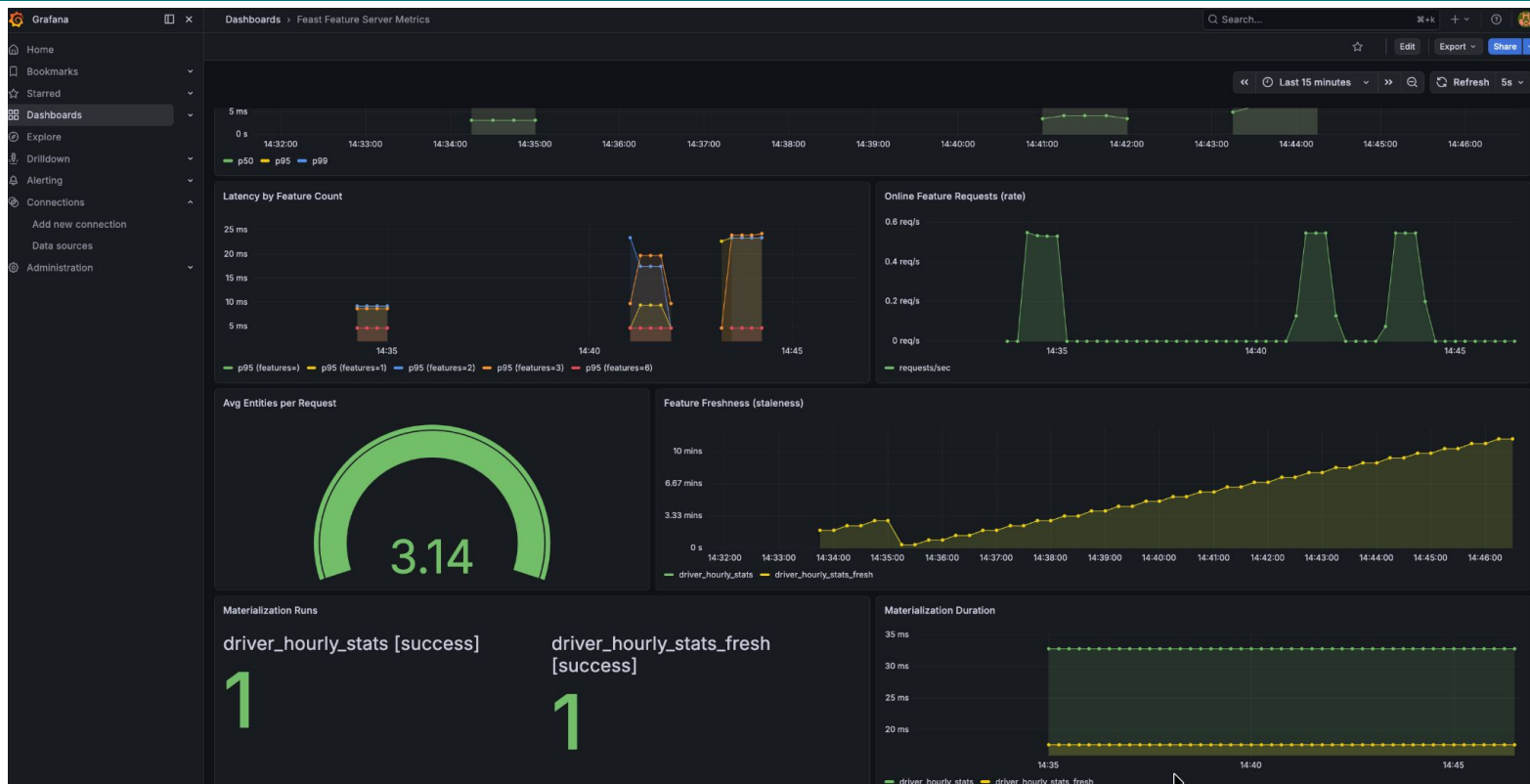


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# Traces



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JAEGER UI Search Compare System Architecture Monitor

Q Lookup by Trace ID...

About Jaeger

Search Upload

Service (3)

feast-test-online

Operation (6)

all

Tags

http.status\_code=200 error=true

Lookback

Last Hour

Max Duration

e.g. 1.2s, 100ms, 500us

Min Duration

e.g. 1.2s, 100ms, 500us

Limit Results

20

Find Traces



20 Traces

Sort: Most Recent

Download Results

Deep Dependency Graph

Compare traces by selecting result items

<input type="checkbox"/>	feast-test-online: GET f98fc2f	4.46ms
1 Span	feast-test-online (1)	Today 9:01:08 pm 2 minutes ago
<input type="checkbox"/>	feast-test-online: GET cf70242	4.43ms
1 Span	feast-test-online (1)	Today 9:01:08 pm 2 minutes ago
<input type="checkbox"/>	feast-test-online: GET 38c8795	4.37ms
1 Span	feast-test-online (1)	Today 9:01:08 pm 2 minutes ago
<input type="checkbox"/>	feast-test-online: GET 26dc01b	4.62ms
1 Span	feast-test-online (1)	Today 9:01:08 pm 2 minutes ago

# OpenLineage Integration

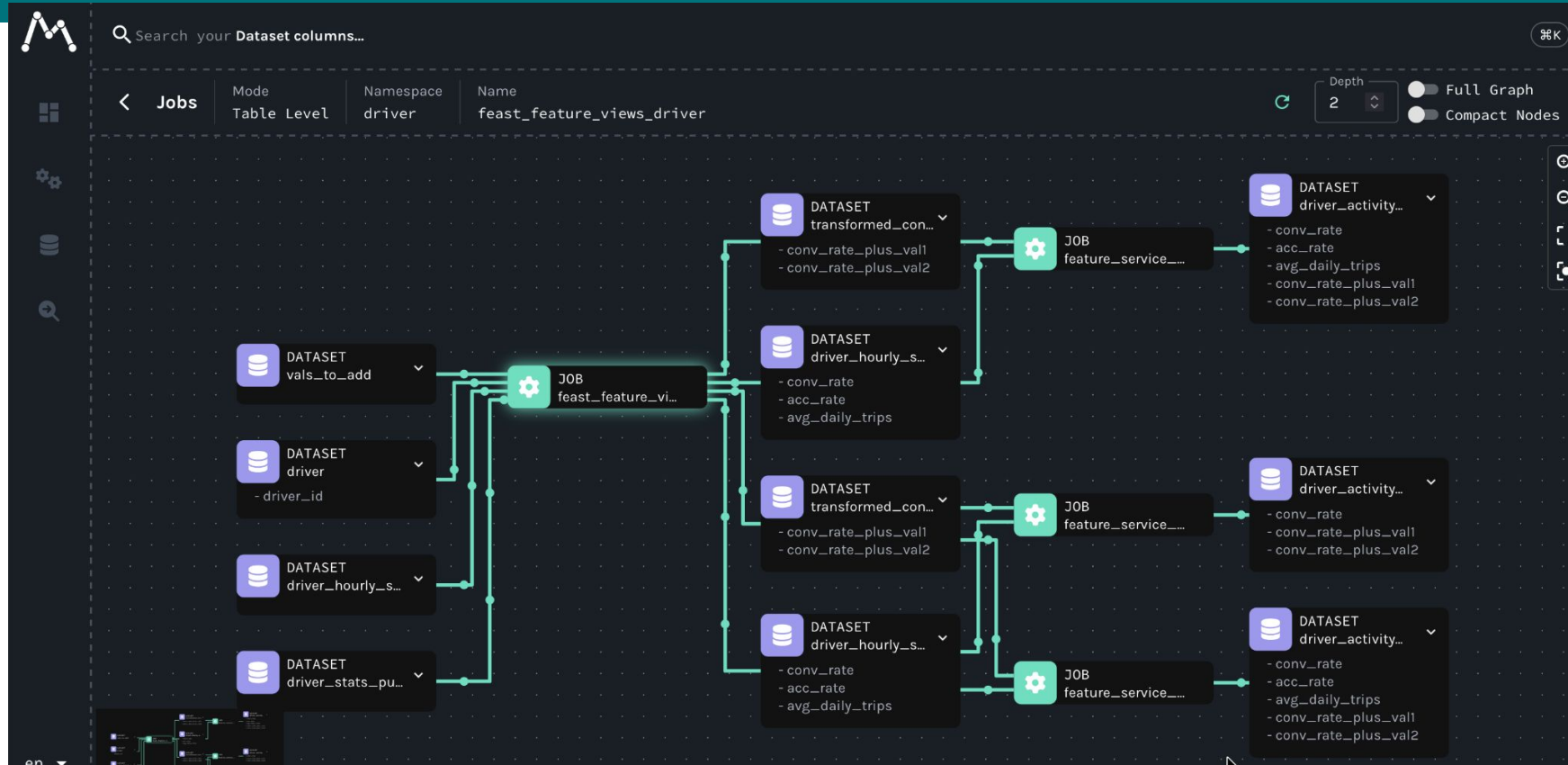


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# Scale Without Surprises

HPA | KEDA | PDB | HA



# Scaling & High Availability



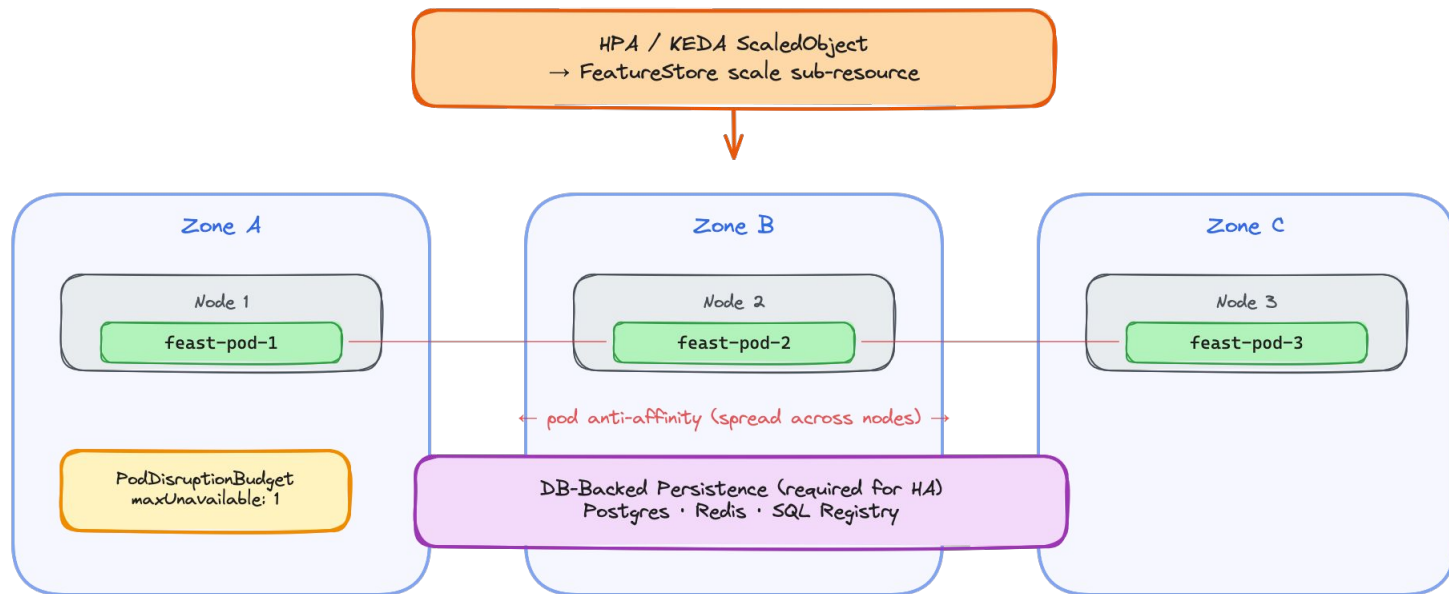
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## Scaling & High Availability



Auto-injected when replicas > 1:

- Pod anti-affinity (prefer spread across nodes)
- Zone topology spread (distribute across AZs)
- RollingUpdate strategy (zero-downtime)
- CEL validation rejects file-based stores



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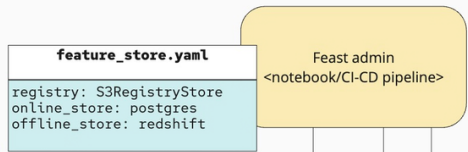
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# RBAC (Role based access control)

Governance | Multi-tenancy | Isolation



## Management layer

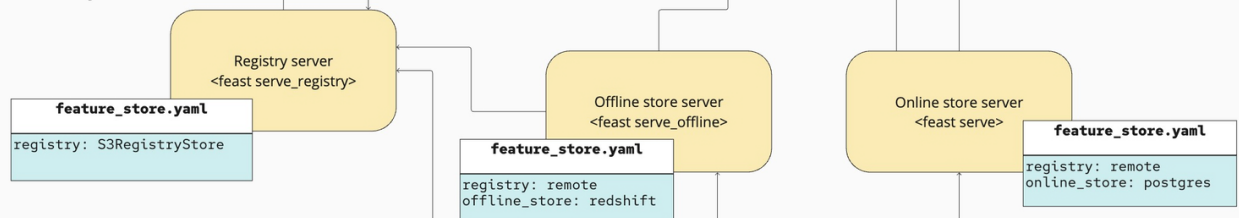


## Data layer

with configured permissions

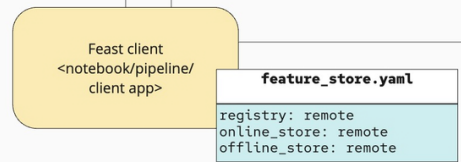


## Service layer



## Security layer

## Client layer



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# Compute Engine

Ray | Spark | Flink



# Compute Engines: DAG Execution



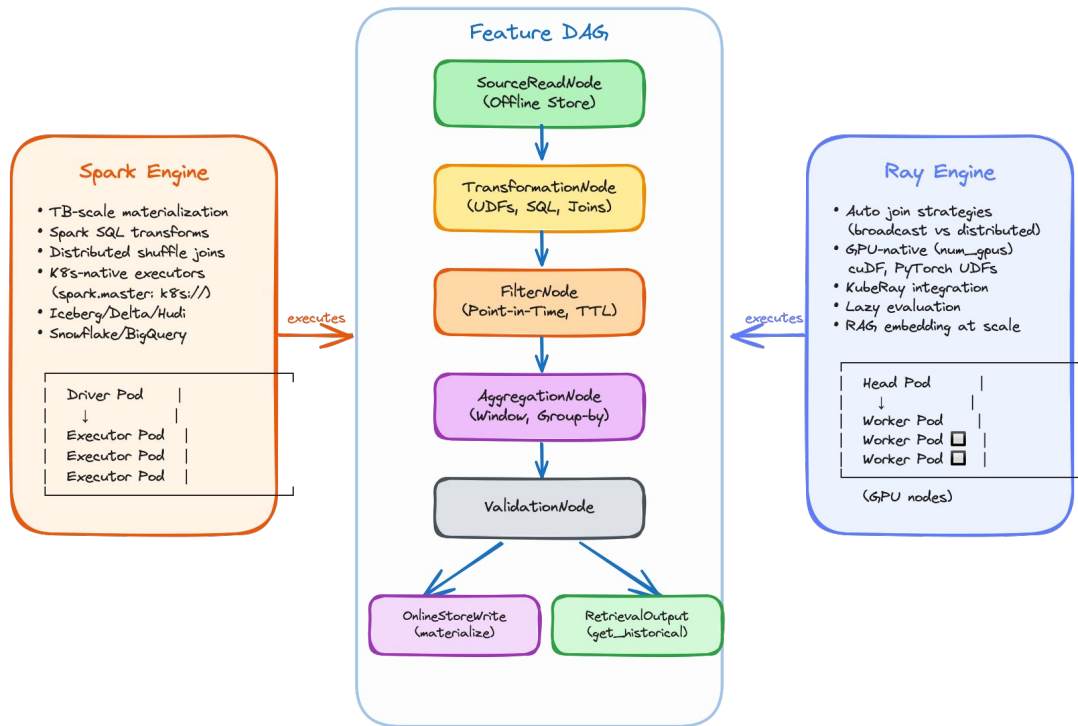
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## Compute Engine: DAG-Based Feature Transformations



Same DAG. Same FeatureViews. Different execution backend.

# Compute Engines: Spark & Ray



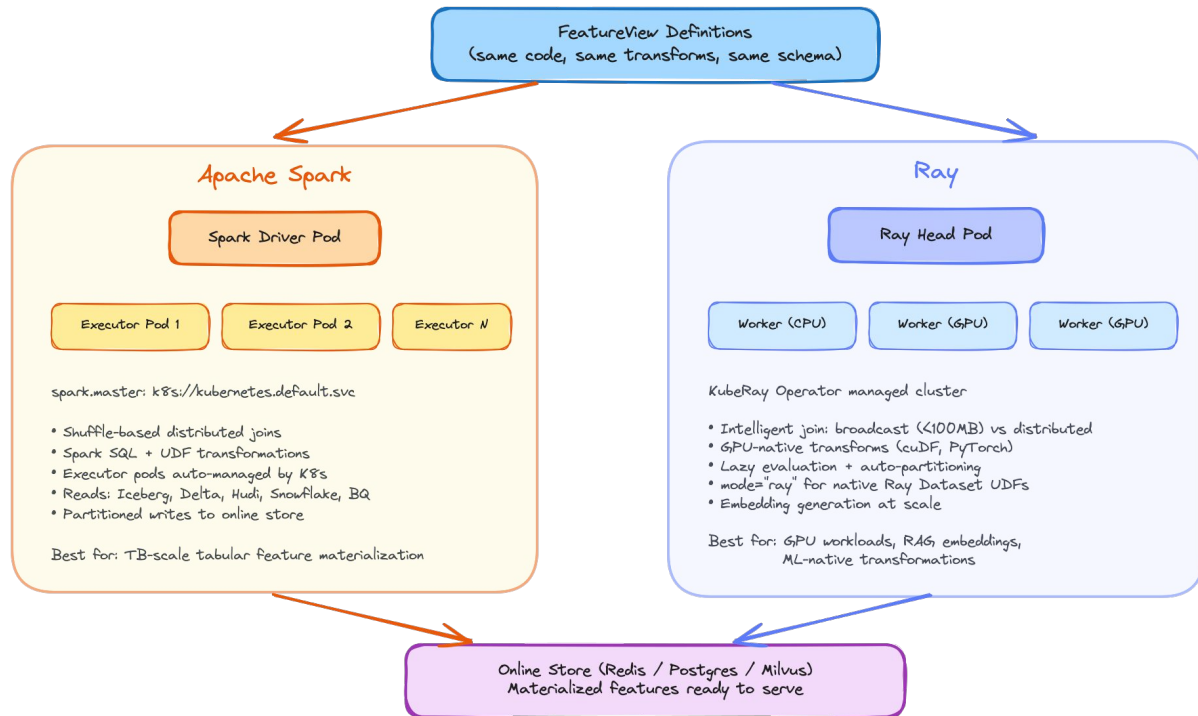
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## Spark vs Ray: Kubernetes Execution Model



Both converge: same Online Store, same Feature Server, same observability



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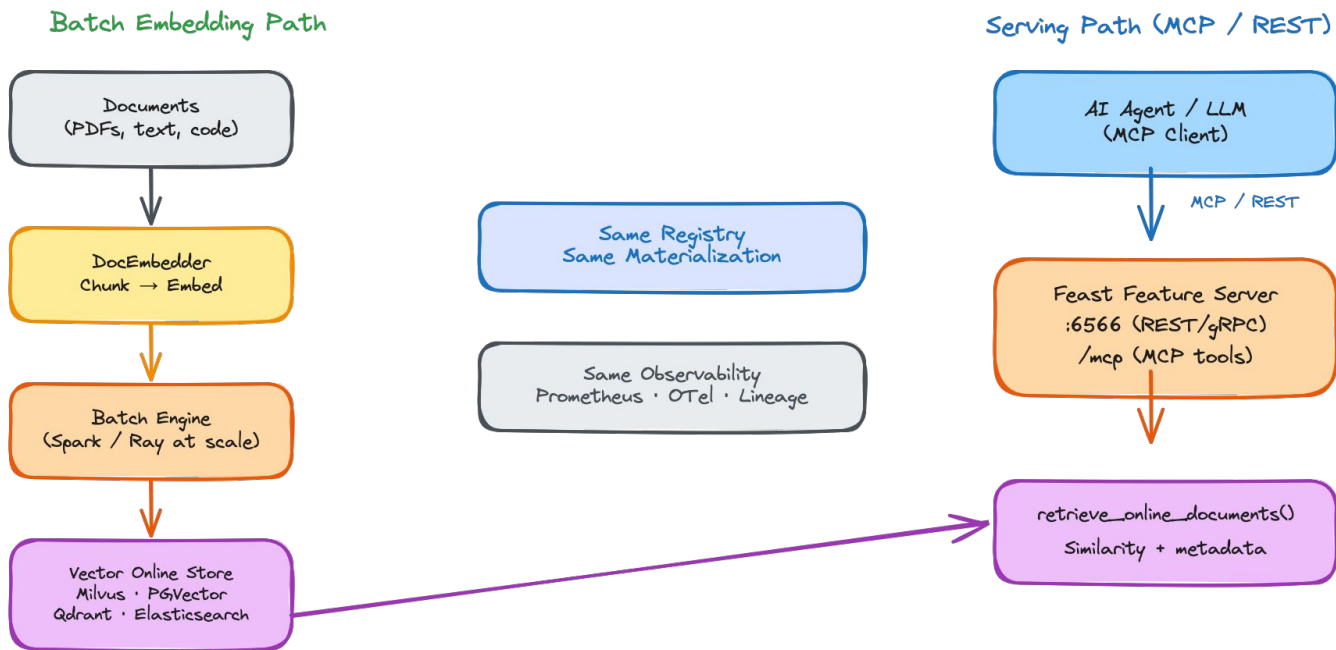
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# GenAI & RAG

Batch ML | Real-time ML | RAG | One stack

## GenAI, RAG & MCP on Feast



Embeddings are features. Agents are clients. One operational model.

# Feast in Open Data Hub

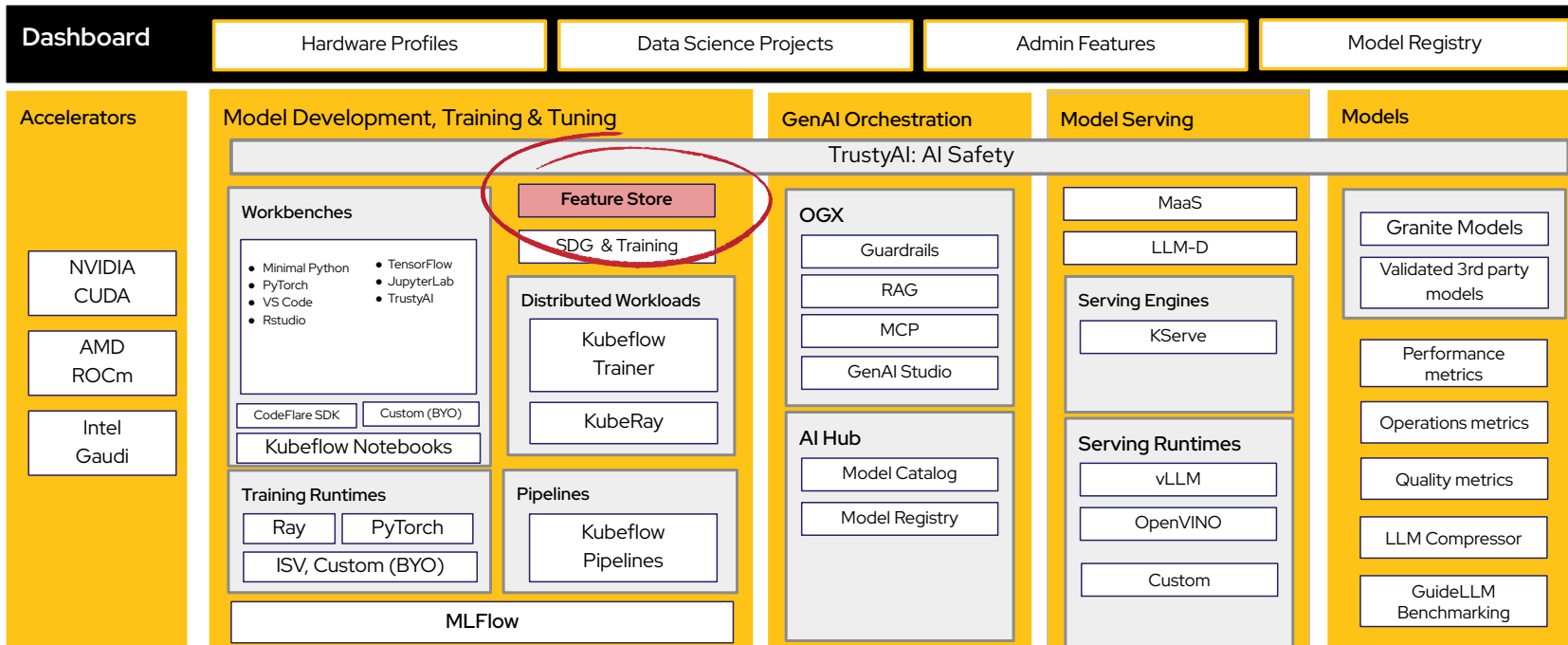


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# Feast in Open Data Hub

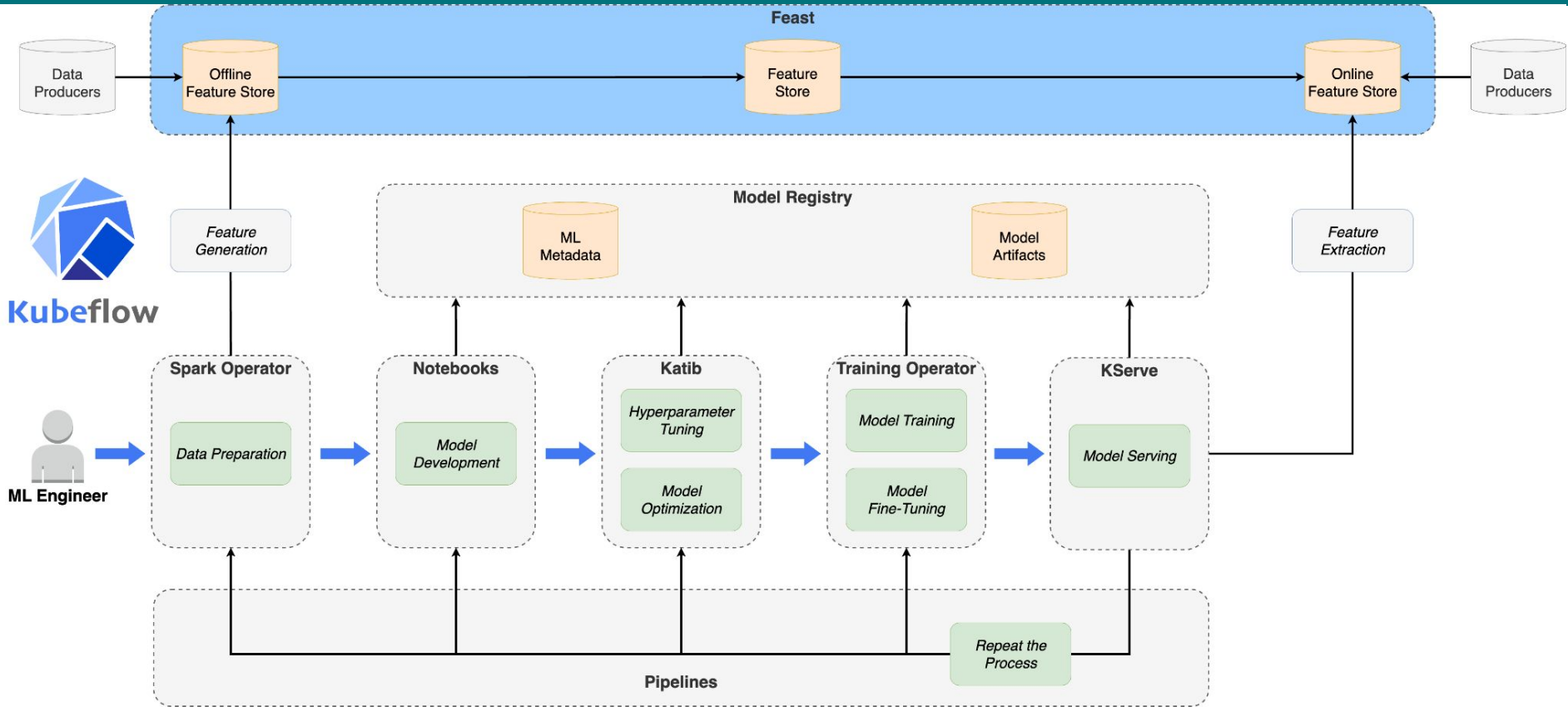


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# Feast in Open Data Hub



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- Home
- Projects
- AI hub
- Develop & train
- Feature store
  - Overview
  - Entities
  - Data sources
  - Datasets
  - Features
  - Feature views
  - Feature services
- Pipelines
- Jobs
- Learning resources
- Applications

## Feature store overview

🔍 Search by name, description, or tag (Example: team=platform)

Feature stores are shared catalogs for defining and managing features across teams. They help ensure consistent feature values from training to production and reduce duplicated feature engineering. Connect your workbenches to use and manage features in your projects. [How to connect workbenches.](#)

Feature store All feature stores ▾ [Manage feature stores](#) [View connected workbenches](#)

Metrics Lineage



### Entities

Entities are collections of related features and can be mapped to the domain of your use case.

9

[Go to Entities](#)



### Data sources

Data sources such as tables or data warehouses contain the raw data from which features are extracted.

10

[Go to Data sources](#)



### Datasets

Datasets are point-in-time-correct snapshots of feature data used for training or validation.

9

[Go to Datasets](#)



### Features

A feature is a single data value used in model training or inference.

118

[Go to Features](#)



### Feature views

Feature views define groups of related features and how to retrieve them from a source.

26



### Feature services

Feature services are groups of related features from one or more feature views that are designed to be retrieved together...

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# Monitoring



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Q Search across Feature Views, Features, Entities, etc. 36K

Project ▼

Home

Resources

- Lineage
- Data Sources (2)
- Entities (3)
- Features (15)
- Feature Views (2)
- Feature Services (2)
- Label Views
- Datasets (0)
- Permissions

Monitoring

Feature View	Granularity	Source	Start Date	End Date	Refresh
All Feature Views	<input checked="" type="checkbox"/> Baseline Daily Weekly Biweekly Monthly Quarterly	All Sources	mm / dd / yyyy N/A for baseline	mm / dd / yyyy N/A for baseline	<a href="#">Refresh</a>

Feature	Feature ...	Type	Distribution	Rows	Null Rate	Health	Mean	Std Dev	Freshness	Source
credit_card_due	credit_history	numeric		364,000	0.0%	Healthy	8489.12	2675.19	13h ago	<a href="#">batch</a>
mortgage_due	credit_history	numeric		364,000	0.0%	Healthy	164986.06	25039.63	13h ago	<a href="#">batch</a>
student_loan_due	credit_history	numeric		364,000	0.0%	Healthy	5985.44	2493.39	13h ago	<a href="#">batch</a>
vehicle_loan_due	credit_history	numeric		364,000	0.0%	Healthy	20004.08	4165.66	13h ago	<a href="#">batch</a>
hard_pulls	credit_history	numeric		364,000	0.0%	Healthy	4.00	2.31	13h ago	<a href="#">batch</a>
missed_payments_2y	credit_history	numeric		364,000	0.0%	Healthy	1.50	1.22	13h ago	<a href="#">batch</a>
missed_payments_1y	credit_history	numeric		364,000	0.0%	Healthy	1.49	1.49	13h ago	<a href="#">batch</a>
missed_payments_6m	credit_history	numeric		364,000	0.0%	Healthy	1.61	1.88	13h ago	<a href="#">batch</a>
bankruptcies	credit_history	numeric		364,000	0.0%	Healthy	0.03	0.17	13h ago	<a href="#">batch</a>
city	zipcode_features	categorical		30,000	0.0%	Healthy	—	—	13h ago	<a href="#">batch</a>

# Monitoring



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Search across Feature Views, Features, Entities, etc.

36K

Project

Home

Resources

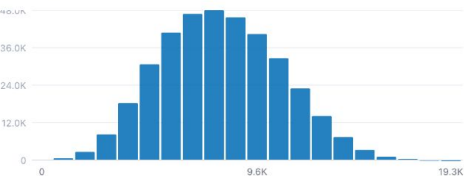
- Lineage
- Data Sources (2)
- Entities (3)
- Features (15)
- Feature Views (2)
- Feature Services (2)
- Label Views
- Datasets (0)
- Permissions

Monitoring

Monitoring / credit\_history / credit\_card\_due

View baseline

Distribution



Statistics

numeric

Row Count

364,000

Null Rate

0.00%

Mean

8489.1213

Std Dev

2675.1948

Min / Max

0 / 19,288

Percentiles

P50: 8,395 | P75: 10,371 | P90: 12,057 | P95: 13,014 |

P99: 14710.0100

Data Source

batch

Granularity

baseline

Computed At

6/12/2026, 6:44:55 PM

Time-Series Analysis

Historical trends for central aggregates and quality signals.

All time

Aggregate Metrics Drift (Mean/P50/P95)



# Monitoring



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Search across Feature Views, Features, Entities, etc.

16K

Project

Home

Resources

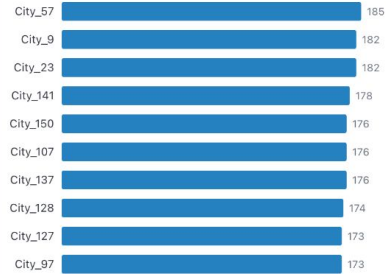
- Lineage
- Data Sources (2)
- Entities (3)
- Features (15)
- Feature Views (2)
- Feature Services (2)
- Label Views
- Datasets (0)
- Permissions

Monitoring

Monitoring / zipcode\_features / city

View baseline

### Category Distribution



200 unique values (28,225 in other categories)

### Statistics

categorical

#### Row Count

30,000

#### Null Rate

0.00%

#### Data Source

batch

#### Granularity

baseline

#### Computed At

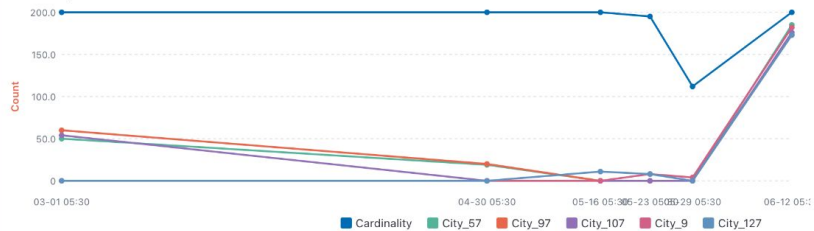
6/12/2026, 6:44:55 PM

### Time-Series Analysis

Historical trends for central aggregates and quality signals.

All time

#### Cardinality over time



# Native MLFlow integration



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Filter by type

All

Select object

All

Filter by permissions

All

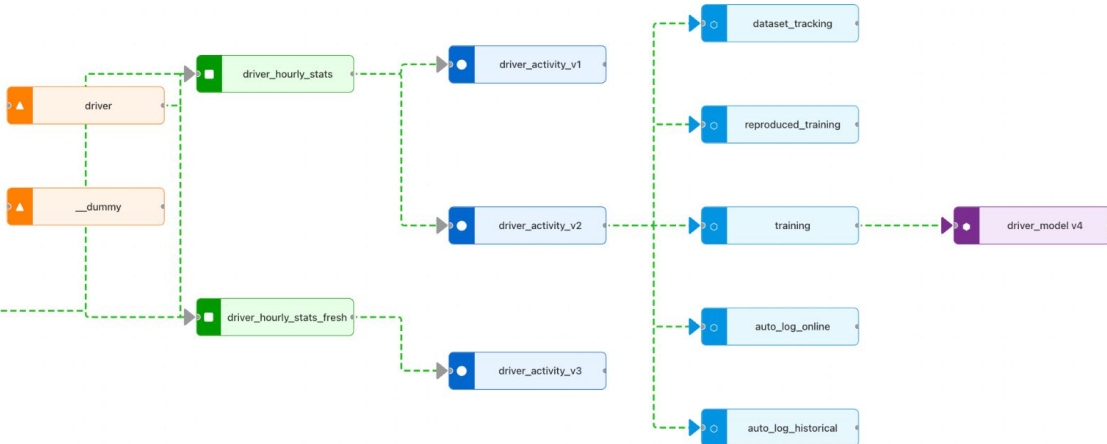
## Lineage

Show Indirect Relationships  Show Objects Without Relationships

### Legend

- Feature Service
- Feature View
- Entity
- Data Source
- MLflow Run
- Registered Model
- vN Version Changed

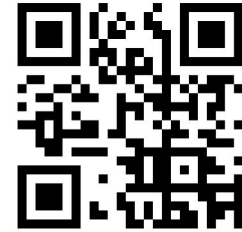
vN Version Changed



# Resources

- <https://feast.dev/>
- <https://github.com/feast-dev/feast>
- <https://opendatahub.io/>
- <https://docs.feast.dev/master/how-to-guides/production-deployment-topologies>
- **Join us at** [feastopen-source.slack.com](https://feastopen-source.slack.com)

Learn More:



<https://feast.dev>

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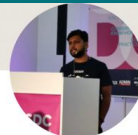
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# Thank You!

## Questions?



**Nikhil Kathole**

Principal Software Engineer, Red Hat

