

mlflow

Platform for Machine Learning Lifecycle

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Outline – Introduction to MLflow: Model Registry Workflows Explained – Module 4

- Model Registry
- Concepts and Motivations
 - MLflow Model Registry
 - Model Registry UI & API Workflow
 - Tutorials on local host
 - Jupyter Lab
- Q & A

<https://github.com/dmatrix/olt-mlflow>

MLflow Components

mlflow Tracking

Record and query experiments: code, data, config, and results

mlflow Projects

Package data science code in a format that enables reproducible runs on any platform

mlflow Models

Deploy machine learning models in diverse serving environments

mlflow Model Registry

Store, annotate and manage models in a central repository

databricks.com/mlflow



mlflow.org



github.com/mlflow



twitter.com/MLflow

databricks

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The Model Management Problem

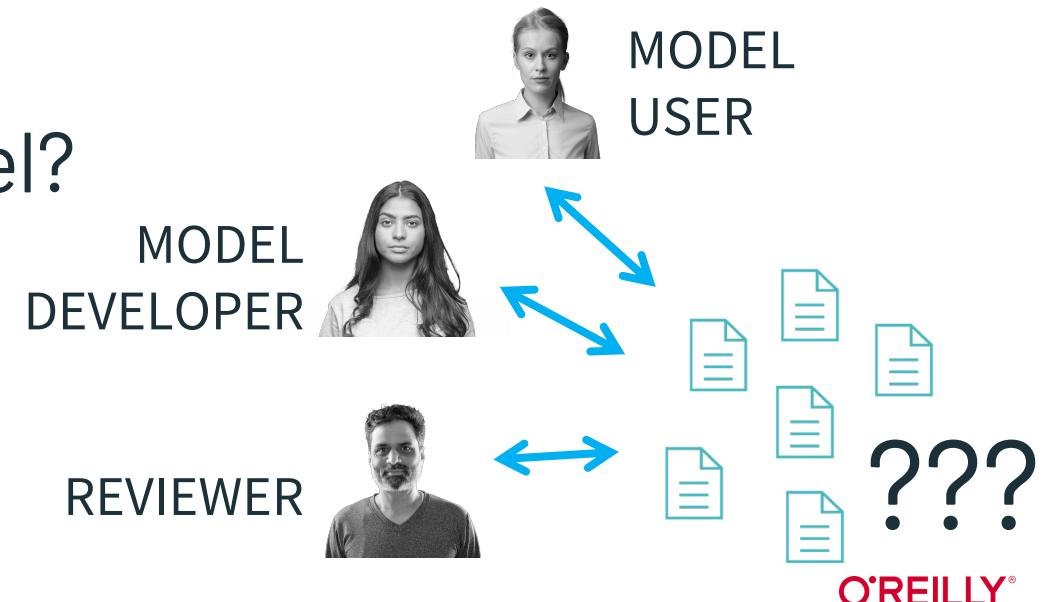
When you're working on one ML app alone, storing your models in files is manageable



The Model Management Problem

When you work in a large organization with many models, many data teams, management becomes a major challenge:

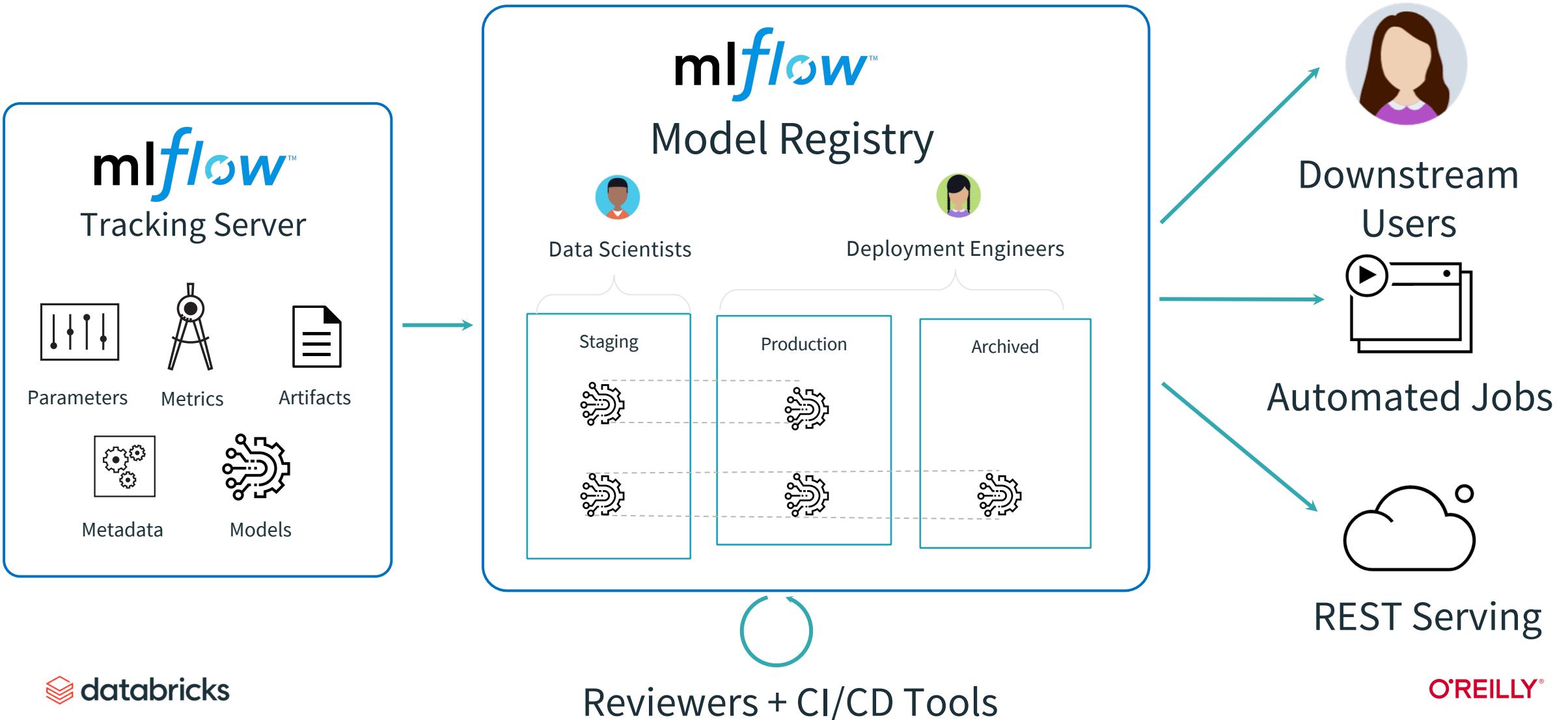
- Where can I find the best version of this model?
- How was this model trained?
- How can I track docs for each model?
- How can I review models?
- How can I integrate with CI/CD?





Model Registry

VISION: Centralized and collaborative model lifecycle management



MLflow Model Registry

- Repository of named, versioned models with controlled Access to Models
- Track each model's stage: none, staging, production, or archived
- Easily inspect a specific version and its run info
- Easily load a specific version
- Provides model description, lineage and activities

The screenshot shows the MLflow Model Registry interface for a registered model named "Airline_Delay_SparkML".

Header: Registered Models > Airline_Delay_SparkML

Created Time: 2019-10-10 15:20:29 **Last Modified:** 2019-10-14 12:17:04

Description: Predicts airline delays (in minutes) using the best Spark RF model from the AutoML Toolkit.

Versions: All Active(1)

Version	Registered at	Created by	Stage
Version 1	2019-10-10 15:20:30	clemens@demo.com	Archived
Version 2	2019-10-10 21:47:29	clemens@demo.com	Archived
Version 3	2019-10-10 23:39:43	clemens@demo.com	Production
Version 4	2019-10-11 09:55:29	clemens@demo.com	None
Version 5	2019-10-11 12:44:44	matei@demo.com	Staging

MLflow Model Registry

The MLflow Model Registry component is a centralized model store, set of APIs, and UI, to collaboratively manage the full lifecycle of an MLflow Model. It provides model lineage (which MLflow experiment and run produced the model), model versioning, stage transitions (for example from staging to production), and annotations.

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 - Listing and Searching MLflow Models
 - Archiving an MLflow Model
 - Deleting MLflow Models



`mlflow.register_model(model_uri, name)` [source]

`mlflow.get_registry_uri()` [source]

`mlflow.set_registry_uri(uri)` [source]

MLflow Model Registry

The MLflow Model Registry component is a centralized model store, set of APIs, and UI, to collaboratively manage the full lifecycle of an MLflow Model. It provides model lineage (which MLflow experiment and run produced the model), model versioning, stage transitions (for example from staging to production), and annotations.

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Model Registry CRUD Operations MLflowClient()

`create_model_version(name, source, run_id, tags=None, run_link=None, description=None)` [\[source\]](#)

`create_registered_model(name, tags=None, description=None)` [\[source\]](#)

`delete_model_version(name, version)` [\[source\]](#)

`get_latest_versions(name, stages=None)` [\[source\]](#)

`transition_model_version_stage(name, version, stage, archive_existing_versions=False)` [\[source\]](#)

Model Registry Workflow UI

Artifacts

sklearn-model

- MLmodel
- conda.yaml
- model.pkl

Full Path: ./mirluns/0/55eb2ad528114c68bd354a0568eca327/artifacts/sklearn-model
Size: 0B

Select a file to preview
Supported formats: image, text, html, geojson files

Register Model



MODEL
DEVELOPER

Tags

Name	Value	Actions
No tags found.		

Add Tag

Artifacts

sklearn-model

- MLmodel
- conda.yaml
- model.pkl

Full Path: ./mirluns/0/55eb2ad528114c68bd354a0568eca327/artifacts/sklearn-model
Size: 0B

Register Model

Model

+ Create New Model

Model Name

SKLearnPowerForecast

Cancel Register

mse ↗ 43186.3

rmse ↗ 207.8

Tags

Name	Value
No tags found.	

Add Tag

Name Value

Artifacts

sklearn-model

- MLmodel
- conda.yaml
- model.pkl

Full Path: ./mirluns/0/55eb2ad528114c68bd354a0568eca327/artifacts/sklearn-model
Size: 0B

Select a file to preview
Supported formats: image, text, html, geojson files

Register Model



Artifacts

sklearn-model

- MLmodel
- conda.yaml
- model.pkl

Full Path: ./mirluns/0/55eb2ad528114c68bd354a0568eca327/artifacts/sklearn-model
Size: 0B

SKLearnPowerForecast, v1
Registered on 2020/05/04

Select a file to preview
Supported formats: image, text, html, geojson files

Register Model

Model

+ Create New Model

Model Name

SKLearnPowerForecast

Cancel Register

Model Registry Workflow UI

This screenshot shows the mlflow Model Registry UI. It displays a single model version (Version 1) for a "SKLearnPowerForecast" model. The details include:

- Registered At: 2020-05-04 11:38:47
- Creator: [redacted]
- Stage: None
- Last Modified: 2020-05-04 11:38:47
- Source Run: Random Forest Regressor: Power Forecasting Model

The "Description" section contains the text: "This version forecasts the power". Below the description is a rich text editor interface with buttons for Write, Preview, H, B, I, S, etc. At the bottom are Save and Cancel buttons.



MODEL
REVIEWER

This screenshot shows the same model version detail page as the first one, but with additional transition options in the "Stage" dropdown:

- Transition to → Staging (highlighted in orange)
- Transition to → Production (highlighted in green)
- Transition to → Archived

The "Description" section includes the text: "This version forecasts the power".



DOWNSTREAM
USERS



AUTOMATED JOBS



REST SERVING

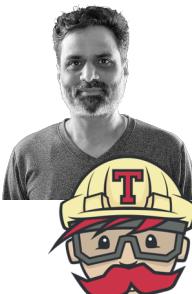
Model Registry Workflow API

```
mlflow.register_model(model_uri, "WeatherForecastModel")  
  
mlflow.sklearn.log_model(model,  
    artifact_path="sklearn_model",  
    registered_model_name= "WeatherForecastModel")
```

MODEL
DEVELOPER



REVIEWERS,
CI/CD TOOLS



```
client = mlflow.tracking.Mlflowclient()  
client.transition_model_version_stage(name="WeatherForecastModel",  
    version=5,  
    stage="Production")
```

```
model_uri= "models:/{{model_name}}/production".format(  
    model_name="WeatherForecastModel")  
model_prod = mlflow.sklearn.load_model(model_uri)  
model_prod.predict(data)
```

DOWNSTREAM
USERS



AUTOMATED JOBS



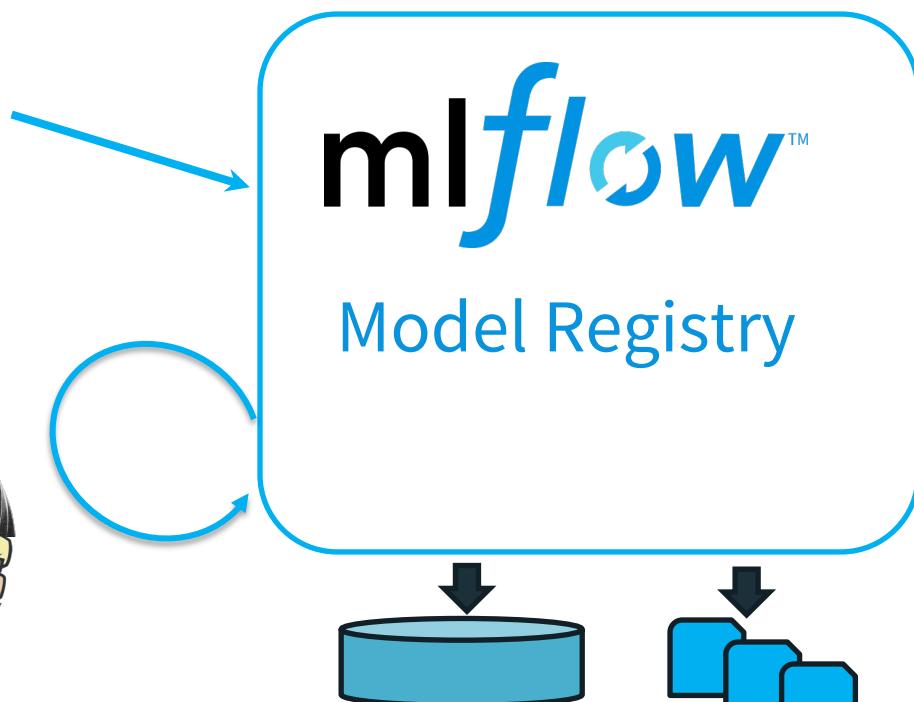
REST SERVING



Model Registry Workflow API

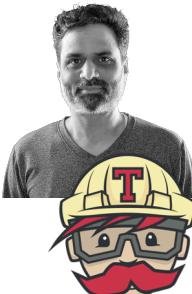
```
mlflow.register_model(model_uri, "WeatherForecastModel")  
  
mlflow.sklearn.log_model(model,  
    artifact_path="sklearn_model",  
    registered_model_name= "WeatherForecastModel")
```

MODEL
DEVELOPER



```
model_uri = "models:/{{model_name}}/production".format(  
    model_name="WeatherForecastModel")  
model_prod = mlflow.pyfunc.load_model(model_uri)  
model_prod.predict(data)
```

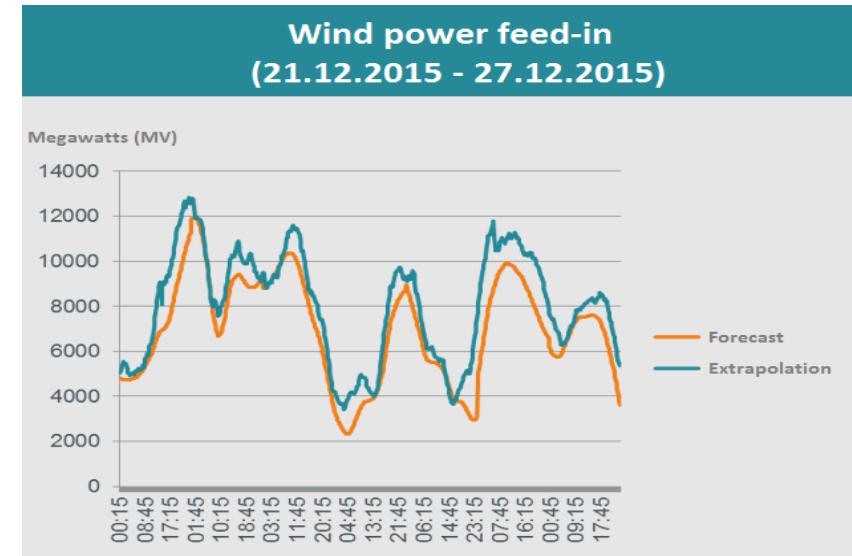
REVIEWERS,
CI/CD TOOLS



```
client = mlflow.tracking.MlflowClient()  
client.transition_model_version_stage(name="WeatherForecastModel",  
    version=5,  
    stage="Production")
```



DOWNSTREAM
FORECASTING APP

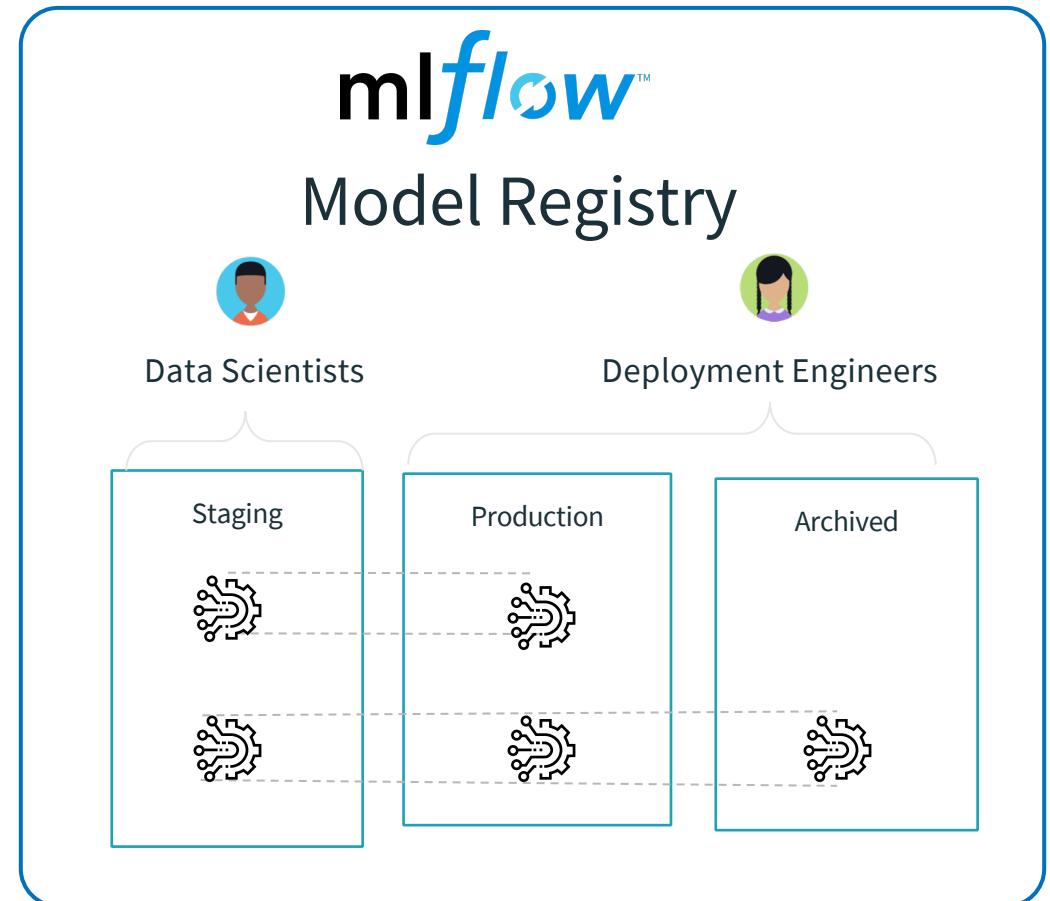


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```
(tutorials) ➔ src git:(master) ✘ sqlite3 ./mlruns.db
SQLite version 3.30.1 2019-10-10 20:19:45
Enter ".help" for usage hints.
sqlite> .databases
main: /Users/julesdamji/gits/mlflow-workshop-part-3/src./mlruns.db
sqlite> .tables
alembic_version      latest_metrics      params          tags
experiment_tags       metrics            registered_models
experiments           model_versions     runs
sqlite> █
```

MLflow Model Registry Recap

- **Central Repository:** Unique named registered models for discovery across data teams
- **Model Registry Workflow:** Provides UI and API for registry operations
- **Model Versioning:** Allow multiple versions of model in different stages
- **Model Stages:** Allow stage transition: none, staging, production, or archived
- **CI/CD Integration:** Easily load a specific version for testing and inspection
- **Model Lineage:** Provides model description, lineage and activities



mlflow Model Registry: Tag and Search APIs

Tags to track custom metadata for a model version, e.g., test results

Search API to automate model management and MLOps actions

The screenshot shows the MLflow Model Registry interface. At the top, there are tabs for 'Experiments' and 'Models'. The 'Models' tab is selected, showing a list of models: 'Models > KNN > Version 12'. Below this, detailed information is provided: 'Created at: 2018-12-04 17:11:06', 'User: test@example.com', 'Stage: Staging', 'Last Modified: 2018-12-04 17:11:06', and 'Source: Run 123'. There are three main sections: 'Notes' (with a link), 'Pending Requests', and 'Tags'. The 'Tags' section is expanded, showing two entries in a table:

Name	Value	Actions
passed-gdpr-review	true	
passed-performance-test	true	

Below the table is a blue 'Add Tag' button. The 'Activity' section shows two recent events: 'Alice requested a stage transition' (None → Staging) and 'Carol approved a stage transition' (None → Staging), both occurring 10 hours ago.

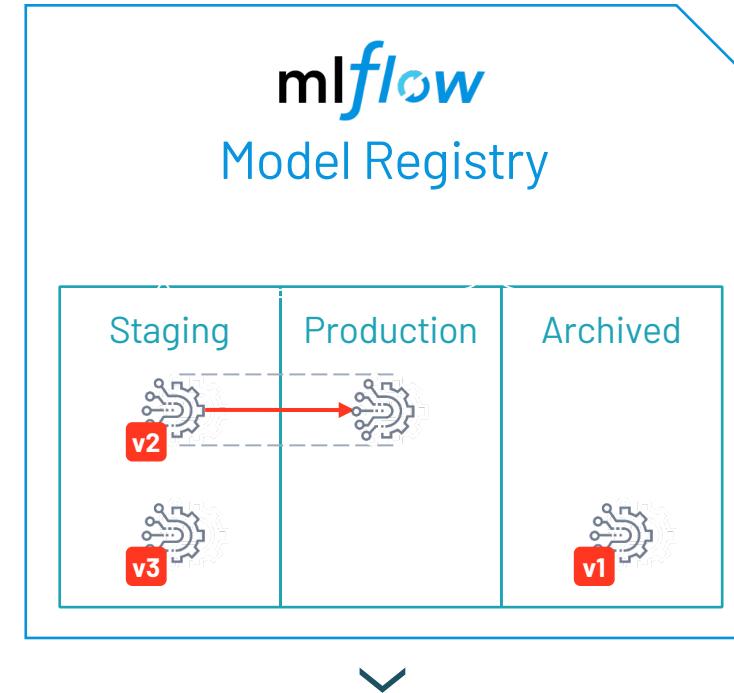
mlflow Model Registry: Webhooks

Databricks Webhooks allow setting callbacks on registry events like stage transitions to run CI/CD tools

MLflow Model Registry on Databricks
Simplifies MLOps With CI/CD Features



by Sue Ann Hong, Ankit Mathur, Jules Damji and Mani Parkhe
Posted in ENGINEERING BLOG | November 19, 2020



`VERSION_REGISTERED: MyModel, v2`

`TRANSITION_REQUEST: MyModel, v2, Staging→Production`

`TAG_ADDED: MyModel, v2, BacktestPassed`

just launched

Human
Reviewers

CI/CD Tools

Batch
Scoring

Real-time
Serving



mlflow Model Registry: Comments

just launched

Comments in the Databricks workspace can now be used to discuss changes on models

The screenshot shows a comments section titled "Activities". A comment from "ci-pipeline@databricks.com" was posted 2 hours ago (edited). The message reads: "Tests failed - see output [here](#)". Below the comment is a placeholder text "Add a comment" with a speech bubble icon.

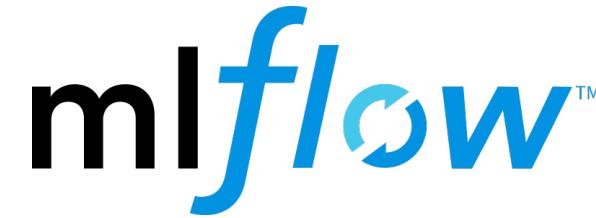
MLflow Model Registry on Databricks Simplifies MLOps With CI/CD Features



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Recap of all parts: What Did We Talk About?



- Modular Components greatly simplify the ML lifecycle
- Easy to install & Great Developer experience
- Develop & Deploy locally; track locally or remotely
- Use Available APIs: Python, Java & R (Soon Scala)
- REST APIs and CLI tools
- Visualize experiments and compare runs
- Centrally register and manage model lifecycle

Model Registry Tutorial

Tutorials: <https://github.com/dmatrix/olt-mlflow>

Thank you! 😊

Q & A

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