Platform for Machine Learning Lifecycle

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MLOps: Production & Engineering World

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
    - Google Colab
- Q & A

https://github.com/dmatrix/tmls-workshop

MLflow Components

mlflow

Tracking

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

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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

new

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Model Registry

Store, annotate and manage models in a central repository

databricks.com/mlflow

mlflow.org

github.com/mlflow

twitter.com/MLflow

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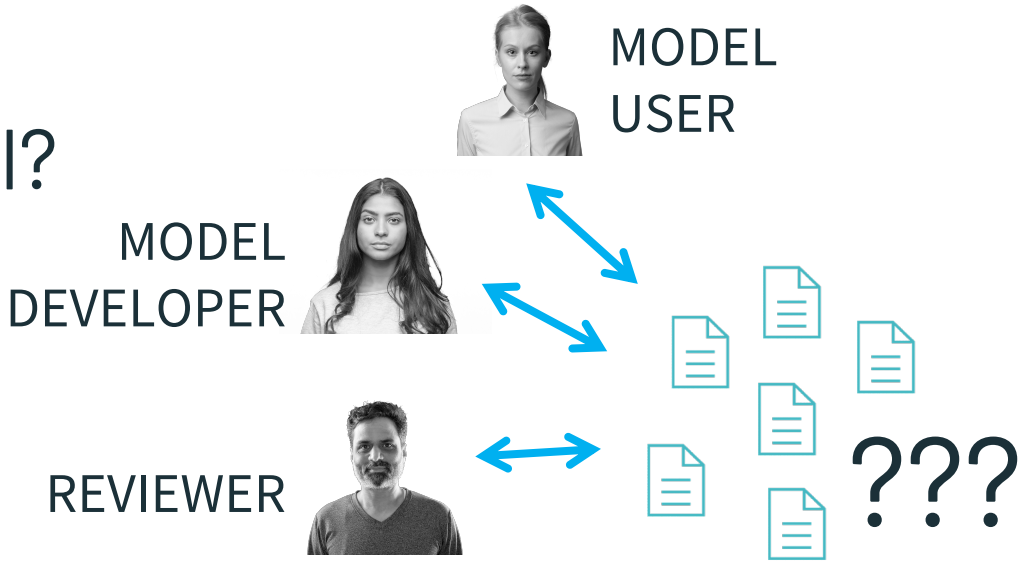
The Model Management Problem

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?



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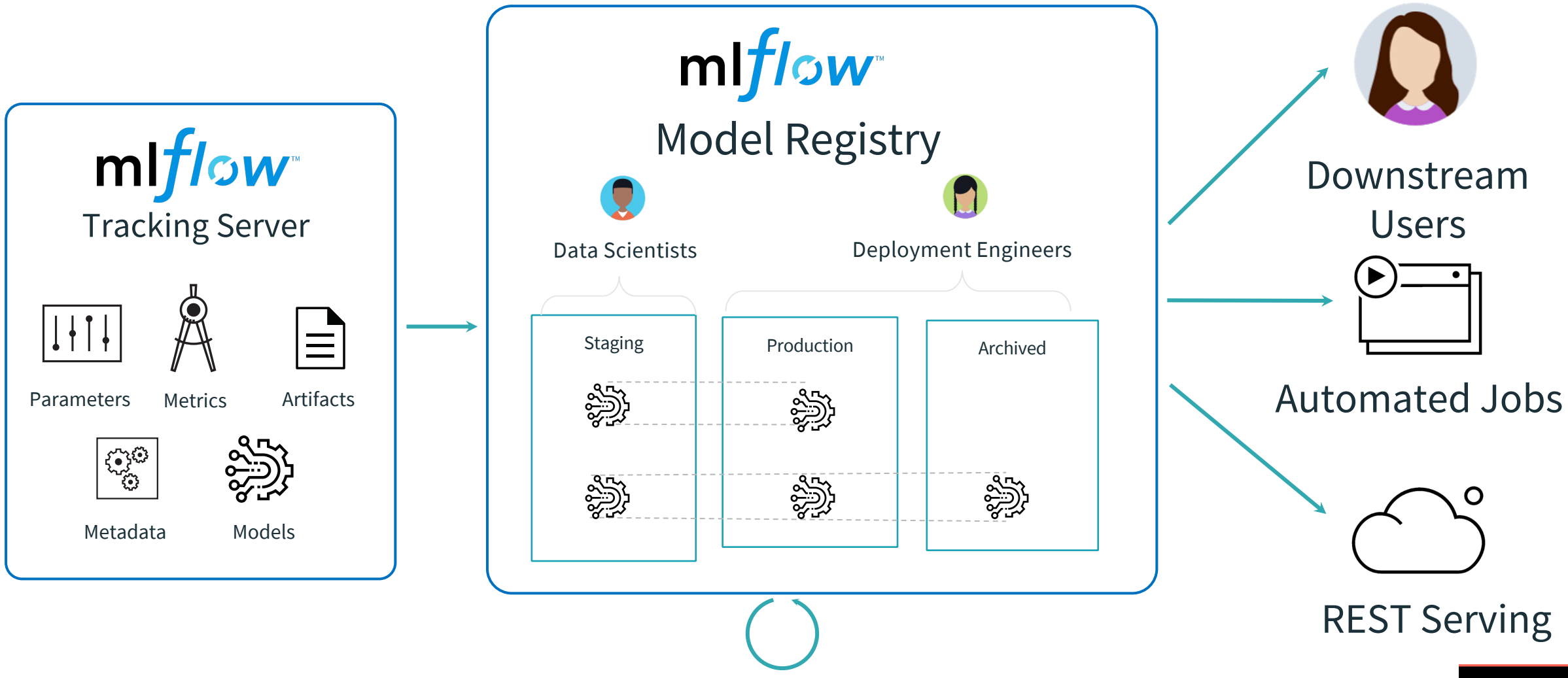
mlflow Model Registry

VISION: Centralized and collaborative model lifecycle management



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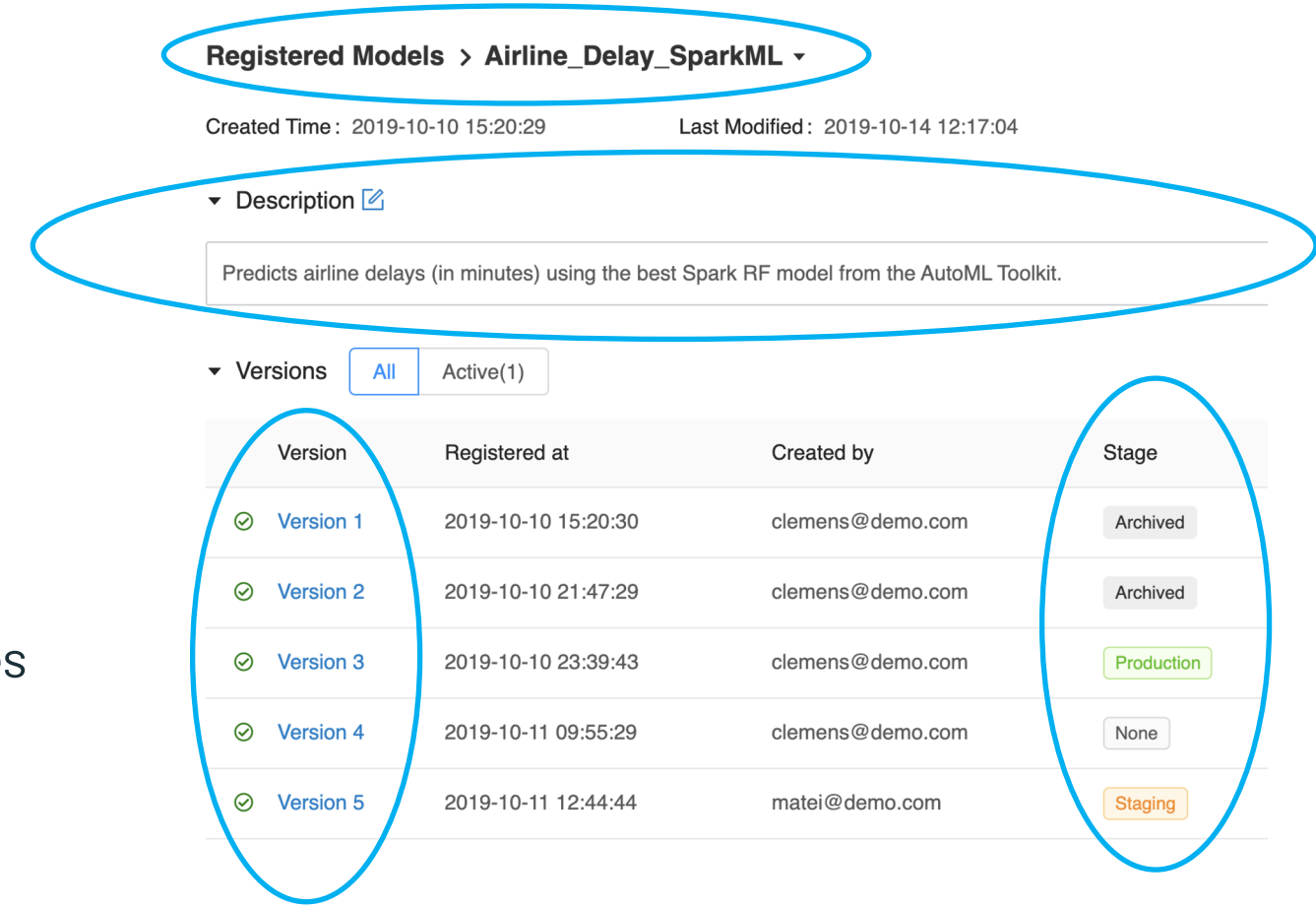
MLflow Model Registry

- 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



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.

Table of Contents

Model Registry CRUD Operations
MLflowClient()

create_model_version(*name, source, run_id, tags=None, run_link=None, description=None*) [source]

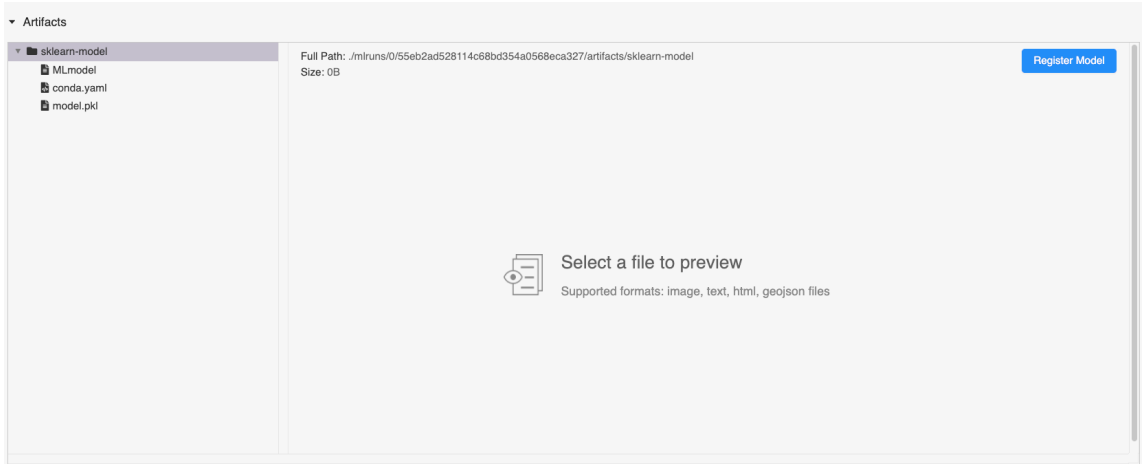
create_registered_model(*name, tags=None, description=None*) [source]

delete_model_version(*name, version*) [source]

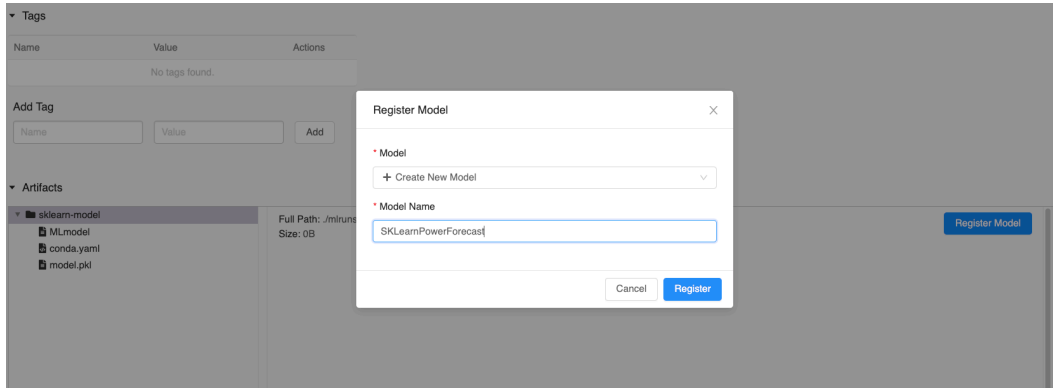
get_latest_versions(*name, stages=None*) [source]

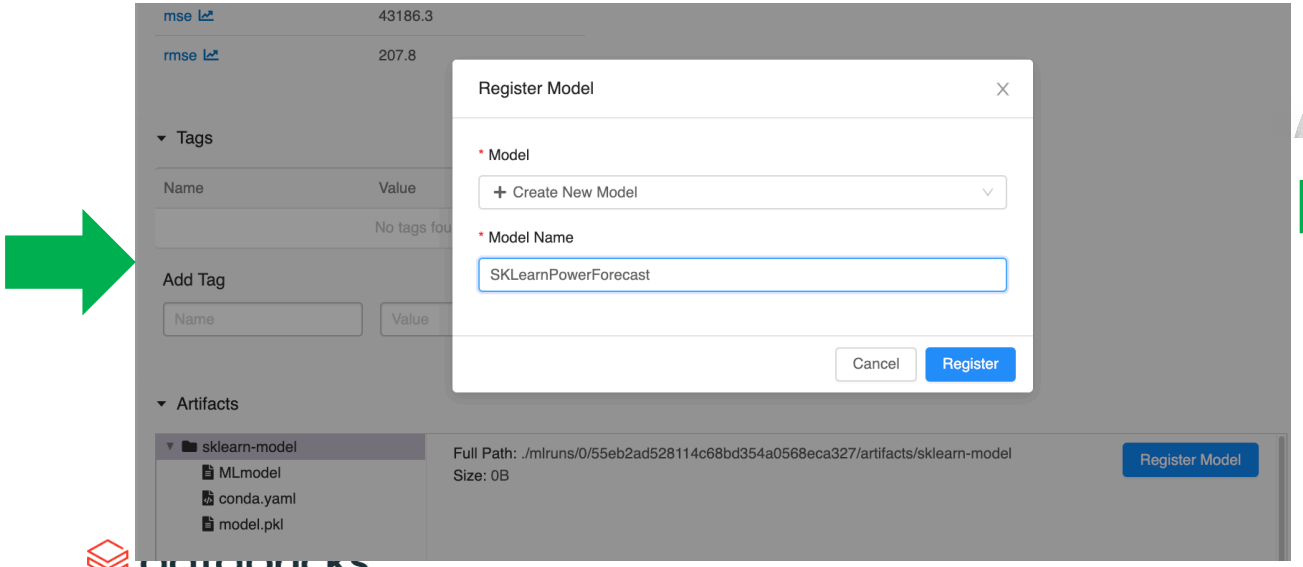
transition_model_version_stage(*name, version, stage, archive_existing_versions=False*) [source]

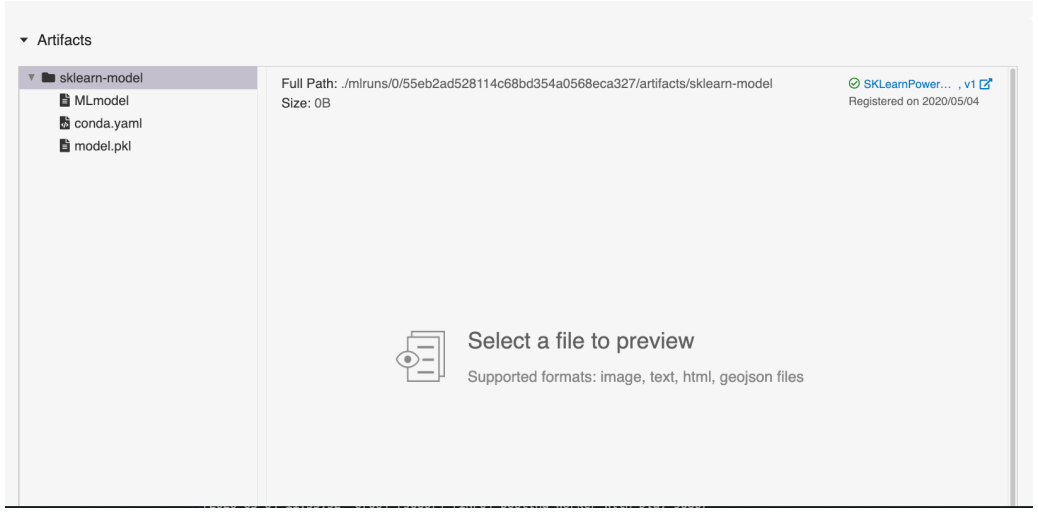
Model Registry Workflow UI



MODEL DEVELOPER







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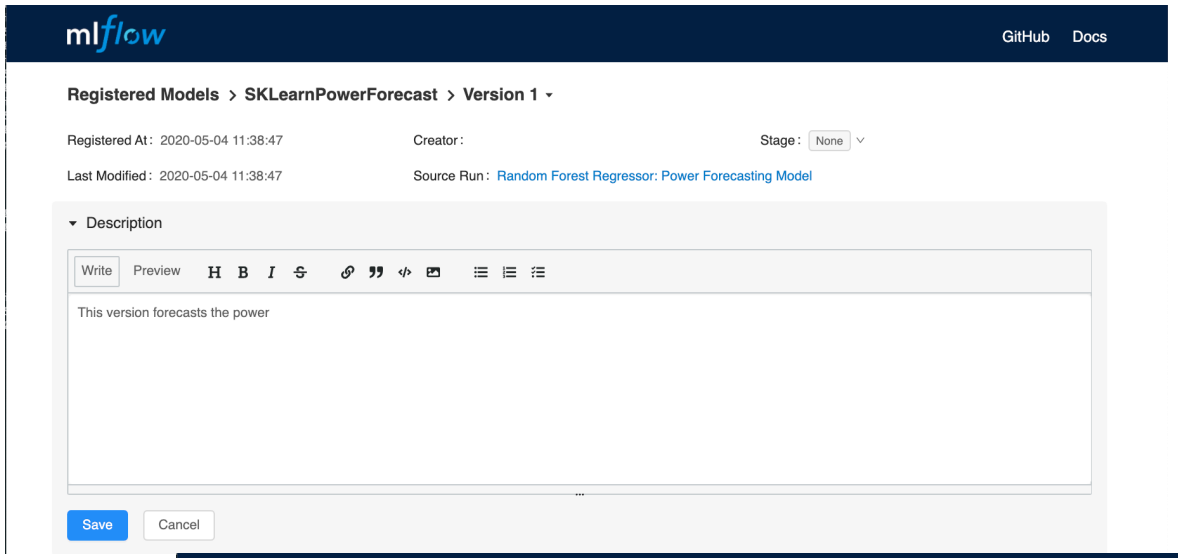
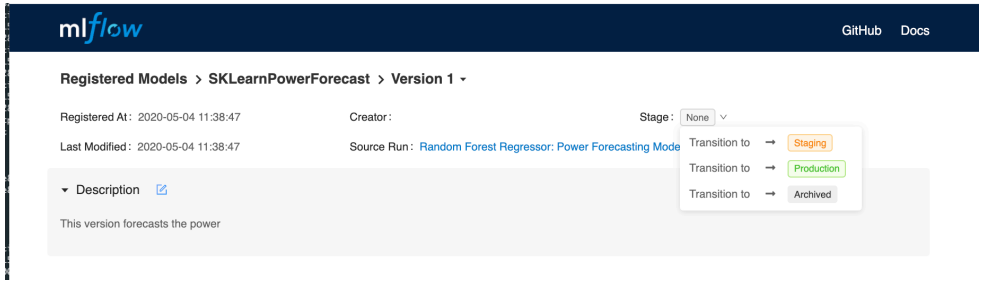
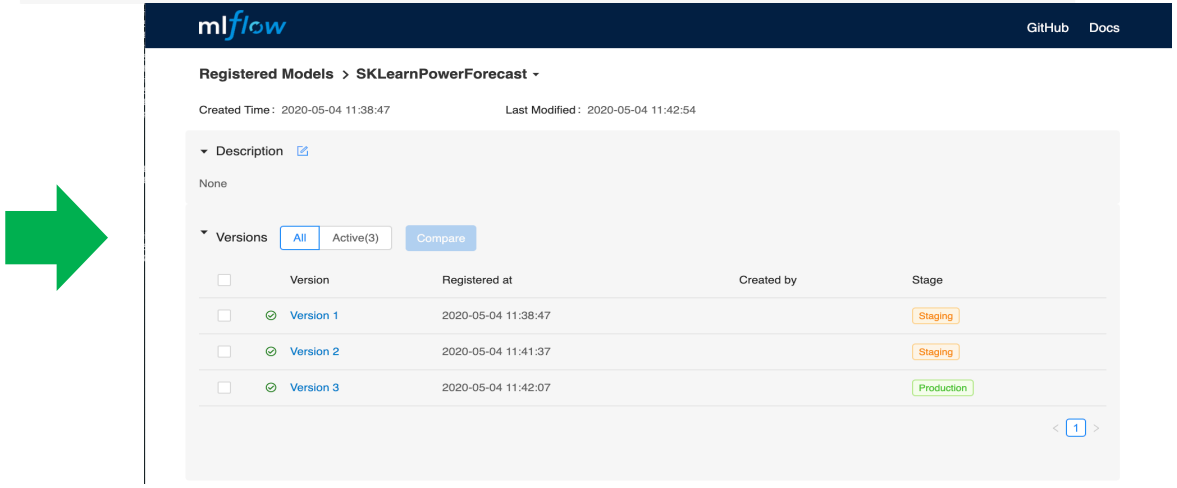
Model Registry Workflow UI



MODEL REVIEWER





DOWNSTREAM USERS

AUTOMATED JOBS

REST SERVING

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Model Registry Workflow API

```
mlflow.register_model(model_uri,"WeatherForecastModel")

mlflow.sklearn.log_model(model,
          artifact_path=”sklearn_model”,
          registered_model_name= “WeatherForecastModel”)
```

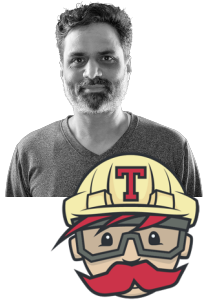
```
model_uri= "models:/{model_name}/production".format(
          model_name="WeatherForecastModel")
model_prod =  mlflow.sklearn.load_model(model_uri)
model_prod.predict(data)
```

MODEL DEVELOPER

DOWNSTREAM USERS

AUTOMATED JOBS

REST SERVING



REVIEWERS, CI/CD TOOLS

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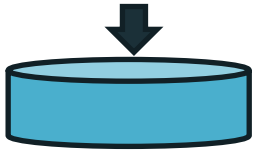
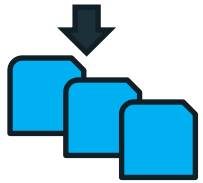
Model Registry





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

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Model Registry Workflow API

```
mlflow.register_model(model_uri,"WeatherForecastModel")

mlflow.sklearn.log_model(model,
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```

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

MODEL DEVELOPER

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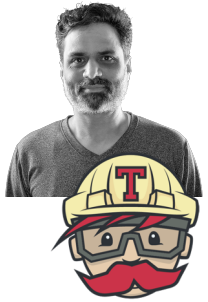
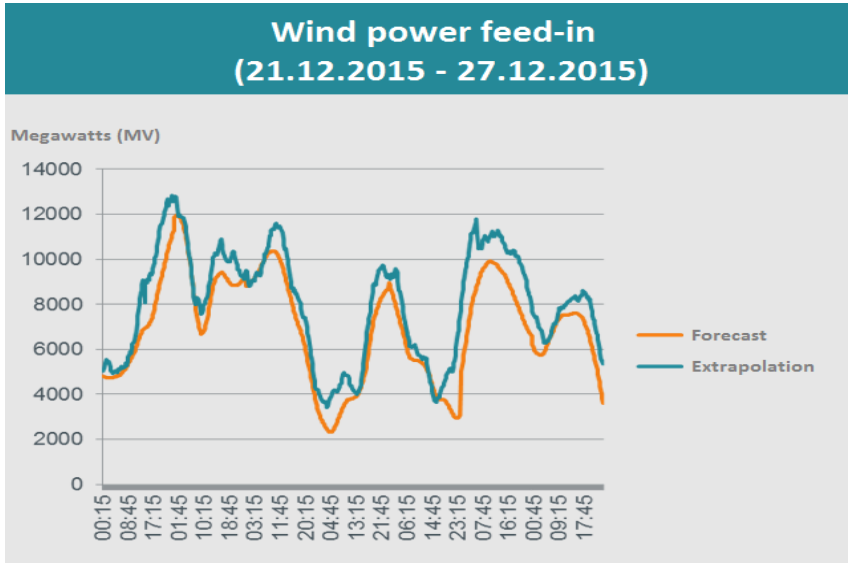
Model Registry

DOWNSTREAM FORCASTING APP

REVIEWERS, CI/CD TOOLS



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client.transition_model_version_stage(name=”WeatherForecastModel”,
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```

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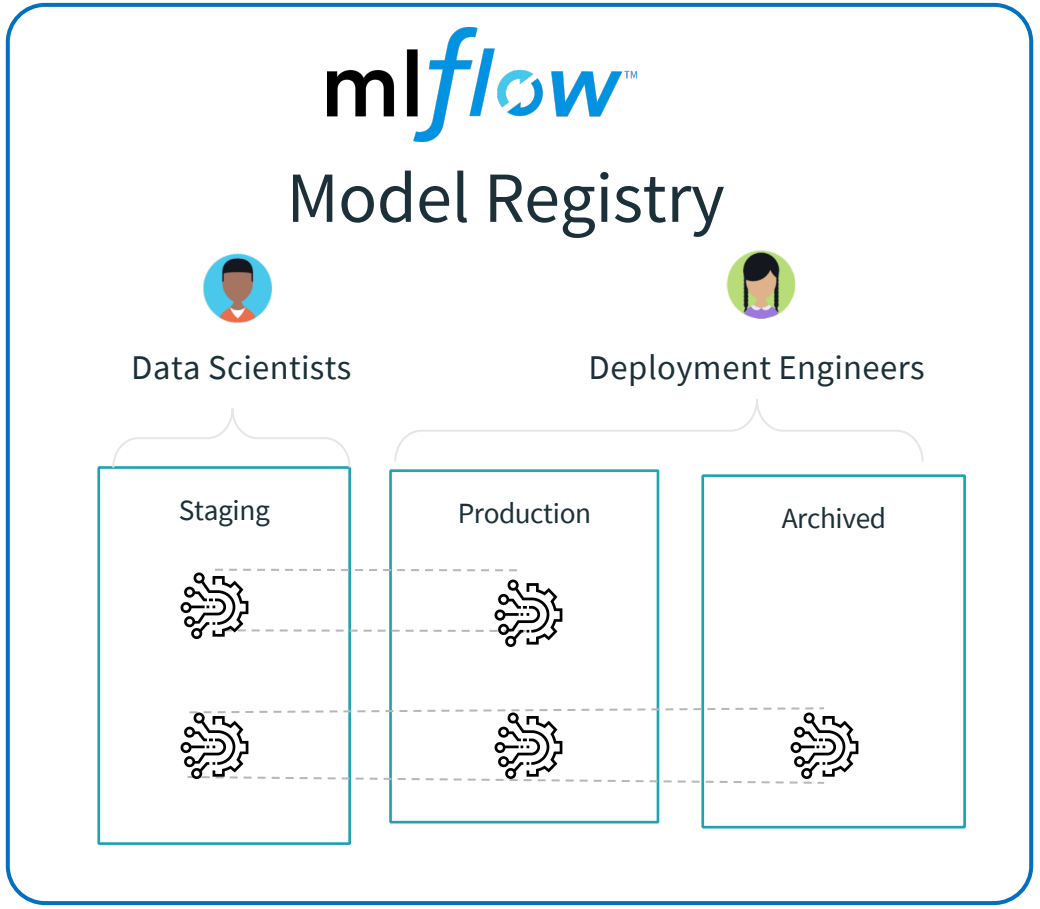
MLflow Model Registry Recap

- 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

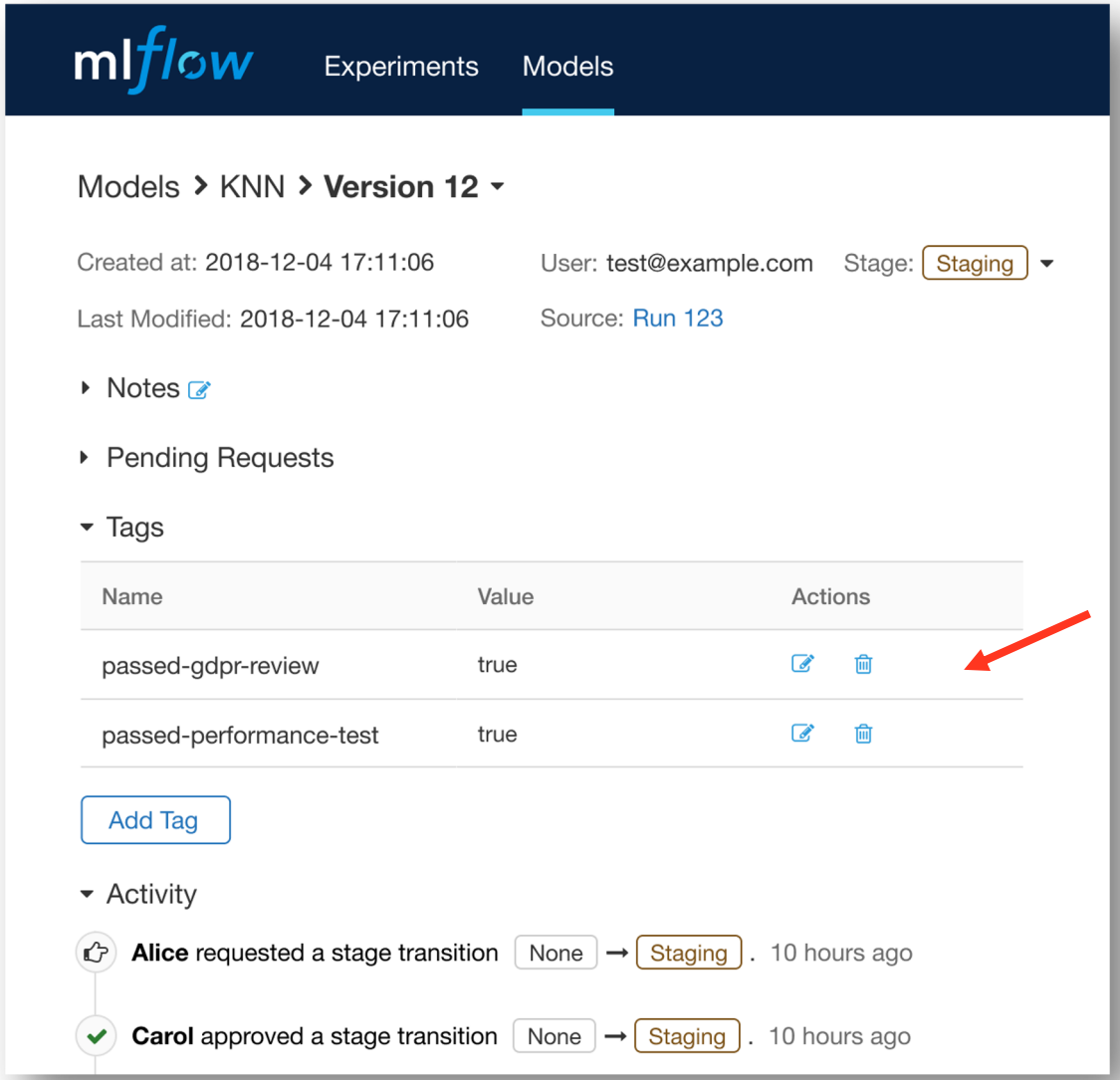


mlflow Model Registry: Tag and Search APIs

in MLflow 1.11

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

Search API to automate model management and MLOps actions



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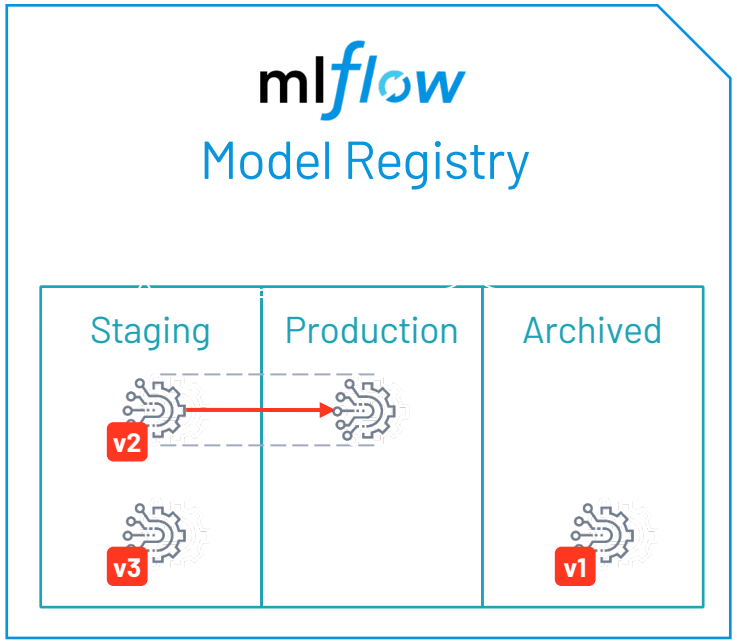
mlflow Model Registry: Webhooks

just launched

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
```

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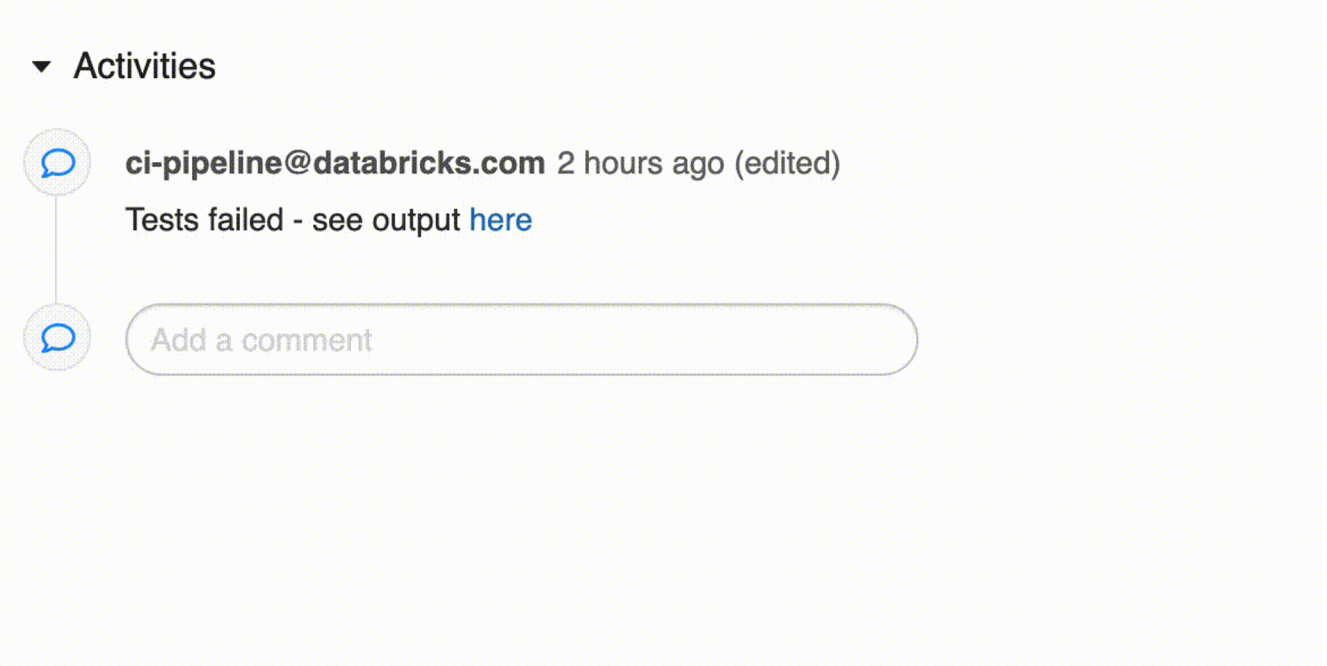
mlflow Model Registry: Comments

just launched

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



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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
- Available APIs: Python, Java & R (Soon Scala)
- REST APIs and CLI tools
- Visualize experiments and compare runs
- Centrally register and manage model lifecycle

Thank you! ☺

Q & A

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