MLflow Workshop

Spyros Stasis

Workshop Outline

MLflow Overview

- Project Lifecycle
- MLflow components

Hands on examples:

- Boston housing prices regression
- MNIST classification

Reproducible Research Tools

Machine Learning Project:

• Stages:

Raw Data → Data Preparation → Model Development → Deployment

- Considerations:
 - 1. Data
 - 2. Environment
 - 3. Parameters
 - 4. Deployment

Tools for tracking/reproducible machine learning pipelines:

- Uber Michelangelo, Facebook FBLearner, Tensorflow TFX
- Data Version Control (DVC)
- Pachyderm

What is MLflow?

- Open source project developed by Databricks
- Main principles:
 - Agnostic
 - Universal
 - Ease of use
 - Flexibility/Scalability



Open source machine learning platform

- Works with any ML library, algorithm, language, etc.
- Open interface design (use with any code you already have)

mlflow Tracking

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

mlflow Projects

Packaging format for reproducible runs and workflows

mlflow Models

General format that standardizes deployment paths



Centralized model management, review & sharing

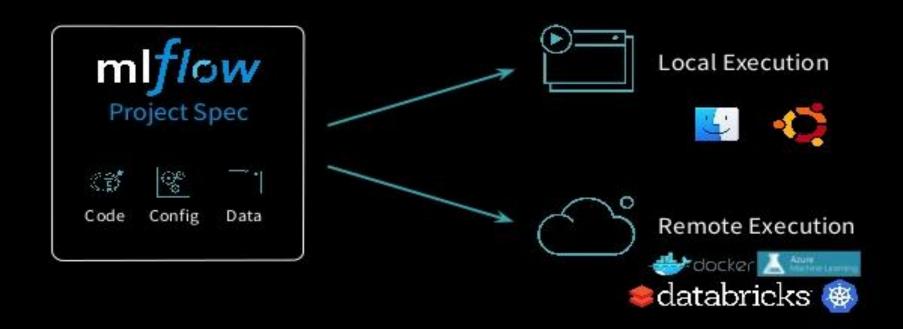
databricks

Slide taken from MLflow: A Platform for Production Machine Learning presentation by Matei Zaharia, (link)

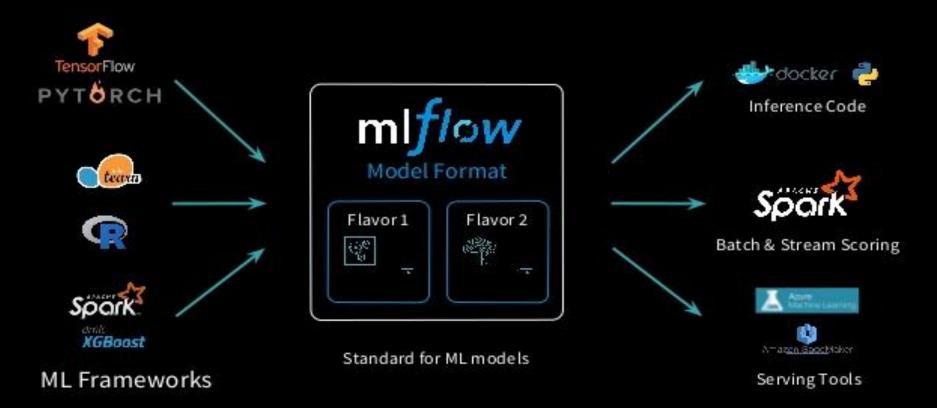
MLflow Tracking



MLflow Projects



MLflow Models



MLflow: Infrastructure for a Complete Machine Learning Life Cycle with Mani Parkhe and Tomas Nykodym (<u>link</u>)

MLflow Model Registry

GitHub-like environment for organizing & reviewing models





Slide taken from MLflow: A Platform for Production Machine Learning presentation by Matei Zaharia, (link)

https://github.com/Pyrsos/qmul_mlflow

Boston housing regression

- Try out:
 - Different regression techniques:
 - Linear Regression
 - Lasso
 - Additional metrics:
 - Max Error
 - Median absolute error
 - Logging artifacts
 - Perform PCA on the input data
 - Store the image in the MLflow logs

MNIST example

- Try out:
 - Different model architectures:
 - Add/remove dense layers
 - Change layer sizes/activation
 - Try and keep track of these changes
 - Custom metrics:
 - Define and log custom metrics
 - Track the metric over time (over epochs/batches)

Resources

- Mlflow Infrastructure for a complete Machine Learning Lifecycel
- Accelerating the Machine Learning Lifecycle with mlflow
- Simplifying model management with mlflow

Thank you! Any questions?