#### amadeus

#### MLflow and beyond

Machine Learning in real life



Generoso PAGANO Davide SANAPO



#### The Speakers



Generoso



- Big-Data products
- Data Mesh architecture
- Machine Learning dev-ops



Davide



#### Agenda

- \_ Machine Learning in real life
- \_ MLflow
- \_ Pyntel



#### ML is everywhere!

\_ ML is reshaping businesses and our daily life

- Price prediction
- SPAM detection
- Product recommendation
- Fraud detection
- Voice recognition
- Chat bots
- Autonomous driving
- •





### ML development is harder than traditional SW development



Matei Zaharia

Author of Apache Spark and MLflow

#### **SW** development

Meet a functional specification

Goal

Quality

Stack

Depends only on code

Defined SW stack for a given task

#### **ML** development

Optimize a metric (e.g. accuracy)

Depends mostly on input data and tuning parameters

Combine many library and frameworks for the same task

#### ML lifecycle





O PyTorch

**†** TensorFlow



Reproducibility



**Raw Data** 

























Model exchange

## mlf/ow

- Open source platform for the ML lifecycle
  - Works with most ML framework
  - Works with any programming languages
  - Scales to Big Data with Apache Spark
- Some of the users...











☆ Star

Contributors 300

🦚 🚱 🗯 🤶

9.5k

약 Fork

2.1k















**Tracking** 

**Projects** 

Registry

**Models** 

# Sinding has not till fifty at the second Harristone A. O.

#### MLflow

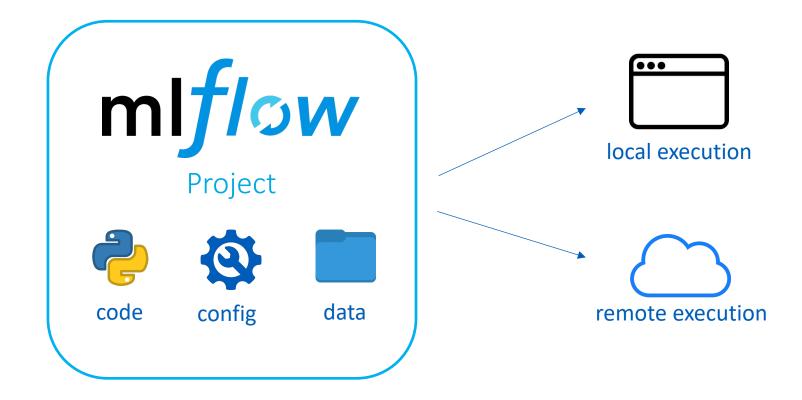
#### Tracking



- \_ API (Python, REST, R, Java) and UI for logging and querying
  - parameters
  - code version
  - metrics
- \_ Analyse and compare results

#### MLflow

#### Projects



- \_ packaging data science code in a reusable and reproducible way
- \_ composing projects

#### MLflow

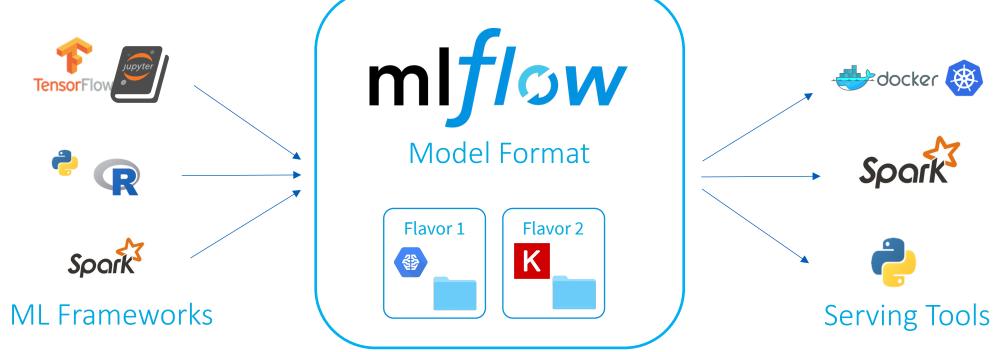
#### Model Registry



- \_ manage the full lifecycle of a model
- \_ model lineage
- \_ model versioning
- \_ stage transitioning

#### **MLflow**

#### Models



- \_ serialization and packaging format for ML Models
- \_ defines model dependencies
- \_ save models from any framework in MLflow format
- \_ deployment APIs

#### Mlflow Model

#### Limitations

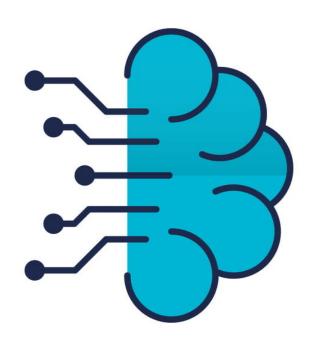
- Model serving through REST API
  - No swagger grammar
- Open model serving
  - No gRPC grammar



#### Pyntel

#### Overview

- Python library
- Extends MLflow Models
- Model signature generation
  - Swagger grammar
  - gRPC schema
- \_ Enrich the model with cloud provider metadata
  - experiment id, subscription, resource group, runId
- Validators
  - prediction validation



#### Conclusions

- Machine Learning in production
- Overview on Mlflow components
- \_ Pyntel
  - Contribution to MLflow planned



#### amadeus

#### MLflow and beyond

Machine Learning in real life



Generoso PAGANO Davide SANAPO

