# We build a ML pipeline after we deploy

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## Agenda

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### 1. Introduction

#### 2. The awesome solution

- ML pipeline out of the box in 1 line of code
- add ML pipeline anytime even in prod env

## Agenda

1. Introduction

2. ML Pipeline:

- why and when
- building blocks
- engineering
- debugging and monitoring
- open-source Python libraries to save your time



## Introduction



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> ML Pipeline: why and when

### Why?

reduce the cost of data science projects:

- focus on new cases/models
- prevent bugs
- audit

## When?

- go from PoC to MVP
- time to scale

ML Pipeline: building blocks

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#### Development and experimentation

Try new ML algorithms with orchestrated experiment steps

**Output:** ML pipeline steps source code





ML pipeline continuous integration Build source code, run tests

**Output:** ML pipeline components to deploy



ML pipeline continuous delivery
Deploy artifacts to the target env
Output: deployed ML pipeline







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> ML Pipeline: engineering

## Engineer around failures

- input checks
- output checks
- model fallback

## Engineer for performance

- scale
- caching
- feedback collection

ML Pipeline: debugging and monitoring

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## ML debugging and monitoring



## Top 3 debugging issues

unpinned libraries
data pre-processing
scattered config for diff envs

### Monitoring

system monitoring
data monitoring
model monitoring



# ML Pipeline: Python libraries

Delta lake, DVC - data versioning Airflow - data processing **Great Expectations** - data validation Feast - feature store Hyperopt, Katib - tuning Kubeflow, Pachyderm, TensorFlow Extended- platform LIME, SHAP, Alibi Explain- model interpretability Mlflow - lifecycle management Seldon Core - model serving FastAPI - restful APIs Pytest, locust - testing Evidently, Alibi Detect - debugging and monitoring

