

# Agenda

- Motivation for MLOps
- MLOps vs DevOps
- MLOps workflow
- Current trends within MLOps
- Reflection & Conclusion



Motivation

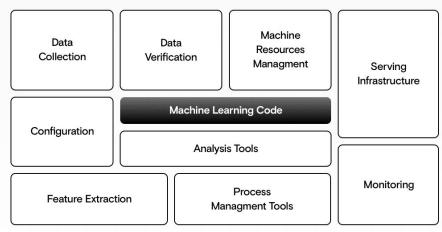
# The problem

22% of companies using ML have successfully deployed a model

 ML applications are complex distributed system (not just algorithms in notebooks)

 The real challenge is to have an <u>integrated system</u> and to <u>continuously operate</u> it in production





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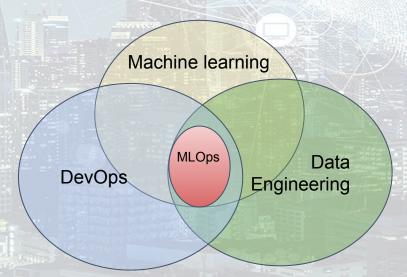
## The solution

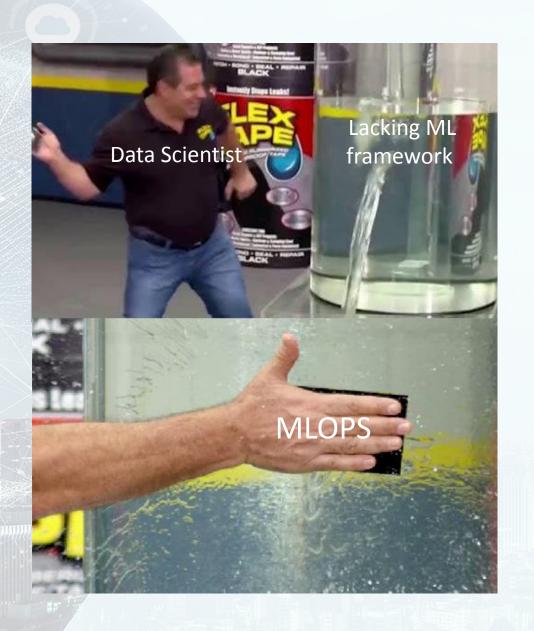
MLOps is a set of practices that combines ML,

DevOps and Data Engineering, which aims to

deploy and maintain ML systems in production

reliably and efficiently





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## MLOps vs DevOps







Dynamics of Machine Learning operations





#### Differences

- CI: <u>testing and validating data</u>, data schemas and models
- CD: A <u>system</u> rather than a single software package or service
- Continuous training (CT): Automatically retraining and serving models

## Workflow

Extraction

Data stream unification

Labeling

Feature engineering

Feature storage

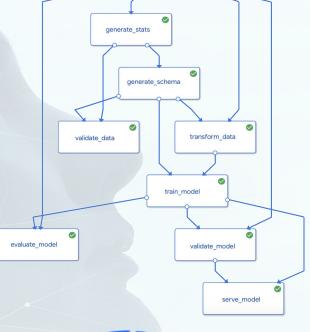
Exploratory data analysis

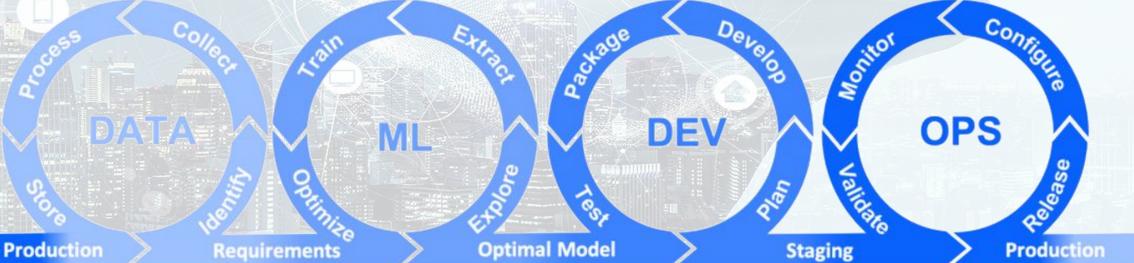
Visualization

Data validation

Model engineering







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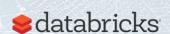
## Current trends

FEAST

PREFECT



- No Jenkins-like tool yet
- Rapid development of new tools
- Mostly task orchestration pipelines
- MLops-as-a-Service
- · Combine with cloud











SELDON

# Conclusion and our thoughts

- Machine Learning systems are much more than building models!
- Improve the maturity of the ML operation workflow:
  - reproducible
  - accountable
  - collaborative
  - continuous
- The missing piece for many companies that enables scaling of ML use

#### The Gartner Hype Cycle

