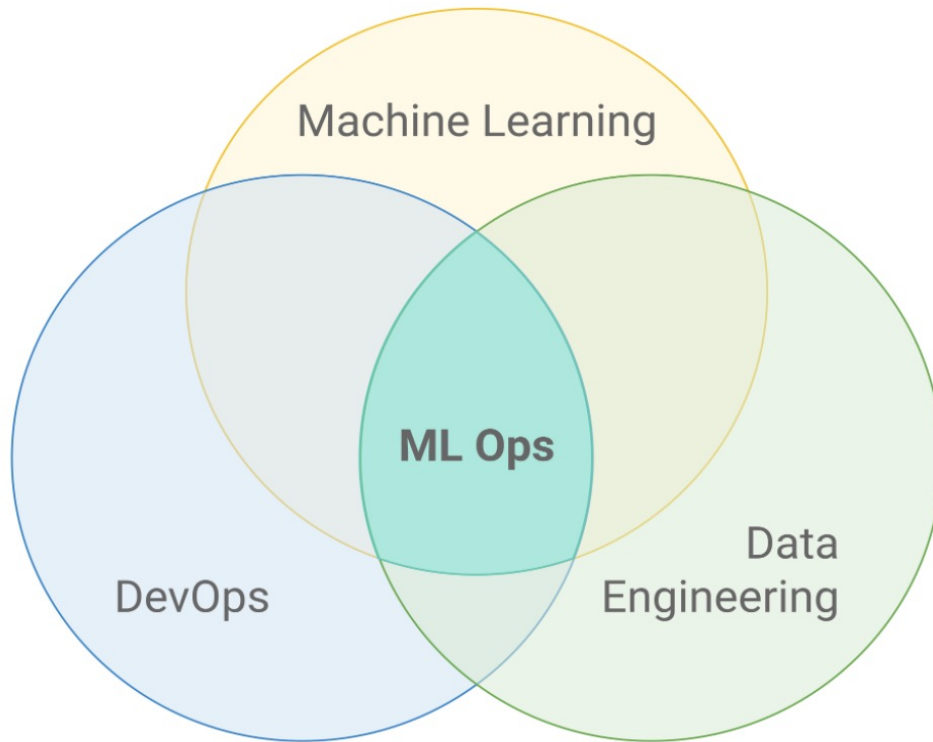




## 2 Value of MLOps



# We need MLOps



## ML systems are complex

- Models don't last forever
- Many teams need to collaborate
- ML systems change with data
- Need to monitor many aspects

## MLOps is growing

Forbes: by 2025 the MLOps business will grow to \$4 billion



# Not a software solution

MLOps is not any particular software solution but:

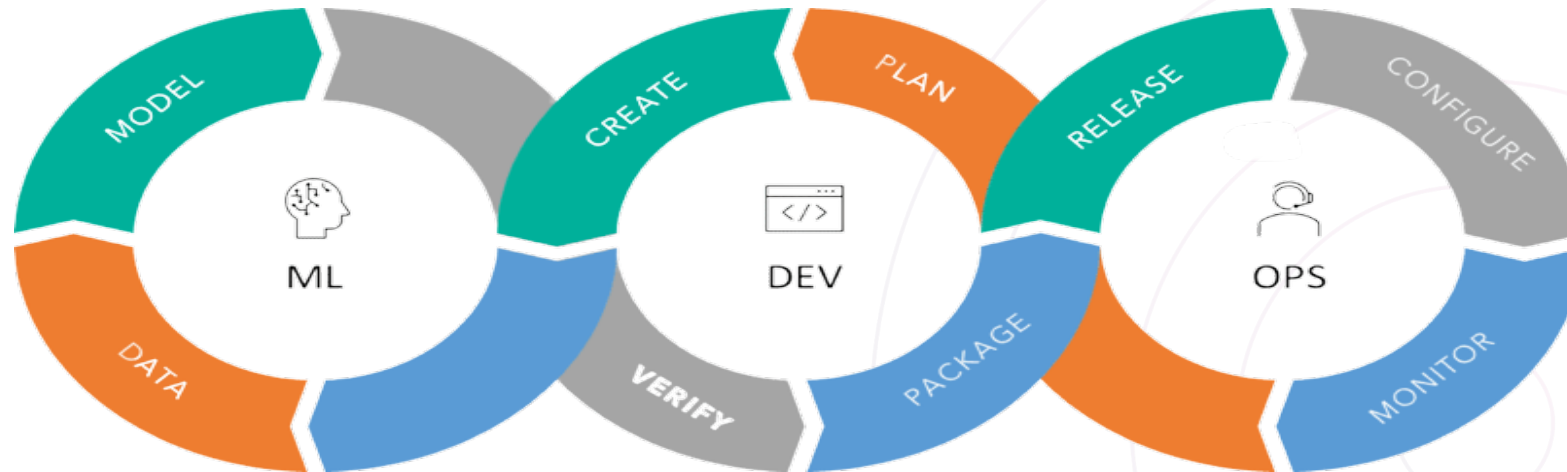
**Tools + Best Practices**

## Focus on:

- Speed up innovation through experimentation
- Create reproducible workflows
- Streamline the deployment process
- Manage the complete ML project lifecycle



# Speeding up innovation



## How:

- Brings all development and operational teams together
- Focuses on incremental improvements in development
- Uses monitoring validation and management systems to check progress



# Reproducible workflows



## Why do we need it?

- Reduce variability across iterations
- Resiliency to failure
- Smart copy

## How do we achieve it?

- Focus on traceability
- Enforce the use of abstractions (e.g. pipelines)
- Track resources (data and models)



# Streamline deployment

## CI/CD is your friend

- MLOps enforces the good use of CI/CD principles
- Continuous and traceable improvement/changes

## Manage infrastructure

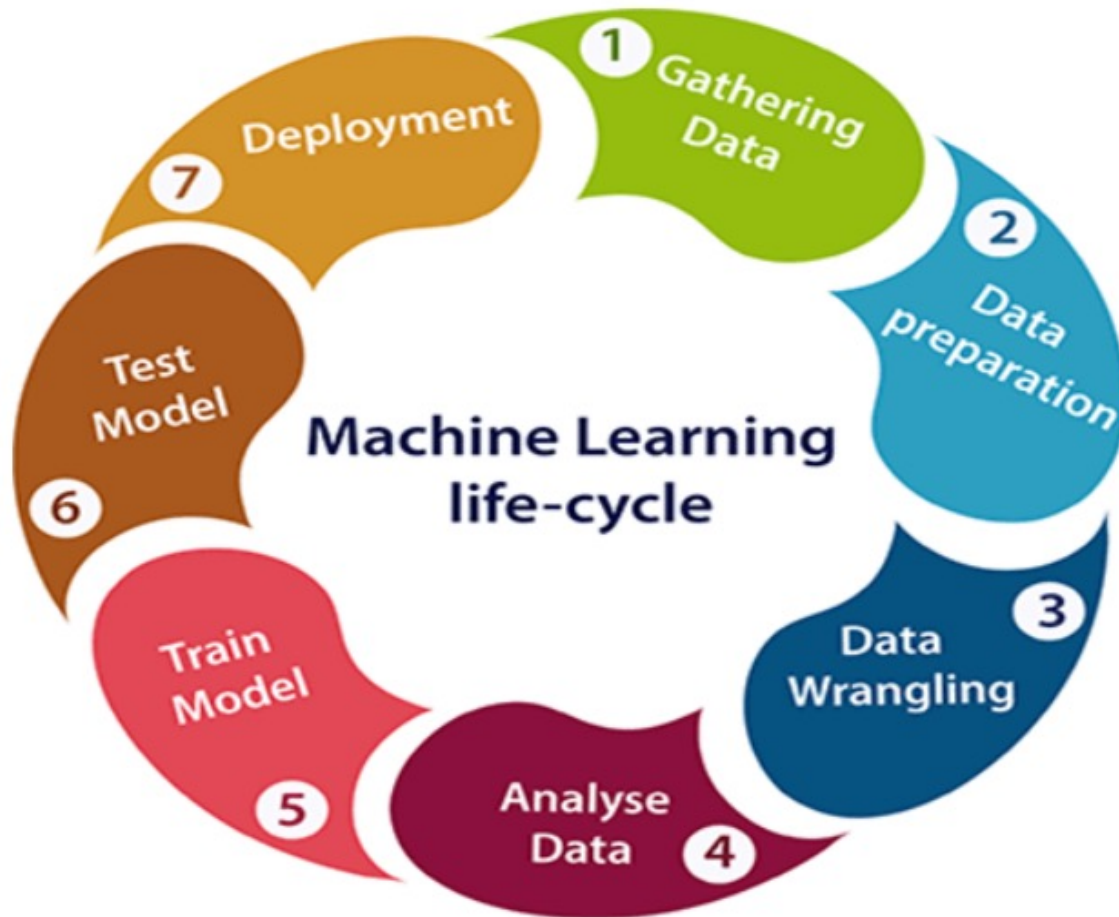
- MLOps also aims to manage the infrastructure
- Enforces scalability and portability

## Manage Configurations

- Abstracts configurations
- Making changes in a small configuration does not imply refactoring



# Manage model lifecycle



## The picture is pretty, however

- In reality this order is often broken
- Each step can be complex

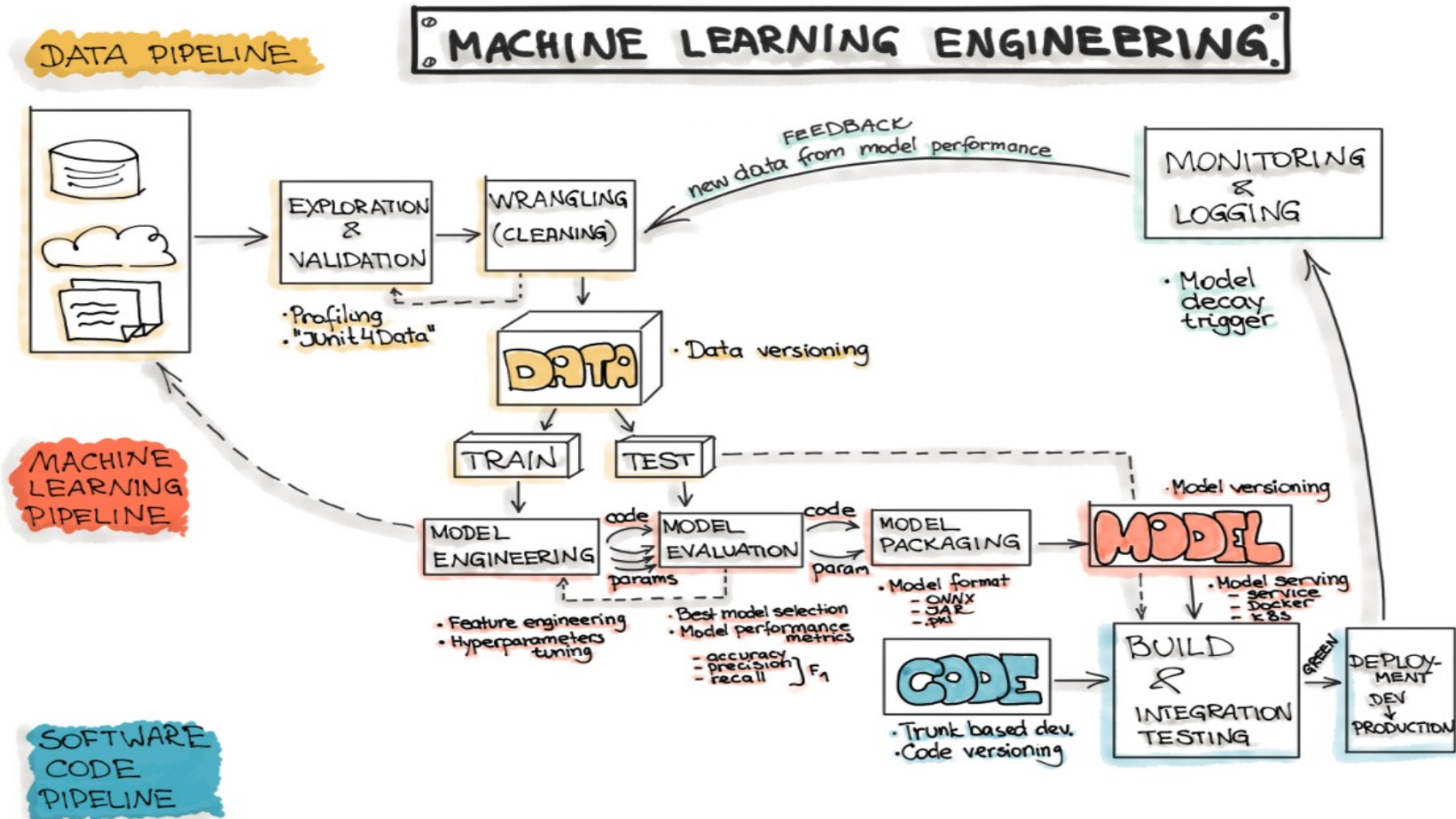
## MLOps to the rescue

- Automate and manage the workflow
- Experiment tracking
- Guidelines for cross-functional teams
- Integrate experimentation with deployment





# Manage the project lifecycle

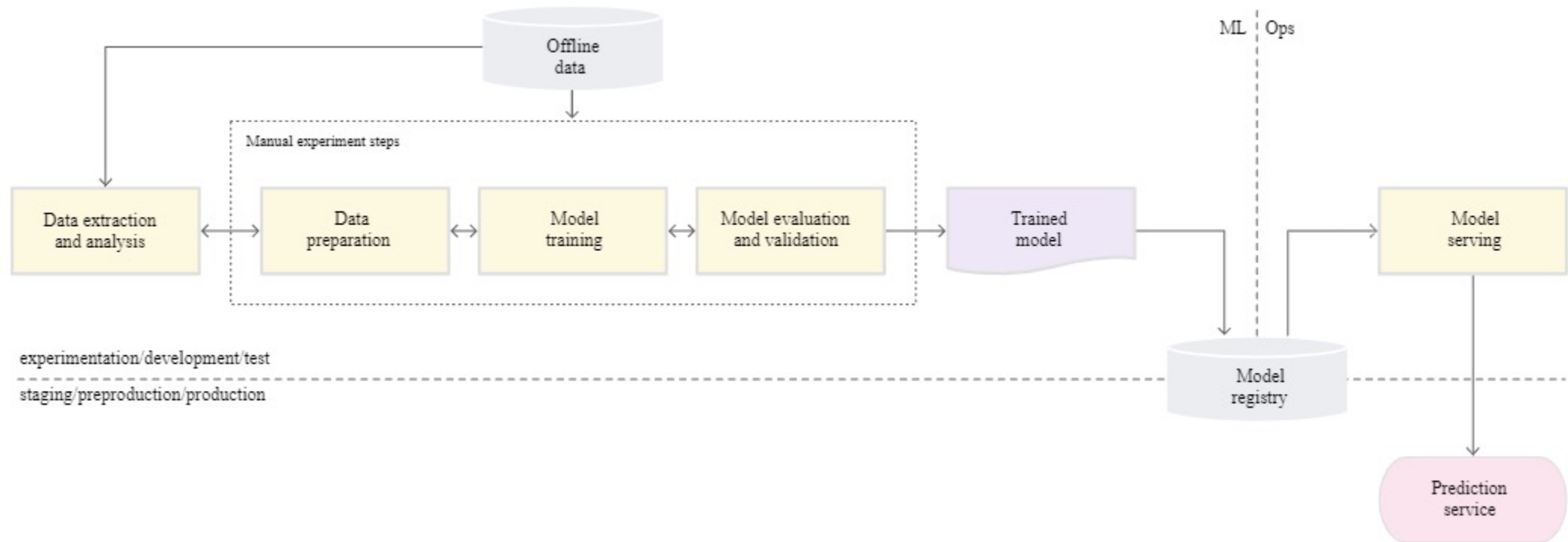






# But is it really necessary

I have a **manual process** that looks like this





# Even if you do

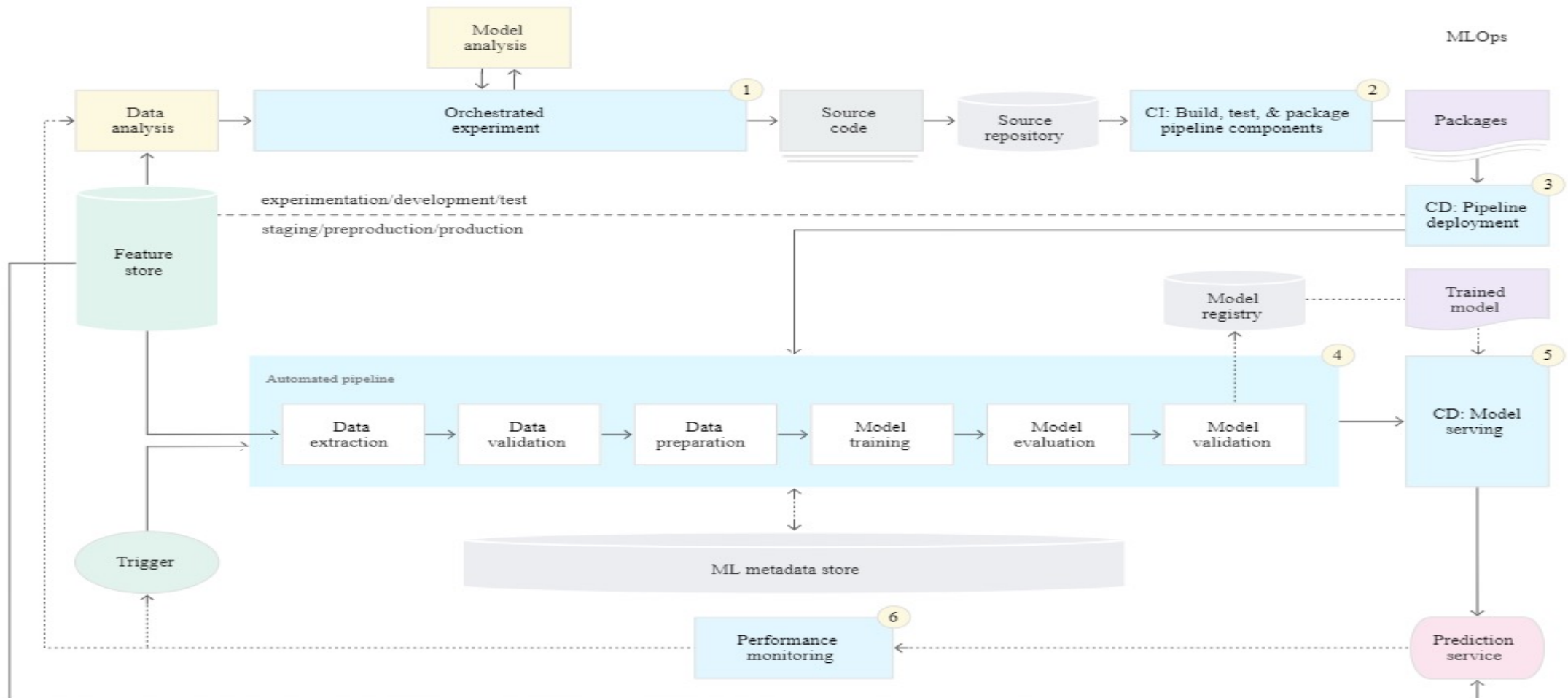
## Even if:

- You have very competent data scientists
- You have a very solid IT department
- They know how to work well together
- They are able to automate model predictions
- And .....

You will still need a **permanent reproducible** record of what happened at every single step + processes to **maintain** and **monitor** your ML system



# Mature framework





# Reduce technical debt

**MLOps main goal is to reduce technical debt**

Some of the question posed by **MLOps**?

- How easily can a new algorithmic approach be tested at full scale?
- What is the transitive closure of all data dependencies?
- How precisely can the impact of a new change to the system be measured?
- Does improving one model or signal degrade others?
- How quickly can new members of the team be brought up to speed?



# Address regulatory challenges



European  
Commission

## Regulation is increasing

- GDPR
- European legal framework for AI

## MLOps related to regulation?

MLOps puts the operations team at the forefront of regulation and best practices

