# Cognizant

## **MLOps Workshop - Kickoff**

Kyllian Broers & Bram van Meurs

October 8<sup>th</sup> 2020 – Version 2.0

## **Workshop Goals**

- 1. Understand the key concepts of MLOps and DevOps
- 2. Be able to explain when and when not to use MLOps

#### You will:

- Get hands-on experience with setting up a DevOps pipeline
- Productionalise a proof of concept ML model, including development and deployment
- Monitor data and detect drift

## The Challenge

#### Develop/Experiment



75% of data science projects never make it to production

## Agenda

- 1. What is DevOps?
- 2. Why MLOps?
- 3. Example within Cognizant
- 4. Workshop contents

## What is DevOps?

### What is DevOps

Way of working

Application lifecycle

Improved **OPERATE** reliability, availability, security

Agile **PLAN** Workflow Collaboration **DEVELOP** Security & Condition of the Condition of Continuous Imphovement APPLICATION **DELIVER** 

Increased quality, stability and productivity

Scalable, repeatable, controllable.
Peace of mind

Cognizant

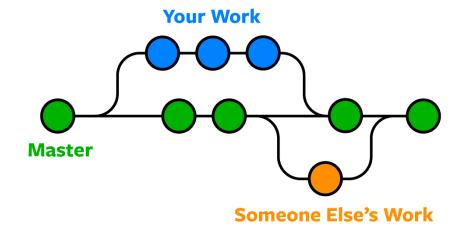
## **Develop**

Programming: Python

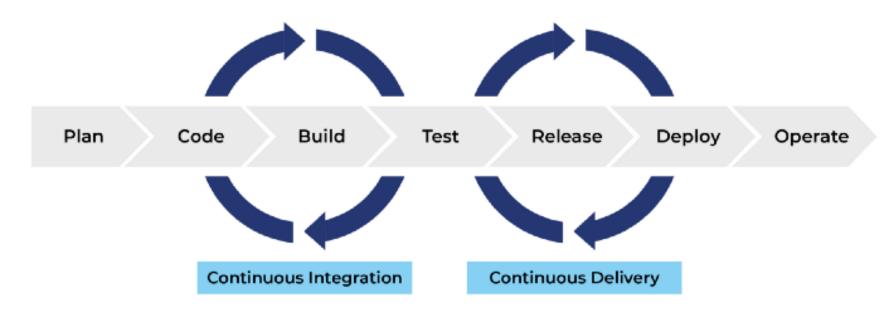
Collaboration: Git

Github

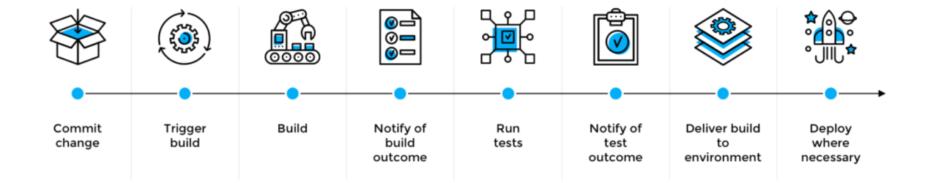
Gitlab



## **DevOps CI/CD**



## **DevOps CI/CD**



## **Benefits of DevOps**

- Shortening the development cycles
- Increasing deployment velocity
- Increasing quality
- Dependable releases
- Ease of rollback

## DevOps and MLOps?

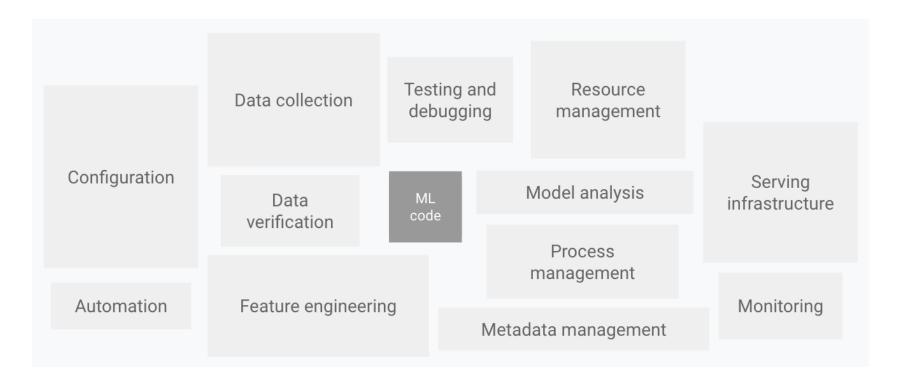
## **DevOps and MLOps**

- Team skills: ML researchers are not experienced software engineers
- Development is experimental
- Testing is more involved
- Deployment includes retraining
- Production: handling evolving data profiles

#### In MLOps

- CI: includes testing and validating data, data schemas and models
- CD: it is about deploying a system
- CT: continuous training

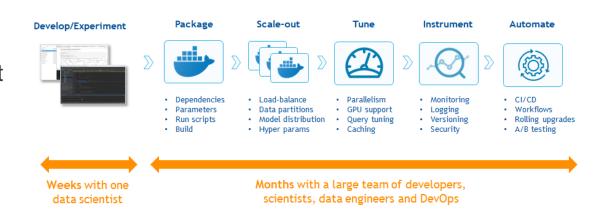
## Parts of a real ML project



# Why MLOps?

### **MLOps Benefits**

- Reproducible model creating, tracking development
- Continuous Training
- Automatic deployment
- Continuous improvement
- Faster time to market
- Higher quality product



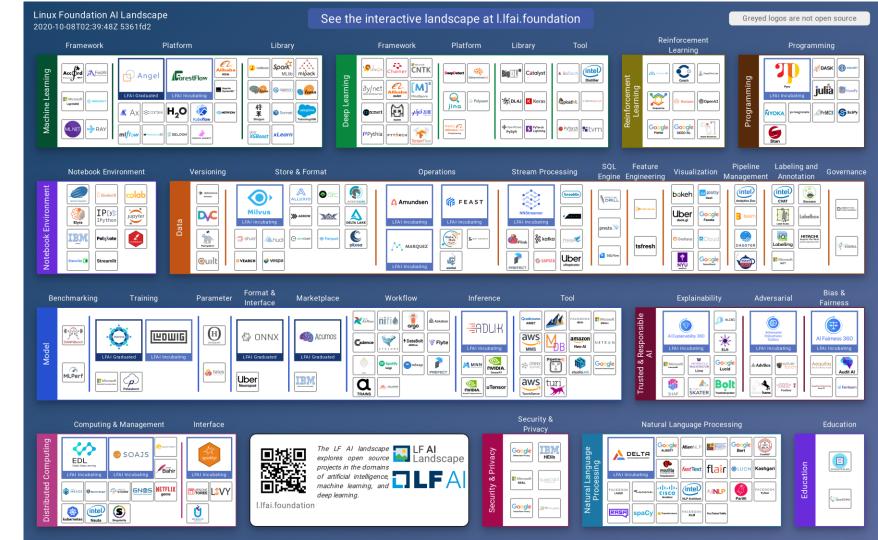
Cloud services & open source tools make MLOps easier

## When not to use MLOps

#### High setup cost

- Develop once, deploy once
- Overhead should not exceed the benefits

## Tools and example

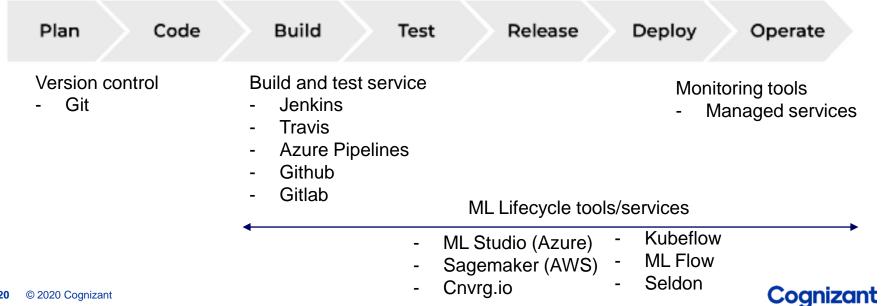


## **DevOps CI/CD**

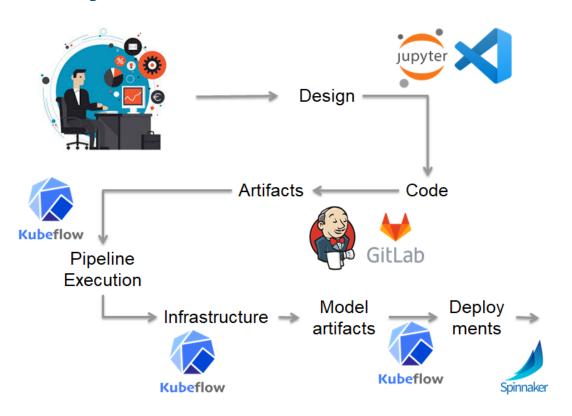
Data storage

Deployment service

- Docker
- Kubernetes / Kubeflow
- Managed service: AWS/Azure



## Macy's



#### **Summary:**

- A data scientist **designs** a model (workflow)
- The design is **coded** using JupyterHub or an IDE and stored in GitLab
- A Continuous Integration (CI) pipeline builds the artifacts automatically with each commit using Jenkins
- One of the artifacts is a Kubeflow pipeline that executes all steps of the workflow
- Kubeflow orchestrates executing of all steps and deploys the steps on the required infrastructure
- 6) Kubeflow stores model artifacts
- A Continous Delivery (CD)
   pipeline updates the services when the model artifacts are refreshed (Spinnaker)

# Workshop

## Workshop

#### Completely in Azure

- As a test to see what azure offers
- Focus is on concepts and general coding
- Azure specific code is avoided as much as possible

#### Assignments

- 1. Getting familiar with Git and DevOps
- 2. Bringing a model from POC to production
- 3. Monitoring the model in production: detecting drift in the data

## **Assignment 1**

Goal: getting familiar with Git and DevOps

- No ML included
- Create a pipeline that performs data analysis
- The pipeline will throw errors: unit tests and linting errors
- Working together with Git
  - Create your own development branch, create pull requests to the assignment-1 branch.

## **Assignment 2**

Goal: going from POC to production using CI/CD

- Transform a notebook into production ready code
- Create a CI pipeline that performs code integration and model training
- Create a CD pipeline that deploys the latest model to container with API

## **Assignment 3**

Goal: monitor a model in production

- Continues at a working assignment 2
- Collect inference data into a dataset
- Use the data monitor tool to detect data drift

### Repo structure

- Each one branch for assignment 1 and 2
- Assignment 3 will branch of assignment 2

Graph	Description	Date	Author	Commit
ور	😢 setup   origin   setup environment	27 Aug 2020	Bram van Me	81f48876
	🔑 assignment-2 origin assignment-2	27 Aug 2020	Bram van Me	423f36ab
	o  assignment-1 origin assignment-1	27 Aug 2020	Bram van Me	b9ea4e80
	😢 documentation   origin   documentation	27 Aug 2020	Bram van Me	42551738

## Workshop Communication

### Communication

#### Use the Teams MLOps Workshop team

- Plenary meeting in general channel
- Team meetings in teams channels
  - Please do not use private calls
- For questions
  - Use mentioning @Bram or @Kyllian, can be both in general or a teams channel.
  - Or add one of us to a channel meeting

### Schedule

09:00 - 10:00 Kickoff

12:30 - 13:00 Discussing Assignment 1

### **Groups**

1: Laura, Vetle, Bart

2: Bas, Stan

3: Timo, Erik

4: Kristoffer, Haoyu

5: Aleksander, Jodie

https://dev.azure.com/MLOpsTraining/group1