MLOps: Put the Data Science Under Control

a crowd surrounds three large columns, in the background the sun rises, Surrealism (DALL-E)



Speaker: Paul PETON

- Lead Tech in Artificial Intelligence
- Data Platform & Al MVP since 2018
- French speaking podcast



• GitHub.com/methodidacte/meetups





Award Categories Al, Data Platform

First year awarded: 2018

Number of MVP Awards:



Everything is against them but they will have to work together...



David Data – Data Scientist



Main goal: to get the best model

- prepares the data
- experiments in a notebook
- plots the results of the various tests

Angie Neer – ML Engineer



Main goal: to guarantee the performance of the service in production

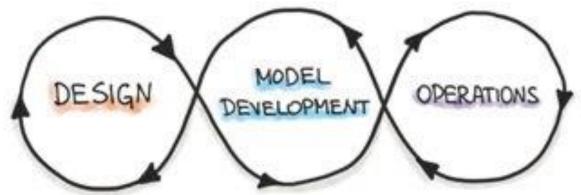
- integrates the deliverables
- deploys new versions
- orchestrates and monitors processes

MLOps: quick definition

Provide an **end-to-end** Machine Learning development process

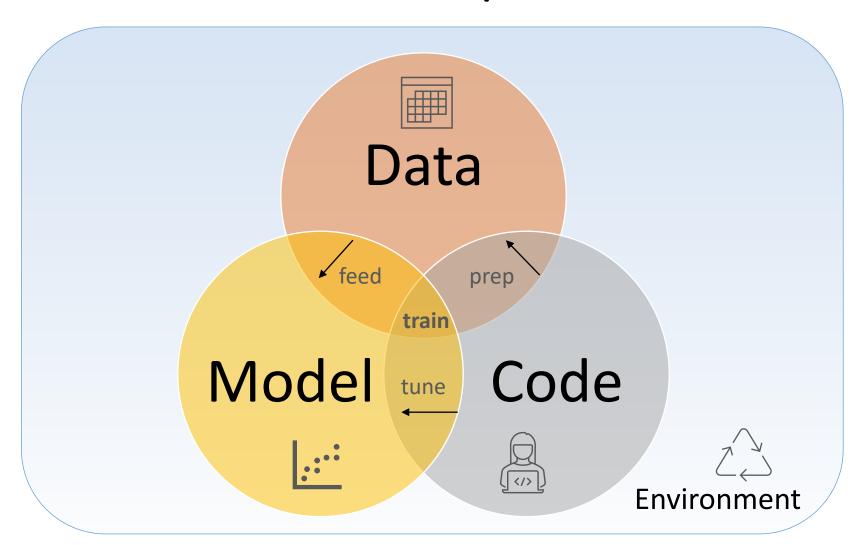
to design, create, and manage

repeatable, testable, and scalable software.



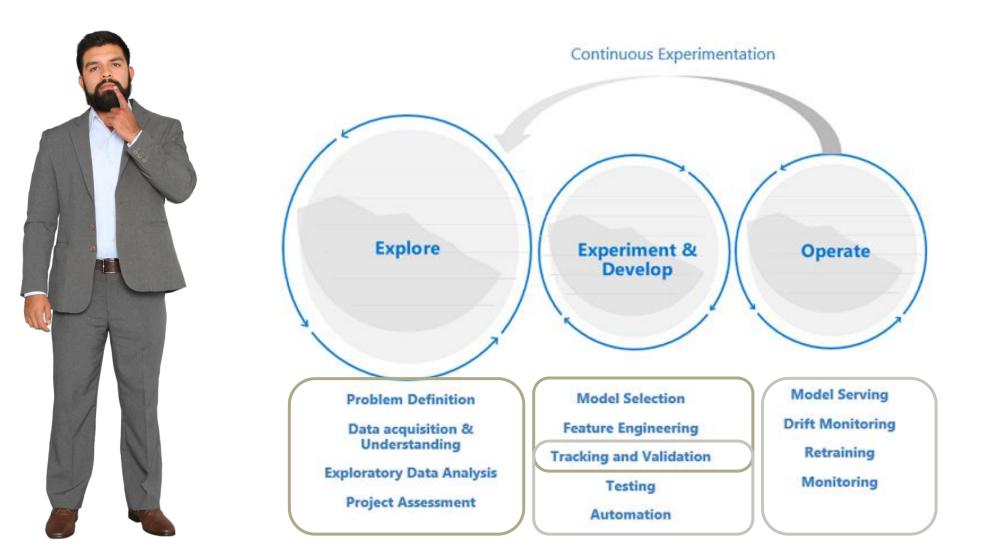
https://ml-ops.org/

What we need to put under control



The environment allows the execution of the code (OS, Python, libraries...).

The three collaborative loops of MLOps

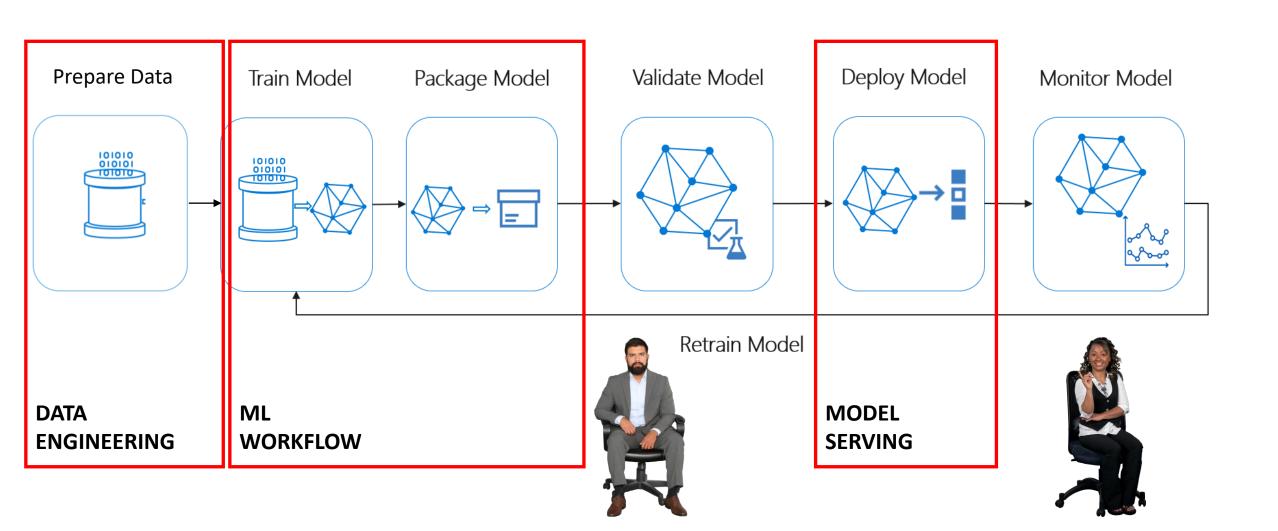




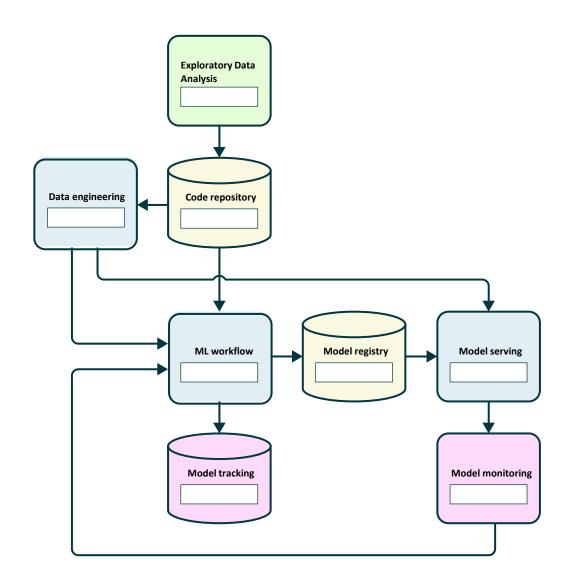
Machine learning DevOps guide - Cloud Adoption Framework | Microsoft Docs

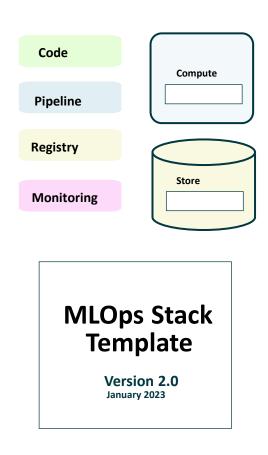


Focus on the three main pipelines



MLOps end to end process





https://valohai.com/blog/the-mlops-stack/https://www.linkedin.com/in/skogstrom/





Experiment with parameters for a Ridge Regression the Diabetes Dataset

This notebook is for experimenting with different parameters to train a ridge regression model on the Diabetes

```
# Change out of the experimentation directory
%cd ..

import azureml.core
from azureml.core import Workspace

+ Code + Markdown

# Load the workspace from the saved config file
ws = Workspace.from_config()

import os, shutil

# Create a folder for the experiment files
training_folder = 'diabetes-training'
```

```
■ Diabetes Ridge Regression Parameter Experimentation.ipynb ×
experimentation > Diabetes Ridge Regression Parameter Experimentation.ipynb > ...
       "cells": [
          "cell_type": "markdown",
          "metadata": {},
          "source": [
           "# Experiment with parameters for a Ridge Regression Model on the Diabetes Da
         "cell_type": "markdown",
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           "This notebook is for experimenting with different parameters to train a ridg
 15
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17
          "cell_type": "code",
          "execution_count": null,
          "metadata": {},
21
          "outputs": [],
          "source": [
          "# Change out of the experimentation directory\n",
           "%cd .."
 25
 26
```

What about notebooks?

- Includes comments, charts, results...
- Can't read the file content in other tool

Improve your code

Reproductibility

Version control

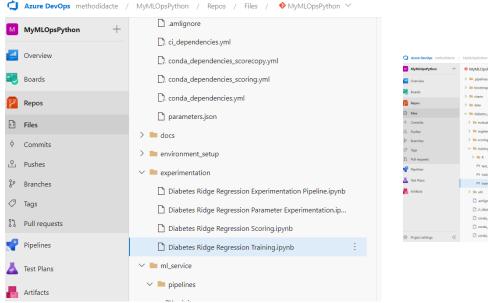
Debug

Test and reuse

CI/CD





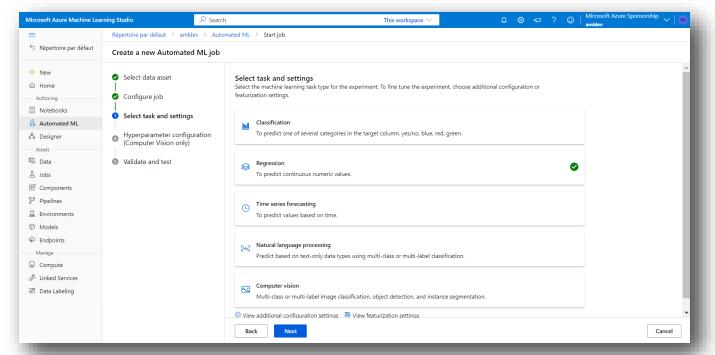


From notebook to python script

- convert the code into functions hosted in scripts
- Use the repository structure to reduce complexity and scale our work

Experiment tracking

• Use automated ML to go quickly to the best model



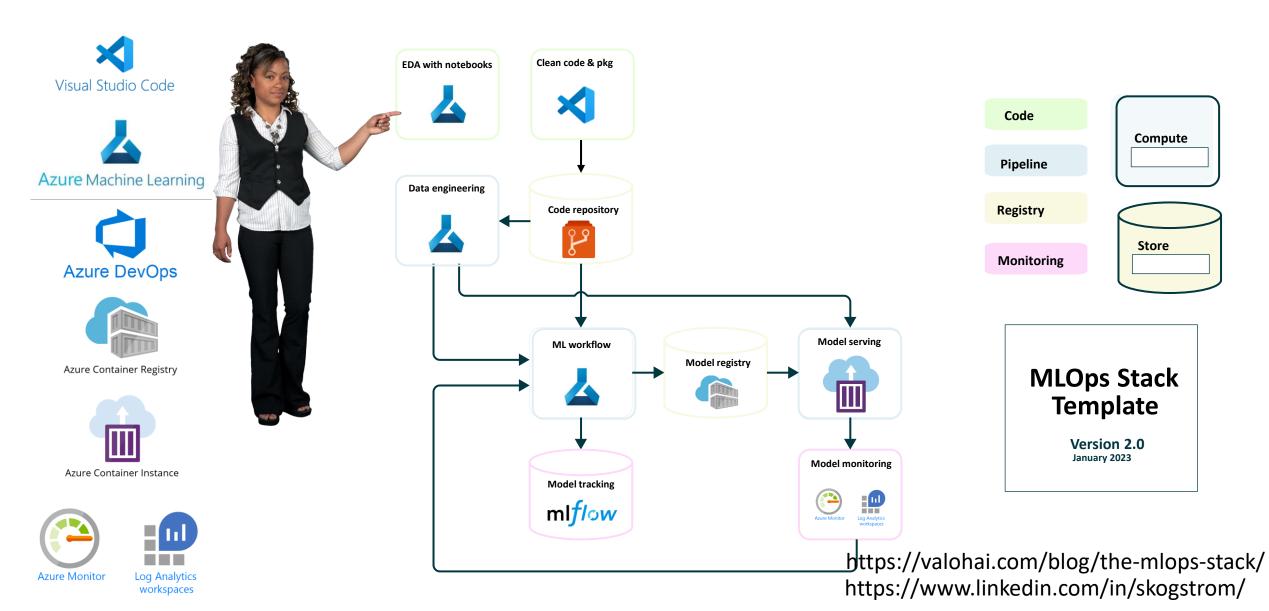
Follow the performance with the evaluation score



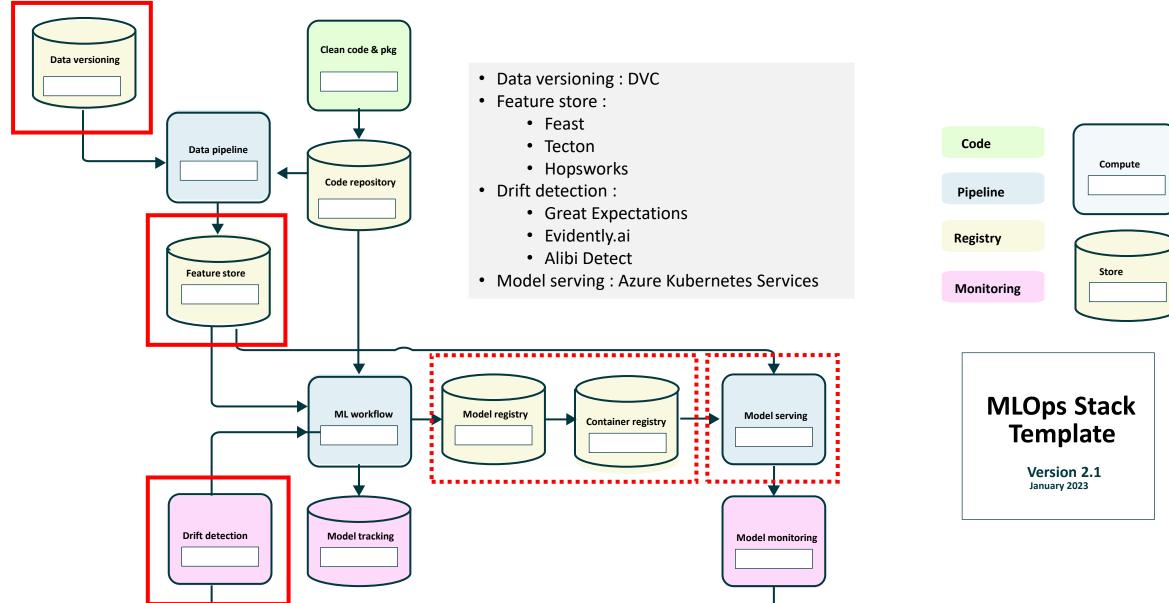
Build the platorm



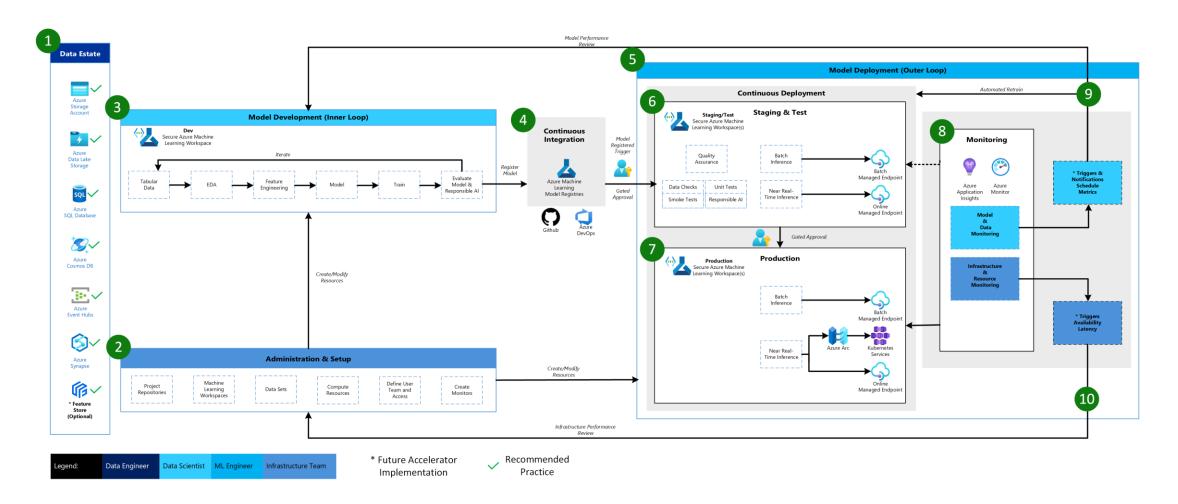
Technical stack components



Technical stack components (advanced version)



MLOps architecture for classical ML on Azure



GitHub - Azure/mlops-v2: Azure MLOps (v2) solution accelerators



DEMO

Automate the deployments



Conclusion : better together !



Share a culture of **software engineering** (clean code, testing) and DevOps (automation)

Advance by **maturity level** (registers, automation, monitoring)

The platform will become **exponentially more efficient** as new use cases are implemented.

