

Description:

Part 1 - Data Governance:

- Make an initial setup using `Data version control` tool. Add a dataset, so another could obtain it via dvc pull after cloning your repository
- Define a DVC pipeline that will:
 - preprocess data
 - train a model
 - evaluate the model
 - generate a feature importance plot with a model agnostic method
- The pipeline should be reproducible using dvc repro
- Run experiments (for example, different Scaling, models) using development environment from previous step and save metrics using dvc metrics

Part 2 - CICD, testing:

- Create unit tests for python code from Part 1
- Create a github action which at least performs:
 - code quality check:
 - auto-formatting with black
 - linting with pylint - fail if less than a threshold [example](#)
 - run unit tests

Criteria:

- DVC pipeline defined in a simple, reproducible manner
- There is an existing remote from which one could pull data (use free tier of AWS/GCP, Google Drive, or any other that would be easy to share)
- Code style / code quality tools are used
- Use github actions for CICD

Materials:

Data governance & CICD:

DVC

- [Intro to CI with DVC, CML and GithubActions](#)
- [DVC for data versioning](#)
- [DVC Pipelines+githubActions](#)

Github Actions

- [GA docs](#)

Testing & monitoring:

- [intro](#)

- DS testing & monitoring (some examples from Google):
 - [concept of drift](#)
 - [behavioral model test](#)

Libraries for testing:

- code testing:
[pytest](#) | [unittest](#)
- code quality (auto-formatting + testing):
[black](#) & [pylint](#)

Explainable ML:

[model agnostic methods for feature importance](#)