ASSIGNMENT 3.4 (a) and (b) on

Model Deployment and Maintance

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All files related to Mlos and train.py are uploaded on my github repository.

Assignment 3.4 (a):

Task 01:

Write a component that will log metadata of your Classification model that you trained on the day dedicated to Supervised Learning. Remember to include all metadata that are important to track for this problem.

```
def track_with_mlflow(model, X_test, Y_test, mlflow, model_metadata):
    mlflow.log_params(model_metadata)
    mlflow.log_metric("accuracy", model.score(X_test, Y_test))
    mlflow.sklearn.log_model(model, "rfc", registered_model_name="sklearn_rfc")
```

Assignment 3.4 (b):

Task 01:

Run your Classification model that you trained on the day dedicated to Supervised Learning in MLFlow.

```
mlops > ♣ train.py > ⊕ main

1 import fire
2 import mlflow
3 import pandas as pd
4 from sklearn.ensemble import RandomForestClassifier
5 from sklearn.preprocessing import StandardScaler
6 from sklearn import datasets
7 from sklearn.pipeline import make_pipeline
8 # from sklearn.metrics import fl_score
9 from sklearn.model_selection import train_test_split

10
11
```

```
mlops >  train.py >  main

idef split_data(wine_data):

wine_x = wine_data.iloc[:,:6]
    wine_y = wine_data['quality']
    x_train, x_test, y_train, y_test = train_test_split(wine_x, wine_y, test_si

return x_train, x_test, y_train, y_test

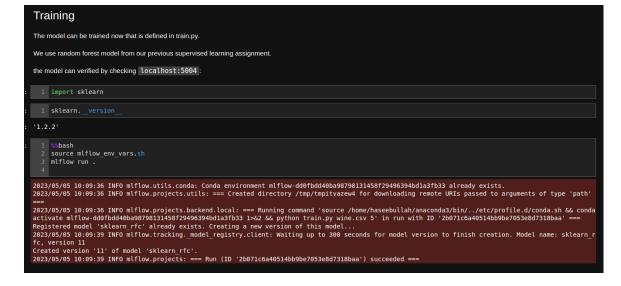
def setup_rfc_pipeline(max_depth):
    rfc_model = RandomForestClassifier(max_depth)
    rfc_pipe = make_pipeline(StandardScaler(), rfc_model)
    return rfc_pipe

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```

```
def track_with_mlflow(model, X_test, Y_test, mlflow, model_metadata):
    mlflow.log_params(model_metadata)
    mlflow.log_metric("accuracy", model.score(X_test, Y_test))
    mlflow.sklearn.log_model(model, "rfc", registered_model_name="sklearn_rfc")
```

MLFlow lab file

```
× ≣ MLproject
                                                 × ≣ train.py
MLFlow Lab.ipynb
# B + % □ □ > ■ C >> Code
                                             y 🕓 git
          Setting up MLFlow tracking server
          Server can be accessed on localhost:5004
            1 %bash --bg
              mlflow server --host 0.0.0.0 \
                  --default-artifact-root ./mlruns
          MLProject file
          Here we have defined entry points and configuered MI flow steps.
          Defined parameters of main function, file and max depth
    [5]: 1 %cat MLproject
          name: basic_mlflow
          # this file is used to configure Python package dependencies.
          # it uses Anaconda, but it can be also alternatively configured to use pip.
          conda_env: conda.yaml
          # entry points can be ran using `mlflow run roject_name> -e <entry_point_name>
          entry_points:
           main:
              # parameters is a key-value collection.
              parameters:
               file_name:
                 type: str
                 default: "wine.csv"
               max_depth:
                 type: int
                 default: 5
              command: "python train.py {file_name} {max_depth}"
```




```
Now predicting the quality of wine
      warning: this might fail at first because the prediction server didn't spin up; in this case wait a minute
          : data='[[7.4,0.7,0,1.9,0.076,11], [7.4,0.7,0,1.9,0.077,10]]'
3 echo data
        5 curl -d "{\"inputs\": $data}" -H 'Content-Type: application/json' 127.0.0.1:5005/invocations
      [[7.4,0.7,0,1.9,0.076,11], [7.4,0.7,0,1.9,0.077,10]]
     % Total % Received % Xferd Average Speed Time Time Time Current
Dload Upload Total Spent Left Speed
100 87 100 23 100 64 7504 20880 ------ 29000
{"predictions": [5, 5]}
          data='[[7.4,0.7,0,1.9,0.076,11], [0.0,0.1,1,1.9,0.077,10]]'echo $data
        5 curl -d "{\"instances\": $data}" -H 'Content-Type: application/json' 127.0.0.1:5005/invocations
     [[7.4,0.7,0,1.9,0.076,11], [0.0,0.1,1,1.9,0.077,10]]
     % Total % Received % Xferd Average Speed Time Time Time Current
Dload Upload Total Spent Left Speed
100 90 100 23 100 67 10962 31935 ------ 45000
{"predictions": [5, 6]}
        1 Wbash
2 data='[[0.4,0.7,1,0.5,0.076,11], [0.0,0.1,0,1.9,0.000,13]]'
3 columns='["fixed acidity", "volatile acidity", "citric acid", "residual sugar", "chlorides", "free sulfur dioxide"]'
4 echo s'data
        6 curl -d "{\"dataframe_split\":{\"columns\":{\"fixed acidity\", \"volatile acidity\", \"citric acid\", \"residual sugar\",\"chlorides\", \"free s
      [[0.4,0.7,1,0.5,0.076,11], [0.0,0.1,0,1.9,0.000,13]]
     % Total % Received % Xferd Average Speed Time Time Current Dload Upload Total Spent Left Speed 100 221 100 23 100 198 6104 52547 ------ 73666 {"predictions": [4, 7]}
      Hurrah!, Now we can see the prediction of our model :-).
[]: 1
                                                                                                                                                                           ⑥↑↓占♀ⅰ
```

Updated ML project file

```
Help
//LFlow_Lab.ipynb
                    × ≣ MLproject
                                                     ≣ train.py
name: basic mlflow
 # this file is used to configure Python package dependencies.
 # it uses Anaconda, but it can be also alternatively configured to use pip.
 conda_env: conda.yaml
 # entry points can be ran using `mlflow run <project name> -e <entry point name>
 entry_points:
   main:
     # parameters is a key-value collection.
     parameters:
      file name:
         type: str
         default: "wine.csv"
       max_depth:
         type: int
         default: 5
     command: "python train.py {file name} {max depth}"
```

Local host 5004: Mlflow

