

An End-to-End Data Science Project

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Workshop overview:

Session 1 Preparation 10.04.2022

Start with the business problem, find data source, preprocess data, set up team process and tech

Session 2 Analytics 17.04.2022

Analyze and understand your data. Gain insights and prepare for the predictive modeling

Session 3 Machine learning 08.05.2022

Build and evaluate prediction model(s), use Mlflow to keep track of the various experiments

Session 4 Production 15.05.2022

Create prediction functions and production class, develop an API, create a dashboard that the user will access and call the API

What you will do:

- Form a team of 3 members
- During the sessions: You will get tasks to be done
- After the sessions:
 - You will complete the whole covered phases
 - Dig deeper into the various technologies discussed



Part 1: Modelling Training

- 1. Clean your data
- 2. Decide on your modelling strategy
- 3. Decide about the evaluation metric
- 4. Train a baseline model
- 5. Train more sophisticated models
- 6. Decide which model will be used



Summary

```
In [9]: # Initialize client and experiment
    client = MlflowClient()
    mlflow.set_experiment(EXPERIMENT_NAME)
    exp = client.get_experiment_by_name(EXPERIMENT_NAME)

In [10]: # Start a new run and track
    with mlflow.start_run(experiment_id=exp.experiment_id):
        mlflow.log_param("pca_var", PCA_VAR) # Track model parameter
        mlflow.log_metric("MSE", mse_test) # Track error value
        mlflow.log_artifact(EXPORT_MODEL_PATH) # Track exported model
```

```
Retrieve experiment
In [3]: # Initialize client
        client = MlflowClient()
        # Get experiment
        exp = client.get_experiment_by_name(EXPERIMENT_NAME)
        runs = mlflow.search_runs([exp.experiment_id])
Out[4]:
           run id
                                              experiment id status
                                                                       artifact_uri
                                                                                                                 metrics.MSE params.pca var
         0 41e1628508fc4a2f83651ceceede6d8a
                                                                                                                 57.884313
                                                             FINISHED | file:///home/deena_gergis/mlflow_illustartion/.
         1 d39decb7e183450d87f679fb044c6e66
                                                             FINISHED file:///home/deena gergis/mlflow illustartion/...
                                                                                                                 31.675892
         2 77c97ff7b2fe46d9becba4c230dd3193
                                                             FINISHED file:///home/deena_gergis/mlflow_illustartion/..
                                                                                                                 31.831237
         3 f11b91bf1c7441c5bacc7e73adfbbe59
                                                              FINISHED file:///home/deena_gergis/mlflow_illustartion/...
                                                                                                                 57.884313
```

Tutorial: Repo:

https://www.linkedin.com/pulse/mlflow-better-way-track-your-models-deena-gergis/ https://github.com/Deena-Gergis/mlflow_tracking



Assignment:

Train, evaluate and track your models



Questions?