## **Group B - DVC Experiment Tasks**

We will experiment with two different algorithms, RandomForestRegression and Linear Regression. A reproducible pipeline has been created and the stages are stored in the "dvc.yaml". You will use the \$dvc exp run" command to execute a new run whenever you make changes to the pipeline files.

Explore the dvc.yaml file where the experiment stages are defined. The first stage of the pipeline is preparing, and that will load the dataset and split them into train and test .csv files. The creating features stage will extract features from the dataset. The train\_model will train the model and output a model. Finally, the evaluation stage will generate the evaluation metrics and output them as a JSON file.

DVC supports flexible ways of tracking and managing ML experiments. In this controlled experiment, we are only interested in how DVC helps with tracking and querying variations of an experiment pipeline. Hence, it is not required to make our experiment persistent to Git.

DVC experiment commands reference

### Preparation

#### **Initialize DVC**

- 1. Go into the dvc-experiment-b folder.
- 2. Initialise Git

```
$ git init
```

3. Initialise DVC: This will initialise DVC in your project and will add the DVC folder in the repo.

```
$ dvc init
```

4. Committing the changes.

```
$ git add .
$ git commit -m "Add DVC config"
```

# Experimenting with algorithm 1 (Linear Regression)

Carry out the following experimental runs using different sets of dataset features and learning parameters as described below. Execute the python script at each run to train and evaluate the model performance by running \$dvc exp run.

Use the -n <name> option of the \$dvc exp run option to add a readable name for each run. After each run, you can use the \$dvc exp show or \$dvc exp list to see completed runs.

- Run 1: Open the terminal and navigate into the dvc-experiment-b folder. What you're going to do first is reproduce the dvc.yaml by running the following command:
   \$ dvc exp run -n run1
- Run 2: In the params.yaml file, change the Linear Regression normalize parameter to False, and run the experiment again.

\$ dvc exp run -n run2

- 3. **Run 3**: In the Diabetes dataset we have a total of 10 features:
  - age age in years
  - sex
  - bmi body mass index
  - bp average blood pressure
  - s1 tc, total serum cholesterol
  - s2 ldl, low-density lipoproteins
  - s3 hdl, high-density lipoproteins
  - s4 tch, total cholesterol / HDL
  - s5 ltg, possibly log of serum triglycerides level
  - s6 glu, blood sugar level

Use the data feature "s1" instead of "age" in create\_features.py on line 13, and run the experiment again

\$ dvc exp run -n run3

- 4. **Run 4**: Modify the ratio of training to test data. Open the params.yaml file and change the split amount from 0.2 to 0.3.
  - \$ dvc exp run -n run4
- 5. **Run 5:** We will now use more features to train the model. Add "s2" in line 13 of create\_features.py. Also, set the "normalize" parameter to True, that's in train\_model.py \$ dvc exp run -n run5

 Run 6: Add even more features, add the following features to line 13 in create\_features.py, and run the experiment 'bmi', 'bp'

\$ dvc exp run -n run6

7. **Run 7:** Revert back to the "run3" using the following command: \$ dvc exp apply run3

Now change the seed value that's in "params.yaml" to 34556, and see if it affects the result. Run the experiment again \$ dvc exp run -n run7

### Experimenting with algorithm 2 (RandomForestRegression)

Now we try a different algorithm by

- Commenting line 16 and uncommenting line 17 in train\_model.py
- Commenting the Ir\_parameters and uncommenting the rfr\_parameters lines in the "params.yaml" file.
- 8. **Run 8:** Execute the experiment using the RFR algorithm.

\$ dvc exp run -n run8

- 9. **Run 9**: We are now going to use the following parameters:
  - n estimators
  - max\_depth
  - min\_samples\_split
  - ccp alpha

In train\_model.py line 17, replace the **RandomForestRegressor()**, with the following code:

```
# Copy this
model = RandomForestRegressor(
    n_estimators=params["n_estimators"], max_depth=params["max_depth"],
    min_samples_split=params["min_samples_split"],
ccp_alpha=params["ccp_alpha"], max_features=params["max_features"],)
```

- 10. **Run 10:** In the params.yaml change the ccp\_alpha and max\_depth to 0.1 and 10 respectively.
  - \$ dvc exp run -n run10
- 11. **Run 11:** Now we will add more features to train the model. Replace the features in line 13 of create\_features.py as shown below:

```
def extract_features(df):
    return df[['s1', 's6', 's2', 'bmi', 'bp']]
```

\$ dvc exp run -n run11

- 12. **Run 12:** Let's try to improve the model. The default value of max\_features is None. Change the max\_features to **log2** in the params.yaml file. \$ dvc exp run -n run12
- 13. **Run 13:** Revert back to "run8" and include all features in the dataset as described in step 3 above.

\$ dvc exp run -n run13

\*Now return to the experiment's questionnaire