

## Machine Learning Classification for Economic Growth Prediction Analyzing OECD NUTS3 Data

### ▪ What does the code do?

This script applies machine learning classification to predict whether a region's economy grew by more than 5% based on OECD data. The dataset includes economic and demographic indicators for NUTS3 regions in 2019. The script loads and preprocesses the data, splits it into training and test sets, and runs three classifiers: Random Forest, Logistic Regression, and Multilayer Perceptron. Each model's performance is evaluated using accuracy, precision, recall, F1-score, and a confusion matrix.

### ▪ How to use the code?

To run the code, ensure you have *Python* installed with required libraries (*pandas*, *numpy*, *sklearn*). Place the dataset files (*assignment\_4\_OECD\_data\_growth.csv* and *assignment\_4\_OECD\_data\_explaining\_vars.csv*) in the same directory as the script. Execute the script (`python script_name.py`), and it will automatically load and process the data, train the classifiers, and display the results. The output includes classification metrics and a confusion matrix, helping to evaluate each model's performance in predicting economic growth.

### ▪ Interpretation of the results

#### Results for MLP Classifier

Precision: 0.7972972972972973

Recall: 0.8872180451127819

F1 score: 0.8398576512455516

Accuracy: 0.7680412371134021

Confusion matrix:

```
[[ 31 30]
```

```
[ 15 118]]
```

#### Results for Logistic Regression

Precision: 0.7245508982035929

Recall: 0.9097744360902256

F1 score: 0.8066666666666666

Accuracy: 0.7010309278350515

Confusion matrix:

```
[[ 15 46]
```

```
[ 12 121]]
```

#### Results for Random Forest

Precision: 0.75

Recall: 0.924812030075188

F1 score: 0.8282828282828283

Accuracy: 0.7371134020618557

Confusion matrix:

```
[[ 20 41]
```

```
[ 10 123]]
```

**MLPClassifier** had the highest accuracy (76.8%) and a strong recall (88.7%), meaning it correctly identified most growing regions. However, its false positive rate (30 cases) indicates that it often misclassified non-growing regions as growing.

**Random Forest** showed a well-balanced performance with the highest recall (92.5%), meaning it detected nearly all growing regions. Its accuracy (73.7%) was slightly lower than MLP, and while it had fewer false negatives (10), it misclassified 41 non-growing regions as growing.

**Logistic Regression** had the lowest accuracy (70.1%) and the highest false positive rate (46 cases), meaning it struggled the most in distinguishing non-growing regions. However, its recall (90.9%) was still strong, showing that it was effective at identifying growth cases.

In conclusion, **MLPClassifier** performed best, offering the highest accuracy and a strong balance of precision and recall. Random Forest is a close second, excelling in recall while maintaining solid accuracy. Logistic Regression performed the weakest, as its high false positive rate makes it less reliable for distinguishing non-growing regions.

It is important to note that results may change from one execution to another because of the random initialization of model parameters, stochastic training processes (especially in MLPClassifier), and floating-point precision differences. Additionally, if the random seed isn't fixed, the train-test split may vary, leading to different model evaluations.