

# Machine Learning

## Programming Assignment II-b

Shuai Wang, Marleen de Jonge and dr. Stevan Rudinac

The following assignment will test your understanding of topics covered in the first five weeks of the course. This assignment **will count towards your grade** and should be submitted through Canvas by **28.11.2024 at 23:59 (CET)**. You must submit this assignment in groups (as registered on Canvas). You can get at most 5 points for this assignment, which is 5% of your final grade.

### 1 Instructions

- Alongside the code for your experiments, you are also required to present a report summarizing the observations and results of each of the experiments. You can use text and graphs/plots (`matplotlib`) for these reports. You should place these report blocks within the Jupyter Notebook in separate text cells. Plots can be appropriately placed near the text explanation. Your final submission should be a single Jupyter Notebook with code and report blocks.
- While it is perfectly acceptable to brainstorm and discuss solutions with other colleagues, please, do not copy code. We will check all submissions for code similarity with each other and with openly-available solutions on the web.
- Please, ensure that all code blocks are functional before you finalize your submission. Points will NOT be awarded for exercises where code blocks are non-functional.

### 2 Submission

You can submit your solutions within a Jupyter Notebook (\*.ipynb). To test the code we will use Anaconda Python (3.10). Please, state the names and student ids of the authors at the top of the submitted file.

### 3 Data

For this assignment, you will continue to use the dataset provided with Programming Assignment II-A. If you are attempting this assignment without attempting part II-A, we advise you to go back and look at the previous assignment for more information on the data.

### 4 Models

The model structure introduced in Programming Assignment II-A uses a **Pipeline** to wrap a *scaler* and a *classifier*. Since scaling is not needed for tree-based models, the resulting pipeline will end up containing only the classifier. Thus, for the tree-based models specified in this assignment, the usage of a **pipeline is NOT required**.

In this assignment, you are asked to use the **DecisionTreeClassifier** and **RandomForestClassifier** as the classifiers. With these models, you must perform the following experiments:

✉ s.wang3@uva.nl, m.r.h.dejonge@uva.nl, s.rudinac@uva.nl

**1. Initialize a `DecisionTreeClassifier` model and perform the following tasks:**

- (a) Fit the classifier on the data and report the model accuracy.
- (b) Report all evaluation metrics listed in Section 5.
- (c) Perform a grid search on the decision tree model parameters `['max_depth', 'max_features']` to evaluate the optimal values for these parameters. Report model hyperparameters for the best classifier model.
- (d) Train the decision tree model for tree depths ranging from 1 to 20. Plot the training and validation accuracy on a line chart, with tree depth on the x-axis and accuracy on the y-axis. Analyze the plot to determine at which depth overfitting begins and discuss its implications for model performance. Please, limit your explanation to a maximum of 50 words.

In less than 50 words, explain your observations and your assessment of the classifier. You can use a text box (i.e., Markdown Cell) in Jupyter to write down your analysis.

**2. Initialize a `RandomForestClassifier` model and perform the following tasks:**

- (a) Fit the classifier on the data and report the model accuracy.
- (b) Report all evaluation metrics listed in Section 5.
- (c) Answer the following questions:
  - i. How are decision tree classifiers different from random forests on a structural level? (max. 50 words)
  - ii. Where would you choose decision trees over random forests and vice-versa? Demonstrate this using an appropriate example from your data. (max. 50 words)
  - iii. Is accuracy an appropriate evaluation metric for this classification task? Justify your answer in less than 20 words.
- (d) Perform a grid search on the random forest model parameters `['max_depth', 'max_features', 'n_estimators']` to evaluate the optimal values for these parameters. Report model parameters for your best classifier model.

In no more than 50 words, explain your observations and your assessment of the classifier. You can use a text box (i.e., Markdown Cell) in Jupyter to write down your analysis.

**3. Test the models' robustness under varying noise levels in the dataset by performing the following tasks:**

- (a) Modify the data by introducing increasing levels of noise to the features (e.g., by adding Gaussian noise with standard deviations of 0, 0.1, 0.2, 0.5, and 1.0).
- (b) Train the `DecisionTreeClassifier` and `RandomForestClassifier` models on the original dataset.
- (c) Evaluate the models on the noisy data sets and record the test accuracy for each noise level. Create a plot showing the impact of noise levels on the accuracy, with on the x-axis the noise level and on the y-axis the accuracy.
- (d) Which model is more robust to noise, and why? How does noise affect the generalization performance of the models? Please, justify your answer in no more than 50 words.

## 5 Model Evaluation

For all models in Programming Assignments II-A and II-B, you must report:

- 1. Accuracy
- 2. Macro and Micro-Averaged Precision and Recall

### 3. F1 Score

Additionally, present your observations on what these scores mean for the models under consideration.

## 6 Grading

Component	Points
Decision Trees & Analysis	1.5
Random Forests & Analysis	1.5
Robustness testing & Analysis	1
Evaluation Metrics	1