# **Machine Learning**

2<sup>nd</sup> module, 2<sup>nd</sup> academic year

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Course Website:

**Instructor's Office Hours:** 

**Class Time:** 

**Room Number:** 

**TAs:** [Names and contact information]

### **Course description**

This is a basic course on Machine Learning that introduces most popular methods and approaches to knowledge extraction from data. Lectures will cover such topics as linear models, case-based methods, decision trees and ensembles. We will also have a look at the basics of neural networks. Students will learn how to preprocess data (including categorical and text cases), choose and analyse quality metrics for a particular task, validate and evaluate models. All topics will be covered with practical homework assignments on Python.

## Course requirements, grading, and attendance policies

Class attendance and participation are encouraged, but not required. The course grade will be based on homework assignments (70% of the grade) and the final exam (30%). There will be both theoretical and practical homework assignments.

#### **Course contents**

- 1. Data types, problem statements and general approach in machine learning.
- 2. Basic machine learning concepts. Quality metrics. Model selection, basic loss functions, cross-validation.
- 3. Linear regression model. Analytical solution and numerical approach. Gradient descent for model learning. Overfitting and regularization. Loss functions and dealing with outliers.

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- 4. Linear classification model. Learning classification models with upper bounds on binary loss function. Quality metrics for classification problems. Multi-class and multi-label problems and their reduction to binary classification.
- 5. Decision trees. Greedy learning approach. Impurity functions. Connection between linear models and decision trees.
- 6. Bagging. Bias-variance decomposition. Analysis of bias and variance for bagging. Random forest.
- 7. Gradient boosting on decision trees. Model correction by fitting residuals. Modern implementations of gradient boosting.
- 8. Unsupervised learning. Data clustering.
- 9. Data visualization. t-SNE method.

## Sample tasks for course evaluation

Write down a linear classification model. How many parameters are there? How to evaluate model quality for an imbalanced dataset? How to fit this model on a large dataset? How to prepare a dataset to speed up gradient descent? How to apply cross-validation to prevent overfitting? How to use regularization here, how to choose regularization strength?

#### **Course materials**

There is no required textbook, but selected chapters from Hastie-Tibshirani-Friedman "The Elements of Statistical Learning" will be useful.

## Academic integrity policy

Cheating, plagiarism, and any other violations of academic ethics at NES are not tolerated.

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