

DOI-MBD-514 Machine Learning I

SEMESTER: Fall
CREDITS: 6 ECTS
LANGUAGE: Spanish (documentation in English)
DEGREES: MBD

Course overview

The purpose of this course is to provide students with a fundamental understanding and an extensive practical experience of how to extract knowledge from an apparently unstructured set of data.

By the end of the course, students will:

- Understand the basic principles behind Machine Learning.
- Have practical experience with the most common Machine Learning algorithms.
- Have well-formed criteria to choose the most appropriate technique for a given application.

Prerequisites

Students willing to take this course should be familiar with linear algebra, basic probability and statistics, and R programming language.

Course contents

Theory

1. Introduction
 - 1.1. Data Mining and Machine Learning
 - 1.2. The learning process
 - 1.3. Types of Machine Learning

2. Classification methods
 - 2.1. The classification problem
 - 2.2. Logistic Regression
 - 2.3. Discriminant analysis
 - 2.4. K-Nearest Neighbors
 - 2.5. Decision Trees
 - 2.6. Support Vector Machines
 - 2.7. Multilayer Perceptrons for classification

3. Regression methods
 - 3.1. The regression problem
 - 3.2. Linear regression. Model selection and regularization.
 - 3.3. Polynomial regression
 - 3.4. Splines
 - 3.5. Generalized Additive Models
 - 3.6. Multilayer Perceptrons for regression
 - 3.7. Radial Basis Function Networks

4. Time series Forecasting
 - 4.1. Stochastic Processes
 - 4.2. Exponential Smoothing
 - 4.3. Decomposition methods
 - 4.4. ARIMA models
 - 4.5. Dynamic Regression models
 - 4.6. GARCH models
 - 4.7. Advanced methods for forecasting

5. Unsupervised Learning
 - 5.1. Probability Density estimation
 - 5.2. Dimensionality Reduction Methods
 - 5.3. Clustering and Vector Quantization
 - 5.4. Self-Organizing Feature Maps

Practice

Each session will be divided in two parts: an introduction of the theoretical concepts and a lab practice using R. Students will work in pairs and will be asked to deliver a report after each lab practice.

Textbook

- Notes prepared by the lecturer (available in Moodle)
- G. James *et al.* (2013), *An Introduction to Statistical Learning*, Springer, ISBN-13 978-1-4614-7137-0

Grading

- **Exams** will account for 50%, of which:
 - Mid-term exams: 15%
 - Final exam: 35%
- **Lab** will account for the remaining 50%, of which:
 - Student work at lab: 10%
 - Lab reports: 40%
- In order to pass the course, the mark of the final exam must be greater or equal to 4 out of 10 points.