



TECHNICAL SHEET OF THE SUBJECT

| Data of the subject | |
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| Subject name | Machine Learning |
| Subject code | E000009337 |
| Quarter | Semestral |
| Credits | 6,0 ECTS |
| Type | Obligatoria |
| Department | Department of Industrial Organization |
| 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 relevant machine learning algorithms. § Have well-form criteria to choose the most appropriate techniques for a given application. |

| Teacher Information | |
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| Teacher | |
| Name | Mario Castro Ponce |
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SPECIFIC DATA OF THE SUBJECT

| Contextualization of the subject |
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| Prerequisites |
| Students willing to take this course should be familiar with linear algebra, basic probability and statistics, and undergraduate-level programming. Previous experience with the R programming language is also desired although not strictly required. |

Competencies - Objectives

THEMATIC BLOCKS AND CONTENTS

| Contents - Thematic Blocks |
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| Contents |
| Unit 1. Introduction |

- 1.1 Data mining & machine learning
- 1.2 The learning process
- 1.3 Smart industry levers and drivers
- 1.4 Types of machine learning

Unit 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

Unit 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

Unit 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

Unit 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

TEACHING METHODOLOGY

General methodological aspects of the subject

EVALUATION AND CRITERIA

Ratings

Regular assessment

Theory will account for 50%, of which:

- Midterm:15%
- Final exam:35%

Lab will account for the remaining 50%

In order to pass the course, the weighted average mark must be greater or equal to 5 out of 10 points, and the mark of the final exam must be greater or equal to 4 out of 10 points. Otherwise, the final grade will be the lower of the two marks.

Retake

Lab marks will be preserved. In addition, all students will take a final exam. The resulting grade will be computed as follows:

Final exam: 50%

Lab practices: 50%

As in the regular assessment period, in order to pass the course, the weighted average mark must be greater or equal to 5 out of 10 points, and the mark of the final exam must be greater or equal to 4 out of 10 points. Otherwise, the final grade will be the lower of the two marks.

BIBLIOGRAPHY AND RESOURCES

Basic Bibliography

-Slides prepared by the lecturer (available in Moodlerooms).

-G. James, D. Witten, T. Hastie, and R. Tibshirani, An Introduction to Statistical Learning with Applications in R, Springer, 2013.