

# Syllabus 2019 - 2020

# **TECHNICAL SHEET OF THE SUBJECT**

Data of the subject		
Subject name	Machine Learning / Analítica Avanzada	
Subject code	E000009337	
Quarter	Semestral	
Credits	6,0 ECTS	
Туре	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**

# SPECIFIC DATA OF THE SUBJECT

### **Contextualization of the subject**

### 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	
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

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- 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



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#### **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.