Course guide
295102 - 295II012 - Data Analysis & Pattern Recognition

Last modified: 26/06/2025

Unit in charge: Barcelona East School of Engineering
Teaching unit: 723 - CS - Department of Computer Science.
749 - MAT - Department of Mathematics.
707 - ESAII - Department of Automatic Control.

Degree: MASTER'S DEGREE IN CHEMICAL ENGINEERING (Syllabus 2019). (Compulsory subject).
MASTER'S DEGREE IN INTERDISCIPLINARY AND INNOVATIVE ENGINEERING (Syllabus 2019). (Compulsory subject).
ERASMUS MUNDUS MASTER IN SUSTAINABLE SYSTEMS ENGINEERING (EMSSE) (Syllabus 2024). (Optional subject).

Academic year: 2025 **ECTS Credits:** 6.0 **Languages:** English

LECTURER

Coordinating lecturer:

Mata Miquel, Christian

Others: Primer quadrimestre:
GERARD ESCUDERO BAKX - Grup: T11, Grup: T12
CRISTIAN MATA MIQUEL - Grup: T11, Grup: T12
FRANCESC POZO MONTERO - Grup: T11, Grup: T12
BARDIA RAFIEIAN - Grup: T11, Grup: T12

DEGREE COMPETENCES TO WHICH THE SUBJECT CONTRIBUTES

Specific:
CEMUEQ-01. To apply knowledge of mathematics, physics, chemistry, biology and other natural sciences, obtained through study, experience and practice, with critical reasoning, to establish economically viable solutions to technical problems
CEMUEQ-03. Conceptualize engineering models, apply innovative methods in the resolution of problems and adequate computer applications, for the design, simulation, optimization and control of processes and systems
CEMUEII-02. Apply techniques of pattern recognition, artificial intelligence and statistical data analysis that allow decisions to be made objectively, quantitatively and reproducibly in problems of a multidisciplinary nature.

Generical:
CGMUEQ-04. To carry out the appropriate research, undertake the design and manage the development of engineering solutions, in new or little known environments, relating creativity, originality, innovation and technology transfer
CGMUEQ-05. Know how to establish mathematical models and develop them through appropriate information technology, as a scientific and technological base for the design of new products, processes, systems and services, and for the optimization of others already developed
CGMUEII-01. Participate in technological innovation projects in multidisciplinary problems, applying mathematical, analytical, scientific, instrumental, technological and management knowledge.

Transversal:
05 TEQ. TEAMWORK. Being able to work as a team player, either as a member or as a leader. Contributing to projects pragmatically and responsibly, by reaching commitments in accordance to the resources that are available.
06 URI. EFFECTIVE USE OF INFORMATION RESOURCES. Managing the acquisition, structure, analysis and display of information from the own field of specialization. Taking a critical stance with regard to the results obtained.
03 TLG. THIRD LANGUAGE. Learning a third language, preferably English, to a degree of oral and written fluency that fits in with the future needs of the graduates of each course.

TEACHING METHODOLOGY

The methodology of the course combines theory lessons, laboratory sessions and autonomous learning through the development of projects and the analysis of real applications.

LEARNING OBJECTIVES OF THE SUBJECT

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STUDY LOAD

Type	Hours	Percentage
Hours small group	27,0	18.00
Hours large group	27,0	18.00
Self study	96,0	64.00

Total learning time: 150 h

CONTENTS

Exploratory data analysis

Description:
• Data visualization (histograms, box-plot, qq-plot, multi-dimensional scatter plots, etc)
• Data clustering (k-means, agglomerative, Gaussian Mixture Models)
• Dimensionality reduction and Principal Component Analysis
• Data representation and feature extraction
• Data metrics, distances, norms, etc.

Related activities:
Laboratory session 1: Data visualization and cluster analysis
Laboratory session 2: Principal Component Analysis and dimensionality reduction

Full-or-part-time: 10h
Theory classes: 6h
Laboratory classes: 4h

Supervised machine learninig

Description:
• Introduction to Classification/Regression problems
• Distance-based methods: kNN & Centroids algorithm
• Probabilistic methods: Naïve Bayes & LDA
• Rule-based methods: Decision Trees & AdaBoost
• Hyperplane-based methods: kernels & SVM

Related activities:
Laboratory session 3: Supervised classification I
Laboratory session 4: Supervised classification II

Full-or-part-time: 12h
Theory classes: 8h
Laboratory classes: 4h

Performance evaluation

Description:
• Type I and type II errors
• Univariate and multivariate hypothesis testing approaches
• Statistical inference and parameter estimation (Maximum-likelihood, Bayesian, bootstrapping)
• Validation procedures: cross-validation; leave-one-out, etc.

Specific objectives:

Related activities:
Laboratory session 5: Hypothesis testing
Laboratory session 6: Model validation procedures

Full-or-part-time: 12h
Theory classes: 8h
Laboratory classes: 4h

Neural networks and deep learning

Description:
• Introduction to artificial neural networks (ANNs) and deep learning
• Feed-forward ANNs for classification and regression
• Training ANNs: backpropagation algorithm, optimization stages, advanced strategies (network complexity, early stopping, dropout, weight regularization)
• Specialized architectures: recurrent neural networks, autoencoders, generative adversarial networks, convolutional neural networks

Related activities:
Laboratory session 7: Artificial Neural Networks
Laboratory session 8: Deep Learning

Full-or-part-time: 10h
Theory classes: 6h
Laboratory classes: 4h

Advanced topics and applications

Description:
Seminars by experts, application projects, analysis of publications, news and trending topics.

Full-or-part-time: 8h
Theory classes: 4h
Laboratory classes: 4h

GRADING SYSTEM

Partial exam 30%
Final exam 30%
Projects and exercises 40%

"Those students who meet the requirements set by the EEBE in their Assessment and Permanence Regulations will be able to access the re-assessment test (https://eebe.upc.edu/ca/estudis/estudis-de-master/documents-masters/assessment-and-academic-progress-regulations-for-bachelors-and-masters-degrees-at-the-eebe.pdf)"

BIBLIOGRAPHY

Basic:

- Bishop, Christopher M. Pattern recognition and machine learning. New York: Springer, cop. 2006. ISBN 9780387310732.
- Duda, Richard O; Hart, Peter E; Stork, David G. Pattern classification. 2nd ed. New York [etc.]: John Wiley & Sons, cop. 2001. ISBN 0471056693.

RESOURCES

Other resources:

Python programming: https://www.python.org/ />Numpy mathematical libraries: http://www.numpy.org/ />Machine learning library: https://scikit-learn.org/stable/ />Graphical representation: https://matplotlib.org/ />Data repository: https://archive.ics.uci.edu/ml/index