Course guides
230202 - CLP - Pattern Classification: Applications in Signal Processing

Unit in charge: Barcelona School of Telecommunications Engineering
Teaching unit: 739 - TSC - Department of Signal Theory and Communications.

Degree: BACHELOR'S DEGREE IN AUDIOVISUAL SYSTEMS ENGINEERING (Syllabus 2009). (Optional subject).
BACHELOR'S DEGREE IN TELECOMMUNICATIONS SCIENCE AND TECHNOLOGY (Syllabus 2010). (Optional subject).
BACHELOR'S DEGREE IN ELECTRONIC SYSTEMS ENGINEERING (Syllabus 2009). (Optional subject).
BACHELOR'S DEGREE IN TELECOMMUNICATIONS SYSTEMS ENGINEERING (Syllabus 2010). (Optional subject).
BACHELOR'S DEGREE IN NETWORK ENGINEERING (Syllabus 2010). (Optional subject).
BACHELOR'S DEGREE IN TELECOMMUNICATIONS TECHNOLOGIES AND SERVICES ENGINEERING (Syllabus 2015). (Optional subject).

Academic year: 2020  ECTS Credits: 6.0  Languages: Catalan, Spanish

LECTURER

Coordinating lecturer: JOSEP VIDAL MANZANO
Others: VERONICA VILAPLANA
MARGARITA CABRERA BEAN
ANTONIO PASCUAL ISERTE

PRIOR SKILLS

Stochastic processes. Signal processing.

REQUIREMENTS

TEACHING METHODOLOGY
LEARNING OBJECTIVES OF THE SUBJECT

The subject gives a review of the pattern classification and recognition from the mathematical point of view and at the same time, applies the methods to several areas of signal processing. The structure of the general problem of pattern recognition (i.e., pre-processing, feature extraction and classification), can be applied to different areas, such as quality control, biomedical applications and diagnosis, communication systems, image processing, and speech recognition. The subject will give a general view of the Bayesian decision theory, maximum likelihood estimation, non-parametric classification techniques and non-supervised learning, with applications to different areas of signal processing, such as classification of biomedical signals, images, signal detection, signal modulation, etc. For each of the selected applications, the work done in class will deal with different classification criteria, in order to analyze the compromise between good performance and computation efficiency of each classifier. The 6 credits of the subject are divided between theoretical classes, and at the same time, practical classes at the laboratory (MATLAB), where the student will develop the selected applications, with emphasis in the applications on medical diagnosis, image processing, and communications. In each of the theoretical parts, the methods and algorithms will be developed so that they can be understood and programmed at the same time. Some advanced techniques will also be presented. In the last weeks of the course all the students will participate in a machine learning competition proposed by the teacher.

STUDY LOAD

<table>
<thead>
<tr>
<th>Type</th>
<th>Hours</th>
<th>Percentage</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hours large group</td>
<td>26,0</td>
<td>17.33</td>
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<tr>
<td>Self study</td>
<td>98,0</td>
<td>65.33</td>
</tr>
<tr>
<td>Hours small group</td>
<td>26,0</td>
<td>17.33</td>
</tr>
</tbody>
</table>

Total learning time: 150 h

CONTENTS

1. Introduction

Description:
1.1 Introduction to pattern classification
1.2 Feature extraction
1.3 Classes and models

Full-or-part-time: 6h
Theory classes: 2h
Laboratory classes: 2h
Guided activities: 1h
Self study: 1h

2. Decision theory

Description:
2.1 Minimization of the Bayesian risk
2.2 Gaussian model
2.3 Linear and quadratic discriminants
2.4 Maximum likelihood estimation

Full-or-part-time: 10h
Theory classes: 4h
Laboratory classes: 4h
Guided activities: 1h
Self study: 1h
3. Feature selection

Description:
3.1 Principal components analysis
3.2 Multiple discriminants analysis

Full-or-part-time: 10h
Theory classes: 4h
Laboratory classes: 4h
Guided activities: 1h
Self study: 1h

4. Non-parametric techniques for supervised learning

Description:
4.1 Parzen windows and k-nearest neighbors
4.2 Support vector machines
4.3 Neural networks
4.4 Decision trees

Full-or-part-time: 24h
Theory classes: 11h
Laboratory classes: 11h
Guided activities: 1h
Self study: 1h

5. Evaluation, combination and selection of classifiers

Description:
5.1 Lack of superiority of any classifier
5.2 Complexity
5.3 Bias and variance
5.4 Resampling for classifier design
5.5 Combining classifiers
5.6 Comparing classifiers
5.7 Launching a ML project

Full-or-part-time: 6h
Theory classes: 2h
Laboratory classes: 2h
Guided activities: 1h
Self study: 1h

6. Unsupervised learning

Description:
6.1 Parametric methods: EM i k-means
6.2 Non-parametric methods: clustering

Full-or-part-time: 8h
Theory classes: 3h
Laboratory classes: 3h
Guided activities: 1h
Self study: 1h
**ACTIVITIES**

**Laboratories**

**Description:**
- PRAC0: Exploratory data analysis
- PRAC1: MAP for Gaussian data
- PRAC2: Data basis and feature selection
- PRAC3: K-Nearest
- PRAC4: SVM
- PRAC5: Neural networks
- PRAC6: Tree classifiers
- PRAC7: Competition

**Material:**
Matlab code and data bases will be available in Atenea.

**Delivery:**
Every second week, a report of the laboratory will have to be uploaded to Atenea.

**Full-or-part-time:** 26h
Laboratory classes: 26h

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**GRADING SYSTEM**

Final exam: 45%
Laboratory: 25%
Competition: 15%
Deliverable exercises: 15%

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**EXAMINATION RULES.**

The use of calculators, mobile phones or class notes is not allowed.

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**BIBLIOGRAPHY**

**Basic:**

**Complementary:**

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**RESOURCES**

**Other resources:**
For the development of the laboratories, applications in Matlab will be provided through Atenea.