Course guides  
230331 - APA - Machine Learning

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

Degree: BACHELOR'S DEGREE IN TELECOMMUNICATIONS TECHNOLOGIES AND SERVICES ENGINEERING (Syllabus 2015). (Optional subject).

Academic year: 2020  
ECTS Credits: 2.0  
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 desision theory, maximum likelihood estimation, non parametric classification techniques and non supervised learning, with application 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 possessing, and communications. In each of the theoretical parts, the methods and algorithms will be developed so that they can be understood and programed 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>14,0</td>
<td>28.00</td>
</tr>
<tr>
<td>Hours small group</td>
<td>6,0</td>
<td>12.00</td>
</tr>
<tr>
<td>Self study</td>
<td>30,0</td>
<td>60.00</td>
</tr>
</tbody>
</table>

**Total learning time:** 50 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 neighbours
- 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

<table>
<thead>
<tr>
<th>Laboratories</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Description:</strong></td>
</tr>
<tr>
<td>PRAC0: Exploratory data analysis</td>
</tr>
<tr>
<td>PRAC1: MAP for Gaussian data</td>
</tr>
<tr>
<td>PRAC2: Data basis and feature selection</td>
</tr>
<tr>
<td>PRAC3: K-Nearest</td>
</tr>
<tr>
<td>PRAC4: SVM</td>
</tr>
<tr>
<td>PRAC5: Neural networks</td>
</tr>
<tr>
<td>PRAC6: Tree classifiers</td>
</tr>
<tr>
<td>PRAC7: Competition</td>
</tr>
<tr>
<td><strong>Material:</strong></td>
</tr>
</tbody>
</table>
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

GRADING SYSTEM

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

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

BIBLIOGRAPHY

**Basic:**

**Complementary:**

RESOURCES

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