270029 - APA - Machine Learning

Coordinating unit: 270 - FIB - Barcelona School of Informatics
Teaching unit: 723 - CS - Department of Computer Science
Academic year: 2019
Degree: BACHELOR'S DEGREE IN INFORMATICS ENGINEERING (Syllabus 2010). (Teaching unit Optional)
ECTS credits: 6
Teaching languages: Catalan

Teaching staff
Coordinator: - Luis Antonio Belanche Muñoz (belanche@cs.upc.edu)
Others: - Javier Béjar Alonso (bejar@cs.upc.edu)

Prior skills
Elementary notions of probability and statistics.
Elementary linear algebra and real analysis
Good programming skills in a high-level language

Requirements
- Prerequisite PE
- Corequisite PROP

Degree competences to which the subject contributes

Specific:
CCO2.1. To demonstrate knowledge about the fundamentals, paradigms and the own techniques of intelligent systems, and analyse, design and build computer systems, services and applications which use these techniques in any applicable field.
CCO2.2. Capacity to acquire, obtain, formalize and represent human knowledge in a computable way to solve problems through a computer system in any applicable field, in particular in the fields related to computation, perception and operation in intelligent environments.
CCO2.4. To demonstrate knowledge and develop techniques about computational learning; to design and implement applications and system that use them, including these ones dedicated to the automatic extraction of information and knowledge from large data volumes.

General:
G4. EFFECTIVE ORAL AND WRITTEN communication: To communicate with other people knowledge, procedures, results and ideas orally and in a written way. To participate in discussions about topics related to the activity of a technical informatics engineer.
### Teaching methodology

The topics exposed in the main classes are very well motivated and motivating (why is it important to know), and supplemented with many examples. The theory lessons introduce all the knowledge, techniques, concepts and results necessary to achieve a well-grounded body of knowledge. These concepts are put into practice in the problem sessions, and in the laboratory.

Prior to each problem-solving session, the teacher proposes problems related to the current topic and the students have time to prepare them using their AA hours of personal work. In class, the students will be divided among several small groups depending on their number. The teacher provides guidance and eventually resolves questions, giving feedback and making the students progress in the resolution. The use of collaborative learning strategies is envisaged, in which one or more students take responsibility to lead the process. Students work again in these problems in their AA time and then deliver them. These deliveries are continued in time, have a uniform load and are evaluated. This strategy is also used to evaluate their skills for effective communication.

The theory lessons are weekly (two hours). The two hours of laboratory classes are twice a month. The problem sessions are weekly.

There is a deliverable practical work that solves a real problem (to be chosen by the student) that integrates all acquired knowledge and skills of the course. This practical work is also used to evaluate their skills for effective communication.

### Learning objectives of the subject

1. Formulate the problem of machine learning from data, and know the different machine learning tasks
2. Organize the flow for solving a machine learning problem, analyzing the possible options and choosing the most appropriate to the problem
3. Decide, defend and criticize a solution to a machine learning problem, arguing the strengths and weaknesses of the approach
4. To compare, judge and interpret a set of results after making a hypothesis about a machine learning problem
5. Understand and know how to apply least squares techniques for solving supervised learning problems
6. Understand and know how to apply techniques for single and multilayer neural networks for solving supervised learning problems
7. Understand and know how to apply support vector machines for solving supervised learning problems
8. Understand and formulate different theoretical tools for the analysis, study and description of machine learning systems
9. Understand and know how to apply the basic techniques for solving unsupervised learning problems
10. Understand and know how to apply basic techniques for solving reinforcement learning problems
11. Understand the most important modern machine learning and computational learning techniques

### Study load

<table>
<thead>
<tr>
<th>Total learning time: 150h</th>
<th>Hours large group: 30h</th>
<th>20.00%</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Hours medium group: 15h</td>
<td>10.00%</td>
</tr>
<tr>
<td></td>
<td>Hours small group: 15h</td>
<td>10.00%</td>
</tr>
<tr>
<td></td>
<td>Guided activities: 6h</td>
<td>4.00%</td>
</tr>
<tr>
<td></td>
<td>Self study: 84h</td>
<td>56.00%</td>
</tr>
</tbody>
</table>
### Content

#### Introduction to Machine Learning

<table>
<thead>
<tr>
<th>Degree competences to which the content contributes:</th>
<th>Description:</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>General information and basic concepts. Overview and approach to the problems tackled by machine learning techniques. Supervised (classification and regression), unsupervised (clustering) and semi-supervised (reinforcement and transductive) learning. Examples of modern applications.</td>
</tr>
</tbody>
</table>

#### Unsupervised machine learning

<table>
<thead>
<tr>
<th>Degree competences to which the content contributes:</th>
<th>Description:</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Definition and approaches for unsupervised machine learning. Clustering algorithms: EM algorithm and k-means algorithm.</td>
</tr>
</tbody>
</table>

#### Supervised Machine Learning Theory

<table>
<thead>
<tr>
<th>Degree competences to which the content contributes:</th>
<th>Description:</th>
</tr>
</thead>
</table>

#### Supervised machine learning (I): classification problems

<table>
<thead>
<tr>
<th>Degree competences to which the content contributes:</th>
<th>Description:</th>
</tr>
</thead>
</table>

#### Supervised machine learning (II): regression problems

<table>
<thead>
<tr>
<th>Degree competences to which the content contributes:</th>
<th>Description:</th>
</tr>
</thead>
</table>

#### Supervised Machine Learning (III): ensembles of models
**Degree competences to which the content contributes:**

**Description:**
Bagging, boosting and ECOC. The bias/variance tradeoff revisited.

---

### Reinforcement learning

**Degree competences to which the content contributes:**

**Description:**
## Planning of activities

<table>
<thead>
<tr>
<th>Development of item 1 of the course</th>
<th>Hours: 10h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Theory classes: 2h</td>
</tr>
<tr>
<td></td>
<td>Practical classes: 0h</td>
</tr>
<tr>
<td></td>
<td>Laboratory classes: 2h</td>
</tr>
<tr>
<td></td>
<td>Guided activities: 0h</td>
</tr>
<tr>
<td></td>
<td>Self study: 6h</td>
</tr>
</tbody>
</table>

### Description:
The student sees an overview and basic concepts of machine learning as well as modern examples of application.

### Specific objectives:
1, 11

<table>
<thead>
<tr>
<th>Development of item 2 of the course</th>
<th>Hours: 8h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Theory classes: 2h</td>
</tr>
<tr>
<td></td>
<td>Practical classes: 0h</td>
</tr>
<tr>
<td></td>
<td>Laboratory classes: 2h</td>
</tr>
<tr>
<td></td>
<td>Guided activities: 0h</td>
</tr>
<tr>
<td></td>
<td>Self study: 4h</td>
</tr>
</tbody>
</table>

### Description:
The teacher explains the theory of unsupervised machine learning, focusing on clustering algorithms.

### Specific objectives:
1, 2, 9

<table>
<thead>
<tr>
<th>Resolution of the problems of item 2 of the course</th>
<th>Hours: 10h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Guided activities: 2h</td>
</tr>
<tr>
<td></td>
<td>Self study: 8h</td>
</tr>
</tbody>
</table>

### Description:
The teacher poses up to 3 problems within the current topic and the students prepare them at home. In the class, the teacher solves difficulties, gives guidance towards full resolution of the problem and eventually answers general questions about the topic. The students work again on these problems and deliver them.

### Specific objectives:
1, 3, 8

<table>
<thead>
<tr>
<th>Development of item 3 of the course</th>
<th>Hours: 12h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Theory classes: 6h</td>
</tr>
<tr>
<td></td>
<td>Practical classes: 0h</td>
</tr>
<tr>
<td></td>
<td>Laboratory classes: 2h</td>
</tr>
<tr>
<td></td>
<td>Guided activities: 0h</td>
</tr>
<tr>
<td></td>
<td>Self study: 4h</td>
</tr>
</tbody>
</table>
### Description:
The teacher exposes the problem of supervised Machine Learning. Explains the differences between regression and classification problems, and the notions of bias-variance tradeoff, over and underfitting as well as other theoretical tools for models selection.

### Specific objectives:
1, 8

### Resolution of the problems of item 3 of the course

#### Description:
The teacher poses up to 3 problems within the current topic and the students prepare them at home. In the class, the teacher solves difficulties, gives guidance towards full resolution of the problem and eventually answers general questions about the topic. The students work again on these problems and deliver them.

#### Specific objectives:
3, 4, 5, 6

<table>
<thead>
<tr>
<th>Hours</th>
<th>Guided activities: 3h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Self study: 10h</td>
</tr>
</tbody>
</table>

### Development of item 4 of the course

#### Description:
The teacher explains the basics of algorithms for separating hyperplanes: the Perceptron algorithm and maximum margin separation. The kernel functions and support vector machines for classification are introduced. Finally two neural networks for classification are covered: the multilayer perceptron and the radial basis function network.

#### Specific objectives:
1, 6, 7, 11

<table>
<thead>
<tr>
<th>Hours</th>
<th>Theory classes: 6h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Practical classes: 0h</td>
</tr>
<tr>
<td></td>
<td>Laboratory classes: 2h</td>
</tr>
<tr>
<td></td>
<td>Guided activities: 0h</td>
</tr>
<tr>
<td></td>
<td>Self study: 4h</td>
</tr>
</tbody>
</table>

### Resolution of the problems of item 4 of the course

#### Description:
The teacher poses up to 3 problems within the current topic and the students prepare them at home. In the class, the teacher solves difficulties, gives guidance towards full resolution of the problem and eventually answers general questions about the topic. The students work again on these problems and deliver them.

<table>
<thead>
<tr>
<th>Hours</th>
<th>Guided activities: 3h</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Self study: 10h</td>
</tr>
</tbody>
</table>
### Specific objectives:

2, 3, 4, 6

### Development of item 5 of the course

**Description:**
The teacher introduces the methods for regression problems: basically least squares (analytical and iterative methods). Error functions for the regression case are introduced. Multilayer neural networks are adapted to the regression case. The support vector machine for regression is introduced.

**Specific objectives:**
1, 5, 6, 7

**Hours:** 13h
- Theory classes: 6h
- Practical classes: 0h
- Laboratory classes: 2h
- Guided activities: 1h
- Self study: 4h

### Resolution of the problems of item 5 of the course

**Description:**
The teacher poses up to 3 problems within the current topic and the students prepare them at home. In the class, the teacher solves difficulties, gives guidance towards full resolution of the problem and eventually answers general questions about the topic. The students work again on these problems and deliver them.

**Specific objectives:**
3, 4, 7

**Hours:** 13h
- Guided activities: 3h
- Self study: 10h

### Development of item 6 of the course

**Description:**
The teacher introduces the basic techniques for ensemble learning: bagging, boosting and ECOC, and explains them in light of the bias / variance tradeoff.

**Specific objectives:**
2, 8, 11

**Hours:** 11h
- Theory classes: 4h
- Practical classes: 0h
- Laboratory classes: 2h
- Guided activities: 1h
- Self study: 4h
Resolution of the problems of item 6 of the course

**Description:**
The teacher poses up to 3 problems within the current topic and the students prepare them at home. In the class, the teacher solves difficulties, gives guidance towards full resolution of the problem and eventually answers general questions about the topic. The students work again on these problems and deliver them.

**Specific objectives:**
2, 3, 4, 9, 10

**Hours:** 10h
- Guided activities: 2h
- Self study: 8h

Development of item 7 of the subject

**Description:**
The teacher explains the basics of reinforcement learning and its applications and briefly introduces transductive learning.

**Specific objectives:**
10, 11

**Hours:** 11h
- Theory classes: 4h
- Practical classes: 0h
- Laboratory classes: 2h
- Guided activities: 1h
- Self study: 4h

Resolution of the problems of item 7 of the subject

**Description:**
The teacher poses up to 3 problems within the current topic and the students prepare them at home. In the class, the teacher solves difficulties, gives guidance towards full resolution of the problem and eventually answers general questions about the topic. The students work again on these problems and deliver them.

**Specific objectives:**
3, 4, 11

**Hours:** 10h
- Guided activities: 2h
- Self study: 8h

Final exam

**Specific objectives:**
5, 6, 7, 8, 9, 10

**Hours:** 3h
- Guided activities: 3h
- Self study: 0h
Delivery of the practical work

<table>
<thead>
<tr>
<th>Specific objectives:</th>
<th>Hours: 0h</th>
</tr>
</thead>
<tbody>
<tr>
<td>2, 3, 4, 5, 6, 7, 8, 9, 10</td>
<td>Guided activities: 0h</td>
</tr>
<tr>
<td></td>
<td>Self study: 0h</td>
</tr>
</tbody>
</table>

Qualification system

The course can be passed with continuous assessment, as follows:

NProbs = Average mark of problems completed during the course  
NPract = Mark for the practical work  
NPart = Mark used to evaluate skills for effective communication

NF1 = 50% NProbs + 40% NPract + 10% NPart

The course can also be passed with a final exam, as follows:

NExF = Mark obtained in a final exam (during the exams period)

NF2 = 40% NExF + 20% NProbs + 30% NPract + 10% NPart

In any case, the final mark is the highest of the two:

FINAL MARK = max (NF1, NF2)
270029 - APA - Machine Learning

Bibliography

Basic:


Complementary:


Others resources:

Hyperlink

http://videolectures.net/Top/Computer_Science/Machine_Learning/

http://cran.r-project.org/

http://www.academicearth.org/courses/machine-learning