

Academic Year: (2019 / 2020)

Review date: 09-05-2020

Department assigned to the subject: Department of Computer Science and Engineering

Coordinating teacher: ALER MUR, RICARDO

Type: Electives ECTS Credits : 3.0

Year : 1 Semester : 2

STUDENTS ARE EXPECTED TO HAVE COMPLETED

Although it is not necessary, it is useful to have taken subjects related to machine learning (data mining, neural networks, evolutionary computation, ...).

COMPETENCES AND SKILLS THAT WILL BE ACQUIRED AND LEARNING RESULTS.

- To know the different types, techniques, and domains, of Automatic Programming (AP)
- To know how AP techniques have been applied to some real domains
- To be able to do research work in PA
- To know the scientific method and being able to perform rigorous evaluation of results
- To be able to use AP and advanced machine learning tools
- To be able to understand and critically assess PA research articles
- To know the different types of publications in PA (surveys, empirical comparisons, original contributions, ...)
- To be able to present scientific contributions
- To be able to know if PA techniques can be applied to specific problems

DESCRIPTION OF CONTENTS: PROGRAMME

1. Introduction
2. Deep Learning:
 - 2.1. Deep Neural Networks
 - 2.2. Recurrent Neural Networks
 - 2.3. Advanced topics in Deep learning
3. Evolutionary-computation
 - 3.1. Genetic Programming
 - 3.2. Estimation of distribution algorithms (Probabilistic Incremental Program Evolution/PIPE)
4. Inductive Logic Programming/ILP
 - 4.1. Basic ILP techniques
 - 4.2. Relational Data Mining
5. Advanced reinforcement-learning
 - 5.1. Dynamic programming
 - 5.2. model-free and model-based methods
 - 5.3. Generalization in reinforcement-learning
 - 5.4. Knowledge-transfer in reinforcement-learning

LEARNING ACTIVITIES AND METHODOLOGY

- Lectures
- Tutorials
- Student's individual work

ASSESSMENT SYSTEM

a maximum of 10 marks will be awarded, by completing the following:

- 1 mark: Attendance
- 2 marks: Completion of one or two small assignments with tools related to lectures.
- 6.5 marks: Theoretical and/or practical assignment. Grading will depend on the complexity of the work. Theoretical works based on literature review will obtain smaller grades than empirical works involving computational experiments.
- 0.5 marks: oral presentation of work (not compulsory)

The extraordinary evaluation will follow the same criteria as the ordinary one.

% end-of-term-examination:	0
% of continuous assessment (assignments, laboratory, practicals...):	100

BASIC BIBLIOGRAPHY

- Dzeroski S. and Lavrac Relational Data Mining., Springer-Verlag, 2001
- Ian Goodfellow, Yoshua Bengio and Aaron Courville Deep Learning, MIT Press (<http://www.deeplearningbook.org>), 2016
- Koza, John R Genetic programming : on the programming of computers by natural selection., MIT PRESS.
- Lavrac, N. and Dzeroski, S Inductive Logic Programming: Techniques and Applications, Ellis Horwood, New York, 1994
- Mitchell, T.M. Machine Learning, McGraw Hill, 1997
- R. Sutton y A. Barto. Reinforcement Learning: an Introduction, The MIT Press, 1998

ADDITIONAL BIBLIOGRAPHY

- Koza, John R. Genetic Programming II: Automatic Discovery of Reusable Programs, MIT PRESS.
- O-Reilly, Una-May et al. (eds.) Evolutionary Computation. Trends in Evolutionary Methods for Program Induction, MIT PRESS.
- Olsson, J. R. Inductive functional programming using incremental program transformation. Artificial Intelligence. Vol. 74:1, 55-83, Elsevier, 1995

BASIC ELECTRONIC RESOURCES

- Ian Goodfellow and Yoshua Bengio and Aaron Courville . Deep Learning: <http://www.deeplearningbook.org/>
- Ricardo Aler . Automatic Inductive Programming tutorial (ICML 2006): <http://www.evannai.inf.uc3m.es/et/icml06/aiptutorial.htm>