Elements of artificial intelligence - course description

| General information | |
|---------------------|--|
| Course name | Elements of artificial intelligence |
| Course ID | 11.4-WE-INFP-EoAl-Er |
| Faculty | Faculty of Computer Science, Electrical Engineering and Automatics |
| Field of study | Computer Science |
| Education profile | academic |
| Level of studies | First-cycle Erasmus programme |
| Beginning semester | winter term 2019/2020 |

| Course information | |
|---------------------|--------------------------------------|
| Semester | 4 |
| ECTS credits to win | 6 |
| Course type | obligatory |
| Teaching language | english |
| Author of syllabus | • dr hab. inż. Marek Kowal, prof. UZ |

| Classes forms | | | | | |
|----------------|--------------------------------|---------------------------|----------------------------------|---------------------------|----------------------|
| The class form | Hours per semester (full-time) | Hours per week (full-time |) Hours per semester (part-time) | Hours per week (part-time |) Form of assignment |
| Lecture | 30 | 2 | • | - | Exam |
| Laboratory | 30 | 2 | • | - | Credit with grade |

Aim of the course

- Familiarize students with the concept of artificial neural networks and their learning algorithms,
- Familiarize students with the concept of fuzzy sets and fuzzy inference mechanism,
- Familiarize students with different graph search strategies.
- Teach students to solve practical engineering problems using artificial intelligence methods.

Prerequisites

Principles of programming, Algorithms and data structures

Scope

Artificial neural networks. Biological neuron. Mathematical model of a neuron. Simple

perceptron. Perceptron learning rule. Perceptron limitations. Models of neurons and their properties. Adaline and Madaline architectures. Multilayer neural networks. Learning of single-layer neural network. Learning of multi-layer neural networks. Error back propagation algorithm. Models of dynamic neurons. Dynamic neural networks. Sample applications of artificial neural networks.

Fuzzy sets and neuro-fuzzy systems. Fuzzy sets and fuzzy logic. Operations on fuzzy sets. Fuzzy inference. Fuzzy rules. Neuro-fuzzy structures and learning algorithms. Sample applications of fuzzy systems.

Graph search strategies. The breadth first search algorithm. The depth first search algorithm. The A* search algorithm. Heuristic functions. Memory and time compelxity. The minimax algorithm. The alpha-beta pruning algorithm. Constrained search.

Teaching methods

Lecture, teaching laboratory classes.

Learning outcomes and methods of theirs verification

| Outcome description | Outcome symbols Methods of verification | The class form |
|---|---|----------------|
| Student is aware of the computational complexity of learned Al methods. | an exam - oral, descriptive, test and other | • Lecture |
| Student can name artificial neuron types and characterize their properties. | • an exam - oral, descriptive, test and other | • Lecture |
| Student can name and characterize fuzzy and neuro-fuzzy systems | s. • an exam - oral, descriptive, test and other | • Lecture |
| Student is able to implement and use fuzzy and neuro-fuzzy systems to solve engineering problems. | a test with score scale an evaluation test an observation and evaluation of activities during the classes | • Laboratory |

| Outcome description | Outcome symbols Methods of verification | The class form |
|---|---|--------------------------------|
| Student can creatively use learned methods of AI in order to solve | activity during the classes | Laboratory |
| new problems. | carrying out laboratory reports | |
| Student can name and define uninformed and heuristic graph search algorithms. | an exam - oral, descriptive, test and oth | er • Lecture |
| Student is able to design and implement a program for heuristic | a test with score scale | Laboratory |
| search. | an evaluation test | |
| | an observation and evaluation of activity | ties |
| | during the classes | |
| Student is able to implement and use artificial neural networks to | a test with score scale | Laboratory |
| solve engineering problems. | an evaluation test | |
| | an observation and evaluation of activity | ties |
| | during the classes | |

Assignment conditions

Lecture - the passing criterion is a sufficient mark from the final test.

Laboratory - the passing criterion are positive marks for laboratory exercises and tests.

Final mark components = lecture: 50% + teaching laboratory: 50%

Recommended reading

- 1. Russell S., Norvig P.: Artificial Intelligence: A Modern Approach, Prentice Hall, 2009.
- 2. Bishop C.M., Hinton G.: Neural Networks for Pattern Recognition, Clarendon Press, Oxford, 1995.
- 3. Edelkamp S., Schroedl S.: Heuristic Search: Theory and Applications, Morgan Kaufmann, 2012.
- 4. Zimmermann H-J.: Fuzzy Set Theory and Its Applications, Springer, 2006.

Further reading

- 1. Bishop C.: Pattern Recognition and Machine Learning, Springer Verlag, 2006.
- 2. Goodfellow I., Bengio Y., Courville A.: Deep Learning, MIT Press, 2016.
- 3. Ross. T.: Fuzzy Logic with Engineering Applications, Wiley, 2004.

Notes

Modified by prof. dr hab. inż. Andrzej Obuchowicz (last modification: 27-10-2019 09:37)

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