

Study programme(s): Applied Mathematics – Data Science				
Level: Master studies				
Course title: Graphical Models and Probabilistic Inference				
Lecturer: Dušan Jakovetić				
Status: elective				
ECTS: 5				
Requirements: Basics of theory of probability				
Learning objectives				
<ul style="list-style-type: none"> - Understanding of theory and practical implementations of graphical models and Belief-Propagation (BP) algorithms for probabilistic inference - Understanding advantages/disadvantages of various graphical models for a given real-world application - Ability to apply graphical models and BP algorithms in MATLAB in real-world problems 				
Learning outcomes				
<ul style="list-style-type: none"> - Ability and experience in modelling, graphical representation, design and analysis of BP algorithms in real-world probabilistic inference problems - Ability to apply the concepts of probabilistic inference on research problems from a wide variety of application areas 				
Syllabus				
<i>Theoretical instruction</i>				
Graphical models for probabilistic systems modeling: directed graphical models - Bayesian Networks; undirected graphical models - Markov Random Fields; Factor Graphs.				
Exact Inference: Efficient marginalization via message-passing Belief-Propagation algorithms; Sum-product algorithm; Max-product (Min-Sum) algorithm.				
Approximate Inference: Loopy Belief-Propagation, Monte Carlo Methods.				
Learning in Graphical Models: ML estimation, Expectation-Maximization algorithm				
<i>Practical instruction</i>				
Application examples in communication systems, image processing, statistical physics, electrical grid (smart grid), computational biology etc.; Implementation methods in MATLAB; Application of selected methods on real-world examples through the course project.				
Literature				
<p>21. D. Koller and N. Friedman: Probabilistic Graphical Models, MIT Press, 2009</p> <p>22. M. J. Wainwright and M. I. Jordan, Graphical models, exponential families, and variational inference, Foundations and Trends in Machine Learning, 2008.</p> <p>23. C. Bishop: Pattern recognition and machine learning, Springer, 2006</p>				
Weekly teaching load			Other: 0	
Lectures: 2	Exercises: 2	Other forms of teaching: 0	Student research: 0	
Teaching methodology				
Lectures; revisions of the material; active students' participation in problem solving; knowledge tests – colloquia; application of the taught material on real-world examples within the course project.				
Grading (maximum number of points 100)				
Pre-exam obligations		points	Final exam	points
Colloquia	Course project	70 = 30 (Colloquia) + 40 (Course project)	written exam	30