

<b>Study program: Artificial intelligence</b>			
<b>Name of the subject: Graphical Models and Probabilistic Inference</b>			
<b>Teacher(s): Miloš Savić, Miloš Radovanović</b>			
<b>Status of the subject: elective</b>			
<b>Number of ECTS credits: 5</b>			
<b>Conditions: none</b>			
<b>Subject goal</b>			
<ul style="list-style-type: none"> <li>- Understanding of theory and practical implementations of graphical models and Belief-Propagation (BP) algorithms for probabilistic inference</li> <li>- Understanding advantages/disadvantages of various graphical models for a given real-world application</li> <li>- Ability to apply graphical models and BP algorithms in MATLAB in real-world problems</li> </ul>			
<b>Outcome of the subject</b>			
<ul style="list-style-type: none"> <li>- Ability and experience in modelling, graphical representation, design and analysis of BP algorithms in real-world probabilistic inference problems</li> <li>- Ability to apply the concepts of probabilistic inference on research problems from a wide variety of application areas</li> </ul>			
<b>Subject content</b>			
<i>Theory</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 learning</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.			
Expert from the industry will be included into the project assignment realization as an external tutor.			
<b>Literature</b>			
<ol style="list-style-type: none"> <li>1. D. Koller and N. Friedman: Probabilistic Graphical Models, MIT Press, 2009</li> <li>2. M. J. Wainwright and M. I. Jordan, Graphical models, exponential families, and variational inference, Foundations and Trends in Machine Learning, 2008.</li> <li>3. C. Bishop: Pattern recognition and machine learning, Springer, 2006</li> </ol>			
<b>Number of active teaching classes</b>	<b>Theoretical teaching: 2</b>	<b>Practical teaching: 2</b>	
<b>Method of carrying out the teaching</b>			
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			
<b>Evaluation of knowledge (maximum number of points 100)</b>			
<b>Pre-exam obligations</b>	points	<b>Final exam</b>	points
Colloquia	30	Written exam	30
Course project	40		