

<b>Study Programme:</b> Computer Science – Master		
<b>Course Unit Title:</b> Advanced Computational Science and Optimization		
<b>Course Unit Code:</b> CS757		
<b>Name of Lecturer(s):</b> Dušan Jakovetić		
<b>Type and Level of Studies:</b> Master Academic Degree		
<b>Course Status (compulsory/elective):</b> Elective		
<b>Semester (winter/summer):</b> Summer		
<b>Language of instruction:</b> Serbian (primary), English (secondary)		
<b>Mode of course unit delivery (face-to-face/distance learning):</b> Face-to-face		
<b>Number of ECTS Allocated:</b> 6		
<b>Prerequisites:</b> Introduction to Computational Science		
<b>Course Aims:</b> <ul style="list-style-type: none"> <li>• Understanding of a wide range of standard and modern numerical methods, with an emphasis on optimization methods</li> <li>• Ability to select an appropriate numerical algorithm for the problem at hand</li> <li>• Ability to implement the taught algorithms in selected programming languages</li> </ul>		
<b>Learning Outcomes:</b> <ul style="list-style-type: none"> <li>• Ability to apply the taught algorithms on real-world problems</li> <li>• Ability to apply the taught algorithms on research problems from various domains of computer science</li> <li>• Ability to customize and analyze efficient numerical algorithms for a given application</li> </ul>		
<b>Syllabus:</b> <i>Theory</i> Iterative methods for solving systems of linear equations: Jacobi, Gauss-Seidel, relaxation methods; First order optimization methods: gradient; projected gradient; line search; proximal gradient; accelerated Nesterov gradient; accelerated gradient for non-smooth optimization (FISTA); Second order optimization methods: Newton; quasi-Newton; Broyden–Fletcher–Goldfarb–Shanno (BFGS); limited memory BFGS; Randomized optimization methods: randomized coordinate gradient; stochastic/online gradient; Parallel and distributed optimization methods: primal decomposition; dual decomposition; augmented Lagrangian; ADMM; distributed gradient. <i>Practice</i> Application examples in various domains of computer science; implementation of the taught methods in selected software languages; application of selected methods on real-world examples.		
<b>Required Reading:</b> <ol style="list-style-type: none"> <li>1. S. Boyd and L. Vandenberghe: Convex Optimization, Cambridge University Press, 2004</li> <li>2. J. Nocedal and S. Wright: Numerical Optimization, Springer, 2011</li> <li>3. D. Bertsekas and J. Tsitsiklis: Parallel and Distributed Computation: Numerical Methods, Prentice-Hall, 1989</li> </ol>		
<b>Weekly Contact Hours:</b> 4	<b>Lectures:</b> 2	<b>Practical work:</b> 2
<b>Teaching Methods:</b> Lectures; revisions of the material; active students' participation in problem solving; knowledge tests – colloquia; application of the taught material on real world examples.		

<b>Knowledge Assessment (maximum of 100 points):</b>			
<b>Pre-exam obligations</b>	points	<b>Final exam</b>	points
Colloquia	40	Oral exam	60
The methods of knowledge assessment may differ; the table presents only some of the options: written exam, oral exam, project presentation, seminars, etc.			