



Subject card

Subject name and code	Optimization methods, PG_00060231								
Field of study	Technical Physics								
Date of commencement of studies	October 2024	Academic year of realisation of subject		2026/2027					
Education level	first-cycle studies		Subject group		Optional subject group Subject group related to scientific research in the field of study				
Mode of study	Full-time studies		Mode of delivery		at the university				
Year of study	3		Language of instruction		Polish				
Semester of study	6		ECTS credits		6.0				
Learning profile	general academic profile		Assessment form		exam				
Conducting unit	Division of Electron Collisions Physics -> Institute of Physics and Applied Computer Science -> Faculty of Applied Physics and Mathematics -> Faculties of Gdańsk University of Technology								
Name and surname of lecturer (lecturers)	Subject supervisor		dr Maciej Kuna						
	Teachers								
Lesson types	Lesson type	Lecture	Tutorial	Laboratory	Project	Seminar	SUM		
	Number of study hours	30.0	0.0	30.0	0.0	0.0	60		
E-learning hours included: 0.0									
Learning activity and number of study hours	Learning activity	Participation in didactic classes included in study plan		Participation in consultation hours		Self-study	SUM		
	Number of study hours	60		5.0		85.0	150		
Subject objectives	The aim of the lecture is to present classical and machine learning inspired optimization methods to allow quick implementation them in the form of appropriate algorithms. There are discussed in the lecture modern optimizing algorithms with an emphasis on their inspiration in physics and biology.								
Learning outcomes	Course outcome		Subject outcome		Method of verification				
	[K6_K01] understands the need to learn and improve professional and personal competencies, inspires and organizes other people's learning process		The student is aware of changes in modern optimization.		[SK5] Assessment of ability to solve problems that arise in practice				
	[K6_W03] has systematized knowledge of higher mathematics, including algebra, analysis, probability theory and numerical methods, allowing for basic description, understanding and modelling of physical phenomena and some technical processes		The student understands how to apply mathematical analysis and algebra tools to evaluate the modeling of phenomena		[SW2] Assessment of knowledge contained in presentation				
[K6_U04] plans and conduct experiments, critically analyzes their results, draw conclusions and forms opinions, has laboratory work experience		The student knows algorithmic methods for verifying and analyzing data.		[SU4] Assessment of ability to use methods and tools					

Subject contents	<p>Course content – lecture</p> <p>Mathematical foundations of optimization. Numerical linear algebra. Vector norms, operations on vectors and matrices. Mathematical analysis. Conditions of existence of extremes of functions of one and many variables and methods of their checking. Methods of finding derivatives.</p> <p>Classic optimization methods. Algorithms for optimizing functions of one variable: dividing the interval into half, golden ratio, Fibonacci, Newton-Raphson and secant method. Algorithms for optimization of multivariable functions: cubic interpolation, Nelder-Mead, conjugate directions, Cauchy, Newton.</p> <p>Optimization issues in machine learning:</p> <p>Linear regression, simple gradient regression, polynomial regression, regularized linear models, logistic regression, linear and non-linear SVM regression, decision trees, team learning, including random forests, dimensionality reduction, deep neural network training.</p> <p>Course content – laboratory</p> <p>PyCharm: IDE for Software Development with Python</p> <p>Linear regression</p> <p>Singular value decomposition (SVD) and pseudo-inverse of a matrix</p> <p>Optimization using KKT conditions</p> <p>Scikit-learn</p> <p>Non-linear SVM (changing kernels and kernel parameters)</p> <p>Bagging classifier</p> <p>VotingClassifier (hard and soft voting)</p> <p>Kernel PCA using radial basis functions</p> <p>Neural Networks: Stochastic Gradient Descent, Momentum, RMSprop, AdaGrad, Adam</p> <p>Tensorflow</p> <p>Keras: functional API</p> <p>Q Learning, Generative Adversarial Networks</p>									
Prerequisites and co-requisites										
Assessment methods and criteria	<table border="1"> <thead> <tr> <th data-bbox="446 743 790 777">Subject passing criteria</th><th data-bbox="790 743 1133 777">Passing threshold</th><th data-bbox="1133 743 1489 777">Percentage of the final grade</th></tr> </thead> <tbody> <tr> <td data-bbox="446 777 790 810">Positive completion of laboratories</td><td data-bbox="790 777 1133 810">50.0%</td><td data-bbox="1133 777 1489 810">50.0%</td></tr> <tr> <td data-bbox="446 810 790 848">Test</td><td data-bbox="790 810 1133 848">50.0%</td><td data-bbox="1133 810 1489 848">50.0%</td></tr> </tbody> </table>	Subject passing criteria	Passing threshold	Percentage of the final grade	Positive completion of laboratories	50.0%	50.0%	Test	50.0%	50.0%
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Example issues/example questions/tasks being completed	<p>Application of linear regression to data optimization.</p> <p>Regularization of linear models.</p> <p>Linear and nonlinear SVM regression.</p> <p>Learning deep neural networks.</p>									
Practical activites within the subject	Not applicable									

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