



Subject card

Subject name and code	Computational basics of artificial intelligence, PG_00068817						
Field of study	Biomedical Engineering, Biomedical Engineering, Biomedical Engineering						
Date of commencement of studies	February 2027	Academic year of realisation of subject				2026/2027	
Education level	second-cycle studies	Subject group				Optional subject group Specialty 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	1	Language of instruction				Polish	
Semester of study	1	ECTS credits				3.0	
Learning profile	general academic profile	Assessment form				assessment	
Conducting unit	Department of Biomedical Engineering -> Faculty of Electronics Telecommunications and Informatics -> Faculties of Gdańsk University of Technology						
Name and surname of lecturer (lecturers)	Subject supervisor	dr inż. Artur Poliński					
	Teachers	dr inż. Artur Poliński					
Lesson types	Lesson type	Lecture	Tutorial	Laboratory	Project	Seminar	SUM
	Number of study hours	15.0	0.0	0.0	15.0	0.0	30
	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	30	5.0	40.0	75		
Subject objectives	The aim of the course is introduction the computational foundations of artificial intelligence						
Learning outcomes	Course outcome	Subject outcome			Method of verification		
	[K7_K02] is ready to provide critical evaluation of received content and to acknowledge the importance of knowledge in solving cognitive and practical problems	has a basic knowledge of computing fundamentals of artificial intelligence			[SK5] Assessment of ability to solve problems that arise in practice		
	[K7_W01] knows and understands, to an increased extent, mathematics to the extent necessary to formulate and solve complex issues related to the field of study	has a basic knowledge of optimization methods			[SW1] Assessment of factual knowledge		
	[K7_U01] can apply mathematical knowledge to formulate and solve complex and non-typical problems related to the field of study by: - appropriate selection of source information and its critical analysis, synthesis, creative interpretation and presentation, - application of appropriate methods and tools	has a basic knowledge of data analysis			[SU1] Assessment of task fulfilment		
	[K7_W03] knows and understands, to an increased extent, the construction and operating principles of components and systems related to the field of study, including theories, methods and complex relationships between them and selected specific issues - appropriate for the curriculum	has a basic knowledge of computing fundamentals of artificial intelligence			[SW1] Assessment of factual knowledge		

Subject contents	<p>Course content – lecture</p> <p>1. Elements of linear algebra and analytical geometry (norms, bilinear mappings, length and distance of vectors, angle between vectors, basis of linear space, orthogonal projection, rotations)</p> <p>2 Matrix decomposition, vectors and eigenvalues, SVD decomposition</p> <p>3 Elements of mathematical analysis (differentiation, Jakobi matrix, Hesse matrix, introduction to gradient methods, Newton's method for equations and systems of nonlinear equations)</p> <p>4 Selected elements of the probability theory (random variable, moments, distributions, Bayes' theorem)</p> <p>5 Optimization methods in artificial intelligence (optimization, optimization with constraints, linear programming)</p> <p>6 Modeling (cost functions, parameter estimation)</p> <p>7 Data analysis using linear regression</p> <p>8 Methods for reducing the dimension of data - principal component analysis</p> <p>9 Methods of heuristic solution search (including simulated annealing)</p>											
Prerequisites and co-requisites	Knowledge of mathematics at the level of engineering studies											
Assessment methods and criteria	<table border="1" data-bbox="448 647 1477 752"> <thead> <tr> <th data-bbox="448 647 794 680">Subject passing criteria</th> <th data-bbox="794 647 1141 680">Passing threshold</th> <th data-bbox="1141 647 1477 680">Percentage of the final grade</th> </tr> </thead> <tbody> <tr> <td data-bbox="448 680 794 714">project</td> <td data-bbox="794 680 1141 714">50.0%</td> <td data-bbox="1141 680 1477 714">50.0%</td> </tr> <tr> <td data-bbox="448 714 794 752">lecture</td> <td data-bbox="794 714 1141 752">50.0%</td> <td data-bbox="1141 714 1477 752">50.0%</td> </tr> </tbody> </table>			Subject passing criteria	Passing threshold	Percentage of the final grade	project	50.0%	50.0%	lecture	50.0%	50.0%
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project	50.0%	50.0%										
lecture	50.0%	50.0%										
Recommended reading	Basic literature	<p>Deisenroth, M. P., Faisal, A. A., &amp; Ong, C. S. (2020). <i>Mathematics for machine learning</i>. Cambridge University Press.</p> <p>Arora, S. A. N. J. E. E. V. (2018, January). Mathematics of machine learning: An introduction. In <i>Proceedings of the International Congress of Mathematicians (ICM 2018)</i> (pp. 377-390).</p> <p>Burges, C. J. (2003, February). Some notes on applied mathematics for machine learning. In <i>Summer School on Machine Learning</i> (pp. 21-40). Springer, Berlin, Heidelberg.</p> <p>Billingsley, P. (2008). <i>Probability and measure</i>. John Wiley &amp; Sons.</p> <p>Von Zur Gathen, J., &amp; Gerhard, J. (2013). <i>Modern computer algebra</i>. Cambridge university press.</p> <p>Rao, S. S. (2019). <i>Engineering optimization: theory and practice</i>. John Wiley &amp; Sons.</p>										
	Supplementary literature	<p>Peterson, J. C., &amp; Smith, R. D. (2015). <i>Mathematics for Machine Technology</i>. Cengage Learning.</p> <p>Bender, E. A. (1996). <i>Mathematical methods in artificial intelligence</i>.</p> <p>Gnedenko, B. V. (2018). <i>Theory of probability</i>. Routledge.</p> <p>Rédei, L. (2014). <i>Algebra</i>. Elsevier.</p> <p>Sra, S., Nowozin, S., &amp; Wright, S. J. (Eds.). (2012). <i>Optimization for machine learning</i>. Mit Press.</p>										
	eResources addresses											

Example issues/ example questions/ tasks being completed	
Practical activities within the subject	Not applicable

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