



## Subject card

Subject name and code	, PG_00066699						
Field of study	Technical Physics						
Date of commencement of studies	February 2026		Academic year of realisation of subject		2025/2026		
Education level	second-cycle studies		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		2.0		
Learning profile	general academic profile		Assessment form		assessment		
Conducting unit	Division of Theoretical Physics and Quantum Informaton -> 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 inż. Patryk Jasik				
	Teachers		dr inż. Patryk Jasik				
Lesson types	Lesson type	Lecture	Tutorial	Laboratory	Project	Seminar	SUM
	Number of study hours	0.0	0.0	30.0	0.0	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		2.0		18.0	50
Subject objectives	The aim of the course is to teach students the practical use of Python in the field of machine learning.						
Learning outcomes	Course outcome		Subject outcome		Method of verification		
	[K7_K01] knows limitations of own knowledge, understands the need to learn and improve professional and personal competencies		The student understands that programming technologies develop quickly, so there is a need to continuously monitor the functionalities of new versions of Python and its packages.		[SK2] Assessment of progress of work		
	[K7_U02] has enhanced knowledge of programming languages and can use software packages		The student has advanced programming skills in Python and its libraries for solving tasks using machine learning methods.		[SU1] Assessment of task fulfilment		

Subject contents	<p>Course content – laboratory Python language</p> <p>1. Basic types of data and operations on them. Conditions. Different types of loops. Exceptions. Lists, tuples, dictionaries.</p> <p>2. Functions. Function with an optional argument. Modules. Creating own module and using it. Writing/reading data to/from files. YAML files. Classes and objects. Introduction to multithreading.</p> <p>3. Analysis of the selected dataset. Reading observations for selected variables. Checking basic statistics for individual variables. Creating histograms. Identification of variables with potentially incorrect data (observations) or missing data. Data repair. Calculation of normalized correlations between individual variables. Conducting linear regression for selected variables, including graphs.</p> <p>5. The scikit-learn package and the linear regression model. The coefficient of determination <math>R^2</math>, MSE, MAE. Division of the data set into a training and test part. Prediction using the created model.</p> <p>6. The scikit-learn and preprocessing package. Polynomial model. Feature engineering. Reduction of variables in the model - the Schwarz criterion (Bayesian Information Criterion - BIC). Operation of the polynomial model in practice.</p> <p>7. The scikit-learn package, the k-nearest neighbors method, decision trees and random forests. Classification problem. The choice of features - predictors and the target variable. Parameters of the model. Model quality assessment - confusion matrix, sensitivity, specificity, precision, accuracy, ROC curve, LIFT curve. Cross-validation: k-fold, n-fold and Monte-Carlo (bootstrap).</p> <p>8. The scikit-learn package and the k-means algorithm as a case of unsupervised learning. Cluster analysis - clustering. Parameters of the model. The Fowlkes-Mallows index, i.e. the consistency between the two divisions of the dataset into clusters. Analysis of the principal components - PCA.</p> <p>9. XGBoost, LightGBM packages as examples of reinforcement learning.</p> <p>10. Introduction to neural networks using TensorFlow and PyTorch.</p>		
Prerequisites and co-requisites	Basic programming skills in the selected language. Basic knowledge of probability calculus and statistics.		
Assessment methods and criteria	Subject passing criteria	Passing threshold	Percentage of the final grade
	The report of the analysis and modeling of the selected data set	60.0%	100.0%
Recommended reading	<p>Basic literature</p> <p>1. Andreas C. Mažller, Sarah Guido, "Introduction to Machine Learning with Python. A Guide for Data Scientists", O'Reilly Media</p> <p>2. Alberto Boschetti, Luca Massaron, "Python Data Science Essentials - Second Edition", Packt Publishing</p> <p>3. Sebastian Raschka, "Python Machine Learning", Packt Publishing</p>		

	Supplementary literature	<p>1. James R. Parker, "Python. An Introduction to Python Programming", Mercury Learning and Information</p> <p>2. Krishna Kumar Mohbey, Brijesh Bakariya, "An Introduction to Python Programming: A Practical Approach", BPB Publications</p> <p>3. Wes McKinney, "Python for Data Analysis. 3rd Edition", O'Reilly Media</p>
	eResources addresses	
Example issues/ example questions/ tasks being completed	<p>Guidelines for creating the report:</p> <ol style="list-style-type: none"> <li>1. Report Title</li> <li>2. Introduction motivation, objectives</li> <li>3. Data Description dataset structure, variables, origin</li> <li>4. Description of the Data Preparation Process step-by-step approach</li> <li>5. Data Analysis assumptions, brief overview of methods and chosen analysis methodology</li> <li>6. Data Modeling assumptions, brief overview of methods and chosen modeling methodology</li> <li>7. Results, Conclusions, and Discussion</li> </ol> <p>The report, along with all code, should be placed in a chosen repository (e.g., GitLab, GitHub).</p>	
Practical activities within the subject	Not applicable	

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