

。 GDAŃSK UNIVERSITY OF TECHNOLOGY

Subject card

Subject name and code	Time series analysis, PG_00063033								
Field of study	Technologies for Industry 5.0								
Date of commencement of studies	October 2024		Academic year of realisation of subject			2027/2028			
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	4		Language of instruction			Polish			
Semester of study	7		ECTS credits			3.0			
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						mputer		
Name and surname	Subject supervisor		dr inż. Patryk						
of lecturer (lecturers)	Teachers							_	
Lesson types and methods	Lesson type	Lecture	Tutorial	Laboratory	Projec	t	Seminar	SUM	
of instruction	Number of study hours	0.0	0.0	30.0	15.0		0.0	45	
	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 S		SUM	
	Number of study hours	45		5.0		25.0		75	
Subject objectives	Gaining knowledge and skills related to time series analysis and modeling.								
Learning outcomes	Course outcome		Subject outcome			Method of verification			
	[K6_W06] demonstrates knowledge related to data analysis and engineering, machine learning, knows the principles of integrating data with management systems to analyze complex engineering and technological problems		The student performs analysis, exploration, and cleaning of time series data sets, is able to utilize statistical models and machine learning models, and can integrate various tools for analytics, data management, and storage.			[SW3] Assessment of knowledge contained in written work and projects			
	[K6_U06] performs analysis, exploration and cleaning of data sets, can use statistical models and machine learning models, integrate various analytical, management and data storage tools		The student demonstrates knowledge in time series analysis and modeling using machine learning methods to analyze complex engineering and technological problems.			[SU1] Assessment of task fulfilment			
	[K6_U01] applies knowledge of mathematics, physics, chemistry, IT tools and other engineering disciplines to solve theoretical, engineering and technological problems		The student applies knowledge of mathematics, physics, chemistry, computational tools, and other engineering disciplines to solve time-dependent theoretical, engineering, and technological problems.			[SU1] Assessment of task fulfilment			

Subject contents	Introduction to Time Series (2 hours)						
	 Definition of Time Series: Introduction to basic concepts and definitions. Types of time series (continuous and discrete). Applications of time series in various fields. Basics of Time Series Visualization: Tools and techniques for visualization. Drawing line and scatter plots. Autocorrelation plots and correlograms. Interactive visualizations 						
	Time Series Decomposition (4 hours)						
	 Decomposition into Components: Theory and practice of time series decomposition. Time series transformations. Additive and multiplicative decomposition. Analysis of trend, seasonality, and residual components. Practical examples of decomposition in Python. 						
	Anomaly Detection Methods (6 hours)						
	 Anomaly Detection Techniques: Overview of methods for detecting anomalies in time series. Statistical methods: deviation detection, Grubbs' and Dixon's tests. Anomaly detection using ARIMA models and regression models. Application of machine learning algorithms (Isolation Forest, One-Class SVM). 						
	Predictive Modeling of Time Series (12 hours)						
	 Statistical Methods: Traditional modeling methods. ARIMA modeling: theory and application. GARCH models for variance modeling. Machine Learning in Time Series Modeling: Use of linear regression and tree-based methods. Ensemble techniques (bagging, boosting). Neural Networks: Modern approaches to time series modeling. Introduction to neural networks and RNN. Long Short-Term Memory (LSTM) and Gated Recurrent Units (GRU). Practical application of LSTM models in forecasting. Latest models and packages. 						
	 Fourier and Wavelet Analysis (6 hours) 1. Fourier Analysis: Theory and application. Fourier Transform: principles and interpretation. Spectral analysis and its applications in time series. Implementation of Fourier Transform in Python. 2. Wavelet Analysis: Theory and practice. Wavelet Transform: principles and differences from Fourier. Applications of wavelet analysis in time series. Implementation of Wavelet Transform in Python. Team Project (max 3 people) (15 hours) Analysis of Real Data: Case studies from various fields. Creation and Presentation of Predictive Models. Solving Real-World Problems Using Learned Methods 						
Prerequisites	Knowledge of Python. Fundamentals	s of statistics and machine learning n	nethods.				
Assessment methods and criteria	Subject passing criteria Team Project (max 3 students)	Passing threshold 60.0%	Percentage of the final grade 100.0%				
Recommended reading	Basic literature	Jason Brownlee, Introduction to Time Series Forecasting with Python, Machine Learning Mastery					
	Supplementary literature	Jason Brownlee, Deep Learning for Time Series Forecasting, Machine Learning Mastery					
	eResources addresses Adresy na platformie eNauczanie:						
Example issues/ example questions/ tasks being completed	 Guidelines for Creating the Project Report: Report Title Introduction - Motivation, Goals Data Description - Data structure, variable description, origin Description of Data Preparation Process - Step-by-step Data Analysis - Assumptions, brief description of methods and analysis methodology Data Modeling - Assumptions, brief description of methods and modeling methodology Results, Conclusions, and Discussion The report, along with all code, should be placed in a chosen repository (e.g., GitLab, GitHub). 						

Work placement	Not applicable

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