

**Karta (sylabus) modułu/przedmiotu**  
**Kierunek studiów: INFORMATYKA**  
 Studia stacjonarne II stopnia

<b>Przedmiot:</b>	<b>Artificial Intelligence Systems</b>
<b>Rodzaj przedmiotu:</b>	<i>Directional</i>
<b>Kod przedmiotu:</b>	I2S1.05
<b>Rok:</b>	1
<b>Semestr:</b>	I
<b>Forma studiów:</b>	<b>Full-time</b>
<b>Rodzaj zajęć i liczba godzin w semestrze:</b>	60
Wykład	30
Ćwiczenia	0
Laboratorium	30
Projekt	0
<b>Liczba punktów ECTS:</b>	4
<b>Sposób zaliczenia:</b>	<i>Exam</i>
<b>Język wykładowy:</b>	English

<b>Goals</b>	
<b>C1</b>	Familiarizing students with advanced models of artificial intelligence
<b>C2</b>	Acquisition by students of practical knowledge in the field of advanced methods of machine learning and artificial intelligence

<b>Prerequisites for knowledge, skills and other competences</b>	
<b>1</b>	Knowledge of basic machine learning models
<b>2</b>	Basic Python programming skills
<b>3</b>	Knowledge of linear algebra, mathematical analysis and statistics

<b>Learning outcomes</b>	
	In terms of knowledge:
<b>EK 1</b>	Has advanced knowledge of developing computational intelligence methods
<b>EK 2</b>	Has a thorough knowledge of the ownership and application of advanced artificial intelligence models depending on the application
	In terms of skills:
<b>EK 3</b>	Can program applications in Python in the range of a wide range of artificial intelligence models
<b>EK 4</b>	He/she can skillfully use machine learning libraries in order to effectively create programs depending on their purpose
	In terms of social competences:
<b>EK 5</b>	He/she is ready to cooperate with colleagues and teachers during laboratory classes
<b>EK 6</b>	Is ready to take on challenges related to the use of modern programming techniques and to critically evaluate these techniques

<b>Course content</b>	
<b>Form of classes - lectures</b>	
	Program content
<b>W1</b>	Review of classic machine learning models
<b>W2</b>	Fuzzy sets
<b>W3</b>	Fuzzy systems

<b>W4</b>	Biologically inspired evolutionary algorithms
<b>W5</b>	Anomaly detection
<b>W6</b>	Information aggregation and fusion. Models of multi-criteria decision making theory
<b>W7</b>	Grouping algorithms
<b>W8</b>	Neural networks models and deep neural networks
<b>W9</b>	Convolutional neural networks
<b>W10</b>	Autoencoders
<b>W11</b>	Recursive networks
<b>W12</b>	AI models in the analysis of images and recordings
<b>W13</b>	Natural language and speech processing
<b>W14</b>	DeepFake. Generative networks
<b>W15</b>	Overview of the latest intelligent models

#### Form of classes – computer laboratories

	Program content
<b>L1</b>	Implementation of classic machine learning models
<b>L2</b>	Uncertainty modeling
<b>L3</b>	Applications of fuzzy systems
<b>L4</b>	Implementation of selected optimization algorithms
<b>L5</b>	Isolation forest and other PyCaret methods
<b>L6</b>	Implementation of aggregation methods
<b>L7</b>	Implementation of selected grouping algorithms
<b>L8</b>	Implementation of selected models of artificial neural networks in the form of a Python package to facilitate code reuse in subsequent projects
<b>L9</b>	Classification of images with the use of convolutional networks. Analysis of the influence of different network topologies on the quality of classification
<b>L10</b>	The use of autoencoders to detect data anomalies
<b>L11</b>	LSTM recursive networks for time series analysis
<b>L12</b>	Using libraries for image analysis on the example of OpenCV and YOLO
<b>L13</b>	Selected NLP techniques: sentiment analysis, the possibilities of the NLTK library and the Bag of Words method
<b>L14</b>	DeepFake detection methods

#### Teaching methods

<b>1</b>	<i>Lecture with multimedia presentation</i>
<b>2</b>	<i>Laboratory exercises: Implementation of tasks with the use of known algorithms and data structures</i>
<b>3</b>	<i>Laboratory exercises: thematic discussion</i>

#### Assessment methods and criteria

Assessment method symbol	Description of the assessment method	Passing threshold
<b>O1</b>	Lecture exam	51%
<b>O2</b>	Credit from the laboratory	51%

#### Basic literature

<b>1</b>	Chollet F., Deep Learning. Praca z językiem Python i biblioteką Keras, Helion, Gliwice 2019
<b>2</b>	Raschka S., Mirjalili V., Python. Uczenie maszynowe. Wydanie II, Helion, Gliwice 2019

#### Supplementary literature

<b>1</b>	Atienza R., Deep learning z TensorFlow 2 i Keras dla zaawansowanych. Sieci GAN i VAE,
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	deep RL, uczenie nienadzorowane, wykrywanie i segmentacja obiektów i nie tylko. Wydanie II, Helion, Gliwice 2022
2	Bonaccorso G., Algorytmy uczenia maszynowego. Zaawansowane techniki implementacji, Helion, Gliwice 2019
3	Foster D., Deep learning i modelowanie generatywne. Jak nauczyć komputer malowania, pisanie, komponowania i grania, Helion, Gliwice 2022
4	Géron A., Uczenie maszynowe z użyciem Scikit-Learn i TensorFlow, Helion, Gliwice 2018
5	Gift N., AI - podejście pragmatyczne. Wprowadzenie do uczenia maszynowego opartego na chmurze, Promise, Warszawa 2018
6	Goodfellow I., Bengio Y., Courville A., Deep Learning. Współczesne systemy uczące się, PWN, Warszawa 2018
7	Grus J., Data science od podstaw. Analiza danych w Pythonie, Helion, 2018
8	Hearty J., Zaawansowane uczenie maszynowe z językiem Python, Helion, Gliwice 2017
9	Karczmarek P., Selected Problems of Face Recognition and Decision-Making Theory, Wydawnictwo Politechniki Lubelskiej, Lublin 2018
10	Osinga D., Deep Learning. Receptury, Helion, Gliwice 2019
11	Patterson J., Gibson A., Deep Learning. Praktyczne wprowadzenie, Helion, Gliwice 2018
12	Rutkowski L., Metody i techniki sztucznej inteligencji, PWN, Warszawa 2012
13	Tadeusiewicz R., Sieci neuronowe, Akademicka Oficyna Wydawnicza, Warszawa 1993

Student workload	
Form of activity	Average number of hours to complete the activity
<b>Contact hours with the lecturer, including:</b>	60
<i>participation in lectures</i>	30
<i>participation in laboratories</i>	30
<b>Student's own work, including:</b>	40
<i>preparation for the exam</i>	20
<i>preparation for the laboratory</i>	20
<b>Total student working time</b>	100
<b>The total number of ECTS points for the course</b>	4

Matrix of learning outcomes					
Learning outcomes	Learning outcome related to the defined field of study	Goals	Course content	Teaching methods	Assessment methods and criteria
LO 1	I2A_W02 ++ I2A_W07 ++ I2A_W08 +++	CO1	L1-L11	1	A1
LO 2	I2A_W07 ++ I2A_W08 +++	CO2	L1-L15	1	A1
LO 3	I2A_U04 ++ I2A_U12 +++ I2A_U13 ++ I2A_U15 +++	CO1, CO2	Lab1-Lab14	2,3	A2
LO 4	I2A_U04 ++ I2A_U12 +++ I2A_U13 ++ I2A_U15 +++	CO1, CO2	Lab1-Lab14	2,3	A2
LO 5	I2A_K01 +	CO2	Lab1-Lab14	2,3	A2

	I2A_K02 +++ I2A_K03 +++				
<b>LO 6</b>	I2A_W02 ++ I2A_W07 ++ I2A_W08 +++	CO2	<i>Lab8-Lab14</i>	2,3	<b>A2</b>

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