# **COURSE OUTLINE**

# (1) GENERAL INFORMATION

SCHOOL	SCHOOL OF APPLIED MATHEMATICAL AND PHYSICAL SCIENCES				
LEVEL OF STUDIES	POSTGRADUATE				
INTERDEPARTMENTAL PROGRAM	Mathematical Modeling in Modern Technologies and Financial Engineering				
COURSE CODE		SEMESTER		2	
COURSE TITLE	Al Hands On				
INDEPENDENT TEACHING ACTIVITIES  In case credit units are awarded separately for different parts of the course, e.g. Lectures, Laboratory Exercises, etc., specify them. If credit units are awarded as a whole for the course, indicate the weekly teaching hours and the total credits.					Credits (ECTS)
Lectures – Exercises			2		ECTS 8
Laboratory			1		
Assignments			3		
COURSE TYI General background, specializ background, specialization, gene knowledge, skill developme	red ral				
PRE- and CO-REQUISITE	· · · · · · · · · · · · · · · · · · ·				
LANGUAGE OF INSTRUCTION AN ASSESSMEN					
COURSE OFFERED TO ERASMU STUDENT (If yes what is the type of course, e reading cours	S: g.				
COURSE WEBSITE (UR	L):   -				

#### (2) LEARNING OUTCOMES

#### **Learning Outcomes**

Here are described the learning outcomes of the course, including the specific knowledge, skills, and competencies that students will acquire upon successful completion (according to the skills and competencies framework of the European higher education chart).

Upon successful completion of the course, students will be able to:

- Understand AI programming foundations, data structures, SQL, and best practices.
- Explain core AI tasks including supervised, unsupervised, and self-supervised learning.
- Apply feature engineering, data preprocessing, and storage techniques.
- Analyze AI models such as clustering, regression, and classification techniques.
- **Develop** practical AI applications, including retrieval-augmented generation (RAG), agent-based systems, and AI-powered customer ranking.
- **Evaluate** the trade-offs between different AI techniques such as dimensionality reduction, transformers, and neural radiance fields (NeRF).
- Implement AI solutions using Python, Docker, and cloud-based AI tools.

#### **General Competencies**

Considering the general competencies expected from a graduate (as these are specified in the Diploma Supplement) which of these does the course aim to develop?

Search, analysis, and synthesis of data and information, using necessary technologies

Adaptation to new situations

Decision-making

Independent work

Teamwork

Working in an international environment

Working in an interdisciplinary environment

Producing new research ideas

Project planning and management

Respect for diversity and multiculturalism

**Environmental awareness** 

Demonstration of social, professional, and ethical responsibility and sensitivity to gender issues

Exercise of critical and self-critical thinking

Promotion of free, creative, and inductive thinking

... Others

#### **General Competencies:**

- Problem-solving & decision-making in Al-driven applications.
- Independent & team-based AI development.
- Interdisciplinary collaboration in AI and data science projects.
- Critical thinking in evaluating AI model performance and ethical considerations.
- Creative AI system design with ethical considerations.
- Project management in Al-related research and applications.

#### (3) COURSE CONTENT

#### **COURSE CONTENT**

### **Module 1: AI Programming Foundations**

- Data structures and libraries (NumPy, Pandas, Matplotlib)
- SQL fundamentals
- Object-oriented & functional programming for Al
- Best coding practices
- Dockerization techniques for Al applications

#### Module 2: Data Fundamentals for AI

- Data types: Structured, unstructured, and semi-structured
- Data preprocessing & feature engineering
- Hot and cold data storage (SQL, NoSQL, Big Data)

#### **Module 3: AI Core Tasks**

- Unsupervised learning: Clustering, dimensionality reduction
- Self-supervised learning: Reinforcement learning, contrastive learning, masked language modeling
- Supervised learning: Classification, regression, ranking, and generative models (text, image, sound, video)

#### Module 4: Modeling

- Classical ML algorithms: PCA, k-Means, SVM, Decision Trees, etc.
- Learning to rank
- Neural networks: CNNs, RNNs, Transformers, LLMs
- Advanced AI techniques: Graph neural networks (GNNs), NeRF, computer vision models

#### **Module 5: AI Applications**

- Customer ranking in marketing
- Retrieval-augmented generation (RAG)
- Retail market product retrieval
- Virtual Try-On & Virtual Try-Off
- Time series forecasting
- Agentic text-to-SQL and function calling
- Ethical AI and fairness considerations

## (4) TEACHING AND LEARNING METHODS – STUDENT EVALUATION

# TEACHING METHODS AND TOOLS Face-to-face, distance learning, etc. USE OF INFORMATION AND COMMUNICATION TECHNOLOGIES (ICT) Use of ICT in teaching, lab training, communication with students Data analysis and visualization tools

#### **COURSE ORGANIZATION**

The teaching methods and approaches are detailed here

Possible activities include:

Lectures Seminars

Laboratory Exercises

Fieldwork

Study & literature analysis

Tutorials
Practical Training
Clinical Training
Artistic Workshop
Interactive Teaching
Study Project
Report Writing
Artistic Creation
Educational Visits

...

Indicate the hours of study for every learning activity as well as the hours of non guided study according to the principles of ECTS

#### STUDENT EVALUATION

Language of Evaluation

Evaluation Methods: Formative or

Summative Multiple Choice Test

Short Answer Questions Essay Questions

Problem Solving
Written Assignment
Report/Presentation

Oral Exam Public Presentation Laboratory Work

Artistic Performance Other

Explicitly define the evaluation criteria and their availability to students

#### **Activity Semester Workload**

Lectures: 26

Study: 26

Homework: 3

Laboratory: 13

Project Work/Presentation: Yes

**Educational Visits: No** 

Total Course Load (in hrs): 65

Evaluation Language: Greek and English (for Erasmus

students)

Evaluation Type Percentage (%)

1st assignment: 25% 2nd assignment: 25% Final assignment: 50% Evaluation Criteria:

- Problem-solving and AI application development.
- Accuracy and efficiency of code implementation.
- Creativity and innovation in AI model design.
- Ethical considerations in AI applications.

Students are **required to complete and pass all three assignments** in order to successfully pass the course. Failure to pass any of the assignments will result in failing the course.

#### (5) RECOMMENDED BIBLIOGRAPHY

**Charu C. Aggarwal** – Neural Networks and Deep Learning: A Textbook.

Ian Goodfellow, Yoshua Bengio, Aaron Courville – Deep Learning.

**David L. Poole & Alan K. Mackworth** – Artificial Intelligence: Foundations of Computational Agents.

**Magnus Ekman** – Efficient Transformers and Quantization Techniques.

Thomas Kipf & Max Welling – Graph Convolutional Networks and Beyond.

MIT OpenCourseWare – Free Course Materials on Neural Networks & AI Systems.

**OpenAI Community Documentation** – LangGraph & AI Agent Frameworks.

## Journal of Machine Learning Research (JMLR)

- Focus: Covers a wide range of machine learning topics, including deep learning, reinforcement learning, and AI applications.
- Website: <a href="https://www.jmlr.org">https://www.jmlr.org</a>

## Artificial Intelligence Journal (AIJ)

- Focus: Publishes research on AI methodologies, algorithms, and applications, including neural networks, NLP, and ethical AI considerations.
- Website: <a href="https://www.journals.elsevier.com/artificial-intelligence">https://www.journals.elsevier.com/artificial-intelligence</a>

## IEEE Transactions on Neural Networks and Learning Systems (TNNLS)

- Focus: A leading journal publishing cutting-edge research in neural networks, deep learning, and AI methodologies.
- Website: <a href="https://ieeexplore.ieee.org/xpl/RecentIssue.jsp?punumber=5962385">https://ieeexplore.ieee.org/xpl/RecentIssue.jsp?punumber=5962385</a>