

## SYLLABUS

### 1. Data about the program of study

|                                      |   |
|--------------------------------------|---|
| 1.1 Institution                      | The Technical University of Cluj-Napoca     |
| 1.2 Faculty                          | Faculty of Automation and Computer Science  |
| 1.3 Department                       | Computer Science                            |
| 1.4 Field of study                   | Computer Science and Information Technology |
| 1.5 Cycle of study                   | Master                                      |
| 1.6 Program of study / Qualification | Data Science / Master                       |
| 1.7 Form of education                | Full time                                   |

### 2. Data about the subject

|   |  |              |   |   |             |
|---|--|--------------|---|---|-------------|
| 2.1 Subject name  | <b>Machine Learning 2</b>  |              |   | Subject code  | <b>7.10</b> |
| 2.2 Course responsible / lecturer                         | Raluca Didona Brehar - <a href="mailto:raluca.brehar@cs.utcluj.ro">raluca.brehar@cs.utcluj.ro</a><br>Prof.dr.eng. Camelia Lemnaru - <a href="mailto:camelia.lemnaru@cs.utcluj.ro">camelia.lemnaru@cs.utcluj.ro</a> |              |   |   |             |
| 2.3 Teachers in charge of seminars / Laboratory / project | As.drd.eng. Jurcă Mihnea - <a href="mailto:mihnea.jurca@cs.utcluj.ro">mihnea.jurca@cs.utcluj.ro</a><br>As.drd.eng. Vlad Negru - <a href="mailto:vlad.negru@cs.utcluj.ro">vlad.negru@cs.utcluj.ro</a>               |              |   |   |             |
| 2.4 Year of study   | I  | 2.5 Semester | 2 | 2.6 Type of assessment (E - exam, C - colloquium, V – verification) | E           |
| 2.7 Subject category                                      | Formative category: DA – advanced, DS – speciality, DC – complementary   |              |   |   | DS          |
|   | Optionality: DI – imposed, DO – optional (alternative), DF – optional (free choice)  |              |   |   | DO          |

### 3. Estimated total time

|  |    |           |        |    |          |   |            |    |         |    |
|--|----|-----------|--------|----|----------|---|------------|----|---------|----|
| 3.1 Number of hours per week   | 3  | of which: | Course | 2  | Seminars | - | Laboratory | 1  | Project | -  |
| 3.2 Number of hours per semester   | 42 | of which: | Course | 28 | Seminars | - | Laboratory | 14 | Project | -  |
| 3.3 Individual study:  |    |           |        |    |          |   |            |    |         |    |
| (a) Manual, lecture material and notes, bibliography                                 |    |           |        |    |          |   |            |    |         | 20 |
| (b) Supplementary study in the library, online and in the field                      |    |           |        |    |          |   |            |    |         | 15 |
| (c) Preparation for seminars/laboratory works, homework, reports, portfolios, essays |    |           |        |    |          |   |            |    |         | 13 |
| (d) Tutoring   |    |           |        |    |          |   |            |    |         | 5  |
| (e) Exams and tests  |    |           |        |    |          |   |            |    |         | 5  |
| (f) Other activities:  |    |           |        |    |          |   |            |    |         | -  |
| 3.4 Total hours of individual study (suma (3.3(a)...3.3(f)))                         |    |           |        |    | 58       |   |            |    |         |    |
| 3.5 Total hours per semester (3.2+3.4)   |    |           |        |    | 100      |   |            |    |         |    |
| 3.6 Number of credit points  |    |           |        |    | 4        |   |            |    |         |    |

### 4. Pre-requisites (where appropriate)

|                |   |
|----------------|---|
| 4.1 Curriculum | Advanced Machine Learning Methods: deep learning,     |
| 4.2 Competence | Linear algebra, programming, logics, basic statistics |

### 5. Requirements (where appropriate)

|                           |   |
|---------------------------|---|
| 5.1. For the course       | white/black-board, projector, PC/laptop |
| 5.2. For the applications | white/black-board, projector, PC/laptop |

## 6. Specific competence

|                              |   |
|------------------------------|---|
| 6.1 Professional competences | <ol style="list-style-type: none"><li>1. Working with advanced mathematical methods and models, engineering and computing specific techniques and technologies.</li><li>2. Development of advanced techniques, methods and methodologies in the domains of software design, programming systems and environments and their applications.</li><li>3. Innovative design of machine learning systems and related software and hardware using the specific tools.</li><li>4. Contextual integration and exploitation of dedicated information systems.</li><li>5. Creative pooling of multidisciplinary knowledge in the field of computers and information technology for research, design, optimization, implementation and testing of theories, algorithms and original methods specific to artificial intelligence and computer vision systems.</li></ol> |
| 6.2 Cross competences        | <ol style="list-style-type: none"><li>1. Proof of knowledge for the economic, ethical, legal and social context associated with the profession, for correct task identification, schedule of activities, responsible decisions, with the final goal the design, preparation and presentation of a scientific paper.</li><li>2. Clear and concise description of professional activity flows, tasks and outcomes obtained by assuming the role of leader / project manager or as a member of a research team, as result of personal skills of domain specific information synthesis, global vision, communication skills with collaborators, ability of task stages identification.</li><li>3. Exercising the skill of continuous self-education and demonstrating critical, innovative and research abilities</li></ol>                                   |

## 7. Expected Learning Outcomes

|           |  |
|-----------|--|
| Knowledge | <p>deep learning fundamental concepts<br/>deep learning architectures<br/>deep learning frameworks<br/>data analytics tools<br/>data models<br/>data storage alternatives<br/>data warehouse principles<br/>database management systems (DBMS)<br/>digital data processing methods<br/>algorithms for dealing with unstructured data<br/>statistics<br/>computer programming<br/>software design principles<br/>software libraries</p> |
|-----------|--|

|                               |  |
|-------------------------------|--|
| Skills                        | <p>The student is able to:</p> <ul style="list-style-type: none"> <li>• create data sets</li> <li>• develop deep learning pipelines</li> <li>• implement deep learning models from scratch</li> <li>• evaluate the performance of deep learning pipelines</li> <li>• deploy deep learning models</li> <li>• establish data processes</li> <li>• manage data</li> <li>• perform dimensionality reduction</li> <li>• interpret technical requirements</li> <li>• use software design patterns</li> <li>• use software libraries</li> <li>• adapt to changes in technological development plans</li> <li>• design user interfaces</li> <li>• implement front-end website designs</li> <li>• use markup languages</li> </ul> |
| Responsibilities and autonomy | <p>The student has the ability to work independently in order to:</p> <ul style="list-style-type: none"> <li>• develop an analytical approach</li> <li>• take a proactive approach</li> <li>• develop strategies to solve problems</li> <li>• be open minded</li> <li>• coordinate engineering teams</li> </ul>  |

#### 8. Discipline objective (as results from the *key competences gained*)

|                         |   |
|-------------------------|---|
| 8.1 General objective   | Understanding basic concepts in deep learning, the main types of architectures, training considerations; understanding the basic elements of reinforcement learning                             |
| 8.2 Specific objectives | Understanding and being able to develop and utilise the most important deep learning and reinforcement learning algorithms. Operate with known frameworks and software tools for deep learning. |

#### 9. Contents

| 9.1 Lectures  | Hours | Teaching methods           | Notes |
|---|-------|----------------------------|-------|
| Introduction to Deep Feedforward Networks                                   | 2     | Presentations, discussions |       |
| Design and optimization of Neural Networks                                  | 2     |                            |       |
| Convolutional Neural Networks   | 2     |                            |       |
| Recurrent Neural Networks   | 2     |                            |       |
| Transformer Models  | 2     |                            |       |
| Generative Models – introduction, autoencoders and variational autoencoders | 2     |                            |       |
| Generative Models - GANs and diffusion models                               | 2     |                            |       |
| Foundation Models   | 2     |                            |       |
| Interpretability of Deep Learning Models                                    | 2     |                            |       |
| Deployment of Deep Learning Models (Quantization, distillation)             | 2     |                            |       |
| Fairness, Bias and Ethics in Machine Learning                               | 2     |                            |       |
| Advanced ML Applications in computer vision                                 | 2     |                            |       |
| Advanced ML Applications in NLP   | 2     |                            |       |
| Review  | 2     |                            |       |

**Bibliography:**

- Deep Learning Book, by Ian Goodfellow, Yoshua Bengio and Aaron Courville, <https://www.deeplearningbook.org/>
- Deep Learning. Foundations and Concepts, Christopher Bishop and Hugo Bishop, Springer Berlin, 2023
- Selected research papers

| 9.2 Applications - Seminars/Laboratory/Project                     | Hours | Teaching methods                        | Notes |
|--|-------|---|-------|
| Introduction to Python and review of usefull Python packages       | 2     | Presentations, discussions, live coding |       |
| FFN for image classification: design, implementation, optimization | 2     |   |       |
| CNN for image classification: design, implementation, optimization | 2     |   |       |
| Implementing an RNN for sequence tagging                           | 2     |   |       |
| Autoencoders   | 2     |   |       |
| Foundation models  | 2     |   |       |
| Interpretability techniques, deployment aspects                    | 2     |   |       |

**Bibliography**

- Selected kaggle.com scripts (<https://www.kaggle.com/>)

*\*Se vor preciza, după caz: tematica seminariilor, lucrările de laborator, tematica și etapele proiectului.*

### 9. Bridging course contents with the expectations of the representatives of the community, professional associations and employers in the field

The contents of this course are in line with the curricula of top universities around the globe (see bibliography sections). Moreover, the contents of the course cover the most important conceptual and technical aspects needed to develop machine learning solutions at industry level).

### 10. Evaluation

| Activity type  | Assessment criteria  | Assessment methods                                 | Weight in the final grade |
|--|--|--|---------------------------|
| Course   | The ability to solve problems specific to the domain. Course participation and involvement | Final Evaluation                                   | 50%                       |
| Seminar  | -  | -  | -                         |
| Laboratory   | The ability to implement and evaluate specific solutions for the proposed problems         | Bi-weekly laboratory assessments, final assignment | 50%                       |
| Project  | -  | -  | -                         |
| Minimum standard of performance: Minimum lab grade 5, Minimum final grade: 5 |  |  |                           |

| Date of filling in:<br>01.09.2025 | Responsible  | Title First name Last name            | Signature |
|-----------------------------------|--------------|---------------------------------------|-----------|
|                                   | Course       | Raluca Didona Brehar, Camelia Lemnaru |           |
|                                   | Applications | Mihnea Jurcă, Vlad Negru              |           |
|                                   |              |                                       |           |

Date of approval in the department  
17.09.2025

Head of department,  
Prof.dr.eng. Rodica Potolea

Date of approval in the Faculty Council  
19.09.2025

Dean,  
Prof.dr.eng. Vlad Mureșan