



## COURSE DESCRIPTION

### 1. Program identification information

1.1 Higher education institution	<b>National University of Science and Technology Politehnica Bucharest</b>
1.2 Faculty	<b>Electronics, Telecommunications and Information Technology</b>
1.3 Department	<b>Applied Electronics and Information Engineering</b>
1.4 Domain of studies	Electronic Engineering, Telecommunications and Information Technology
1.5 Cycle of studies	Masters
1.6 Programme of studies	Electric Vehicle Propulsion and Control

### 2. Date despre disciplină

2.1 Course name (ro)				2.6. Evaluation type	E	2.7 Course regime	Ob
(en)				Machine Learning for Autonomous Systems			
2.2 Course Lecturer				Prof. Dr. Ing. Anamaria RĂDOI			
2.3 Instructor for practical activities				Prof. Dr. Ing. Anamaria RĂDOI			
2.4 Year of studies	1	2.5 Semester	II	2.6. Evaluation type	E	2.7 Course regime	Ob
2.8 Course type	S	2.9 Course code	5	2.10 Tipul de notare			
							Nota

### 3. Total estimated time (hours per semester for academic activities)

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3.1 Number of hours per week	3	Out of which: 3.2 course	2.00	3.3 seminary/laboratory	1
3.4 Total hours in the curricula	42.00	Out of which: 3.5 course	28	3.6 seminary/laboratory	14
Distribution of time:					hours
Study according to the manual, course support, bibliography and hand notes					50
Supplemental documentation (library, electronic access resources, in the field, etc)					
Preparation for practical activities, homework, essays, portfolios, etc.					
Tutoring					0
Examinations					8
Other activities (if any):					0
3.7 Total hours of individual study	58.00				
3.8 Total hours per semester	100				
3.9 Number of ECTS credit points	4				

### 4. Prerequisites (if applicable) (where applicable)



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4.1 Curriculum	Completion of the following subjects: <ul style="list-style-type: none"><li>• Algebra</li><li>• Probability theory and statistics</li><li>• Information transmission theory</li><li>• Decision and estimation in information processing</li><li>• Statistical Signal Processing and Estimation Theory</li></ul>
4.2 Results of learning	Obtain: <ul style="list-style-type: none"><li>• Notions of vectorial algebra, operations with arrays, gradients</li><li>• Notions of probabilities and statistics</li><li>• Notions of decision theory and classification</li><li>• Neural network architectures</li><li>• Notions of signal processing</li></ul>

**5. Necessary conditions for the optimal development of teaching activities** (where applicable)

5.1 Course	The course will take place in a room equipped with video-projector and computer.
5.2 Seminary/ Laboratory/Project	The application sessions will take place in a room with adequate equipment (blackboard) and must include a video-projector.

**6. General objective** *(Referring to the teachers' intentions for students and to what the students will be thought during the course. It offers an idea on the position of course in the scientific domain, as well as the role it has for the study programme. The course topics, the justification of including the course in the curricula of the study programme, etc. will be described in a general manner)*

This subject is studied within the field of Electronic Engineering, Telecommunications and Information Technologies / Master's program Electrical Vehicle Propulsion and Control (EPIC) and aims to familiarize students with the main approaches in the field of Artificial Intelligence, e.g. machine learning techniques, Deep Learning.

As a specific topic, the discipline addresses basic notions (probability & statistics, working with arrays and the calculation of gradients), as well as advanced ones (classification, semantic annotation, regression), concepts and principles specific to the domain. The concepts in the course syllabus have practical applications in various fields, such as shape recognition, signal / image classification, automotive, robotics (human-machine interfaces). The discipline thus contributes to the formation of an overview of the field, providing graduates with the necessary skills, as well as a scientific and technical training adequate to the current requirements at an international level.

**7. Competences** *(Proven capacity to use knowledge, aptitudes and personal, social and/or methodological abilities in work or study situations and for personal and professional growth. They reflect the employers requirements.)*

<b>Specific Competences</b>	Understanding and usage of fundamental concepts for developing neural network architectures Combine knowledge Applies the gained knowledge in practice Applies methods and standard domain-specific instruments for identifying problems and optimal solutions Correctly analyses the application context of basic domain knowledge by using key concepts in terms of subject and specific methodology Oral and written communication in English: demonstrate understanding of the vocabulary related to the field, in a foreign language.
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<b>Transversal (General) Competences</b>	<p>The ability to make decisions in order to solve current problems that arise in Deep Learning</p> <p>The ability to constantly inform and document for personal and professional development by reading specialized literature</p> <p>The ability to analyze and synthesize information</p> <p>Autonomy and critical thinking</p> <p>Flexibility in using new systems and technologies within a team where members together achieve a well-defined goal while taking on different roles or tasks</p> <p>Compliance with the principles of academic ethics</p> <p>The ability to work in stressful situations and optimal time management.</p>
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**8. Learning outcomes** (*Synthetic descriptions for what a student will be capable of doing or showing at the completion of a course. The learning outcomes reflect the student's accomplishments and to a lesser extent the teachers' intentions. The learning outcomes inform the students of what is expected from them with respect to performance and to obtain the desired grades and ECTS points. They are defined in concise terms, using verbs similar to the examples below and indicate what will be required for evaluation. The learning outcomes will be formulated so that the correlation with the competences defined in section 7 is highlighted.*)

<b>Knowledge</b>	<p><i>The result of knowledge acquisition through learning. The knowledge represents the totality of facts, principles, theories and practices for a given work or study field. They can be theoretical and/or factual.</i></p> <ul style="list-style-type: none"><li>• Defines domain-specific notions such as linear layer, fully connected layer, convolutional layer, neural network, Deep Learning, classification, regression, performance matrix.</li><li>• Describe/classify decision, classification and regression systems.</li><li>• Highlights consequences of using certain parameter optimization techniques.</li><li>• Knows decision criteria and parameter optimization techniques.</li></ul>
<b>Skills</b>	<p><i>The capacity to apply the knowledge and use the know-how for completing tasks and solving problems. The skills are described as being cognitive (requiring the use of logical, intuitive and creative thinking) or practical (implying manual dexterity and the use of methods, materials, tools and instrumentation).</i></p> <ul style="list-style-type: none"><li>• Select relevant information for designing classification and regression systems, based on neural networks.</li><li>• Work in a team through discussions related to solving some requirements during the laboratory.</li><li>• Solve applications during the laboratory.</li><li>• Analyze and compare various classification and regression methods.</li><li>• Formulate conclusions over the experiments.</li></ul>
<b>Responsability and autonomy</b>	<p><i>The student's capacity to autonomously and responsibly apply their knowledge and skills.</i></p> <ul style="list-style-type: none"><li>• Select appropriate bibliographic sources.</li><li>• Respect the principles of academic ethics.</li><li>• Demonstrate adaptability to learning in new contexts.</li><li>• Collaborate with other colleagues and teaching staff during the performance of teaching activities.</li><li>• Demonstrate autonomy in problem solving.</li><li>• Demonstrate social responsibility through active involvement in events in the academic community.</li><li>• Realize the value of his contribution in the field of engineering to the identification of solutions to real problems of a social and economic nature, demonstrating social responsibility.</li><li>• Apply principles of professional ethics/deontology in identifying optimal solutions.</li><li>• Demonstrate effective time management skills.</li></ul>



**9. Teaching techniques** (*Student centric techniques will be considered. The means for students to participate in defining their own study path, the identification of eventual fallbacks and the remedial measures that will be adopted in those cases will be described.*)

The teaching process will explore both expository (lecture, exposition) and conversational-interactive teaching methods, based on discovery learning models facilitated by direct and indirect exploration of reality, but also on action-based methods, such as exercise, activities practice and problem solving. The oral communication methods are exposition, problematization and conversation.

Lectures will be used in the teaching activity, based on presentations in .pdf format and notes written on the blackboard. Each course will start with a short review of the chapters already covered, with an emphasis on the concepts covered in the last course.

The presentations use images / diagrams and connections with current technology so that the information presented is easy to understand, assimilate and apply in various contexts. Active listening and assertive communication techniques will be applied, as well as bi-directional feedback mechanisms.

Teamwork skills will be practiced to solve different learning tasks in seminars and laboratories.

## 10. Contents

COURSE		
Chapter	Content	No. hours
1	Introduction in Machine Learning. Probability theory review. Vector algebra review.	2
2	Decision rules. Bayes criteria.	2
3	Model parameter estimation. MAP and MSE estimation. Estimation for an unknown a-priori distribution. Evaluation of the estimator's quality.	2
4	Clustering. K-Means algorithm. Gaussian Mixture Models. Parameter-free unsupervised learning.	4
5	Support Vector Machine.	4
6	Neural networks for classification. Activation functions. Minimizing a cost function. The gradient method. Optimization. Regularization.	8
7	Convolutional neural networks. Standard architectures.	6
	<b>Total:</b>	28

### Bibliography:

1. Anamaria Radoi, curs Machine Learning for Autonomous Systems, <https://curs.upb.ro/2021/course/view.php?id=9894>
2. Christopher M. Bishop, "Pattern Recognition and Machine Learning (Information Science and Statistics)", Springer-Verlag, Berlin, Heidelberg, 2006

LABORATORY		
Crt. no.	Content	No. hours
1	Introduction in Python.	2
2	Clustering using K-means and Gaussian Mixture Models.	2
3	Algoritmul K-Nearest Neighbors.	2
4	Support Vector Machine.	2
5	Optimization using the gradient method.	2
6	Liniar neural network.	2



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7	Convolutional neural network.	2
	<b>Total:</b>	14

**Bibliography:**

1. Anamaria Radoi, curs Machine Learning for Autonomous Systems, <https://curs.upb.ro/2021/course/view.php?id=9894>
2. Christopher M. Bishop, "Pattern Recognition and Machine Learning (Information Science and Statistics)", Springer-Verlag, Berlin, Heidelberg, 2006

**11. Evaluation**

Activity type	11.1 Evaluation criteria	11.2 Evaluation methods	11.3 Percentage of final grade
11.4 Course	Rigorous knowledge of basic theoretical concepts (clusterization, supervised learning, SVM, neural networks, optimization for minimizing a cost function, the gradient method, quadratic estimation, MAP estimation) and methods to operate with these notions.	Final written exam (during Exam session)	40%
	Test	Written exam	20%
11.5 Seminary/laboratory/project	Problem solving during the laboratory	Grading during semester	40%
11.6 Passing conditions			
• Obtaining 50% of the total score.			

**12. Corroborate the content of the course with the expectations of representatives of employers and representative professional associations in the field of the program, as well as with the current state of knowledge in the scientific field approached and practices in higher education institutions in the European Higher Education Area (EHEA)**

- Through the activities carried out, students develop skills to offer solutions to problems and to propose ideas for improving the existing situation in the Automotive field by expanding some notions and concepts associated with the field of Artificial Intelligence.
- The course has a similar content to the courses held by Stanford University.

Date

Course lecturer

Instructor(s) for practical activities

19.09.2025

Prof. Dr. Ing. Anamaria  
RĂDOI

Prof. Dr. Ing. Anamaria RĂDOI



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Date of department approval

Head of department

Conf. dr. ing. Bogdan Cristian FLOREA

Date of approval in the Faculty  
Council

Dean

Prof. Dr. Ing. Mihnea Udrea