

Incoming student mobility

Name of UNIOS University Unit: SCHOOL OF APPLIED MATHEMATICS AND
INFORMATICS

COURSES OFFERED IN FOREIGN LANGUAGE
FOR ERASMUS+ INDIVIDUAL INCOMING STUDENTS

Department or Chair within the UNIOS Unit	School of Applied Mathematics and Informatics
Study program	<ul style="list-style-type: none"> Undergraduate university study programme in Mathematics and Computer Science Undergraduate university study programme in Mathematics
Study level	Undergraduate (Bachelor)
Course title	Machine Learning
Course code	M096
Language of instruction	English
Brief course description	<p>Syllabus.</p> <p>Supervised Learning:</p> <ol style="list-style-type: none"> 1. Introduction to the theory of machine learning. 2. Techniques of nonparametric learning. K-nearest neighbour method. Decision tree. 3. Bayesian learning. Bayesian naive and optimal classifier. 4. Neural network: representation and learning. 5. Regression and classification. Linear regression. Locally weighted linear regression. Logistical regression and classification. Regularization. 6. Support Vector Machine (SVM): Hyperplane separation. Optimal marginal classification. Dual problem. Kernel method. Sequential minimal optimization. 7. Theory of statistical learning. Vapnik-Chervonenkis dimension. <p>Unsupervised Learning:</p> <ol style="list-style-type: none"> 1. Introduction and motivation. Definitions. Different examples of applications <p>Representative of the finite set from \mathbb{R} in least squares (LS) sense and in least absolute deviations (LAD) sense. Representative of the finite set from \mathbb{R}^2. Distance-like function in \mathbb{R}^2. Centroid, median and geometric median in plane. Representative of the</p>

	<p>finite set from \mathbb{R}^n: centroid, median, geometric median. Applications of Mahalanobis distance-like function. Representative of the data on unit circle.</p> <ol style="list-style-type: none"> 2. Data clustering methods. K-means algorithm. EM (Expectation Maximization) algorithm. K-medoid method. Agglomerative clustering 3. Dimension reduction. Principal Component Analysis. 4. Appropriate number of clusters in a partition: Indexes. 5. Spectral clustering methods and theory of graphs. 6. Probabilistic and statistical aspects of data clustering
Form of teaching	Consultative teaching.
Form of assessment	<p>Lectures and exercises are illustrated by ready-made software packages. Exercises are partially auditory and partially laboratory, with the use of computers. Lectures, exercises and seminars are obligatory. Final exam consists of a written and oral part, and it is taken after the completion of lectures. Acceptable results achieved in mid-term exams throughout the semester replace the written part of the exam. Students may influence their final grade by doing homework or writing a seminar paper during the semester. Homework expands course contents, and students are expected to be independent and creative. Seminar papers are understood as an extension of homework.</p>
Number of ECTS	7
Class hours per week	3+2+0
Minimum number of students	
Period of realization	Summer semester
Lecturer	Kristian Sabo Domagoj Matijević