

Course Title	Advanced Topics in Machine Learning				
Course Code	AI620				
Course Type	Elective				
Level	Master (2 nd Cycle)				
Year / Semester	1 st Year/2 nd Semester or 2 nd Year/1 st Semester				
Teacher's Name	TBA				
ECTS	7	Lectures / week	Up to 6 Teleconferences	Laboratories / week	None
Course Purpose and Objectives	This course focuses on advanced topics of machine learning, including clustering, reinforcement learning, ensembles methods, cost-sensitive learning and class imbalance.				
Learning Outcomes	<p>Upon successful completion of this course students should be able to:</p> <ul style="list-style-type: none"> • Discover structure in unlabeled data in the form of clusters. • Design and implement sequential decision making agents interacting with their environment and learning the optimal policy of acting. • Deal with issues that arise when working with real world data in practical applications, such as class imbalance and the need to optimize the particular (business) cost/benefit in a task, instead of the typical cost function or evaluation measure. 				
Prerequisites	AI600		Co-requisites	None	

Course Content	<p><u>1-4) Unsupervised Learning:</u> k-means, hierarchical clustering, density-based clustering, Principal Components Analysis (Kernel PCA, SPARSE PCA), Self-organizing Maps, Independent Components Analysis</p> <p><u>5-6) Ensemble methods:</u> bagging, (gradient) boosting, output flipping and smearing, random subspaces and patches, rotation forest, stacking</p> <p><u>7) Cost-sensitive learning:</u> cost-based evaluation, minimizing the expected cost, probability calibration, sampling methods, example weighting</p> <p><u>Class imbalance:</u> types of class imbalance, in(appropriate) evaluation measures, sampling methods, generating synthetic examples</p> <p><u>8) Active learning:</u> active learning scenaria, uncertainty sampling, query-by-committee, error-based methods, density-weighted methods, similarities/differences with semi-supervised learning</p> <p><u>9-13) Reinforcement learning:</u> multi-armed bandits, Markov decision processes, dynamic programming, Monte Carlo methods, temporal difference learning, on-policy prediction with approximation.</p>
Teaching Methodology	E-Learning
Bibliography	<p>Hastie T., Tibshirani R., Friedman J. (Latest Edition). The elements of statistical learning, Latest Edition, Springer</p> <p>Introduction to Data Mining (Latest Edition). Pang-Ning Tan, Michael Steinbach, Anuj Karpatne, Vipin Kumar, Pearson Education, Limited</p> <p>Sutton, R. S., & Barto, A. G. (Latest Edition). Reinforcement learning: an introduction, Latest Edition, MIT Press</p> <p>Settles, B. (Latest Edition). Active learning. Morgan & Claypool.</p> <p>Zhou, Z.-H. (Latest Edition). Ensemble methods: Foundations and algorithms. Chapman and Hall/CRC.</p> <p>He, H., Garcia, E. (Latest Edition) Learning from Imbalanced Data, IEEE Transactions on Knowledge and Data Engineering, 21(9), pp. 1263-1284</p> <p>Branco, P., Torgo, L., & Ribeiro, R. P. (Latest Edition). A Survey of Predictive Modeling on Imbalanced Domains. ACM Computing Surveys.</p> <p>Elkan, C. (Latest Edition). The foundations of cost-sensitive learning. In IJCAI International Joint Conference on Artificial Intelligence (pp. 973–978)</p>

Assessment	Final Examination Assignments/On-going evaluation	50%	
Language	English		
		50%	
		100%	