

ROBOTICS AND INTELLIGENT SYSTEMS MASTER PROGRAM  
COURSE SYLLABUS

Course Details				
<b>Code</b>	<b>Academic Year</b>			<b>Semester</b>
RIS511	1			1
<b>Title</b>	<b>T</b>	<b>A</b>	<b>L</b>	<b>ECTS</b>
Machine Learning	2	2	0	7
<b>Language</b>	English			
<b>Level</b>	<b>Undergraduate</b>		<b>Graduate</b>	<b>X</b>
<b>Department / Program</b>	Robotics and Intelligent Systems			
<b>Forms of Teaching and Learning</b>	Face-to-Face, Group Study, Individual Study, programming.			
<b>Course Type</b>	<b>Compulsory</b>		<b>Elective</b>	<b>X</b>
<b>Objectives</b>	The course will give the student the basic ideas and intuition behind modern machine learning methods as well as a formal understanding of how, why, and when they work; and gain the ability to use this knowledge in the development of various learning models.			
<b>Content</b>	<ul style="list-style-type: none"> <li>- Perceptron, Convergence, Generalization</li> <li>- Linear Regression, Bias and Variance</li> <li>- Logistic Regression</li> <li>- Over- and Underfitting, Regularisation</li> <li>- Maximum Margin Classification, Support Vector Machines (SVM)</li> <li>- Non-linear Predictions, Kernels</li> <li>- Neural Networks, Multilayer Perceptron, Backpropagation, Intro to Deep Learning</li> <li>- Unsupervised Learning, K-Means Algorithm</li> <li>- Principal Component Analysis (PCA)</li> <li>- Model selection, model selection criteria</li> </ul>			
<b>Prerequisites</b>				
<b>Coordinator</b>	DI Dr. Canan Yıldız			
<b>Lecturer(s)</b>	DI Dr. Canan Yıldız			
<b>Assistant(s)</b>				
<b>Work Placement</b>				
Recommended or Required Reading				
<b>Books / Lecture Notes</b>	<ul style="list-style-type: none"> <li>- <a href="#">Maschine Learning</a>, Tom Mitchell, McGraw-Hill, 1997.</li> <li>- <a href="#">Artificial Intelligence: A Modern Approach</a>, S. Russel und P. Norvig, Prentice Hall, Englewood Cliffs, 2003.</li> </ul>			
<b>Other Sources</b>	<ul style="list-style-type: none"> <li>- Hands-on machine learning with Scikit-Learn, Keras and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems, Aurélien Géron, O'Reilly Media, 2019.</li> </ul>			
Additional Course Material				
<b>Documents</b>	-			
<b>Assignments</b>	-			
<b>Exams</b>	-			

**ROBOTICS AND INTELLIGENT SYSTEMS MASTER PROGRAM  
COURSE SYLLABUS**

Course Composition			
Mathematics und Basic Sciences	20		%
Engineering			%
Engineering Design			%
Social Sciences			%
Educational Sciences			%
Natural Sciences			%
Health Sciences			%
Expert Knowledge	80		%
Assessment			
Activity	Count		Percentage (%)
Midterm Exam	1		40
Quiz			
Assignments	1		10
Attendance			
Recitations			
Projects			
Final Exam	1		50
		<b>Total</b>	<b>100</b>
ECTS Points and Work Load			
Activity	Count	Duration	Work Load (Hours)
Lectures	14	2	28
Self-Study	1	94	98
Assignments	9	4	36
Presentation / Seminar Preparation			
Midterm Exam	1	3	3
Recitations	14	2	28
Laboratory			
Projects			
Final Exam	1	3	3
		<b>Total Work Load</b>	<b>196</b>
		<b>ECTS Points (Total Work Load / 28)</b>	<b>7</b>
Learning Outcomes			
1	Understand the complexity of Machine Learning algorithms (regression, classification, clustering, and dimensionality reduction) and their limitations.		
2	Select the appropriate machine learning algorithms for real-life applications.		

## ROBOTICS AND INTELLIGENT SYSTEMS MASTER PROGRAM COURSE SYLLABUS

3	Be capable of confidently applying common Machine Learning algorithms in practice and implementing their own.
4	Be capable of performing experiments in Machine Learning using real-world data.
5	Assess the model quality in terms of relevant performance/error metrics for each application.

### Weekly Content

1	Introduction, supervised and unsupervised learning, model representation, cost function
2	Gradient Descent, Gradient Descent for Linear Regression
3	Multiple Variables, Feature Scaling, Learning Rate, Polynomial Regression
4	Classification, Logistic Regression
5	Decision Boundary, Multiclass Prediction, One-vs-All
6	Neural Networks, Model Representation
7	Cost Function and Backpropagation, Gradient Checking, Random Initialization
8	Evaluating Learning Algorithms, Train/Validation/Test Sets, Bias and Variance, Learning Curves
9	Midterm Exams
10	Large Margin Classification, Optimization Objective, Intuition
11	Kernels, Support Vector Machines
12	Unsupervised Learning, K-Means Algorithm
13	Dimensionality Reduction, Data Compression, Principal Component Analysis
14	Principal Component Analysis (cont.)
15	Summary, Recitation

### Contribution of Learning Outcomes to Program Objectives (1-5)

	P1	P2	P3
1	5	5	4
2	5	5	4
3	5	5	4
4	5	5	4
5	5	5	4

**Contribution Level**      1: Low 2: Low-intermediate 3: Intermediate 4: High 5: Very High

**Compiled by:**      DI Dr. Canan Yıldız

**Date of Compilation:**      26.05.2021s