

Full Syllabus



| Course Title | |
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| Introduction to Machine Learning | |
| Lecturer | |
| Dr. Nadav Cohen | |
| Semester | |
| Fall | |
| Course requirements | |
| See catalog | |
| Final grade components | |
| 20% homework assignments, 80% final exam | |
| Course schedule | |
| Class no. / Date | Subject and Requirements (assignments, reading materials, tasks, etc.) |
| 1 | Introduction to the course and to machine learning |
| 2 | Probably-approximately-correct (PAC) learning model |
| 3 | Generalization bounds, VC dimension, bias-variance tradeoff, model selection |
| 4 | Regularization, optimization |
| 5 | Support vector machines (SVM) |
| 6 | Kernels |
| 7 | Multi-class learning, deep learning |
| 8 | Decision trees |
| 9 | Boosting and ensemble methods |
| 10 | Regression, online learning |
| 11 | Principal component analysis (PCA), clustering, generative models |
| 12 | Gaussian mixture model (GMM) and expectation maximization (EM) |
| 13 | Summary, supplements, |
| Required course reading | |

Optional course reading

 $https://www.cs.huji.ac.il/\circshais/UnderstandingMachineLearning/understanding-machine-learning-theory-algorithms.pdf$

Comments