Introduction to Machine Learning



Course Syllabus

What You Will Learn	Topics
Getting started	Machine Learning, Classifications of ML problems (Supervised, Reinforcement and Unsupervised) with examples
Tools for Machine	Python, Conda, JupyterLab, Python libraries for ML(SciPy, NumPy,
1)av 1	Matplotlib, Pandas), Data visualization, Linear Algebra for ML
Learning	· //
Day 2 Linear regression	Cost function, Model and hypotheses representation, Gradient
	Descent, Feature scaling and Mean normalization
Day 3 Logistic regression	Classification and decision boundary, Sigmoid function, Multiclass
Logistic regression	classification, Bias and Variance, Regularization
Day 4 More Supervised Learning	Linear Discriminant Analysis, K-Nearest Neighbors, Decision Trees
measurement	(Precision, Recall, F1 Score), Confusion matrix, K-fold improvement
Support Vector	Large Margin classifiers, Kernels (Gaussian and Linear), SVM software
Machines	packages
Unsupervised Learning	K-means clustering, Dimensionality reduction and Principal
Algorithms	Component Analysis
Day 8 Anomaly Detection	Density estimation, Gaussian distribution, Multivariate Gaussian
	distribution
Day 9 Recommender systems	Content-based recommendations, Collaborative filtering, Predicting
	content ratings
Hands on project	Applied Machine Learning
W/L-+ +- J+9	Advice for applying Machine Learning, Deep Learning and Neural
what to do next?	Networks, Project ideas and future plans
	Getting started Tools for Machine Learning Linear regression Logistic regression More Supervised Learning Performance measurement Support Vector Machines Unsupervised Learning Algorithms Anomaly Detection Recommender systems