Course: 11751(9): Machine Learning

Instructor: Dr. Walid Gomaa and Dr. Noha Yousri

Lectures: Mon 10am-1:30pm and Thur 10am-1:30pm

 ${\bf Email: wgomaa@alex.edu.eg and noha_yousri@yahoo.com}$

Course description: This course provides a broad overview of the current paradigms and techniques of machine learning, in particular statistical learning and modeling and self-organizing systems. By the end of the term the students should be familiar with the central and modern concepts of machine learning.

- 1. Bayessian Learning:
 - (a) Bayes theorem
 - (b) MAP algorithm and application to concept learning
 - (c) Maximum Likelihood and least-squared error optimization
 - (d) Bayes optimal classifier
 - (e) Naive Bayes classifier
 - (f) Bayesian belief networks
- 2. Artificial Neural Networks:
 - (a) Perception training rule
 - (b) Gradient descent and the backpropagation algorithm
- 3. Linear methods for regression:
 - (a) Residual Sum of Squares RSS objective function
 - (b) Variance reduction methods including subset selection, forward stepwise and backward stepwise
 - (c) Shrinkage methods including ridge regression and Lasso regrission
- 4. Non-linear methods for regression: piecewise polynomials and splines
- 5. Decision tree learning:
 - (a) The ID3 algorithm
 - (b) The C4.5 algorithm
- 6. Case-Based Reasoning
- 7. Classification:
 - (a) Introduction to basic classifiers: Euclidean distance classifiers, k NN classifiers
 - (b) Linear discriminants: reward-punishment, least mean square, Ho-Kashyap

- (c) A brief lecture on SVMs
- (d) Non-parametric density estimation: histogram, k-NN and parzen windows
- (e) Classifier evaluation
- (f) Bagging and Boosting
- 8. Clustering:
 - (a) Clustering problem, and challenges
 - (b) Classical algorithms: k-means, k-medoids, single, average and complete linkage, divisive algorithms
 - (c) Graph-based and density-based algorithms: Zahn's MST algorithm, DBScan, Mitosis
 - (d) Cluster validity: Dunn's index, Xie-Beni, novel validity measures
- 9. Applications: Selected topic: Bioinformatics

References:

- 1. 'Machine Learning', Thomas Mitchell, McGraw Hill Higher Education, 1st edition, 1997.
- 2. 'The Elements of Statistical Learning: Data Mining, Inference, and Prediction', Trevor Hastie, Robert Tibshirani, and Jerome Friedman, Springer, 2nd ed. 2009.

Grading: Presentations, term papers, exams, and class participation.