

CompSci 273A : Introduction to Machine Learning
Fall 2015
Instructor: Eric Mjolsness

Rough outline of topics:

Introduction

- Example: Perceptron Learning

- ML tasks

 - data types: inputs and outputs
 - supervised vs. unsupervised vs ...
 - classification, regression, relationship learning
 - clustering, dimension reduction

- ML techniques

 - data models
 - training
 - testing
 - generalization theory

Supervised learning

- regression

 - linear regression
 - features & bases
 - overfitting
 - regression for classification (logistic regression)
& vice versa (linear output units)

 - Kernal methods

- classification

 - two-class vs. multiclass
 - decision trees
 - neural networks
 - k-nearest-neighbors
 - SVMs

- ensemble methods

Unsupervised learning

- clustering (k-means, EM-MoG)
- dimensionality reduction (PCA, NNMF, manifold learning)

Theory

- VC dimension
- Probabilistic & Markov models

Advanced

- Deep & structured neural networks
- belief propagation
- relaxation neural networks

Assignments and Grading:

There will be a several homework sets worth 40%, a midterm exam worth 30% and a group project worth 30%. The Group Projects will be described in a separate document.

Midterm exam: Tuesday, November 3, in class. Note that this is the first class after the end of Daylight Savings time.

Final Projects due: Roughly, at the regularly scheduled Final Exam time for this class.

References

Required:

C. Bishop: Pattern Recognition and Machine Learning, Springer.

Optional:

P Flach: Machine Learning, Cambridge U. Press. (*Elementary and tutorial but entirely missing important topics including neural networks.*)

Advanced (not stocked in bookstore):

K. Murphy, Machine Learning: A Probabilistic Approach, MIT Press.

T. Hastie, R. Tibshirani, J. Friedman: The Elements of Statistical Learning (Springer).

G. Montavon, G. Orr, K.-R. Muller: Neural Networks: Tricks of the Trade, Springer.

Scholkopf and Smola: Learning with Kernals (*Specialized to kernal methods*).