### Introduction to Machine Learning

Fall Semester

# Recitation 13

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# 13.1 Informal Course Overview

# 13.1.1 Unsupervised

### Clustering

- K-Means
- Gaussian Mixture Model (GMM), learned with EM
- Spectral Clustering

#### **Dimensionality Reduction**

• PCA (and Probabilistic PCA)

# 13.1.2 Supervised

### **Binary Classification**

- k-Nearest Neighbors, kernel k-NN
- Naïve Bayes
- Hyperplane Classifiers:
  - Perceptron (and Margin Perceptron)
  - Winnow
  - SVM:
    - \* Hard Margin and Soft Margin SVM
    - \* Kernel SVM: Polynomial and Gaussian Kernels, others
- Decision Tree (e.g. with Decision Stumps)
- Boosting:

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- AdaBoost (e.g. with Decision Stumps)
- Bagging, Stacking
- Random Forest

### **Multiclass Classification**

- K-Nearest Neighbors
- Naïve Bayes
- Multiclass SVM: One vs. All, One vs. One

### Regression

- K-Nearest Neighbors Regression
- Linear Regression
- Linear Regression with Regularization:
  - Ridge Regression
  - Lasso Regression
- Logistic Regression

# 13.1.3 Tools and Theory

#### **Model Selection**

- Train vs. Test, k-fold Cross Validation
- Penalty

#### **Statistics**

- Maximum Likelihood (ML) Estimators
- Maximum Aposteriori (MAP) Estimators

### Optimization

- Gradient Descent (and Coordinate Gradient Descent)
- Linear Programming, Quadratic Programming
- Expectation-Maximization (EM)

### Others

- PAC Framework
- Singular Value Decomposition (SVD)