XCS229 Resource Handout Map

Below is an outline of all concepts covered in both XCS229i ML and XCS229ii ML Strategy and Intro to RL.

1 Supervised Learning

1.1 Supervised Learning & Discriminative Algorithms

- 1. Linear Regression
- 2. Batch/Stochastic Gradient Descent
- 3. Normal Equation
- 4. Locally Weighted Regression
- 5. Probabilistic Interpretation of Linear Regression
- 6. Logistic Regression (Further reading: BinaryClassification & Logistic Regression)
- 7. Newton's Method
- 8. Perceptron (Further reading: Perceptron)
- 9. Logistic Regression
- 10. Exponential Family
- 11. Generalized Linear Models (GLM)
- 12. Softmax Regression

1.2 Generative Learning Algorithms

- 1. Gaussian Discriminant Analysis
- 2. Generative & Discriminative Analysis
- 3. Naive Bayes
- 4. Laplace Smoothing
- 5. Event Models

1.3 Kernels and Support Vector Machines

- 1. Support Vector Machines (Further reading: Representer Functions)
- 2. Kernels

1.4 Decision Trees

- 1. Decision Trees
- 2. Ensemble Methods (Further reading: Boosting)

2 Unsupervised Learning

- 1. K-means Clustering
- 2. Mixture of Gaussians
- 3. Expectation Maximization (EM) Algorithm
- 4. Factor Analysis
- 5. Independent Component Analysis (Further reading: Principal Component Analysis)

3 Deep Learning (Neural Networks)

- 1. Backpropagation (Further reading: Supplemental notes on backpropagation)
- 2. Activation Functions
- 3. Vanishing and Exploding Gradients
- 4. Initialization Schemes

4 Machine Learning Theory

- 1. Bias/Variance Tradeoff (Further reading: Bias Variance Calculations)
- 2. Error Analysis
- 3. Regularization and Model Selection

5 Reinforcement Learning

- 1. Reinforcement Learning and Control
- 2. LQR, DDP, and LQG