

6.867 Machine Learning

List of topics, problems, concepts, and methods

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Recommended prerequisites

The course assumes some basic knowledge of probability theory and linear algebra; for example, you should be somewhat familiar with

- Joint and marginal probability distributions
- Normal (Gaussian) distribution
- Expectation and variance
- Statistical correlation and statistical independence

Problems, concepts, methods, and tools

Listed here are important topics we intend to cover in the course. The list is not all inclusive.

Problems:

- Regression and classification
- Active learning
- Feature selection
- Density estimation
- Clustering
- Model selection
- Inference

Concepts:

- Estimation, bias, variance, loss

- Empirical risk, maximum likelihood
- Generalization, overfitting
- Regularization
- Capacity, VC-dimension
- Generative/discriminative models
- Representation, model structure
- Minimum description length

Models and methods:

- linear regression, additive models
- Generalized Linear Models
- Neural networks
- Support Vector Machine (SVM)
- Boosting
- Mixture models, mixtures of experts
- Kernel density estimation ,
- Markov chain/processes
- Hidden Markov Models (HMM) , ,
- Belief networks, Markov random fields ,

Tools:

- Cross-validation
- Gradient descent
- Quadratic programming ,
- EM algorithm ,
- Forward-backward algorithm ,
- Junction tree algorithm
- Gibbs sampling