Learning Goals 575



Overall Goals

Intuition while choosing between methods

Confidence while considering use of methods

Cognizance of choices being made in an analytics workflow

Successfully reason about tweaking methods in a principled manner

A Note on Practice

The goal of the assignments is to get you up to speed with R and practice with a few simple datasets

The goal of the lectures is to ask 'why' for every step we take (or choice we make) in an analytics workflow

These two complement each other so the learner can get a complete view of the subject.





Understand the difference between supervised and unsupervised learning



Able to use vector and matrix notations for data, inputs and outputs



Understand what is meant by 'linear' in a linear model



Understand the procedure behind the nearest neighbor classifier



Understand the statistical assumptions being made for k-nearest neighbor and linear models



Understand what the best (but unrealizable) regression and classification models are



Understand that curse of dimensionality is the motivation for looking at other models



Connect the dots between linear model and k-nn to other 'richer' models



Know how to compare models using bias and variance



Understand the assumptions behind statistical properties of linear regression coefficients



How to change the bias and variance of a model family via regularization



Completely understand the mechanics of cross validation

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Why is subset selection, ridge or LASSO needed versus ordinary least squares?



How are they related to a likelihood interpretation?



What is the basic intuition behind Ridge regression? LASSO?



How can you do classification via regression?



What assumptions underlie logistic regression?



What assumptions underlie linear discriminant analysis?



Review of bias variance tradeoff, and their relation to model selection and assessment.



Review of cross validation



The notion of Bootstrap



MLE as a unifying point of view for supervised and unsupervised learning



Review of the first 6 lectures



What is a Gaussian mixture model?



Why do we need to sample? How can we sample using Gibbs sampling?



The motivation for the Expectation Maximization algorithm



What are generalized linear models?

575: Goals for Lecture: Data and Decisions



Overview of three industry problems



Learn the details of a marketing decision problem and how predictions help



What is the newsvendor problem and how does machine learning help?



See the application of logistic regression for ICU mortality prediction (an application in clinical decision support)



Why and how are tree based methods nonlinear? How to optimize their 'parameters'?



What is the idea behind boosting (with adaboost as an example)?



What are some practical issues with tree based models?



Lets demystify gradient boosting methods!



What is bagging? How does Random forest improve over a single tree?



Walkthrough of a nonlinear model from scratch: the Multivariate adaptive regression splines model



How to interpret a learned random forest? (variable importance)



The geometric idea behind support vector machines



SVMs: What are support vectors?



Unsupervised learning from the point of view of a joint distribution



SVMs: Extension to nonlinear settings via Kernel trick



Association rules: what is the core idea? Extending beyond market basket data.



What is dissimilarity?



Principal components: why are they needed? How to compute them?



How does K-means work? We will look at it from an optimization point of view.



Spectral clustering: what is spectral? spectral Understand it as a spectral decomposition followed by k-means.



What is different in time series datasets compared to supervised learning datasets?



What are the core ideas behind the ARMA model construction?



How to model dependencies across time?



Can supervised learning methods be used for this? (hint: yes!)

Make use of the resources, reach out to the teaching staff and have fun learning!