

Final Exam Topics

The in-class final exam will be cumulative, with an emphasis on the topics covered after the midterm.

Post Midterm Topics (in addition to topics covered up to the midterm (clickable))

- Supervised learning - Classical classification
 - Fisher's linear discriminants
 - Linear and quadratic discriminants for normal distributions
 - K-NN method
- Unsupervised learning - Cluster analysis
 - K-means
 - Hierarchical clustering
 - Mixture Models for cluster analysis
 - (likelihood function, latent membership variable approach, MLE setup, EM approach)
- Expectation - Maximization (EM) algorithm basics
 - EM imputation for missing data
 - EM algorithm for Maximum Likelihood Estimation
 - The latent variable approach in EM
- Independent Component Analysis
 - Basic assumptions and results of ICA
 - Non-Gaussian-ness measures: Skewness, Kurtosis, Entropy.
- High dimensional data challenges and modern approaches
 - Regularized regression Methods: Ridge, Lasso, ElasticNet regression
 - (bias-variance trade-off in regularized regression models)
 - Sparse PCA (using regularized regression methods)
 - Support Vector Machine (SVM) for classification (introduction only, not including SVM kernel method)
- (Required for 32950 only)
 - Introduction to tree based methods: regression tree, classification tree, Random Forest.