KEY TOPICS FOR ML

- 1. Basic mathematical concepts in ML Regularization-- and how they would apply in production.
- 2. Basic Statistics.
- 3. Data collection and noise reduction, overfitting and why they occurs etc
- 4. Kernel Methods
- 5. Supervised Learning Algorithms
 - K Nearest Neighbors
 - Decision Trees
 - Naive Bayes and Probability theory
 - Support Vector Machines. Optimization problem to solve to train a support vector machine
 - Regression, Linear, Logistic Regression, Tree-based regression
- **6.** Unsupervised Learning Algorithms
 - K-Means clustering
 - FP tree
 - Apriori Analysis
- 7. PageRank
- 8. Expectation Maximization
- **9. Dimensionality Reduction,** detecting redudant features using filters, PCA, KPCA, ICA uses
- **10. Reinforcement Learning**
- 11. Bayesian Networks
- 12. Ensemble Learning
- 13. Gaussian Processes
- 14. Anomaly Detection
- 15. Tools for parallelizing machine learning algorithms.
- 16. Useful Links

Edwin Chen's blog

ML Theory

Simply Statistics