Course Overview CSci 8980: ML at Large Scale and High Dimensions

Instructor: Arindam Banerjee

January 22, 2014

Instructor: Arindam Banerjee Course Overview

General Information

- Course Number: CSci 8980
- Class: Mon Wed 01:00-02:15 pm
- Location: Akerman Hall 227
- Instructor: Arindam Banerjee
- Office Hours: Keller Hall 6-213 Mon Wed 02:30 03:30 pm
- Web page: http://wwwusers.cselabs.umn.edu/classes/Spring-2014/csci8980-lshd/
- Moodle page:

https://ay13.moodle.umn.edu/course/view.php?id=10099

- Email:
 - banerjee@cs.umn.edu
 - Mention 8980 in the title, so I know

- Please read the syllabus carefully
- Paper Reviews: 45% of total grade
 - Individual activity
 - Review 15 (out of 21) papers from the list
 - Each paper is 3% of total grade
- Paper Presentations: 20% of total grade
 - Present in paper specific groups of (at least) 2
 - Present (at least) 2 papers, each paper is 10%
- Project: 35% of total grade
 - Group activity, groups of 2
 - Proposal, progress report, final report, presentation

Course Activities: Paper Reviews

- Individual activity
- Choose 15 (of 21) papers you want to review
 - Do not have to submit the list to me upfront
- Each paper is 3% of total grade
- 2 page (max) writeups, including references
 - 11 pt font, 1 inch margin, in pdf
 - Latex template will be provided
- Guidelines:
 - Problem considered, related work
 - Main result, associated results
 - Key insights/ideas used or introduced
 - Potential limitations, if any, and future directions
- Due before class, in moodle

- Each student will do (at least) 2 paper presentations
- Each paper will be presented by (at least) 2 students
 - Group will be paper specific
 - Determined by preferences for presentation
- Submit preferences over all 21 papers:
 - Ranking (preference list) over all 21 papers
 - Due: Tue, Jan 28 in moodle
- Assignments finalized by Thu, Jan 30
 - Each paper will be assigned to 2 students
- Plan a 45 min presentation
 - Work together on the presentation

Course Activities: Project

- Groups of 2 students, form groups by Fri Feb 28
- Project
 - Proposal: 1-page, due Fri, Mar 14
 - Progress Report: 2-page, due Fri, Apr 11
 - Final Report: 5-page + references, etc., due Fri, May 09
 - Presentation: Last two weeks of the semester (see schedule)
- Scope of project has to be aligned with the course topics
 - High-dimensional statistics
 - Large scale optimization
 - Graphical model structure learning
 - Structured sparsity, e.g., low 'rank' matrices/tensors, etc.
 - Superposition or 'dirty' models, e.g., low rank + sparse, etc.
 - ...
- Cannot simply apply a ML method to a dataset

- Paper Reviews: 45 $\% = 15 \times 3 \%$
- Paper Presentation: 20 % = 2 \times 10 %
- Project: 35 % = 5 + 5 + 15 + 10 % (Proposal + Progress Report + Final Report + Presentation)
- Grading is absolute: A = 90-100, A- = 85-90, B+ = 80-85, B = 70-80, B- = 65-70, C+ = 60-65, C = 50-60, F = less than 50.

• Sparsity

- Initiated in the 90s (studied even earlier)
- Models with sparsity and structured sparsity
- Focus on theoretical advances

Structure Learning

- Conditional independence for high-dimensional distributions
- Provable recovery using sparse estimation

Optimization

- Large scale non-smooth optimization
- Online and stochastic optimization
- Methods for structured sparsity

- Opt 1 John Duchi, Shai Shalev-Shwartz, Yoram Singer, and Ambuj Tewari, Composite Objective Mirror Descent, Conference on Learning Theory (COLT), 2010.
- Opt 2 Anatoli Juditsky and Arkadi Nemirovski, First-Order Methods for Nonsmooth Convex Large-Scale Optimization, I: General Purpose Methods, Optimization for Machine Learning, 2011.
- Spt 1 Martin J. Wainwright, Sharp Thresholds for High-Dimensional and Noisy Sparsity Recovery Using L1-Constrained Quadratic Programming (Lasso), IEEE Transactions on Information Theory, 55:5, 2183-2202, 2009.
- Spt 2 Peter J. Bickel, Ya'acov Ritov, and Alexandre Tsybakov, Simultaneous analysis of Lasso and Dantzig Selector, The Annals of Statistics, 37:4, 1705-1732, 2009

- Opt 3 John Duchi, Elad Hazan, and Yoram Singer, Adaptive Subgradient Methods for Online Learning and Stochastic Optimization, Journal of Machine Learning Research, 12, 2121-2159, 2011.
- Opt 4 Huahua Wang and Arindam Banerjee, Online Alternating Direction Method, International Conference on Machine Learning (ICML), 2012.
- Spt 3 Sahand N. Negahban, Pradeep Ravikumar, Martin J. Wainwright, and Bin Yu, A Unified Framework for High-Dimensional Analysis of M-Estimators with Decomposable Regularizers, Statistical Science, 27:4, 538–557, 2012.
- Spt 4 Venkat Chandrasekaran, Benjamin Recht, Pablo A. Parrilo, and Alan S. Willsky, The Convex Geometry of Linear Inverse Problems, Foundations of Computational Mathematics, 12, 805-849, 2012.

- Spt 5 Yaniv Plan and Roman Vershynin, Robust 1-bit compressed sensing and sparse logistic regression: A convex programming approach, IEEE Transactions on Information Theory, 59:1, 482-494, 2013.
- Opt 5 John Duchi, Alekh Agarwal, and Martin Wainwright, Dual Averaging for Distributed Optimization: Convergence Analysis and Network Scaling, IEEE Transactions on Automatic Control, 57:3, 592-606, 2012.
- Opt 6 Ohad Shamir and Tong Zhang, Stochastic Gradient Descent for Non-smooth Optimization: Convergence Results and Optimal Averaging Schemes, International Conference on Machine Learning (ICML), 2013.
- Opt 7 Sebastian Bubeck and Nicolo Cesa-Bianchi, Regret Analysis of Stochastic and Nonstochastic Multi-armed Bandit Problems, Foundations and Trends in Machine Learning, 5:1, 1-122, 2012.

- Spt 6 Garvesh Raskutti, Martin J. Wainwright, and Bin Yu, Restricted eigenvalue properties for correlated Gaussian designs, Journal of Machine Learning Research, 11, 2241-2259, 2010.
- Spt 7 Mark Rudelson and Shuheng Zhou, Reconstruction from anisotropic random measurements, IEEE Transactions on Information Theory, 59:6, 3434-3447, 2013.
- Opt 8 Lei Yuan, Jun Liu, and Jieping Ye, Efficient Methods for Overlapping Group Lasso, IEEE Transactions on Pattern Analysis and Machine Intelligence, 35:9, 2104-2116, 2013.
- Opt 9 Andreas Argyriou, Rina Foygel, and Nathan Srebro, Sparse Prediction with the k-Support Norm, Advances in Neural Information Processing Systems (NIPS), 2012.

- Opt 10(a) Ambuj Tewari, Pradeep Ravikumar, and Inderjit Dhillon, Greedy Algorithms for Structurally Constrained High Dimensional Problems, Advances in Neural Information Processing Systems (NIPS), 2011.
- Opt 10(b) Martin Jaggi, Revisiting Frank-Wolfe: Projection-Free Sparse Convex Optimization, International Conference on Machine Learning (ICML), 2013.
 - StL 1 Pradeep Ravikumar, Martin J. Wainwright, Garvesh Raskutti, and Bin Yu, High-dimensional covariance estimation by minimizing l1-penalized log-determinant divergence, Electronic Journal of Statistics, 5, 935-980, 2011.
 - StL 2 Tony Cai, Weidong Liu and Xi Luo, A Constrained L1 Minimization Approach to Sparse Precision Matrix Estimation, Journal of American Statistical Association, 106:494, 594-607, 2011.

- StL 3(a) Han Liu, Fang Han, Ming Yuan, John Lafferty, and Larry Wasserman, High Dimensional Semiparametric Gaussian Copula Graphical Models, The Annals of Statistics, 40:4, 2293-2326, 2012.
- StL 3(b) Lingzhou Xue and Hui Zou, Regularized Rank-Based Estimation of High-Dimensional Nonparanormal Graphical Models, The Annals of Statistics, 40:5, 2541-2571, 2012.
 - StL 4 Pradeep Ravikumar, Martin J. Wainwright and John Lafferty, High-Dimensional Ising Model Selection using L1-regularized Logistic Regression, The Annals of Statistics, 38:3, 1287-1319, 2010.