

Matrix Algebra and Optimization for Statistics and Machine Learning

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- ▶ Yiyuan She, OSB 209F
- ▶ Tuesday and Thursday 12:30pm–1:45pm, OSB 215
- ▶ Office hour: Thursday 2:00pm–3:00pm and by appointment

Course Webpage

- ▶ <http://campus.fsu.edu>
- ▶ Syllabus
- ▶ Announcements
- ▶ Slides

Books

None. Some reference books though (optional)

- ▶ Matrix Algebra From A Statistician's Perspective, Harville, 1997
- ▶ Matrix Algebra, Gentle, 2007
- ▶ Convex Optimization, Boyd and Vandenberghe (<https://stanford.edu/~boyd/cvxbook/>)
- ▶ Numerical Optimization, Nocedal and Wright, 1999

Prerequisite

- ▶ Regression
- ▶ Linear algebra
- ▶ Optimization (basic)

Software

- ▶ R + necessary packages

Grades

- ▶ Project

Check the syllabus online.

Project

- ▶ Deadline: Dec 5 by 5pm
- ▶ Paper: typed, 10-15 pages
- ▶ It can either computational or theoretical
- ▶ (Potential arrangement: oral presentation)

Possible Topics

- ▶ Eigenvalues and eigenvectors, singular value decomposition, Cholesky decomposition, QR decomposition, partitioned matrices, Schur complement, generalized inverses, Hadamard products, Kronecker products, vectorization, projections, subspace intersection and canonical angles
- ▶ Matrix differentiation: Jacobian, the chain rule
- ▶ [Matrix inequalities]

Possible Topics (cont'd)

- ▶ Convex functions and convex optimization
- ▶ Lagrange duality
- ▶ Gradient descent and Newton's method
- ▶ Line search methods
- ▶ Inexact Newton, quasi-Newton, sketched Newton
- ▶ Proximal methods and linearization
- ▶ Bregman divergence and mirror descent
- ▶ MM algorithms
- ▶ Block coordinate descent and randomization
- ▶ Augmented Lagrangian, operator splitting
- ▶ J-L lemma, random sampling and random projection
- ▶ Stochastic approximation and optimization

Possible Topics (cont'd)

- ▶ Co-integration, support vector machines, SDP relaxation, canonical correlation analysis, low-rank matrix approximation, back-propagation in neural networks, multi-task learning, generalized linear models, graph learning, bandit algorithms, network clustering and ranking, approximate message passing, manifold learning, structural parsimony pursuit, boosted trees
- ▶ [Optional: gradient boosting, manifold optimization]

Examples

- ▶ Gaussian graph learning
- ▶ Multivariate meta-analysis
- ▶ Matrix completion
- ▶ Sufficient dimension reduction
- ▶ Neural network
- ▶ K-means for clustering
- ▶ Multi-task learning