140,778 ADVANCED STATISTICAL COMPUTING

(Biostatistics – 2nd term, 3 units)

COURSE SYLLABUS

Instructor: Hongkai Ji

Contact

Office: 615 N. Wolfe St., Rm. E3638 Office hours: Thursday 3:00-4:00pm

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Course page: http://www.biostat.jhsph.edu/~hji/courses/computing/

Class times

Tuesday 8:30 - 9:50am Thursday 8:30 - 9:50am

Location

Wolfe W4013

Description

This course covers the theory and application of common algorithms used in statistical computing. Topics include root finding, optimization, numerical integration, Monte Carlo, Markov chain Monte Carlo, stochastic optimization and bootstrapping. Some specific algorithms discussed include: Newton-Raphson, EM, Metropolis-Hastings algorithm, Gibbs sampling, simulated annealing, Gaussian quadrature, Romberg integration, etc. Applications of these algorithms to real research problems will be discussed.

Course Learning Objectives

Upon successfully completing this course, students will be able to: 1) describe common deterministic statistical algorithms, such as root finding, numerical integration methods, Newton-Raphson, quasi-Newton methods, EM, MM; 2) describe common stochastic algorithms used in statistics, such as Monte Carlo methods, Markov Chain Monte Carlo, stochastic optimization, Gibbs sampling, Metropolis-Hastings method; 3) understand mathematical properties of common statistical algorithms; 4) implement statistical algorithms using a high-level statistical programming language.

Prerequisites

Prior programming experience; at least one year of doctoral-level statistics/biostatistics theory and methods courses; 140.776

Texts & References

Numerical Analysis for Statisticians, Kenneth Lange, *Springer* Monte Carlo Strategies in Scientific Computing, Jun S. Liu, Springer

Grading policy

Attendance: 10% Homework: 60% Final Project: 30%

Schedule

Oct 20 (Thu): Introduction

Oct 25 (Tue): Solution of nonlinear equations: bisection, functional iteration, Newton's method

Oct 27 (Thu): Convergence rate, Optimization (homework 1)

Nov 1 (Tue): Newton-Raphson, Scoring

Nov 3 (Thu): EM algorithm, Applications of EM

Nov 8 (Tue): Convergence rate of EM, Model-based Clustering

Nov 10 (Thu): Mixtures of Factor Analyzers, MM algorithm (homework 1 due; homework 2)

Nov 15 (Tue): Random number generation

Nov 17 (Thu): Markov chains

Nov 22 (Tue): Markov Chain Monte Carlo (MCMC), Metropolis-Hasting

Nov 24 (Thu): Thanksgiving, no class

Nov 29 (Tue): Gibbs sampler, MCMC in DNA motif discovery

Dec 1 (Thu): MCMC in DNA motif discovery (homework 2 due, homework 3)

Dec 6 (Tue): Marginalization, General conditional sampling

Dec 8 (Thu): Importance sampling

Dec 13 (Tue): Variance reduction, Gaussian quadrature Dec 15 (Thu): Romberg integration (homework 3 due)

Final Project Due Date

Dec 30 (Thu)