Disclaimer
This syllabus is to be used as a guideline only. The information provided is a summary of topics to be covered in the class.

Information contained in this document such as assignments, grading scales, due dates, office hours, required books and materials may be from a previous semester and are subject to change. Please refer to your instructor for the most recent version of the syllabus.

EEE 598: STATISTICAL MACHINE LEARNING: FROM THEORY TO ALGORITHMS SPRING 2019 Course Information

Instructor: Gautam Dasarathy

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Office: GWC 324 Tel: 480-865-5035

Class Meetings: MW 3:05pm - 4:20pm @ ECG 152

Office Hours: TBD

Course Description: This course will serve as a primer in statistical learning theory and as a platform for exploring emerging algorithms and theory in large scale data analytics. This study is at the intersection of information processing, statistical theory, and computational sciences. The class will contain a healthy mix of topics from all of these disciplines.

Prerequisites: Background in probability, statistics, linear algebra, and applied mathematics. Some facility in a programming language like Python, R, or Matlab will be assumed.

Instructor Consent: Future enrollment in the course would require instructor consent. Please fill out this form in its entirety: https://goo.gl/forms/Z6AhczqiIBdT2EEt1

Topics: A tentative list of topics that we will cover:

- 1. Introduction [motivation, definitions, terminology, probability, statistics, and linear algebra tools]
- 2. Concentration of measure and Empirical Risk Minimization
- 3. Vapnik-Chervonenkis Theory and binary classification
- 4. Structural risk minimization
- 5. Support Vector Machines
- 6. Ensemble methods [weak and strong learning, boosting, Adaboost]
- 7. Convex Losses and Radamacher Complexity
- 8. Logistic regression and LASSO
- 9. Johnson-Lindenstrauss Lemma, Sparsity, and Compressed Sensing
- 10. Graphical Models: Inference, Structure Learning, and applications
- 11. Clustering [k-means, k-means++, hierarchical, spectral]
- 12. Stochastic Gradient Descent (SGD) and Large Scale Optimization
- 13. Multi-armed Bandits, Active Learning
- 14. Online Learning
- 15. Ranking
- 16. Minimax lower bounds

Grading: Grades will be based on class participation, quizzes, homework assignments, presentations and/or project reports.

Textbooks: A textbook will not be followed in this course. A collection of notes, relevant papers and materials will be prepared and distributed.

Academic Integrity: Absolute professionalism in matters of academic integrity is expected. The minimum penalty for collaboration on an examination in this class will be a grade of E for the class. Other forms of academic misconduct will also incur harsh penalties. You are urged to review the ABOR Student Code of Conduct and disciplinary procedures.

http://students.asu.edu/files/StudentCodeofConduct.pdf

http://students.asu.edu/files/StudentDisciplinaryProceduresChapter5.pdf