**Instructor:** Urbashi Mitra, Professor  
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**Course Objectives:** To provide a fundamental understanding of concepts and techniques in sparse approximation as applied to electrical engineering. The emphasis will be on developing the analysis and design tools needed to apply such methods to research.

**Prerequisites:** EE441 (linear algebra) and (EE464 or EE465)(probability)

**Other Requirements:** Basic computer skills (i.e. programming and plotting, familiarity with Matlab).

**Recommended Courses:** EE563 (estimation theory), EE568 (coding theory), EE553 (optimization)

**Text:** there is no required text, we shall exploit course notes and seminal journal articles; however, the following are recommended


3. An Introduction to Compressive Sensing, Richard Baraniuk, Mark Davenport, Marco Duarte, Chinmay Hegde, Wotao Yin, Mona Sheikh, and Jason Laska, Connexions 2011 (http://cnx.org/content/col11133/1.5)


**Grading:** (tentative) 30% Homework and journal article summaries  
30% Midterm  
40% Final Project

Final grades will be assigned by a combination of student score distribution (curve) and the discretion of the instructor.

**Students with Disabilities:** Any student requesting academic accommodations based on a disability is required to register with Disability Services and Programs (DSP) each semester. A letter of verification for approved accommodations can be obtained from DSP. Please be sure the letter is delivered to me (or to TA) as early in the semester as possible. DSP is located in STU 301 and is open 8:30 a.m. - 5:00 p.m., Monday through Friday. The phone number for DSP is (213) 740-0776.

**Cheating:** Cheating or plagiarism will not be tolerated on homework or exams. You may discuss homework problems among yourselves but each person must do their own work. Copying or turning in identical homework sets is cheating. The penalty ranges from 0 points on the homework or exam, to an F in the course, to recommended expulsion. See:  
http://www.usc.edu/dept/publications/SCAMPUS/gov/academic_integrity.html
Outline: (each item roughly corresponds to one week’s material)

1. Vector spaces
2. Estimation and detection
3. Large Deviations theory
4. Compressible and sparse signals
5. $l_1$ minimization
6. Probabilistic approach to compressed sensing
7. Deterministic approach to compressed sensing
8. Sensing matrices
9. Sparse Approximation in noise
10. Optimality of sparse approximation/compressed sensing
11. Low-rank matrix recovery
12. Nuclear-norm minimization
13. Exploiting structure
14. Applications
15. Final Project Presentations