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5. J. Meditch, "Stochastic Optimal Linear Estimation and Control," McGraw-Hill, 1969
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7. H. Sorenson, "Least squares estimation from Gauss to Kalman," *IEEE Spectrum*, vol 7, pp 63-68, 1970
8. A. Gelb, "Applied Optimal Estimation," MIT Press, Cambridge, MA 1974
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10. R. Bellman, "Some Consequences of the Nonnegativity of the Elements of the Matrix Exponential," *Nonlinear Analysis, Theory, Methods, and Applications*, vol 4, no 4, pp 735-736, 1980
11. H. Sorenson, "Parameter Estimation," Marcel-Dekker, 1980
12. G. Goodwin and K. Sin, "Adaptive Filtering, Prediction, and Control," Prentice-Hall, 1984
13. H. Sorenson, "Kalman Filtering: Theory and Applications," IEEE Press, 1985
14. G. Golub and C. Van Loan, "Matrix Computations," The Johns Hopkins University Press, 1989
15. L.L. Scharf, "Statistical Signal Processing: Detection, Estimation, and Times series Analysis," Addison-Wesley, 1991
16. J.M. Mendel, "Lessons in Estimation Theory for Signal Processing," Prentice-Hall, 1995
17. E. Brookner, "Tracking and Kalman Filtering Made Easy," John Wiley and Sons, 1998

18. B. Ristic, S. Arulampalam, and N. Gordon, "Beyond the Kalman Filter: Particle Filters for Tracking Applications," Artech House, Norwell, MA 2004
19. D. Simon, "Optimal Estimation: Kalman, H_∞ , and Nonlinear Approaches," Wiley-Interscience, 2006

Languages of Instruction: English, Mathematics, and MATLAB

Exams and Percent of Grade:

Homework and programs	50%
Midterm	25%
Final	25%

Homework: Assigned every Thurs and due at the start of class the next Thurs.

MATLAB Programs: Approximately 6, included as part of the weekly homework.

Course Objectives:

Detection, estimation, and time series analysis are the main branches of statistical signal processing. Estimation theory is perhaps the most fundamental of these three, as it plays an important role in the other two. Our objective is to lay the probabilistic foundations for estimation and then to develop its main lines, from parameter estimation to recursive least squares to Wiener and Kalman filtering. Then we address selected topics among linear prediction, modal analysis, multi-sensor array processing, time-frequency analysis, Gaussian sums, particle filtering, and the like.

Course Outline (provisional after first 6 topics, depending on student interest)

1. Deterministic least squares, including RLS
2. Stochastic least squares, including channel and filtering models
3. The Wiener filter, including spectral factorization
4. The Kalman filter
5. Linear prediction and modal analysis
6. Fast algorithms of the Levinson, Schur, conjugate gradient type
7. Multi-sensor array processing
8. Gaussian sums and particle filters
9. Time-Frequency analysis
10. HMM, Baum-Welch, and EM
11. Compressed sensing

MATLAB Experiments:

1. various experiments in numerical linear algebra
2. more to come

Assumed Input Skills:

1. Linear Systems Theory
2. Probability Theory
3. Rudiments of Complex Analysis

Presumed Output Skills:

1. Command of the course syllabus