Department of Electrical & Computer Engineering University of California, Santa Barbara ECE 240A Winter 2010 Shynk H.O. #2

# ECE 240A OPTIMAL ESTIMATION AND FILTERING TENTATIVE COURSE OUTLINE

LINEAR MODEL (Lesson 2)

State-space formulation Random and deterministic signals Autoregressive model Moving average model Filtering, smoothing, and prediction

#### LEAST-SQUARES ESTIMATION (Lessons 3, 4, and 5)

Batch processing Orthogonality condition Singular value decomposition Recursive processing Information and covariance forms Initial conditions

### PROPERTIES OF ESTIMATORS (Lessons 6, 7, 8, and 9)

Small-sample properties Unbiasedness and efficiency Cramer-Rao inequality and Fisher's information Large-sample properties Stochastic convergence and consistency Properties of least-squares estimators Best linear unbiased estimation

#### SUFFICIENT STATISTICS (Lesson A)

Factorization theorem Exponential families of distributions Complete and sufficient statistics Uniformly minimum-variance unbiased estimation

### MAXIMUM LIKELIHOOD (ML) ESTIMATION (Lessons 10, 11, and 12)

Likelihood ratio Multiple hypotheses Maximum-likelihood method Log-likelihood function Multivariate Gaussian random variables

# MEAN-SQUARED (MS) ESTIMATION (Lesson 13)

Mean-square error Orthogonality principle Conditional mean estimator Nonlinear estimation

### MAXIMUM A POSTERIORI (MAP) ESTIMATION (Lesson 14)

Bayesian estimation Conditional likelihood function Detection theory Comparison of ML, MS, and MAP estimation

## STATE ESTIMATION (Lessons 15 and 16)

Gauss-Markov random sequences State-variable model Single-stage predictor Innovations process State prediction, filtering, and smoothing

## KALMAN FILTER (Lessons 17, 18, 19, 20, and 21)

Recursive estimation Properties of Kalman filter Whitening filter Steady-state Kalman filter Relationship to Wiener filter Smoothing: fixed interval, fixed point, and fixed lag