

## **Preliminary Course Outline for SIOG 223A, Winter 2020**

Instructor: Cathy Constable  
cconstable@ucsd.edu, x43183

4 units, 3 hour meeting/week, homework, S/U grades permitted. Expected time commitment 10 hours /week

**Prerequisites:** graduate standing or consent of instructor; familiarity with calculus, at least through multivariate calculus, and with vectors, vector spaces, and matrices

The goal of this class is to introduce MS and PhD students to fundamental methods for time domain geophysical data analysis that will be useful in their research. Frequency domain methods are covered in the second quarter SIOG 223B. In SIOG 223A basic statistics and methods for parameter estimation and model fitting are discussed in the context of geophysical examples. Students are expected to complete homework involving both analytical calculations and computations using Matlab or other programming language of their choice. Collaboration on homework is encouraged. Instructor reserves the right to require a final take-home exam if necessary for evaluation. Topics covered will include:

### **Chapter 0: Communicating your results: Graphical Rhetoric**

- Displaying what you want to show
- Using variables that can be decoded

### **Chapter 1: Introduction**

- Probability and statistics compared: Reality, models, and inference
- What kinds of questions can you ask? Estimation, Hypothesis testing

### **Chapter 2: Probability and random variables**

- Probability for events
- Conditional probability [and Bayes' Theorem]
- PDF's and CDF's; Lebesgue's Decomposition Theorem
- Expectations, means, variances, moments
- The Central Limit Theorem

### **Chapter 3: Some distributions**

- Uniform, Normal, Poisson, chi-square, Exponential, gamma, lognormal, Weibull, chi-squared, t and F, von Mises and Fisher

### **Chapter 4: Multivariate Random Variables, Correlation and Error Propagation**

- Multivariate PDFs
- Conditionals and Marginals
- Moments of Multivariate PDFs
- Independence and Correlation
- Regression
- Multivariate Normal Distribution

### **Chapter 5: Parameter Estimation**

- The simplest estimation: Method of Moments
- Order Statistics
- Trimmed estimates
- Sampling distributions for Statistics
- Monte Carlo Methods
- Bootstrap Methods
- Confidence Limits for Statistics
- Desirable properties for estimators:
  - unbiasedness, efficiency, minimum Mean Square Error, consistency, robustness
- Maximum likelihood
- Cramer-Rao Inequality
- $L_1$  norm estimation

### **Chapter 6: Hypothesis testing**

- How does hypothesis testing work
- what does "95% confidence" mean and why do you care?

The general framework

Examples: the Schuster tests

Tests for the same mean

Test for pdfs: Kolmogorov-Smirnov test;  $\chi^2$  test for goodness of fit; Q-Q Plots

### **Chapter 7: Least Squares Estimation**

Least squares estimation

Assessing Fit

Correlation and Regression

Normal equations in matrix form

Statistical Properties of LS estimates, inferences about derived parameters

Weighted LS

Numerical issues

### **Chapter 8: Total Least Squares and Robust methods**

Total Least squares and the bootstrap

Robustness, Non-Gaussian data errors, and M-type Estimation

### **Chapter 9: Non-parametric Density Function Estimation**

Density estimates and sample distribution functions - comparing data and theory

Adaptive Estimation: Nearest Neighbors and Variable Kernels

Maximum Penalized likelihood estimators

### **Chapter 10: Basic Introduction to Inverse Methods**

Difference between inversion and parameter estimation

Ambiguity and non-uniqueness

Mathematical Optimization

Stochastic Methods