Instructor: Elly Kaizar  
221 Cockins Hall  
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(614) 247-2585

Office Hours: Wednesday 4:00-5:00pm, or by appointment

Course website: http://carmen.osu.edu

Course description:  
This class will review statistical analysis for complete data and provide an introduction to the models and methods for the dataset with missing values. The course has a significant component of statistical computations dealing with missing data. It is intended for those who already have some experience with standard statistical methods for complete data and want to extend them to handle the missing data in practice.

Course Objectives:  
After the completion of this course, the students are expected to
1. Understand the missing data mechanism, the underlying assumptions and identify different patterns of missing data
2. Understand the difference in statistical analysis between missing data problem and complete data problem (including weighted methods)
3. Be able to perform simple missing data analysis with single imputation; comprehend its weakness
4. Be able to implement likelihood-based analysis with ignorable missing response; implement EM algorithm with some statistical package
5. Understand the principle of Bayesian analysis with missing data; implement multiple imputation with some statistical package
6. Understand missing data models in contingency tables
7. Have a basic understanding of the recent development of statistical methods to deal with non-ignorable missing data
8. Be able to implement and interpret statistical methods for missing data in a practical scenario.

Required Texts:  

Reference:  
Software:
The use of computer in data analysis is essential for this course. There are many different algorithms used in analysis of missing data, and so we will be discussing several different analysis packages. The software program R will be primarily used for examples, homework and exams.

If you are not already familiar with R programming, there are a plethora of online tutorials available. I recommend the video tutorials by Dan Goldstein you can find here: http://www.decisionsciencenews.com/2007/09/26/r-video-tutorial-number-1/

Prerequisites:
STAT 6201, 6302 (623), or 6802 (622), and STAT 6450 (645), 6950, PubHBio 6203, or 703; or permission of instructor. Students must be familiar with maximum likelihood estimation and regression analysis.

Lectures:
M/W/F 3:00 - 3:55PM, Scott Lab N0054

Tentative Schedule:

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<th>Readings</th>
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<td>Missing data mechanism</td>
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<td>Likelihood-based approach</td>
<td>Chap 6.1-6.3, 7.1-7.4</td>
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<td>6-8</td>
<td>Bayesian approach and Multiple Imputation (and single value imputation)</td>
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<td>EM algorithm and large sample inference for ignorable missing data</td>
<td>Chap 8.1-8.5</td>
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<td>9-10</td>
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Evaluation:

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<td>Final Project</td>
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Homework:
I anticipate 7 graded homework assignments. The lowest grade among these will be dropped from the final grade calculation. The remaining 6 grades will be weighted equally. Homework assignments should be submitted either in hard copy at class time or in my office, or electronically to the Carmen Dropbox. In both cases, your name should appear at the top of the first page. Electronic submissions must be in .pdf format. No late assignments will be accepted. The dropped homework grade is meant to give you flexibility when you have an emergency. If you find yourself in extraordinary circumstances where this allowance does not seem appropriate, please see me asap.

Exams:
There will be 2 in-class exams. The second exam is cumulative. You may bring a calculator (not a cell phone or other communication device) to both. You may bring a single 8.5x11 sheet of paper with your own notes on both sides to the first exam, and two such sheets of paper to the second exam. These sheets may be prepared any way you like, and will not be collected.

_Tentative_ exam dates:
   September 18
   November 1

Final Project:
The culmination of this course will be a final project that will be completed in groups. Your group must produce:
   1. Written proposal (1 page maximum) that describes your idea and data source
   2. Final written report first draft
   3. Poster presentation
   4. Written critique of other groups’ project reports and presentations
   5. Final written report (5 page maximum, excluding figures and appendices)
More information about the final project will be provided in the first week of the course.

Academic integrity:
Cheating, plagiarism and other forms of academic dishonesty will not be tolerated. Any violation will be prosecuted to the fullest extent as set out in University Rule 3335-31-02.

Accommodation for special needs
Any student who feels they may need an accommodation based on the impact of a disability should contact the instructor privately to discuss your specific needs. You should also contact the Office of Disability Services at 292-3307 or in 150 Pomerene Hall to coordinate reasonable accommodations for students with documented disabilities.