## BSTA 7820 Statistical Methods for Incomplete Data Spring, 2024 Blockley 418

Instructor:	Qi Long, 201 Blockley, (215)573-0659, qlong@upenn.edu
Lectures:	January 22 – May 1, Monday and Wednesday, 1:45-3:15 pm No lectures on March 4 and 6 (Spring Break)
	Canceled lectures: March 11 and 13 (2024 ENAR Spring Meeting); March 20 (GGEB Work-in-Progress)
Office Hours:	Qi Long, Blockley 201, Monday 3:15–4pm (cancelled when lecture is cancelled) or by appointment
Textbooks:	Statistical Analysis with Missing Data, 3rd Edition, by Little, R.J.A, and Rubin, D., John Wiley & Sons (2019).
Prerequisites:	BSTA 621/622, BSTA 632, BSTA 651, or their equivalents; permission of instructor. Knowledge about Bayesian modeling, though not required, can be helpful.
Course Description:	This course reviews the theory and methodology of incomplete data, covering missing data patterns, missing data mechanisms including MCAR, MAR, and MNAR, potential impacts of missing data on data analysis; imputation methods; likelihood-based methods for handling missing data; computational methods such as the EM algorithm and its extensions; semiparametric methods for missing data such as IPW and AIPW; methods for MNAR and nonignorable missingness includ- ing sensitivity analysis. If time permits, it will also cover additional advanced topics on analysis of incomplete data

## **Outline of Lectures**

	• Part 1: Introduction (missing data patterns; missing data mechanisms; overview of missing data methods).
	• Part 2: Ad hoc methods for handling missing data (complete-case analysis; available-case analysis; LOCF).
	• Part 3: Single and multiple imputation methods.
	• Part 4: Likelihood-based methods; EM algorithm.
	• Part 5: Inverse Probability Weighting (IPW) and Augmented IPW (AIPW) methods.
	• Part 6: Methods for handling Missing Not At Random (MNAR) including pattern mixture models, selection models, and sensitivity analysis.
	• Part 7 (if time permits): Advanced topics for analysis of incomplete data, e.g., fairness in analysis of incomplete data, and machine learning (ML) and deep learning (DL) imputation models for incomplete high-dimensional data etc.
Grading Policy:	
	• Attendance and Participation @ $40\%$
	• Homework @ 40%: 4 homework assignments with 10% for each
	• Final presentation @ $20\%$
Grades:	
	• $(85, 100] \approx A$
	• $(75, 85] \approx B$
	• $(59,75] \approx C$
	• +/- grades will be given accordingly.
Final Presentation:	The final presentation will entail a review of 1-2 papers related to analy- sis of incomplete data and is scheduled for April 24, April 29 and May 1 at 1:45-3:15pm in Blockley 418. Each student will sign up for a 15-min slot including Q&A. Students are expected to provide the instructor by April 10 the list of paper(s) that will be presented.

Misc Notice: All course materials (e.g., outlines, handouts, syllabus exams, PowerPoint presentations, lectures, audio and video recordings, et.c) are proprietary. Students are prohibited from posting, sharing, or selling any such course materials without the express written permission of the professor teaching this course.