



Syllabus

BST 462 – Biostatistical Methods II Spring, 2022

Instructors:

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Office Hours: By appointment

Classes: Monday / Wednesday, 1:30-3:10 PM, SRB 1.406.

Dates: January 12 – April 27

Course Description

This is a course on advanced statistical methods commonly employed in the analysis of data in biomedical research. The course will focus on application of statistical methods rather than their underlying theory. Inferential procedures (hypothesis testing, interval estimation), checking of model assumptions, and interpretation of results will be emphasized. SAS and R software will be used to illustrate the methods. More advanced courses (BST 426 – Linear Models; BST 479 – Generalized Linear Models; BST 513 – Analysis of Longitudinal / Dependent Data; BST 514 – Survival Analysis; BST 531 – Nonparametric Inference) are devoted to the theory underlying most of the methods covered in this course.

The following is a list of the major topics to be covered in the course (approximate timing in brackets):

- *Generalized linear models*: Multicategory logistic regression; Poisson regression; negative binomial regression [Weeks 1-2; Professor Wu]
- *Analysis of normally-distributed dependent data*: Random and mixed effects models; covariance pattern models; random coefficient models; best linear unbiased predictors [Weeks 2-4; Professor McDermott]
- *Missing data*: Missing data mechanisms; limitations of single imputation methods; direct likelihood methods; multiple imputation; inverse probability weighting [Weeks 5-6; Professor McDermott]



- *Analysis of time-to-event data*: Censoring; Kaplan-Meier estimator; log-rank test; Cox proportional hazards model; time-dependent covariates; analysis of grouped time-to-event data; analysis of left-truncated data [Weeks 6-10; Professor McDermott]
- *Shrinkage estimation and variable selection*: Variable selection; dimension reduction; ridge regression; lasso and its variations; elastic nets; more general penalty functions [Weeks 10-12; Professor Wu]
- *Nonparametric methods*: Wilcoxon signed rank and rank sum tests; Kruskal-Wallis test; rank correlation [Week 12; Professor Wu]
- *Bootstrap methods*: Nonparametric bootstrap; standard error estimation; percentile-based confidence intervals; bootstrap failure; bias-corrected and accelerated (BCa) bootstrap [Week 13; Professor McDermott]
- *Presentation of student projects* [Week 14]

Prerequisites

BST 461 – Biostatistical Methods I, or permission of instructor(s).

Course Aims and Objectives

The course has three major objectives:

1. To provide a detailed understanding of the application of advanced statistical methods to yield inferences about the target population that address the scientific questions / hypotheses of interest.
2. To illustrate the use of descriptive tools to examine the appropriateness of the assumptions underlying the statistical methods that are used.
3. To provide an introduction to the use of SAS and R software to appropriately apply the methods covered in the course.

Course Policies and Expectations

The following policies/expectations apply to this course:

- You are expected to attend all classes. There is a required component of class participation and questions are encouraged.
- You should have your video turned on during all Zoom sessions if at all possible.
- Cell phones should be turned off or set to vibrate for emergencies.
- We do not have a fixed schedule for office hours, but we are more than happy to make appointments with students via Zoom or in-person to address questions or anything else related to the class.
- You are encouraged to discuss homework assignments with other students in the class, but you must write up the final solutions yourself. Please refer to the department policy on student collaboration: <https://www.urmc.rochester.edu/biostat/courses/studentcollaboration.aspx>
- Late homework assignments will be permitted only in case of emergency or with prior approval of the instructor.

Course Materials



The primary material for the course will be the presentation slides and output from SAS and R programs. The following books may also be useful:

1. Agresti A. An Introduction to Categorical Data Analysis, Second Edition. Hoboken, NJ: John Wiley and Sons, 2007. This is available in the department library (CAT 5.705) and through the University of Rochester Library as a free eBook (<https://ebookcentral.proquest.com/lib/rochester/detail.action?docID=290465>).
2. Brown H, Prescott R. Applied Mixed Models in Medicine, Third Edition. Chichester, UK: John Wiley and Sons, 2015. This is available through the University of Rochester Library as a free eBook (<https://ebookcentral.proquest.com/lib/rochester/detail.action?docID=1896285>).
3. Fitzmaurice GM, Laird NM, Ware JH. Applied Longitudinal Analysis, First Edition. Hoboken, NJ: John Wiley and Sons, 2004. This is available in the department library (LINM 148.100).
4. Molenberghs G, Kenward MG. Missing Data in Clinical Studies. Chichester, UK: John Wiley and Sons, 2007. This is available in the department library (APPL 327.900).
5. Hosmer DW Jr, Lemeshow S, May S. Applied Survival Analysis: Regression Modeling of Time-to-Event Data, Second Edition. Hoboken, NJ: John Wiley and Sons, 2008. This is available in the department library (APPL 211.114) and through the University of Rochester Library as a free eBook (<https://www.oreilly.com/library/view/applied-survival-analysis/9781118211588/?ar>).
6. Hastie T, Tibshirani R, Friedman J. The Elements of Statistical Learning, Second Edition. New York, NY: Springer, 2009. This is available for free on the first author's website (<https://web.stanford.edu/~hastie/ElemStatLearn/download.html>).
7. Rosner B. Fundamentals of Biostatistics, Sixth Edition. Belmont, CA: Thomson Brooks/Cole, 2006. This is available in the department library (TEXT 380.102).
8. Davison AC, Hinkley DV. Bootstrap Methods and their Application. Cambridge, UK: Cambridge University Press, 1997. This is available in the department library (STAT 104.001, STAT 104.002, STAT 104.010).

Material for the course will be drawn from all of these sources (and others).

The homework assignments will involve the use of SAS and R software. There will be in-class instruction on how to write SAS and R programs to analyze data using the methods introduced in the course. The following resources should be helpful:

- Delwiche LD, Slaughter SJ. The Little SAS Book: A Primer, Sixth Edition. Cary, NC: The SAS Institute, Inc., 2019. This is available through the University of Rochester Library as a free eBook (<https://library-books24x7-com.ezp.lib.rochester.edu/toc.aspx?site=WLK0Y&bookid=148247> or <https://www.oreilly.com/library/view/the-little-sas/9781642953435/?ar>).
- SAS documentation
https://documentation.sas.com/doc/en/pgmsascdc/9.4_3.5/pgmsassyntaxwlc/home.htm
- Statistical computing from UCLA
<https://stats.idre.ucla.edu/r/>
<https://stats.idre.ucla.edu/other/examples/icda/>



- Introduction to R
<http://cran.r-project.org/doc/manuals/R-intro.pdf>

Assignments and Grading Procedures

The final grade will be based on homework assignments (60%), class participation (10%), and a final project (30%). In general, homework assignments will be due 1-2 weeks after being distributed. Graded assignments will be returned within one week.

Each student needs to complete a project involving the analysis of a data set of his/her choosing. One of the instructors will be assigned to supervise the student on the project. The student will meet with the assigned instructor to discuss the data set, research question, and analytic methods to be used. The project has to be approved by the instructor. Methods and techniques learned in this course should be applied to analyze the chosen data set. Students will give 20-minute presentations of their projects during the final two classes of the semester. A written report must be submitted during the final examination period and should include the following:

Introduction: A description of the background to the project should be provided, including a brief account of the scientific rationale and objectives of the study.

Methods: Some topics to consider are type of study (observational or experimental); study population; method of sampling and choice of sample size; description of intervention(s), if any; method of randomization, if any; response variables; and covariates and potential confounding variables. A description of the statistical methodology used should be provided, with references to the literature where appropriate. The description should be detailed enough to allow another competent statistician, given the same data and computing facilities, to reproduce major steps in the analysis.

Results: A summary of the major results from the analysis should be provided using both descriptive and inferential methods. The assumptions of any statistical models used should be checked using appropriate methods; the results of model verification should also be reported.

Conclusions: A clear interpretation of the findings of the study should be provided. This section must be written for general readers and should be in non-statistical language.

Academic Integrity

Academic integrity is a core value of the University of Rochester. Students who violate the University of Rochester "University Policy on Academic Honesty" are subject to disciplinary penalties, including the possibility of failure in the course and/or dismissal from the University. Since academic dishonesty harms the individual, other students, and the integrity of the University, policies on academic dishonesty are strictly enforced. For further information on the University of Rochester Policy on Academic Honesty, please visit the following website:

http://www.rochester.edu/college/honesty/docs/Academic_Honesty.pdf

Accommodations for Students with Disabilities



Students needing academic adjustments or accommodations because of a documented disability must contact the Disability Resource Coordinator for the school in which they are enrolled (see link below for contact information).

<http://www.rochester.edu/eoc/DisabilityCoordinators.html>

Course Schedule

Classes will be held twice per week, but please note that there will be no class held on January 17 (Martin Luther King, Jr. Day).