17.800: Quantitative Research Methods I

Fall 2021

Instructor: Teppei Yamamoto

TAs: Emilia Simison & Lukas Wolters

Department of Political Science

MIT

Contact Information

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<td>W 2-3 pm</td>
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<td>Office # / Zoom ID:</td>
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Logistics

- Lectures: M & W 9:30–11:00 am in Room E51-372
- Recitations: F 10-11 am in Room TBD
- Office Hours: See “Contact Information” above

Note that the first class meets on September 8, and the last class meets on December 8. There will be no class/recitation on the following days: Monday, October 11 (Indigenous Peoples Day); Wednesday, November 24 - Friday, November 26 (Thanksgiving).

Please also note that enrollment is capped at 30 students due to capacity constraints and priority is given to political science graduate students. Therefore, we cannot guarantee a spot for students from other departments or outside MIT. In case there is excess demand, students will be notified whether they are permitted to remain enrolled in the class by the end of the second week of class.

All class activities (lectures, recitation sessions, and office hours) are going to happen in person on campus, with the exception of Lukas’s office hours which will be held on Zoom. In addition, the recitation session on October 8th will be held online via Zoom. We ask you to strictly follow MIT’s pandemic regulations on on-campus behavior when we meet in person; in particular, well-fitted face coverings are mandatory at all times and eating during class is prohibited. This is subject to modification in case of significant changes in public health conditions and/or MIT regulations, or if the instructor or TA becomes unable to provide instruction in person due to their health conditions. Should you be barred from attending a class session in person due to your own health condition and/or per instruction by MIT Medical, contact us immediately so we can figure out alternative arrangements.
Overview and Goals

This is the first course in a four-course sequence on quantitative political methodology. Political methodology is a major subfield of political science which deals with the development and application of statistical methods to problems in political science. The subsequent courses in the sequence are 17.802, 17.804, and 17.806. By the end of the sequence, students will be capable of understanding and confidently applying a variety of statistical methods and research designs that are essential for political science research.

This first course provides a graduate-level introduction to regression models, along with the basic principles of probability and statistics which are essential for understanding how regression works. Regression models are routinely used in political science, policy research, and other disciplines in social science. The principles learned in this course also provide a foundation for the general understanding of quantitative political methodology. If you ever want to collect quantitative data, analyze data, critically read an article which presents a data analysis, or think about the relationship between theory and the real world, then this course will be helpful for you.

You can only learn statistics by doing statistics. In recognition of this fact, the homework for this course will be extensive. In addition to the lectures and weekly homework assignments, there will be required and optional readings to enhance your understanding of the materials. You will find it helpful to read these not only once, but multiple times (before, during, and after the corresponding homework).

The class is open to interested graduate students from other departments. Qualified undergraduates can also take the course subject to permission of the instructors.

Prerequisites

This course has the following prerequisites:

- Understanding of the basic linear algebra and calculus equivalent to the contents covered in the department’s math camp. The notes used in the camp are posted on the Canvas course website (see below for the URL) for your reference.

- Basic knowledge of the R language.

- Willingness to work hard on unfamiliar materials.

In addition, you will benefit more from the class if you have taken one (or more) undergraduate classes in quantitative methodology (e.g., 17.803).

Course Requirements

Grades will be based on:

- Weekly homework assignments (45% of final grade)
- Final homework assignment (20% of final grade)
- Three in-class quizzes (30% of final grade)
  - Monday, October 4
  - Monday, November 8
• Participation (5% of final grade).

The weekly homework assignments will consist of analytical problems, computer simulations, and data analysis. They will usually be assigned on Wednesday night and due the following Wednesday, prior to lecture. All assignments must be submitted electronically through the class Canvas site. No late homework will be accepted unless you ask for special permission from the instructor in advance of the deadline. All sufficiently attempted homework (i.e. a typed and well organized write-up with all problems attempted) will be graded on a three-point scale (✓+, ✓, ✓−). You may re-write one assignment over the semester and have it regraded. If you choose to submit a re-write, it is due before the Wednesday lecture one week after the assignment is returned.

We encourage students to work together on the assignments, but you always need to write up and submit your own solutions. For the writeup, we strongly recommend software that is specifically designed for scientific writing. We especially recommend \LaTeX, which is used widely in political science and across scientific disciplines. Another option is R Markdown, which is simpler and particularly suited for documents that contain R code (but less commonly used for general scientific writing). We recommend against Microsoft Word, which is not designed for documents containing code or math. We also require that you make a solo effort at all the problems before consulting others in your group, and that you write the names of your co-workers on your assignments.

The final homework assignment of the term will be a special problem set, which will be longer than a regular problem set and weighted more heavily toward the calculation of the final grade. You will not be allowed to collaborate with anybody on the final problem set. This is to test if you have developed sufficient experience to work through problems on your own. No rewrite is permitted on the final assignment.

There will be three closed-book, in-class quizzes conducted on the dates listed above. The quizzes will occur during the first 30 minutes of the regular class time on those dates.

Although this course is primarily lecture-based, participation is important. We expect you to attend lectures and recitation sessions regularly. During the sessions, we encourage you to actively participate by asking questions and reacting to our questions. Participation in the Piazza discussion board (see below) will also count towards the participation grade.

Finally, please note that no incompletes will be given in this course except in highly unusual circumstances.

Notes on Regrading Requests

If you think that there is an error in grading your problem set or quiz answer, you should always start by reading the solution carefully one more time, making sure you understand it well. Then, read your own answer as if you were reading it for the first time, and ask yourself whether any of the differences might reasonably have caused the point reduction. Note that it is what you actually wrote in your answer, not what you intended to communicate by writing your answer, that should be considered a reasonable basis for the possible grading error. That is, your answer needs to be self-contained and cannot be supplemented by any information you provide outside of your answer sheet.

If you still think there was truly an error in your grade, you can request a re-grading by emailing one of us about your intent. The TA who graded your answer will get in touch with you and discuss your concerns. Unless the request is about a simple factual error the TA can resolve without any uncertainty (e.g. scores were not added up correctly), the TA will forward your entire problem set or quiz for re-grading by either the other TA or the instructor. The new grader will re-grade the
whole problem set or quiz anew, not only the particular problem you had a dispute about, without regard to the original scores. The new grade will then replace your original grade for the problem set or quiz and become final, regardless of whether it is higher or lower.

Notes on Academic Integrity

Please respect and follow the rules written in MIT’s handbook on academic integrity, which is available at:

http://web.mit.edu/academicintegrity/

In particular, the following is a (partial) list of the acts we will consider academically dishonest:

- Obtaining or consulting course materials from previous years
- Sharing course materials with people outside of the class, such as problem sets and solutions
- Copying and pasting someone else’s answers to problem sets electronically, even if you collaborated with the person in a legitimate way (as specified above)

Recitation Sessions and TA Help

Weekly recitation sessions will be held on Fridays from 10-11am in Room TBD. The session will cover a review of the theoretical material and also provide help with computing issues. The teaching assistant will run the sessions and can give more details. Attendance is strongly encouraged.

In addition to recitation, each TA will hold office hours once per week. See “Contact Information” above for the time and location. Questions will be addressed on a first-come, first-served basis.

The TAs for the course are very valuable resource for getting help on homework and understanding the material. While you are encouraged to talk with the TAs in office hours or via piazza if you run into problems with any aspect of the class, please also respect their time. In general, TAs are only expected to answer questions submitted via Piazza (see below) within 24 hours and they are not expected to answer queries during the weekend. In addition, in-person queries should be restricted to the TA’s posted office hours or recitation.

Course Website

The course website is located at the following URL:

https://canvas.mit.edu/courses/3556

This site will provide copies of lecture slides, recitation materials, homework assignments, data sets, and links to reading materials.

Questions about Course Materials

In this course, we will utilize an online discussion board called Piazza. This is a question-and-answer platform that is easy to use and designed to get you answers to questions quickly. We encourage you to use the Piazza Q&A board when asking questions about lectures, problem sets, and other course materials outside of recitation sessions and office hours. You can access the Piazza course page either directly from the below address or the link posted on the Canvas course website:
Using Piazza will allow students to see other students’ questions and learn from them. Both the TAs and the instructor will regularly check the board and answer questions posted, although everyone else is also encouraged to contribute to the discussion. A student’s respectful and constructive participation on the forum will count toward his/her class participation grade. Do not email your questions directly to the instructors or TAs (unless they are of personal nature) — we will not answer them!

Notes on Computing

We teach this course in R, an open-source statistical computing environment that is very widely used in statistics and political science. You can download it for free from www.r-project.org. We recommend that you use the RStudio IDE (integrated development environment) to work with R, which can be downloaded for free from http://www.rstudio.com.

If you did not attend the R portion of the department’s math camp, the web provides many great tutorials and resources that can provide you with equivalent knowledge. This list is a good list to start. A quick nice way to start you off is the R tutorial created by DataCamp: here. R runs on a wide variety of UNIX-based platforms (including Mac OS X), Windows and Linux – you can download and use it even if your computer is 10 years old. R makes programming very easy, has strong graphical capabilities, and also contains canned functions for most commonly used statistical procedures.

Books

Required Books

There will be required readings for each section of the course. Students are expected to complete them before the relevant materials are covered in the lectures. The following textbooks are required and will be used throughout the course.


If you are new to R and did not participate in the department’s math camp, you are expected to work through the following tutorial (or an equivalent) as early as possible during the first few weeks of the semester: *Introduction to R* by DataCamp.

Optional Books

The following books are optional but may prove useful to students looking for additional coverage of some of the course topics. Some books are provided by MIT Libraries. Students can access available titles using the links below.

*Other good textbooks:*

• Andrew, Gelman and Jennifer Hill. *Data Analysis Using Regression and Multilevel/Hierarchical Models*. Cambridge University Press. (regression modeling) [Link](#).

• Fox, John and Sanford Weisberg. *An R Companion to Applied Regression*. 2nd ed. (R, with focus on regression modeling)

*For math background:*


*For visualizing data (conceptual):*

• Cleveland, William S. *Visualizing Data*. Summit, NJ: Hobart Press.


*For visualizing data (implementation in R):*


• Wickham, Hadley. *ggplot2: Elegant Graphics for Data Analysis*. Springer. [Link](#).

• Sarkar, Deepayan. *Lattice: Multivariate Data Visualization with R*. Springer. [Link](#).
Course Contents and Reading Assignments

1 Introduction
- Overview and Course Requirements
- Course Outline

2 Elementary Probability Theory
- Why Do We Need Probability?
- Probability Axioms
- Marginal, Joint and Conditional Probability
- Law of Total Probability
- Bayes’ Rule
- Independence

Required Readings:
- Bertsekas and Tsitsiklis, Chapter 1
- Wooldridge, Appendix A

3 Random Variables and Probability Distributions
- Discrete and Continuous Random Variables
- Measures of Location
- Measures of Dispersion
- Probability Distributions

Required Readings:
- Bertsekas and Tsitsiklis, Chapters 2.1–2.4 & 3.1–3.3
- Wooldridge, Appendix B.1 & B.3

4 Multiple Random Variables
- Joint and Conditional Distributions
- Conditional Expectation
- Covariance and Independence

Required Readings:
- Bertsekas and Tsitsiklis, Chapters 2.5–2.8, 3.4–3.7, 4.2 & 4.3
- Wooldridge, Appendix B.2 & B.4–B.5
5  Univariate Statistical Inference

5.1  Point Estimation
   • Properties of Estimators
   • Sampling Distribution
   • Elementary Asymptotic Theory

5.2  Interval Estimation
   • Confidence Intervals

5.3  Hypothesis Testing
   • Logic of Statistical Testing
   • p-Values

Required Readings:
   • Wooldridge, Appendix C
   • Bertsekas and Tsitsiklis, Chapter 5

6  What is Regression?
   • Nonparametric Regression
   • Linear Regression
   • Bias-Variance Tradeoff

Required Readings:
   • Wooldridge, Chapter 1

7  Simple Linear Regression
   • Mechanics of Ordinary Least Squares
   • Linear Model Assumptions
   • Properties of the Least Squares Estimator
   • Gauss-Markov Theorem
   • Testing and Confidence Intervals
   • Large Sample Inference

Required Readings:
8 Linear Regression with Two Regressors

8.1 Mechanics of Regression with Two Regressors
- Motivation for Multiple Regression
- Mechanics and Inference in OLS with Two Regressors

8.2 Omitted Variables and Multicollinearity
- Omitted Variable Bias
- Multicollinearity

8.3 Dummy Variables, Interactions and Polynomials
- Dummy Variables
- Interaction Terms
- Polynomials and Logarithms

Required Readings:
- Wooldridge, Chapters 3–7

9 Multiple Linear Regression

9.1 Mechanics of Multiple Regression
- Review of Matrix Algebra and Vector Calculus
- Mechanics of Multiple Linear Regression

9.2 Statistical Inference with Multiple Regression
- Statistical Inference for Multiple Linear Regression
- Testing Multiple Hypotheses

Required Readings:
- Wooldridge, Appendix D & E
10 Diagnosing and Fixing Problems in Linear Regression

10.1 Outliers and Influential Observations
- Plotting Residuals
- Standardized and Studentized Residuals
- Added Variable and Component Residual Plots
- Leverage and Influence

10.2 Heteroskedasticity, Serial Correlation and Clustering
- Weighted Least Squares
- Generalized Least Squares
- Heteroskedasticity-robust Standard Errors
- Cluster-robust Standard Errors
- Autocorrelation

10.3 Measurement Error
- Types of Measurement Errors
- Measurement Error in the Dependent Variable
- Measurement Error in an Independent Variable

Required Readings:
- Wooldridge, Chapters 8–9

Optional Readings:
- Wand, Jonathan; Kenneth Shotts; Jasjeet Sekhon; Walter Mebane; Michael Herron; and Henry Brady. 2001 “The Butterfly Did It: The Aberrant Vote for Buchanan in Palm Beach County, Florida.” APSR. 95: 793-810.

11 Extensions and Advanced Topics (time permitting)
- Nonlinear Regression Models
  - Logit and Probit Models
  - Generalized Linear Models
- Semiparametric and Nonparametric Regression Models
– Generalized Additive Models

Required Readings:

• Wooldridge, Chapter 17.1

Optional Readings: