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NOTE: Syllabus is under construction and will change before the start of term.

General Information
How can we detect voting irregularities? What causes individuals to vote? In what sense (if any) does democracy (or trade) reduce the probability of war? Quantitative political scientists address these questions and many others by using and developing statistical methods that are informed by theories in political science and the social sciences more generally. In this course, we provide an introduction to the tools used in basic quantitative social science research. The first four weeks of the course cover introductory univariate statistics, while the remainder of the course focuses on linear regression models. Furthermore, the principles learned in this course provide a foundation for the future study of more advanced topics in quantitative political methodology.
While the tools of statistical inference are worth studying in their own right, the primary goal of this course is to provide graduate students (and some undergraduates) with the necessary skills to critically read, interpret, and replicate the quantitative content of many political science articles. As such, the statistical methods covered in this course will be presented within the context of a number of articles. Throughout the term, we will reanalyze the data and revisit the conclusions from various prominent papers in the social sciences.

**Who should take which course number?**

We have designed the class with a great deal of flexibility in mind and have various course numbers that correspond to students with different backgrounds. Note that all sections of the course will use the R statistical computing environment. You are more than welcome to use other programs for your computation, but we will not support them.

**Gov 2000**  This is the default course number for all graduate students who will be doing *any* empirical research in political science or the social sciences more generally. Even if you think you’re going to use only qualitative methods in your research, you should still take this course to give yourself a solid footing and understanding of quantitative methods. Talk to older graduate students and faculty—you’ll need to know more methods than you think both for the dissertation and the job market. This section will teach you to be flexible data analysts, capable of tailoring standard methods to the unique situation of each task. You will control the tools, not the other way around. You will learn to write and adjust code to replicate and critique results from the literature.

**Gov 2000e**  This course number designed for those students who plan to do *absolutely no empirical work* in political science. Even if this describes you, you would likely be better served in the long term by taking 2000. Students in this section will focus on the analysis and critique of methods and empirical work. Even if you have no plans to do empirical work, you will still need to evaluate that work, be it for a job talk or as referee. You will still do some data analysis, but the coding aspect of the class will be less emphasized. In its place, you will be expected to produce a higher and more competent level of analysis/criticism in all assignments (no free lunches).

**Gov 1000**  This is the default course number for undergraduate students and roughly covers the same material as Gov 2000 with special tailoring for undergraduates.

**Stat E-190**  This is the course number for Harvard Extension School students.
Prerequisites

The most important prerequisite is a willingness to work hard on possibly unfamiliar material. Statistical methods is like a language and it will take time and dedication to master its vocabulary, its grammar, and its idioms. This presents a challenge for us as instructors to give you the best intuition and a challenge for you as a student to work hard to internalize that intuition.

Formally, the prerequisites vary for different types of students. For graduate students in the Government Department there are no course prerequisites. For other graduate students, undergraduate students, and Extension School students, the prerequisite is GOV 50, GOV E-1005, or the equivalent.

For any student who meets the prerequisites yet is concerned with his or her preparedness for the course, we strongly encourage the following in advance of the semester. First, we recommend reading and working through the exercises in David Freedman, Robert Pisani, and Roger Purves, Statistics, 2007 (any of the older editions should suffice as well). Next, we encourage familiarization with the R for the section of the course the student intends on taking. Moreover, if the student plans on typesetting problem set answers in \LaTeX, familiarity with the \LaTeX markup language would be helpful. Resources on R and \LaTeX are available under the “Resources” tab on the class website.

Course Details

Reading

There are readings for each topic and they mostly cover the theory of the method along with some applications. Obviously, read the required readings and any others that pique your curiosity. In addition, though, engage with the readings: take notes, write down your impressions or confusions, talk with your classmates, preferably through Canvas (see below for more details). All of your classes should be pushing your research forward and you will be more creative the more you actively read.

Grading

- weekly homework assignments (50% of final grade)
- a midterm exam (10% of final grade)
- cumulative take-home final exam (30% of final grade)
- participation (10% of final grade).
Homeworks

Methods are tools and it isn't very instructive to read a lot about hammers or watch someone else wield a hammer. You need to get your hands on a hammer or two. Thus, in this course, you will have homeworks on a weekly basis. They will be a mix of analytic problems, computer simulations, and data analysis. For all sections, the homework will be due before the start of section (Thursdays at 5:00pm). Solutions will be posted on Thursday night after section. Students have the option to “self correct” one homework over the course of the term on the basis of the solution key (due the following Thursday at 5:00pm). These corrections should take the form of an updated homework with comments added to indicate where mistakes were made and that demonstrate an understanding of those mistakes.

These homeworks should be typed and well-formatted, with tables and figures incorporated into the text. We will grade on a (+, ✓, −) basis (including half grades between these categories). No late homework will be accepted except in the case of a documented emergency.

Midterm

The midterm will be a checkout exam, that should only take a few hours to complete, and only involves short analytical problems. You will have five hours to complete exam, but it should take less time than this. This exam will be available for checkout one week after we finish the material on univariate statistics, and it is designed to ensure that all students understand the foundational material before we move to regression. Both FAS and Extension School students will upload the completed exam to the course Canvas site.

Take-home Final

The take-home final exam will be handed out on Thursday, December 3, one week before the last day of reading period. It will be due at 5:00pm on Thursday, December 10, the last day of reading period. The take-home final is an exercise in guided replication and primarily involves data analysis and interpretation. Note that the format and goals for the take-home exam are very different from the format and goals for the midterm exam. Both FAS and Extension School students will upload the completed exam to the course Canvas site.

1 All sufficiently attempted homework will be typed and well organized with all problems attempted, and all sufficiently corrected homework will include typed and well organized comments integrated into the original homework. The instructor will determine sufficiency in borderline cases.
**Collaboration Policy**

We encourage students to work together on the homework assignments, but you must write your own solutions (this includes computer code), and you must write the names of your collaborators on your assignment. I also strongly suggest that you make a solo effort at all the problems before consulting others. The midterm and the final will be very difficult if you have no experience working on your own. **There is no collaboration allowed on either the midterm or the final exam.**

**Participation**

Ten percent of the grade will be awarded for class participation, quality of presentation on the homework, and reading comments. A preliminary version of the lecture notes will be posted on Friday evening with references to pages of the textbook on the notes. Posting questions on Canvas about the assigned reading or the lecture notes will count towards class participation. These comments and questions provide feedback for tailoring the Tuesday lecture to the needs of the students in the course.

**Discussion Sections**

There will be two discussion sections for this course. Both are on Thursday evenings, the first from 6-7pm and the second from 7-8pm. These sections will be taped and made available to all students.

**Course Canvas Site & Discussion Board**

We will be using Canvas to host the course website this year. You can find the site at the following URL: [https://canvas.harvard.edu/courses/4533](https://canvas.harvard.edu/courses/4533). On Canvas you will find a Discussion Board for class-related discussion. The quicker you begin asking questions on Canvas, the quicker you’ll benefit from the collective knowledge of your classmates and instructors. This is an ideal forum for posting questions regarding the course material and/or computing. I encourage students to reply to each other’s questions, and a student’s respectful and constructive participation on Canvas will count toward his/her class participation grade.

**Office Hours and Availability**

My office doors are almost always open during regular business hours and you should feel free to swing by to see me whenever. If you must need to see me at a particular time, please send me an email to make an appointment.

The office hours for Mayya Komisarchik and Anton Strezhnev are posted above and will be held in the CGIS Cafe, known as the Fisher Family Commons. If you have ques-
tions about the course material, computational issues, or other course-related issues please do not hesitate to set up an appointment with either me, Mayya, or Anton.

If you have a general question, you can also post it on Canvas. This is almost always the fastest way to get an answer. However, you can also email me directly at mblackwell@gov.harvard.edu. If the question is of general interest, I will forward the question and my answer to the class. Make sure to tell me explicitly in your email if you would like to stay anonymous.

**Required Books**

The following textbook is **required** for this course:

- Wooldridge, Jeffrey M. *Introductory Econometrics.* New York: South-Western. 5th edition. (earlier editions are fine)

**Optional Books**


- Berksekas, Dimitri P. and John N. Tsitsiklis, *Introduction to Probability.* Athena Scientific. (Also available as lecture notes online.)


- Hernán, Miguel A. and James M. Robins. 2012. *Causal Inference.* Forthcoming, Cambridge University Press. (Note that this book is still being written and you can find draft PDFs on the linked page above.)


**Computing**

We’ll use R in this class, which you can download for free at [http://www.r-project.org](http://www.r-project.org). R is open source and available on all major platforms (including Solaris, so no
excuses). You can find a virtually endless set of resources for R on the internet, including this Getting Started With R page. You may also be interested in using RStudio, an editor and development environment for R. If you are completely new to R, you should complete this online short course, Try R.

**Preliminary Schedule**

The following is a preliminary schedule of course topics. We may adjust the schedule due to comprehension, time, and interest. *Note: this schedule is subject to change.*

§1  **Introduction**

- Course Details and Requirements
- What are the goals of the course?

§2  **Basic Probability**

- The basics of probability
- Marginal, joint, and conditional probability
- Law of total probability
- Independence

Reading

- Bertsekas & Tsitsiklis, 1.1-1.5

§3  **Random Variables and Probability Distributions**

- Types of random variables
- Measures of central tendency
- Measures of spread
- Random variables, distributions
- Probability distributions

Reading
• Wooldridge, Appendix B.1, B.3, B.5
• Gelman and Hill, Ch. 2.1
• Bertsekas & Tsitsiklis, 2.1–2.4 & 3.1–3.3

§4 Multiple Random Variables
• Joint and conditional distributions
• Conditional expectation
• Covariance, correlation, and independence

Reading
• Wooldridge, Appendix B.2, B.4

§5 Inferences about a Single Variable
• Populations, samples, estimation
• Small samples, large samples, and asymptopia
• Properties of estimators
• Hypothesis testing, confidence intervals

Reading
• Wooldridge, Appendix C

§6 Regression, Causality, and the Statistical Model
• Potential outcomes and causal inference
• Difference in means
• Nonparametric regression
• Parametric models and linear regression
• Bias-variance tradeoff

Reading
• Wooldridge, Ch. 1
§7  **Simple Linear Regression**  
- Mechanics of Ordinary Least Squares  
- Assumptions of the linear model  
- Properties of least squares  
- Gauss-Markov Theorem  
- Inference with regression  

**Reading** 
- Wooldridge, Ch. 2

§8  **Linear Regression with Two Regressors**  
- Mechanics of regression with two regressors  
- Omitted variables and multicollinearity  
- Dummy variables, interactions, and polynomials  

**Reading** 
- Wooldridge, Ch. 3, 6.2, 7.1-7.4

§9  **Multiple Linear Regression**  
- Matrix algebra and mechanics of multiple linear regression  
- Inference in a multiple linear regression model  

**Reading** 
- Wooldridge, Ch. 4-5

§10  **Diagnosing and Fixing Problems**  
- Functional form  
- Model fit, outliers, and influential observations  
- Heteroskedasticity and non-Normal errors
· Measurement error

Reading

· Wooldridge, Ch. 6.3, 8, 9.4-9.6

§11 Panel Data Models (time permitting)

· Fixed effects
· Random effects
· Clustered standard errors
· Hierarchical models

Reading

· Wooldridge, Ch. 14