Syllabus
POLISCI 150C/355C
Causal Inference for Social Science
Spring 2023

Instructor: Alain Schläpfer
TA: Victoria Liu

Time & Location

Class: Monday and Wednesday, 1:30-2:50 pm, ECON 139

Section: Friday 9:30-10:20 am, 10:30-11:20 am

Contact Information

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Office Hours:	Wednesday 9-11am	Wednesday 4:30-5:30pm
Appointment Link:	https://calendly.com/alainsch	https://calendly.com/victoria_liu_cx/polisci-150c-oh

Overview and Class Goals

Causal inference methods have revolutionized the way we use data, statistics, and research design to move from correlation to causation and rigorously learn about the impact of some potential cause (e.g., a new policy or intervention) on some outcome (e.g., election results, levels of violence, poverty). This course provides an introduction to the toolkit of modern causal inference methods as they are now widely used across academic fields, government, industry, and non-profits. Topics include experiments, matching, regression, difference-in-differences, instrumental variable estimation, and regression discontinuity designs. We will illustrate and apply the methods with examples drawn from various fields including policy evaluation, political science, public health, economics, business, and sociology. We will further discuss in class relevant articles from the Political Science literature.

Prerequisites

Political Science 150A or an equivalent is required. The course assumes that you are proficient in `R`, as covered in 150A.

Class Requirements

Reading
The syllabus lists the required readings for each part of the course. Part of the reading is based on the text book, and should be completed prior to the corresponding lectures.
Additionally, we will discuss research studies from the political science literature that uses the tools and methods of causal inference. For these discussions, it is essential that students have all ready the relevant articles carefully and are prepared to discuss the content in class. Participation grades will mainly reflect active contributions during these discussions. The articles will be added to the syllabus as the term progresses.

Homework
This is a methodological course, developing skills in understanding and applying statistical methods. You can only learn statistics by doing statistics, and therefore the homework for this course is extensive, including a problem set roughly every two weeks. No late homework will be accepted. We encourage students to work together on the assignments, but you always need to write your own solutions, and we ask that you make a solo effort at all the problems before consulting others. In particular, you should not copy someone else's answers or computer code. For analytical questions, you should include your intermediate steps, as well as comments on those steps when appropriate. For data analysis questions, include annotated R code as part of your answers. All results should be presented so that they can be easily understood.

Exams and Project
There will be two midterm exams and a final presentation and report. The midterms will take place during weeks 4 and 8. Both midterms will consist of an open book exam. The final report will be due at the end of the last week of classes. Students are expected to write a short empirical paper that details a research design on a causal research question of their choice, drawing from the tools and methods discussed in class. The paper should be 5-8 pages in length and present a clear research question. Then the paper should explain how one of the methods of causal inference could be used to answer this question. The paper should further discuss the potential shortcomings and threats to the validity of the estimates.

Grading
Grades will be based on:

- Homework assignments (35 %)
- Midterm exams (30 %)
- Final presentation and project report (30 %)
- Class participation (5 %)

Sections

A weekly section will be held to review the material and provide help with computing issues. Victoria Liu will run the sections and provide more detail.

Sections will be held 9:10 and at 10:30 on Friday morning. Please enroll through Canvas.

Computation

In this course, we will use R.

To refresh your R, you are expected to work through one of the following free tutorials unless you are well familiar with this material. All three tutorials cover similar material, just pick the one you like best:
Owen. *The R Guide.* At: http://cran.r-project.org/doc/contrib/Owen-TheRGuide.pdf
Venables and Smith. *An Introduction to R.* At: http://cran.r-project.org/doc/manuals/R-intro.pdf
Verzani. *Simple R.* At: http://cran.r-project.org/doc/contrib/Verzani-SimpleR.pdf

In addition, there are many resources for R both new and advanced users:

- Wickham, Hadley and Garrett Grolemund. 2017 *R for Data Science.* Available for free online at here. (focused on 'tidyverse')
- Fox, John and Sanford Weisberg. 2010. *An R Companion to Applied Regression.* Sage Publications. (focused on regression analysis)
- Venables, W. N. and B. D. Ripley. 2002. *Modern Applied Statistics with S*, 4th ed. Springer. (general statistics)
- Teetor, Paul. 2011. *R Cookbook.* O'Reilly Press.

- For specific questions about R, searching the CRAN website with appropriate keywords will often yield satisfactory results.

Course Website

This course uses Ed-Discussion, which you can access through Canvas. Please use the Ed-Discussion board to ask any questions relevant to the course. This will allow students to see and learn from other students' questions. We will both regularly check the board and answer questions posted, although everyone else is also encouraged to contribute to the discussion. *Do not email your questions directly to the instructors or TAs* (unless they are of personal nature) — we will not answer them!

Books

- *Most required readings are from the following textbook.*
  - Angrist, Joshua D. and Jörn-Steffen Pischke. 2014. *Mastering Metrics.* Princeton University Press.

As an additional source, I can recommend *Causal Inference: The Mixtape* by Scott Cunningham, which can also be accessed freely online and includes example code. For a more advanced treatment of the material, I further recommend Angrist, Joshua D. and Jörn-Steffen Pischke, *Mostly Harmless Econometrics: An Empiricist's Companion.*

Students with Documented Disabilities

Students who may need an academic accommodation based on the impact of a disability must initiate the request with the Office of Accessible Education (OAE). Professional staff will evaluate the request with required documentation, recommend reasonable accommodations, and prepare an Accommodation Letter for faculty. Unless the student has a temporary disability, Accommodation letters are issued for the entire academic year. Students should contact the OAE as soon as possible since timely notice is needed to coordinate accommodations. The OAE is located at 563 Salvatierra Walk (phone: 650-723-1066, URL: https://oae.stanford.edu/).

Preliminary Schedule

The following is a preliminary schedule of course topics. Notice that required readings are marked with a ($\star$).

1 The Potential Outcome Model

- Counterfactual Responses and the Fundamental Identification Problem
- Estimands and Assignment Mechanisms
- Heterogeneity and Selection

Readings:

- *Mastering Metrics*, Ch. 1 p.1-17

2 Randomized Trials

- Estimation and Hypothesis testing
- Covariate adjustment, Examples, Blocking
- Complete, Block, and Cluster Randomization

Readings:

- *Mastering Metrics*, Ch. 1: p. 17-33

Articles:

- *Misperceived Social Norms: Women Working Outside the Home in Saudi Arabia*, Bursztyn, González, and Yanagizawa-Drott, American Economic Review, 2020.
- Clingingsmith, David, Asim Ijaz Khwaja, and Michael Kremer , *Estimating the Impact of the Hajj: Religion and Tolerance in Islam's Global Gathering*, Clingingsmith, Khwaja, and Kremer, Quarterly Journal of Economics, 2009.

3 Selection on Observables and Matching

- Identification under Selection on Observables
- *Readings: Selection on Observables*
  - *Mastering Metrics*, Ch. 2
- Matching
- *Readings: Matching*
  - Angrist and Pischke, *Mostly Harmless Econometrics*, Ch. 3, Section 3.3.1

Articles:

- *Identifying Judicial Empathy: Does Having Daughters Cause Judges to Rule for Women's Issues?*, Glynn and Senn, American Journal of Political Science, 2014.
- *Estimating the Effects of Human Rights Treaties on State Behavior*, Hill Jr., Journal of Politics, 2010.

4 Instrumental Variables

- Identification: Using Exogenous Variation in Treatment Intake Given by Instruments
- Imperfect Compliance in Randomized Studies
- Wald Estimator, Local Average Treatment Effects, 2SLS

Readings:

- *Mastering Metrics*, Ch. 3

Articles:

- *Conservation and development: Evidence from Thai protected areas*, Sims, Journal of Environmental Economics and Management, 2010.
- *Economic Shocks and Civil Conflict: An Instrumental Variables Approach*, Miguel, Satyanath, and Sergenti, Journal of Political Economy, 2004.

5 The Regression Discontinuity Design

- Sharp and Fuzzy Designs, Identification, Estimation, Falsification Checks

Readings:

- *Mastering Metrics*, Ch. 4

Articles:

- *What Happens When Extremists Win Primaries?* Hall, American Political Science Review, 2015
- *Nationalism and Conflict: Lessons from International Sports*, Bertoli, International Studies Quarterly, 2017.

6 Panel Data Methods

- Panel Data Methods and Fixed Effects Estimation

Readings:

- Angrist and Pischke, *Mostly Harmless Econometrics*, Ch. 5

7 Difference-in-Differences

- Identification, Estimation, Falsification tests

Readings:

- *Mastering Metrics*, Ch. 5

Articles:

- *Tabloid Media Campaigns and Public Opinion: Quasi-Experimental Evidence on Euroscepticism in England*, Foos and Bischof, American Political Science Review, 2021
- *The Economic Costs of Conflict: A Case Study of the Basque Country*, Abadie and Gardeazabal, American Economic Review, 2003.

Calendar (preliminary and subject to change)

Week	Tuesday	Thursday	Friday
1 (4/3&4/5)	PO 1	PO 2	
2 (4/10&4/12)	RT 1	RT 2	
3 (4/17&4/19)	Observables 1	Observables 2	PS 1 due
4 (4/24&4/26)	Observables 3	Exam	
5 (5/1&5/3)	IV 1	IV 2	PS 2 due
6 (5/8&5/10)	IV 3	RDD 1	
7 (5/15&5/17)	RDD 2	RDD 3	PS 3 due
8 (5/22&5/24)	Exam	Panel 1	
9 (5/29&5/31)	Panel 2	DiD 1	PS 4 due
10 (6/5&6/7)	DiD 2	Presentations	Project due