
Microeconometrics

Dmitry Arkhangelsky
darkhangel@cemfi.es

Fall, 2023

Logistics

Schedule

Lectures: Thu 9:30-11:00, 11:30 - 13:00;

Exercises: Mon 17:00-18:30;

Office hours: By appointment

TA: Juan Segura

Home assignments

Home assignments will include theoretical, computational, and empirical exercises. I strongly encourage you to use R for the coding problems. Home assignments should be done in groups with a minimal size of two and a maximal size of three students.

Grading

Grades will be based on group home assignments (10%), presentations (10%), and a final exam (80%).

Course Objectives

The primary goal of this course is to guarantee that students can (a) understand, (b) implement, and (c) discuss possible improvements of any empirical paper in economics/finance (from the econometric perspective). We will develop the necessary theory and will illustrate it with empirical exercises. A secondary goal is to introduce modern statistical techniques (ML and high-dimensional statistics) and show how they can be used to improve standard estimation procedures.

Prerequisites

First-year statistical and econometric courses should be sufficient. I will ask you to implement some of the procedures in R, which requires specific (pretty modest) coding skills.

Plan

1. Prediction: Empirical Risk Minimization and Regularization;
2. Basic Toolkit: Experiments, Weak Instruments, Unconfoundedness, RDD, Two-Way Models;
3. Beyond MHE: Selection Models, Marginal Treatment Effects;
4. General view: GMM and Related Problems;
5. Topics: Bipartite Networks, Quantile Methods, Shift-Share Designs;
6. Uncertainty Quantification: Design-Based vs. Model-Based Inference;

Literature

Textbooks

Lecture notes will be uploaded weakly on the course website. There is no textbook for the course, but the references below can be useful:

- Econometrics textbook by Bruce Hansen;
- “Causal Inference for Statistics, Social, and Biomedical Sciences: An Introduction”, by Guido W. Imbens and Donald B. Rubin;
- Manuel Arellano’s lecture notes (available on his website);
- Stefan Wager’s lecture notes for STATS 361 (available on his website)
- “Mostly Harmless Econometrics” by Joshua D. Angrist & Jorn-Steffen Pischke;
- “Elements of Statistical Learning” by Hastie, Tibshirani, Friedman (available online).
- “Computer Age Statistical Inference” by Hastie and Efron (available online);

Papers

Lecture notes will be partially based on the papers below (the list will be updated throughout the course). I do not expect you to read all or any of them (some are very technical), but I will let you know the most useful ones.

References

- A. Abadie, A. Diamond, and J. Hainmueller. Synthetic control methods for comparative case studies: Estimating the effect of california's tobacco control program. *Journal of the American statistical Association*, 105(490):493–505, 2010.
- A. Abadie, S. Athey, G. W. Imbens, and J. M. Wooldridge. Finite population causal standard errors. Technical report, National Bureau of Economic Research, 2014.
- A. Abadie, S. Athey, G. W. Imbens, and J. Wooldridge. When should you adjust standard errors for clustering? Technical report, National Bureau of Economic Research, 2017.
- R. Adao, M. Kolesár, and E. Morales. Shift-share designs: Theory and inference. Technical report, National Bureau of Economic Research, 2018.
- S. Agrawal and N. Goyal. Analysis of thompson sampling for the multi-armed bandit problem. In *Conference on learning theory*, pages 39–1. JMLR Workshop and Conference Proceedings, 2012.
- I. Andrews, J. H. Stock, and L. Sun. Weak instruments in instrumental variables regression: Theory and practice. *Annual Review of Economics*, 11:727–753, 2019.
- J. Angrist and M. Kolesár. One instrument to rule them all: The bias and coverage of just-id iv. Technical report, National Bureau of Economic Research, 2021.
- M. Arellano and S. Bond. Some tests of specification for panel data: Monte carlo evidence and an application to employment equations. *The review of economic studies*, 58(2):277–297, 1991.
- M. Arellano and S. Bonhomme. Nonlinear panel data estimation via quantile regressions, 2016.
- M. Arellano and S. Bonhomme. Quantile selection models with an application to understanding changes in wage inequality. *Econometrica*, 85(1):1–28, 2017.
- D. Arkhangelsky and G. Imbens. The role of the propensity score in fixed effect models. Technical report, National Bureau of Economic Research, 2018.
- D. Arkhangelsky, S. Athey, D. A. Hirshberg, G. W. Imbens, and S. Wager. Synthetic difference in differences. Technical report, National Bureau of Economic Research, 2019.
- T. B. Armstrong and M. Kolesár. Optimal inference in a class of regression models. *Econometrica*, 86(2):655–683, 2018.
- O. C. Ashenfelter and D. Card. Using the longitudinal structure of earnings to estimate the effect of training programs, 1984.
- S. Athey and G. W. Imbens. Identification and inference in nonlinear difference-in-differences models. *Econometrica*, 74(2):431–497, 2006.
- S. Athey, G. W. Imbens, and S. Wager. Approximate residual balancing: debiased inference of average treatment effects in high dimensions. *Journal of the Royal Statistical Society: Series B (Statistical Methodology)*, 2016.

- S. Athey, M. Bayati, N. Doudchenko, G. Imbens, and K. Khosravi. Matrix completion methods for causal panel data models. *arXiv preprint arXiv:1710.10251*, 2017.
- J. Bai. Panel data models with interactive fixed effects. *Econometrica*, 77(4):1229–1279, 2009.
- A. Belloni, V. Chernozhukov, and C. Hansen. Inference on treatment effects after selection among high-dimensional controls. *The Review of Economic Studies*, 81(2):608–650, 2014.
- M. Bertrand, E. Duflo, and S. Mullainathan. How much should we trust differences-in-differences estimates? *The Quarterly journal of economics*, 119(1):249–275, 2004.
- S. Bonhomme and E. Manresa. Grouped patterns of heterogeneity in panel data. *Econometrica*, 83(3):1147–1184, 2015.
- S. Bonhomme, T. Lamadon, and E. Manresa. Discretizing unobserved heterogeneity. *University of Chicago, Becker Friedman Institute for Economics Working Paper*, (2019-16), 2017.
- S. Bonhomme, T. Lamadon, and E. Manresa. A distributional framework for matched employer employee data. *Econometrica*, 87(3):699–739, 2019.
- S. Bonhomme, K. Holzheu, T. Lamadon, E. Manresa, M. Mogstad, and B. Setzler. How much should we trust estimates of firm effects and worker sorting? Technical report, National Bureau of Economic Research, 2020.
- K. Borusyak, P. Hull, and X. Jaravel. Quasi-experimental shift-share research designs. Technical report, National Bureau of Economic Research, 2018.
- K. Borusyak, X. Jaravel, and J. Spiess. Revisiting event study designs: Robust and efficient estimation. *arXiv preprint arXiv:2108.12419*, 2021.
- G. Chamberlain. Asymptotic efficiency in estimation with conditional moment restrictions. *Journal of Econometrics*, 34(3):305–334, 1987.
- G. Chamberlain. Efficiency bounds for semiparametric regression. *Econometrica: Journal of the Econometric Society*, pages 567–596, 1992.
- V. Chernozhukov, D. Chetverikov, M. Demirer, E. Duflo, C. Hansen, et al. Double machine learning for treatment and causal parameters. *arXiv preprint arXiv:1608.00060*, 2016.
- V. Chernozhukov, W. Newey, and J. Robins. Double/de-biased machine learning using regularized iesz representers. *arXiv preprint arXiv:1802.08667*, 2018.
- M. Cytrynbaum. Designing representative and balanced experiments by local randomization. *arXiv preprint arXiv:2111.08157*, 2021.
- M. Das, W. K. Newey, and F. Vella. Nonparametric estimation of sample selection models. *The Review of Economic Studies*, 70(1):33–58, 2003.
- C. De Chaisemartin and X. d’Haultfoeuille. Two-way fixed effects estimators with heterogeneous treatment effects. *American Economic Review*, 110(9):2964–96, 2020.

- D. Ghanem, P. H. Sant'Anna, and K. Wüthrich. Selection and parallel trends. *arXiv preprint arXiv:2203.09001*, 2022.
- A. Goodman-Bacon. Difference-in-differences with variation in treatment timing. *Journal of Econometrics*, 225(2):254–277, 2021.
- B. Graham and A. de Paula. *The Econometric Analysis of Network Data*. Academic Press, 2019.
- L. P. Hansen. Large sample properties of generalized method of moments estimators. *Econometrica: Journal of the Econometric Society*, pages 1029–1054, 1982.
- D. A. Hirshberg and S. Wager. Balancing out regression error: efficient treatment effect estimation without smooth propensities. *arXiv preprint arXiv:1712.00038*, 2017.
- G. Imbens and S. Wager. Optimized regression discontinuity designs. *arXiv preprint arXiv:1705.01677*, 2017.
- M. Kasy and A. Sautmann. Adaptive treatment assignment in experiments for policy choice. *Econometrica*, 89(1):113–132, 2021.
- D. S. Lee, J. McCrary, M. J. Moreira, and J. R. Porter. Valid t-ratio inference for iv. Technical report, National Bureau of Economic Research, 2021.
- C. F. Manski. Identification of endogenous social effects: The reflection problem. *The review of economic studies*, 60(3):531–542, 1993.
- R. Meager. Aggregating distributional treatment effects: A bayesian hierarchical analysis of the microcredit literature. *American Economic Review*, 112(6):1818–47, 2022.
- H. R. Moon and M. Weidner. Linear regression for panel with unknown number of factors as interactive fixed effects. *Econometrica*, 83(4):1543–1579, 2015.
- D. B. Rubin. Estimating causal effects of treatments in randomized and nonrandomized studies. *Journal of educational Psychology*, 66(5):688, 1974.
- P. H. Sant'Anna and J. Zhao. Doubly robust difference-in-differences estimators. *Journal of Econometrics*, 219(1):101–122, 2020.
- S. M. Schennach. Recent advances in the measurement error literature. *Annual Review of Economics*, 8:341–377, 2016.
- L. Sun and S. Abraham. Estimating dynamic treatment effects in event studies with heterogeneous treatment effects. *Journal of Econometrics*, 225(2):175–199, 2021.
- S. Wager, W. Du, J. Taylor, and R. J. Tibshirani. High-dimensional regression adjustments in randomized experiments. *Proceedings of the National Academy of Sciences*, 113(45):12673–12678, 2016.
- J. M. Wooldridge. Two-way fixed effects, the two-way mundlak regression, and difference-in-differences estimators. *Available at SSRN 3906345*, 2021.