

## **Modern Sampling Methods: Design and Inference**

**Syllabus** (last updated: January 11, 2022)

### **Instructors:**

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### **Course Description:**

The way that data are collected can have important implications for their subsequent use in model estimation and evaluation, parameter inference, and policy choice. Complex experimental schemes are gaining increasing interest in economics and many other fields. In this course we examine the interplay between data design, and statistical inference and decisions. We will examine classic sampling and experimental designs, but also alternatives such as adaptive randomization designs and multi-armed bandits which raise the possibility of improving inference and decisions at lower cost. Some key themes we will examine include: the connection between design and identification; statistical efficiency considerations; and alternative objectives for data analysis beyond point estimation and classical statistical inference.

### **Outline of Topics:**

1. Introduction: Examples and Key Issues; Random Samples for Population Inference
2. Randomized Experiments
3. Publication Bias and Preanalysis Plans
4. Treatment and Policy Choice
5. Multi-Wave Experiments
6. Covariate-Adaptive Randomization
7. Bandit Algorithms and Response-Adaptive Experiments
8. Applications of Bandits and Adaptive Designs
9. Statistical Inference with Adaptively Generated Data
10. Window Choice in Time Series

## Topics and Readings:

Especially useful readings, recommended to read before the lecture, are marked with a “\*”. Additional references and readings will be added as this syllabus is revised.

### 1. Examples and Background

- Athey, S., and Haile, P. A., (2002), “Identification of Standard Auction Models,” *Econometrica* 70 (6): 2107-2140.
- Karlan, D., and Wood, D. H., (2017), “The Effect of Effectiveness: Donor Response to Aid Effectiveness in a Direct Mail Fundraising Experiment,” *Journal of Behavioral and Experimental Economics* 66: 1-8.
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- Marschak, J., (1953) “Economic Measurements for Policy and Predictions,” in *Studies in Econometric Method*, Cowles Foundations Monograph No. 14, Wiley.

### 2. Randomized Experiments

- Abadie, A., and Cattaneo, M., (2018), “Econometric Methods for Program Evaluation,” *Annual Review of Economics* 10: 465-503.
- Athey, S., and Imbens, G. W., (2017), “The Econometrics of Randomized Experiments,” in *Handbook of Economic Field Experiments*, Volume 1, Elsevier.
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### 3. Publication Bias and Preanalysis Plans

- Andrews, I., and Kasy, M., (2019), “Identification of and Correction for Publication Bias,” *American Economic Review* 109(8), 2766-2794.
- Camerer, C., et al, (2016), “Evaluating Replicability of Laboratory Experiments in Economics,” *Science* 351(6280): 1433-1436.
- Casey, K., Glennerster, R., and Miguel, E. (2012), “Reshaping Institutions: Evidence on Aid Impacts Using a Preanalysis Plan,” *Quarterly Journal of Economics*, 1755-1812.
- Christensen, G., and Miguel, E., (2018) “Transparency, Reproducibility, and the Credibility of Economics Research,” *Journal of Economic Literature* 56(3): 920-980.
- Coffman, L. and Niederle, M. (2015), “Pre-Analysis Plans Have Limited Upside, Especially Where Replications Are Feasible,” *Journal of Economic Perspectives*, 29(3): 81-98.

- \* Duflo, E., Banerjee, A., Finkelstein, A., Katz, L. F., Olken, B. A., and Sautmann, A., (2020), “In Praise of Moderation: Suggestions for the Scope and Use of Pre-Analysis Plans for RCTs,” NBER Working Paper No. 26993.
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- Ioannidis, J. P. A., (2009), “Why Most Discovered True Associations are Inflated,” *Epidemiology* 19(5): 640-648.
- Leamer, E. (1974), “False Models and Post-Data Model Construction,” *Journal of the American Statistical Association*, 69(345), 122-131.
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- McCrary, J., Christensen, G., and Fanelli, D., (2016), “Conservative Tests under Satisficing Models of Publication Bias,” *PLoS ONE* 11(2): e0149590.
- Miguel, E. (2021), “Evidence on Research Transparency in Economics,” *Journal of Economic Perspectives*, 35(3), 193-214.
- Olken, B. (2015), “Promises and Perils of Pre-Analysis Plans,” *Journal of Economic Perspectives*, 29(3), 61-80.
- Open Science Collaboration (2015), “Estimating the Reproducibility of Psychological Science,” *Science*, 349(6251), aac4716.
- Spiess, J., (2018), “Optimal Estimation when Researcher and Social Preferences are Misaligned,” working paper.  
<https://gsb-faculty.stanford.edu/jann-spiess/files/2021/01/alignedestimation.pdf>
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#### 4. Treatment and Policy Choice

- Athey, S., and Wager, S., (2021), “Policy Learning with Observational Data,” *Econometrica* 89(1): 133-161.
- Chamberlain, G., (2011), “Bayesian Aspects of Treatment Choice,” in *The Oxford Handbook of Bayesian Econometrics*, ed. by J. Geweke, G. Koop, and H. van Dijk, Oxford University Press.
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- Hirano, K., and Porter, J. R., (2009), “Asymptotics for Statistical Treatment Rules,” *Econometrica* 77, 1683-1701.
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- Stoye, J., (2009), “Minimax Regret Treatment Choice with Finite Samples,” *Journal of Econometrics* 151, 70-81.

## 5. Multi-Wave Experiments

- \* Hahn, J., Hirano, K., and Karlan, D., (2011), “Adaptive Experimental Design Using the Propensity Score,” *Journal of Business and Economic Statistics* 29(1): 96-108.
- Jennison, C., and Turnbull, B. W., (2000), *Group Sequential Methods with Applications to Clinical Trials*, CRC Press.
- Tabord-Meehan, M., (2021), “Stratification Trees for Adaptive Randomization in Randomized Controlled Trials,” working paper.

## 6. Covariate-Adaptive Randomization

- Bai, Y., Romano, J., and Shaikh, S. (2021), “Inference in Experiments with Matched Pairs,” *Journal of the American Statistical Association*.
- Bruhn, M., and McKenzie, D. (2000), “In Pursuit of Balance: Randomization in Practice in Development Field Experiments” *American Economic Journal: Applied Economics*, 1(4), 200-232.
- \* Bugni, F. A., Canay, I. A., and Shaikh, A. M., (2018), “Inference under Covariate Adaptive Randomization,” *Journal of the American Statistical Association* 113 (524), 1784-1796.
- Bugni, F.A., Canay, I. A., Shaikh, A. M., (2019), “Inference under Covariate Adaptive Randomization with Multiple Treatments,” *Quantitative Economics* 10(4) 1747-1785.
- Duflo, E., Glennerster, R., and Kremer, M. (2007), “Using Randomization in Development Economics Research: A Toolkit,” *Handbook of Development Economics*, 4, 3895-3962.
- Markaryan, T., and Rosenberger, W., (2010), “Exact Properties of Efron’s Biased Coin Randomization Procedure,” *The Annals of Statistics*, 38, 1546-1567.
- Zelen, M. (1974), “The Randomization and Stratification of Patients to Clinic Trials,” *Journal of Chronic Diseases*, 27, 365-375.

## 7-8. Response-Adaptive Experiments and Multi-Armed Bandits

- Adusumilli, K., (2021), “Risk and Optimal Policies in Bandit Experiments.”  
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- Rothschild, M., (1974), “A Two-Armed Bandit Theory of Market Pricing,” *Journal of Economic Theory* 9: 185-202.
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- Thompson, W. R., (1933), “On the Likelihood that One Unknown Probability Exceeds Another in View of the Evidence of Two Samples,” *Biometrika* 25: 285-294.

## 9. Statistical Inference with Data from Bandits

- Deshpande, Y., Mackey, L., Syrgkantis, V., and Taddy, M., (2020), “Accurate Inference for Adaptive Linear Models,” working paper. <https://arxiv.org/abs/1712.06695>
- Dimakopoulou, M., Zhou, Z., Athey, S., and Imbens, G., “Estimation Considerations in Contextual Bandits,” working paper. <https://arxiv.org/abs/1711.07077>
- Fang, Z., and Santos, A., (2019), “Inference on Directionally Differentiable Functions,” *Review of Economic Studies* 86, 377-412.
- Hadad, V., Hirshberg, D. A., Zhan, R., Wager, S., and Athey, S., (2021), “Confidence Intervals for Policy Evaluation in Adaptive Experiments,” *Proceedings of the National Academy of Sciences* vol. 118 no. 15. <https://doi.org/10.1073/pnas.2014602118>
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#### 10. Window Choice in Time Series (if time permits)

- Andrews, D. W. K., (1993), “Tests for Parameter Instability and Structural Change with Unknown Change Point,” *Econometrica* 61(4): 821-856.
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#### T. Additional Technical Background

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- van der Vaart, A. W., (1998), *Asymptotic Statistics*, Cambridge University Press.
- Vershynin, R., (2018), *High-Dimensional Probability: An Introduction with Applications in Data Science*, Cambridge University Press.
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