

Design and analysis of experiments

HU | Spring 2026 (Winter term 25-26)

Instructor	Teaching assistant
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Overview: The course addresses the design and analysis of experiments. It is a hands-on course in which theoretical results are introduced through lecture, demonstrated in practice, and worked on in groups. In-class discussions focus on major works, features of designs, and experimental strategies. Topics include randomization schemes, analysis strategies, design evaluation, and workflows.

1 Class resources

- All resources: <https://macartan.github.io/ci>
- This syllabus: https://macartan.github.io/ci/syllabus_experiments_2026.pdf
- Slides: https://macartan.github.io/ci/slides_experiments_2026.html
- Student survey
- Git repo <https://github.com/macartan/ci>: This repo is for both my causal inference class and this experiments class.

2 Times and locations

Meetings will be held at the WZB

Date	Time	Location	Topic
9 Jan 2026	9-13	WZB A 305	Getting started
16 Jan 2026	9-13	WZB A 305	Doing experiments
30 Jan 2026	9-13	WZB A 305	Causality and causal inquiries
6 Feb 2026	9-13	WZB A 305	Answer strategies
20 Feb 2026	9-13	WZB A 305	Data strategies and design evaluation
27 Feb 2026	9-13	WZB B 001	Topics: downstream experimentation; spillover experiments; patient preference trials; survey experiments

The sessions will be structured roughly as follows.

- 9:00 - 11:00: lecture or discussions of readings
- 11:00 - 11:15: break
- 11:15 - 13:00: discussions, team presentations and in-class exercises

Instructor office hours: Fridays, 13:00 - 15:00, in person

Teaching assistant office hours: Wednesdays, 13:00 - 14:00 on [Zoom](#) (excluding 25 Feb)

3 Expectations

3.1 Required (6 ECTS)

- Do the readings and take part in discussion (30%)
- Take part in field experiment I (30%)
- Design an experiment as a candidate for survey experiment II (40%)

3.2 Optional (+4 ECTS)

- Prepare a research design or short paper, perhaps building on existing work. Typically this contains:
 - a problem statement
 - a description of an experimental strategy to address the problem
 - empirical (ideally) or simulation based results (ok)
 - a discussion of implications
 - A passing paper will illustrate identify a causal effect of substantive interest credibly; a good paper will innovate on method or topic.
- Deadline: August 31, 2026

4 Sessions and readings

The course combines lectures, discussion, and activities.

All readings are linked from this document.

I will draw material especially from Blair, Coppock, and Humphreys (2023): [Research Design in the Social Sciences](#)

Recommended but not open access: [Gerber and Green \(2012\)](#) with supplementary material available here: <https://isps.yale.edu/FEDAI>

Readings for each session are given below.

4.1 Intro (9 January)

Aims:

- Introduce the class
- Practice running experiments by hand
- Introduce and discuss field [experiment I](#)
- To prepare for next session: proposed adaptation of field experiment 1

Course outline, tools

4.2 Experimentation (16 January)

Aims:

- Review lifecycle of an experiment (discussion)
- Discuss ethical challenges to experimentation
- Discuss a set of strong experiments
- Develop group field experiment 1, decide on strategy, and start ethics application

To do before class: 1. the readings; 2. develop specific proposals in small groups for field experiment 1 (4 slides), completing individually or collectively a copy of the [EGAP research design form](#)

Required reading:

on ethics:

- [Johnson: Protecting the Community: Lessons from the Montana Flyer Project](#)
- [APSA guidelines](#)

some cases:

- Example 1: Gerber, Green, and Larimer (2008)
- Example 2: Lowe (2021)

- The Hijab penalty and the Appendix

Optional pieces on experimentation:

- Humphreys (2015) on ethics in experiments
- Design lifecycle Blair, Coppock, and Humphreys (2023) p 319 - 358
- DD, Ch 2 introduces ideas at a high level
- Alvarez, Key, and Núñez (2018)
- Humphreys, De la Sierra, and Van der Windt (2013) on fishing and registration

Optional substantive pieces:

- Kurzban, Tooby, and Cosmides (2001)
- Haas et al. (2025)
- Mousa (2020)
- Dietrich and Sands (2023)

4.3 Causality (30 January)

Aims:

- Understand the potential outcomes framework
- Understand identification challenges and the role of randomization
- Understand varieties of estimands

To do before class: 1. the readings; 2. collectively complete a pre-analysis plan and [ethics application](#) for Experiment 1

Required:

- Holland is a beautiful classic reading on the fundamental problem of causal inference
- DD, Ch 7

Optional:

- II Ch 2, goes over both potential outcomes and DAGs
- Keele (2015) gives a good high level overview of many key ideas in this course
- Imai, King, and Stuart (2008)
- Hernán and Robins (2024) (Section 3.1, 7.2, 8.4, 10.1)
- Hernán and Robins (2024) Ch 6
- II Ch 4
- Lundberg, Johnson, and Stewart (2021)

4.4 Answer strategies (6 February)

Aims:

- Understand alternative strategies to estimating treatment effects
- Understand design based strategies for estimating standard errors and p -values
- Understand confidence intervals
- Understand the role of covariates in experimental design

To do before class: 1. Collective presentation on the implementation and findings from Experiment 1

Optional:

- Freedman (2008) helps make connections between design based inference and regression
- Bind and Rubin (2020) on RI
- Lin (2012) relieves some worries you might have after reading Freedman (2008)
- Chang, Middleton, and Aronow (2024) follows up on Lin (2012).

4.5 Experimental Design and Evaluation (20 February)

Aims:

- Understand alternative assignment strategies
- Understand statistical power
- Understand how to assess alternative diagnosands

To do before class: 1. The readings 2. Share initial design for for survey based experiment.

Required:

- [DD, Ch 7](#)
- [DD, Ch 8](#) on data strategies
- [DD, Ch 9](#) on answer strategies
- [DD, Ch 13](#) gives more practical guidance on getting started

4.6 Topics (27 February)

To do before class: 1. The readings 2. Round-robin report on design performance. Teams receive a class-mate's design, declare a model that produces possible data from the design (without consulting the authors design declaration, if any) and applies the original authors answer strategy, reporting performance.

Required reading:

4.6.1 survey experiments

- [Hainmueller, Hopkins, and Yamamoto \(2014\)](#)

4.6.2 mediation experiments

- [Imai, Keele, and Tingley \(2010\)](#) on mediation

4.6.3 Lab experiments

- [Roth \(1986; Palfrey 2009\)](#)

4.6.4 Experiments on networks

- [Baird et al. \(2014\)](#) [Optional: [Sinclair, McConnell, and Green \(2012\)](#); [Aronow and Samii \(2017\)](#)]

4.6.5 downstream experimentation

- [Coppock and Green \(2016\)](#)

Not required:

4.6.6 patient preference trials

- [Benedictis-Kessner et al. \(2019\)](#)

5 Prerequisites and tools

You should already have background in statistics up to the point of feeling comfortable with regression.

In addition you should know some R. Really, the more you can invest on getting on top of R before the class the better.

Before the first class please make sure your R is up-to-date and that you are working in R studio. Then make sure you have the following packages installed.

```
pacman::p_load(  
  rstan,  
  dagitty,  
  DeclareDesign,  
  CausalQueries,  
  tidyverse,  
  knitr  
)
```

Resources:

- Resources for learning R: <http://www.r-bloggers.com/how-to-learn-r-2/>
- Hadley Wickham's [R for Data Science](#)
- [Grant McDermott on data science](#)

5.1 Writing with qmd

Please plan to do drafting in Quarto. This is a simple markup language that lets you integrate writing and coding. This document is written in Quarto and the slides will be also.

The key thing is that you can insert code chunks like this.

```
# Define a random number  
x <- rnorm(1)
```

Code like this is run as the document compiles, and results can be accessed as needed, like this: we just sampled the random number $x = -1.5230672$.

- See [Wickham, Çetinkaya-Rundel, and Gromelund \(2023\)](#) on [Workflow](#) and [Quarto](#)

I recommend using Rstudio as an editor. More information here: <https://quarto.org/docs/get-started/hello/rstudio.html>

5.2 AI

AI can be a great aid to help understand this material and also to develop code. However it should only be used as an aid: it makes many many mistakes; you remain responsible for your code and more importantly for your comprehension.

References

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