Experimental methodology¹

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¹ Some sources: James Tremewan's course thought at MUNI in May 2022, Handbook of Research Methods and Applications in Experimental Economics and Handbook of Experimental Economic Methodology



Some kind of experimental design necessary for causal inference.

Intro • O O

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Intro

You can look for experimental designs in observational data (Štěpán Mikula).

Some kind of experimental design necessary for causal inference.

You can look for experimental designs in observational data (Štěpán Mikula).

Or you can produce your own experimental data – popular in many fields:

- Gender and labor market huge literature in the lab and field
- Environmental economics common lab and field studies
- Management/adaptation common experiments: (1, 2)
- DSGE lab studies: (1, 2)

Intro OO

Overview

Economic experiments (as compared to surveys) have two main features:

randomization

Intro O • O

• incentivization (exceptions: survey experiment, some lab studies)

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- randomization
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Running **experiments** (as opposed to using experiments in the data):

Design

Intro

- informed by theory
- to be able to analyze data efficiently (power, avoid assumptions)
- Admin
 - Pre-registration
 - Ethics approval
- Data collection (depends on the type of the experiment)
- Data analysis

MU Experimental Economics Laboratory (MUEEL)

We offer research infrastructure:

Intro

- experimental laboratory hroot, otree programmers, support (lab manager, payments)
- access to software and equipment (Qualtrics, Veyon, Air Quality, ...)
- experience with lab/field/survey experiments and agent-based models
- network of experts local (CERGE-EI, VŠE, WU) and broader

Questionnaires in the lab

Scientific disciplines are as precise as the measurements of basic concepts.

 $\label{eq:main_example} \mbox{Main example in economics: } \mbox{{\bf preferences}} \mbox{ (consistency, stability, } \mbox{{\bf values}/distributions)}$

- risk preferences
- time preferences: discount rate, present bias
- social preferences: e.g. altruism, trust, reciprocity

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Main example in economics: **preferences** (consistency, stability, values/distributions)

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Global preference survey:



Theory and experiments (1/2)

Economic theory provides structure for the examination of how people behave in economic situations.

Experiments not motivated by theory may lead to duplication of effort.

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Experiments not motivated by theory may lead to duplication of effort.

Experiments are best at:

- testing theories make comparative-static predictions (about direction and not about size/parameters of the model)
- testing assumptions field work is based on

Theory and experiments (2/2)

Economic theory

- is strong and works: e.g. market experiments
- is strong and does not work: e.g. expected utility theory
- is weak: e.g. ultimatum game
 - selfish preferences are not theory
 - experience

Theory and experiments (2/2)

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Adjustments to make theory more useful, e.g.

- other-regarding preferences (Fehr & Schmidt, 1999; Bolton & Ockenfels, 2000)
- quantal response equilibrium (McKelvey & Palfrey, 1995)
- one-one shot interactions level-k theory (Stahl & Wilson, 1995)

Design choices: treatments

Random assignment to treatment \to causal inference Treatments should differ only in one thing!

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Sometimes experiments involve multiple factors, e.g. communication (with/without) and stake-size (low/high) in a dictator game.

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Rule of thumb: Use the minimum number of treatments possible to test your hypothesis!

- Why?
 - More treatments problem of multiple testing.
 - Fewer treatments = more observations per treatment
- Avoid
 - intermediate treatment unless interested in non-linearities.
 - full factorial design unless interested in the interaction term.

Design choices: within/between subject design

Within-subject design: Each subject in more than one treatments

- Advantages:
 - Effect on individual
 - Greater power
- Disadvantages:
 - Subjects may want to choose the same to be consistent
 - Experimental demand effect = changes in behavior due to cues about what constitutes appropriate behavior
 - Potential order effects = order of tasks affects decisions

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Between-subject design: Each subject only in one treatment

- Advantages solves problems of demand effects and consistency
- Disadvantages power and subjective judgment (Birnbaum, 1999)

Design choices: incentives

- Transparent (minimum payment in lab experiments)
- Incentive compatible
- Pay all vs. random incentive system (RIS) (Clot et al., 2018)

Other design choices

- One-shot vs. repeated play of a game/situation
 - advantages: allow learning, interested in learning effects
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- Direct response vs. strategy method in sequential games:
 - direct-response: P1 makes a decision, P2 is informed of the decision then makes their own decision.
 - strategy method: P1 makes a decision, and simultaneously P2 decides for all possible choices of P1. The strategies of the two players are combined and the outcome determined.

Power calculations

Power is the probability of finding an effect that is really there.

Underpowered studies (too few subjects):

- less likely to detect real treatment effects.
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Useful exercise. **Problem**: you need to know the effect size.

Especially useful to get some idea about number of subjects when interested in more complicated effects (e.g. interaction effect)

Replication crisis

- Psychology: 36/100 studies replicated, effect sizes 50% of original effect (Open Science Collaboration, Science, 2015)
- Experimental economics: 11/18 (61%) studies in AER or QJE replicated, effect sizes 66% of original effect (Camerer et al, *Science*, 2016)
- Social sciences: 13/21 (62%) studies in Nature or Science, effect sizes 50% of original effect (Camerer et al, *Science*, 2018)

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Reasons:

- Publication bias
- The way tests work (sample size, α , reasonable hypotheses)
- P-hacking
- Testing multiple hypotheses
- ..

P-hacking

- Selectively removing outliers
- Running multiple tests reporting lowest p-value
- Running multiple regressions specifications and reporting only the one that works.
- Including pilot data to the analysis
- Stopping gathering data as soon as p < 0.05

Multiple testing

Problem when using multiple simultaneous tests (common in experimental economics – multiple outcomes/subgroups/treatments).

Family-wise error rate (FWER): Probability of at least one false rejection for k statistically independent tests at α level:

$$\mathsf{FWER} = 1 - (1 - \alpha)^k$$

For $\alpha = 0.05$ and k = 2, 3, 4, ..., FWER = 0.098, 0.143, 0.185,...

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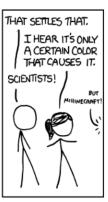
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There are methods used for adjusting for multiple hypotheses:

- Bonferroni
 - Holm-Bonferonni
 - List (2019)
 - •







WE FOUND NO LINK BETWEEN PURPLE JELLY BEANS AND ACNE (P > 0.05)



WE FOUND NO LINK BETWEEN BROWN JELLY BEANS AND ACNE (P > 0.05),



WE FOUND NO LINK BETWEEN PINK JELLY BEANS AND ACNE (P > 0.05)



WE FOUND NO LINK BETWEEN BLUE JELLY BEANS AND ACNE (P > 0.05).



WE FOUND NO LINK BETWEEN TEAL JELLY BEANS AND ACNE (P > 0.05).



WE FOUND NO LINK BETWEEN SALMON JELLY BEANS AND ACNE (P>0.05).



WE FOUND NO LINK BETWEEN RED JELLY BEANS AND ACNE (P > 0.05).



WE FOUND NO LINK BETWEEN TURQUOISE JELLY BEANS AND ACNE (P>0.05).



WE FOUND NO LINK BETWEEN MAGENTA JELLY BEANS AND ACNE (P > 0.05).



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WE FOUND NO LINK BETWEEN TAN JELLY BEANS AND ACNE (P > 0.05).



WE FOUND NO LINK BETWEEN CYAN JELLY BEANS AND ACNE (P>0.05).



WE FOUND A LINK BETWEEN GREEN JELLY BEANS AND ACNE (P < 0.05).



WE FOUND NO LINK BETWEEN MAUVE JELLY BEANS AND ACNE (P > 0.05).



WE FOUND NO LINK BETWEEN BEIGE JELLY BEANS AND ACNE (P>0.05).



WE FOUND NO LINK BETWEEN LILAC JELLY BEANS AND ACNE (P > 0.05)



WE FOUND NO LINK BETWEEN BLACK JELLY BEANS AND ACNE (P > 0.05).

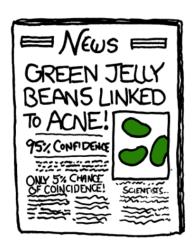


WE FOUND NO LINK BETWEEN PEACH JELLY BEANS AND ACNE (P > 0.05).



WE FOUND NO LINK BETWEEN ORANGE JELLY BEANS AND ACNE (P > 0.05).





Should we adjust for multiple hypotheses?

- Yes, if the null hypothesis is that jelly beans cause acne. Why test it separately by color, then?
- Perhaps not, if you want to test the link between colour and acne (if you have a theory why each tested color might cause acne)
- Always important to report the tests to all hypotheses (and to pre-register)

Pre-registration

You register your desing prior to data collection (AEA RCT Registry, AsPredicted).

What do you register?

- number of observations problem with stopping
- hypotheses multiple comparisons

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When important/necessary?

- design leads to many potential hypotheses (problematic for exploratory studies)
- expected by journals field experiment, now increasingly common in lab experiments

Ethics approval

- Design (social importance)
- Procedures
- Hiring
- Ethics questions

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Necessary/more important when

- expected by the journal (in some countries obligatory whenever research on people)
- we collect personal information
- subjects can be harmed (e.g. we induce stress or strong emotions)/use of vulnerable population
- we take biological material (measuring hormone levels)
- we use deception

Deception

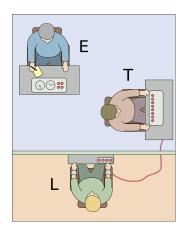
- widespread in (social) psychology, factually banned in laboratory experimental economics
- reputational rather than ethical concerns (ex ante informed consent, ex post debriefing)
- acts of commission (e.g. lying to subjects) vs. acts of omission (not telling subjects things)
- evidence (Ortmann and Hertwig, 2002) shows that economists should avoid acts of commission and acts of omission that violate default assumptions
- and what about field experiments? difficult to get informed consent ethically problematic (audit studies)

 Design
 Admin
 Data collection
 Data analysis
 External validity
 Examples
 Reference

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Examples: Milgram experiment

Milgram, S. (1963). Behavioral study of obedience. The Journal of abnormal and social psychology



Public Announcement

WE WILL PAY YOU \$4.00 FOR ONE HOUR OF YOUR TIME

Persons Needed for a Study of Memory

"We will pay fire hundred New Haven men to help us complete a scientific study of memory and learning. The study is being done at Yale University.

"Each person who participates will be paid \$4.00 (plus S0c carlars) for approximately 1 hour's time. We need you for only one hour: there are no fairther obligations. You may choose the time you would like to come (evenings, weekdars, or weekends).

*No special training, education, or experience is needed. We want: Factory workers Businessmen Construction workers City employees Clerks Salespeople Laborers Professional people White-collar workers

Barbers Telephone workers Others
All persons must be between the ages of 20 and 50. High shool and college
students cannot be used.
"If you meet these qualifications, fill out the coupon below and mail it
most to Professor Stanley Migram, Department of Psychology, Yale University,

New Haven. You will be notified later of the specific time and place of the study. We reserve the right to decline any application.

You will be raid \$4.00 (blus 50c cattage) as soon as you arrive at the

	TO: PROF: STANLEY MILGRAM, DEPARTMENT OF PSYCHOLOGY YALE UNIVERSITY, NEW HAVEN, CONN. I want to take part it this study of memory and learning. I am between the ages of 20 an SO. I will be paid \$4.00 (plus SOc cafface) fil participate.
	NAME (Please Print).
	ADDRESS
	TELEPHONE NO Best time to call you
	AGEOCCUPATIONSEX
	WEEKDAYS EVENINGS WEEKENDS

Examples: Color of a free ride

Mujcic, R., Frijters, P. (2021). The colour of a free ride. *The Economic Journal*

OPINION Bus driver research by UQ's Frijters and Mujcic deserves publication



04 MAR 2015

Controversy/opinion:

Research by University of Queensland economists Paul Frijters and Redzo Mujcic revealing discrimination by Brisbane bus drivers deserves to be published, writes Andreas Ortmann.



Types of experiments

- laboratory experiments
- field experiments (experiment run outside of laboratory)
 - artefactual (lab-in-the-field) non-standard subjects (Henrich et al. 2001)
 - framed the same as an artefactual, except that it incorporates important elements of the context of the naturally-occurring environment
 - natural natural environment and subjects do not know that they are participants in an experiment
 - strange category: survey experiments often population-based, typically without incentives

Laboratory experiments

Subject pool management (hroot, Orsee, Sona, Prolific, mTurk, ...)

- ethics (clear rules, ability to stop any time, anonymity)
- motivation (monetary incentives, regular invitations to experiments, good/fair experience)

Experimental laboratory (or online environment)

- privacy during choices (dividers)
- privacy during payments
- experimental environment (zTree, oTree, Qualtrics, ...)

Field experiments

- useful when the "right" population
- natural experiments are more valuable, but ethical concerns
- cooperation with authorities often needed

Hypotheses and p-values

Tests:

- null (H₀) vs. alternative hypothesis (H₁)
- one-sided vs. two-sided test
 - ullet strong theoretical reasons o one-sided
 - competing theories (direction is unclear) → two-sided

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Interpret p-values carefully:

- It is not the probability the null hypothesis is true
- Never "accept the null": p > 0.1 is not evidence the null is true
- A small p-value does not mean an effect is important

Types of tests

Treatment tests:

- Non-parametric tests do not assume that data comes from a particular probability distribution, e.g. normal distribution.
 - important because experimental data often has small sample size and is non-normally distributed
- At the cost of stronger (typically untestable) assumptions about true distribution of data, parametric methods allow us to do more interesting stuff, e.g.:
 - regression analysis (controls, non-independent data)
 - treating with heterogenous types and zeros
 - structural models (estimating utility functions)

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 - regression analysis (controls, non-independent data)
 - treating with heterogenous types and zeros
 - structural models (estimating utility functions)
- Exact tests calculate the p-value exactly (true asymptotically)
- Non-exact tests are OK for a big-enough sample (but "big enough" depends on the true distribution of the data, which is unknown)

Overview of statistical methods used²

- Treatment tests
 - looking for least assumptions (non-parametric) and most power
 - depends on type of data (binary/multi-valued) and comparison (within/between subject)

²Moffatt, P. (2020). Experimetrics: Econometrics for experimental economics. Bloomsbury Publishing.

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 - accounting for dependence between observations

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- Structural modelling = estimating parameters of a utility function

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What is external validity

External validity (ecological validity, generalizability) – validity of applying the conclusions of a scientific study outside the context of that study (across other situations, people, stimuli, and times).

Internal validity – reflects how well the study is conducted and hence to what extent do the observed results describe in the population we are studying.

Levitt and List (2007)

Levitt and List (2007) – behavior in the lab (when measuring **social preferences**) is influenced by factors other than monetary incentives:

- the presence of moral and ethical considerations
- the nature and extent of scrutiny of one's actions by others
- the context in which the decision is embedded
- self-selection of the individuals making the decisions
- the **stakes** of the game

Camerer's reply to Levitt and List

Camerer (2011) comes with three arguments against the criticism:

- The goal of experimental economics is to establish a general theory linking economic factors to behaviour. Generalizability from the lab to the field is not a primary concern in a typical experiment.
- The factors listed by Levitt and List are not essential for all lab experiments (except for obtrusiveness, because of human subjects protection) and there is little evidence that typical lab features undermine generalizability.
- Economics experiments designed to test lab-field generalizability show that laboratory findings could be generalized to comparable field settings.

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Comparing lab and field: Laboratory experiments are more easily replicated whereas field experiments are less obtrusive.

Lab experiments focus on qualitative effect

Kessler and Vesterlund (2015) argue that

- the debate concentrates around a straw-man version of external validity that quantitative results are externally valid
- for most laboratory studies it is only relevant to ask whether the qualitative results (= direction or the sign of the estimated effect) are externally valid
- laboratory studies are conducted to identify general principles of behavior and therefore promise to generalize

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Different methodologies are not in competition. They are complementary.

Case: Are bankers cheaters? (1/6)

Cohn et al. (Nature, 2014) run a lab-in-the-field experiment with 128 bankers

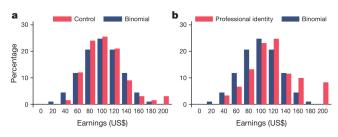
- Priming professional identity with seven questions
 - T: about their professional background (e.g. "At which bank are you presently employed?" or "What is your function at this bank?")
 - C: that were unrelated tot heir profession (e.g. "How many hours per week do you watch television on average?")
- ullet Playing a **cheating game** (10 coin throws earning \$20 or 0 each) + paid only if higher than earnings of a randomly drawn person from a pilot
- Manipulation check converting word fragments into meaningful words (__oker = broker or smoker)

Design Admin Data collection Data analysis External validity Examples Reference

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Case: Are bankers cheaters? (2/6)

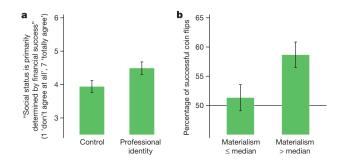
Several channels though which priming may have worked:

- Competitive behaviour in banking is intrinsically desirable (no evidence)
- Salience of competitive incentive schemes (no difference between core and support units)
- Norm obedience what I should do/what others do (no difference in beliefs what others will do)
- Evidence on materialistic values below

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Case: Are bankers cheaters? (3/6)

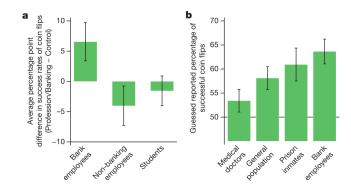
Other tests and robustness:

- other industries
- beliefs

Case: Are bankers cheaters? (3/6)

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- beliefs



Case: Are bankers cheaters? (4/6)

Criticism:

- Vranka and Houdek (FrontPsy, 2015) criticize interpretation:
 - stereotype threat instead of social norms
 - what is treatment? control group primed by leisure activities call for multiple control groups
 - priming bankers with money instead of professional identity
 - why not replicate when they had bankers giving expectations?
- Hupe (2018) criticize the analysis (not published)
 - propose a different H0: cheating the same in all six groups
 - criticizes the use of MW test (ties)

Case: Are bankers cheaters? (5/6)

Replications:

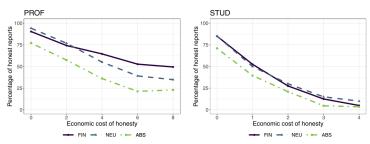
- Rahnwan et al. (*Nature*, 2019) do not replicate the result (n = 768)
- Cohn et al. (Nature, 2019) why the replication by Rahnwan not valid:
 - Media attention → selection (2 out of 27 banks)
 - Disclosed purpose at the beginning (demand effect)
 - Manipulation check did not work in Middle East
 - Basic retail services vs. investment/private banking
 - Study tests whether "prevailing scandals involving fraud and unethical behaviour in the banking industry were partly the result of a problematic business culture, rather than, for example, the employment of dishonest people in the banking industry." instead of "the prevailing business culture in the banking industry weakens and undermines the honesty norm"
- Huber and Huber (JEBO, 2020)

Case: Are bankers cheaters? (6/6)

- Huber and Huber (*JEBO*, 2020):
 - Lying game (T: 31; F: 37) in which individual lying is observed
 - Framing:
 - Abstract: Imagine there are two possible states of nature and one of your tasks is to report the current state.
 - Neutral: Imagine you are a security clerk at a museum and one of your tasks is to inform the manager each week about the average number of visitors in the preceding week.
 - Financial: Imagine you are the Chief Executive Officer (CEO) of a publicly listed company and one of your tasks is to inform shareholders each quarter about the course of business and the earnings per share.

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Some more examples

A good strategy is to combine experiments with observational data:

- Bursztyn, Leonardo, Thomas Fujiwara, and Amanda Pallais. "'Acting Wife': Marriage Market Incentives and Labor Market Investments." American Economic Review 107.11 (2017): 3288-3319.
- Kuziemko, I., Buell, R. W., Reich, T., Norton, M. I. (2014). "Last-place aversion": Evidence and redistributive implications. The Quarterly Journal of Economics, 129(1), 105-149.

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Some Czech examples:

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