

# Ethics of Electoral Experiments: Design-Based Recommendations

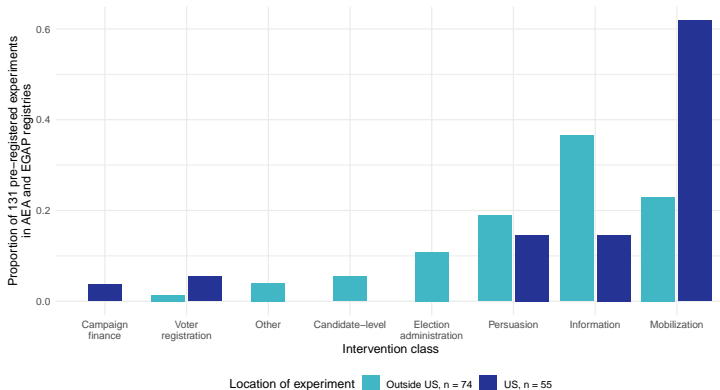
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# Electoral experiments as a tool

- Experiments on **real elections** represent a popular tool in studies of elections, political behavior, and political accountability.
- Developments in past  $\approx 10$  years:
  - Modal experiment is now **outside US** → wider variety of institutional contexts.
  - Some experiments conducted at **massive scale**.



## The concern: downstream social harms

- Many of the treatments in survey of PAPs: low individual risk of harm.
- But by changing electoral behavior, we might **change election outcomes**.
- Changing the outcome of a contested election will **harm** some actors, e.g., the candidate made to lose and her supporters.
  - Downstream social impacts (Teele, 2013; Phillips, 2021; McDermott and Hatemi, 2020)
  - Often unpredictable (Baele, 2013; Carlson 2020)
  - Potential for disparate welfare impacts across groups (McDermott and Hatemi, 2020; Gubler and Selway, 2016; and Zimmerman 2016)
- APSA (2020) guidance (p. 15): interventions are of “**minimal social risk** if they are not done at a scale liable to alter electoral outcomes.”

## Surveying current practice

- Are we currently designing experiments with an eye to **aggregate effects**?
  - **No**: Aggregate electoral impact discussed in 2/129 PAPs
- Why do we need tools to do this better?
  - We need to consider mapping to the level of vote aggregation, the **district** → typically not unit of assignment or outcome measurement.
  - Ethical consideration is about changing **any election outcome**, not an average causal effect.
  - We need to think through these possibilities **ex-ante**.

Framework



## Three key elements

1. The **counterfactual** to the experiment: What would have happened in the absence of a randomized experiment?
  - No intervention?
  - **Partner** conducts (likely) non-randomized intervention.
2. A model of **voter response** to the intervention:
  - Whose behavior might change in response → **interference** assumptions
  - A model of **how voters respond**.
3. A **decision rule** that compares potential aggregate electoral impact to predicted outcomes.

## Experiments and their counterfactuals

- Critical question: who gets the intervention without **absent** the randomized experiment?

Stratum	Intervention	
Voter is...	$\pi(e)$	$\pi(\neg e)$
Always assigned to intervention	1	1
If-experiment assigned to intervention	1	0
If non-experiment assigned to intervention	0	1
Never assigned to intervention	0	0

- Typically if “always assigned” or “if non-experiment assigned” is non-empty  
→ **a partner** is doing the intervention anyway.
  - But **randomizing** can have a distinct impact.
  - This is the impact we should be guarding against.
  - Distinct from current treatments of partners as a “get out of jail free” card.

# Modeling voter behavior

- **Interference assumption**: which voters' behaviors can be affected by intervention?
  - Baseline: SUTVA + no within-cluster inference: intervention only impacts those directly assigned.
  - Extension: only SUTVA
  - Extension: no SUTVA → but we need to bound interference.
- How many **votes change**?
  - **Manski (extreme value) bounds** provide a maximally agnostic model of voter response to treatment.
  - Outcome defined as:

$$Y_i = \begin{cases} 1 & \text{vote for ex-ante marginal winner} \\ 0 & \text{else} \end{cases}$$

- EV bounds will be **conservative** if assumption about interference is correct (or conservative).



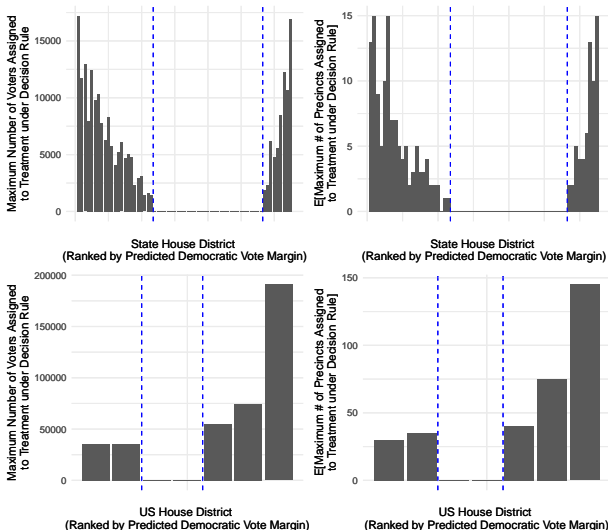
## Maximal aggregate electoral impact + decision rule

- Collectively we can characterize **Maximal Aggregate Electoral Impact** in district  $d$  from:
  1. Who is treated due to experiment? → **counterfactual** to experiment
  2. Whose behavior might be affected? → **interference** assumption
  3. How many votes are changed? → **Manski bounds**
- We can also **predict** margin of victory.
  - Better prediction in some places than others.
  - But we can examine the predictive distribution.
- Decision rule in district  $d$ : Run an experiment iff:

$$2MAEI < \hat{F}^{-1}(0.05)$$

# What does this get us?

- Example from **simulation** on 2018 state house and US House elections in Colorado



## Discussion



## Five recommendations + tradeoffs for learning

### ○ Five recommendations:

1. Select treatments to improve the plausibility of **restricted interference**.
2. Experiment in **FPTP** races.
3. Implement interventions in **larger electoral districts**.
4. Avoid implementing experimental interventions in close or **unpredictable** races.
5. Reduce the number of subjects per district **assigned because of experiment**.

### ○ Posit **tradeoffs** for:

- The **types of treatments** we study experimentally
- **Generalizability** of measured effects (#2-#4)
- **Power** (#5, possibly #4)

## Is the ethical objective too strict? Too permissive?

- Ethical objective is to **avoid changing aggregate outcomes**
- But there are other arguments that may lead to a stricter or more permissive approach to electoral experiments.

Justifications for <b>stricter</b> rules	Justifications for more <b>permissive</b> rules
1. Lack of consent	1. Intervention increases welfare
2. Self-determination	2. Election outcomes have many causes

- I view impetus to avoid changing aggregate outcomes as a **useful default**.

# Conclusion

- Common criticism of electoral experiments: changing electoral outcomes  
→ **social harms**
  - Extant ethical guidance generally suggests we shouldn't be changing elections around the world.
- ... but not reflected in the design of **preregistered electoral experiments** surveyed.
- I provide tools to design experiments that are unlikely to change electoral outcomes.
- Framework shows which **design levers** we can use to reduce aggregate electoral impact.
  - Study of "permissible" designs allows us to understand **limits** of what we can learn from electoral experimentation.

Thanks!

