

Gov 51: Inference for Experiments

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Comparison between groups

- More interesting to compare across groups.
 - Differences in public opinion across groups
 - Difference between treatment and control groups.
- Bedrock of causal inference!

Social pressure experiment

- Back to the Social Pressure Mailer GOTV example.
 - Primary election in MI 2006
- Treatment group: postcards showing their own and their neighbors' voting records.
 - Sample size of treated group, $n_T = 360$
- Control group: received nothing.
 - Sample size of the control group, $n_C = 1890$

Outcomes

- Outcome: $X_i = 1$ if i voted, 0 otherwise.
- Turnout rate (sample mean) in treated group, $\bar{X}_T = 0.37$
- Turnout rate (sample mean) in control group, $\bar{X}_C = 0.30$
- Estimated **average treatment effect**

$$\widehat{ATE} = \bar{X}_T - \bar{X}_C = 0.07$$

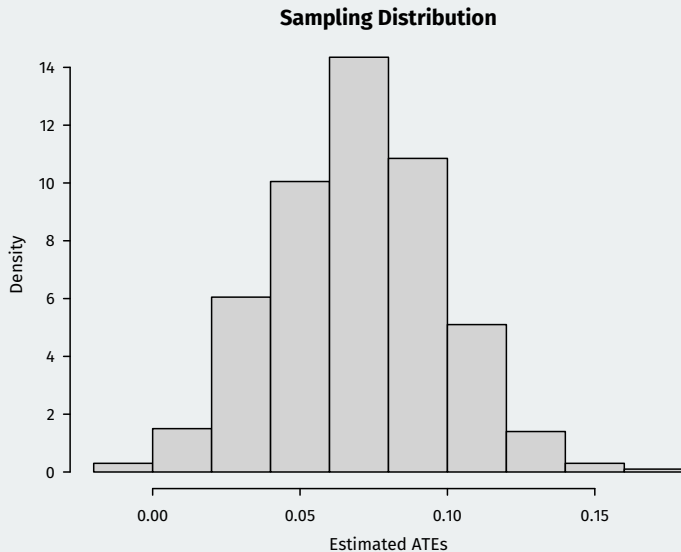
Inference for the difference

- Parameter: **population ATE** $\mu_T - \mu_C$
 - μ_T : Turnout rate in the population if everyone received treatment.
 - μ_C : Turnout rate in the population if everyone received control.
- Estimator: $\widehat{ATE} = \bar{X}_T - \bar{X}_C$
- \bar{X}_T is a r.v. with mean $\mathbb{E}[\bar{X}_T] = \mu_T$
- \bar{X}_C is a r.v. with mean $\mathbb{E}[\bar{X}_C] = \mu_C$
- $\rightsquigarrow \bar{X}_T - \bar{X}_C$ is a r.v. with mean $\mu_T - \mu_C$
 - Sample difference in means is on average equal to the population difference in means.

- What if these were the true population means? We would still expect some **variation** in our estimates:

```
xt.sims <- rbinom(1000, size = 360, prob = 0.37) / 360
xc.sims <- rbinom(1000, size = 1890, prob = 0.30) / 1890

hist(xt.sims - xc.sims, freq = FALSE, xlab = "Estimated ATEs",
     main = "Sampling Distribution")
```



Standard error

- Is an $\widehat{ATE} = 0.07$ big?
- How much variation would we expect in the difference in means across repeated samples?
- **Variance** of our estimates:

$$\begin{aligned}\mathbb{V}(\widehat{ATE}) &= \mathbb{V}(\bar{X}_T - \bar{X}_C) = \mathbb{V}(\bar{X}_T) + \mathbb{V}(\bar{X}_C) \\ &= \frac{\mu_T(1 - \mu_T)}{n_T} + \frac{\mu_C(1 - \mu_C)}{n_C}\end{aligned}$$

- **Standard error** is the square root of this variance:

$$\widehat{SE}_{\widehat{ATE}} = \sqrt{\frac{\bar{X}_T(1 - \bar{X}_T)}{n_T} + \frac{\bar{X}_C(1 - \bar{X}_C)}{n_C}} = 0.028$$

- SE represents how far, on average, $\bar{X}_T - \bar{X}_C$ will be from $\mu_T - \mu_C$.

Confidence intervals

- We can construct confidence intervals based on the CLT like last time.

$$\begin{aligned}CI_{95} &= \widehat{ATE} \pm 1.96 \times \widehat{SE}_{\widehat{ATE}} \\ &= 0.07 \pm 1.96 \times 0.028 \\ &= 0.07 \pm 0.054 \\ &= [0.016, 0.124]\end{aligned}$$

- Range of possibilities taking into account plausible chance errors.
- 0 not included in this CI \rightsquigarrow chance error as big as the estimated effect unlikely.