

Power calculations in RD designs

Northwestern Causal Inference Workshop

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Overview: power and sample size calculations

- Statistical power: prob of rejecting $H_0 : \tau = 0$
 - ▶ Prob of detecting a non-zero effect
- Power calculation:
 - ▶ Statistical power given a sample size and $\tau = \tau_A \neq 0$
- Sample size calculation:
 - ▶ Number of obs required to achieve a certain power for a given effect
- Minimum detectable effect (MDE) calculation:
 - ▶ Smallest effect detectable at a given power and sample size

Introduction: Experimental Design

- ATE estimator:

$$\hat{\tau} = \bar{Y}_1 - \bar{Y}_0 = \frac{\sum_{i=1}^n Y_i D_i}{N_1} - \frac{\sum_{i=1}^n Y_i (1 - D_i)}{N_0}$$

where

$$N_1 = \sum_{i=1}^n D_i, \quad N_0 = \sum_{i=1}^n (1 - D_i)$$

- Central limit theorem:

$$\sqrt{n}(\hat{\tau} - \tau_{\text{ATE}}) \rightarrow_{\mathcal{D}} \mathcal{N}(0, \mathbf{V}), \quad \mathbf{V} = \frac{\sigma_1^2}{p} + \frac{\sigma_0^2}{1-p}$$

where $p = \mathbb{P}[D_i = 1]$

Power function

$$H_0 : \tau_{ATE} = 0 \quad vs \quad H_A : \tau_{ATE} \neq 0$$

- Statistic:

$$T = \frac{\hat{\tau}}{\sqrt{V/n}}$$

- Power function:

$$\beta(\tau) = \mathbb{P}_{\tau}[\text{reject } H_0] = \mathbb{P}_{\tau} \left[\frac{|\hat{\tau}|}{\sqrt{V/n}} \geq z_{1-\alpha/2} \right]$$

Approximating the power function

- We can use the CLT to approximate the power function $\beta(\tau)$:

$$\beta(\tau) \approx 1 - \Phi\left(\frac{\sqrt{n}\tau}{\sqrt{V}} + z_{1-\alpha/2}\right) + \Phi\left(\frac{\sqrt{n}\tau}{\sqrt{V}} - z_{1-\alpha/2}\right)$$

- Power increases when:
 - ▶ Sample size n is large
 - ▶ Variance of the estimator V is small

Power and sample size calculation

$$\beta(\tau) \approx 1 - \Phi\left(\frac{\sqrt{n}\tau}{\sqrt{V}} + z_{1-\alpha/2}\right) + \Phi\left(\frac{\sqrt{n}\tau}{\sqrt{V}} - z_{1-\alpha/2}\right)$$

- Suppose we know or can estimate V
- Power calculation:
 - ▶ Set τ , n and V , find β
- Sample size calculation:
 - ▶ Set τ , β and V , find n

Relative sample size

- How do we divide treated and controls given a total sample size n ?
- Asymptotic variance of $\hat{\tau}$:

$$\frac{V}{n} = \frac{1}{n} \left(\frac{\sigma_1^2}{p} + \frac{\sigma_0^2}{1-p} \right)$$

- Minimizing the variance with respect to $p = \mathbb{P}[D_i = 1]$,

$$p^* = \frac{\sigma_1}{\sigma_1 + \sigma_0}$$

- ▶ Assign more people to group with higher s.d.
- ▶ If $\sigma_1 = \sigma_0$, $p^* = 1/2$

Power calculations in RD designs

- Same general idea: use CLT to approximate power function
- But need to incorporate robust bias-corrected inference
- Robust bias-corrected inference:

$$\frac{\hat{\tau}^{\text{bc}} - \tau_{\text{RD}}}{\sqrt{\hat{V}^{\text{bc}}}} \rightarrow_{\mathcal{D}} \mathcal{N}(0, 1), \quad \hat{\tau}^{\text{bc}} = \hat{\tau} - \hat{B}$$

- ▶ $\hat{\tau}^{\text{bc}}$ is the bias-corrected estimator
- ▶ \hat{V}^{bc} is the robust variance estimator

Power calculations in RD designs

- Power function:

$$\beta^{\text{bc}}(\tau) = 1 - \Phi\left(\frac{\tau}{\sqrt{\widehat{V}^{\text{bc}}}} + z_{1-\alpha/2}\right) + \Phi\left(\frac{\tau}{\sqrt{\widehat{V}^{\text{bc}}}} - z_{1-\alpha/2}\right)$$

where

$$\widehat{V}^{\text{bc}} = \frac{\widehat{V}_+^{\text{bc}}}{nh_+} + \frac{\widehat{V}_-^{\text{bc}}}{nh_-}$$

- Power depends on:

- ▶ Effective sample size: n and h_+, h_-
- ▶ Size of the effect τ
- ▶ Outcomes variances, kernel, order of polynomial, etc: $\widehat{V}_+^{\text{bc}}, \widehat{V}_1^{\text{bc}}$

Sampling in RD designs

- $\beta^{\text{bc}}(\tau)$ can be used to calculate required sample size
- Relative sample sizes:

$$p^* = \frac{\sqrt{\widehat{V}_+^{\text{bc}}}}{\sqrt{\widehat{V}_+^{\text{bc}}} + \sqrt{\widehat{V}_-^{\text{bc}}}}$$

- How to sample:
 - ▶ Order obs according to distance to cutoff
 - ▶ Begin with closest obs, continue in order of distance to cutoff