

# Causal Inference Assignment 1

Due noon Saturday 6/21

ICPSR Session 1 (June 16, 2025)

1. Refer to the the **acorn** data set that accompanies this assignment. Using the mean of turnout proportions in treatment group precincts,  $n_1^{-1}\mathbf{Z}'\mathbf{y}$ , as test statistic, simulate its rerandomization distribution under the null hypothesis of strictly no effect, reporting:

- (a) your simulation  $p$ -value;
- (b) your simulation approximation of  $E[n_1^{-1}\mathbf{Z}'\mathbf{y}]$ , the null expected value of the test statistic;
- (c) your simulation approximation of  $\text{Var}[n_1^{-1}\mathbf{Z}'\mathbf{y}]$ , this test statistic's variance under the null.

Note that  $n_1$  is the number of treated units,  $n$  is the total number of units,  $\mathbf{Z}$  is the random assignment variable and  $\mathbf{y}$  is the observed outcome variable

2. Calculate  $E[n_1^{-1}\mathbf{Z}'\mathbf{y}]$ , the expected value of the sample mean under the strict null hypothesis of no effect, from first principles — i.e., without simulations — using data in  $\mathbf{y}$ .
3. In this setup,  $\text{Var}[\bar{y}_1] = \frac{1}{n_1} \frac{n_0}{n} \frac{\sum_{i=1}^n (y_i - \bar{y})^2}{n-1}$ , where  $\bar{y}_1 = n_1^{-1}\mathbf{Z}'\mathbf{y}$  is the mean of  $y$ s among the group assigned to treatment ( $\{i : Z_i = 1\}$ ) while  $\bar{y} = n^{-1}\mathbf{1}'\mathbf{y}$  is the mean of  $y$  over the full study population. If you've seen formulas for the sampling variance of the mean in earlier stats courses, you probably saw

$$\widehat{\text{Var}}[\bar{y}_1] = \frac{s_y^2}{n_1} = \frac{1}{n_1} \frac{\sum_{i=1}^{n_1} (y_i - \bar{y}_1)^2}{n_1 - 1}, \text{ and/or } \text{Var}[\bar{y}_1] = \frac{\sigma_y^2}{n_1} = \frac{1}{n_1} \frac{\sum_{i=1}^N (y_i - \bar{y})^2}{N}.$$

Which gives a larger value, our new  $\text{Var}[\bar{y}_1]$ -formula or the  $\text{Var}[\bar{y}_1]$  formula immediately above? Explain why it makes sense that the formula we've given would differ in the direction it does from this other formula.

4. Calculate  $\text{Var}[n_1^{-1}\mathbf{Z}'\mathbf{y}]$  using the appropriate formula. Determine the error of the simulation-based approximation to this quantity that you reported in question 1, expressing it as a percentage of  $\text{Var}[n_1^{-1}\mathbf{Z}'\mathbf{y}]$ .
5. Determine the Normal theory approximation to  $\Pr(n_1^{-1}\mathbf{Z}'\mathbf{y} \geq n_1^{-1}\mathbf{z}'\mathbf{y})$ . (Hints: Use the variance and expected values calculated in 4 and 2 to transform your observed treatment group mean into a corresponding “z-score.” To determine Normal quantiles corresponding to z-scores in R, use `pnorm()`; type `?pnorm` for help. And  $\mathbf{z}'$  refers to the observed Acorn data treatment assignment.)
6. A researcher plans to ask six subjects to donate time to an adult literacy program. Each subject will be asked to donate either 30 ( $Z = 0$ ) or 60 ( $Z = 1$ ) minutes. The researcher is considering three methods for randomizing the treatment. Method I is to make independent decisions for each subject, tossing a coin each time. Method C is to write “30” and “60” on three playing cards each, and then shuffle the six cards. Method P tosses one coin for each of the 3 pairs (1, 2), (3, 4), (5, 6), asking for 30 (60) minutes from exactly one member of each pair.
  - a Discuss strengths & weaknesses of each method.
  - b How would your answers to (a) change if  $n : 6 \mapsto 600$ ?
  - c Determine  $E[Z_1]$  under each method.
  - d Determine  $E[Z_1 + Z_2 + \dots + Z_6]$  under each method.
  - e Calculate  $E[\mathbf{Z}'\mathbf{1}]$  under each of the three methods.
  - f For which of the methods does  $E[(\mathbf{Z}'\mathbf{1} - E[\mathbf{Z}'\mathbf{1}])^2] = 0$ ?<sup>1</sup>
  - h For two of the three methods, the linearity property of expected value<sup>2</sup> entails that  $E[\frac{\mathbf{Z}'\mathbf{x}}{\mathbf{Z}'\mathbf{1}}] = (x_1 + x_2 + \dots + x_6)/6$ . Which two are these, and why doesn't the argument work for the third?

<sup>1</sup>I.e., for which does  $\text{Var}[\mathbf{Z}'\mathbf{1}] = 0$ ? (In general,  $\text{Var}[V] = E[V - E[V]]^2$ .)

<sup>2</sup>For constant  $\alpha$  and  $\beta$ ,  $E(\alpha + \beta Y) = \alpha + \beta E(Y)$ . Also  $E(\alpha X + \beta Y) = \alpha E(X) + \beta E(Y)$ .