## **Assignment 7: Causal Inference**

PSCI 8357, Spring 2016 March 24, 2016

This assignment must be turned in by the start of class on **Thursday**, **March 31**. You must follow the instructions for submitting an assignment.

## Main Task

- 1. Generate N = 500 observations of the following data.
  - Four binary covariates, *X*<sub>1</sub>,...,*X*<sub>4</sub>, independent of each other and distributed Bernoulli(0.5) (i.e., 50-50 chance of being 0 or 1).
  - A binary treatment variable, *T*, where

$$\Pr(T=1) = \frac{4 + X_1 + X_2 + X_3 + X_4}{12}$$

• A continuous response variable, *Y*, where

$$Y = T + \exp(X_1 + X_2 + X_3 + X_4 + X_1 X_2 + X_3 X_4) + \epsilon$$

with white noise error  $\epsilon \sim N(0, 1)$ .

Under this model, there is a constant treatment effect of  $\tau = 1$ .

Since you are simulating data at random, remember to use set.seed() at the start of your script so your results are reproducible.

- 2. Use a naïve difference of means test to estimate the average treatment effect. How does the estimate compare to the true ATE?
- 3. Use subclassification to estimate the average treatment effect. How does the estimate compare to the true ATE?

In the unlikely but not-impossible event that there is a grouping with no variation in the treatment, just drop that grouping from the subclassification.

- 4. Use OLS of *Y* on  $(T, X_1, X_2, X_3, X_4)$  to estimate the average treatment effect. How does the estimate compare to the true ATE?
- 5. Use a modified form of subclassification to estimate the average treatment effect. Instead of grouping on the full combination of covariates, just group on the sum  $X_1 + X_2 + X_3 + X_4$ . How does the estimate compare to the true ATE?
- 6. Use a loop to repeat the first five steps M = 1,000 times. Calculate the (approximate) bias, standard error, and mean squared error of the three estimators. (Remember that MSE = Bias<sup>2</sup> + Std. Error<sup>2</sup>.) Test the hypothesis that each estimator is unbiased. Which of the estimators is best overall?
- 7. What if, instead of being binary, each covariate  $X_j$  were uniformly distributed between 0 and 1? Based on the results of your analysis here, how would you choose to estimate the average treatment effect? (**Don't** run a new simulation—try to make an inference from the results with binary covariates.)

## Visualization Challenge

Depict the *joint* relationship between each *X* variable and (1) the probability of treatment and (2) the expected value of the response.