## **Section 5 Activity**

## Main

The goal of this question is to practice simulating data and running regressions in R. It is taken from your textbook.

- 1. Set the RNG seed to be 1
- 2. Create a vector **x** that has 30 observations randomly drawn from a standard normal distribution. (Hint: use the **rnorm()** function).
- 3. Create a second vector **eps** that has 30 observations randomly drawn from a normal distribution with a mean of 0 and a standard deviation of 0.25.
- 4. Using x and eps create a vector y according to the following data generating process

$$Y = -1 + 0.5X + \epsilon$$

- 5. What is the length of y? What are the values of  $\beta$  in the DGP?
- 6. Create a data frame called dgp with the variables created in 2-4.
- 7. Using ggplot2 create a scatterplot of the relationship between x and y.
- 8. Run the regression of y on x and report the summary of the model. Call this m1. Comment on why you expect the result. (Hint: consider the discussion of the Conditional Expectation Function from lecture)
- 9. Using ggplot2 add the least squares line to your previous plot. Give it a color other than black. Draw the population regression line on the plot in a different color.
- 10. Create a second model m2 that adds a squared term  $x^2$  to the model. Is there evidence that the term improves the model fit? Which model is "correct"?
- 11. For both models, manually predict the result of y when x = 4. Would you trust either prediction?

## **Bonus**

- 1. Add a new variable **z** to the dgp data frame that has 30 observations randomly drawn from a Poisson distribution. Set lambda=3.
- 2. Update the y variable in the dgp data frame so that Y is now drawn from the following data generating process.

$$-1 + 0.5X + .25Z + .75(XZ) + \epsilon$$

3. Run a new model called  ${\tt m3}$  that would perfectly estimate the CEF in expectation. Report the summary of this model.