To the extent that it helps you to learn, you may work with fellow students on this assignment. R and Python have extensive libraries online that can guide you on this assignment. [25 points – each question for 5 points]

- 1. In words, the "standard normal distribution" is a normal distribution with mean zero and variance one, denoted here as N(0,1). For a random variable X that is distributed as a standard normal, mathematically we write  $X \sim N(0,1)$ . [5 points]
  - a. Using R or Python, write code to draw at random 10 observations from a N(0,1) random variable. Instruct the machine to calculate the mean, variance and standard deviation of your draws.
  - b. Repeat this exercise using 10,000 draws from a N(0,1), instructing again the machine to calculate the mean, variance and standard deviation of your draws.
  - c. Repeat this exercise with 1,000,000 draws from a N(0,1), instructing again the machine to calculate the mean, variance and standard deviation of your draws.
  - d. What conclusions, if any, do you draw from increasing the sample size?
  - e. Submit your code and results.
- 2. We discussed at some length the bivariate linear regression model,  $y_i = \beta_0 + \beta_1 x_i + \epsilon_i$ . [5 points]
  - a. Go to <a href="http://www.random.org/integers/">http://www.random.org/integers/</a> and generate two series of 1,000 random integers with values between 0 and 9. Call one series y and the other x.
  - b. Using Python or R, fit the bivariate linear regression model.
  - c. Examine your t-statistic to evaluate whether it is greater than two in absolute value. Would you reject or fail to reject that there is any relationship between these two series?
  - d. Submit your series, your code, and your results.
- 3. Go to Yahoo!Finance (or the source of your choice, such as Bloomberg) and download a daily price series for a particular publicly-traded stock of your choice for a ten-year time period (don't use Apple), as well as the daily price series on the exchange on which it trades. [5 points]
  - a. Using R or Python, calculate the log returns of each series as the natural log of the ratio of (price today/price yesterday). Use the reported closing price for this exercise.
  - b. Using R or Python, generate a histogram of log returns of the stock of your choice.
  - c. Using R or Python, generate a scatterplot that relates the log returns of your stock of choice to the log returns of the exchange on which it is traded.
  - d. Finally, using R or Python, fit a linear model to obtain estimates of what finance folks call the "alpha" and the "beta". Is "alpha" significantly different than zero at a 95% level?

    Does a 95% confidence level for "beta" include one?
  - e. Submit all code and results.
- 4. Download the file train.dta from the course website. These data are formatted as a Stata dataset. [5 points]

- a. Read this dataset into R or Python. (For R, you may find the "foreign" library of use. For Python, check out Pandas. The goal here is to get you familiar with reading datasets with alternative formats.)
- b. Generate summary statistics for the following variables in the data:
  - d, which is an indicator for whether a particular email is spam
  - x1, which is an attribute of the email
- c. Using least squares, regress d on x1. (For R, check out Im. For Python, check out StatsModels.) Congratulations, you have created a support vector machine that you will use to forecast whether an incoming email with a different attribute is spam.
- d. Suppose you set the threshold that an email is spam if the predicted value exceeds 1.<sup>1</sup> I give you a new email with an attribute value 0.65. Would you classify it as spam or not spam?
- e. I give you another new email with an attribute value of 0.99. Would you classify it as spam or not spam?
- f. Submit code and results.
- 5. This is a very challenging question, but it addresses several key topics in data analytics. You should work collectively on a solution with the recognition that you may not complete it. The phrase "data generating process" (or DGP) is often used to describe the <u>hypothetical process</u> by which observations arise in the real world. We discussed at some length the bivariate linear regression model,  $y_i = \beta_0 + \beta_1 x_i + \epsilon_i$ . In this problem, we will work with a specific DGP and evaluate features of  $\widehat{\beta_1}$ , the least squares estimate of  $\beta_1$ . [5 points]
  - a. Suppose your DGP is  $y_i = 1 + 2x_i + \epsilon_i$ , where  $x \sim N(0,1)$  and  $\epsilon \sim N(0,1)$ .
  - b. Using R or Python, write code to generate 1,000 draws for x and  $\epsilon$ . Use these draws to generate y in accordance with the DGP in a.
  - c. Using R or Python, write code to estimate the bivariate model,  $y_i = \beta_0 + \beta_1 x_i$  and summarize the findings.
  - d. Repeat b. and c. above five times for a new set of random draws for each replication. (This effort is called Monte Carlo simulation.)
  - e. Given what you've done in d., Suppose you wrote code to repeat b. and c. above 1,000 times, each time recording the estimated value of  $\beta_1$ . What do you think a histogram of these 1,000 replications of the estimate value of  $\beta_1$  would show?
  - f. Suppose that you were not interested in the estimate of  $\beta_1$  per se, but instead in some functional transformation, such as the estimate of  $\exp(\beta_1)$ . What might you do with your 1,000 replications from e. above to inform you about the distribution of the estimate of  $\exp(\beta_1)$ ?
  - g. Submit code and results.

<sup>1</sup> Remember that the predicted value is simply the estimated coefficient from your regression times the value of the attribute.