**DS 710**

**Homework 6**

**R assignment**

1. Can we detect when a marketing campaign has been successful?

1. On homework 4, you simulated data from the TableFarm salad chain before and after the implementation of a new marketing campaign.  Read the combined data (both before and after) into R.  (You could do this by saving the data as a .csv file and using read.csv(), or by copying the data into a text file, separating the values by commas, and enclosing the data in c( … ) to make a vector.)
2. Make a scatterplot of the data.  Add a vertical line to mark the month in which the new marketing campaign began, and add a legend to your plot.
3. Make side-by-side boxplots of the revenue before and after implementing the marketing campaign.  Write a few sentences describing and comparing the boxplots, and relating them to the underlying model you used to simulate the data.
4. Based on the way you simulated the data, you know that the marketing campaign was successful; that is, the data after implementing the marketing campaign was simulated from an underlying model with a higher mean than before the marketing campaign.  However, in real life we probably wouldn’t know this.  Based on the scatterplot and boxplots, would you be confident in claiming that the marketing campaign was successful?  Why or why not?
5. Write the null and alternative hypotheses for a test of whether the marketing campaign was successful.  (I.e., whether the mean revenue with the marketing campaign is higher than the mean revenue before the marketing campaign.)
6. In a few sentences, explain why a 2-sample t-test is appropriate for testing the hypotheses in part e.
7. Conduct a 2-sample t-test in R.  Include the R output and state your conclusion in the context of the problem.

2.  Can we detect an association between chocolate consumption and Nobel prizes?

1. On homework 4, you simulated data on countries’ per-capita chocolate consumption and number of Nobel Prize winners, using an error term (representing random “noise”).  Read these data into R and make a scatterplot of the number of Nobel Prize winners versus chocolate consumption.
2. Fit a linear model to the data.  What is the equation of the line of best fit?  How does it compare to the theoretical model you used to simulate the data?  Graph the line of best fit with the scatterplot.
3. State the null and alternative hypotheses for a test of whether the number of Nobel Prize winners (per 10 million population) is associated with per-capita chocolate consumption.
4. State your conclusion about the hypotheses in part c, in the context of the problem.
5. Graph the diagnostic plots for the regression.  Explain what they tell us.

3.  In homework 5, you counted the frequencies of letters in two encrypted texts.  In this problem, you will use statistical analysis to identify the language in which the text was written, and decrypt it.

1. Read the letter frequencies from encryptedA into R and attach the data.  Use the following code to make a barplot of the letter frequencies, with the letters listed in order of increasing frequency:  (Here I’ve assumed that your columns were named “key” and “count”.)

encrypt\_order = order(count)

barplot( count[encrypt\_order], names.arg = key[encrypt\_order] )

Be sure you understand what this code does.

1. The file Letter Frequencies.csv contains data on the frequencies of letters in different languages.  (Source:  <http://www.sttmedia.com/characterfrequency-english> and <http://www.sttmedia.com/characterfrequency-welsh>, accessed 21 August 2015. Used by permission of Stefan Trost.)  Read these data into R.
2. In a single graphing window, display two bar plots:  A plot on top showing the encrypted frequencies, and a plot below it showing the frequencies of letters in English.  Each plot should be sorted in order of increasing frequency.  Each plot should also have a title telling whether it is from the encrypted text or from plain English.
3. Based on the **shape** of the plots, do you think it is likely that the encrypted text came from English?  Explain.

(Note: The order of the letters along the horizontal axis of each plot will be quite different, because one plot shows the frequencies in plain English, and the other shows the frequencies in the encrypted text. So, you should ignore what letter is written below each bar when answering this question. Instead, look at things like the relative frequency of the most-common letter and the second-most common.)

1. We want to conduct a hypothesis test to be more precise about whether it is plausible that the text came from English.  To do this, we will pair up each letter in the encrypted text with a letter in English, based on the order of frequency.  So, encrypted1 “u” is paired with English “e”, encrypted1 “s” is paired with English “t”, etc.  Then we will test whether the resulting letter frequencies plausibly come from a random sample of English words.

To pair up the letters, sort the vector of counts from the encrypted text in order of increasing frequency, and store it as a new vector. Then do the same thing with the vector of frequencies from English.

1. To pair up the letters, we need the data (the counts of letters from encryptedA.txt) and the probability model (the theoretical frequencies from Letter Frequencies.csv) to have the same number of letters. Depending on how you formatted your output from Python, your letter counts may include 20 or 26 letters. This is due to the fact that some letters did not appear in the encrypted text, so they appeared 0 times. If necessary, prepend 6 zeroes to the *count* vector to make it the same length as the theoretical frequencies:

count = c( rep(0, 6), count )

1. State the null and alternative hypotheses for a chi-squared Goodness of Fit test of this question.
2. To satisfy the assumptions of a Goodness of Fit test, we need the expected counts of each category to be greater than or equal to 5. Find the total number of letters in the encrypted text. Then multiply this number by the probabilities from Letter Frequencies.csv to get the expected counts.
3. Combine categories (letters) to get expected counts that are greater than or equal to 5. For example, if you decided to combine the first two categories, you could use the code

sortEnglish\_combined = c( sum(sortEnglish[1:2]), sortEnglish[3:26] )

Combine the same categories in the encrypted counts.

1. Use R to conduct the chi-squared Goodness of Fit test.

1. State your conclusion in the context of the problem.
2. Repeat steps h-k for Welsh, and then repeat for both languages for encryptedB.  Based on the hypothesis tests, which text do you think came from which language?  How confident are you in your assessment?
3. Optional: Try to decrypt the English text. Simon Singh’s Black Chamber website (<http://www.simonsingh.net/The_Black_Chamber/substitutioncrackingtool.html>) will automatically substitute letters for you, so you can test different possibilities for what English plaintext letter is represented by each letter in the ciphertext. Start by substituting the letter E for the most common letter in the ciphertext. Then use frequencies of letters in the ciphertext, common patterns of letters, and experimentation to determine other substitutions.

Submit a .doc, .docx, .rmd, or .pdf file to GitHub containing your R code, R output and graphs, and your written interpretations and explanations. (You may include your responses for problems 1, 2, and 3 in the same file.)