# Statistics 5014: Homework 5

Due Tuesday Sep 26, 9am 2018-09-26

## Problem 3: What makes a good figure?

A good figure needs to tell a good story about the data in question. In order to do so, it should be concise, clear, and easy to read in a minimal amount of time. Simultaneously, it should contain sufficient information to send a few key messages, and the information should be categorized in colors, shapes, and patterns that are easily distinguishable.

#### Problem 4

a. Create a function that computes the proportion of successes in a vector. Use good programming practices.

```
pSuccess <- function(bin_vect){
    # Check if bin_vect is a binary vector and execute an error if not
    if( any(which(bin_vect[which(bin_vect != 1)] != 0)) ){
        stop("Input should be a binary vector")
    }
    # Calculate proportion of success
    length(which(bin_vect == 1))/length(bin_vect)
}</pre>
```

b. Create a matrix to simulate 10 flips of a coin with varying degrees of "fairness" as follows:

```
set.seed(12345)
# Original code, repeats a single vector of 10 random flips 10 times,
# and so all 10 columns are the same.
P4b_data <- matrix(rbinom(10, 1, prob = (30:40)/100), nrow = 10, ncol = 10)</pre>
```

c. Use your function in conjunction with apply to compute the proportion of success in P4b\_data by column and then by row. What do you observe? What is going on?

When applying the function over the columns, the proportion of success of each column vector is returned, resulting in a vector showing the probability of each column. The same performed with the rows when applying the function over the rows. However, the given matrix P4b\_data repeats a single vector of 10 random flips 10 times, and so all 10 columns are the same. This results in the probabilities of all the columns to be equal, as shown in success\_col, while success\_row only returns probabilities of 0 or 1, since each element in the same row is repeated.

```
success_col <- apply(P4b_data, 2, pSuccess) # Compute probabilities by col
success_col</pre>
```

```
## [1] 1 1 1 1 0 0 0 0 1 1
```

d. You are to fix the above matrix by creating a function whose input is a probability and output is a vector whose elements are the outcomes of 10 flips of a coin. Now create a vector of the desired

probabilities. Using the appropriate apply family function, create the matrix we really wanted above. Prove this has worked by using the function created in part a to compute and tabulate the appropriate marginal successes.

```
rSuccess <- function(prob_vect){
  # Check if prob_vect is a vector of probabilities and execute an error if not
  if( any(prob_vect < 0 | prob_vect > 1) ){
    stop("Vector elements are probabilities. They should range from 0 to 1.")
  # Create a matrix that is dependent on the length and values of prob vect
  # and returns a matrix of 10 randomly generated binary values
  # for each element of prob_vect
  matrix(rbinom(10*length(prob_vect), 1, prob = prob_vect),
         ncol = length(prob_vect), byrow = T)
}
# Test run of rSuccess and pSuccess
test_vect <- runif(12, 0, 1) # Randomly create a vector of probabilities
test_matrix <- rSuccess(test_vect)</pre>
# Perform Monte-Carlo simulation to confirmthe original probabilities
set.seed(8034)
test_list <- list() # Make an empty list for the simulated data
M <- 2000 # Number of simulations
# Make an empty dataset for simulated probabilities
test_prob <- data.frame(matrix(NA, nrow = M, ncol = length(test_vect)))</pre>
# Fill the list with simulated data
for(i in 1:M){
  test list[[i]] <- rSuccess(test vect)</pre>
  test_prob[i,] <- apply(test_list[[i]], 2, pSuccess)</pre>
test_results_col <- apply(test_prob, 2, mean) # Average probabilities by column
test_results_row <- apply(test_matrix, 1, mean) # Average probabilities by row
kable(as.data.frame(rbind(test_vect, test_results_col),
      row.names = c("Actual probability", "Simulated probability")),
      caption = "Marginal successes by column", digits = 2, format = "pandoc")
```

Table 1: Marginal successes by column

	X1	X2	Х3	X4	X5	X6	X7	X8	X9	X10	X11	X12
Actual probability	0.03	0.15	0.74	0	0.39	0.46	0.39	0.4	0.18	0.95	0.45	0.33
Simulated probability	0.03	0.15	0.74	0	0.39	0.47	0.39	0.4	0.18	0.95	0.45	0.33

Table 2: Marginal successes by row

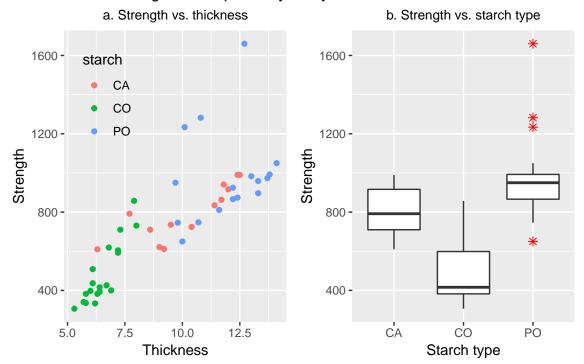
	V1	V2	V3	V4	V5	V6	V7	V8	V9	V10
Probability of success	0.42	0.33	0.42	0.58	0.5	0.42	0.42	0.5	0.58	0.42

### Problem 5

Overall, there is a positive relationship between strength and thickness. Among the different starch types, there is some overlap in the distributions of strength and thickness. The median strength of the PO type starch appears to be the greatest, and a few outliers can be observed for the group.

```
# Store URL for starch data
url <- "http://www2.isye.gatech.edu/~jeffwu/book/data/starch.dat"</pre>
starch_data <- fread(url, header = T, fill = T, sep = " ") # Import data
# Scatterplot of numeric data
starch_plot1 <- ggplot(starch_data, aes(x = thickness, y = strength,</pre>
                                         col = starch)) + geom_point() +
  theme(plot.title = element_text(size = 10, hjust = 0.5),
        legend.justification=c(0,1), legend.position=c(0.05, 0.95),
        legend.background = element blank(), legend.key = element blank()) +
  labs(title = "a. Strength vs. thickness", x = "Thickness", y = "Strength")
# Box plot of strength
starch_plot2 <- ggplot(starch_data, aes(x = starch, y = strength)) +</pre>
  geom boxplot(outlier.colour="red", outlier.shape = 8, outlier.size = 2) +
  labs(title = "b. Strength vs. starch type", x = "Starch type",
       y = "Strength") +
  theme(plot.title = element_text(size = 10, hjust = 0.5))
grid.arrange(starch_plot1, starch_plot2, nrow = 1,
             top = "Figure 1. Exploratory analysis of starch data")
```

Figure 1. Exploratory analysis of starch data



## Problem 6: Create an annotated map of the US

Part a. Get and import a database of US cities and states.

```
#we are grabbing a SQL set from here
# http://www.farinspace.com/wp-content/uploads/us_cities_and_states.zip
#download the files, looks like it is a .zip
library(downloader)
download("http://www.farinspace.com/wp-content/uploads/us_cities_and_states.zip", dest = "us_cities_sta
unzip("us_cities_states.zip", exdir="./")
#read in data, looks like sql dump, blah
states <- fread(input = "./us_cities_and_states/states.sql", skip = 23, sep = "'",
                 sep2 = ",", header = F, select = c(2,4))
cities <- fread(input = "./us_cities_and_states/cities.sql", sep = "'",</pre>
                 sep2 = ",", header = F, select = c(2,4))
colnames(states) <- c("state", "abbr")</pre>
colnames(cities) <- c("city", "abbr")</pre>
states_50 <- states[states$abbr != "DC"] # Limit states to the 50
Part b. Create a summary table of the number of cities included by state.
# Remove rows that have states not listed in the states dataset
uscities <- subset(cities, abbr %in% states_50$abbr)
table(uscities$abbr) # View summary table
##
                                                                                IN
##
     ΑK
          AL
                AR
                     AZ
                          CA
                                CO
                                     CT
                                          DE
                                                FL
                                                     GA
                                                          ΗI
                                                                ΙA
                                                                     ID
                                                                           IL
##
    229
                                                                               738
         579
               605
                    264 1239
                               400
                                    269
                                           57
                                               524
                                                    629
                                                           92
                                                               937
                                                                    266 1287
##
    KS
          ΚY
               LA
                     MA
                          MD
                               ME
                                     ΜI
                                          MN
                                                MO
                                                     MS
                                                          MT
                                                                NC
                                                                     ND
                                                                           NE
                                                                                NH
##
    634
         803
               479
                    511
                         430
                               461
                                    885
                                         810
                                               942
                                                    440
                                                          360
                                                               762
                                                                    373
                                                                          528
                                                                               255
##
               NV
                                                     SC
                                                          SD
                                                                                VA
    NJ
          NM
                     NY
                          OH
                                OK
                                     OR
                                          PA
                                                RΙ
                                                                TN
                                                                     TX
                                                                          UT
##
    579
         346
                99 1612 1069
                              585
                                    379 1802
                                                70
                                                    377
                                                         364
                                                               548 1466
                                                                          250
                                                                               839
##
     VT
          WΑ
                WT
                     WV
                          WY
##
    288
         493
              753
                   753
                         176
citynum <- data.frame(table(uscities$abbr))</pre>
colnames(citynum) <- c("abbr", "num")</pre>
states_50 <- cbind(states_50, num = citynum$num)</pre>
```

Part c. Create a function that counts the number of occurrences of a letter in a string. The input to the function should be "letter" and "state\_name". The output should be a scalar with the count for that letter.

Create a for loop to loop through the state names imported in part a. Inside the for loop, use an apply family function to iterate across a vector of letters and collect the occurrence count as a vector.

```
# Function that counts the number of letters in a string
getCount <- function(letter, state_name){
   state_lower <- tolower(state_name)
   count <- sum(strsplit(state_lower, "")[[1]] == letter)
   return(count)
}

# Collect the occurrence count of state names as a matrix
letter_count <- data.frame(matrix(NA, nrow = 50, ncol = 26))

for(i in 1:50){
   letter_count[i,] <- sapply(letters, getCount, state_name = states_50$state[i])</pre>
```

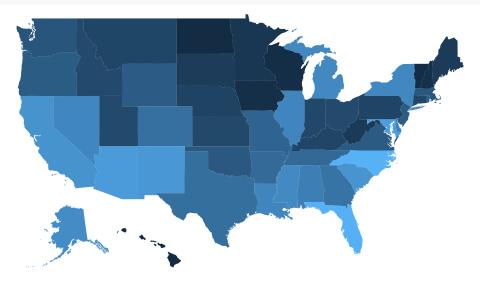
}

### Part d.

Create 2 maps to finalize this. Map 1 should be colored by count of cities on our list within the state. Map 2 should highlight only those states that have more than 3 occurrences of ANY letter in their name.

Quick and not so dirty map:

```
#https://cran.r-project.org/web/packages/fiftystater/vignettes/fiftystater.html
library(fiftystater)
data("fifty_states") # this line is optional due to lazy data loading
crimes <- data.frame(state = tolower(rownames(USArrests)), USArrests)</pre>
# map id creates the aesthetic mapping to the state name column in your data
p1 <- ggplot(crimes, aes(map_id = state)) +
  # map points to the fifty_states shape data
  geom_map(aes(fill = Assault), map = fifty_states) +
  expand_limits(x = fifty_states$long, y = fifty_states$lat) +
  coord_map() +
  scale x continuous(breaks = NULL) +
  scale_y_continuous(breaks = NULL) +
  labs(x = "", y = "") +
  theme(legend.position = "bottom",
  panel.background = element_blank())
р1
```





```
# Map states with more than 3 of the same letter
states_50$state <- tolower(states_50$state)
state_3lett <- states_50[apply(letter_count, 1, max) > 3]

p2 <- ggplot(state_3lett, aes(map_id = state)) +
    geom_map(aes(fill = num), map = fifty_states) +
    expand_limits(x = fifty_states$long, y = fifty_states$lat) +</pre>
```

```
coord_map() +
scale_x_continuous(breaks = NULL) +
scale_y_continuous(breaks = NULL) +
labs(x = "", y = "") +
theme(legend.position = "bottom", panel.background = element_blank())
p2
```





