HW5_Do_Quyen

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Problem 3

I believe a good figure should provide us with an overall impression about the data like its size, structure, relationship between variables. It should also provide insight into the data and inform the next steps in the analysis process. Some important aspects of a good figures are reasonable scaling, interpretability, size and the use of colors.

Problem 4

a. A function computing the proportion of successes in a vector

```
count_success <- function (vect, value = 1) {
    # Compute the proportion of successes in a vector

# Args:
    # vect: the vector on which the proportion of successes will be computed
    # value: the value presented "success" value in the vector. Default value is 1

#Return:
    # A real number from 0 to 1

length(vect[which(vect==value)])/length(vect)
}</pre>
```

b. Create a simuluated matrix

```
set.seed(12345)
P4b_data <- matrix(rbinom(10,1,prob =(30:40)/100),nrow = 10, ncol =10)</pre>
```

c. Checking the proportion of success

```
# Calculate the proportion of success across matrix row
prop_row <- apply(P4b_data,1,count_success)
prop_col <- apply(P4b_data,2,count_success)

prop_mat <- matrix(c(prop_row,prop_col),nrow=2,ncol=10,byrow = TRUE, dimnames = list(c("By Row","By Col
prop_mat

## 1 2 3 4 5 6 7 8 9 10
```

```
## By Row 1.0 1.0 1.0 1.0 0.0 0.0 0.0 1.0 1.0 ## By Col 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6
```

The matrix of the simulated binomial using the code in b didn't produce the data as the intention. Instead of applying different probabilities to draw success among 10 rows of the matrix, the function seems to apply p = 1 to all the rows instead.

```
d.

simulate_binom <- function(probability) {

# Simulate 10 random binomial variables
```

```
# of n = 10 and given probability
  # Args:
  # probability: the probability for the binomial distribution
  # a vector containing 10 RVs drawn from binomial distribution
  return(rbinom(10, 1, prob = probability))
# A vector of probability
prob_vect <- (31:40)/100
# apply simulate_binom on each element of prob_vect
correct_mat <- sapply(prob_vect,simulate_binom)</pre>
# Calculate the proportion of success
# across rows and columns of correct_mat
prop_row2 <- apply(correct_mat,1,count_success)</pre>
prop_col2 <- apply(correct_mat,2,count_success)</pre>
prop_mat2 <- matrix(c(prob_vect,prop_row2,prop_col2),nrow=3,ncol=10,byrow = TRUE, dimnames = list(c("Tr</pre>
prop_mat2
##
                             2
                                  3
                                       4
                                            5
## True probability 0.31 0.32 0.33 0.34 0.35 0.36 0.37 0.38 0.39 0.4
                    0.70 0.30 0.50 0.50 0.30 0.10 0.80 0.40 0.10 0.2
## By Row
## By Col
                    0.20 0.30 0.40 0.30 0.40 0.60 0.30 0.30 0.50 0.6
```

To fix the code in b, I created a vector of probabilities, then used the sapply on each element of that vector and pasted it onto simulation function. "sapply" on the probability vector return a matrix of data whose columns are 10 random data points drawn from the binomial distribution having corresponding probability from the vector.

The "By col" information tell us the probabilities of success for each true probability. Compared with the true probability, we saw that they are pretty close. Any difference is due to randomization.

Problem 5

```
#Import raw data from url
url <- "https://www2.isye.gatech.edu/~jeffwu/book/data/starch.dat"
starch.dat <- read.csv(url, header=TRUE,sep="")

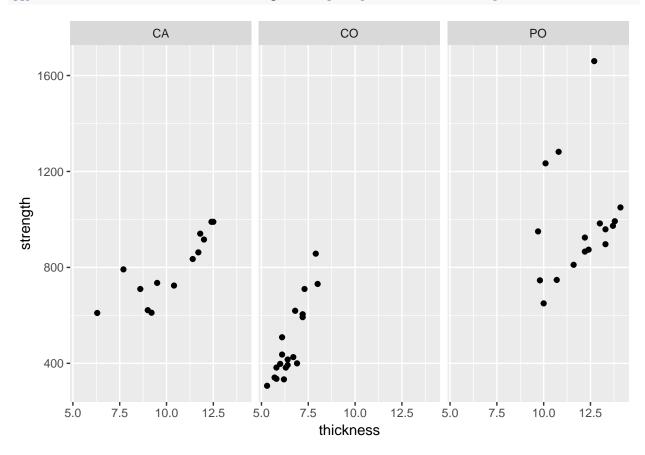
#Summary
str(starch.dat)

## 'data.frame': 49 obs. of 3 variables:
## $ starch : Factor w/ 3 levels "CA","CO","PO": 1 1 1 1 1 1 1 1 1 1 1 1 ...
## $ strength : num 792 610 710 941 990 ...
## $ thickness: num 7.7 6.3 8.6 11.8 12.4 12 11.4 10.4 9.2 9 ...
starch.dat$starch <- factor(starch.dat$starch)
knitr::kable(summary(starch.dat))</pre>
```

starch	strength	thickness
CA:13	Min.: 306.4	Min. : 5.300
CO:19	1st Qu.: 508.8	1st Qu.: 6.700
PO:17	Median: 735.4	Median: 9.500
NA	Mean: 737.0	Mean: 9.388
NA	3rd Qu.: 924.4	3rd Qu.:12.000
NA	Max. $:1660.0$	Max. $:14.100$

```
{\it \#Multipane\ plot\ using\ ggplot\ and\ ggpubr}
p1 <- ggplot(starch.dat,aes(x=strength)) + geom_histogram(colour= "black",bins=10,fill="darkred")</pre>
p2 <- ggplot(starch.dat,aes(x=starch,y=strength ,group=starch,fill=starch))</pre>
p2 <- p2 + geom_boxplot() + guides(fill=FALSE) + labs(x="starch")</pre>
p3 <- ggplot(starch.dat,aes(x=thickness)) + geom_histogram(colour= "black",bins=10,fill="darkgreen")
p4 <- ggplot(starch.dat,aes(x=starch,y=thickness ,group=starch,fill=starch))
p4 <- p4 + geom_boxplot() + guides(fill=FALSE) + labs(x="starch")
p5 <- ggplot(starch.dat,aes(thickness,strength,colour=starch)) + geom_point() + labs(x="thickness",y="s
ggarrange(ggarrange(p1,p2,p3,p4,ncol = 2,nrow=2), p5, nrow = 2)
    12.5 -
10.0 -
7.5 -
5.0 -
2.5 -
0.0 -
                                                       strength 1600 - 1200 - 800 - 400 -
 count
                                                                                   co
                                          1500
                                                                      СA
                 500
                             1000
                                                                                                 ΡO
                          strength
                                                                                  starch
                                                       thickness
 7.5 -
5.0 -
2.5 -
0.0 -
                                                          12.5 -
10.0 -
7.5 -
5.0 -
                                                                      CA
                6
                                                                                   co
                        8
                                10
                                        12
                                                14
                                                                                                 ΡO
                         thickness
                                                                                 starch
    1600 -
    1200 -
 strength
     800 -
     400 -
          5.0
                                   7.5
                                                                                    12.5
                                                           10.0
```

thickness



Problem 6

##

[1,]

325 1587

989

756

961

725

```
##
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
##
       between, first, last
## The following object is masked from 'package:purrr':
##
##
       transpose
Part b. A summmary table of the number of cities included by state
#The number of cities included by states
\#knitr::kable(t(table(cities\$state\_code)), caption = "Number of cities by state")
t(table(cities$state_code))
##
##
                                        CO
                                             CT
                                                             FL
                                                                        ΗI
                                                                              ΙA
             AK
                  AL
                        AR
                             AZ
                                  CA
                                                  DC
                                                        DE
                                                                   GA
##
     [1,]
           273
                 838
                      709
                            532 2651
                                       659
                                            438
                                                  284
                                                        98 1487
                                                                  972
                                                                       139 1060
##
##
                       IN
                             KS
                                                   MD
                                                        ME
                                                                        MO
                                                                              MS
             ID
                  IL
                                  ΚY
                                        LA
                                             MA
                                                             MI
                                                                   MN
```

619

489 1170 1031 1170

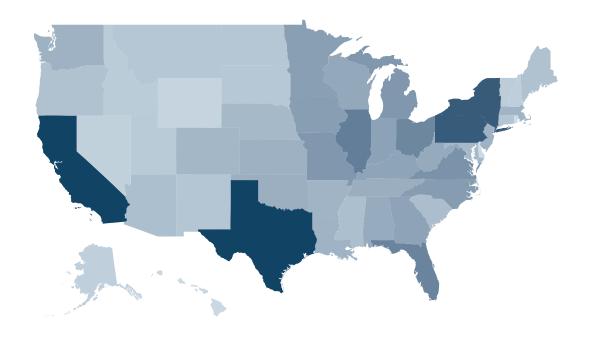
533

703

```
##
##
            MT
                 NC
                       ND
                            NF.
                                NH
                                                  NV
                                                       NY
                                                            OΗ
                                                                  ΩK
                                                                       OR.
                                                                            PA
                                       N.T
                                            MM
##
     [1,] 405 1090
                      407
                           620
                                 284
                                      733
                                           426
                                                 253 2207 1446
                                                                 774
                                                                      484 2208
##
##
            PR
                  RI
                       SC
                            SD
                                  TN
                                       TX
                                            UT
                                                  VA
                                                       VT
                                                            WA
                                                                  WI
                                                                       WV
                                                                             WY
##
     [1,] 176
                 91 539 394
                               795 2650 344 1238
                                                     309
                                                          732
                                                                 898
                                                                      859
                                                                           195
city_count_data <- as.data.frame(table(cities$state_code))</pre>
names(city count data) <- c("state code", "city count")</pre>
Part c. Function that counts occurances of a letter in a string
count.occurances <- function(string, letter){</pre>
  #Count the occurance of a letter from a given string
  #Args:
  # string: the string from which the letter will be calculated from
  # letter: the letter whose occurances in the string will be calculated
  #Returns:
  # the number of occurances of the letter in the string
  # Split the string into a vector of characters
  char.vect <- strsplit(string,split = NULL)[[1]]</pre>
  # Ensure lower case is across the two variables
  char.vect <- tolower(char.vect)</pre>
  letter <- tolower(letter)</pre>
 return (sum(char.vect==letter))
}
letter_count <- data.frame(matrix(NA,nrow=51,ncol=26))</pre>
for (i in 1:51){
 letter count[i,] <- sapply(LETTERS,count.occurances,string=states$state name[i])</pre>
names(letter_count) <- LETTERS</pre>
#Merge the information from states data.frame onto letter_count
letter_count$state_name <- states$state_name</pre>
letter_count$state_code <- states$state_code</pre>
letter_city_total <- merge(letter_count,city_count_data,by="state_code")</pre>
#Exclude "district of columbia"
#letter_city_total <- letter_city_total[letter_city_total$state_code != "DC",]</pre>
Part d.
#https://cran.r-project.org/web/packages/fiftystater/vignettes/fiftystater.html
library(ggplot2)
library(fiftystater)
#Create US map colored by city count
```

data("fifty_states") # this line is optional due to lazy data loading

```
# Set up dataset to make sure state_names match that of fifty_states
letter_city_total$state_code <- tolower(letter_city_total$state_code)</pre>
letter_city_total$state_name <- tolower(letter_city_total$state_name)</pre>
# Color for the map
#(reference: https://medium.com/@NickDoesData/visualizing-geographic-data-in-r-fb2e0f5b59c5)
low_color='#ccdbe5'
high_color="#114365"
legend_title = 'City counts'
# US Maps colored with city counts
# map_id creates the aesthetic mapping to the state name column in your data
p <- ggplot(letter_city_total, aes(map_id = state_name))</pre>
# map points to the fifty_states shape data
p <- p + geom_map(aes(fill = city_count), map = fifty_states)</pre>
p <- p + expand_limits(x = fifty_states$long, y = fifty_states$lat) + coord_map()
p <- p + scale_x_continuous(breaks = NULL) + scale_y_continuous(breaks = NULL)</pre>
#Set gradient color for city counts
p <- p + scale_fill_continuous(low = low_color, high = high_color, guide = guide_colorbar(title = legen
p <- p + labs(x = "", y = "") + theme(legend.position = "bottom", panel.background = element_blank())
```

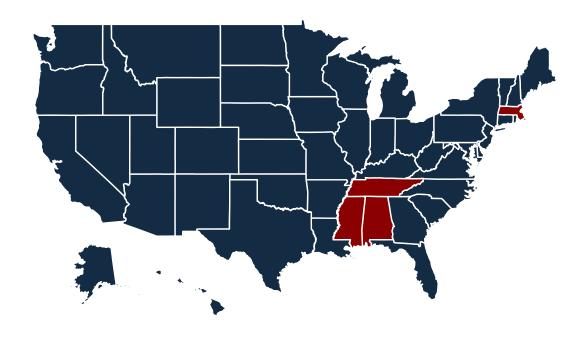




```
#Create a new variable signaled state_name with more than 3 occurances of any letter
letter_city_total$more_than_3 <- c()</pre>
for (i in 1:51) {
  letter_city_total$more_than_3[i] <- ifelse(sum(letter_city_total[i,2:27]>3) >=1,1,0)
  if (sum(letter_city_total[i,2:27]>3))
    print(letter_city_total[i,"state_name"])
}
## [1] "alabama"
## [1] "massachusetts"
## [1] "mississippi"
## [1] "tennessee"
# US Maps highlighted by states that have more than 3 occurrences of any letter in their name
high color <- "darkred"
legend_title <- "State with name having 3 or more occurances of some letter"</pre>
# map id creates the aesthetic mapping to the state name column in your data
letter_city_total$id <- letter_city_total$state_name</pre>
p <- ggplot(letter_city_total, aes(map_id = state_name))</pre>
# map points to the fifty_states shape data
p <- p + geom_map(aes(fill = more_than_3), map = fifty_states,color = "#ffffff")</pre>
p <- p + expand_limits(x = fifty_states$long, y = fifty_states$lat) + coord_map()</pre>
```

```
# ATTEMPT TO ADD STATE ABBREVIATION - NOT WORKING!
# letter_city_total$id <- letter_city_total$state_name
# p <- p + geom_text(data = fifty_states %>%
# group_by(id) %>%
# summarise(lat = mean(c(max(lat), min(lat))),
# long = mean(c(max(long), min(long)))) %>%
# mutate(state = id) %>%
# left_join(letter_city_total, by = "id"), aes(x = long, y = lat, label = state_code ))

p <- p + scale_x_continuous(breaks = NULL) + scale_y_continuous(breaks = NULL)
#Set gradient color for city counts
p <- p + scale_fill_continuous(high = high_color, guide = guide_colorbar(title = legend_title),labels = p <- p + labs(x = "", y = "") + theme(legend.position = "bottom", panel.background = element_blank())
p</pre>
```



State with name having 3 or more occurances of some letter No Yes