# Homework 5

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### Problem 3

A good figure is dependent on your medium, audience and itent, but in general I think a good figure is one that shows information that would be otherwise hard to communicate and communicates it efficiently. If you are making a figure to use for yourself, the figure just has to show the relationship you are looking for, wheras if you are making it for a more general audience, it needs to be colored nicely (in a way that makes the figure easier to understand), have titles, captions and labels and be simple enough that people who may not know a lot about the topic can make use of it. Additionally, if the figure is in a paper or online, it can be more complex than if you are putting it in a presentation where your audience sees it for less time and therefore it has to be simpler so it can be understood quickly. The figure should attempt to convey information clearly and objectively

### Problem 4

```
#Initialize function that accepts arguments
#for your data and the ability to define what counts as a success
propSuccess <- function(data, success = 1){</pre>
 counter <- length(data) #determine total number of elements in data set
 num <- 0 #counter to record the number of successes
 for(i in 1:counter){ #checks each element in the data set
    #to see if it is success and updates num if it is
   if(data[i] == success){
    num <- num+1
   }
 }
 prop <- num/counter #calculate proportion of</pre>
                     #successes by dividing successes by total
 return(prop)
# Set seed for generating random flip values
   set.seed(12345)
   # make a matrix of 10 observations for 10
    # different degrees of fairness between .3 and .4
   P4b_data <- matrix(rbinom(10, 1, prob = (30:40)/100), nrow = 10, ncol = 10)
# Use apply to find proportions of success for each column and each row
resultsCols <- apply(P4b_data, 2, propSuccess)
resultsRows <- apply(P4b_data, 1, propSuccess)</pre>
resultsCols
```

**##** [1] 1 1 1 1 0 0 0 0 1 1

resultsRows

Each column is exactly identical with 6/10 entries being a 1, while each row is either all 1s or all 0s. The problem is that it is only generating one set of 10 random numbers and then copying that set of numbers for each of the columns in the matrix, or maybe it is generating each time from the same seed start point so its generating the same random numbers each time

```
makeTenFlips <- function(probability){</pre>
  output <- rbinom(10, 1, prob = probability)</pre>
}
flipProbs <- (30:40)/100
flipMat \leftarrow matrix(rep(0, 100), nrow = 10, ncol = 10)
flipMat <- replicate(10, makeTenFlips(flipProbs))</pre>
resultsCols <- apply(flipMat, 2, propSuccess)</pre>
resultsRows <- apply(flipMat, 1, propSuccess)</pre>
outtable <- data.frame(rbind(resultsCols, resultsRows))</pre>
rownames(outtable) <- c("Column Proportions", "Row Proportions")</pre>
colnames(outtable) <- c(1:10)</pre>
outtable
##
                                   3
                                            5
## Column Proportions 0.2 0.3 0.4 0.4 0.4 0.6 0.2 0.3 0.5 0.3
## Row Proportions
                        0.6 0.2 0.5 0.4 0.3 0.1 0.7 0.4 0.1 0.3
```

### Problem 5

```
p5dat <- readLines("https://www2.isye.gatech.edu/~jeffwu/book/data/starch.dat")
p5elements <- strsplit(p5dat, split = " ")
starch <- vector()</pre>
strength <- vector()</pre>
thickness <- vector()</pre>
position <- 1
for(i in 2:length(p5elements)){
  if(length(p5elements[[i]])<=3){</pre>
  starch[position] <- p5elements[[i]][1]</pre>
  strength[position] <- p5elements[[i]][2]</pre>
  thickness[position] <- p5elements[[i]][3]</pre>
  }
  else{
  starch[position] <- p5elements[[i]][1]</pre>
  strength[position] <- p5elements[[i]][length(p5elements[[i]])-1]</pre>
  thickness[position] <- p5elements[[i]][length(p5elements[[i]])]</pre>
  }
  position <- position+1</pre>
}
strength <- as.numeric(strength)</pre>
thickness <- as.numeric(thickness)</pre>
p5table <- data.frame(starch, strength, thickness)
colnames(p5table) <- c(p5elements[[1]][1:3])</pre>
p5table
```

```
## starch strength thickness
## 1 CA 791.7 7.7
```

```
## 3
                 710.0
                              8.6
           CA
## 4
           CA
                 940.7
                             11.8
## 5
                 990.0
                             12.4
           CA
## 6
           CA
                 916.2
                             12.0
## 7
                 835.0
           CA
                             11.4
## 8
                 724.3
                             10.4
           CA
## 9
                 611.1
                              9.2
           CA
## 10
           CA
                 621.7
                              9.0
## 11
                              9.5
           CA
                 735.4
## 12
           CA
                 990.0
                             12.5
## 13
           CA
                 862.7
                             11.7
## 14
                              8.0
           CO
                 731.0
## 15
                 710.0
           CO
                              7.3
## 16
           CO
                 604.7
                              7.2
## 17
           CO
                 508.8
                              6.1
## 18
           CO
                 393.0
                              6.4
## 19
           CO
                 416.0
                              6.4
## 20
           CO
                 400.0
                              6.9
## 21
           CO
                 335.6
                              5.8
## 22
           CO
                 306.4
                              5.3
## 23
           CO
                 426.0
                              6.7
## 24
           CO
                 382.5
                              5.8
## 25
           CO
                 340.8
                              5.7
## 26
           CO
                 436.7
                              6.1
## 27
           CO
                 333.3
                              6.2
## 28
           CO
                 382.3
                              6.3
## 29
           CO
                 397.7
                              6.0
## 30
           CO
                 619.1
                              6.8
## 31
                 857.3
                              7.9
           CO
## 32
           CO
                 592.5
                              7.2
## 33
           PO
                 983.3
                             13.0
## 34
           P0
                 958.8
                             13.3
## 35
           PO
                 747.8
                             10.7
## 36
           PO
                 866.0
                             12.2
## 37
          PO
                 810.8
                             11.6
## 38
           PO
                 950.0
                              9.7
## 39
           PO
                1282.0
                             10.8
## 40
           PO
                1233.8
                             10.1
## 41
           P0
                1660.0
                             12.7
## 42
           PO
                 746.0
                              9.8
## 43
           PO
                 650.0
                             10.0
## 44
                 992.5
                             13.8
           PO
## 45
                 896.7
           PO
                             13.3
## 46
           PO
                 873.9
                             12.4
## 47
                 924.4
           PO
                             12.2
## 48
           PO
                1050.0
                             14.1
## 49
           PO
                 973.3
                             13.7
# create mat for scatter plots
layMat \leftarrow matrix(c(1:3, 0, 4, 0), ncol = 3, nrow = 2, byrow = TRUE)
layout(layMat)
# make vectors with just the data from one of each starch type
```

## 2

610.0

CA

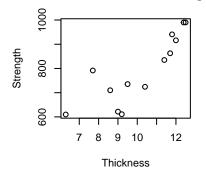
6.3

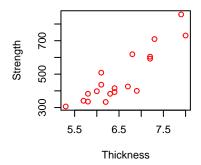
```
CA <- p5table[which(p5table[,1] == "CA"), 2:3]
CO <- p5table[which(p5table[,1]== "CO"), 2:3]</pre>
PO <- p5table[which(p5table[,1] == "PO"), 2:3]
# Make scatter plots of each factor and one graph with
# them all combined but color coded with legend
plot(CA$thickness, CA$strength, main = "Canna thickness vs strength",
       xlab = "Thickness", ylab = "Strength", col = 1)
plot(CO$thickness, CO$strength, main = "Corn thickness vs strength",
       xlab = "Thickness", ylab = "Strength", col = 2)
plot(PO$thickness, PO$strength, main = "Potato thickness vs strength",
       xlab = "Thickness", ylab = "Strength", col = 3)
plot(thickness, strength, col = p5table$starch, main = "Thickness vs Strength",
       sub = "All starch types", xlab = "Thickness", ylab = "Strength",
       xlim = c(4,14), ylim = c(300,1700)
legend("topleft", legend = c("Canna", "Corn", "Potato"),
         col = 1:3, pch = 1, cex = .8)
```

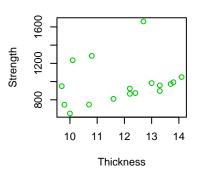
### Canna thickness vs strength

### Corn thickness vs strength

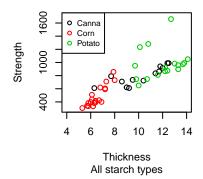
### Potato thickness vs strength



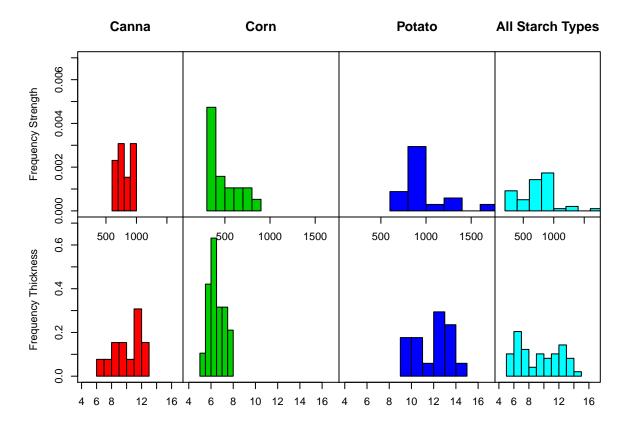




### Thickness vs Strength



```
c(4,0,0,0), c(4,0,0,0), c(4,0,0,4))
mylist <- list(CA[,1], CO[,1], PO[,1], strength, CA[,2], CO[,2], PO[,2], thickness)
# Make layout mat
layout((matrix(c(1:8), ncol=4, nrow = 2, byrow = \frac{TRUE}{}), widths = c(rep(2.5/10, 4)),
         heights = c(5/10, 5/10))
# For loops make each histogram and add to layout mat
counter <- 1</pre>
for(i in 1:4){ #top row of mat
  par(mar = margins[,i])
  if(i == 1){
  hist(mylist[[i]], main = dataTitle[i], ylab = datax[i], col = i+1, freq = FALSE,
         xlab = NULL, ylim = c(0,.007), xlim = c(100,1700))
  }
  else{
    hist(mylist[[i]], main = dataTitle[i], ylab = datax[i], col = i+1, freq = FALSE,
           xlim = c(100,1700), xlab = NULL, yaxt = "n", ylim = c(0,.007))
  }
  box()
  counter <<- counter + index[i]</pre>
for(i in 5:8){ # bottom row of mat
  par(mar = margins[,i])
  if(i == 5){
  hist(mylist[[i]], main = NULL, col = i-3, ylab = "Frequency Thickness", freq = FALSE,
         ylim = c(0,.7), xlab = NULL, xlim = c(4,17))
  }
  else{
    hist(mylist[[i]], yaxt = "n", ann = FALSE, ylab = datax[i], col = (i-3),
           xlim = c(4,17), freq = FALSE, main = NULL, ylim = c(0,.7))
  }
  box()
  counter <- counter + index[i]</pre>
}
```



```
# Function that takes a data set and outputs a table of
# summary statistics for exploratory data analysis
dataSummary <- function(dataset, v1, v2, ...){</pre>
dataOut <- rbind.data.frame(summary(dataset[,2]), summary(dataset[,3]))</pre>
sumNames <- c("Min", "Q1", "Med", "Mean", "Q3", "Max")</pre>
rown <- c(paste("Overall", v1), paste("Overall", v2))</pre>
factors <- unique(dataset[,1])</pre>
headers <- colnames(dataset)</pre>
# Find summary information by factor and add
# to data frame
for(i in 1:length(factors)){
  dataOut <- rbind.data.frame(dataOut, summary</pre>
                  (dataset[which(dataset[,1] == factors[i]), v1]),
                   summary(dataset[which(dataset[,1] == factors[i]), v2]))
  rown <- c(rown, paste(factors[i], v1), paste(factors[i], v2))</pre>
}
# Add names to data table
colnames(dataOut) <- sumNames</pre>
rownames(dataOut) <- rown</pre>
# Calculate IQRs and Ranges and add to data summary table
dataIQR <- dataOut[,"Q3"]-dataOut[,"Q1"]</pre>
dataRange <- dataOut[,"Max"]-dataOut[,"Min"]</pre>
```

```
dataOut <- cbind.data.frame(dataOut, dataIQR, dataRange)</pre>
# Initialize sd vector and calculate standard deviation within each factor for v1 and v2
dataSD <- c(sd(dataset[,2]), sd(dataset[,3]))</pre>
for(i in 1:length(factors)){
  dataSD <- c(dataSD, sd(dataset[which(dataset[,1] == factors[i]), v1]),</pre>
              sd(dataset[which(dataset[,1] == factors[i]), v2]))
# Initialize variance vector and calculate variance within each factor for v1 and v2
dataVar <- c(var(dataset[,2]), var(dataset[,3]))</pre>
for(i in 1:length(factors)){
  dataVar <- c(dataVar, var(dataset[which(dataset[,1] == factors[i]), v1]),</pre>
               var(dataset[which(dataset[,1] == factors[i]), v2]))
}
# add remaining columns and column names and print summary data
dataOut <- cbind.data.frame(dataOut, dataVar, dataSD)</pre>
sumNames <- c(sumNames, "IQR", "Range", "Variance", "Standard Deviation")
colnames(dataOut) <- sumNames</pre>
dataOut
# Create data summary of starch data
dataSummary(p5table, "strength", "thickness")
##
                                                        QЗ
                       Min
                               Q1
                                    Med
                                              Mean
                                                              Max
                                                                    IQR Range
## Overall strength 306.4 508.80 735.4 736.975510 924.40 1660.0 415.6 1353.6
                                          9.387755 12.00
## Overall thickness 5.3
                             6.70
                                    9.5
                                                             14.1
                                                                    5.3
                                                                           8 8
## CA strength
                     610.0 710.00 791.7 795.292308 916.20 990.0 206.2
                                                                         380.0
## CA thickness
                     6.3
                             9.00 10.4 10.192308 11.80
                                                             12.5
                                                                    2.8
                                                                           6.2
## CO strength
                     306.4 382.40 416.0 482.826316 598.60 857.3 216.2 550.9
                             6.05 6.4 6.531579
                                                                           2.7
## CO thickness
                     5.3
                                                     7.05
                                                              8.0
                                                                    1.0
## PO strength
                     650.0 866.00 950.0 976.429412 992.50 1660.0 126.5 1010.0
## PO thickness
                       9.7 10.70 12.2 11.964706 13.30 14.1
                                                                    2.6
                         Variance Standard Deviation
## Overall strength 7.979212e+04
                                         282.4749963
## Overall thickness 7.696930e+00
                                           2.7743342
## CA strength
                    1.933448e+04
                                         139.0484775
## CA thickness
                     3.884103e+00
                                           1.9708127
## CO strength
                     2.483643e+04
                                         157.5957693
## CO thickness
                     5.489474e-01
                                           0.7409098
## PO strength
                     5.654676e+04
                                         237.7956207
## PO thickness
                     2.291176e+00
                                           1.5136633
```

## Problem 6

```
#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_st")
```

```
unzip("us_cities_states.zip", exdir = ".")
#read in data, looks like sql dump, blah
   library(data.table)
   states <- fread(input = "./us_cities_and_states/states.sql",skip = 23,sep = "'",</pre>
                      sep2 = ",", header = F, select = c(2,4), data.table = FALSE)
states <- states[,1]
states <- states[-8]
states <- tolower(states)</pre>
   #PART B
   # Read in cities_extended and shorten by filtering out repeated city names
   cities <- fread(input = "./us_cities_and_states/cities_extended.sql",sep = "'",</pre>
                      sep2 = ",", header = F, select = c(2,4))
   citiesshort <- unique(cities)</pre>
   # Give some headers
   colnames(citiesshort) <- c("City", "State")</pre>
   #Create summary table of the number of cities included by state
   numbystate <- data.frame(table(citiesshort$State))</pre>
   numbystate < numbystate[c(-8,-40),]
```

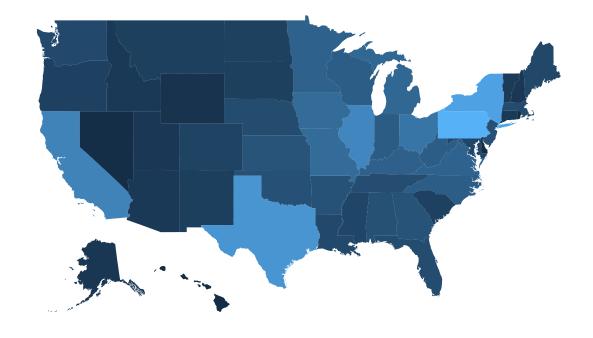
### PART C

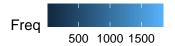
```
letter_count <- data.frame(matrix(NA,nrow=50, ncol=26))</pre>
letterCounter <- function(letter, stateName){</pre>
  temp <- strsplit(stateName, split = "")</pre>
  tf <- temp[[1]] == letter
  counter <- sum(tf)</pre>
  return(counter)
}
alphabet <- c("a", "b", "c", "d", "e", "f", "g", "h", "i", "j", "k", "l", "m",
                "n", "o", "p", "q", "r", "s", "t", "u", "v", "w", "x", "y", "z")
for(i in 1:length(states)){
  letter_count[i,] <- sapply(alphabet, letterCounter, stateName = states[i])</pre>
}
colnames(letter_count) <- alphabet</pre>
rownames(letter_count) <- states</pre>
letter_count
##
                  abcdefghijklmnopqrstuvwxyz
```

```
## delaware
## florida
        1 0 0 1 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 0 0 0 0 0
        10001020100000100100000000
## georgia
        ## hawaii
## iowa
        ## idaho
        ## illinois
        0 0 0 0 0 0 0 0 3 0 0 2 0 1 1 0 0 0 1 0 0 0 0 0 0
        ## indiana
## kansas
        20000000000100100002000000
        ## kentucky
## louisiana
        2000000000001011000010100000
        2 0 1 0 1 0 0 1 0 0 0 0 1 0 0 0 0 4 2 1 0 0 0 0
## massachusetts
        20010000001111000100000010
## maryland
## maine
        ## michigan
        ## minnesota
        ## missouri
        ## mississippi
        2000000000001210000100000
## montana
## north dakota
        2001000100100120010200000
## nebraska
        2 1 0 0 1 0 0 0 0 0 1 0 0 1 0 0 0 1 1 0 0 0 0 0 0
## new hampshire 1 0 0 0 2 0 0 2 1 0 0 0 1 1 0 1 0 1 1 0 0 0 1 0 0
        0 0 0 0 3 0 0 0 0 1 0 0 0 1 0 0 0 1 1 0 0 0 1 0 1 0
## new jersey
        ## new mexico
## nevada
        ## new york
        ## ohio
        ## oklahoma
        ## oregon
        2 0 0 0 1 0 0 0 1 0 0 1 0 3 0 1 0 0 1 0 0 1 0 0 1 0
## pennsylvania
## rhode island
        ## south dakota
        ## tennessee
## texas
        10001000000000000011000100
## utah
        10000001000000000001100000
## virginia
        10000010300001000100010000
## vermont
        1 0 0 0 0 0 1 1 1 0 0 0 0 2 1 0 0 0 1 1 0 0 1 0 0
## washington
## wisconsin
        ## west virginia 1 0 0 0 1 0 1 0 3 0 0 0 0 1 0 0 0 1 1 1 0 1 1 0 0 0
        0\; 0\; 0\; 0\; 0\; 0\; 1\; 0\; 1\; 0\; 0\; 0\; 1\; 1\; 1\; 0\; 0\; 0\; 0\; 0\; 0\; 0\; 1\; 0\; 1\; 0
## wyoming
tflc <- letter_count >= 3
Three.Or.More <- apply(tflc, 1, sum)
statesum <- cbind(states, Three.Or.More)</pre>
```

#### Map 1

```
#https://cran.r-project.org/web/packages/fiftystater/vignettes/fiftystater.html
library(ggplot2)
library(fiftystater)
```





### ### Map 2

```
mystates <- data.frame(state = states, statesum)
# map_id creates the aesthetic mapping to the state name column in your data
q <- ggplot(mystates, aes(map_id = state)) +
# map points to the fifty_states shape data
geom_map(aes(fill = Three.Or.More), map = fifty_states) +
expand_limits(x = fifty_states$long, y = fifty_states$lat) +
coord_map() +
scale_x_continuous(breaks = NULL) +</pre>
```

