

Assignment 3:

Problem 1:

#Step 1: Load packages you will use in server script, including shiny package.

```
library(tm)
library(wordcloud)
library(quantmod)
library(RMySQL)
library(hunspell)
library(e1071)
```

Result preparation - Method 1: outside server part.

Use the same text data from Lecture 9.

You can choose to do your analysis outside "server" part and use the result in the server part

Load libraries

.....

Problem 1

.....

load driver for SQL

```
drv <- dbDriver("MySQL")
```

```
con <- dbConnect(drv, user = "gang", password = "gang",
  host = "localhost", dbname = "fe513_twitter")
```

#checking table

```
dbListTables(con)
```

```
# Step 1 - creating a data frame for unique user_ids
```

```
SQLtext1 <- "SELECT DISTINCT(user_id) FROM twitter_message"
```

```
unique_userid <- dbSendQuery(con, SQLtext1)
```

```
unique_users <- dbFetch(unique_userid)
```

```
unique_users
```

```
# Step 2 - creating a data frame for random user_ids
```

```
random_users <- data.frame(unique_users[sample(nrow(unique_users), 3, replace = FALSE, prob = NULL),])
```

```
random_users
```

```
names(random_users)[1] <- paste("user_id")
```

```
random_users
```

```
random_users[1,1]
```

```
random_users[2,1]
```

```
random_users[3,1]
```

```
dbWriteTable(con, "random_users", random_users, overwrite = TRUE)
```

```
dbReadTable(con, "random_users")
```

```
# step 3 - create query to extract all of those user tweets
```

```
SQLtext2 <- "SELECT * FROM twitter_message where user_id IN (SELECT user_id FROM random_users)"
```

```
r_user_query <- dbSendQuery(con, SQLtext2)
```

```
r_user_tweets <- dbFetch(r_user_query)
```

step 4 - combine tweets from the same user into one variable. As a result, you will have 3 variables(long strings) in R.

```
user1 <- r_user_tweets[which(r_user_tweets$user_id == random_users[1,1]), ]
```

```
user2 <- r_user_tweets[which(r_user_tweets$user_id == random_users[2,1]), ]
```

```
user3 <- r_user_tweets[which(r_user_tweets$user_id == random_users[3,1]), ]
```

```
combined_user1_tweets <- paste(user1$tweets, collapse = " ")
```

```
combined_user2_tweets <- paste(user2$tweets, collapse = " ")
```

```
combined_user3_tweets <- paste(user3$tweets, collapse = " ")
```

step 5 - Creating list of all non-english words from each set and removing non-ASCII characters

```
combined_user1_tweets <- iconv(combined_user1_tweets, "latin1", "ASCII", sub="")
```

```
combined_user2_tweets <- iconv(combined_user2_tweets, "latin1", "ASCII", sub="")
```

```
combined_user3_tweets <- iconv(combined_user3_tweets, "latin1", "ASCII", sub="")
```

```
u1_nonen <- unlist(hunspell_find(combined_user1_tweets))
```

```
u2_nonen <- unlist(hunspell_find(combined_user2_tweets))
```

```
u3_nonen <- unlist(hunspell_find(combined_user3_tweets))
```

```
nonenglishset <- c(u1_nonen, u2_nonen, u3_nonen)
```

step 5.1 - remove all punctuations

```
ut <- Corpus(VectorSource(c(combined_user1_tweets, combined_user2_tweets,  
combined_user3_tweets)))
```

```
utnopunc <- tm_map(ut, removePunctuation)
```

```
uttolower <- tm_map(utnopunc, content_transformer(tolower))
```

```
# step 5.2 - remove all numbers
```

```
utnonum <- tm_map(uttolower, removeNumbers)
```

```
# step 5.3 - remove stop words
```

```
utnostop <- tm_map(utnonum, removeWords, stopwords("en"))
```

```
# step 5.4 - remove non-english words
```

```
utnoeng <- tm_map(utnostop, removeWords, nonenglishset)
```

```
# step 6 - Create the term-document matrix and plot the wordclouds
```

```
TM1 <- TermDocumentMatrix(utnoeng)
```

```
M1 <- as.matrix(TM1)
```

```
final <- data.frame(word = rownames(M1), freq1=M1[,1], freq2 = M1[,2], freq3 = M1[,3])
```

```
par(mfrow = c(1, 3))
```

```
wordcloud(words = final$word, freq = final$freq1, min.freq = 2, colors=brewer.pal(8, "Dark2"))
```

```
wordcloud(words = final$word, freq = final$freq2, min.freq = 2, colors=brewer.pal(8, "Dark2"))
```

```
wordcloud(words = final$word, freq = final$freq3, min.freq = 2, colors=brewer.pal(8, "Dark2"))
```



```
# labels in total.
```

```
cleaned_user1_tweets <- iconv(r_user_tweets$tweets, "latin1", "ASCII", sub="")
```

```
PS2_cleaned_table <- cbind(r_user_tweets$user_id, cleaned_user1_tweets)
```

```
PS2 <- as.data.frame(PS2_cleaned_table)
```

```
print.sum <- summary(PS2)
```

```
# step 5.1 - remove all punctuations
```

```
utps2 <- Corpus(VectorSource(c(cleaned_user1_tweets)))
```

```
utnopuncps2 <- tm_map(utps2, removePunctuation)
```

```
uttolowerps2 <- tm_map(utnopuncps2, content_transformer(tolower))
```

```
# step 5.2 - remove all numbers
```

```
utnonumps2 <- tm_map(uttolowerps2, removeNumbers)
```

```
# step 5.3 - remove stop words
```

```
utnostopps2 <- tm_map(uttolowerps2, removeWords, stopwords("en"))
```

```
# step 5.4 - remove non-english words
```

```
utnoengps2 <- tm_map(utnostopps2, removeWords, ps2_nonen)
```

```
# step 6 - Create the term-document matrix
```

```
TMPS2 <- DocumentTermMatrix(utnoengps2)
```

```
M2 <- as.matrix(TMPS2)
```

```
ncol(M2)
```

```
# DFM2 <- as.data.frame((M2))
```

```
# DFM2[,1][DFM2[,1] == "1"] <- random_users[1,1]
# DFM2[,2][DFM2[,2] == "1"] <- random_users[2,1]
# DFM2[,3][DFM2[,3] == "1"] <- random_users[3,1]
```

```
# Step 7 - run kmeans clustering
```

```
kmRes <- kmeans(M2, 3, nstart = 20)
```

```
# Step 8 - use table() function in R showing the difference between cluster results and user id label.
```

```
ktable <- table(res = kmRes$cluster, real = PS2$V1)
```

```
      real
res 20659892 241205643 39221596
  1         84         42         20
  2          0          4          0
  3          0          2          0
```

```
#returns cluster label
```

```
cmRes <- cmeans(M2, centers = 3, iter.max = 100)
```

```
head(cmRes$cluster)
```

```
1 2 3 4 5 6
2 2 2 2 2 2
```

```
#return membership (the probability of one data point belongs to one group. )
```

```
#the cluster label is based on max(membership)
```

```
head(cmRes$membership)
```

```
      1      2      3
1 0.3333251 0.3333495 0.3333254
2 0.3333333 0.3333334 0.3333333
3 0.3333321 0.3333358 0.3333321
4 0.3333330 0.3333339 0.3333330
5 0.3333333 0.3333335 0.3333333
6 0.3333329 0.3333342 0.3333329
```

```
# Step 7 - run hierarchical cluster
```

```
d <- dist(M2, method = "euclidean")
```

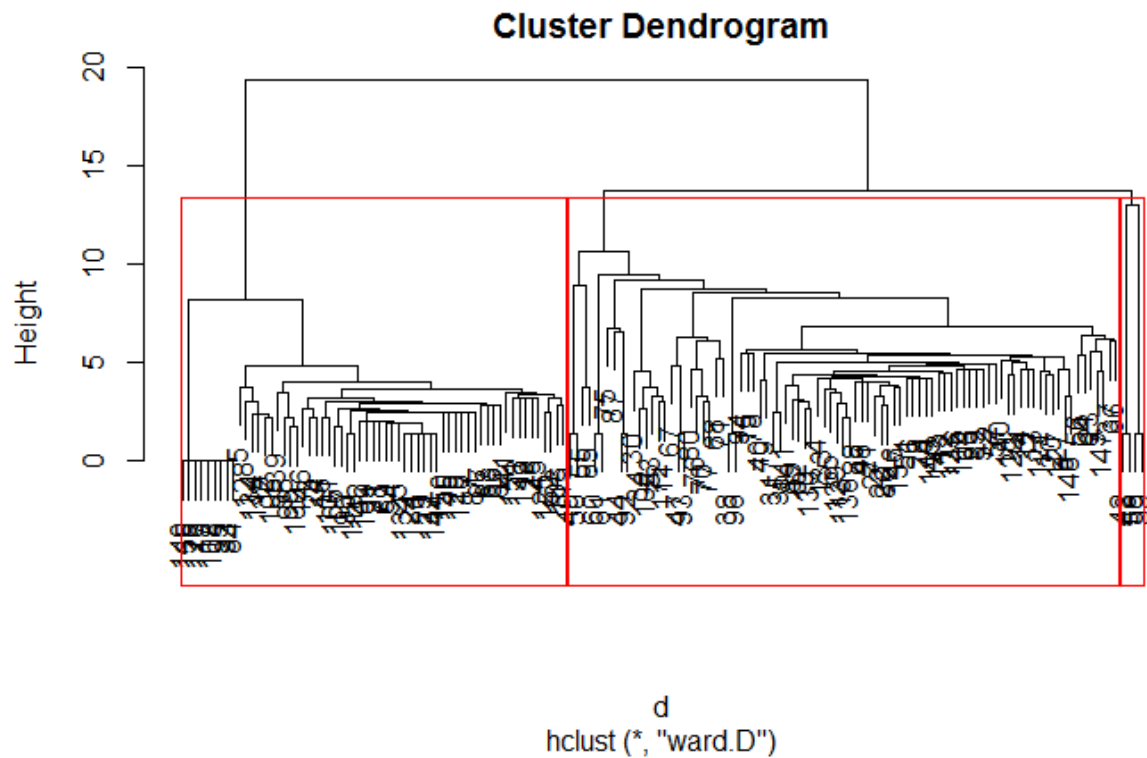
```
HclustResult <- hclust(d, method="ward.D")
```

```
plot(HclustResult)
```

```
groups <- cutree(HclustResult, k=3) # cut tree into n clusters
```

```
# draw dendrogram with red borders around the n clusters
```

```
rect.hclust(HclustResult, k=3, border="red")
```



```
# Step 8 - use table() function in R showing the difference between cluster results and user id label.
```

```
htable <- table(res = groups, real = PS2$V1)
```

	real			
res	20659892	241205643	39221596	
1	53	6	2	
2	31	38	18	
3	0	4	0	

Problem 3 Screenshots and Code:

☒ wordcloud color?

Wordcloud min frequency

2

Wordcloud min frequency

1

Wordcloud min frequency

2

[1] 18949452 102285890 592001965

☒ wordcloud color?

Wordcloud min frequency

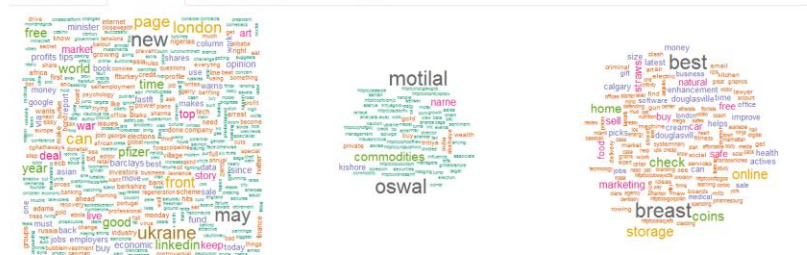
2

Wordcloud min frequency

1

Wordcloud min frequency

2



FE513 (Name: Assignment 3 App)

☒ wordcloud color?

Wordcloud min frequency

2

Wordcloud min frequency

1

Wordcloud min frequency

2

The widget on the side panel controls the word clouds. This page loads slowly due to the processing of words in the database.

E1.SQL and Names E1.Wordcloud E2.Summary E2.Results E2.Compare

[1] "102285890:10"

[1] "18949452:371"

[1] "592001965:119"

V1 cleaned_user1_tweets

18949452 Asian economies now need more debt to generate growth: <http://t.co/CToQpYXSH0> <http://t.co/de6LhZVOLF>

18949452 A Swift way to curb Putin's ambitions <http://t.co/RNzH1H5tcc> <http://t.co/feQFrEPiWP>

18949452 Ouch. "Google has helped give the world attention deficit disorder while destroying privacy," writes @LukeJohnsonRCP <http://t.co/6NqXdxWoOJ>

18949452 Free to read: What the Google privacy ruling means for the internet? Can it even be enforced? <http://t.co/33OLgBF1F>

18949452 Bob Diamond's Africa vehicle raises money for second time in less than 6 months: <http://t.co/Hir7unASDK>

18949452 Can Asia overcome its addiction to debt? <http://t.co/Zo4625O4y> <http://t.co/K7XXvwbCqp>

18949452 Sneaky rhetoric: The art of saying something without appearing to say it <http://t.co/yVOKdnSSRq>

592001965 Phytoceramides supplements have been called your "secret to cheat your age" by Dr. Oz <http://t.co/34uX9jeQje> -> <https://t.co/VxNagGP1ku>

592001965 \$205,000 - Miami, FL Condo For Sale - 133 NE 2nd Ave -> <http://t.co/kDgCNKV9ldn>

592001965 \$390,000 - Aventura, FL Condo For Sale - 3500 MYSTIC POINTE DR -> <http://t.co/5yeapRbJ>

592001965 Body Wraps -> <https://t.co/oubLff8xY2>

592001965 find best electric percolators, choose the best electric percolators -> <http://t.co/QMTOQwF9U>

592001965 Vemma - The Real Truth -> <http://t.co/MSIUDFyNt5>

592001965 Professional social media services buy facebook likes -> <http://t.co/IOUq4YoV7w6>

592001965 Check out this exclusive review of the Article Factory Pro software by Josh Zamora -> <https://t.co/Pe5HlR2Baz>

18949452 Turkish mine death toll at 157, says mayor <http://t.co/09fmTLDeAS>

18949452 UN Syria envoy Brahimi to step down <http://t.co/LCHQDlRyJ9q>

18949452 Front page of the Financial Times US for Wednesday, May 14 <http://t.co/BRPzDuQpnH>

18949452 "For those who find the commercial world draining, I suggest the intellectual succour of the Stoics": <http://t.co/bx3QEJK9Er>

18949452 Saudi Arabia moves to ease Iran tensions <http://t.co/Zmw0IO8ByH>

18949452 US steel imports surge on shale boom <http://t.co/hqQ918Z5UI>

18949452 Goldmans Twitter banker joins hedge fund <http://t.co/es135xGSHS>

FE513 (Name: Assignment 3 App)

☒ wordcloud color?

Wordcloud min frequency

2

Wordcloud min frequency

1

Wordcloud min frequency

2

The widget on the side panel controls the word clouds. This page loads slowly due to the processing of words in the database.

E1.SQL and Names E1.Wordcloud E2.Summary E2.Results E2.Compare

The information below is the size of the clusters.

[1] 494 1 5

The information below indicates variance between clusters, within clusters, and of clusters.

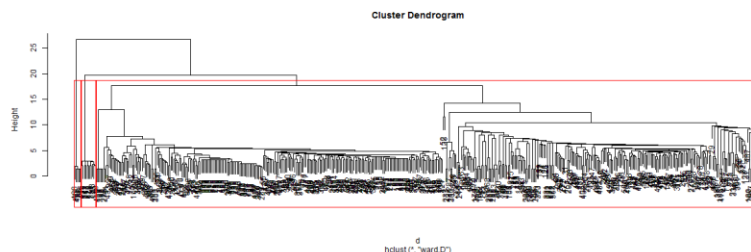
[1] 170.6798

[1] 4381.666

[1] 4206.186 0.000 4.800

[1] 4210.986

Call: hclust(d = d, method = "ward.D") Cluster method: ward.D Distance: euclidean Number of objects: 500



FE513 (Name: Assignment 3 App)

☒ wordcloud color?

Wordcloud min frequency

2

Wordcloud min frequency

1

Wordcloud min frequency

2

The widget on the side panel controls the word clouds. This page loads slowly due to the processing of words in the database.

[E1.SQL and Names](#) [E1.Wordcloud](#) [E2.Summary](#) [E2.Results](#) [E2.Compare](#)

The table below is for K means.

res	real	Freq
1	102285890	10
2	102285890	0
3	102285890	0
1	18949452	371
2	18949452	0
3	18949452	0
1	592001965	113
2	592001965	1
3	592001965	5

The table below is for Hierarchical Clustering.

res	real	Freq
1	102285890	10
2	102285890	0
3	102285890	0
1	18949452	360
2	18949452	11
3	18949452	0
1	592001965	114
2	592001965	0
3	592001965	5

Server File:

#Step 1: Load packages you will use in server script, including shiny package.

```
library(tm)
```

```
library(wordcloud)
```

```
library(quantmod)
```

```
library(RMySQL)
```

```
library(hunspell)
```

```
library(e1071)
```

Result preparation - Method 1: outside server part.

Use the same text data from Lecture 9.

You can choose to do your analysis outside "server" part and use the result in the server part

Load libraries

```

# ~~~~~
# Problem 1
# ~~~~~

# load driver for SQL
drv <- dbDriver("MySQL")

con <- dbConnect(drv, user = "gang", password = "gang",
                 host = "localhost", dbname = "fe513_twitter")

#checking table

dbListTables(con)

# Step 1 - creating a data frame for unique user_ids
SQLtext1 <- "SELECT DISTINCT(user_id) FROM twitter_message"

unique_userid <- dbSendQuery(con, SQLtext1)

unique_users <- dbFetch(unique_userid)
unique_users

# Step 2 - creating a data frame for random user_ids
random_users <- data.frame(unique_users[sample(nrow(unique_users), 3, replace = FALSE, prob =
NULL),])

random_users
names(random_users)[1] <- paste("user_id")
random_users

```

```
random_users[1,1]
```

```
random_users[2,1]
```

```
random_users[3,1]
```

```
dbWriteTable(con, "random_users", random_users, overwrite = TRUE)
```

```
dbReadTable(con, "random_users")
```

```
# step 3 - create query to extract all of those user tweets
```

```
SQLtext2 <- "SELECT * FROM twitter_message where user_id IN (SELECT user_id FROM random_users)"
```

```
r_user_query <- dbSendQuery(con, SQLtext2)
```

```
r_user_tweets <- dbFetch(r_user_query)
```

```
# step 4 - combine tweets from the same user into one variable. As a result, you will have 3  
variables(long strings) in R.
```

```
user1 <- r_user_tweets[which(r_user_tweets$user_id == random_users[1,1]), ]
```

```
user2 <- r_user_tweets[which(r_user_tweets$user_id == random_users[2,1]), ]
```

```
user3 <- r_user_tweets[which(r_user_tweets$user_id == random_users[3,1]), ]
```

```
combined_user1_tweets <- paste(user1$tweets, collapse = " ")
```

```
combined_user2_tweets <- paste(user2$tweets, collapse = " ")
```

```
combined_user3_tweets <- paste(user3$tweets, collapse = " ")
```

```
# step 5 - Creating list of all non-english words from each set and removing non-ASCII characters
```

```
combined_user1_tweets <- iconv(combined_user1_tweets, "latin1", "ASCII", sub="")
```

```
combined_user2_tweets <- iconv(combined_user2_tweets, "latin1", "ASCII", sub="")
```

```
combined_user3_tweets <- iconv(combined_user3_tweets, "latin1", "ASCII", sub="")
```

```
u1_nonen <- unlist(hunspell_find(combined_user1_tweets))
```

```
u2_nonen <- unlist(hunspell_find(combined_user2_tweets))
```

```
u3_nonen <- unlist(hunspell_find(combined_user3_tweets))
```

```
nonenglishset <- c(u1_nonen, u2_nonen, u3_nonen)
```

```
# step 5.1 - remove all punctuations
```

```
ut <- Corpus(VectorSource(c(combined_user1_tweets, combined_user2_tweets,  
combined_user3_tweets)))
```

```
utnopunc <- tm_map(ut, removePunctuation)
```

```
uttolower <- tm_map(utnopunc, content_transformer(tolower))
```

```
# step 5.2 - remove all numbers
```

```
utnonum <- tm_map(uttolower, removeNumbers)
```

```
# step 5.3 - remove stop words
```

```
utnostop <- tm_map(utnonum, removeWords, stopwords("en"))
```

```
# step 5.4 - remove non-english words
```

```
utnoeng <- tm_map(utnostop, removeWords, nonenglishset)
```

```
# step 6 - Create the term-document matrix and plot the wordclouds
```

```
TM1 <- TermDocumentMatrix(utnoeng)
```

```
M1 <- as.matrix(TM1)
```

```
final <- data.frame(word = rownames(M1), freq1=M1[,1], freq2 = M1[,2], freq3 = M1[,3])
```

```
par(mfrow = c(1, 3))
```

```
wordcloud(words = final$word, freq = final$freq1, min.freq = 2, colors=brewer.pal(8, "Dark2"))
wordcloud(words = final$word, freq = final$freq2, min.freq = 2, colors=brewer.pal(8, "Dark2"))
wordcloud(words = final$word, freq = final$freq3, min.freq = 2, colors=brewer.pal(8, "Dark2"))
```

```
#.....
```

```
# Problem 2
```

```
#.....
```

```
ps2_nonen <- unlist(hunspell_find(r_user_tweets$tweets))
```

```
# step 4 - make a label(index) vector for those tweets/documents.
```

```
# The label is the user id you have. Thus, all documents posted
```

```
# by one user should have the same label, and you will have 3 unique
```

```
# labels in total.
```

```
cleaned_user1_tweets <- iconv(r_user_tweets$tweets, "latin1", "ASCII", sub="")
```

```
PS2_cleaned_table <- cbind(r_user_tweets$user_id, cleaned_user1_tweets)
```

```
PS2 <- as.data.frame(PS2_cleaned_table)
```

```
print.sum <- summary(PS2)
```

```
# step 5.1 - remove all punctuations
```

```
utps2 <- Corpus(VectorSource(c(cleaned_user1_tweets)))
```

```
utnopuncps2 <- tm_map(utps2, removePunctuation)
```

```
uttolowerps2 <- tm_map(utnopuncps2, content_transformer(tolower))
```

```
# step 5.2 - remove all numbers
```

```
utnonumps2 <- tm_map(uttolowerps2, removeNumbers)
```

```
# step 5.3 - remove stop words
```

```
utnostopps2 <- tm_map(uttolowerps2, removeWords, stopwords("en"))
```

```
# step 5.4 - remove non-english words
```

```
utnoengps2 <- tm_map(utnostopps2, removeWords, ps2_nonen)
```

```
# step 6 - Create the term-document matrix
```

```
TMPS2 <- DocumentTermMatrix(utnoengps2)
```

```
M2 <- as.matrix(TMPS2)
```

```
ncol(M2)
```

```
# DFM2 <- as.data.frame((M2))
```

```
# DFM2[,1][DFM2[,1] == "1"] <- random_users[1,1]
```

```
# DFM2[,2][DFM2[,2] == "1"] <- random_users[2,1]
```

```
# DFM2[,3][DFM2[,3] == "1"] <- random_users[3,1]
```

```
# Step 7 - run kmeans clustering
```

```
kmRes <- kmeans(M2, 3, nstart = 20)
```

```
# Step 8 - use table() function in R showing the difference between cluster results and user id label.
```

```
ktable <- table(res = kmRes$cluster, real = PS2$V1)
```

```
#returns cluster label
```

```
cmRes <- cmeans(M2, centers = 3, iter.max = 100)
```

```
head(cmRes$cluster)
```

```
#return membership (the probability of one data point belongs to one group. )
```



```
#the cluster label is based on max(membership)
head(cmRes$membership)
```

```
# Step 7 - run hierarchical cluster
```

```
d <- dist(M2, method = "euclidean")
```

```
HclustResult <- hclust(d, method="ward.D")
```

```
plot(HclustResult)
```

```
groups <- cutree(HclustResult, k=3) # cut tree into n clusters
```

```
# draw dendrogram with red borders around the n clusters
```

```
rect.hclust(HclustResult, k=3, border="red")
```

```
# Step 8 - use table() function in R showing the difference between cluster results and user id label.
```

```
htable <- table(res = groups, real = PS2$V1)
```

```
# Step 2: Make server script.
```

```
# Analyze data/model, make output and plots.
```

```
# The outputs needs to be defined in render functions.
```

```
# If you have selection in UI, we need reactive function.
```

```
shinyServer(function(input, output) {
```

```
  # Tab Results
```

```
  output$kmmeans <- renderPrint({
```

```
    kmRes$size
```

```
  })
```

```
  output$ktotss <- renderPrint({
```

```
kmRes$totss
```

```
}}
```

```
output$kwwith <- renderPrint({
```

```
  kmRes$withinss
```

```
}}
```

```
output$ktotwith <- renderPrint({
```

```
  kmRes$tot.withinss
```

```
}}
```

```
output$kbet <- renderPrint({
```

```
  kmRes$betweenss
```

```
}}
```

```
output$Hclust <- renderPrint({
```

```
  HclustResult
```

```
}}
```

```
output$Hplot <- renderPlot({
```

```
  plot(HclustResult)
```

```
  groups <- cutree(HclustResult, k=3)
```

```
  rect.hclust(HclustResult, k=3, border="red")
```

```
}}
```

```
# Tab Table Comparisons
```

```
output$htable <- renderTable({
```

```
  ktable
```

```
}}
```

```

output$htable <- renderTable({
  htable
})

# Tab 2 - wordcloud

# We choose the doc and color in UI, and pass them to p_col and freq_column

# p_col defines whether we have color or black/white plot.

# Then freq_column is passed to wordcloud plot.

p_col <- reactive({
  if(input$Colorinp2){
    brewer.pal(8, "Dark2")
  } else {
    NA
  }
})

# wordcloud plot

# It make use of Ninp2 to define the min frequency in the plot

output$pout2 <- renderPlot({
  # plot wordcloud
  par(mfrow = c(1, 3))
  wordcloud(words = final$word, freq = final$freq1, min.freq = input$Ninp2, colors=p_col())
  wordcloud(words = final$word, freq = final$freq2, min.freq = input$Ninp2.1, colors=p_col())
  wordcloud(words = final$word, freq = final$freq3, min.freq = input$Ninp2.2, colors=p_col())
})

# Print out table format with top rows from final data frame

# The number of rows is defined by the user.

# Decide which doc

```

```
output$tout2 <- renderTable({  
  head(docContent(), n = input$Ninp3)  
})
```

```
# Tab Summary
```

```
output$tout3 <- renderTable({  
  PS2  
})
```

```
output$summaryu1 <- renderPrint({  
  print.sum[1]  
})
```

```
output$summaryu2 <- renderPrint({  
  print.sum[2]  
})
```

```
output$summaryu3 <- renderPrint({  
  print.sum[3]  
})
```

```
# # Tab 4 - plotly
```

```
# output$plotly4 <- renderPlotly({  
  # plot_ly(stockdata(), x = ~rownames(stockdata()), y = ~stockdata()[,4], type = "scatter", mode =  
  "lines", name = "Close")%>%  
  # add_trace(y = ~stockdata()[,1], name = 'Open')  
  # })  
# #
```

```
# # Tab SQL and Names

output$printsql1 <- renderText({

  SQLtext1

})
```

```
output$printsql2 <- renderText({

  SQLtext2

})
```

```
#Print exactly like the output in R console

output$randomusers <- renderPrint({

  random_users[,1]

})

})
```

UI File:

#Step 1: load packages

```
library(plotly)
```

```
library(ggvis)
```

```
library(shinythemes)
```

#UI define the layout and format of whole web interface

```
shinyUI(
```

```
  fluidPage(
```

```
    # Use the theme in shinythemes package
```

```
    theme = shinytheme("united"),
```

```
    # import .css file (external file)
```

```
    # You need to put the css file into www folder
```

```
    #theme = "style.css",
```

```
##### Set header #####
```

```
# Application title
```

```
headerPanel("FE513 (Name: Assignment 3 App)",
```

```
##### Set sidebar #####
```

```
# Sidebar with a slider input for number of observations
```

```
sidebarPanel(
```

```
  # Input option for colors on wordcloud
```

```
  checkboxInput("Colorinp2", "wordcloud color?", FALSE),
```

```
  # Define a input numeric block with default value = 2, each word cloud can have a different frequency
```

```
  numericInput("Ninp2", "Wordcloud min frequency", 2),
```

```
  numericInput("Ninp2.1", "Wordcloud min frequency", 2),
```

```
  numericInput("Ninp2.2", "Wordcloud min frequency", 2)
```

```
),
```

```
##### Set mainpanel #####
```

```
# Show a plot of the generated distribution
```

```
mainPanel(
```

```
  h3("The widget on the side panel controls the word clouds. This page loads slowly due to the processing of words in the database."),
```

```
  tabsetPanel(
```

```

#    tabPanel("1.Random Line", code("data <- rnorm(N)"),plotOutput("pout11"),
plotOutput("pout12")),

    tabPanel("E1.SQL and Names", textOutput("printsqli1"), p("\n"), textOutput("printsqli2"), p("\n"),
textOutput("randomusers")),

    tabPanel("E1.Wordcloud", plotOutput("pout2")),

    tabPanel("E2.Summary", textOutput("summaryu1"), p("\n"), textOutput("summaryu2"), p("\n"),
textOutput("summaryu3"), p("\n"), tableOutput("tout3")),

    tabPanel("E2.Results", h3("The information below is the size of the clusters."),
textOutput("kmeans"), p("\n"), h3("The information below indicates variance between clusters, within
clusters, and of clusters."), textOutput("kbet"), p("\n"), textOutput("ktotss"), p("\n"),
textOutput("kwith"), p("\n"), textOutput("ktotwith"), p("\n"), textOutput("Hclust"), p("\n"),
plotOutput("Hplot")),

    tabPanel("E2.Compare", h3("The table below is for K means."), p("\n"), tableOutput("ktable"),
p("\n"), h3("The table below is for Hierarchical Clustering."), p("\n"), tableOutput("htable"))

)

)

)

)

```