

# Midterm

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1)  $U \leftarrow c(\text{"Maine"}, \text{"Texas"}, \text{"Delaware"}, \text{"Oregon"}, \text{"Utah"}, \text{"Vermont"}, \text{"Ohio"})$

```
U <- c("Maine", "Texas", "Delaware", "Oregon", "Utah", "Vermont", "Ohio")
str(U)
```

```
## chr [1:7] "Maine" "Texas" "Delaware" "Oregon" "Utah" "Vermont" "Ohio"
```

a) Is U an atomic vector or a list ?

U is an atomic vector since they are all character vectors.

b) Use and show R code that will extract the elements “Maine” and “Vermont”.

```
U[c(1,6)]
```

```
## [1] "Maine" "Vermont"
```

c) Use and show R code that will extract all elements except “Texas”.

```
U[-2]
```

```
## [1] "Maine" "Delaware" "Oregon" "Utah" "Vermont" "Ohio"
```

d) Use and show R code that will produce length of U.

```
length(U)
```

```
## [1] 7
```

2)  $V = \text{list}(\text{"Chicago"}, k = \text{list}(2, 6, 18, 24), \text{FALSE}, 13, 1.3, y = 1:10)$

```
V = list("Chicago", k = list(2, 6, 18, 24), FALSE, 13, 1.3, y = 1:10)
str(V)
```

```
## List of 6
## $ : chr "Chicago"
## $ k:List of 4
## ..$ : num 2
```

```
## ..$ : num 6
## ..$ : num 18
## ..$ : num 24
## $ : logi FALSE
## $ : num 13
## $ : num 1.3
## $ y: int [1:10] 1 2 3 4 5 6 7 8 9 10
```

a) Is **V** an atomic vector or a list ?

V is list since it has more than one type of vectors.

b) Use and show R code that will extract the 5th element of **V**.

```
V[5]
```

```
## [[1]]
## [1] 1.3
```

c) If the vector **V** is a list, use and show R code to identify the type of each object in **V**.

```
str(V)
```

```
## List of 6
## $ : chr "Chicago"
## $ k:List of 4
## ..$ : num 2
## ..$ : num 6
## ..$ : num 18
## ..$ : num 24
## $ : logi FALSE
## $ : num 13
## $ : num 1.3
## $ y: int [1:10] 1 2 3 4 5 6 7 8 9 10
```

3) Copy paste and run the tribble given below.

```
tribble( ~John, ~Raymond, ~Martha,~Alice, ~Juan,
          86,    77,    81,    88,    90,
          79,          78,          85,          81,          78,
          76,          75,          88,          94,          81,
          84,          90,          71,          84,          89,
          100,         80,          93,          85,          84,
          90,          73,          70,          88,          93,
) -> TestScores
TestScores

## # A tibble: 6 x 5
##   John Raymond Martha Alice Juan
##   <dbl>   <dbl>   <dbl> <dbl> <dbl>
## 1    86     77     81    88    90
## 2    79     78     85    81    78
```

```
## 3    76    75    88    94    81
## 4    84    90    71    84    89
## 5   100    80    93    85    84
## 6    90    73    70    88    93
```

a) Use and show R code (a map function) to find the median for each column.

```
map_dbl(TestScores,median)
```

```
##      John Raymond  Martha   Alice    Juan
##      85.0      77.5    83.0    86.5    86.5
```

b) Use and show R code (a map function) to find the cube root of each column element.

```
TestScores %>%
  map(~.(1/3)) -> T
T
```

```
## $John
## [1] 4.414005 4.290840 4.235824 4.379519 4.641589 4.481405
##
## $Raymond
## [1] 4.254321 4.272659 4.217163 4.481405 4.308869 4.179339
##
## $Martha
## [1] 4.326749 4.396830 4.447960 4.140818 4.530655 4.121285
##
## $Alice
## [1] 4.447960 4.326749 4.546836 4.379519 4.396830 4.447960
##
## $Juan
## [1] 4.481405 4.272659 4.326749 4.464745 4.379519 4.530655
```

c) Use and show R code (a map function) to convert each column value to 0.

```
zero <-TestScores %>%
  map(~. * 0)
zero
```

```
## $John
## [1] 0 0 0 0 0 0
##
## $Raymond
## [1] 0 0 0 0 0 0
##
## $Martha
## [1] 0 0 0 0 0 0
##
## $Alice
## [1] 0 0 0 0 0 0
##
## $Juan
## [1] 0 0 0 0 0 0
```

4) Use and show R code, as demonstrated in class to produce the following matrix

```
z <- matrix( nrow = 3, ncol = 4)
for (m in 1:3) {
  for (n in 1:4) {
    z[m, n] <- -(m+n)^2
  }
}
print(z)
```

```
##      [,1] [,2] [,3] [,4]
## [1,]   -4   -9  -16  -25
## [2,]   -9  -16  -25  -36
## [3,]  -16  -25  -36  -49
```

5)

a) Show and use a census API key that gives you access to the ACS data. Do not use my API key, use and show your own key.

```
# census_api_key("4009f73e21670e9fb8801c8067991ecb855c1632", install=TRUE)
census_api_key("4009f73e21670e9fb8801c8067991ecb855c1632", overwrite=TRUE)
```

```
## To install your API key for use in future sessions, run this function with `install = TRUE`.
# For line 102 I can not knit that is why I use #
```

b) Using ACS census data from 2015, show and use R code to do the following to produce a tibble that shows the median income estimates and the margin of errors for white males ages 35 - 44 in the counties of California. The required variable code starts with the characters BO1001. Use the table to find the other characters.

```
c3544 <- load_variables(2015,"acs5",cache = TRUE)

ca <- get_acs(geography = "county",
              variables = c(medincome = "BO1001A_011"),
              state = "06",
              year = 2015)
```

```
## Getting data from the 2011-2015 5-year ACS
```

```
ca
```

```
## # A tibble: 58 x 5
##   GEOID NAME                variable estimate moe
##   <chr> <chr>                <chr>      <dbl> <dbl>
## 1 06001 Alameda County, California medincome 51644 667
## 2 06003 Alpine County, California medincome    50  26
## 3 06005 Amador County, California medincome 1809  72
```

```
## 4 06007 Butte County, California      medincome      9962      128
## 5 06009 Calaveras County, California  medincome      1927       74
## 6 06011 Colusa County, California    medincome      1147       79
## 7 06013 Contra Costa County, California medincome    42756      605
## 8 06015 Del Norte County, California medincome      1629       90
## 9 06017 El Dorado County, California medincome      8609      141
## 10 06019 Fresno County, California   medincome    34979      714
## # ... with 48 more rows
```

```
head(ca,5)
```

```
## # A tibble: 5 x 5
##   GEOID NAME      variable estimate   moe
##   <chr> <chr>      <chr>      <dbl> <dbl>
## 1 06001 Alameda County, California medincome    51644    667
## 2 06003 Alpine County, California medincome         50     26
## 3 06005 Amador County, California medincome    1809     72
## 4 06007 Butte County, California medincome    9962    128
## 5 06009 Calaveras County, California medincome    1927     74
```

c) Use dplyr functions to change your table of part a so that it reflects estimates that are greater than \$30,000 dollars and list the estimates in descending order.

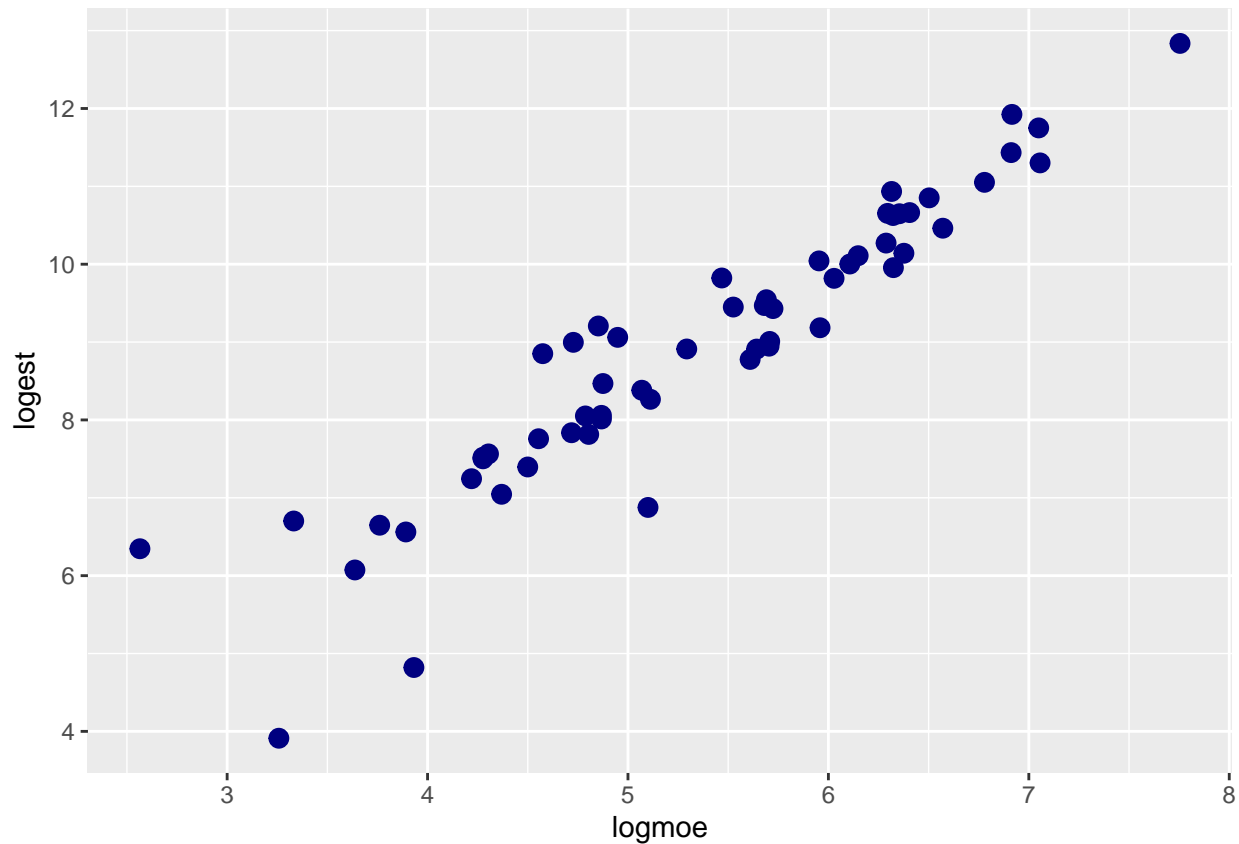
```
ca %>%
  filter(estimate>= 30000) %>%
  arrange(desc(estimate)) ->ca30
ca30
```

```
## # A tibble: 13 x 5
##   GEOID NAME      variable estimate   moe
##   <chr> <chr>      <chr>      <dbl> <dbl>
## 1 06037 Los Angeles County, California medincome  375435  2332
## 2 06073 San Diego County, California medincome  150891  1008
## 3 06059 Orange County, California medincome  126819  1152
## 4 06065 Riverside County, California medincome   92346  1004
## 5 06071 San Bernardino County, California medincome   80925  1160
## 6 06085 Santa Clara County, California medincome   63036   879
## 7 06067 Sacramento County, California medincome   56066   553
## 8 06001 Alameda County, California medincome   51644   667
## 9 06013 Contra Costa County, California medincome   42756   605
## 10 06075 San Francisco County, California medincome   42307   542
## 11 06029 Kern County, California medincome   42121   575
## 12 06111 Ventura County, California medincome   41155   557
## 13 06019 Fresno County, California medincome   34979   714
```

d) Use and show ggplot R coding to produce a scatter plot that features x = natural log of moe plotted against y = natural log of estimate. Does your plot suggest a linear relationship between the variables ? If so, what general trend can be inferred? (Use the full data table that you generated for part b)

```
log(ca$moe) ->logmoe
log(ca$estimate)-> logest
```

```
ggplot(ca,mapping =aes(x =logmoe,y=logest))+
  geom_point(color= "navy blue",size=3,alpha=0.5)
```

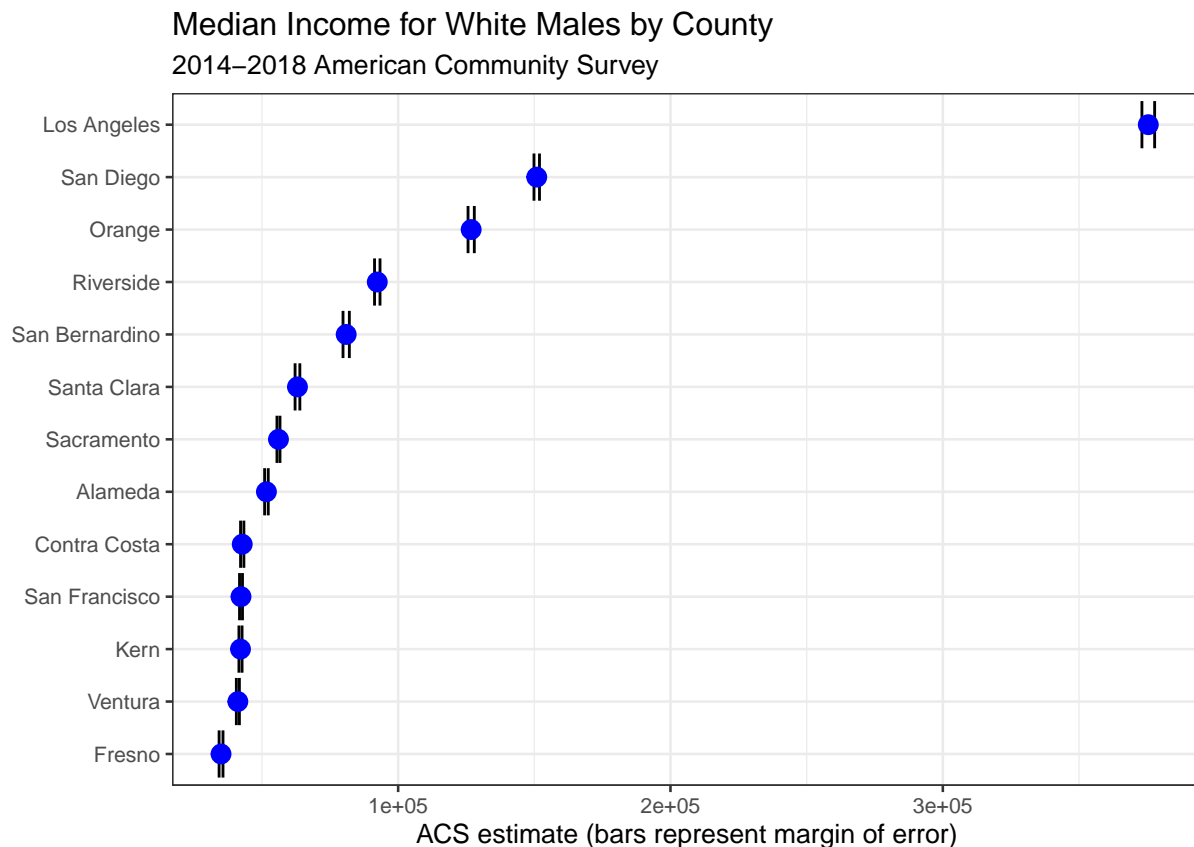


```
calog <- lm(logest~logmoe,ca)
```

I use ggplot to create a scatter plot it shows that natural log of moe and natural log of estimate has a positive linear relationship in here and when logmoe increase log of estimate will also increase so we can say that natural log of moe and natural log of estimate is a linear relationship.

e) Use and show R code that will produce the following graph ing graph for the data generated in part c

```
ca30 %>%
  mutate(NAME = gsub(" County, California", "", NAME)) %>%
  ggplot(aes(x = estimate, y = reorder(NAME, estimate))) +
  geom_errorbarh(aes(xmin = estimate - moe, xmax = estimate + moe)) +
  geom_point(color = "Blue", size = 3) +
  labs(title = "Median Income for White Males by County",
       subtitle = "2014-2018 American Community Survey",
       y = "",
       x = "ACS estimate (bars represent margin of error)") +
  theme_bw(base_size = 10)
```



6) Provided below is the famous poem “ Stopping by the Wood On a Snowy Evening” by Robert Frost. Use the text mining sequence of steps and the R code modeled in class to

- create a tibble
- find line locations of words
- produce a word frequency table, and
- and create a bar graph data visualization plot that will also display word frequency trends.

Hint( do not forget to process a single spaced body of text ; be careful about commas and double quotation marks. Use the examples demonstrated in class.)

```
text <- c("Whose woods these are I think I know.",
          "His house is in the village though;",
          "He will not see me stopping here",
          "To watch his woods fill up with snow.",
          "My little horse must think it queer",
          "To stop without a farmhouse near",
          "Between the woods and frozen lake",
          "The darkest evening of the year.",
          "He gives his harness bells a shake",
          "To ask if there is some mistake.",
```

```

    "The only other sound's the sweep",
    "Of easy wind and downy flake.",
    "The woods are lovely, dark and deep,",
    "But I have promises to keep,",
    "And miles to go before I sleep,",
    "And miles to go before I sleep.")

```

```
text
```

```

## [1] "Whose woods these are I think I know."
## [2] "His house is in the village though;"
## [3] "He will not see me stopping here"
## [4] "To watch his woods fill up with snow."
## [5] "My little horse must think it queer"
## [6] "To stop without a farmhouse near"
## [7] "Between the woods and frozen lake"
## [8] "The darkest evening of the year."
## [9] "He gives his harness bells a shake"
## [10] "To ask if there is some mistake."
## [11] "The only other sound's the sweep"
## [12] "Of easy wind and downy flake."
## [13] "The woods are lovely, dark and deep,"
## [14] "But I have promises to keep,"
## [15] "And miles to go before I sleep,"
## [16] "And miles to go before I sleep."

```

```
text_tibble <- tibble(line = 167:182, text = text)
```

```
text_tibble
```

```

## # A tibble: 16 x 2
##   line text
##   <int> <chr>
## 1 167 Whose woods these are I think I know.
## 2 168 His house is in the village though;
## 3 169 He will not see me stopping here
## 4 170 To watch his woods fill up with snow.
## 5 171 My little horse must think it queer
## 6 172 To stop without a farmhouse near
## 7 173 Between the woods and frozen lake
## 8 174 The darkest evening of the year.
## 9 175 He gives his harness bells a shake
## 10 176 To ask if there is some mistake.
## 11 177 The only other sound's the sweep
## 12 178 Of easy wind and downy flake.
## 13 179 The woods are lovely, dark and deep,
## 14 180 But I have promises to keep,
## 15 181 And miles to go before I sleep,
## 16 182 And miles to go before I sleep.

```

```

text_tibble %>%
  unnest_tokens(word, text) -> tibble1
tibble1

```

```

## # A tibble: 108 x 2
##   line word
##   <int> <chr>

```

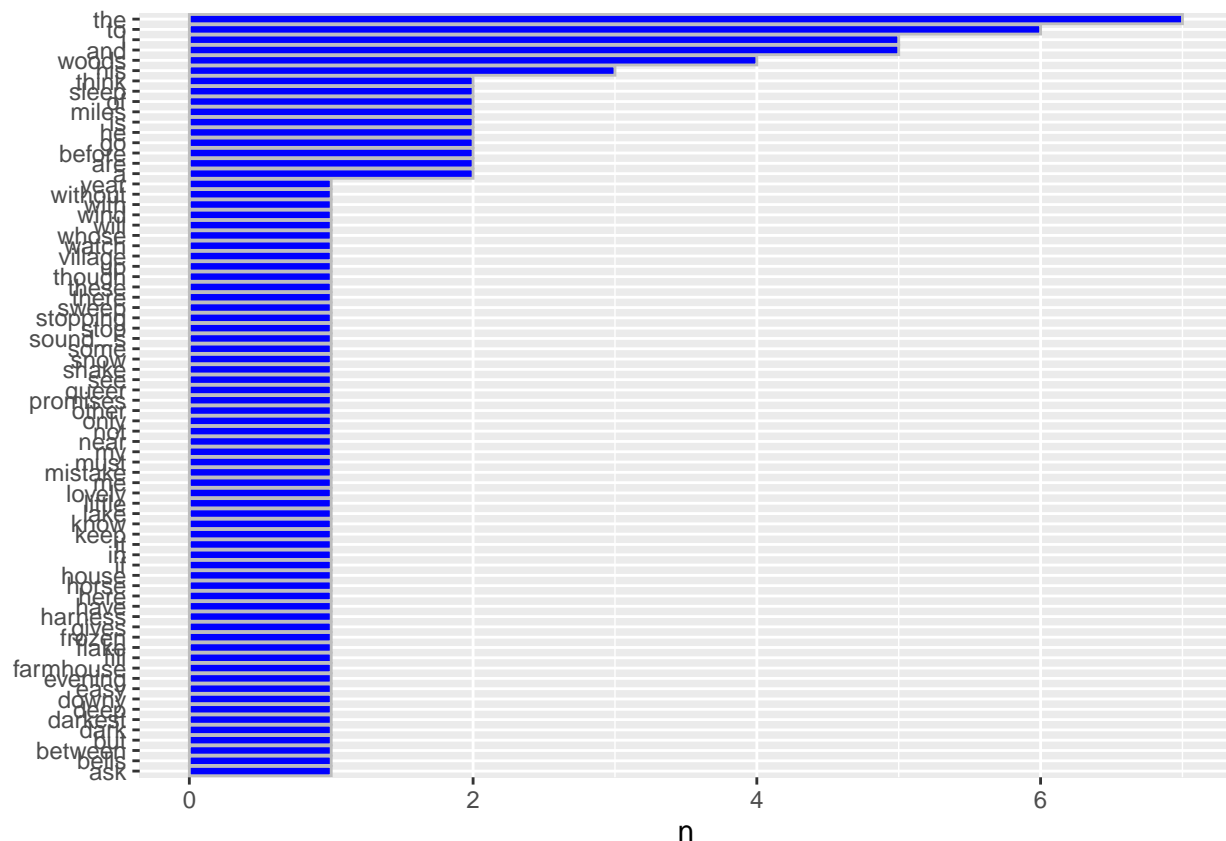


```
## 1 167 whose
## 2 167 woods
## 3 167 these
## 4 167 are
## 5 167 i
## 6 167 think
## 7 167 i
## 8 167 know
## 9 168 his
## 10 168 house
## # ... with 98 more rows
```

```
tibble1%>%
  count(word, sort =TRUE) %>%
  filter(n >= 1)
```

```
## # A tibble: 74 x 2
##   word      n
##   <chr> <int>
## 1 the      7
## 2 to       6
## 3 and      5
## 4 i        5
## 5 woods    4
## 6 his      3
## 7 a        2
## 8 are      2
## 9 before   2
## 10 go      2
## # ... with 64 more rows
```

```
tibble1%>%
  count(word, sort =TRUE) %>%
  filter(n >= 1) %>%
  mutate(word = reorder(word, n)) %>%
  ggplot(aes(n, word)) +
  geom_col(fill = "blue", color = "grey") +
  labs(y = NULL)
```



7) Now using the same body of text found in Problem 6, use and show R code to create a word cloud. You can use the coding and methods that were illustrated in class or you can use alternate coding of your choice to create the word cloud.

```
text1 = read_table("midterm.txt")

##
## -- Column specification -----
## cols(
##   Whose = col_character(),
##   woods = col_character(),
##   these = col_character(),
##   are = col_character(),
##   I = col_character(),
##   think = col_character(),
##   I_1 = col_character(),
##   know. = col_character()
## )

docs <- Corpus(VectorSource(text1))
docs

## <<SimpleCorpus>>
## Metadata: corpus specific: 1, document level (indexed): 0
```

```
## Content: documents: 8
```

```
inspect(docs)
```

```
## <<SimpleCorpus>>
```

```
## Metadata: corpus specific: 1, document level (indexed): 0
```

```
## Content: documents: 8
```

```
##
```

```
## [1] c("His", "He", "To", "My", "To", "Between", "The", "He", "To", "The", "Of", "The", "But", "And",
```

```
## [2] c("house", "will", "watch", "little", "stop", "the", "darkest", "gives", "ask", "only", "easy",
```

```
## [3] c("is", "not", "his", "horse", "without", "woods", "evening", "his", "if", "other", "wind", "are
```

```
## [4] c("in", "see", "woods", "must", "a", "and", "of", "harness", "there", "sound's", "and", "lovely,
```

```
## [5] c("the", "me", "fill", "think", "farmhouse", "frozen", "the", "bells", "is", "the", "downy", "da
```

```
## [6] c("village", "stopping", "up", "it", "near", "lake", "year.", "a", "some", "sweep", "flake.", "a
```

```
## [7] c("though;", "here", "with", "queer", NA, NA, NA, "shake", "mistake.", NA, NA, "deep,", NA, "slee
```

```
## [8] c(NA, NA, "snow.", NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA)
```

```
toSpace <- content_transformer(function(x, pattern) gsub(pattern, " ", x))
```

```
docs <- tm_map(docs, toSpace, "/")
```

```
docs <- tm_map(docs, toSpace, "@")
```

```
docs <- tm_map(docs, toSpace, "\\|")
```

```
docs <- tm_map(docs, content_transformer(tolower))
```

```
docs <- tm_map(docs, removeNumbers)
```

```
docs <- tm_map(docs, removeWords, stopwords("english"))
```

```
docs <- tm_map(docs, removeWords, c("blabla1", "blabla2"))
```

```
docs <- tm_map(docs, removePunctuation)
```

```
docs <- tm_map(docs, stripWhitespace)
```

```
dtm <- TermDocumentMatrix(docs)
```

```
m <- as.matrix(dtm)
```

```
v <- sort(rowSums(m), decreasing=TRUE)
```

```
d <- data.frame(word = names(v), freq=v)
```

```
head(d, 10)
```

```
##           word freq
```

```
## woods      woods   3
```

```
## miles      miles   2
```

```
## sleep      sleep   2
```

```
## ask        ask     1
```

```
## chouse     chouse   1
```

```
## darkest    darkest  1
```

```
## easy       easy     1
```

```
## gives      gives    1
```

```
## little     little   1
```

```
## stop       stop     1
```

```
set.seed(1234)
```

```
wordcloud(words = d$word, freq = d$freq, min.freq = 1,  
           max.words=200, random.order=FALSE, rot.per=0.35,  
           colors=brewer.pal(8, "Blues"))
```



8) Go to the link “<https://www.imdb.com/list/ls096735829>” and use Selector Gadget , as demonstrated in class to collect data specific to the movie names, the directors of the movies, the movie ratings, and the running times of the movies. Your code should produce the final resulting table given below. Hint: (you may have to first create a data frame, and then convert the data frame to the tibble shown below.

```
Link <- "https://www.imdb.com/list/ls096735829"
page = read_html(Link)
Movies2020 = page%>% html_nodes(".list-item-header a")%>%
  html_text()

page = read_html(Link)
Directors2020 = page%>% html_nodes(".text-muted a:nth-child(1)")%>%
  html_text()

page = read_html(Link)
Ratings2020 = page%>% html_nodes(".ipl-rating-star.small .ipl-rating-star__rating")%>%
  html_text()

page = read_html(Link)
Runtime2020 = page%>% html_nodes(".runtime")%>%
  html_text()
```

```
moviesdataframe = data.frame(Movies2020,Directors2020, Ratings2020,Runtime2020)

as_tibble(moviesdataframe)
```

```
## # A tibble: 47 x 4
##   Movies2020      Directors2020    Ratings2020 Runtime2020
##   <chr>          <chr>          <chr>        <chr>
## 1 Weathering with You Makoto Shinkai    7.5         112 min
## 2 The Empty Man      David Prior       6.2         137 min
## 3 Monsters of Man     Mark Toia        5.4         131 min
## 4 Songbird           Adam Mason       4.7          84 min
## 5 Tesla              Michael Almereyda 5.1         102 min
## 6 Underwater         William Eubank    5.8          95 min
## 7 Greenland          Ric Roman Waugh   6.4         119 min
## 8 Possessor          Brandon Cronenberg 6.5         103 min
## 9 Tenet              Christopher Nolan  7.3         150 min
## 10 The Phenomenon     James Fox        7.4         100 min
## # ... with 37 more rows
```

9)

a) Use and show R code that shows both column variables x and y, of the diamonds datatable contain the value 4.93. How many times does the number 4.93 appear in each column ?

```
diamonds %>%
  filter(x== 4.93)-> x4.93
x4.93
```

```
## # A tibble: 42 x 10
##   carat cut      color clarity depth table price      x      y      z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1  0.44 Ideal    G      SI2     61.1    55   694  4.93  4.95  3.02
## 2  0.43 Premium  H      SI1     59      58   716  4.93  4.96  2.92
## 3  0.41 Ideal    E      SI2     59.8    55   876  4.93  4.81  2.91
## 4  0.46 Ideal    I      VS1     62.3    56   911  4.93  4.96  3.08
## 5  0.43 Premium  H      SI1     60.2    57   919  4.93  4.91  2.96
## 6  0.45 Premium  F      VS2     61.4    60   945  4.93  4.87  3.01
## 7  0.46 Ideal    J      VVS1     61.8    56   953  4.93  4.97  3.06
## 8  0.43 Premium  D      VS2     59.7    59   963  4.93  4.89  2.93
## 9  0.46 Very Good E      SI1     61.8    57   968  4.93  4.98  3.06
## 10 0.45 Ideal    J      VVS1     61.9    56   978  4.93  4.95  3.06
## # ... with 32 more rows
```

```
diamonds %>%
  filter(y== 4.93)-> y4.93
y4.93
```

```
## # A tibble: 50 x 10
##   carat cut      color clarity depth table price      x      y      z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1  0.44 Ideal    H      SI2     62.3    54   654  4.89  4.93  3.06
## 2  0.41 Ideal    E      SI2     59.8    55   683  4.81  4.93  2.91
```

```
## 3 0.43 Very Good H SI1 60.2 57 716 4.91 4.93 2.96
## 4 0.44 Premium H SI2 59.6 60 759 4.95 4.93 2.96
## 5 0.44 Premium G SI1 61 60 772 4.88 4.93 2.99
## 6 0.43 Premium D SI1 60.1 58 830 4.89 4.93 2.95
## 7 0.44 Ideal D SI1 61.6 56 838 4.9 4.93 3.03
## 8 0.43 Ideal F VS2 61.4 54 848 4.9 4.93 3.02
## 9 0.43 Very Good G VS2 59.1 60 867 4.88 4.93 2.9
## 10 0.43 Premium D VS2 59.7 59 901 4.89 4.93 2.93
## # ... with 40 more rows
```

```
nrow(x4.93)
```

```
## [1] 42
```

```
nrow(y4.93)
```

```
## [1] 50
```

b) Use and show R code that shows neither column variable x or y, contain the value 3.62.

```
diamonds %>%
  filter(x!=3.62 | y!= 3.62)-> neither
head(neither)
```

```
## # A tibble: 6 x 10
##   carat cut      color clarity depth table price      x      y      z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1 0.23 Ideal    E      SI2      61.5  55  326  3.95  3.98  2.43
## 2 0.21 Premium  E      SI1      59.8  61  326  3.89  3.84  2.31
## 3 0.23 Good     E      VS1      56.9  65  327  4.05  4.07  2.31
## 4 0.29 Premium  I      VS2      62.4  58  334  4.2   4.23  2.63
## 5 0.31 Good     J      SI2      63.3  58  335  4.34  4.35  2.75
## 6 0.24 Very Good J      VVS2      62.8  57  336  3.94  3.96  2.48
```

```
tibble(neither)
```

```
## # A tibble: 53,940 x 10
##   carat cut      color clarity depth table price      x      y      z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1 0.23 Ideal    E      SI2      61.5  55  326  3.95  3.98  2.43
## 2 0.21 Premium  E      SI1      59.8  61  326  3.89  3.84  2.31
## 3 0.23 Good     E      VS1      56.9  65  327  4.05  4.07  2.31
## 4 0.29 Premium  I      VS2      62.4  58  334  4.2   4.23  2.63
## 5 0.31 Good     J      SI2      63.3  58  335  4.34  4.35  2.75
## 6 0.24 Very Good J      VVS2      62.8  57  336  3.94  3.96  2.48
## 7 0.24 Very Good I      VVS1      62.3  57  336  3.95  3.98  2.47
## 8 0.26 Very Good H      SI1      61.9  55  337  4.07  4.11  2.53
## 9 0.22 Fair     E      VS2      65.1  61  337  3.87  3.78  2.49
## 10 0.23 Very Good H      VS1      59.4  61  338  4     4.05  2.39
## # ... with 53,930 more rows
```

c) Now show and use R code to find all values that the column variables x and y have in common.

```
diamonds %>%
  filter(x==y & y==x) ->common
common
```

```
## # A tibble: 17 x 10
##   carat cut      color clarity depth table price      x      y      z
##   <dbl> <ord>    <ord> <ord>    <dbl> <dbl> <int> <dbl> <dbl> <dbl>
## 1  0.3 Ideal     H      VVS2    62.5   54   567   4.3   4.3   2.7
## 2  0.27 Very Good F      VVS1    62    55   591   4.16  4.16  2.59
## 3  1      Very Good H      VS2    63.3   53  5139    0     0     0
## 4  1.14 Fair     G      VS1    57.5   67  6381    0     0     0
## 5  1      Premium E      VS2    60     60  6600   6.43  6.43  3.89
## 6  1      Premium E      VS2    60     60  6720   6.43  6.43  3.89
## 7  1.22 Premium  G      SI2    62.4   61  6969   6.79  6.79  4.23
## 8  1.56 Ideal     G      VS2    62.2   54  12800    0     0     0
## 9  1.2 Premium  D      VVS1    62.1   59  15686    0     0     0
## 10 2.25 Premium  H      SI2    62.8   59  18034    0     0     0
## 11 0.32 Ideal     D      VVS2    62.1   54   858   4.4   4.4   2.74
## 12 0.42 Ideal     H      VVS1    62.8   57  1108   4.79  4.79  3.01
## 13 0.61 Premium  G      SI1    60.8   60  1255   5.42  5.42  3.31
## 14 0.48 Ideal     F      VS2    62.4   54  1279   5.03  5.03  3.15
## 15 0.51 Premium  F      SI1    61.4   59  1421   5.13  5.13  3.16
## 16 0.71 Good     F      SI2    64.1   60  2130    0     0     0
## 17 0.71 Good     F      SI2    64.1   60  2130    0     0     0
```