# Applied Statistical Programming - Spring 2022

#### Problem Set 3

Due Wednesday, March 16, 10:00 AM (Before Class)

#### Instructions

- 1. The following questions should each be answered within an Rmarkdown file. Be sure to provide many comments in your code blocks to facilitate grading. Undocumented code will not be graded.
- 2. Work on git. Continue to work in the repository you forked from https://github.com/johnsontr/Appli edStatisticalProgramming2022 and add your code for Problem Set 4. Commit and push frequently. Use meaningful commit messages because these will affect your grade.
- 3. You may work in teams, but each student should develop their own Rmarkdown file. To be clear, there should be no copy and paste. Each keystroke in the assignment should be your own.
- 4. For students new to programming, this may take a while. Get started.

### tidyverse

## Rows: 16661 Columns: 33

Your task in this problem set is to combine two datasets in order to observe how many endorsements each candidate received using only dplyr functions. Use the same Presidential primary polls that were used for the in class worksheets on February 28 and March 2.

```
library(fivethirtyeight)
## Some larger datasets need to be installed separately, like senators and
## house_district_forecast. To install these, we recommend you install the
## fivethirtyeightdata package by running:
## install.packages('fivethirtyeightdata', repos =
## 'https://fivethirtyeightdata.github.io/drat/', type = 'source')
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5
                     v purrr
                              0.3.4
## v tibble 3.1.6
                     v dplyr
                              1.0.8
## v tidyr
           1.2.0
                     v stringr 1.4.0
## v readr
           2.1.2
                     v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                 masks stats::lag()
# URL to the data that you've used.
url <- 'https://jmontgomery.github.io/PDS/Datasets/president_primary_polls_feb2020.csv'
polls <- read_csv(url)</pre>
```

```
## -- Column specification ------
## Delimiter: ","
## chr (21): state, pollster, sponsors, display_name, pollster_rating_name, fte...
## dbl (8): question_id, poll_id, cycle, pollster_id, pollster_rating_id, samp...
## lgl (3): internal, tracking, nationwide_batch
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
Endorsements <- endorsements_2020 # from the fiverthirtyeight package</pre>
```

First, create two new objects polls and Endorsements. Then complete the following.

- Change the Endorsements variable name endorsee to candidate\_name.
- Change the Endorsements dataframe into a tibble object.
- Filter the poll variable to only include the following 6 candidates: Amy Klobuchar, Bernard Sanders, Elizabeth Warren, Joseph R. Biden Jr., Michael Bloomberg, Pete Buttigieg and subset the dataset to the following five variables: candidate\_name, sample\_size, start\_date, party, pct
- Compare the candidate names in the two datasets and find instances where the a candidates name is spelled differently i.e. Bernard vs. Bernie. Using only dplyr functions, make these the same across datasets.
- Now combine the two datasets by candidate name using dplyr (there will only be five candidates after joining).
- Create a variable which indicates the number of endorsements for each of the five candidates using dplyr.
- Plot the number of endorsement each of the 5 candidates have using ggplot(). Save your plot as an object p.
- Rerun the previous line as follows: p + theme\_dark(). Notice how you can still customize your plot without rerunning the plot with new options.
- Now, using the knowledge from the last step change the label of the X and Y axes to be more informative, add a title. Save the plot in your forked repository.

```
# Change variable name endorsee in Endorsements of to candidate_name

# Change endorsements of to tibble

# Filter poll variable to include 6 candidates

# Subset the data to include the five variables

# Find differently spelled names and unify them (use only dplyr)

# Variable where number of endorsements is counted

# Plot each of the 6(?) candidates

# add theme_dark()

# Add titles, and save plot
```

## Text-as-Data with tidyverse

For this question you will be analyzing Tweets from President Trump for various characteristics. Load in the following packages and data:

```
# Change eval=FALSE in the code block. Install packages as appropriate.
library(tidyverse)
library(tm)
## Loading required package: NLP
## Attaching package: 'NLP'
## The following object is masked from 'package:ggplot2':
##
##
       annotate
library(lubridate)
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
       date, intersect, setdiff, union
library(wordcloud)
## Loading required package: RColorBrewer
trump_tweets_url <- 'https://politicaldatascience.com/PDS/Datasets/trump_tweets.csv'</pre>
tweets <- read_csv(trump_tweets_url)</pre>
## Rows: 32974 Columns: 6
## -- Column specification -----
## Delimiter: ","
## chr (3): source, text, created_at
## dbl (2): retweet_count, favorite_count
## lgl (1): is_retweet
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

- First separate the created\_at variable into two new variables where the date and the time are in separate columns. After you do that, then report the range of dates that is in this dataset.
- Using dplyr subset the data to only include original tweets (remove retweents) and show the text of the President's top 5 most popular and most retweeted tweets. (Hint: The match function can help you find the index once you identify the largest values.)
- Create a *corpus* of the tweet content and put this into the object Corpus using the tm (text mining) package. (Hint: Do the assigned readings.)
- Remove extraneous whitespace, remove numbers and punctuation, convert everything to lower case and remove 'stop words' that have little substantive meaning (the, a, it).
- Now create a wordcloud to visualize the top 50 words the President uses in his tweets. Use only words that occur at least three times. Display the plot with words in random order and use 50 random colors. Save the plot into your forked repository.

- Create a document term matrix called DTM that includes the argument control = list(weighting = weightTfldf)
- Finally, report the 50 words with the the highest tf.idf scores using a lower frequency bound of .8.