Problem Set 2: Data Wrangling

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Background

Political advertising has traditionally been focused on the medium of television, but in recent cycles, online advertising has become much more popular. In this problem set, you will explore a dataset that has information on Facebook ad spending and impressions by candidates in the 2018 election cycle in the United States. The variables in this data are described below.

Name	Description
cand_id	unique identifier code for candidate
cand_name	full name of the candidate
cand_name_last	last name of the candidate
party	party affiliation of the candidate ($R = Republican, D = Democrat$)
office	office being sought by candidate
state	state in which the candidate is running
incumbency	incumbency status of candidate (incumbent, challenger, or open seat)
spend	estimated total spending on Facebook ads by candidate
impressions	estimated total impressions of Facebook ads
ad_tone_attack	proportion of FB ads that mention candidate's opponent only
ad_tone_promote	proportion of FB ads that mention candidate only
ad_tone_contrast	proportion of FB ads that mention candidate and candidate's opponent

Question 1 (8 points)

Load the data using the read_csv function and save it as fb_ads (using this will automatically make fb_ads a tibble). In the text, describe how many candidates there are in the dataset.

Use dplyr functions to create a table with the number of candidates in each type of incumbency status in the data set. Save this table output as incumbency_table (for the autograder). Use the function knitr::kable() on this table to have a nicely formatted table produced in the knitted output.

Rubric: 2pt for loading the data (autograder); 1pt for describing the number of candidates (PDF); 3pts for creating the table (autograder); 2pt for using kable() to nicely format the output (PDF)

Answer 1

```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
fb_ads <- read.csv("data/fb_ads.csv")</pre>
incumbency_table <- fb_ads |>
  count(incumbency) |>
  knitr::kable(col.names = c("Incumbency Status", "Amount"))
incumbency_table
```

Incumbency Status	Amount
Challenger	2510
Incumbent	2022
Open Seat	2482

There are 7,014 candidates in the dataset.

Question 2 (7 points)

Filter the data to just US House and US Senate races and use this to create a tibble called party_incumbent_promote that has 6 rows that summarizes the average of ad_tone_promote for each combination of party and incumbency. Call the variable summarizing the promote variable as promote_prop and be sure to remove any missing values when computing the averages.

Use knitr::kable() to produce a nicely formatted table. In this call, set the digits arguments to 3 and use the col.names argument to pass a nicer set of names. You can use the following as a template:

In the writeup, describe which type of candidate sponsored the most promoting ads on average.

Rubric: 3pts for creating party_incumbent_promote correctly (autograder); 2pt for a nicely formatted table (PDF); 1pt for changing the column names of the output table (PDF); 1pt for correctly identifying the type of candidate with highest average (PDF)

Answer 2

```
party_incumbent_promote <- fb_ads |>
  filter(office == "US House" | office == "US Senate") |>
  group_by(party, incumbency) |>
  summarize(promote_prop = mean(ad_tone_promote, na.rm = TRUE)) |>
  head()
```

'summarise()' has grouped output by 'party'. You can override using the '.groups' argument.

```
knitr::kable(party_incumbent_promote, col.names = c("Party", "Incumbency", "Promotion"), digits = 3)
```

Party	Incumbency	Promotion
D	Challenger	0.833
D	Incumbent	0.854
D	Open Seat	0.845
R	Challenger	0.813
R	Incumbent	0.792
R	Open Seat	0.828

party_incumbent_promote

```
## # A tibble: 6 x 3
## # Groups: party [2]
     party incumbency promote_prop
##
     <chr> <chr>
                              <dbl>
## 1 D
           Challenger
                              0.833
## 2 D
           Incumbent
                              0.854
           Open Seat
## 3 D
                              0.845
## 4 R
           Challenger
                              0.813
## 5 R
           Incumbent
                              0.792
## 6 R
           Open Seat
                              0.828
```

The type of candidate with the highest average sponsored self-promotions are incumbent democrats.

Question 3 (7 points)

Create a new variable called impressions_millions that is the total Facebook ad impressions in millions (as opposed to single impressions). Make sure to save the resulting dataset back as fb_ads.

Create a histogram of this variable for just the US House races. Save the ggplot output as plot_q3 and also print it to produce a plot in the output. In the text, describe the shape of the histogram and tell the reader if most of the House candidates had more than 10 million ads impressions on Facebook.

Rubric: 2pt for creating the new variable (autograder); 3pts for creating the histogram object (autograder); 2pts for answering the question about the histogram (PDF)

Answer 3

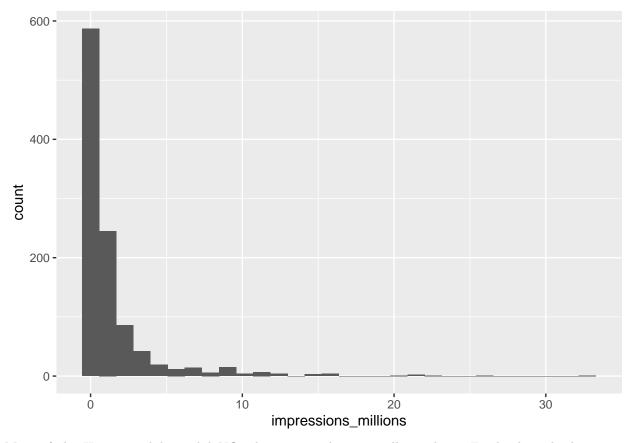
```
library(ggplot2)

fb_ads <- fb_ads |>
    mutate(impressions_millions = impressions/1000000)

plot_q3 <- fb_ads |>
    filter(office == "US House") |>
    ggplot(mapping = aes(x = impressions_millions)) +
    geom_histogram()

plot_q3
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



Most of the House candidates did NOT have more than 10 million ads on Facebook. The histogram is strongly skewed to the right, with the vast majority of the House candidates achieving 0-5 million impressions.

Question 4 (13 points)

Let's now recreate the following plot that shows the top 15 House candidates in terms of Facebook ad impressions.



You should save the ggplot output as fb_top_plot. You should also write fb_top_plot on its own line in the chunk to produce the actual plot. The key features of this graph that you should replicate for the autograder are:

- The barplot should have candidate last names on the y-axis and the impressions_millions variable from question 3 on the x-axis.
- The data feeding into the ggplot call should only have US House candidates and only the candidates with the highest 15 impressions_millions values.
- The y-axis should be ordered in ascending values of impressions_millions so that the lowest values are at the bottom. You may want to manipulate cand_name_last to achieve this.
- The fill color of the bar plot should be mapped to the party variable (but not globally!).

You do not need to exactly match the labels, but you should have informative labels. The color does not need to match, but if you want to change the fill colors, you can use the scale_fill_manual(values = c(R = "red", D = "blue")) function (where you can change the red and blue to whatever you want).

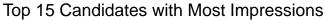
Rubric: 3pts for correct axes (autograder); 3pts for correct data fed into ggplot (autograder); 3pts for the correct ordering of the y-axis (PDF); 3pts for fill being mapped to party (autograder); 1pt for plot being in knitted output and having informative labels (PDF)

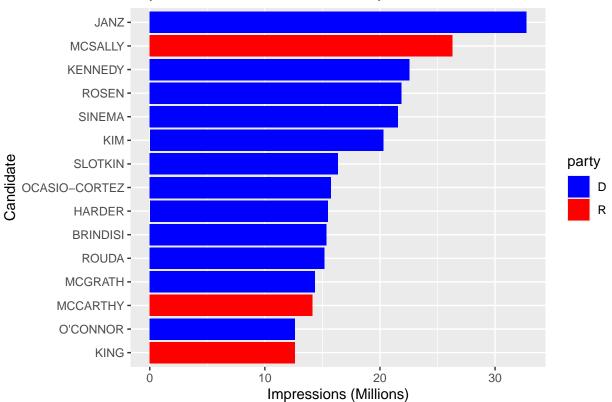
Answer 4

library(tidyverse)

Warning: package 'tidyverse' was built under R version 4.3.3

```
## -- Attaching core tidyverse packages ----
                                                                   ----- tidyverse 2.0.0 --
## v forcats
             1.0.0
                      v stringr
                                    1.5.1
                        v tibble
## v lubridate 1.9.3
                                    3.2.1
              1.0.2
                                    1.3.1
## v purrr
                        v tidyr
## v readr
              2.1.5
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
fb_top_plot <- fb_ads |>
 filter(office == "US House") |>
  slice_max(impressions_millions, n = 15) |>
  ggplot(mapping = aes(x = impressions_millions,
                                     y = fct_reorder(cand_name_last, impressions_millions))) +
 geom_col(mapping = aes(fill = party)) +
  scale_fill_manual(values = c(R = "red", D = "blue")) +
 labs( x = "Impressions (Millions)",
       y = "Candidate",
       title = "Top 15 Candidates with Most Impressions")
fb_top_plot
```





Code

```
options(width = 100)
library(dplyr)
fb_ads <- read.csv("data/fb_ads.csv")</pre>
incumbency_table <- fb_ads |>
  count(incumbency) |>
  knitr::kable(col.names = c("Incumbency Status", "Amount"))
incumbency_table
party_incumbent_promote <- fb_ads |>
  filter(office == "US House" | office == "US Senate") |>
  group_by(party, incumbency) |>
  summarize(promote_prop = mean(ad_tone_promote, na.rm = TRUE)) |>
  knitr::kable(party_incumbent_promote, col.names = c("Party", "Incumbency", "Promotion"), digits = 3)
party_incumbent_promote
library(ggplot2)
fb ads <- fb ads |>
  mutate(impressions_millions = impressions/1000000)
plot_q3 <- fb_ads |>
  filter(office == "US House") |>
  ggplot(mapping = aes(x = impressions_millions)) +
  geom_histogram()
plot_q3
library(tidyverse)
fb_top_plot <- fb_ads |>
  filter(office == "US House") |>
  slice_max(impressions_millions, n = 15) |>
  ggplot(mapping = aes(x = impressions_millions,
                                       y = fct_reorder(cand_name_last, impressions_millions))) +
  geom_col(mapping = aes(fill = party)) +
  scale_fill_manual(values = c(R = "red", D = "blue")) +
  labs( x = "Impressions (Millions)",
        y = "Candidate",
        title = "Top 15 Candidates with Most Impressions")
fb_top_plot
```