Final Project* PSTAT 231

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Contents

- 1. What makes voter behavior prediction (and thus election forecasting) a hard problem?
- 2. What was unique to Nate Silver's approach in 2012 that allowed him to achieve good predictions?
- 3. What went wrong in 2016? What do you think should be done to make future predictions better?
- 4. Remove summary rows from election.raw data:
- Federal-level summary into a election_federal.
- State-level summary into a election_state.
- Only county-level data is to be in election.

```
election_raw <- read.csv(here("data", "election", "election.csv")) %>%
    as_tibble()

census_meta <- read.csv(here("data", "census", "metadata.csv"), sep = ";") %>%
    as_tibble()

census <- read.csv(here("data", "census", "census.csv")) %>%
    as_tibble() %>%
    mutate(CensusTract = as.factor(CensusTract))

election_federal <- election_raw %>%
    filter(fips == "US")

election_state <- election_raw %>%
    filter(state != "US", is.na(county))

election <- election_raw %>%
    filter(!is.na(county))
```

5. How many named presidential candidates were there in the 2016 election? Draw a bar chart of all votes received by each candidate

There were 31 explicitly mentioned presidential candidates, plus a category of None of these acandidates. Figure 1 shows the votes (on a log_{10} -scale) that each candidate received.

```
election_federal %>%
  group_by(candidate) %>%
  summarize(votes = sum(votes, na.rm = T)) %>%
  ungroup() %>%
  mutate(candidate = fct_reorder(.f = candidate, .x = votes)) %>%
  ggplot(aes(x = candidate, y = votes)) +
  geom_col() +
  coord_flip() +
```

^{*}Code available on GitHUb at: https://github.com/jcvdav/PSTAT231/tree/master/final_project

```
scale_y_continuous(trans = "log10") +
labs(x = "Candidate", y = "Votes (log-10 Scale)")
```

6. Create variables county_winner and state_winner by taking the candidate with the highest proportion of votes. Hint: to create county_winner, start with election, group by fips, compute total votes, and pct = votes/total. Then choose the highest row using top_n (variable state_winner is similar).

Selecting by pct

Selecting by pct

7. Draw county-level map by creating counties = map_data("county"). Color by county

```
my_fun <- function(long, lat) {</pre>
  # long <- data$long
  # lat <- data$lat
  data <- cbind(long, lat) %>%
    rbind(cbind(long[1], lat[1]))
  st_sfc(st_polygon(list(as.matrix(data))))
}
state_dictionary <- tibble(abb = state.abb, state = tolower(state.name))</pre>
states <- map_data("state") %>%
  group_by(group, region) %>%
  summarize(geometry = my_fun(long, lat)) %>%
  ungroup() %>%
  st_sf(crs = 4326) %>%
  left_join(state_dictionary, by = c("region" = "state"))
counties <- map_data("county") %>%
  group_by(group, region, subregion) %>%
  summarize(geometry = my_fun(long, lat)) %>%
  ungroup() %>%
  st_sf(crs = 4326)
ggplot(data = counties,
       mapping = aes(fill = subregion)) +
```

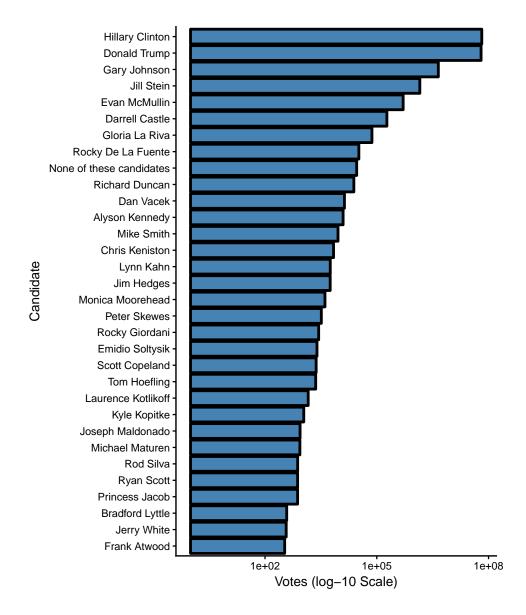
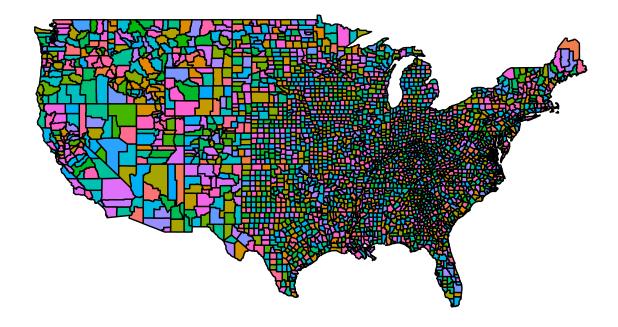


Figure 1: Number of votes that each presidential candidate received in the 2018 Presidential Elections.

```
geom_sf(data = counties) +
ggtheme_map() +
theme(legend.position = "None")
```



Now color the map by the winning candidate for each state. First, combine states variable and state_winner we created earlier using left_join(). Note that left_join() needs to match up values of states to join the tables; however, they are in different formats: e.g. AZ vs. arizona. Before using left_join(), create a common column by creating a new column for states named fips = state.abb[match(some_column, some_function(state.name))]. Replace some_column and some_function to complete creation of this new column. Then left_join(). Your figure will look similar to state_level New York Times map.

```
states %>%
  left_join(state_winner, by = c("abb" = "state")) %>%
  ggplot(mapping = aes(fill = winner)) +
  geom_sf(color = "white") +
  ggtheme_map() +
  scale_fill_brewer(palette = "Set1") +
  guides(fill = guide_legend(title = "Winner"))
```

