# 2024 General Election Forcasting Model

POLSCI 239 - Assignment Four

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### Methodology

The data for this model is borrowed from ABC's 538 general election state polling dataset. (Full citation in README)

```
polling_data <- read_csv("data/president_polls.csv", show_col_types = FALSE)
glimpse(polling_data)</pre>
```

```
Rows: 15,971
Columns: 52
$ poll_id
                       <dbl> 88806, 88806, 88836, 88836, 88817, 88817, 88~
$ pollster_id
                       <dbl> 770, 770, 1895, 1895, 1741, 1741, 770, 770, ~
                       <chr> "TIPP", "TIPP", "Quantus Insights", "Quantus~
$ pollster
                       <dbl> NA, NA, 2184, 2184, NA, NA, NA, NA, NA, NA, NA, ~
$ sponsor_ids
                       <chr> NA, NA, "TrendingPolitics", "TrendingPolitic~
$ sponsors
                       <chr> "TIPP Insights", "TIPP Insights", "Quantus I~
$ display_name
$ pollster_rating_id
                       <dbl> 144, 144, 859, 859, 721, 721, 144, 144, 338,~
$ pollster_rating_name
                       <chr> "TIPP Insights", "TIPP Insights", "Quantus I~
$ numeric_grade
                       <dbl> 1.8, 1.8, NA, NA, NA, NA, 1.8, 1.8, 0.7, 0.7~
$ pollscore
                       <dbl> -0.4, -0.4, NA, NA, NA, -0.4, -0.4, 0.6,~
$ methodology
                       <chr> "Online Panel", "Online Panel", "Online Pane~
                       <dbl> 3.0, 3.0, 5.5, 5.5, 8.0, 8.0, 3.0, 3.0, 4.0,~
$ transparency score
                       <chr> NA, NA, "Pennsylvania", "Pennsylvania", "Flo~
$ state
                       <chr> "10/18/24", "10/18/24", "10/17/24", "10/17/2~
$ start date
                       <chr> "10/20/24", "10/20/24", "10/20/24", "10/20/2~
$ end date
$ sponsor_candidate_id
                       $ sponsor_candidate
                       $ sponsor_candidate_party
                       $ endorsed_candidate_id
                       $ endorsed_candidate_name
```

```
$ endorsed_candidate_party
                      <dbl> 213459, 213459, 213538, 213538, 213472, 2134~
$ question_id
$ sample_size
                      <dbl> 1244, 1244, 840, 840, 400, 400, 1254, 1254, ~
$ population
                      <chr> "lv", "lv", "lv", "lv", "lv", "lv", "lv", "lv", "l~
$ subpopulation
                       <chr> "lv", "lv", "lv", "lv", "lv", "lv", "lv", "lv", "l~
$ population_full
$ tracking
                      <lgl> TRUE, TRUE, NA, NA, NA, TRUE, TRUE, NA, ~
                      <chr> "10/21/24 08:43", "10/21/24 08:43", "10/21/2~
$ created_at
                      $ notes
$ url
                      <chr> "https://tippinsights.com/tipp-tracking-poll~
                      <chr> "https://tippinsights.com/tipp-tracking-poll~
$ url_article
                      <chr> NA, NA, "https://docs.google.com/document/d/~
$ url_topline
                      $ url_crosstab
$ source
                      $ internal
                      <lgl> NA, NA, FALSE, FALSE, FALSE, FALSE, NA, NA, ~
                      <chr> NA, NA, "REP", "REP", NA, NA, NA, NA, "REP",~
$ partisan
$ race_id
                      <dbl> 8914, 8914, 8872, 8872, 8778, 8778, 8914, 89~
$ cycle
                      <dbl> 2024, 2024, 2024, 2024, 2024, 2024, 2024, 20~
$ office_type
                      <chr> "U.S. President", "U.S. President", "U.S. Pr~
$ seat number
                      $ seat name
                      <chr> "11/5/24", "11/5/24", "11/5/24", "11/5/24", ~
$ election date
$ stage
                      <chr> "general", "general", "general", "general", ~
$ nationwide_batch
                      <lgl> FALSE, FALSE, FALSE, FALSE, FALSE, FA~
$ ranked_choice_reallocated <lgl> FALSE, FALSE, FALSE, FALSE, FALSE, FALSE, FA-
$ ranked_choice_round
                      <lg1> FALSE, FALSE, FALSE, FALSE, FALSE, FA-
$ hypothetical
                       <chr> "DEM", "REP", "DEM", "REP", "DEM", "REP", "D~
$ party
                      <chr> "Harris", "Trump", "Harris", "Trump", "Harri~
$ answer
$ candidate_id
                      <dbl> 16661, 16651, 16661, 16651, 16661, 16651, 16~
$ candidate_name
                      <chr> "Kamala Harris", "Donald Trump", "Kamala Har~
                      <dbl> 47.0, 48.0, 48.2, 50.3, 45.4, 54.6, 47.0, 49~
$ pct
```

# **Data Cleaning**

The model will only calculate win percentages for toss up states.

```
toss_up_states <- c("Michigan", "Nevada",
      "Arizona", "New Mexico",
      "Wisconsin", "Pennsylvania",
      "North Carolina", "Georgia")
polling_data <- polling_data |>
  select(
    poll_id,
    state,
   end_date,
    sample_size,
    candidate_name,
    pct
  ) |>
  rename("dem_pct" = pct) |>
  filter(candidate_name == "Kamala Harris" & state %in% toss_up_states) |>
  mutate(end_date = as.Date(end_date, format = "%m/%d/%y")) |>
  arrange(end_date) |>
  drop_na(sample_size)
glimpse(polling_data)
```

## **Summary Statistics**

```
options(pillar.sigfig = 7)

polling_data |>
    group_by(state) |>
    summarise(
    poll_count = n(),
    raw_harris_approval = mean(dem_pct),
    ealiest_poll = min(end_date),
    most_recent_poll = max(end_date)
)
```

```
# A tibble: 8 x 5
 state
               poll_count raw_harris_approval ealiest_poll most_recent_poll
 <chr>
                    <int>
                                       <dbl> <date> <date>
1 Arizona
                                  46.50595 2023-11-03 2024-10-18
                      111
2 Georgia
                      119
                                   46.41235 2023-11-03 2024-10-18
3 Michigan
                                    47.50210 2023-11-03 2024-10-18
                      124
4 Nevada
                      80
                                    46.96812 2023-11-03 2024-10-18
5 New Mexico
                                   48.96
                                            2024-08-03 2024-10-18
                      10
6 North Carolina
                                   47.18766 2024-02-16 2024-10-18
                      111
7 Pennsylvania
                      166
                                   47.52759 2023-11-03 2024-10-20
8 Wisconsin
                      126
                                   48.46984 2023-11-03
                                                        2024-10-18
```

# Weighting Data by Sample Size

Each poll was weighted using a function based on its sample size. Specifically, I take the square root of the median sample size for each state and then multipled the Harris approval percentage for each poll by the square root of the poll's sample size divided by that states' square-rooted median sample size. This methodology was adopted from 538's weighting guidelines and then adjusted to fit the specifications of the dataset. A new "adjusted\_pct" variable was applied to each poll in the dataset.

```
options(pillar.sigfig = 7)

square_root_median_sample_size_by_state <- polling_data |>
    group_by(state) |>
    summarize(
    square_root_median_sample_size = sqrt(median(sample_size, na.rm = TRUE))
    )

as_tibble(square_root_median_sample_size_by_state)
```

```
# A tibble: 8 x 2
  state
                 square_root_median_sample_size
  <chr>
                                            <dbl>
1 Arizona
                                         27.85678
2 Georgia
                                         28.26659
3 Michigan
                                         26.22975
4 Nevada
                                         26.01922
5 New Mexico
                                         22.94559
6 North Carolina
                                         28.28427
7 Pennsylvania
                                         28.33725
8 Wisconsin
                                         26.45751
```

```
polling_data <- polling_data |>
  mutate(adjusted_pct = case_when(
    state == "Arizona" ~ sqrt(sample_size)/27.85678*dem_pct,
    state == "Georgia" ~ sqrt(sample_size)/28.26659*dem_pct,
    state == "Michigan" ~ sqrt(sample_size)/26.22975*dem_pct,
    state == "Nevada" ~ sqrt(sample_size)/26.01922*dem_pct,
    state == "New Mexico" ~ sqrt(sample_size)/22.94559*dem_pct,
    state == "North Carolina" ~ sqrt(sample_size)/28.28427*dem_pct,
    state == "Pennsylvania" ~ sqrt(sample_size)/28.33725*dem_pct,
    state == "Wisconsin" ~ sqrt(sample_size)/26.45751*dem_pct
)
```

)

#### glimpse(polling\_data)

### **Exponentially Weighted Moving Average**

In an EWMA calculation, recent data points are assigned more weight than older points. This makes the average more responsive to recent changes in the data. The lambda variable controls how much weight is assigned to more recent data points. I assign a lambda value of 0.95 in order to assign greater importance to more recent polls. The smoothed average provides the a value for Harris' approval rating that is then used as the Dem win percentage in my forcast. More documentation on the EWMA can be found here (https://www.investopedia.com/articles/07/ewma.asp).

```
options(pillar.sigfig = 7)
calculate_ewma <- function(data, raw_average, lambda) {</pre>
  ewma <- numeric(length(data[[raw average]]))</pre>
  ewma[1] <- data[[raw_average]][1]</pre>
  for (i in 2:length(data[[raw_average]])) {
    \verb|ewma[i]| <- lambda * data[[raw_average]][i] + (1 - lambda) * ewma[i - 1]|
  return(ewma[length(ewma)])
polling_data |>
  group_by(state) |>
  summarise(
    ewma_adjusted_pct = calculate_ewma(cur_data(), "adjusted_pct", 0.95)
Warning: There was 1 warning in `summarise()`.
i In argument: `ewma_adjusted_pct = calculate_ewma(cur_data(), "adjusted_pct",
  0.95).
i In group 1: `state = "Arizona"`.
Caused by warning:
! `cur_data()` was deprecated in dplyr 1.1.0.
i Please use `pick()` instead.
```

4	Nevada	41.80885
5	New Mexico	68.36112
6	North Carolina	47.53028
7	Pennsylvania	49.37062
8	Wisconsin	44.52851

#### **Additional Considerations and Data Limitations**

This dataset introduces several inconsistencies to the model which will be addressed here. First, the inconsistent number of polls conducted within each state creates uncertainty in the accuracy of the data. Second, the variability of polling sources opens the data to potential bias. FiveThirtyEight uses extensive guidelines when choosing polls to include within their data in order to account for bias; however, this is mostly a subjective science and isn't statistically grounded in my model. Information on 538's polling policy can be found here (https://fivethirtyeight.com/features/polls-policy-and-faqs/). Third, this model uses a ruidmentary modeling algorithm that adjusts based on sample size and time decay. Other weights such as pollster rating and margin of error are common strategies, but are not considered in this model.

Weighting and averaging data admits a certain level of subjectivity into the data as the methods by which the data is adjusted are largely statistically insignificant. The weighting and averaging methods I chose were subjective choices influenced by common practice but are not scientifically grounded as the best practice.