

My title*

My subtitle if needed

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October 24, 2024

First sentence. Second sentence. Third sentence. Fourth sentence.

```
#install.packages("tidymodels")
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.5
v forcats    1.0.0      v stringr    1.5.1
v ggplot2     3.5.1      v tibble     3.2.1
v lubridate  1.9.3      v tidyr      1.3.1
v purrr       1.0.2
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
library(ggplot2)
analysis_data <- read_csv(here::here("data/02-analysis_data/cleaned_US_voting.csv"))
```

Rows: 1683 Columns: 11

```
-- Column specification -----
Delimiter: ","
chr  (5): pollster_rating_name, methodology, state, candidate_name, populati...
dbl  (4): numeric_grade, sample_size, percent, transparency_score
date (2): start_date, end_date
```

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.

*Code and data are available at: [<https://github.com/wyx827/2024USpresidentialelection.git>].

1 Introduction

Overview paragraph

Estimand paragraph

Results paragraph

Why it matters paragraph

Telegraphing paragraph: The remainder of this paper is structured as follows. Section [2](#)....

2 Data

2.1 Overview

We use the statistical programming language R (R Core Team 2023).... Following Alexander (2023), we consider...

Overview text

2.2 Measurement

Some paragraphs about how we go from a phenomena in the world to an entry in the dataset.

2.3 Outcome variables

Add graphs, tables and text. Use sub-sub-headings for each outcome variable or update the subheading to be singular.

Some of our data is of penguins (fig-bills),

Talk more about it.

And also planes (fig-planes). (You can change the height and width, but don't worry about doing that until you have finished every other aspect of the paper - Quarto will try to make it look nice and the defaults usually work well once you have enough text.)

Talk way more about it.

2.4 Predictor variables

Add graphs, tables and text.

Use sub-sub-headings for each outcome variable and feel free to combine a few into one if they go together naturally.

3 Model

To model Donald Trump’s polling percentages over time, we employed a Bayesian linear regression framework. Bayesian methods provide a flexible approach to inference by incorporating prior beliefs and updating them with observed data, allowing us to quantify uncertainty in both parameter estimates and predictions.

Here we briefly describe the Bayesian model used to investigate the winning probability of Trump. Background details and diagnostics are included in [Appendix D](#).

3.1 Multiple Linear Regression Model Overview

The model now predicts Trump’s polling percentage (percent) using the following predictors:

- Numeric Grade (numeric_grade): Reflects the quality rating of the pollster.
- Sample Size (sample_size): The number of respondents in the poll.
- State (state): A categorical variable for different U.S. states.
- Transparency Score (transparency_score): A measure of how transparent the polling data and methodology are.
- End Date (end_date): The date the poll was completed, which might capture trends over time.

The model takes the form: The model takes the form:

3.2 Interpretation of Coefficients:

- Intercept (β_0): This is the predicted Trump polling percentage when all predictors (numeric grade, sample size, state, transparency score, and end date) are at their baseline or zero value.
- Numeric Grade (β_1): This coefficient measures how much Trump’s polling percentage changes as the pollster’s numeric grade increases. A positive and significant coefficient would indicate that higher-rated pollsters report better polling numbers for Trump, while a negative coefficient would suggest the opposite.

- Sample Size (_2): This measures the impact of the number of respondents on Trump’s polling percentage. A positive coefficient would indicate that larger sample sizes are associated with higher polling percentages for Trump.
- State (_3): The coefficients for the state variable represent differences in Trump’s polling percentage in each state compared to the reference state (baseline category). For example, if the coefficient for Florida is negative, it means Trump polls lower in Florida compared to the reference state.
- Transparency Score (_4): This coefficient shows how much Trump’s polling percentage is affected by the transparency of the poll. A positive coefficient would indicate that polls with higher transparency tend to report higher polling percentages for Trump, whereas a negative coefficient would imply the opposite.
- End Date (_5): The end date is a time-related variable, capturing trends over time. A positive and significant coefficient would suggest that Trump’s polling percentage has increased as the election date approaches, while a negative coefficient would suggest a decrease in his polling numbers over time.

3.3 Interpretation

The posterior distributions of the parameters allow us to quantify the uncertainty around each effect: - The coefficient for end_date informs us about how Trump’s polling percentages have evolved over time. A positive coefficient would suggest an upward trend, while a negative coefficient would indicate a decline. - The coefficient for numeric_grade captures the impact of pollster quality on the polling percentage. High-quality pollsters may produce different estimates compared to lower-quality ones. - The state-level effects account for regional differences in Trump’s support. Some states may show significantly higher or lower levels of support, even after adjusting for the time of the poll and pollster quality.

4 Model Evalutation

- R-squared Table 1 shows the summary table for the model. And then evaluate it on the test set. It appears as though the model is having difficulty identifying Trump supporters.

Table 1: Relationship between wing length and width

Table 1: Regression Model Coefficients

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-67.184	14.445	-4.651	0.000
numeric_grade	0.914	1.006	0.909	0.364

	Estimate	Std. Error	t value	Pr(> t)
sample_size	0.000	0.000	0.205	0.838
stateArizona	-5.020	2.582	-1.944	0.052
stateArkansas	5.569	4.380	1.272	0.204
stateCalifornia	-18.345	2.738	-6.699	0.000
stateColorado	-13.340	3.002	-4.443	0.000
stateConnecticut	-13.480	3.574	-3.772	0.000
stateFlorida	-3.214	2.740	-1.173	0.241
stateGeorgia	-4.814	2.573	-1.871	0.062
stateIdaho	3.467	4.379	0.792	0.429
stateIllinois	-14.064	3.275	-4.294	0.000
stateIndiana	0.902	3.268	0.276	0.783
stateIowa	-4.692	3.101	-1.513	0.131
stateKansas	-0.798	3.275	-0.244	0.808
stateMaine	-10.182	2.726	-3.735	0.000
stateMaryland	-20.244	2.998	-6.753	0.000
stateMassachusetts	-22.543	2.753	-8.190	0.000
stateMichigan	-6.635	2.576	-2.576	0.010
stateMinnesota	-10.250	2.684	-3.819	0.000
stateMissouri	1.087	2.803	0.388	0.698
stateMontana	1.379	2.784	0.495	0.620
stateNational	-7.935	2.546	-3.117	0.002
stateNebraska	-7.528	2.782	-2.706	0.007
stateNevada	-5.617	2.594	-2.165	0.031
stateNew Hampshire	-8.778	2.655	-3.306	0.001
stateNew Jersey	-15.189	3.577	-4.246	0.000
stateNew Mexico	-10.980	3.099	-3.543	0.000
stateNew York	-15.239	2.633	-5.789	0.000
stateNorth Carolina	-4.783	2.582	-1.853	0.064
stateNorth Dakota	2.970	4.380	0.678	0.498
stateOhio	-2.796	2.677	-1.045	0.296
stateOklahoma	8.450	4.385	1.927	0.054
stateOregon	-14.974	4.385	-3.415	0.001
statePennsylvania	-6.375	2.568	-2.483	0.013
stateRhode Island	-13.374	2.994	-4.467	0.000
stateSouth Carolina	-1.939	4.376	-0.443	0.658
stateSouth Dakota	0.394	3.096	0.127	0.899
stateTennessee	3.624	4.378	0.828	0.408
stateTexas	-3.825	2.681	-1.427	0.154
stateUtah	-3.243	4.385	-0.740	0.460
stateVermont	-23.452	3.581	-6.549	0.000
stateVirginia	-9.518	2.663	-3.575	0.000

	Estimate	Std. Error	t value	Pr(> t)
stateWashington	-15.876	3.264	-4.864	0.000
stateWest Virginia	8.265	4.379	1.887	0.059
stateWisconsin	-6.444	2.571	-2.506	0.012
stateWyoming	15.921	4.377	3.637	0.000
transparency_score	-0.646	0.088	-7.383	0.000
end_date	0.006	0.001	8.880	0.000

5 Diagnostic plots for the model

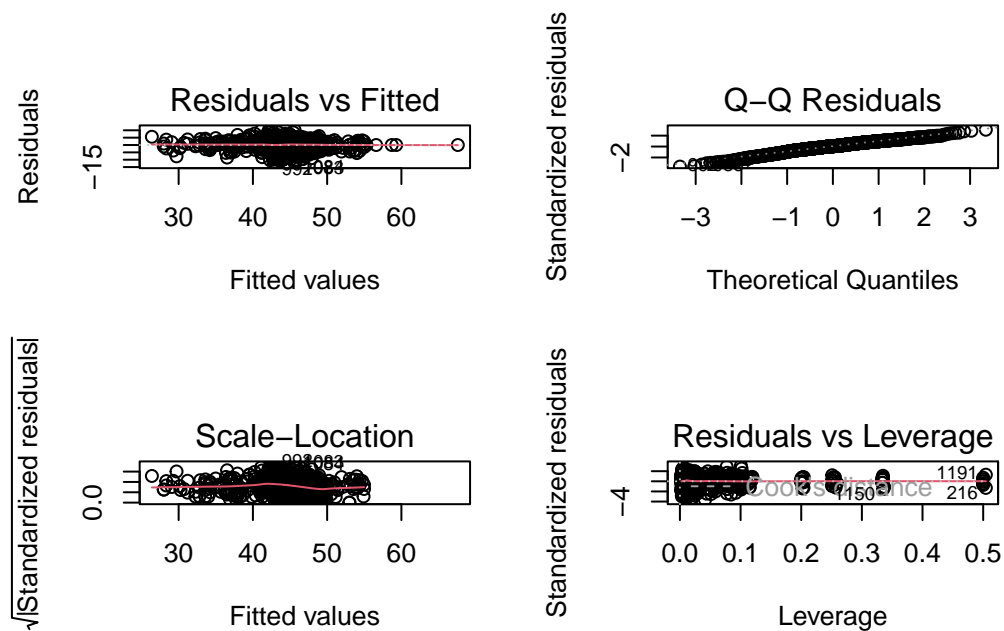


Figure 1

```
library(car)
```

Loading required package: carData

Attaching package: 'car'

The following object is masked from 'package:dplyr':

```
recode
```

The following object is masked from 'package:purrr':

```
some
```

```
vif(regression_model)
```

	GVIF	Df	GVIF ^{1/(2*Df)}
numeric_grade	1.294110	1	1.137590
sample_size	1.289750	1	1.135672
state	2.082151	44	1.008369
transparency_score	1.263558	1	1.124081
end_date	1.309491	1	1.144330

```
summary(regression_model)
```

Call:

```
lm(formula = percent ~ numeric_grade + sample_size + state +  
    transparency_score + end_date, data = analysis_data)
```

Residuals:

Min	1Q	Median	3Q	Max
-12.813	-1.999	0.215	2.477	10.401

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-6.718e+01	1.444e+01	-4.651	3.68e-06	***
numeric_grade	9.142e-01	1.006e+00	0.909	0.363667	
sample_size	4.266e-05	2.082e-04	0.205	0.837710	
stateArizona	-5.020e+00	2.582e+00	-1.944	0.052119	.
stateArkansas	5.569e+00	4.380e+00	1.272	0.203739	
stateCalifornia	-1.834e+01	2.738e+00	-6.699	3.25e-11	***
stateColorado	-1.334e+01	3.002e+00	-4.443	9.70e-06	***
stateConnecticut	-1.348e+01	3.574e+00	-3.772	0.000170	***
stateFlorida	-3.214e+00	2.740e+00	-1.173	0.240935	
stateGeorgia	-4.814e+00	2.573e+00	-1.871	0.061596	.
stateIdaho	3.467e+00	4.379e+00	0.792	0.428696	
stateIllinois	-1.406e+01	3.275e+00	-4.294	1.90e-05	***

stateIndiana	9.016e-01	3.268e+00	0.276	0.782695	
stateIowa	-4.692e+00	3.101e+00	-1.513	0.130540	
stateKansas	-7.979e-01	3.275e+00	-0.244	0.807553	
stateMaine	-1.018e+01	2.726e+00	-3.735	0.000197	***
stateMaryland	-2.024e+01	2.998e+00	-6.753	2.27e-11	***
stateMassachusetts	-2.254e+01	2.753e+00	-8.190	6.77e-16	***
stateMichigan	-6.635e+00	2.576e+00	-2.576	0.010130	*
stateMinnesota	-1.025e+01	2.684e+00	-3.819	0.000141	***
stateMissouri	1.087e+00	2.803e+00	0.388	0.698117	
stateMontana	1.379e+00	2.784e+00	0.495	0.620435	
stateNational	-7.935e+00	2.546e+00	-3.117	0.001871	**
stateNebraska	-7.528e+00	2.782e+00	-2.706	0.006915	**
stateNevada	-5.617e+00	2.594e+00	-2.165	0.030551	*
stateNew Hampshire	-8.778e+00	2.655e+00	-3.306	0.000974	***
stateNew Jersey	-1.519e+01	3.577e+00	-4.246	2.35e-05	***
stateNew Mexico	-1.098e+01	3.099e+00	-3.543	0.000412	***
stateNew York	-1.524e+01	2.633e+00	-5.789	9.10e-09	***
stateNorth Carolina	-4.783e+00	2.582e+00	-1.853	0.064166	.
stateNorth Dakota	2.970e+00	4.380e+00	0.678	0.497805	
stateOhio	-2.796e+00	2.677e+00	-1.045	0.296409	
stateOklahoma	8.450e+00	4.385e+00	1.927	0.054220	.
stateOregon	-1.497e+01	4.385e+00	-3.415	0.000659	***
statePennsylvania	-6.375e+00	2.568e+00	-2.483	0.013175	*
stateRhode Island	-1.337e+01	2.994e+00	-4.467	8.70e-06	***
stateSouth Carolina	-1.939e+00	4.376e+00	-0.443	0.657716	
stateSouth Dakota	3.944e-01	3.096e+00	0.127	0.898654	
stateTennessee	3.624e+00	4.378e+00	0.828	0.407898	
stateTexas	-3.825e+00	2.681e+00	-1.427	0.153882	
stateUtah	-3.243e+00	4.385e+00	-0.740	0.459694	
stateVermont	-2.345e+01	3.581e+00	-6.549	8.64e-11	***
stateVirginia	-9.518e+00	2.663e+00	-3.575	0.000365	***
stateWashington	-1.588e+01	3.264e+00	-4.864	1.31e-06	***
stateWest Virginia	8.265e+00	4.379e+00	1.887	0.059354	.
stateWisconsin	-6.444e+00	2.571e+00	-2.506	0.012344	*
stateWyoming	1.592e+01	4.377e+00	3.637	0.000288	***
transparency_score	-6.461e-01	8.752e-02	-7.383	2.93e-13	***
end_date	6.148e-03	6.923e-04	8.880	< 2e-16	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.564 on 1174 degrees of freedom

Multiple R-squared: 0.5165, Adjusted R-squared: 0.4968

F-statistic: 26.13 on 48 and 1174 DF, p-value: < 2.2e-16

5.1 Model justification

We expect a positive relationship between the size of the wings and time spent aloft. In particular...

We can use maths by including latex between dollar signs, for instance θ .

6 Results

6.1 Predicted Electoral Outcomes

We applied a regression model to predict the percentage of votes Trump is expected to receive in each state. The model results, combined with each state's electoral vote allocation, allowed us to predict the winner in each state. Based on this, we calculated the total number of electoral votes for both Trump and Harris.

The table below (Table 2) summarizes the predicted results, showing Trump's predicted percentage, the number of electoral votes in each state, and the predicted winner (either Trump or Harris). For instance: - Alabama: Trump is predicted to win 53% of the vote, securing all 9 electoral votes. - California: Trump is predicted to receive 34.39% of the vote, resulting in a victory for Harris, who takes California's 55 electoral votes. - Florida: The model predicts a close race, with Trump at 48.89% of the vote, resulting in a Harris win in this critical battleground state.

Electoral Vote Count: - Trump Electoral Votes: 78 - Harris Electoral Votes: 416 These results indicate that based on the current model predictions, Harris is expected to win the 2024 U.S. Presidential Election, securing 416 electoral votes, compared to Trump's 78 electoral votes. The election outcome hinges on several battleground states, where the vote margins are predicted to be narrow.

[1] "Trump Electoral Votes: 78"

[1] "Harris Electoral Votes: 416"

Table 2: Prediction for Trump

Table 2: Prediction for Trump by State

State	Trump Predicted %	Electoral Votes	Winner
Alaska	53.00	3	Trump
Arizona	47.62	11	Harris
Arkansas	56.60	6	Trump
California	34.39	55	Harris
Colorado	37.60	9	Harris
Connecticut	39.09	7	Harris
Florida	48.89	29	Harris
Georgia	47.71	16	Harris
Idaho	54.50	4	Trump
Illinois	36.37	20	Harris

State	Trump Predicted %	Electoral Votes	Winner
Indiana	53.73	11	Trump
Iowa	46.68	6	Harris
Kansas	49.63	6	Harris
Maine	41.13	2	Harris
Maryland	32.88	10	Harris
Massachusetts	29.80	11	Harris
Michigan	45.92	15	Harris
Minnesota	42.40	10	Harris
Missouri	52.51	10	Trump
Montana	54.20	3	Trump
Nebraska	45.16	5	Harris
Nevada	47.00	6	Harris
New Hampshire	42.86	4	Harris
New Jersey	37.60	14	Harris
New Mexico	42.05	5	Harris
New York	37.45	29	Harris
North Carolina	47.98	16	Harris
North Dakota	54.00	3	Trump
Ohio	49.58	18	Harris
Oklahoma	58.70	7	Trump
Oregon	35.30	6	Harris
Pennsylvania	46.28	20	Harris
Rhode Island	39.12	4	Harris
South Carolina	50.60	9	Trump
South Dakota	52.18	3	Trump
Tennessee	55.30	11	Trump
Texas	48.51	38	Harris
Utah	47.00	6	Harris
Vermont	28.00	3	Harris
Virginia	43.22	13	Harris
Washington	36.33	12	Harris
West Virginia	59.30	5	Trump
Wisconsin	45.96	10	Harris
Wyoming	67.60	3	Trump

6.2 Predicted Electoral Outcomes by State

The following map (Figure 2) shows the predicted winner for each state in the 2024 U.S. Presidential Election, based on the regression model's predicted vote percentages for Trump and Harris.

Predicted Winner of the 2024 U.S. Presidential Election by State

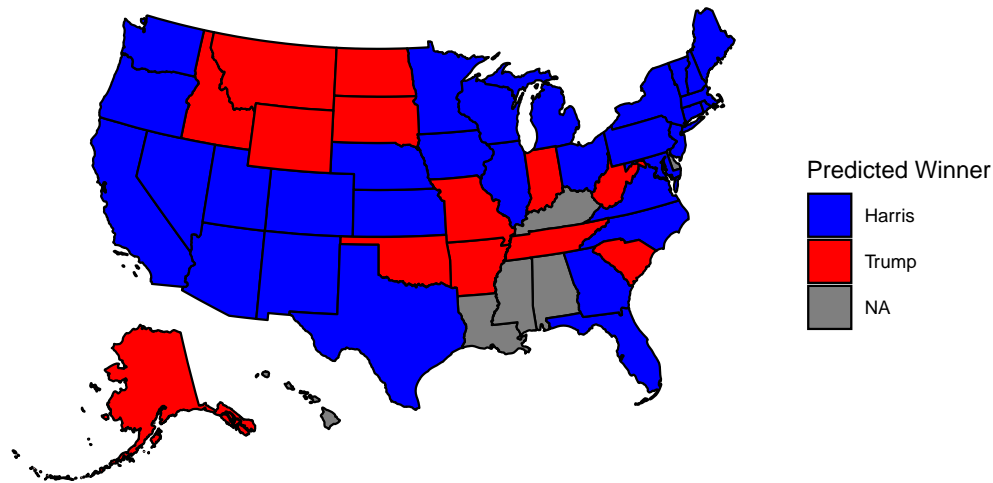


Figure 2

The predicted outcome in the regression model reflects the geographic voting patterns, with Trump winning in traditionally Republican-leaning states like Alabama, Missouri, and Wyoming, while Harris dominates in Democratic strongholds such as California, New York, and Illinois. However, key battleground states such as Florida and Arizona are predicted to favor Harris, potentially determining the overall election outcome.

7 Discussion

7.1 What Is Done in This Paper?

This paper uses a regression model to predict the outcome of the 2024 U.S. Presidential Election between Donald Trump and Kamala Harris. By analyzing historical voting data, the model forecasts the percentage of votes Trump is expected to receive in each state. These predictions are combined with the state's electoral vote allocation to determine the overall winner. The findings indicate a significant electoral victory for Harris, projecting 416 electoral votes for her compared to 78 for Trump. A geographic map is also presented to visualize state-level outcomes and highlight regional voting trends.

7.2 What Do We Learn About the World?

The study highlights the persistence of geographic voting patterns in U.S. elections. Even with changes in candidates, many states follow their historical voting tendencies. Republican-

leaning states like Alabama and Wyoming are projected to support Trump, while Democratic-leaning states such as California and New York favor Harris. This pattern shows the strong influence of regional political loyalties, which continue to shape election outcomes.

7.3 Another Thing We Learn About the World

The analysis underscores the importance of battleground states in determining the election's outcome. States like Florida and Arizona, predicted to lean toward Harris, play a decisive role in the projected electoral victory. This illustrates how close contests in a few states can influence the overall election result, emphasizing the significant impact these states have in shaping the presidency.

7.4 Weaknesses of What Was Done

While the model offers a useful prediction, it has limitations. It relies heavily on past voting data, which may not fully capture future changes in voter behavior or external factors that could affect the election. The model also simplifies the election by focusing on state-level percentages, without accounting for variables such as turnout, third-party candidates, or unexpected political shifts. Additionally, the assumption that past voting patterns will continue might not be accurate in a rapidly changing political landscape.

7.5 What Is Left to Learn or How Should We Proceed in the Future?

Future research should incorporate more detailed data, including demographic shifts and economic factors that may influence voter preferences. It would be valuable to examine how emerging political movements and changes in communication strategies affect elections, particularly in closely contested states. Post-election analysis comparing predictions with actual results will also be essential to improving the accuracy of models and understanding how unforeseen events may alter outcomes.

7.6 Weaknesses and next steps

The logistic regression model is limited by the lack of interaction terms between states, which may obscure regional trends or external factors that influence neighboring states. Additionally, the model assumes that polling data accurately reflects voter intentions, but biases or missing data in key regions could affect the predictions.

Future iterations of this model should incorporate voter turnout predictions, demographic variables, and real-time polling updates to better capture the dynamics of voter behavior. Including social media sentiment analysis and analyzing cross-state interactions could refine the prediction model, offering more nuanced insights into election forecasts.

Appendix

A Pollster Methodology Overview and Evaluation

A.1 Overview of SurveyUSA

SurveyUSA is a privately held opinion research company that operates nationwide, across all 50 U.S. states. Since its founding, the company has conducted over 40,000 research projects, serving a client base of 400 organizations, including media outlets, corporations, non-profits, government agencies, and academic institutions. Known for its expertise in localized opinion research, SurveyUSA focuses on gathering data at the city, county, and regional levels. The company offers timely, cost-effective surveys tailored to meet specific client needs, distinguishing itself from larger global firms.

A.2 Population, Frame, and Sample

- Target Population: U.S. citizens eligible to vote in the 2024 presidential election.
- Sample Frame: U.S. households with either home telephones or access to devices such as phones or tablets.
- Sample Size: Sample sizes vary across different polls. For the 2024 U.S. presidential election cycle, SurveyUSA conducted 49 polls, with sample sizes ranging from 507 to 2,330 for registered voters or likely voters. The average sample size for these polls is approximately 1,045 households.

A.3 Recruitment

SurveyUSA employs a mixed-method approach to recruitment, including online panels, telephone calls, and a text-to-web method. Some respondents are recruited through Random Digit Dialing (RDD) using telephone samples purchased from Aristotle, while others, who do not use home telephones, are invited to complete the survey on an electronic device such as a phone or tablet. Respondents from non-probability online panels are selected randomly by Cint/Lucid Holdings LLC.

A.4 Sampling approach and Trade-offs

SurveyUSA uses a blend of probability and non-probability sampling methods. Some respondents are drawn from non-probability online panels, while others are recruited using probability-based telephone sampling. Responses are weighted based on the latest U.S.

Census estimates for age, gender, ethnicity, and region, ensuring alignment with the target population. Questions and answer choices are rotated to reduce order bias, recency effects, and latency effects.

- **Advantages:**

The diverse sampling approach not only ensures a broad range of opinions is captured but also complements probability-based sampling, which accurately reflects the overall population. Furthermore, reweighting the data according to U.S. Census demographics strengthens the credibility of the results by ensuring demographic accuracy. Additionally, rotating questions and answer choices helps mitigate bias, further improving the reliability of the data. Finally, the use of online surveys offers a cost-effective solution for efficient data collection.

- **Disadvantages:**

Phone-based data collection tends to be time-consuming and can be affected by interviewer effects during telephone interviews. Additionally, challenges like non-response issues, such as busy signals or refusals to participate, can hinder the effectiveness of the data collection process.

A.5 Non-response Handling

In cases of non-response, SurveyUSA attempts follow-up calls if interviews are interrupted by answering machines or busy signals. Weighting is applied to adjust for non-response bias, although this doesn't completely eliminate challenges posed by unreachable or unwilling participants.

A.6 Questionnaire Evaluation

- **Positive Aspects:** A logical flow between questions facilitates easy navigation for respondents throughout the survey, while simple wording promotes inclusivity by enabling individuals from diverse backgrounds to comprehend the questions. Furthermore, all questions are directly relevant to analyzing the 2024 U.S. presidential election, and providing predefined response options simplifies the choices for participants.
- **Negative Aspects:** Static options for party affiliation and ideology may fail to capture the nuances of respondents' political beliefs. These rigid categories could oversimplify complex political identities.

A.7 Summary Evaluation

SurveyUSA’s methodology reflects a balanced approach, leveraging various sampling approach and method to reach a representative sample. While its blend of probability and non-probability methods has strengths, such as cost-effectiveness and broad reach, it faces challenges related to telephone interview logistics, potential interviewer bias, and the limitations of fixed questionnaire options. Nevertheless, the inclusion of data weighting and question rotation adds credibility to its results, making SurveyUSA a reliable pollster for localized opinion research.

B Appendix B: Idealized Methodology and Survey

B.1 Objective and Overview

The goal of this survey methodology is to accurately forecast the outcome of the U.S. presidential election by collecting high-quality, representative data from a diverse set of respondents across the country. With a budget of \$100,000, this methodology incorporates sophisticated sampling techniques, robust respondent recruitment strategies, and rigorous data validation protocols. The approach is designed to maximize accuracy, reduce bias, and account for various demographic, geographic, and political factors that influence voting behavior.

B.2 Core Objectives

- Obtain a representative sample of the U.S. electorate.
- Ensure data quality through rigorous validation.
- Leverage statistical modeling and poll aggregation for an accurate prediction.

B.3 Sampling Strategy

The sampling strategy is designed to ensure that the survey reaches a broad, representative section of the voting population. To achieve this, we will use **stratified random sampling** combined with **quota sampling** for key demographics. This ensures that each important subgroup within the population is adequately represented.

B.3.1 Stratification Variables

- **Age Groups:** 18-29, 30-44, 45-64, 65+
- **Gender:** Male, Female, Non-binary/Other
- **Race/Ethnicity:** White, Black, Hispanic/Latino, Asian, Indigenous, Other

- **Education Level:** No high school, High school graduate, College graduate, Post-graduate
- **Income Bracket:** <\$30,000, \$30,000-\$60,000, \$60,000-\$100,000, >\$100,000
- **Geographic Region:** Northeast, Midwest, South, West

B.3.2 Sample Size

A total of **10,000 respondents** will be surveyed, providing a margin of error of approximately $\pm 1\%$ at a 95% confidence level. This sample size will allow for detailed subgroup analysis (e.g., by state, demographic group), yielding statistically robust predictions.

B.3.3 Weighting

Post-stratification weights will be applied to adjust for any oversampling or undersampling of specific demographic groups. For example, younger voters or underrepresented minorities will be weighted to reflect their true proportions in the voting population.

B.4 Recruitment Strategy

To maximize respondent diversity and ensure accurate sampling, the survey will employ **multi-channel recruitment**:

- **Digital Advertisements:** Targeted ads on platforms like Facebook, Instagram, and Google will recruit respondents based on their demographic profiles (age, gender, location, political interest).
- **Email Outreach:** If permissible, voter registration databases will be accessed to send email invitations to registered voters.
- **Partnerships with Civic Organizations:** Partnering with non-profits and civic organizations that engage diverse communities will further boost respondent diversity.
- **Incentives:** Each participant will be entered into a lottery with a chance to win a \$100 gift card to encourage participation.

B.5 Data Validation and Quality Assurance

Maintaining data integrity and ensuring high-quality responses are critical to the accuracy of the election forecast. Several measures will be put in place to validate responses and reduce noise in the dataset.

B.5.1 Data Validation Protocols

- **Real-time Captcha Verification:** This will prevent automated bots from submitting responses.
- **Email/Phone Verification:** Respondents will verify their email or phone number to ensure authenticity and prevent duplicate submissions.
- **Time on Task Monitoring:** The survey platform will monitor the time respondents spend on each question. Responses completed suspiciously quickly will be flagged for review.
- **Voter Registration Cross-Check:** If feasible, respondents will be cross-referenced with voter registration records to ensure eligibility.
- **Response Audits:** Randomly selected respondents will be contacted to verify the accuracy of their responses, ensuring dataset integrity.

B.6 Poll Aggregation and Data Analysis

B.6.1 Poll Aggregation

This survey will be combined with results from reputable polling firms (e.g., YouGov, Ipsos, Gallup) to strengthen the forecast through a **poll-of-polls** approach.

- **Weighting by Methodology and Recency:** Poll results will be weighted based on the rigor of their methodology and the recency of the poll.
- **Handling Bias and Variability:** Aggregated results will adjust for pollster biases and variability between polls to ensure that no single poll dominates the prediction.

B.6.2 Modeling Approach

Bayesian hierarchical models will account for variability across different states, demographics, and regions. This will allow for modeling the popular vote and potentially translating it into **Electoral College predictions**.

B.7 Budget Allocation

- **Respondent Recruitment (Targeted ads, outreach):** \$70,000
- **Incentives (e.g., lottery prizes):** \$10,000
- **Survey Platform (Google Forms, Qualtrics subscription):** \$5,000
- **Data Validation Tools:** \$5,000
- **Poll Aggregation & Analysis Software:** \$10,000

B.8 Survey Implementation

The survey will be implemented via **Google Forms**, which offers a cost-effective platform for data collection. You can access the survey at the following link: [Google Form Survey](#)

B.8.1 Survey Structure:

Introduction:

Thank you for taking part in this survey aimed at predicting the outcome of the 2024 US Presidential election. Your insights are valuable to our research.

Please note:

- **All responses will be kept strictly confidential.**
- **Your participation is entirely voluntary.**
- **We kindly request that you answer all questions honestly and to the best of your knowledge.**
- **The survey is estimated to take approximately 10 minutes to complete.**

If you have any inquiries or concerns regarding this survey, please don't hesitate to contact the research team at shaw.wei@mail.utoronto.ca. (Yuxuan Wei, Xuanle Zhou, Yongqi Liu)

Your contribution to this study is greatly appreciated! Each participant will be entered into a lottery with a chance to win a \$100 gift card!

Section 1: Eligibility Screening:

Are you a U.S. citizen? - Yes - No [If No, end survey]

Will you be 18 or older by Election Day (November 5, 2024)? - Yes - No [If No, end survey]

Are you registered to vote in the United States? - Yes - No - Not sure - Plan to register before the election

Section 2: Demographic Information:

What is your age group? - 18-29 - 30-44 - 45-64 - 65 or older - Prefer not to say

What is your gender? - Male - Female - Non-binary/Other - Prefer not to say

What is your race/ethnicity? (Select all that apply) - White - Black or African American - Hispanic or Latino - Asian - American Indian or Alaska Native - Native Hawaiian or Pacific Islander - Prefer not to say - Other: [Short text answer]

What is your highest level of education completed? - No high school - High school graduate or equivalent - Some college, no degree - Bachelor's degree - Graduate or professional degree - Prefer not to say

What was your total household income in 2023? - Less than \$30,000 - \$30,000 - \$59,999 - \$60,000 - \$99,999 - \$100,000 - \$149,999 - \$150,000 or more - Prefer not to say

In which region of the United States do you currently reside? - Northeast (ME, NH, VT, MA, RI, CT, NY, NJ, PA) - Midwest (OH, IN, IL, MI, WI, MN, IA, MO, ND, SD, NE, KS) - South (DE, MD, DC,

VA, WV, NC, SC, GA, FL, KY, TN, AL, MS, AR, LA, OK, TX) - West (MT, ID, WY, CO, NM, AZ, UT, NV, WA, OR, CA, AK, HI)

Section 3: Political Views and Voting Intentions:

How likely are you to vote in the 2024 Presidential election? - Definitely will vote - Probably will vote - Might or might not vote - Probably will not vote - Definitely will not vote

Generally speaking, do you usually think of yourself as a: - Democrat - Republican - Independent - Prefer not to say - Other: [Short text answer]

If the 2024 Presidential election were held today, who would you vote for? - Kamala Harris (Democrat) - Donald Trump (Republican) - Undecided - Prefer not to say - Other: [Short text answer]

How certain are you about your choice? - Very certain - Somewhat certain - Not very certain - Not at all certain - Prefer not to say

Which THREE issues are most important to you in deciding your vote? (Select exactly three) - Economy and jobs - Healthcare - Immigration - Climate change - National security - Education - Gun policy - Social justice/racial equality - Taxes - Crime and public safety - Foreign policy - Other: [Short text answer]

Section 4: Information Sources and Engagement:

What is your primary source of political news? (Select all that apply) - Network TV news (ABC, CBS, NBC) - Cable TV news (CNN, Fox News, MSNBC) - News websites - Social media - Radio - Print newspapers - Friends and family - Other: [Short text answer]

How closely have you been following news about the 2024 Presidential election? - Very closely - Somewhat closely - Not too closely - Not at all - Not sure

Section 5: Validation and Consent:

Please verify that you are a human by selecting “Blue” from the following options: - Red - Green - Blue - Yellow

Consent Statement: “I understand that my participation in this survey is voluntary and that my responses will be kept confidential. I agree that my responses may be used for research purposes.” - Yes, I agree - No, I do not agree

Email Address: [Email field]

End Message:

“Thank you for completing this survey. Your response has been recorded. If you have any questions about this survey or would like to be informed about the results, please contact at shaw.wei@mail.utoronto.ca.”

B.9 Survey Design Considerations

- **Question Wording:** All questions are designed to avoid bias or leading responses.
- **Neutrality:** Political questions are framed neutrally to avoid influencing respondents' answers.
- **Pilot Testing:** The survey will undergo a pilot test to identify and resolve any issues before full deployment.

C Additional data details

D Model details

D.1 Posterior predictive check

Examining how the model fits, and is affected by, the data

D.2 Diagnostics

Checking the convergence of the MCMC algorithm

E Acknowledgements

R Core Team (2023)

References

- Alexander, Rohan. 2023. *Telling Stories with Data*. Chapman; Hall/CRC. <https://tellingstorieswithdata.com/>.
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