

# Forecasting the 2024 US Presidential Election\*

My subtitle if needed

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October 21, 2024

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

## 1 Introduction

*Overview paragraph:* Provide a brief overview of the context of the upcoming US presidential election and the importance of forecasting its results.

*Estimand paragraph:* Define the estimand clearly (e.g., predicting the probability of a candidate winning the election based on poll data).

*Results paragraph:* Summarize the key findings of the model, highlighting its predictive accuracy and implications.

*Why it matters paragraph:* Explain the broader significance of accurately forecasting election results for politics, society, and policymaking.

*Telegraphing paragraph:* The remainder of this paper is structured as follows: Section 2 details the data and measurement process; Section 3 covers model development and results; Section 4 discusses implications and future steps. The remainder of this paper is structured as follows. Section 2....

## 2 Data

### 2.1 Overview

We sourced the “Presidential General Election Polls” dataset from FiveThirtyEight (FiveThirtyEight 2024) and performed an in-depth analysis using the statistical programming language R (R Core Team 2023).

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\*Code and data are available at: <https://github.com/DavidFJ207/USPresidentialForecast>

Our primary objective was to clean, organize, and analyze U.S. presidential election polling data to provide insights into voter preferences by state. This involved selecting data from reputable pollsters, organizing it by state, and addressing any missing or incomplete entries. Additional datasets were merged where necessary, and after thoroughly addressing missing values, we finalized a polished dataset ready for analysis. Below is a table summarizing the polling data by state, showcasing voter preferences by party and demographic breakdowns. The table highlights key trends and differences across states in terms of party lean and demographic distribution. **?@fig-predictors-pct-summary** provides a snapshot of these findings.

Latest Poll Points by Party

State	Democrat Pct	Republican Pct
Arizona	47.6	51.7
Arkansas	24.4	56.6
California	60.8	38.0
Colorado	41.0	35.0
Connecticut	49.4	40.0
Florida	47.6	51.1
Georgia	48.2	50.3
Idaho	25.8	54.5
Illinois	43.4	33.7
Indiana	40.7	57.9
Iowa	39.6	48.2
Kansas	31.0	47.0
Kentucky	26.3	54.6
Maine	50.9	40.1
Maryland	65.3	33.3
Massachusetts	54.0	33.7
Michigan	50.0	49.1
Minnesota	51.0	49.0
Missouri	43.4	55.2
Montana	42.5	57.5
National	50.2	48.6
Nebraska	30.9	46.8
Nevada	48.8	49.0
New Hampshire	52.4	47.6
New Jersey	45.7	39.4
New Mexico	53.6	46.4
New York	55.5	41.6
North Carolina	49.1	49.9
North Dakota	17.0	54.0
Ohio	44.7	54.3
Oklahoma	27.0	55.0
Oregon	50.7	35.3
Pennsylvania	49.3	48.8
South Carolina	36.7	50.6
South Dakota	26.4	55.3
Tennessee	22.0	55.3
Texas	46.7	52.3
Utah	33.6	47.0
Virginia	52.7	46.2
Washington	49.4	38.7
West Virginia	22.5	59.3
Wisconsin	48.7	50.3
Wyoming	14.8	67.6

Figure 1

State	High Educated	Low Educated	Female	Male	Nonbinary	Caucasian	Minority Ethnicity	Young Age	Older Age
Arizona	22.1	24.3	51.8	46.4	1.8	80.2	19.8	41.6	58.4
Arkansas	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
California	18.6	25.5	49.3	49.2	1.5	87.9	12.1	49.6	50.3
Colorado	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Connecticut	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Florida	18.6	25.5	49.3	49.2	1.5	87.9	12.1	49.6	50.3
Georgia	23.8	24.1	49.8	49.5	0.7	86.5	13.5	41.3	58.6
Idaho	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Illinois	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Indiana	28.7	19.5	51.9	47.1	1.0	49.1	50.9	50.1	49.8
Iowa	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Kansas	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Kentucky	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Maine	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Maryland	18.6	25.5	49.3	49.2	1.5	87.9	12.1	49.6	50.3
Massachusetts	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Michigan	27.2	22.6	53.0	46.2	0.8	60.3	39.7	45.4	54.7
Minnesota	30.1	17.7	52.5	46.7	0.8	68.3	31.7	40.3	59.7
Missouri	18.6	25.5	49.3	49.2	1.5	87.9	12.1	49.6	50.3
Montana	23.8	24.1	51.8	47.0	1.2	71.2	28.8	43.0	57.0
National	23.6	26.3	51.1	47.2	1.7	64.8	35.2	48.0	52.0
Nebraska	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Nevada	27.2	22.6	53.0	46.2	0.8	60.3	39.7	45.4	54.7
New Hampshire	30.1	17.7	52.5	46.7	0.8	68.3	31.7	40.3	59.7
New Jersey	21.1	24.7	50.8	48.6	0.0	77.7	22.3	0.0	0.0
New Mexico	25.1	20.0	51.4	47.6	1.0	89.1	10.9	46.0	54.0
New York	20.8	24.0	52.1	47.3	0.6	65.2	34.8	45.4	54.6
North Carolina	22.1	24.3	51.8	46.4	1.8	80.2	19.8	41.6	58.4
North Dakota	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Ohio	20.5	18.5	53.0	46.7	0.3	54.9	45.1	45.1	59.8
Oklahoma	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Oregon	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Pennsylvania	25.8	21.8	53.7	44.6	1.7	60.9	39.1	54.7	45.4
South Carolina	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
South Dakota	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Tennessee	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Texas	20.8	24.0	52.1	47.3	0.6	65.2	34.8	45.4	54.6
Utah	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Virginia	20.8	24.0	52.1	47.3	0.6	65.2	34.8	45.4	54.6
Washington	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
West Virginia	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Wisconsin	28.7	19.5	51.9	47.1	1.0	49.1	50.9	50.1	49.8
Wyoming	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

Figure 2

## 2.2 Measurement

For this analysis, we selected Emerson as our primary pollster. The rationale behind this choice is explained in detail in Section 3. Emerson’s polls provided essential details, including polling date ranges, party affiliations, sample sizes, and the percentage breakdown of support for different political parties.

We focused on analyzing voter preferences based on party affiliation rather than individual candidate support, as this was more relevant to predicting overall trends. By capturing the broader party dynamics, our model aims to forecast how states are likely to swing in future elections, especially with respect to swing states. This focus will be elaborated in our results section, Section 5.

Additionally, the Emerson dataset included critical demographic data, such as respondents’ political preferences, motivations for voting in the 2024 election, opinions on Donald Trump’s legal battles, and perspectives on major political figures and issues. These details, coupled with information on party affiliation, sources of political news, and other personal attributes, allowed us to paint a richer picture of the voting landscape.

To ensure uniformity across all states, we concentrated on questions that were consistently asked across the entire dataset. This enabled us to perform state-level aggregation and analysis without being limited by state-specific polling questions, ensuring a standardized approach across the board.

## 2.3 Outcome variables

The initial analysis of the dataset revealed some key trends in party preferences across different states. In Figure 3, we illustrate which states exhibit the strongest leanings toward either the Democratic or Republican party. This visualization plays a crucial role in identifying the pivotal swing states that lie close to the center of the graph. These swing states, which show near-equal support for both parties, are of particular interest for future election predictions, as they are likely to determine the overall election outcome.

By visualizing the difference between Democratic and Republican preferences across states, we are able to pinpoint which states are firmly partisan and which fall closer to the center, indicating their potential to swing either way in future elections.

## 2.4 Predictor variables

Next, we take a deeper dive into the various demographic and social factors that influence voter preferences. The predictors we consider include education levels, gender, ethnicity, and age group. These variables are crucial as they shape political preferences and are likely to influence which way a state leans politically.

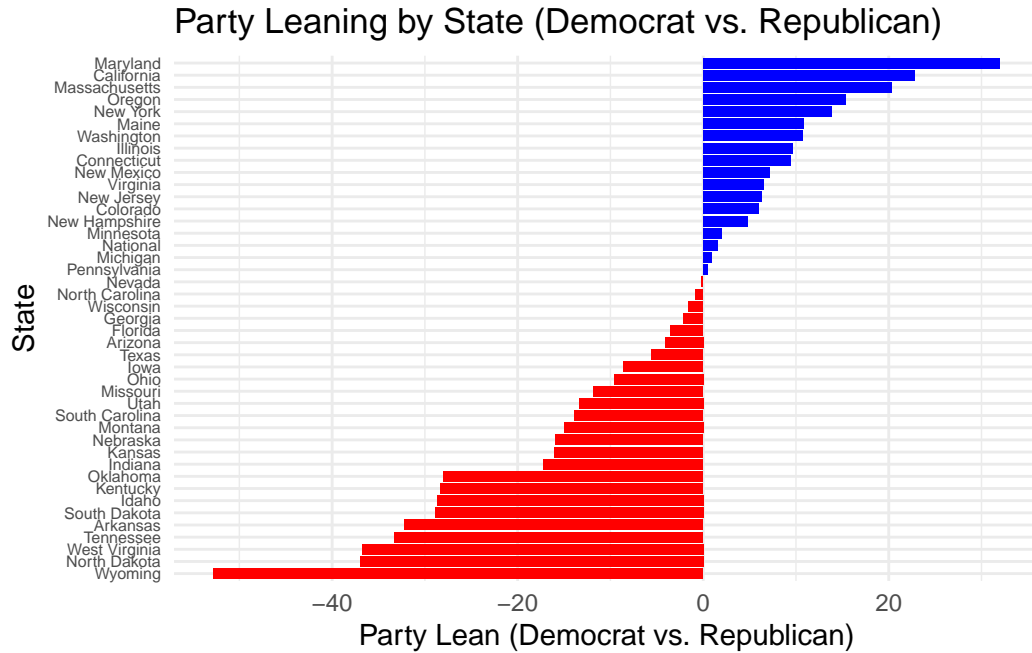


Figure 3: Example Predictor Variable Visualization

In Figure 4, we explore how these factors correlate with party lean across the U.S. states. For instance, higher levels of education may correlate with a greater likelihood of supporting the Democratic party, while other demographic factors such as ethnicity and age may play different roles in shaping political affiliation. This analysis helps us to identify which factors have the greatest influence in each state, offering a clearer understanding of what drives the electorate.

The correlation matrix in Figure 4 provides a detailed look at how various predictors are related to party preferences. From this, we can see which predictors—such as education or gender—are more strongly associated with Democratic or Republican leanings. The blue shading indicates stronger correlations with Democratic support, while red represents stronger correlations with Republican support.

This analysis offers valuable insights into how specific demographic factors shape the political landscape in different states, allowing us to better predict electoral outcomes based on shifting demographic trends.

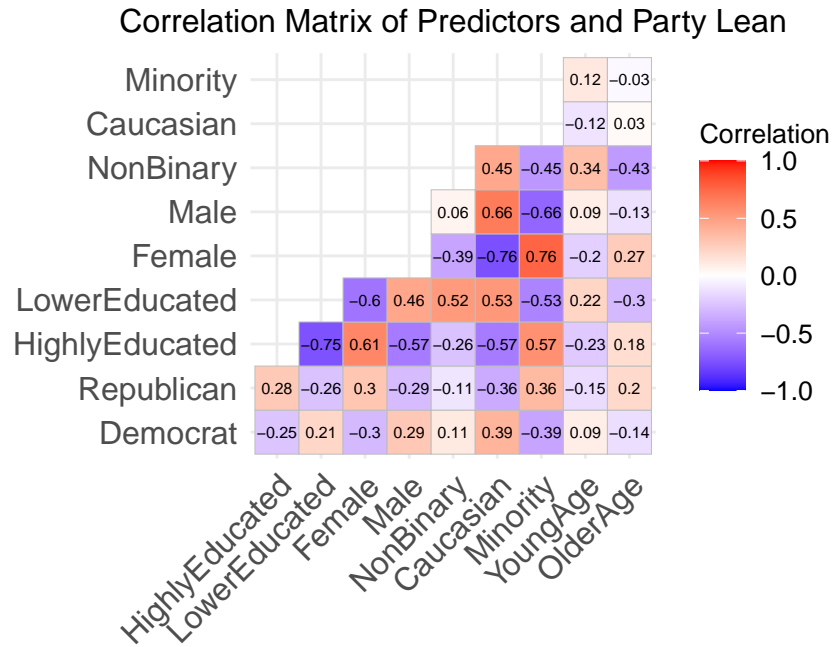


Figure 4: Example Predictor Variable Visualization

### 3 Pollster Methodology Overview (Appendix A)

Append a detailed review of a chosen pollster’s methodology, survey techniques, strengths, and weaknesses.

## 4 Model

### 4.1 Model Development

The goal of our model is to forecast the popular vote outcome of the 2024 US presidential election.

*Person B:* Choose a linear or generalized linear model. Justify the selection based on your research goal and data structure.

### 4.2 Model Set-up

Define the model mathematically and contextually (e.g., linear model predicting vote share).

$$y_i | \mu_i, \sigma \sim \text{Normal}(\mu_i, \sigma) \quad (1)$$

$$\mu_i = \alpha + \beta_i x_i \quad (2)$$

$$\alpha \sim \text{Normal}(0, 2.5) \quad (3)$$

$$\beta \sim \text{Normal}(0, 2.5) \quad (4)$$

$$\sigma \sim \text{Exponential}(1) \quad (5)$$

We run the model in R (R Core Team 2023) using the `rstanarm` package of Goodrich et al. (2022). We use the default priors from `rstanarm`.

*Person B:* Define all model components and assumptions.

### 4.3 Model Justification

Our results are summarized in Table ???. Discuss the implications of the findings and their predictive accuracy.

## 5 Results

Our results are summarized in Table ??.

*Person B:* Visualize the model's results and include any performance metrics (e.g., RMSE, test/train split).

## 6 Discussion

### 6.1 First discussion point

If my paper were 10 pages, then should be be at least 2.5 pages. The discussion is a chance to show off what you know and what you learnt from all this.

### 6.2 Second discussion point

Please don't use these as sub-heading labels - change them to be what your point actually is.



### **6.3 Third discussion point**

Discuss what the model reveals about the election forecast and its potential impact on understanding voting behavior.

*Person C:* Discuss limitations of the model and areas for further improvement.

### **6.4 Weaknesses and next steps**

Weaknesses and next steps should also be included.

## Appendix

### .1 Appendix A: Pollster Methodology Overview

*Person A:* Provide a detailed methodology review for the chosen pollster, including survey design, population sampling, non-response handling, etc.

### .2 Appendix B: Idealized Survey Design for \$100K Budget

*Person C:* Develop and describe an ideal survey design for forecasting the election with a \$100K budget, and include survey link.

### .3 Additional Data & Model Details

Include any technical details on data cleaning, model diagnostics, and posterior checks.

## A Additional data details

## B Model details

### B.1 Posterior predictive check

In `?@fig-ppcheckandposteriorvsprior-1` we implement a posterior predictive check. This shows...

In `?@fig-ppcheckandposteriorvsprior-2` we compare the posterior with the prior. This shows...

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

### B.2 Diagnostics

`?@fig-stanareyouokay-1` is a trace plot. It shows... This suggests...

`?@fig-stanareyouokay-2` is a Rhat plot. It shows... This suggests...

Checking the convergence of the MCMC algorithm

## References

- FiveThirtyEight. 2024. “Dataset: US Presidential General Election Polls.” [https://projects.fivethirtyeight.com/polls/data/president\\_polls.csv](https://projects.fivethirtyeight.com/polls/data/president_polls.csv).
- Goodrich, Ben, Jonah Gabry, Imad Ali, and Sam Brilleman. 2022. “rstanarm: Bayesian applied regression modeling via Stan.” <https://mc-stan.org/rstanarm/>.
- R Core Team. 2023. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.