

Trump’s 2024 Winning Coalition: Insights from the 2022 Comprehensive Election Study*

Logistic Regression Analysis Reveals Support for Trump Among Young Men and Infrequent Voters

Talia Fabregas

December 3, 2024

First sentence. Second sentence. Third sentence. Fourth sentence. Key Trump 2024 voting blocs, including young men without a college education were less likely to vote in the 2022 midterms when he was not on the ballot compared to 2020

Table of contents

1	Introduction	3
2	Data	4
2.1	Overview	4
2.2	Measurement	5
2.3	Outcome variables	5
2.4	Predictor variables	7
2.4.1	Race and gender	7
2.4.2	Age	7
2.4.3	Education	9
2.4.4	Civic engagement and trust in government	9
3	Model	10
3.1	Model set-up	12
3.1.1	Vote choice model	12
3.1.2	Turnout model	13
3.2	Model justification	14
3.3	Model Weaknesses and Limitations	14

*Code and data are available at: <https://github.com/taliafabs/US-Midterms-2022.git>.

4	Results	14
4.1	Voters that favor Trump	15
4.1.1	Young men	15
4.2	2022 midterm election turnout	15
4.2.1	Voters with low interest in politics and low civic engagement are less likely to vote in midterm elections	16
4.2.2	Low propensity voters that favored Trump in 2020 were less likely to vote in 2022	16
5	Discussion	17
5.1	First discussion point	17
5.2	Overlapping predictors for Trump support and not voting in an election where he is not on the ballot	17
5.3	Polling misses and Trump’s unique appeal to infrequent, low-propensity voters	17
5.4	2024 Polling Misses and Non-Response Bias	17
5.5	Weaknesses and next steps	17
	Appendix	18
A	Surveys, Sampling, and Observational Data	18
A.1	Deep dive into 2022 Cooperative Election Study (CES) survey methodology . .	18
A.1.1	Overview and survey objectives	18
A.1.2	Population and sampling frame	18
A.1.3	Sampling approach	18
A.1.4	Data validation	19
A.1.5	Weighting and data adjustments	19
A.1.6	Strengths and limitations	20
A.2	Second Point: Idealized Survey Methodology to Study 2024 Down-Ballot Voter Attrition	20
A.3	Social Desirability Bias and Identifying Non-Voters	20
B	Additional data details	20
B.1	Data cleaning	20
B.1.1	Identifying non-voters	21
C	Model details	21
C.1	Model results	21
C.1.1	Vote preference model	21
C.1.2	Turnout model	21
C.2	Posterior predictive check	24
C.2.1	Turnout model	24
C.3	Credibility Intervals	24
C.3.1	Turnout model	24

C.4 Diagnostics	25
C.4.1 Turnout model	25
References	26

1 Introduction

Overview paragraph

This study aims to estimate how race, gender, education, trust in the federal government, civic engagement, and interest in politics impact voting behavior. The first thing that this study will examine is the effects of age, gender, race, education, trust in the federal government, civic knowledge, and civic engagement on the probability that an individual supports Trump. The second thing that this study will examine is the effects of past presidential vote choice, age, highest level of education completed, trust in the federal government, knowledge of which party holds a majority in Congress, and interest in politics on the probability that an individual who voted for Donald Trump or Joe Biden in the 2020 U.S. presidential election would also vote in the 2022 U.S. midterm election. The estimands, which can never be known with complete certainty, include:

- The true effects of age, gender, race, education, trust in the federal government, knowledge of which party has a majority in the U.S. House, and interest in politics on the likelihood that an individual supports Trump.
- The true effects of presidential vote preference, age, education, trust, civic knowledge, and interest in politics on the likelihood that an individual who voted in a presidential election will also vote in the subsequent midterm election (Alexander 2023) .

Based on the results of applying the logistic regression model to predict vote preference, I estimate that young men, voters without a college education, and voters with low trust in government are likely to support Trump. The logistic regression vote preference model estimates ...

Based on the results of applying the logistic regression model to predict voter turnout, I estimate that voters with

Why it matters paragraph

In the 2024 election, Democratic Senate candidates (both incumbent and non-incumbent) were elected in states that President-Elect Trump carried: Michigan, Wisconsin, Nevada, and Arizona. In Nevada and Wisconsin, Senators Jackie Rosen and Tammy Baldwin

The remainder of this paper is structured as follows. Section 2 contains an overview of the survey data set from Schaffner, Ansolabehere, and Shih (2023) that was used, visualizations of different variables, and summary statistics. Section 3 contains the logistic regression models

used to predict the probability that an individual who voted for either Trump or Biden in the 2020 presidential election would also vote in the 2022 midterm election and the probability that an individual would support Trump over Biden in 2020. Section 4 contains tables and data visualizations that present results about what characteristics made individuals who voted in the 2020 presidential election more likely to also vote in the 2022 midterm election and what characteristics make individuals more likely to support Trump. Section 5 contains detailed a detailed discussion about the results presented in Section 4, including demographics where support for Trump is strong, why some 2020 presidential election voters were more likely than others to vote in the 2022 midterm election, and how the 2016, 2020, and 2024 polling misses can be explained by low civic engagement and distrust in government among Trump voters. Appendix A contains an in-depth discussion about and evaluation of CES 2022 survey methodologies.

2 Data

2.1 Overview

The 2022 Comprehensive Election Study Common Content (CES) data set from Schaffner, Ansolabehere, and Shih (2023) was used for this project. The data was obtained from Harvard Dataverse on November 27, 2024. It is a nationally-representative survey that aims to study voting behavior in the United States. It contains 60,000 observations and over 700 variables.

I cleaned the 2022 CES data set to only include respondents who voted for either Donald Trump or Joe Biden in the 2020 U.S. presidential election. I used the *pres20vote* variable, which is each participant’s response to a post-election survey wave question about who they voted for in 2020, to determine who each respondent voted for. This is not a perfect way to find out who each respondent voted for because it is subject to social desirability bias (Silver 2024). I am confident that most Trump and Biden 2020 voters are correctly labeled because most CES 2022 respondents’ 2020 presidential votes are consistent with their party affiliation or ideological leanings (Schaffner, Ansolabehere, and Shih 2023). However, there is still a chance respondents who actually voted for Trump may have said that they voted for Biden in 2020 because of Trump’s controversial nature or to say that they voted for the winning candidate (Silver 2024). I discuss this in depth in Section 5.5. After sub-setting the data to only include identified Trump and Biden 2020 voters, I used TargetSmart voter file match status to determine which respondents voted in the 2022 midterm election and which respondents did not. Additional data details can be found in [?@sec-additional-data](#).

I used the statistical programming language R (R Core Team 2023) and the `dplyr`, `janitor`, `ggplot2`, and `kable` packages to clean the data, prepare the data, and create tables and data visualizations.

2.2 Measurement

The measurement task is to capture how Americans view their representatives, how they hold the different levels of government to account during elections, how they voted, their electoral experiences, and how voting behavior and experiences vary across different regions, demographics, and social contexts (Schaffner, Ansolabehere, and Shih 2023). Surveys are a widely-used instrument for measuring public opinion during election cycles (Alexander 2023). During U.S. presidential and midterm election cycles, pollsters and researchers conduct surveys that measure candidate preferences, public opinion, and how Americans' diverse geography, demographics, and experiences impact them. Researchers often use these measurements to predict election outcomes, analyze election outcomes, and analyze which factors predict voting behavior and vote preference.

The Cooperative Election Study (CES) survey data set that I used is a collection of 60,000 responses from a nationally-representative sample of American adults. Each entry represents the political preferences, voting intentions, ideological leanings, demographics, issue evaluations, and past voting decisions of one respondent. The CES survey has been conducted every year since 2006 (Schaffner, Ansolabehere, and Shih 2023). In presidential and midterm election years, it consists of a pre-election wave and a post-election wave. The pre-election wave aims to measure the opinions, vote preferences, vote intentions of the American public, and demographics (Schaffner, Ansolabehere, and Shih 2023). The post-election wave aims to measure how different factors, including geography, demographics, issue-evaluations, and the state of the economy influenced Americans' decisions about who to vote for or whether to vote at all in the recent election (Schaffner, Ansolabehere, and Shih 2023).

The transformation of an individual American adult's opinion to an entry in the CES 2022 data set follows three steps, as outlined by Schaffner, Ansolabehere, and Shih (2023):

1. **Survey data collection:** selected voters from a nationally representative sample respond to a Common Content survey.
2. **Weighting:** survey responses are weighted to adjust for any imbalances that exist in the sample. Respondents who are less likely to answer a survey are given higher weights.
3. **Reporting:** the weighted survey results are recorded as entries in the data set. The data set then serves as a snapshot of American public opinion in the weeks leading up to and shortly after the midterm election.

2.3 Outcome variables

Firstly, I will use age, gender, race, highest level of education completed, trust in the government, and knowledge of which party controls the U.S. House of Representatives to predict whether an individual supports Donald Trump.

Then, I will use 2020 presidential vote, age, highest level of education completed, trust in the government, knowledge of which party holds a majority in the U.S. House of Representatives,

and interest in politics to predict whether someone who did vote for either Donald Trump or Joe Biden in the 2020 presidential election would also vote in the 2022 midterm election, or more generally, in an election where neither Trump nor Biden was on the ballot.

It is possible that variables that predict support for Trump can also be used to predict turnout in the 2022 midterm election. Support for Trump is high among voters who are male, do not have a college education, do not trust the government, and have low civic engagement. These factors make them less likely to respond to a survey, and it is possible that these factors also make them less likely to vote in a midterm election where Trump is not on the ballot.

Table 1: Biden voters are overrepresented in the CES 2022 survey data set.

2020 Vote	Num respondents	%
Donald Trump	17442	41.56
Joe Biden	24526	58.44

The 2020 presidential election had record-setting turnout (Grant 2023). As seen in Table 2, the majority of 2022 CES respondents who voted for either Trump or Biden in the 2020 presidential election were also identified as 2022 midterm election voters. However, it appears that around 30% of 2022 CES respondents who did vote for one of the major candidates in the 2020 presidential election abstained from the 2022 midterm election when neither Trump nor Biden was on the ballot. Hartig et al. (2023) found that midterm elections typically have lower turnout than the preceding presidential election and the responses to the 2022 CES survey are consistent with that. The remainder of this project aims to study what affects vote preference and which presidential election voters are most likely to vote in the subsequent midterm election and which presidential election voters are most likely to abstain from the subsequent midterm election.

Table 2: 70.64% of respondents who voted in the 2020 presidential election also voted in the 2022 midterm election

Voting Status	Num respondents	%
Did Not Vote in 2022	12323	29.36
Voted in 2022	29645	70.64

Table 3 shows that 71.4% of respondents who voted for Biden in 2020 voted in the 2022 midterm election, compared to 69.9% of respondents who voted for Trump. This 1.5-point midterm turnout gap between Biden 2020 and Trump 2020 voters in the 2022 CES survey could be related to the fact that Trump voters are more likely to distrust the government and have low civic engagement, and are therefore less likely to respond to a survey or vote in an election where he is not on the ballot (Silver 2024). As shown in Table 1, Trump voters are

underrepresented in the 2022 CES survey, so non-response bias will be a factor throughout this project.

Table 3: 2022 midterm turnout was comparable among survey respondents who voted for Trump versus survey respondents who voted for Biden in 2020, but the turnout rate of Biden voters was 1.5 points higher than that of Trump voters.

presvote2020	voted_status	n	percentage
Donald Trump	Did Not Vote in 2022	5310	30.44
Donald Trump	Voted in 2022	12132	69.56
Joe Biden	Did Not Vote in 2022	7013	28.59
Joe Biden	Voted in 2022	17513	71.41

2.4 Predictor variables

The tables and visualizations below present possible relationships between predictor variables and either support for Trump or voting in the 2022 midterm election.

2.4.1 Race and gender

In the 2024 U.S. presidential election, there were gender and racial gaps in support for Trump, with more women supporting Vice President Harris and more men supporting Trump. Since 2016, a higher percentage of white voters than voters of color has supported Trump, but in the 2024 presidential election, Trump increased his vote share among voters of color, especially Hispanic voters and black men. As shown in Table 3, white male survey respondents were more likely to support Trump, while black female respondents favored Biden in 2020. The gender gap in support for Trump is also shown, with white men more likely to support Trump than white women in the 2022 CES survey (Schaffner, Ansolabehere, and Shih 2023). Support for Trump varies by race and gender, but these are not the only variables that predict vote preference.

2.4.2 Age

As shown in Figure 2, there was lower turnout in 2022 midterm election than the 2020 presidential election across all age brackets. Voters in the 18-29 age bracket, especially those who voted for Trump in 2020, are under-represented in the 2022 CES survey.

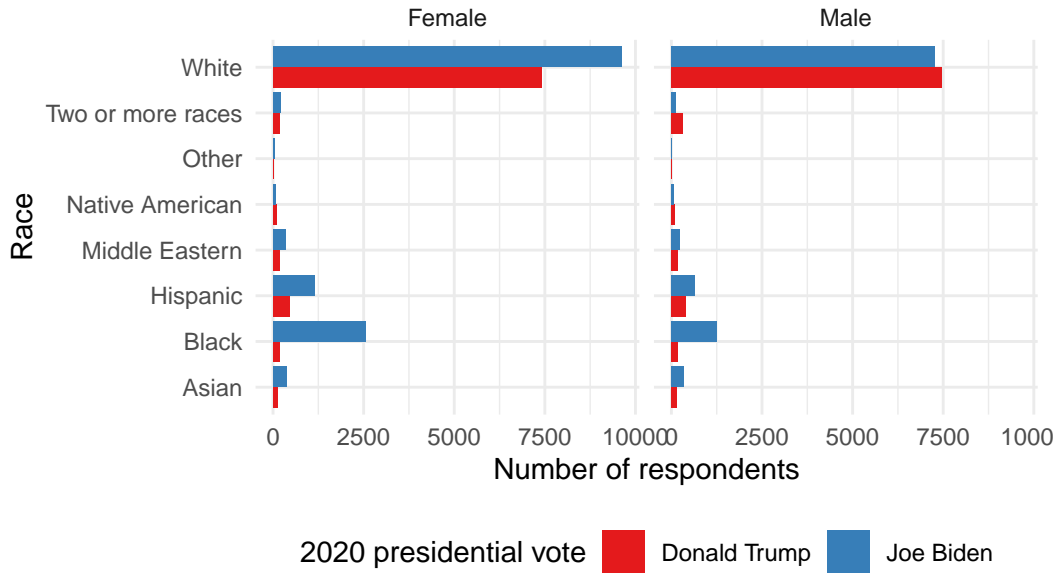


Figure 1: Support for Truo varied among CES respondents, with white men most likely to support Trump

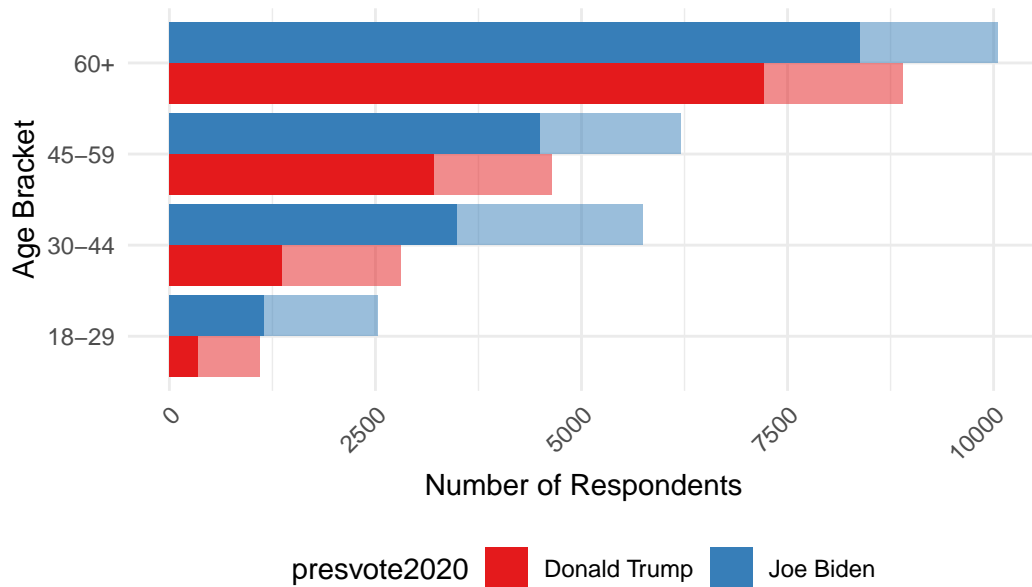


Figure 2: More than half of 18-29 year old respondents who voted for Trump in 2020 presidential election did not vote in the 2022 midterm election. The solid portion of the red and blue bars shows the number of respondents who voted in the 2020 election that also voted in the 2022 midterm.

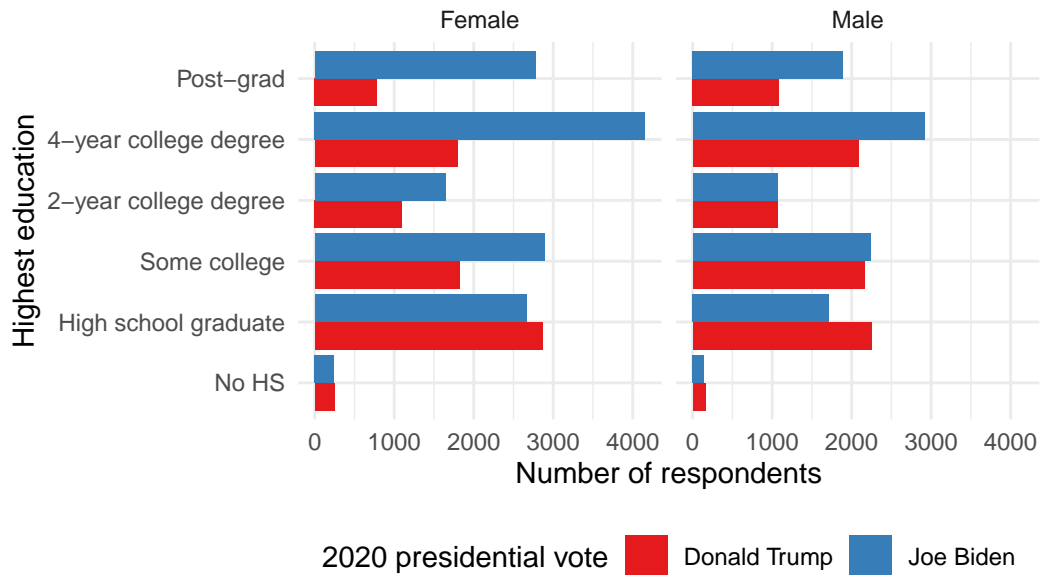


Figure 3: College-educated respondents were less likely to support Trump in 2020.

2.4.3 Education

Education gap in American politics.

Race, gender, highest level of education, income, and religion are widely used when predicting political preference. In 2016, 2020, and 2024, pre-election polls and predictive models underestimated support for Trump.

2.4.4 Civic engagement and trust in government

The 2022 CES survey does not contain an explicit question about civic engagement and there is no direct way to determine who is a low-propensity or low-information voter. I have used the following variables to measure civic engagement and trust in government:

- **trustfed**: response to the question about how much they trust the federal government)
- **newsint**: response to survey question about interest in politics
- **CC22_310b**: response to survey question asking which party has a majority in the U.S. House of Representatives.

In my data preparation process, I labeled respondents with little to no trust in the government, little to no interest in politics, and who do not know which party controlled the U.S. House as low-propensity voters.

As shown in Figure 4, survey respondents with less trust in the government favored Trump in 2020.

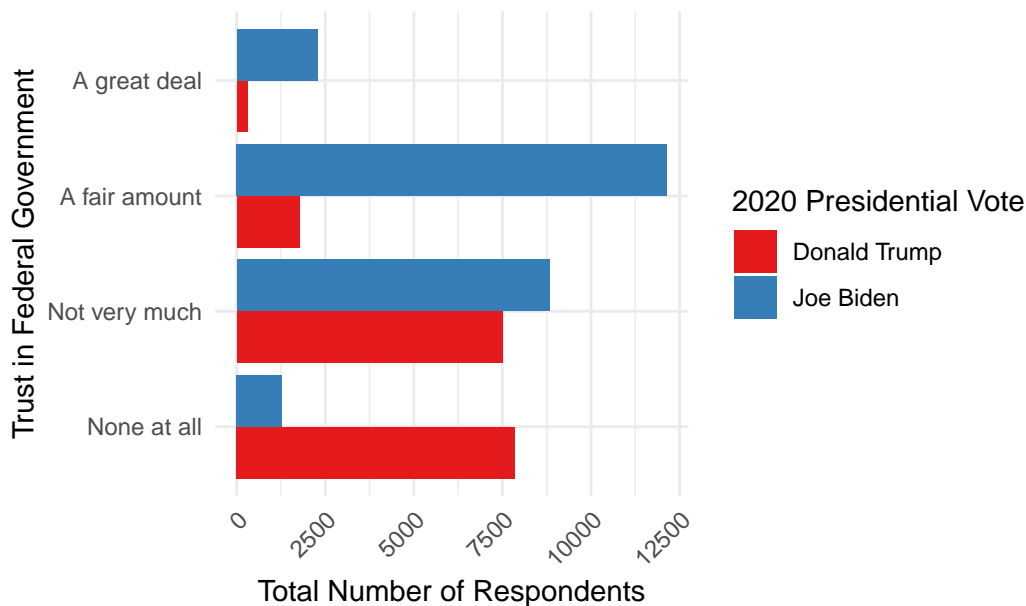


Figure 4: Voters with less trust in the government favored Trump in 2020, while voters with more trust favored Biden in 2020. The bars show the number of respondents with each level of trust in government who voted for Trump and Biden in 2020.

As shown in Figure 5 and Figure 6, respondents who voted in 2020, have at least some interest in politics, and knew which party held a majority in the U.S. House of Representatives turned out to vote at a higher rate in the 2022 midterm election compared to respondents who do not.

3 Model

The goal of my modeling strategy is twofold.

I use two logistic regression models: one to model the probability that an individual who voted in the 2020 presidential election would also vote in the 2022 midterm election, and one to model the probability that an individual voter supports Trump.

The model to predict *voted_for_trump* uses the following predictors:

- `age_bracket`
- `gender`

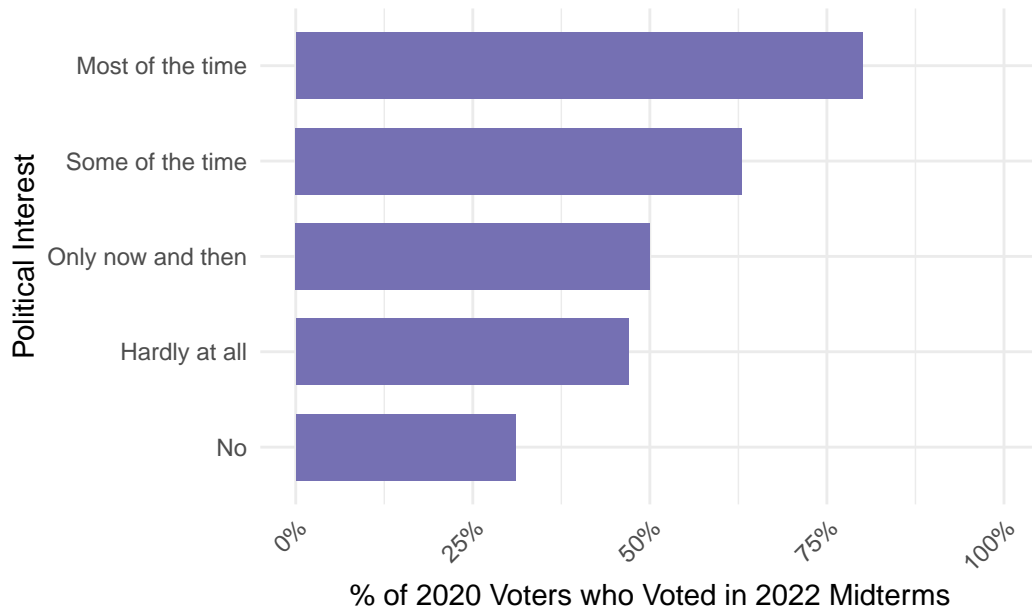


Figure 5: Respondents with low interest in politics who voted in the 2020 presidential election were less likely to vote in the 2022 midterm election.

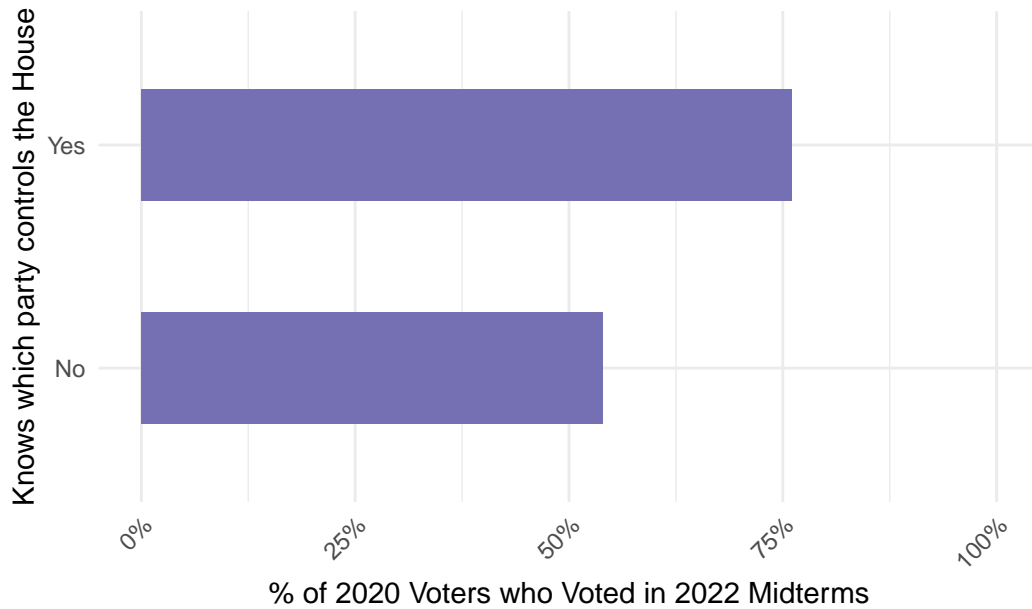


Figure 6: There is a nearly 15-point gap in 2022 turnout amongst respondents who voted for Trump or Biden in 2020 and know which party has a majority in the U.S. House of Representatives versus those who do not.

- race
- educ
- trust
- know_us_house

The model to predict *voted_in_2022* uses the following predictors:

- presvote2020
- age_bracket
- educ
- trust
- know_us_house
- political_interest

Here, I provide a brief description of how I used logistic regression to predict voter turnout and vote preference, and how my models are set up. Model details and diagnostics can be found in [Appendix C](#).

3.1 Model set-up

3.1.1 Vote choice model

Define π_i as the probability that respondent i supports Trump, $age_bracket_i$ as their age bracket, $gender_i$ as their gender, $race_i$ as their race, $educ_i$ as the highest level of education that they completed, $trust_i$ as their self-reported level of trust in the government, and $know_us_house_i$ as whether or not they know which party had a majority in the U.S. House of Representatives as of 2022:

$$\begin{aligned}
y_i | \pi_i &\sim \text{Bern}(\pi_i) \\
\text{logit}(\pi_i) &= \beta_0 + \beta_1 \cdot \text{age_bracket}_i + \beta_2 \cdot \text{gender}_i + \beta_3 \cdot \text{race}_i + \beta_4 \cdot \text{educ}_i \\
&\quad + \beta_5 \cdot \text{trust}_i + \beta_6 \cdot \text{know_us_house}_i \\
\beta_0 &\sim \text{Normal}(0, 2.5) \\
\beta_1 &\sim \text{Normal}(0, 2.5) \\
\beta_2 &\sim \text{Normal}(0, 2.5) \\
\beta_3 &\sim \text{Normal}(0, 2.5) \\
\beta_4 &\sim \text{Normal}(0, 2.5) \\
\beta_5 &\sim \text{Normal}(0, 2.5) \\
\beta_6 &\sim \text{Normal}(0, 2.5)
\end{aligned}$$

3.1.2 Turnout model

Define π_i as the probability that survey respondent i who voted in 2020 also voted in the 2022, presvote2020_i as who they voted for in the 2020 election (Trump or Biden), educ_i as the highest level of education they completed, trust_i as their level of trust in the government,

$$\begin{aligned}
y_i | \pi_i &\sim \text{Bern}(\pi_i) \\
\text{logit}(\pi_i) &= \beta_0 + \beta_1 \cdot \text{presvote2020}_i + \beta_2 \cdot \text{educ}_i + \beta_3 \cdot \text{trust}_i + \beta_4 \cdot \text{know_us_house}_i \\
&\quad + \beta_5 \cdot \text{political_interest}_i \\
\beta_0 &\sim \text{Normal}(0, 2.5) \\
\beta_1 &\sim \text{Normal}(0, 2.5) \\
\beta_2 &\sim \text{Normal}(0, 2.5) \\
\beta_3 &\sim \text{Normal}(0, 2.5) \\
\beta_4 &\sim \text{Normal}(0, 2.5) \\
\beta_5 &\sim \text{Normal}(0, 2.5) \\
\beta_6 &\sim \text{Normal}(0, 2.5)
\end{aligned}$$

The models are run in R (R Core Team 2023) using the `rstanarm` package and default priors from `rstanarm` (Goodrich et al. 2022).

3.2 Model justification

I used two different logistic regression models to estimate support for Trump and voter turnout.

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 . Confounders have not been eliminated -

I found that trust in the federal government is related to both support for Trump and likelihood that someone who voted in 2020 also voted in 2022. education same thing.

3.3 Model Weaknesses and Limitations

The models were trained using a random subset of the 2022 CES survey data set. My analysis data set (after data cleaning) contained 41,968 observations. I took a random subset of 5000 observations to train my models. This was a trade-off because

Schaffner, Ansolabehere, and Shih (2023) warn that using a small sub sample of the CES survey data to train a model because measurement error can lead to flawed inferences. The CES is a large survey and as a whole, it provides a sufficient number of observations to analyze small sub-populations (Schaffner, Ansolabehere, and Shih 2023).

Trained using a random subset of the data. Refer to codebook about the risks of that. I randomly sampled 5,000 observations from the 2022 CES survey data set. This was a tradeoff; training a model on more observations would have had a significantly longer runtime. However, randomly subsetting the 2022 CES survey data set is risky because ...

4 Results

The model results are summarized in Appendix [C.1](#).

4.1 Voters that favor Trump

4.1.1 Young men

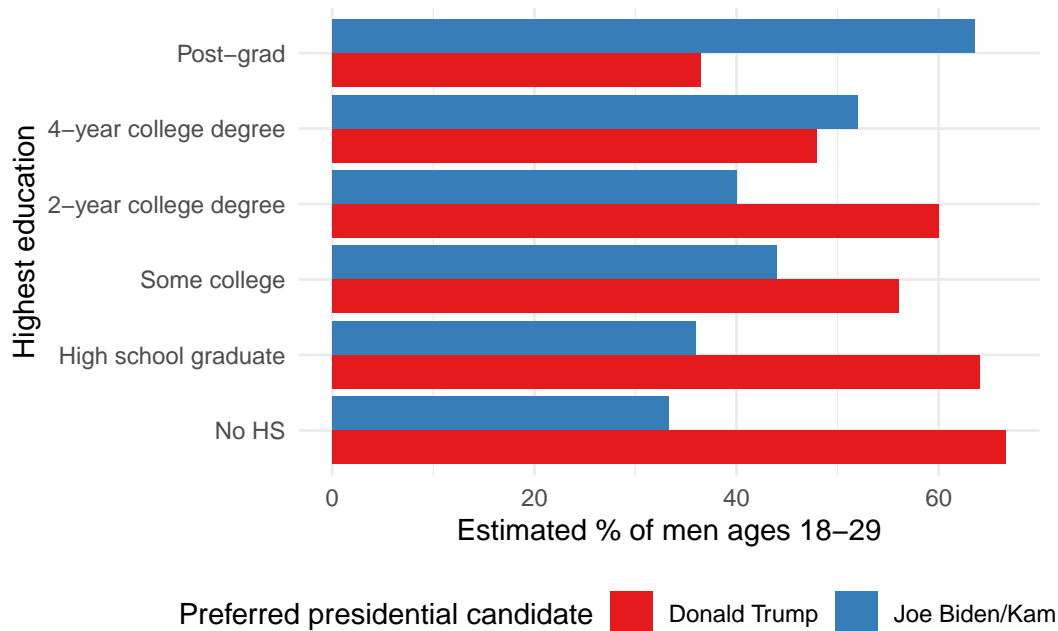


Figure 7: The majority of men ages 18-29 without a 4-year college degree support Trump, based on model estimates. The blue bars show the estimated % of young men with at each level of education that are Biden/Harris/Democratic supporters and the red bars show the % of young men at each level of education that support Trump.

4.2 2022 midterm election turnout

Table 4: Overall, Trump 2020 voters may have been a bit more likely than Biden 2020 voters to vote in the 2022 midterm election based on model estimates.

Biden 2020 voters	Trump 2020 voters
74.66	75.48

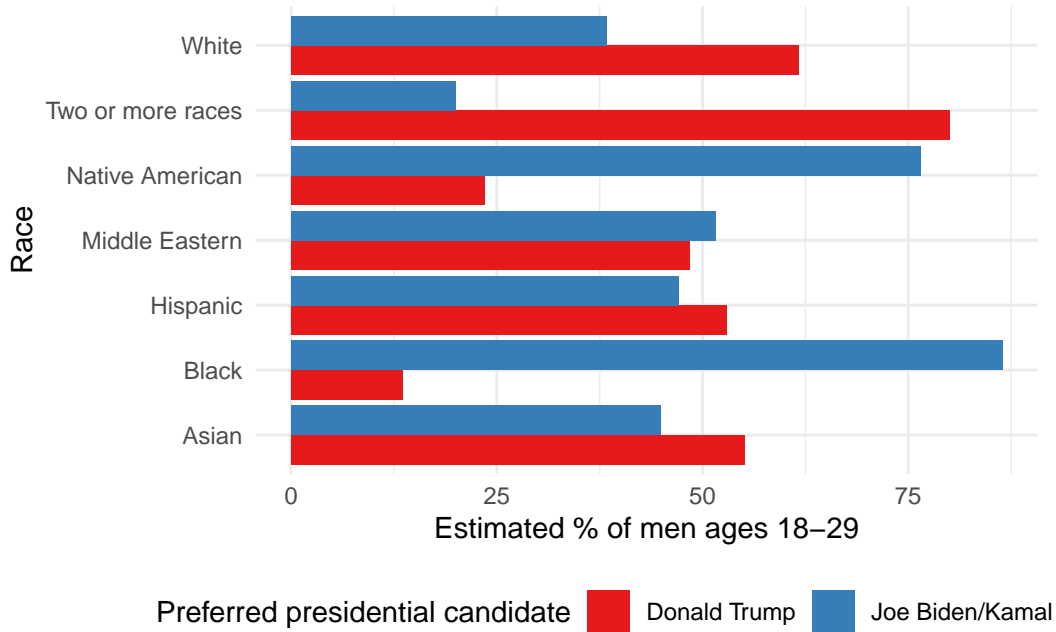


Figure 8: The majority of white, hispanic, asian, and multiracial men ages 18-29 support Trump based on model estimates. The red bars

4.2.1 Voters with low interest in politics and low civic engagement are less likely to vote in midterm elections

4.2.2 Low propensity voters that favored Trump in 2020 were less likely to vote in 2022

This result foreshadowed the fact that young men, especially those without a college degree who are infrequent voters, would become a key part of Trump's winning coalition in 2024.

Table 5: Based on model estimates, 2020 presidential election voters with no college education, low trust in government, and low civic engagement/knowledge were likely to support Trump but less likely than the average 2020 voter to also vote in the 2022 midterm election.

Estimated Trump %	Estimated 2022 turnout %
69.54	62.56

5 Discussion

5.1 First discussion point

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

5.2 Overlapping predictors for Trump support and not voting in an election where he is not on the ballot

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

5.3 Polling misses and Trump's unique appeal to infrequent, low-propensity voters

Young men who are not college-educated, do not know which party controls the U.S. House of Representatives,

5.4 2024 Polling Misses and Non-Response Bias

5.5 Weaknesses and next steps

Weaknesses and next steps should also be included.

One weakness is correctly identifying 2020 Trump voters and 2022 midterm non-voters in the 2022 CES survey data set.

Appendix

A Surveys, Sampling, and Observational Data

A.1 Deep dive into 2022 Cooperative Election Study (CES) survey methodology

A.1.1 Overview and survey objectives

The 2022 CES survey is a collaboration between 60 research teams that aims to study public opinion and voting behavior across the United States (Schaffner, Ansolabehere, and Shih 2023). It is part of the ongoing Cooperative Election Study (CES), previously known as the Cooperative Congressional Election Study (CCES). It consists of 60,000 responses from a nationally-representative sample of American adults. Each research team purchased a 1000 person nationally representative survey from YouGov (Schaffner, Ansolabehere, and Shih 2023). The 2022 CES survey was conducted online by YouGov and 60,000 American adults were surveyed in two waves: the pre-election wave (September 29 -November 8, 2022) and the post-election wave (November-December 2022) (Schaffner, Ansolabehere, and Shih 2023). Sample matching was used to build a nationally representative sample and weighting was used to further ensure that the sample is representative (Schaffner, Ansolabehere, and Shih 2023).

A.1.2 Population and sampling frame

The target population of a survey is the population that it aims to represent and speak about (Alexander 2023). The target population of the 2022 CES survey is all American adults (Schaffner and Kuriwaki 2022). Researchers using the 2022 CES survey dataset can subset it to only include respondents who matched to a validated voter registration record if they wish to study registered voters (Schaffner, Ansolabehere, and Shih 2023). The sampling frame of a survey is all potential respondents from its target population (Alexander 2023). The 2022 CES survey’s sampling frame is American adults who are YouGov panelists and receive notifications about surveys [Schaffner and Kuriwaki (2022)].

A.1.3 Sampling approach

The 2022 CES uses an online survey conducted by YouGov to collect responses (Schaffner, Ansolabehere, and Shih 2023).

A.1.4 Data validation

A.1.5 Weighting and data adjustments

Table 6: Respondents without a college education were given more weight because they are less likely to respond to a survey.

Highest Education Completed	Average weight
No HS	2.3837276
High school graduate	0.8909107
Some college	0.6702328
2-year college degree	0.6637045
4-year college degree	0.7888546
Post-grad	0.7993014

Table 7: Respondents who voted for Trump in 2020 were given more weight.

2020 presidential vote	Average weight
Donald Trump	0.9091925
Joe Biden	0.7284850

Table 8: Respondents had less trust in government were given more weight because they are less likely to respond to a survey.

Trust in government	Average weight
None at all	0.8871197
Not very much	0.8097035
A fair amount	0.7397012
A great deal	0.8132750

A.1.6 Strengths and limitations

A.2 Second Point: Idealized Survey Methodology to Study 2024 Down-Ballot Voter Attrition

A.3 Social Desirability Bias and Identifying Non-Voters

B Additional data details

B.1 Data cleaning

Data cleaning and preparation was done using R (R Core Team 2023) and the `dplyr` (Wickham et al. 2023) and `janitor` (Firke 2023) packages. The original dataset from Schaffner, Ansolabehere, and Shih (2023) contained 60,000 observations and over 700 variables. 50,981 of the 60,000 respondents took the post-election survey and I required responses to post-election questions, so I subsetting to only include respondents who participated in the post-election survey. I then subsetting the data to include only respondents who voted for either Trump or Biden in 2020 using the `filter` function from Wickham et al. (2023). My analysis did not require the use of over 700 variables, so I used the `select` function from Wickham et al. (2023) to select the ones I needed.

The variables from the raw data that I used in my analysis data include:

- *TS_g2022*: which method of voting the respondent used in the 2022 midterm election. NA if the respondent was not matched to a TargetSmart voter file (Schaffner, Ansolabehere, and Shih 2023).
- *presvote2020post*: who the respondent voted for in the 2020 presidential election
- *race*: the respondent's race
- *birthyr*: year of birth (this was used to construct the age, and eventually *age_bracket* variables).
- *gender4*: gender identity
- *educ*: highest level of education completed
- *CC22_423*: the respondent's self-reported level of trust in the Government in Washington, D.C.
- *CC22_310a*: whether or not the respondent knew which party had a majority in the U.S. House of Representatives
- *newsint*: how often the respondent is interested in following what's going on in U.S. government and politics.

B.1.1 Identifying non-voters

I constructed a new variable, *voted_in_2022*, which is equal to 1 if the respondent voted in the 2022 midterm election and 0 if they did not. I identified non-voters as respondents who did not match to a TargetSmart voter file (respondents where *TS_g2022* was NA). According to Schaffner, Ansolabehere, and Shih (2023), identifying un-matched voters as non-voters is a justified approach because the most common reason why a respondent will not have a TargetSmart voting record is because they are not registered to vote, and therefore did not vote and self-reported non-voting rates are very high among respondents who did not match to a TargetSmart record. There is a small chance that this approach could falsely label un-matched voters without a TargetSmart match as non-voters, but it is the best way to make sure that no non-voters are falsely labeled as voters.

Schaffner, Ansolabehere, and Shih (2023) outlined two alternative approaches for identifying non-voters: labeling only matched non-voters as non-voters or labeling both matched non-voters and respondents who did not match to a TargetSmart voter file, but self-reported that they did not vote as non-voters. I opted not to use either of these approaches because un-matched respondents are most likely non-voters and using self-reported participation to determine whether un-matched respondents voted relies on the assumption that social desirability bias was not a factor (Schaffner, Ansolabehere, and Shih 2023). Social desirability bias happens when a survey respondent becomes inclined to answer a question in a way that they think would make them look good to others (Alexander 2023). There is a chance that un-matched respondents might be inclined to say that they did vote in 2022 because participating in the democratic process is seen as socially desirable behavior.

C Model details

This section provides additional details about the models, including model summaries, posterior predictive checks, credibility intervals, and model diagnostics.

C.1 Model results

C.1.1 Vote preference model

The model summary is shown in Table 9.

C.1.2 Turnout model

The model summary is shown in Table 10.

Table 9: Explaining whether someone voted for Trump based on race, gender, age, education, trust in government, civic engagement

	Voter Turnout Model
(Intercept)	−2.92 (0.29)
age_bracket30-44	0.36 (0.12)
age_bracket45-59	0.69 (0.12)
age_bracket60+	0.89 (0.11)
genderMale	0.46 (0.07)
genderNon binary	−3.36 (1.04)
genderOther	0.25 (1.13)
raceBlack	−1.56 (0.21)
raceHispanic	−0.42 (0.21)
raceMiddle Eastern	−0.70 (0.29)
raceNative American	−0.44 (0.41)
raceOther	−1.54 (1.16)
raceTwo or more races	0.65 (0.37)
raceWhite	0.60 (0.17)
educ4-year college degree	−0.32 (0.12)
educHigh school graduate	0.45 (0.13)
educNo HS	0.11 (0.19)
educPost-grad	−0.63 (0.13)
educSome college	−0.15 (0.13)
trustfedA great deal	−0.13 (0.17)
trustfedNone at all	3.71 (0.10)
trustfedNot very much	1.85 (0.08)
know_us_houseYes	−0.31 (0.08)
political_interestMost of the time	0.29 (0.17)
political_interestNo	−76.69 (64.18)
political_interestOnly now and then	0.79 (0.19)
political_interestSome of the time	0.57 (0.18)
Num.Obs.	7500

Table 10: Explaining whether someone who voted in the 2020 presidential election also voted in the 2022 midterm election, based on who they voted for, age, education, trust in government, civic knowledge, and political interest

Voter Turnout Model		
(Intercept)		−0.79 (0.21)
presvote2020Joe Biden		0.24 (0.08)
age_bracket30-44		0.55 (0.10)
age_bracket45-59		1.16 (0.11)
age_bracket60+		1.79 (0.11)
educ4-year college degree		0.36 (0.11)
educHigh school graduate		−0.27 (0.12)
educNo HS		−1.01 (0.17)
educPost-grad		0.42 (0.13)
educSome college		0.10 (0.12)
trustfedA great deal		−0.86 (0.13)
trustfedNone at all		0.35 (0.11)
trustfedNot very much		0.39 (0.08)
truststateA great deal		−0.07 (0.11)
truststateNone at all		0.15 (0.10)
truststateNot very much		−0.07 (0.08)
know_us_houseYes		0.36 (0.07)
know_us_senateYes		0.02 (0.06)
political_interestMost of the time		0.28 (0.15)
political_interestNo		−1.07 (1.12)
political_interestOnly now and then		−0.32 (0.17)
political_interestSome of the time		0.00 (0.16)
Num.Obs.		7500
R2	23	0.134
Log.Lik.		−3270.706
WAIC		6606.0
RMSE		0.42

C.2 Posterior predictive check

C.2.1 Turnout model

In ?@fig-ppcheckandposteriorvsprior-1 we implement a posterior predictive check. This shows the comparison of the outcome variable *vote_2022*, with simulations from the posterior distribution (Alexander 2023).

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

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

C.3 Credibility Intervals

C.3.1 Turnout model

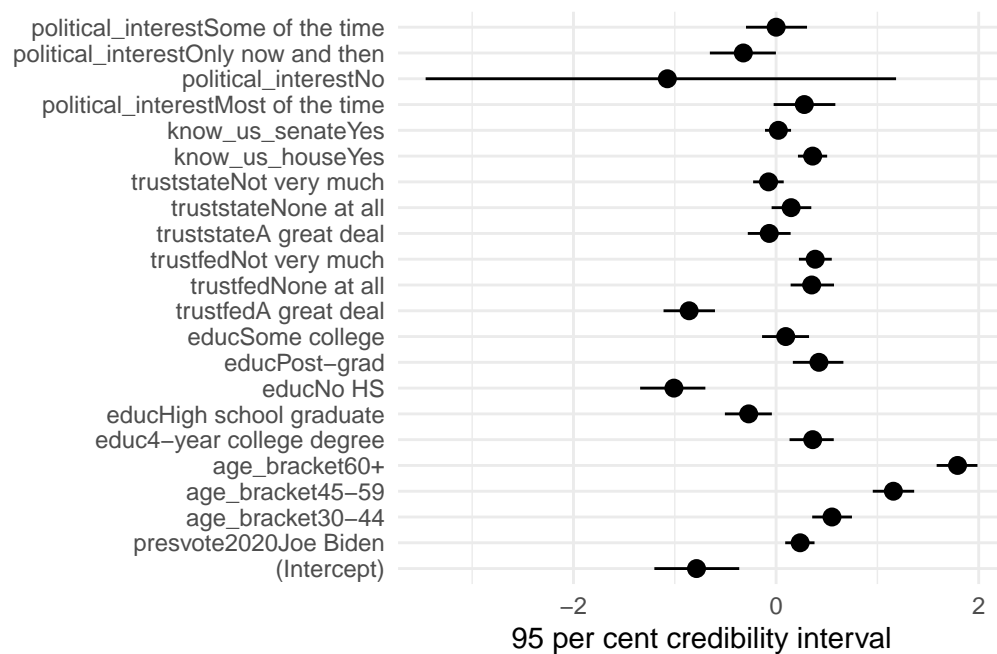


Figure 9: Credible intervals for predictors of voter turnout in the 2022 U.S. midterm elections

C.4 Diagnostics

C.4.1 Turnout model

Figure 10a is a trace plot. It shows... This suggests...

Figure 10b is a Rhat plot. It shows... This suggests...

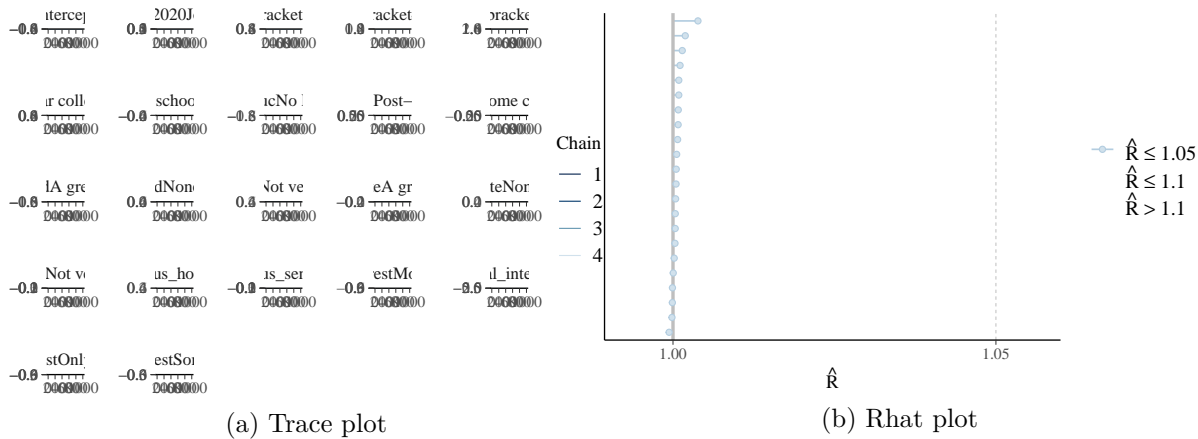


Figure 10: Checking the convergence of the MCMC algorithm

References

- Alexander, Rohan. 2023. *Telling Stories with Data*. Chapman; Hall/CRC. <https://tellingstorieswithdata.com/>.
- Firke, Sam. 2023. *Janitor: Simple Tools for Examining and Cleaning Dirty Data*. <https://CRAN.R-project.org/package=janitor>.
- Goodrich, Ben, Jonah Gabry, Imad Ali, and Sam Brilleman. 2022. “rstanarm: Bayesian applied regression modeling via Stan.” <https://mc-stan.org/rstanarm/>.
- Grant, Tracy. 2023. *United States Midterm Elections*. Britannica. <https://www.britannica.com/topic/midterm-election>.
- Hartig, Hannah, Andrew Daniller, Scott Keeter, and Ted Van Green. 2023. *Republican Gains in 2022 Midterms Driven Mostly by Turnout Advantage*. Pew Research Center. <https://www.pewresearch.org/politics/2023/07/12/voter-turnout-2018-2022/>.
- R Core Team. 2023. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.
- Schaffner, Brian, Stephen Ansolabehere, and Marissa Shih. 2023. “Cooperative Election Study Common Content, 2022.” Harvard Dataverse. <https://doi.org/10.7910/DVN/PR4L8P>.
- Schaffner, Brian, and Shiro Kuriwaki. 2022. *Frequently Asked Questions*. Harvard University. <https://cces.gov.harvard.edu/frequently-asked-questions>.
- Silver, Nate. 2024. “Nate Silver: Here’s What My Gut Says about the Election, but Don’t Trust Anyone’s Gut, Even Mine.” https://www.nytimes.com/2024/10/23/opinion/election-polls-results-trump-harris.html?unlocked_article_code=1.UU4.pFkQ.F2hD-woxmiEj&smid=url-share.
- Wickham, Hadley, Romain François, Lionel Henry, Kirill Müller, and Davis Vaughan. 2023. *Dplyr: A Grammar of Data Manipulation*. <https://CRAN.R-project.org/package=dplyr>.