US President Prediction*

Predict the results of 2024 US President Election

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October 22, 2024

In this study, we aim to study the 2024 U.S. Presidential Election with the methodology of polls of polls across various states. With the insightful analyses on polls provided by Redfiled & Wilton Strategies, the conducted survey on the population highlights various key point for the prediction of US president. Using Generalized Linear Models (GLMs), this report analyzes the accuracy of prediction in president votes through several key variables from pollster. The model incorporates variables such as pollscore, methodology, transparency_score and hypothetical. The prediction finding highlights the significance of candicate trustworthiness among the voter decision, providing insights for swing voters. This analysis offers valuable information into the potential electoral outcomes driving the 2024 US president election.

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....

^{*}Code and data are available at: https://github.com/younazhao/US-President-Prediction/tree/main.

2 Data

2.1 Overview

We use the statistical programming language R (R Core Team 2023).... Our data.... 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

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

The goal of our modelling strategy is twofold. Firstly,...

Here we briefly describe the Bayesian analysis model used to investigate... Background details and diagnostics are included in Appendix B.

3.1 Model set-up

Define y_i as the percentage of support for a candicate in a poll.

 $y_i = \beta_0 + \beta_1 pollscore + \beta_2 methodology + \beta_3 transparency score + \beta_4 hypothetical + \epsilon transparency score + \beta_4 hypothetical + \delta_4 hypothetical + \delta_5 hypothetica$

pollscore: A numeric value representing the score or reliability of the pollster in question (e.g., -1.1). "The error and bias we can attribute to a pollster. Negative numbers are better. Stands for"Predictive Optimization of Latent skill Level in Surveys, Considering Overall Record, Empirically.

methodology: The method used to conduct the poll (e.g., Online Panel).

transparency_score: A score reflecting the pollster's transparency about their methodology (e.g., 9.0). "A grade for how transparent a pollster is, calculated based on how much information it discloses about its polls and weighted by by recency. The highest Transparency Score is 10."

hypothetical: Indicates whether the poll is about a hypothetical match-up.

We run the model in R (R Core Team 2023).

3.1.1 Model justification

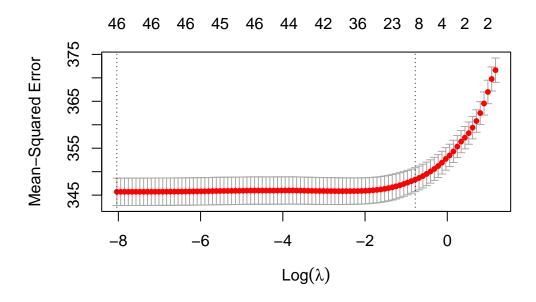
Table 1: model 1

	Std.			
	Estimate	Error t	value	$\Pr(> \mid \! t \mid)$
(Intercept)	63.979667 5.	0205698 12.	7435072	0.0000000
pollscore	- 0.	3209748	-	0.0000000
	3.533768	11.	0094857	
methodology Email	- 6.	5340436	-	0.0013902
	20.892998	3.1	975602	
methodologyEmail/Online Ad	- 9.	0526392	-	0.2493112
	10.429453	1.1	520898	
methodologyIVR	- 6.	0164883	-	0.0666730
	11.034652	1.8	340686	
methodologyIVR/Live Phone/Text	- 9.	6688032	-	0.1649743
	13.426432	1.3	886343	
methodologyIVR/Live Phone/Text/Online	- 5.	5574331	-	0.0018617
Panel/Email	17.296261	3.1	122752	
methodologyIVR/Online Panel	- 5.	2482548	-	0.0093971
	13.633674	2.5	977538	

	Std.	
	Estimate Error t value	$\Pr(> \! t)$
methodologyIVR/Online Panel/Email	- 5.1426875 -	0.0001926
	19.181827 3.7299227	
$methodology IVR/Online\ Panel/Text-to-Web$	- 5.0543198 -	0.0001065
	19.594610 3.8768045	
methodologyIVR/Online	- 5.3029985 -	0.0026983
Panel/Text-to-Web/Email	15.913793 3.0009047	
methodologyIVR/Text	- 5.1695978 -	0.0050499
	14.497848 2.8044440	
methodologyIVR/Text-to-Web	- 5.3274275 -	0.0000353
	22.047187 4.1384303	
methodologyIVR/Text-to-Web/Email	- 9.0613151 -	0.4113279
	7.444712 0.8215928	
methodologyLive Phone	- 4.9948811 -	0.0000106
	22.009337 4.4063785	
methodologyLive Phone/Email	- 5.4088784 -	0.0003710
	$19.261395 \qquad \qquad 3.5610700$	
methodologyLive Phone/Online Panel	- 5.1500099 -	0.0001147
	19.871909 3.8586158	
methodologyLive Phone/Online Panel/App	- 7.0126693 -	0.0000514
Panel	28.407277 4.0508507	
methodologyLive Phone/Online Panel/Text	- 5.6998052 -	0.0030673
	16.880690 2.9616257	
methodologyLive Phone/Online	- 5.1969825 -	0.0058012
Panel/Text-to-Web	14.340535 2.7593964	
methodologyLive Phone/Online	- 5.8205235 -	0.0129858
Panel/Text-to-Web/Text	14.461653 2.4845966	
methodologyLive Phone/Probability Panel	- 6.7920148 -	0.1233767
	10.465689 1.5408814	
methodologyLive Phone/Text	- 14.0219151 -	0.7781333
	3.950813 0.2817599	0.0000180
methodologyLive Phone/Text-to-Web	- 5.0227875 -	0.0000130
41 11 T. DI /D + 1771/A	21.910971 4.3623129	0.0018060
methodologyLive Phone/Text-to-Web/App	- 6.0056340 -	0.0013268
Panel (Translation Planel)	19.284147 3.2110093	0.5050004
methodologyLive Phone/Text-to-Web/Email	- 14.0268863 -	0.5252664
	8.910865 0.6352703	0.0550550
methodologyLive	- 6.4684401 -	0.2773772
Phone/Text-to-Web/Email/Mail-to-Web	7.026590 1.0862882	0.0000000
methodologyLive Phone/Text-to-	- 7.6956175 - 4.1740494	0.0000302
Web/Email/Mail-to-Web/Mail-to-Phone	32.121834 4.1740424	

	Std.	
	Estimate Error t value	$\Pr(> t)$
methodologyLive Phone/Text-to-Web/Online	- 7.6818415 -	0.0049558
Ad	21.589862 2.8105061	
methodologyLive Phone/Text/Online Ad	- 7.4749796 -	0.0586983
	14.132632 1.8906582	
methodologyLive Phone/Text/Online Panel	- 8.2277139 -	0.0026087
	24.775149 3.0111826	
methodologyMail-to-Web/Mail-to-Phone	- 6.4005546 -	0.0019972
	19.786943 3.0914419	
methodologyOnline Ad	- 5.4662855 -	0.0000026
	25.708962 4.7031869	
methodologyOnline Panel	- 4.9695024 -	0.0003204
	17.887315 3.5994179	
methodologyOnline Panel/Email	- 7.0214561 -	0.1314368
,	10.592440 1.5085816	
methodologyOnline	- 7.4842549 -	0.0099732
Panel/Email/Text-to-Web	19.288720 2.5772399	
methodologyOnline Panel/Online Ad	- 6.2383664 -	0.0940351
GV /	10.447035 1.6746427	
methodologyOnline Panel/Probability Panel	- 7.0121219 -	0.0156039
, and a second s	16.958524 2.4184582	
methodologyOnline Panel/Text	- 9.0638772 -	0.4706695
	6.538836 0.7214171	
methodologyOnline Panel/Text-to-Web	- 5.1077118 -	0.0011365
	16.626980 3.2552699	0.00==000
methodologyOnline Panel/Text-to-Web/Text	- 5.1391024 -	0.0003649
	18.323110 3.5654301	0.000000
methodologyProbability Panel	- 5.0203042 -	0.0011957
	16.269851 3.2408098	0.001100.
methodologyText	- 6.1228937 -	0.0253295
monodology text	13.694893 2.2366701	0.0200200
methodologyText-to-Web	- 5.3181586 -	0.0000041
methodology text-to- web	24.519690 4.6105602	0.0000041
methodologyText-to-Web/Online Ad	- 5.4576811 -	0.0198471
methodology text-to- web/ Omme Ad	12.714183 2.3295943	0.0150471
transparency score	- 0.0825576 -	0.0000000
transparency_score	1.437809 17.4158377	0.0000000
hypotheticalTRUE	- 0.4346440 -	0.0000000
ny pouncticar i ito is	7.695244 17.7047033	0.0000000
	1.030244 11.1041030	

Table 2: ?(caption)



3.2 Lasso Regularization

Lasso Regularization is performed to see if the selected variable in our model would best predict the results of polls percentage of a candidate. If the variable do not align with our actual numbers, we should consider other variables or a interaction variable.

Computing the Mean Square Error would give us a sense of our lasso model prediction.

Mean Squared Error: 342.4718

R-squared: 0.07872276

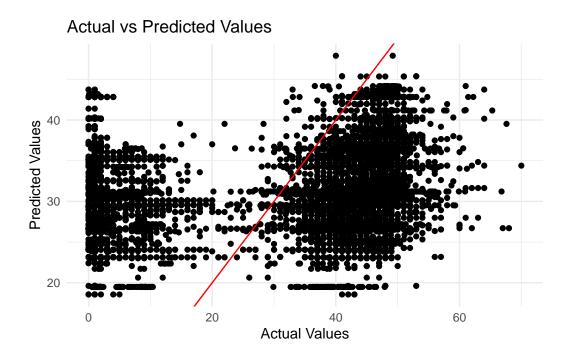
Table 3: Mean Squared Error

Metric	Value
SST SSE	3.802858e+06 3.503487e+06
R-squared	7.872280e-02

From this graph, we can see that the cluster points do not perfectly align on the line of the best fit so we should still investigate in other variables.

Table 4: actual vs predicted

Actual	Predicted
50	43.17871
46	43.17871
48	37.02916
45	37.02916
1	37.02916
1	37.02916



Appendix

A Additional data details

A.1 pollster methodology analysis

This survey was conducted by Redfield & Wilton Strategies to assess the voting intentions of eligible voters in key U.S. swing states ahead of the 2024 Presidential Election. The primary goal of this poll is to provide an accurate and timely snapshot of public opinion in states where electoral outcomes are uncertain and could have a decisive impact on the overall result of the election. Swing states, due to their political volatility and diverse voter bases, are critical in determining the balance of power in the U.S. electoral system. Understanding voter preferences in these states is essential for political analysts, campaigns, and the general public.

A.1.1 Population of Interest

The population of interest for this survey consists of all eligible voters residing in major U.S. swing states, specifically Arizona, Florida, Georgia, Michigan, North Carolina, and Pennsylvania. These states are known for their fluctuating political alignments and are expected to play a crucial role in the upcoming election.

A.1.2 Sampling Frame

The population sampled includes eligible voters from Arizona, Florida, Georgia, Michigan, North Carolina, and Pennsylvania. Participants were selected via an online panel.

A.1.3 Sample

The sample sizes for each state were as follows:

Arizona: 750 respondents

Florida: 1,350 respondents

Georgia: 927 respondents

Michigan: 970 respondents

North Carolina: 880 respondents

Pennsylvania: 1,070 respondents

A.2 Weakness & Strength of the methodology

In terms of strengths, Redfield & Wilton has a great reputation for producing reliable polling data. They have employ a mix of online and telephone survey, which add in various resources of collecting their data. This approach can help reduce bias. In addition, Redfield & Wilton often target swing states, which makes their polls results important to the US president election.

The weakness of their methodology would incorporate a certain potential bias based on their political leaning of their clients or media. This could cause a certain neutrality in their dataset.

Overall, Redfield & Wilton has been considered as a competent, reputable and reliable pollster with his variaty and methodology on polls. Even though the pollster contained a potential bias, it has been a reliable resource in prediction of US president election.

B Model details

References

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

R Core Team. 2023. R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for Statistical Computing. https://www.R-project.org/.