

Models & Markets

Comparing 2018 Midterm Predictions

Question

- What is the best way to predict elections?
- When does each method perform best? Why?

Hypothesis

- Prediction **markets** will outperform **models**

Why Predict Elections?

- Resource allocation
- Strategy adjustment
- Quantitative journalism
- Uncertainty is scary!

How to Predict Elections?

1. Opinion polling
2. Poll aggregation
3. Forecasting models
4. Prediction markets

Forecasting Models

Forecasting Models

- 21st Century Invention
- Quantitative inputs
- Express uncertainty
- Probabilistic
- Proprietary



Who will win the presidency?



Chance of winning



Electoral votes

■ Hillary Clinton	302 . 2
■ Donald Trump	235 . 0
■ Evan McMullin	0 . 8
■ Gary Johnson	0 . 0

Popular vote

■ Hillary Clinton	48 . 5%
■ Donald Trump	44 . 9%
■ Gary Johnson	5 . 0%
■ Other	1 . 6%

FORECAST

PRESIDENT SENATE

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CLINTON
98.0%

TRUMP
1.7%



Model Process

Most election models work in something like the following way: [1] they calculate the most likely outcome in a particular state... and then [2] they determine the degree of uncertainty around that estimate.

– Nate Silver

[1] Model Inputs

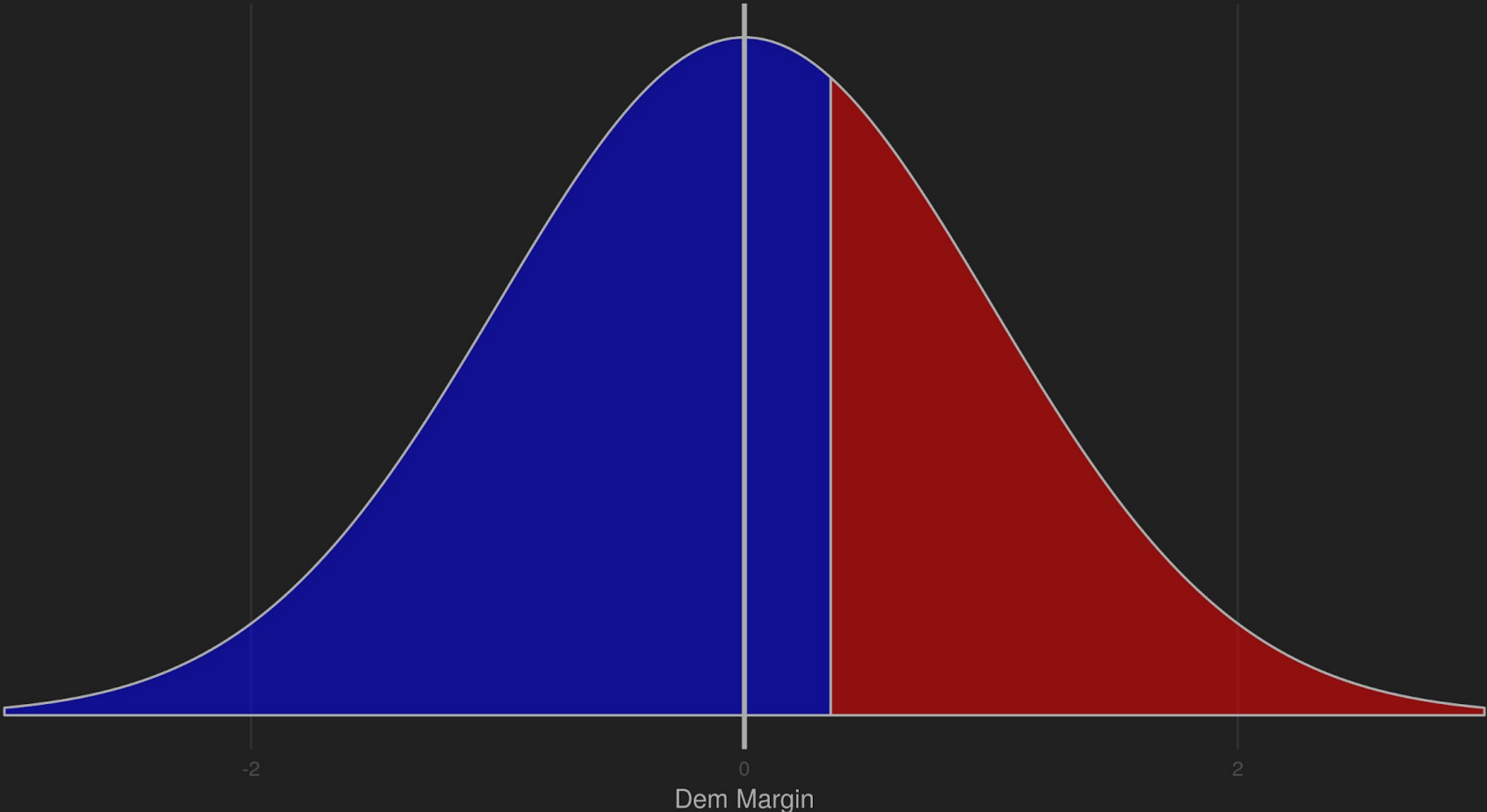
1. Weighted poll aggregation
2. CANTOR imputation
3. Fundamentals
 - incumbency
 - fundraising
 - previous election
 - scandals, etc

[2] Model Uncertainty

- Fewer polls
- Lopsided race
- Further away
- Polls disagree
 - With one another
 - With fundamentals

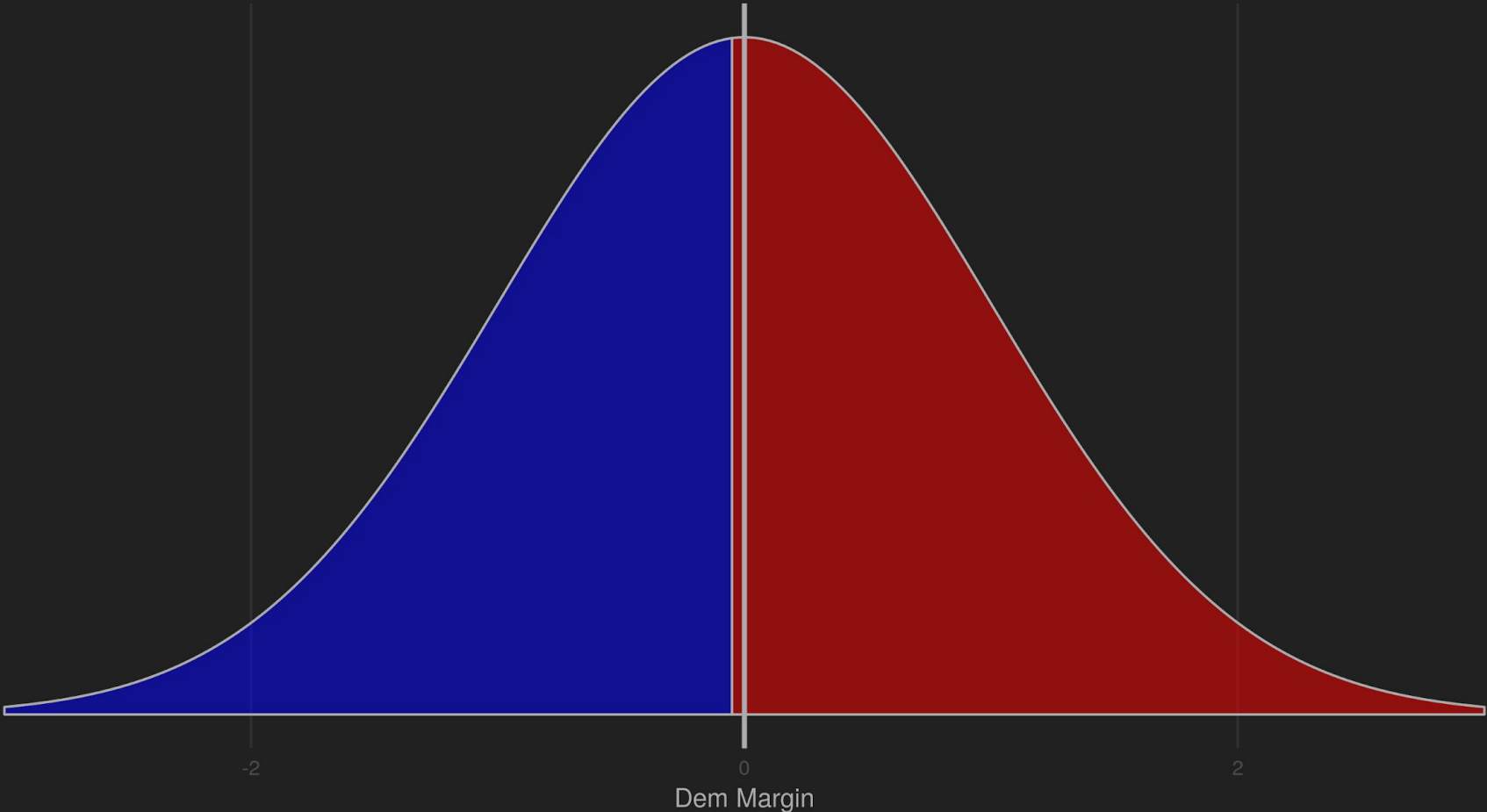
Model Probability Distribution

Winner = DEM



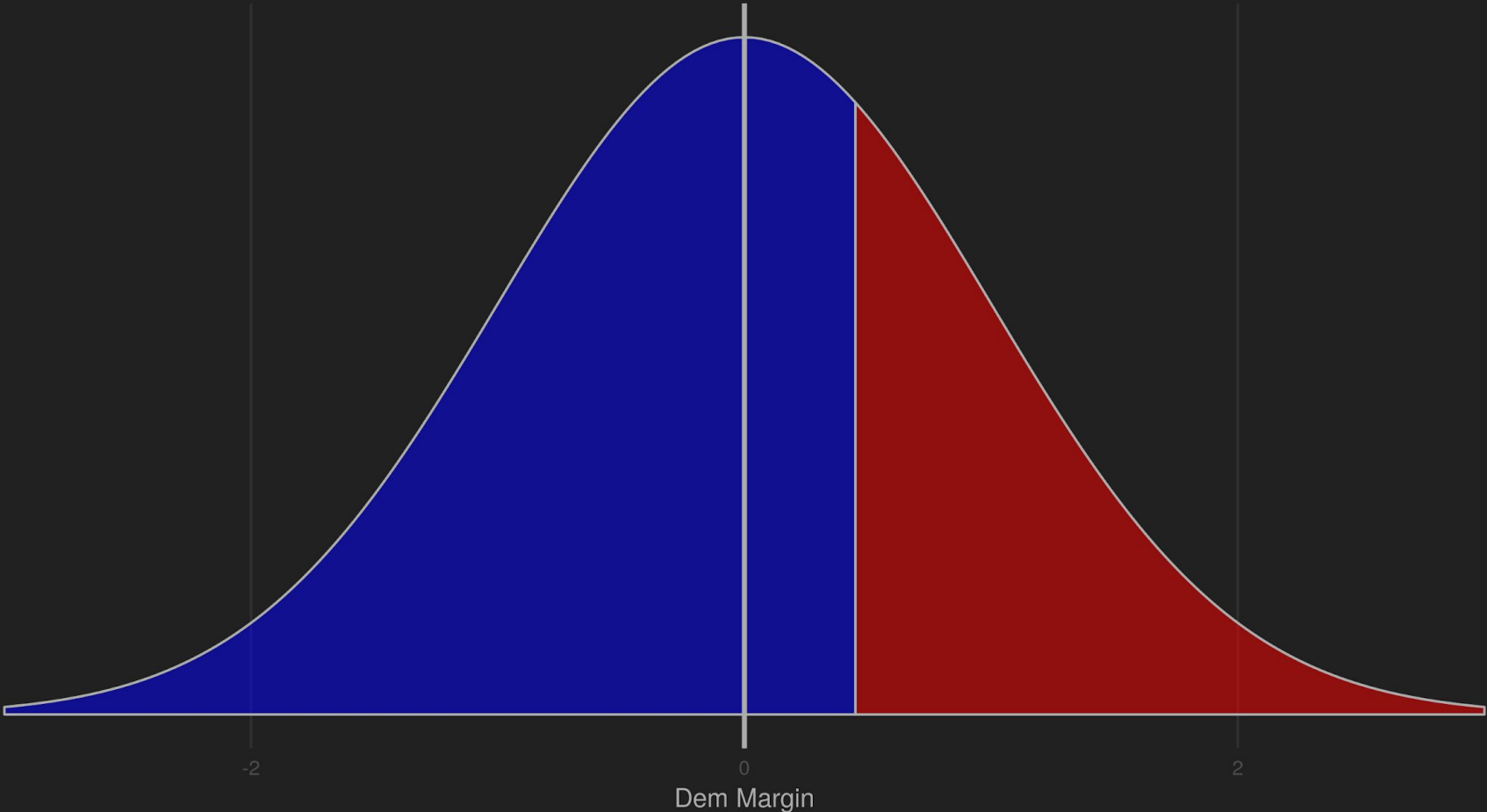
Model Probability Distribution

Winner = GOP



Model Probability Distribution

Winner = DEM



Model Variables

1. Date
2. Election
3. Party
4. Special
5. Incumbency
6. Voteshare(s)
- 7. Probability**







Prediction Markets

Prediction Markets

- Exchange-traded binary options
- Efficient market hypothesis
- Elections, terrorism, shipwrecks
- Dubious legality
 - Academic waiver from CFTC

Market Contracts

- \$850 limit
- Buyers on either side
- \$1 or \$0 based on outcome
- Sell at any time
- Price change

Contract	Latest Yes Price	Best Offer	Best Offer
	Joe Biden 24¢ 1¢↑	25¢ <div> Buy Yes Buy No </div>	76¢
	Bernie Sanders 22¢ 1¢↓	22¢ <div> Buy Yes Buy No </div>	79¢
	Pete Buttigieg 18¢ NC	19¢ <div> Buy Yes Buy No </div>	82¢
	Kamala Harris 14¢ NC	15¢ <div> Buy Yes Buy No </div>	86¢
	Andrew Yang 9¢ 1¢↓	10¢ <div> Buy Yes Buy No </div>	91¢
	Elizabeth Warren 8¢ 1¢↑	9¢ <div> Buy Yes Buy No </div>	93¢

Market Variables

1. ID
2. Question
3. Name
4. Contract
5. Date
6. Volume
- 7. Price**

Comparing Methods

Reading data

```
"https://fivethirtyeight.com/data/..." %>%  
  read_memento(timestamp = "2018-11-06") %>%  
  read_csv(col_types      = cols(  
    Democrat_Won      = col_logical(),  
    Republican_Won    = col_logical(),  
    uncalled           = col_logical(),  
    forecastdate       = col_date("%m/%d/%y"),  
    category           = col_factor())
```

Formatting data

```
model_district %>%  
  bind_rows(model_seat, .id = "chamber") %>%  
  unite(state, district,  
        col = race,  
        sep = "_") %>%  
  rename(date = forecastdate,  
         prob = win_probability) %>%  
  filter(name != "Others") %>%  
  arrange(date, name)
```

Models

1. Date
2. Election
3. Party
4. Voteshare
- 5. Probability**

Markets

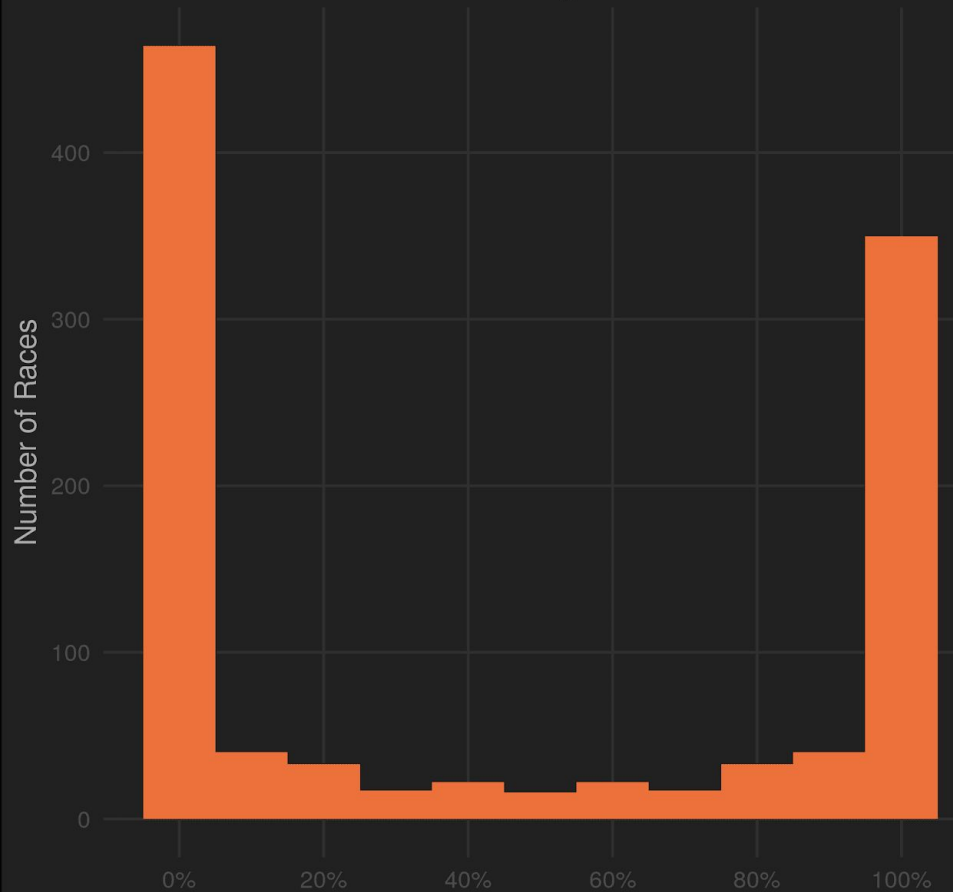
1. Date
2. Market
3. Contract
4. Volume
- 5. Price**

Comparing data

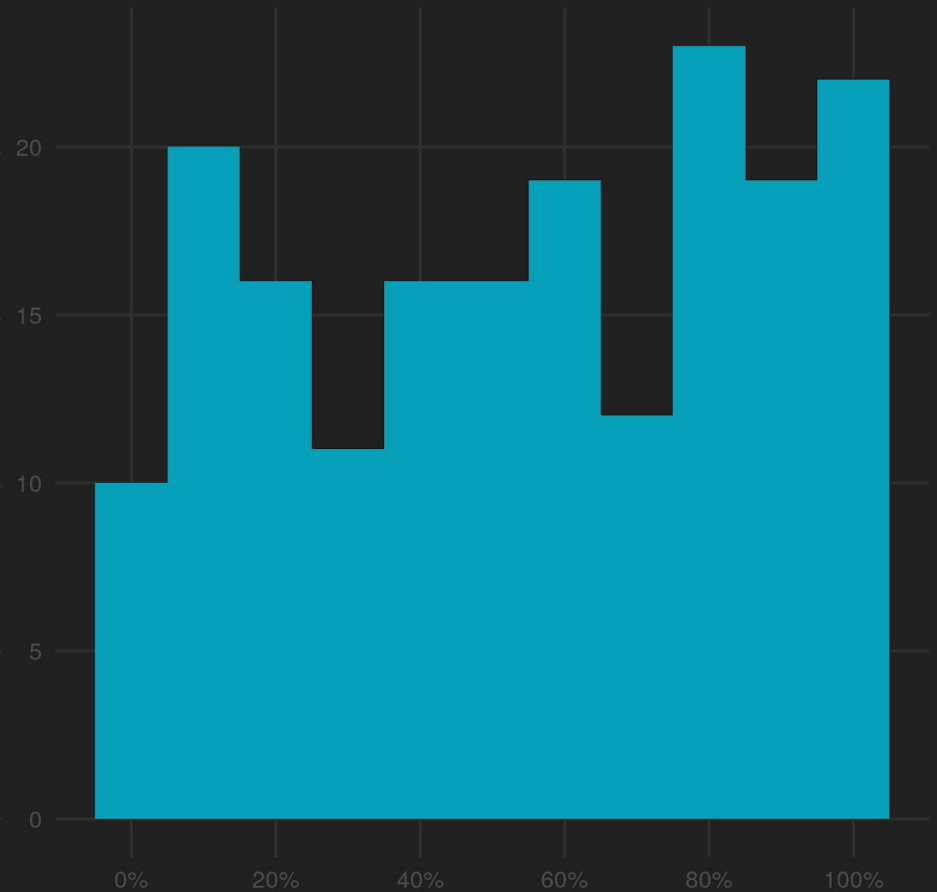
```
markets_dem %>%  
  inner_join(model_dem) %>%  
  rename(model = prob,  
         market = close) %>%  
  gather(key = method,  
         value = prob) %>%  
  inner_join(results) %>%  
  mutate(brier = (winner - prob)^2) %>%  
  mutate(hit = (prob > 0.5) == winner)
```

Distribution of Race Probabilities by Predictive Method

Forecasting Model



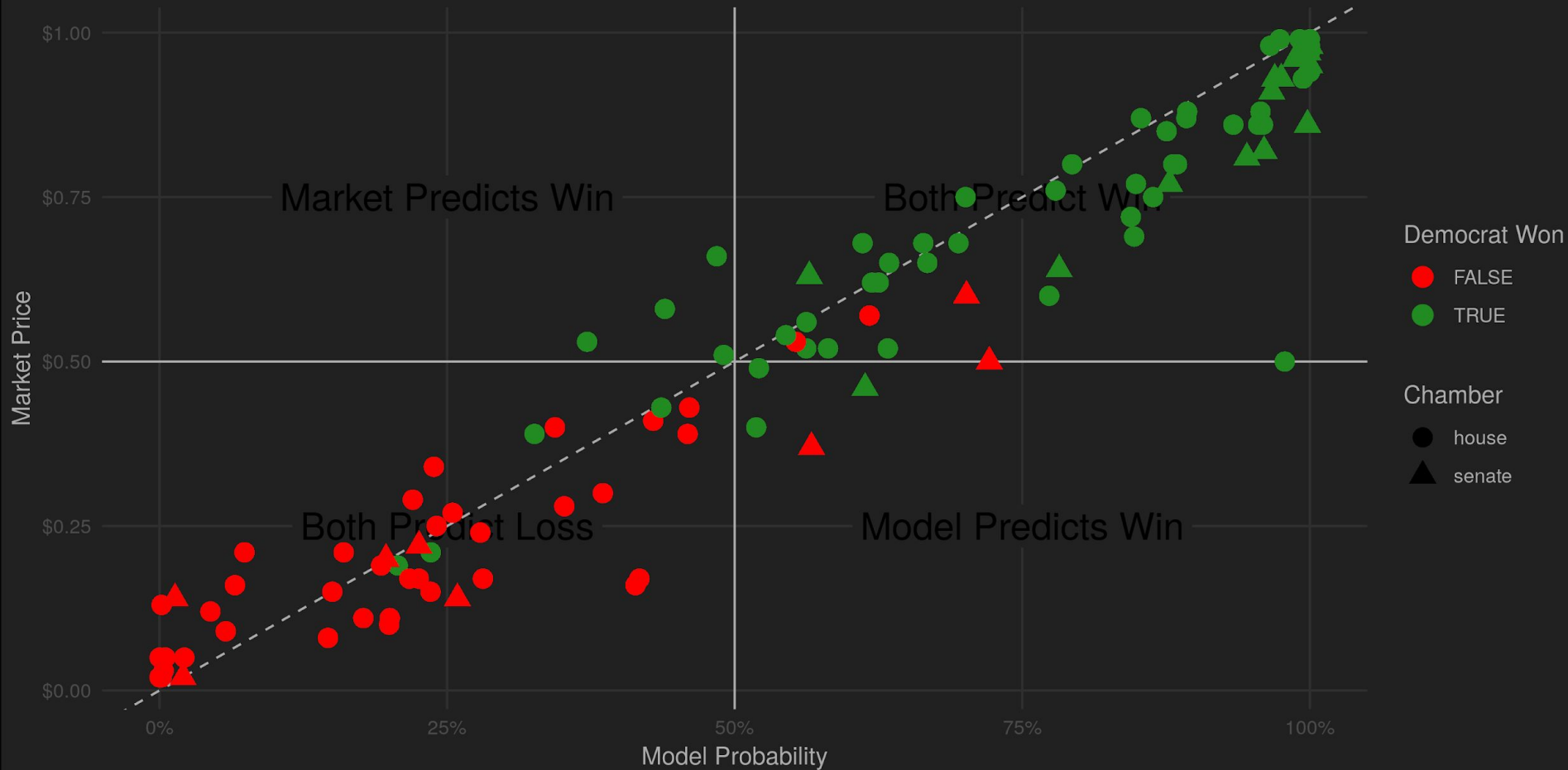
Prediction Markets



Democratic Win Probability

Midterm Races by Democrat's Chance of Winning

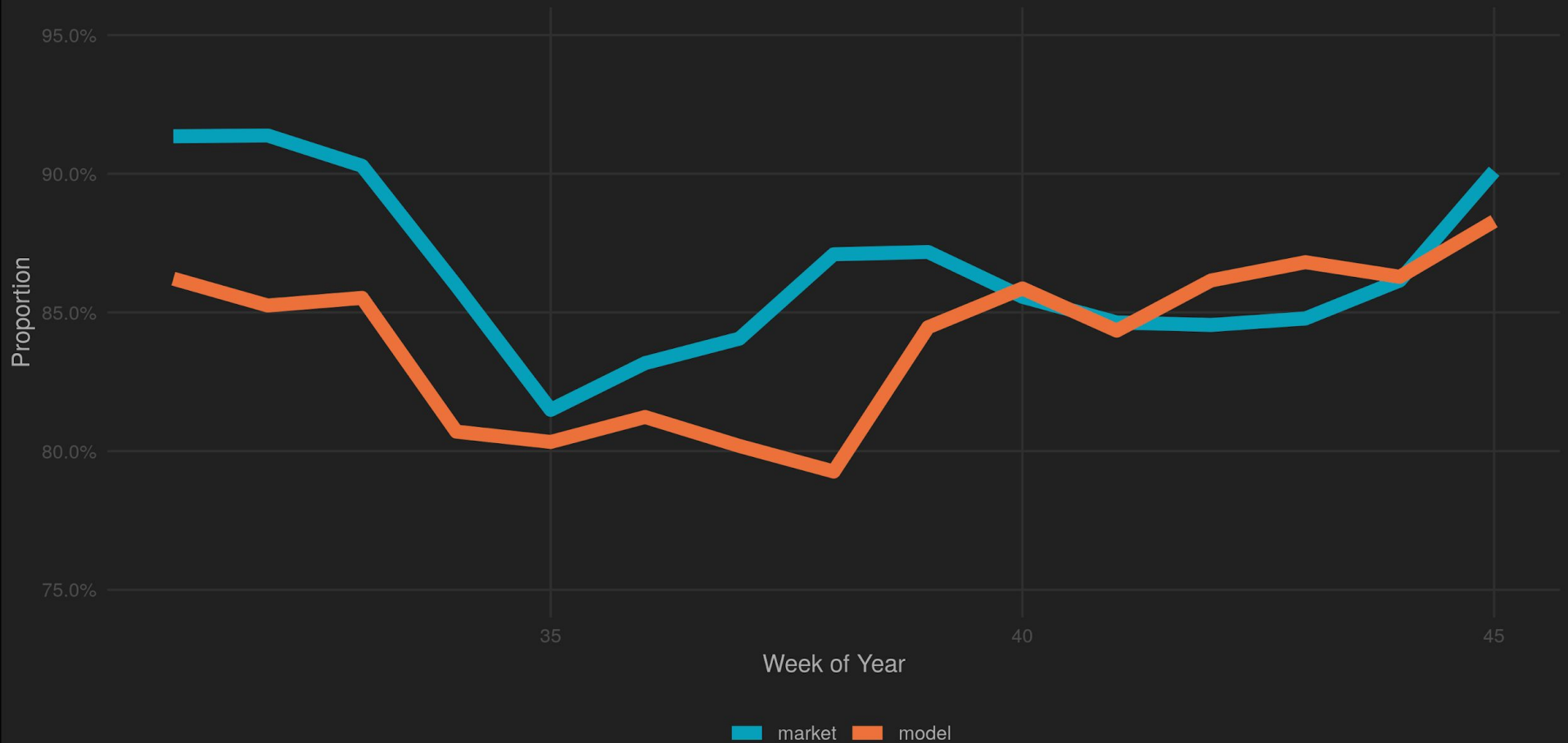
November 5th, Night Before Election Day



<i>Date</i>	<i>Race</i>	<i>Method</i>	<i>Prob</i>	<i>Pred</i>	<i>Win</i>	<i>Hit</i>
8/1	AZ-S1	Market	66	1	1	1
8/1	AZ-S1	Model	74	1	1	1
8/1	CA-12	Market	91	1	1	1
8/1	CA-12	Model	100	1	1	1
8/1	CA-22	Market	30	0	0	1
8/1	CA-22	Model	5	0	0	1
8/1	CA-39	Market	61	1	1	1
8/1	CA-39	Model	38	0	1	0

Proportion of Correct Predictions by Week

PredictIt Markets and FiveThirtyEight Model



Proportions Test

- Market proportion: 86.0343%
- Model proportion: 83.8057%
- Lower bound: 1.158%
- Upper bound: 3.2999%
- X-squared = 16.794
- p-value = 0.000042
- Alternative hypothesis: Two sided

		TRUE		FALSE	
		Markets	Model	Markets	Model
TRUE		79.5 (51.2)	84.5 (51.2)	40.6 (9.7)	36.5 (9.7)
FALSE		59.3 (4.3)	63.7 (6.5)	23.0 (34.3)	16.8 (32.1)

Brier Score

- Democrat has 70% chance
- Democrat wins on election day
- Brier Score = $(0.70 - 1)^2 = 0.09$

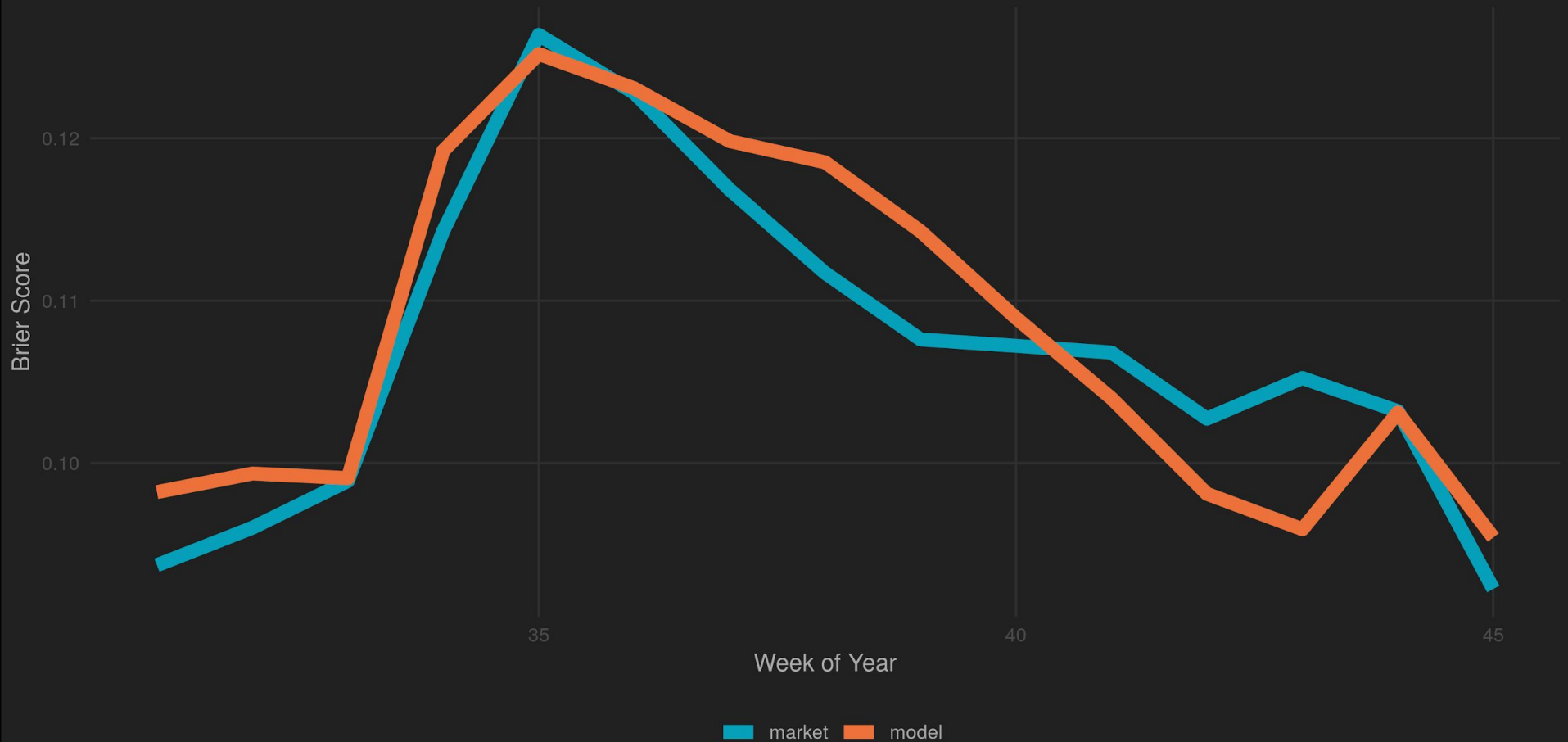
$$BS = \frac{1}{N} \sum_{t=1}^N (f_t - o_t)^2$$

Brier T-Test

- Market mean: 0.1084
- Model mean: 0.1091
- Lower bound: -0.005
- Upper bound: 0.0035
- $t =$ -0.33902
- p-value = 0.7346
- Alternative hypothesis: Difference

Brier Score Predictions by Week

PredictIt Markets and FiveThirtyEight Model



Expected Probabilities and Actual Proportions of Democratic Victory

Expected probabilities binned by rounding to the nearest 10%

