# Where's the Money? Campaign Finance and Election Outcomes in the 2016 U.S. Congressional Races

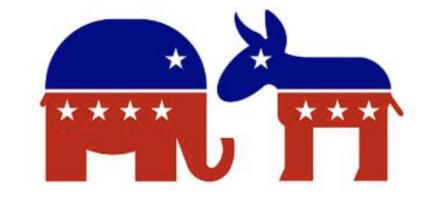
# Project Overview

#### Interest

 Voters, candidates, and news stations often ask what factors go into a winning campaign

## **Goal & Project Type**

- Predict whether a candidate is a "Winner" or "Loser" based on campaign finance data
- Binary Classification



## Data Processing

## **Feature Review and Engineering**

• Raw Data: 1814 candidates, 51 features

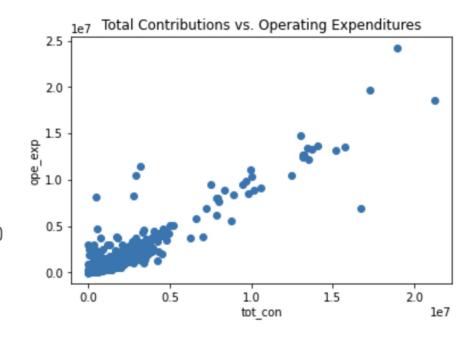
- Processed data: 1656 candidates, 20 features
  - o 17 numeric
  - 3 categorical variables

	can_id	can_nam	can_off	can_off_sta	can_off_dis	can_par_aff	can_inc_cha_ope_sea	can_str1	can_str2	can_cit	•••	cas_on_han
0	H2GA12121	ALLEN, RICHARD W	Н	GA	12.0	REP	INCUMBENT	2237 PICKENS RD	NaN	AUGUSTA		
1	H6PA02171	EVANS, DWIGHT	Н	PA	2.0	DEM	CHALLENGER	PO BOX 6578	NaN	PHILADELPHIA		
2	H6FL04105	RUTHERFORD, JOHN	н	FL	4.0	REP	OPEN	3817 VICKERS LAKE DRIVE	NaN	JACKSONVILLE		
3	H4MT01041	ZINKE, RYAN K	Н	МТ	0.0	REP	INCUMBENT	409 2ND ST W	NaN	WHITEFISH		
4	H8CA09060	LEE, BARBARA	Н	CA	13.0	DEM	INCUMBENT	409 13TH ST, 17TH FL	NaN	OAKLAND		

# Data Processing (continued)

## **Data Summary**

- Feature highlights
  - Contributions
  - Loans
  - Operating Expenditures
  - Length of Campaign
  - Missing Values
- Final Sample distribution: 1186 losers, 470 winners



## Model 1: Decision Trees

### **Decision Tree Classifier**

- One-Hot Encoder for categorical variables
  - Challenger, Incumbent, or Open Race Status
  - House or Senate Race
  - o Democrat, Republican, Other
- Training on 80% of data, Testing on 20% of data
- Weights: Winners at 2.5, Losers at 1
- Parameter Setting

# Model 1: Results Analysis

#### **Decision Tree Classifier**

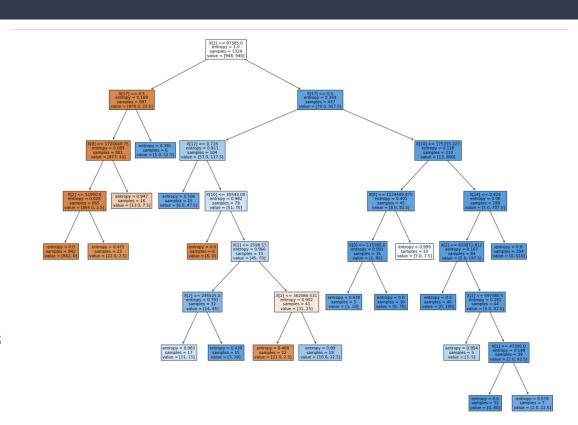
• Accuracy: 94.3%

#### Top Features:

- Other Committee Contributions (83%)
- Incumbent Status (10%)

#### Results Analysis

- 19 prediction errors out of 332 candidates
- Overpredicted that candidates won races
- Over-reliant on PACs and Super PACS contribution data



## Model 2: Random Forest Classifier

## **Random Forest Classifier**

- Similar setup as decision tree classifier
- Parameter Settings
- Accuracy: 95.78%
- 14 prediction errors out of 332 candidates

# Model 2: Results Analysis

	feature	importances				
2	oth_com_con	0.184804				
17	(Incumbent,)	0.126577				
4	tot_con	0.112342				
14	can_con_tot_con_ratio	0.103925				
13	oth_con_tot_con_ratio	0.092812				
8	tot_rec	0.079900				
10	cas_on_han_clo_of_per	0.058828				
9	ope_exp	0.055860				
11	campaign_days_total	0.044882				
0	ind_con	0.039824				
12	ind_con_tot_con_ratio	0.028748				
16	(Challenger,)	0.024326				

#### **Random Forest Classifier**

- Top Features
  - Super PACs
  - Incumbent Status
  - Total Contributions
  - Contribution Ratios
- By generating multiple decision trees and then taking the prediction that most of the trees return, the resulting model weights more features evenly rather than relying heavily on one feature.
- Similar to the other model, we see that the model mislabeled losers as winners.

## Conclusion

 Important features: PACs and Super PACs contributions, as well as incumbent status

- Limitations:
  - Effects of geography on outcome
  - Challenging to use election data because of unbalanced samples