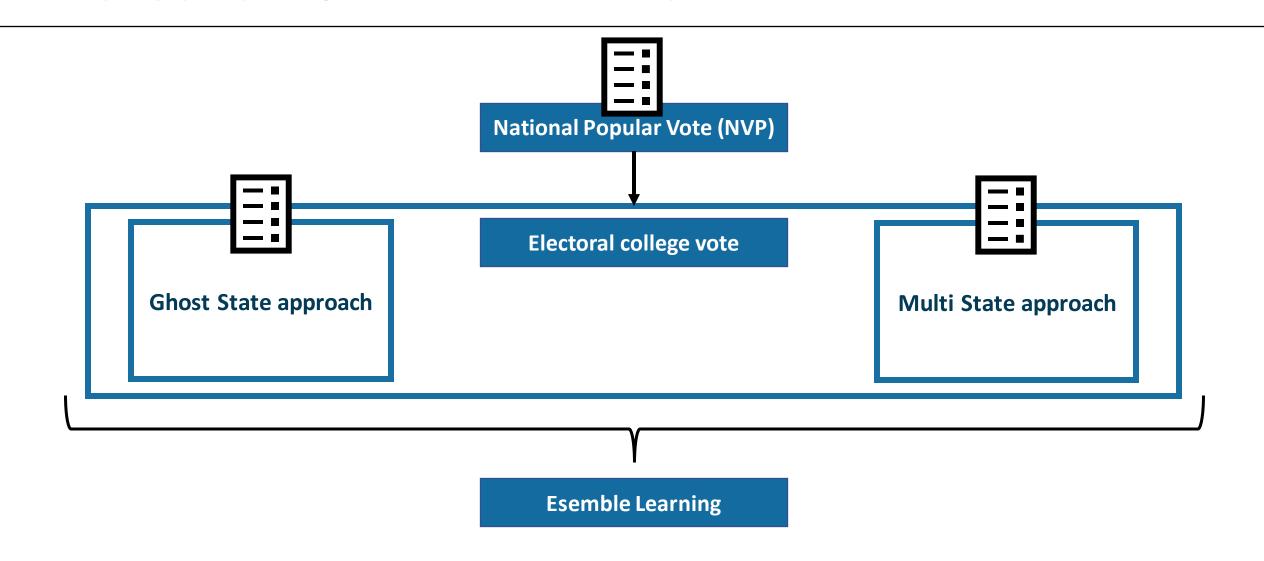
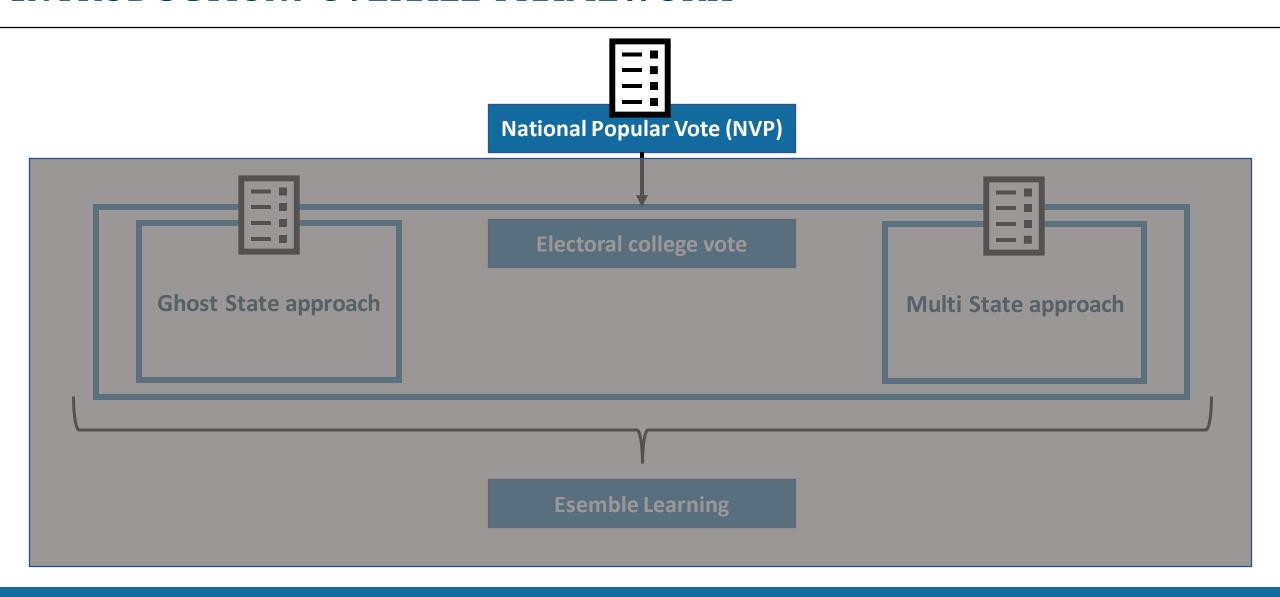
# PREDICTING THE US 2020 PRESIDENTIAL ELECTIONS







# NATIONAL POPULAR VOTE



#### **Database**

For every year since 1968:

- MacroEconomic data drawn from FRED
- Polls drawn from 5.38

**Dataset**: 26 features and 18 observations (one per election year)

How to handle such situation with sparse data?



#### **Modelling**

Favor sparse-like models:

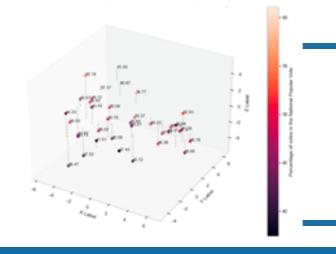
- Feature Engineering combing different predictors
- PCA
- Lasso-Like models
- Cross Validation based on Accuracy

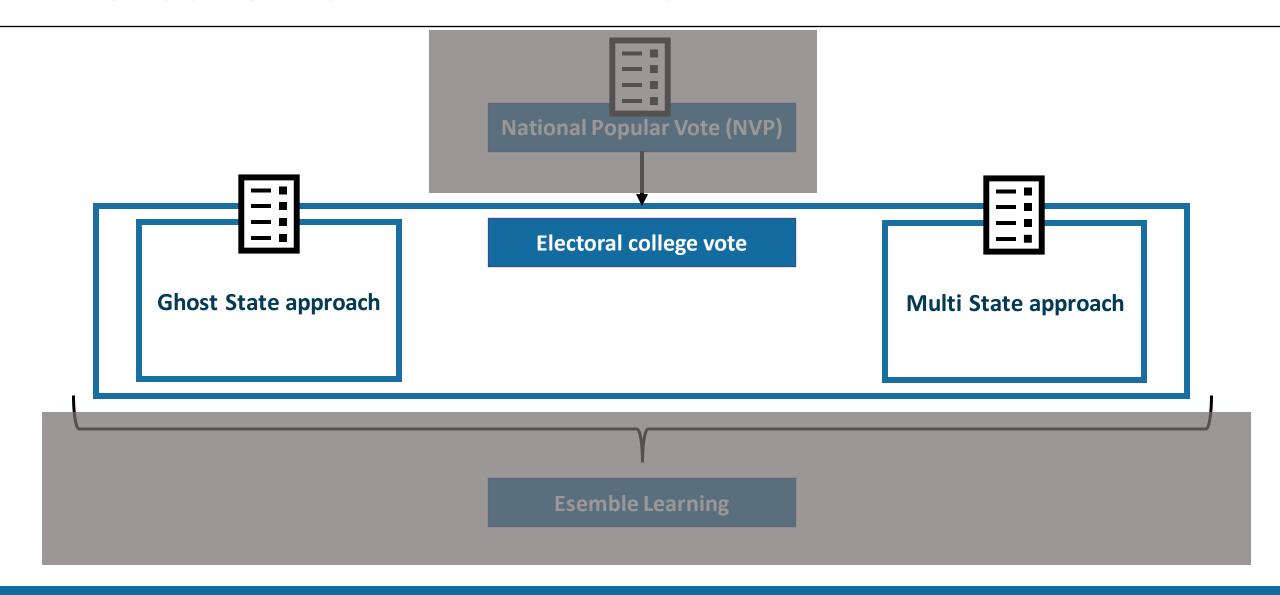


#### Results

Lasso Regression, with accuracy score 0.73 Predictions:

- For Donald Trump: NPV = 46.525 +/- 1.39
- For Joe Biden: NPV = 52.46 +/- 1.88





- A. Ghost State Approach
- B. Multi State Approach



#### **Database**

- 9 Features, 444 Observations (394 for Train/CV, 50 for Test)
- Polling data (5.38), Macroeconomic data (FRED)



#### **Modelling**

#### **Tested Models:**

- LogisticRegressionCV, DecisionTreeClassifier, RandomForestClassifier, AdaBoostClassifier, NeuralNetwork
- Feature Engineering using Polynomial Features, PCA
- Lasso Regularization, Dropout, EarlyStopping
- Cross-Validation based on classification accuracy for parameters tunning and model selection

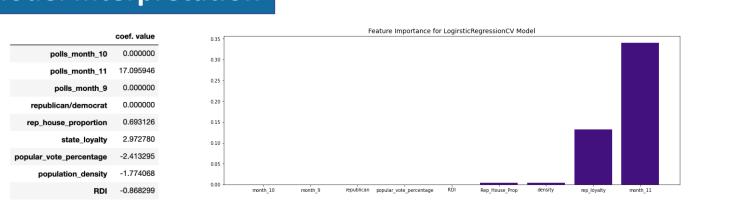
	LogisticRegressionCV	DecisionTreeClassifier	RandomForestClassifier	AdaBoostClassifier	SequentialNeuralNetwork
Accuracy	0.88585	0.88075	0.8987	0.87831	0.8992

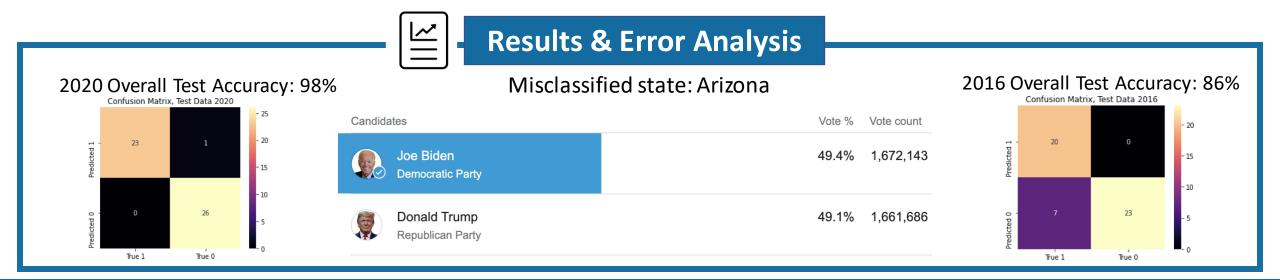
• Cross-validation accuracies dictated by the data, rather than model's expressive power LogisticRegressionCV chosen as final model for the Ghost State Approach

- A. Ghost State Approach
- B. Multi State Approach

### **Model Interpretation**

- polls\_month\_11 most important feature
- pairwise correlation between polls
- positive relationship: rep\_house\_proportion, state\_loyalty
- negative relationship: population density
- multicolinearity: popular\_vote\_percentage





- A. Ghost State Approach
- B. Multi State Approach

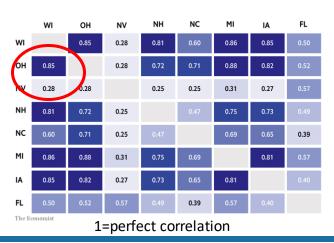
Predict each state seperately

# **Not swing states**

- Deterministic approach
- Based on loyalty feature of the state

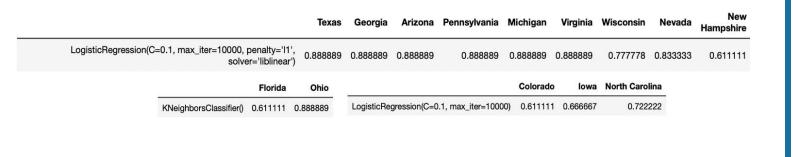
# Predict each state seperately Construct the dataframes

Correlation matrix from The Economist



# **Swing states**

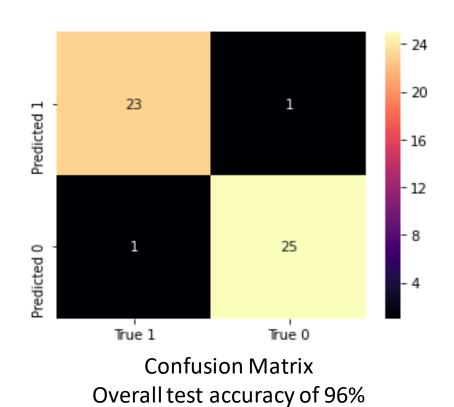
- More data points → add correlated states
- Model can learn the correlation between states



#### A. Ghost State Approach

#### B. Multi State Approach

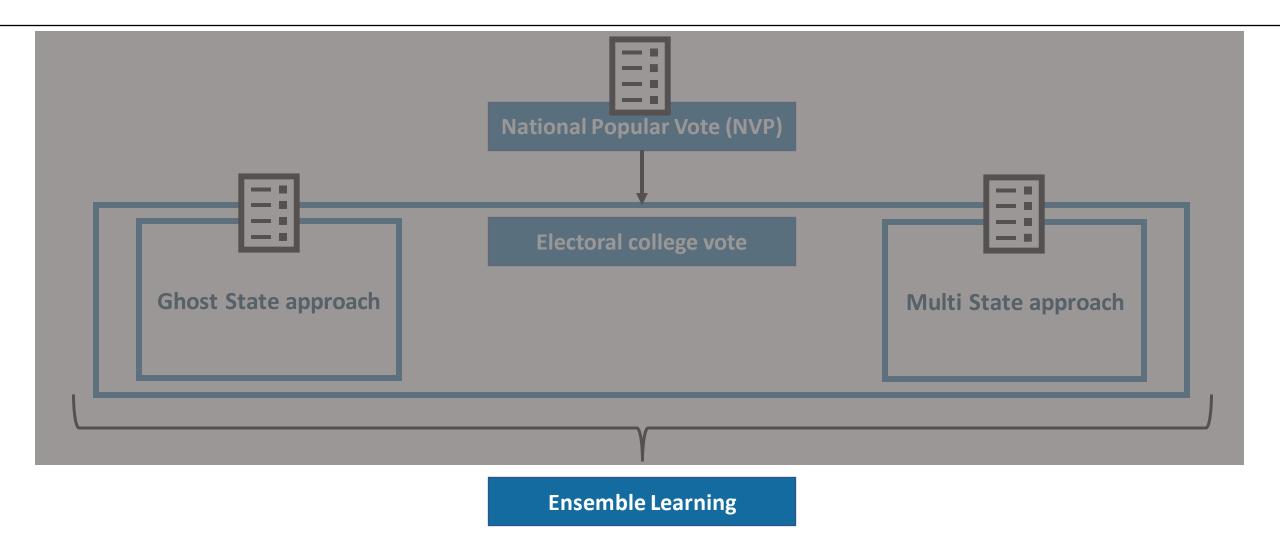
# **Overall results**



Average feature weighs for PC 1 Average feature weighs for PC 2 Average feature weighs for PC 3 Weight Features Features month\_10 month\_9 popular\_vote\_percentage Rep\_House\_Prop

Average feature importance across all swing states models

→ The models doesn't rely on polls



# **ENSEMBLE LEARNING**



#### **Database**

Use the output from the Multi State Model and the Ghost State Model

 Probabilities output on every election for every state for the two models

Dataset: 394 examples, 2 features



## Modelling

Favor interpretable results:

- Logistic Regression (being a weighted sum of the two predictors)
- Cross Validation based on Accuracy



#### Results

- Final model: 0.98 classification accuracy
- Misclassified state: Arizona
- Weights to the different models: 4.89 for the single state, 0.92 for the multi state