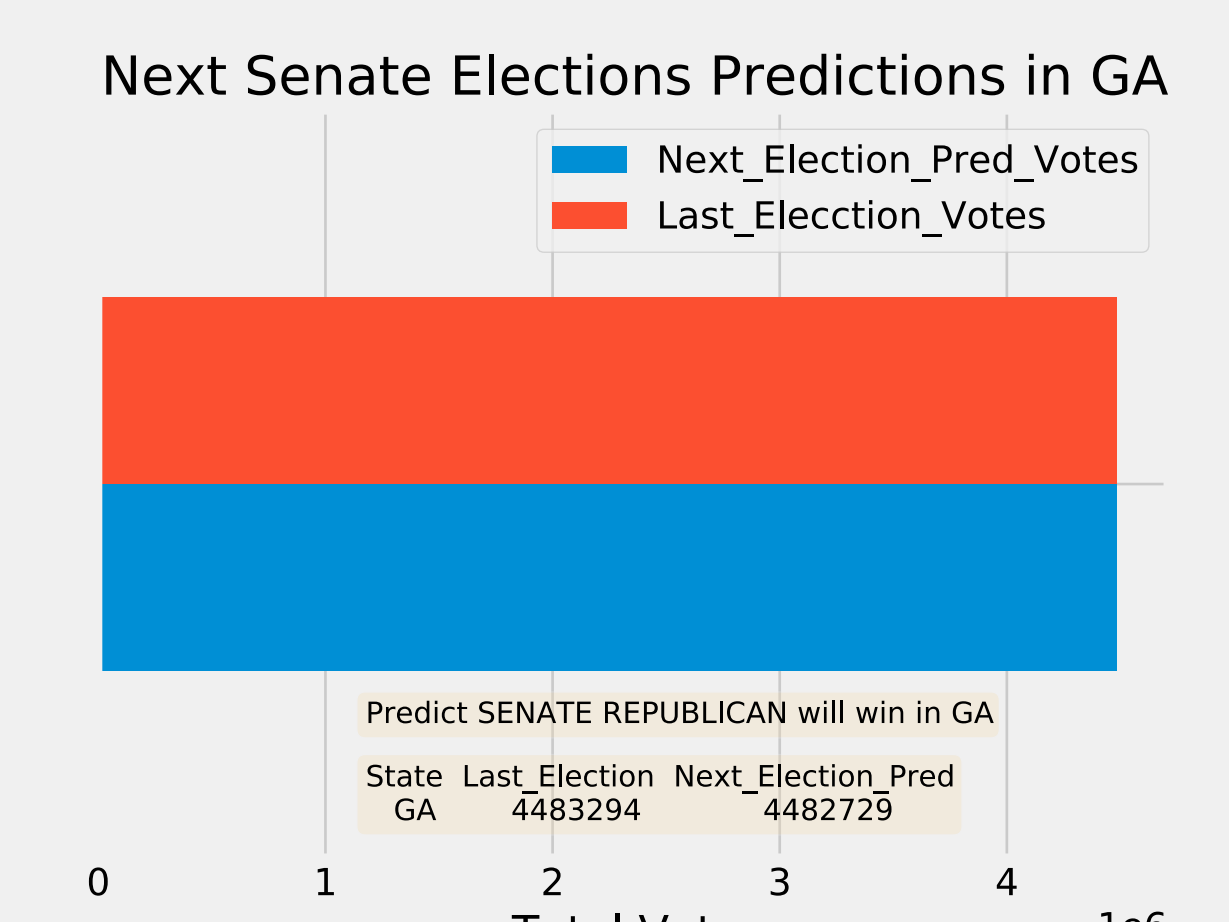
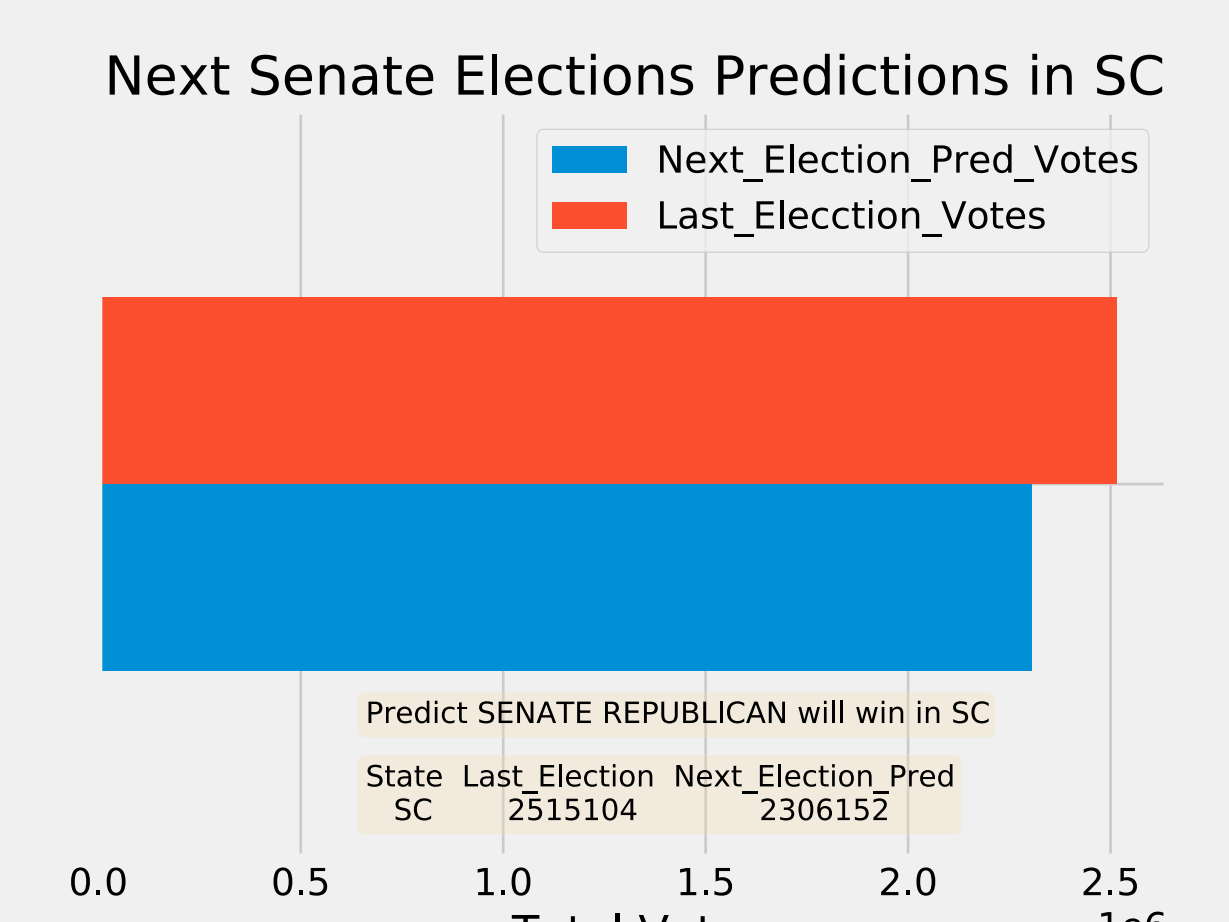
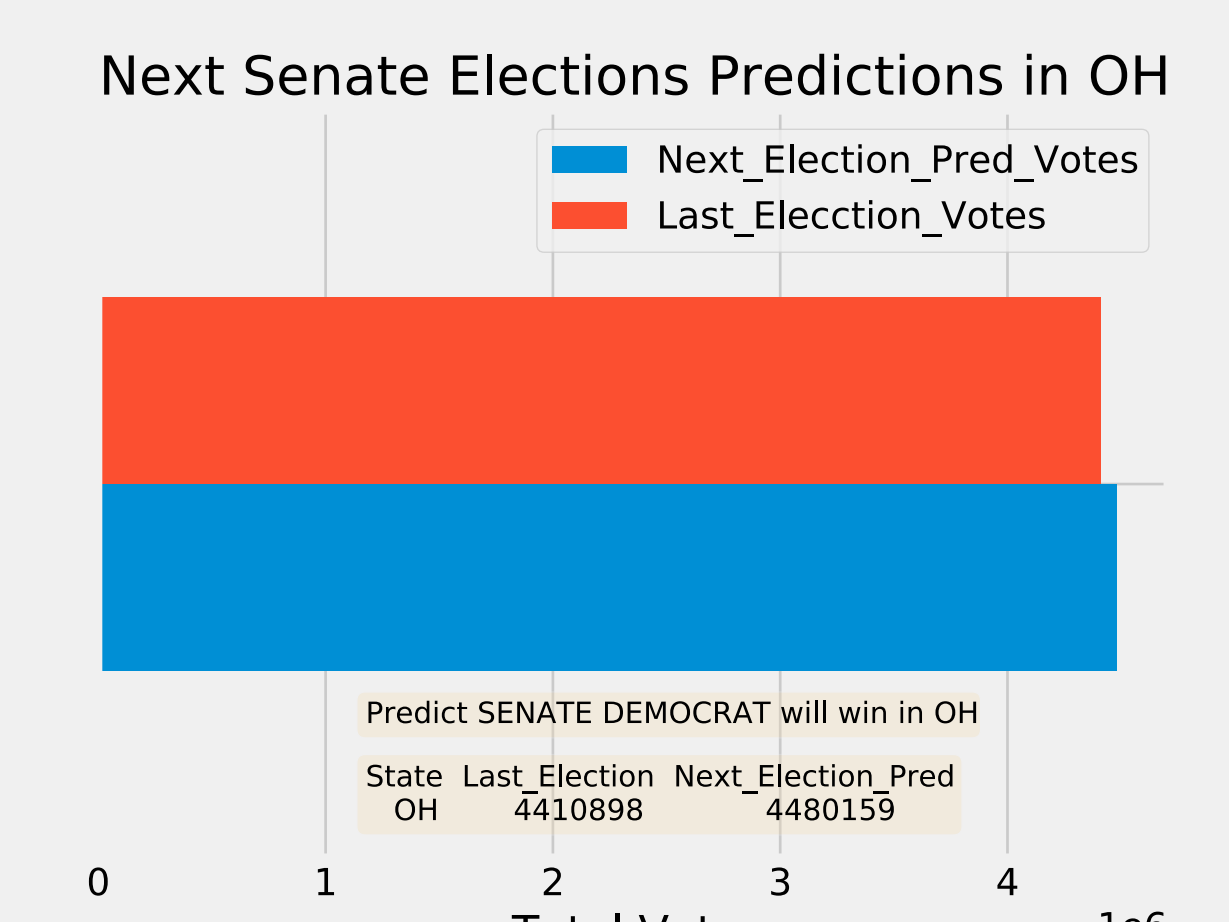
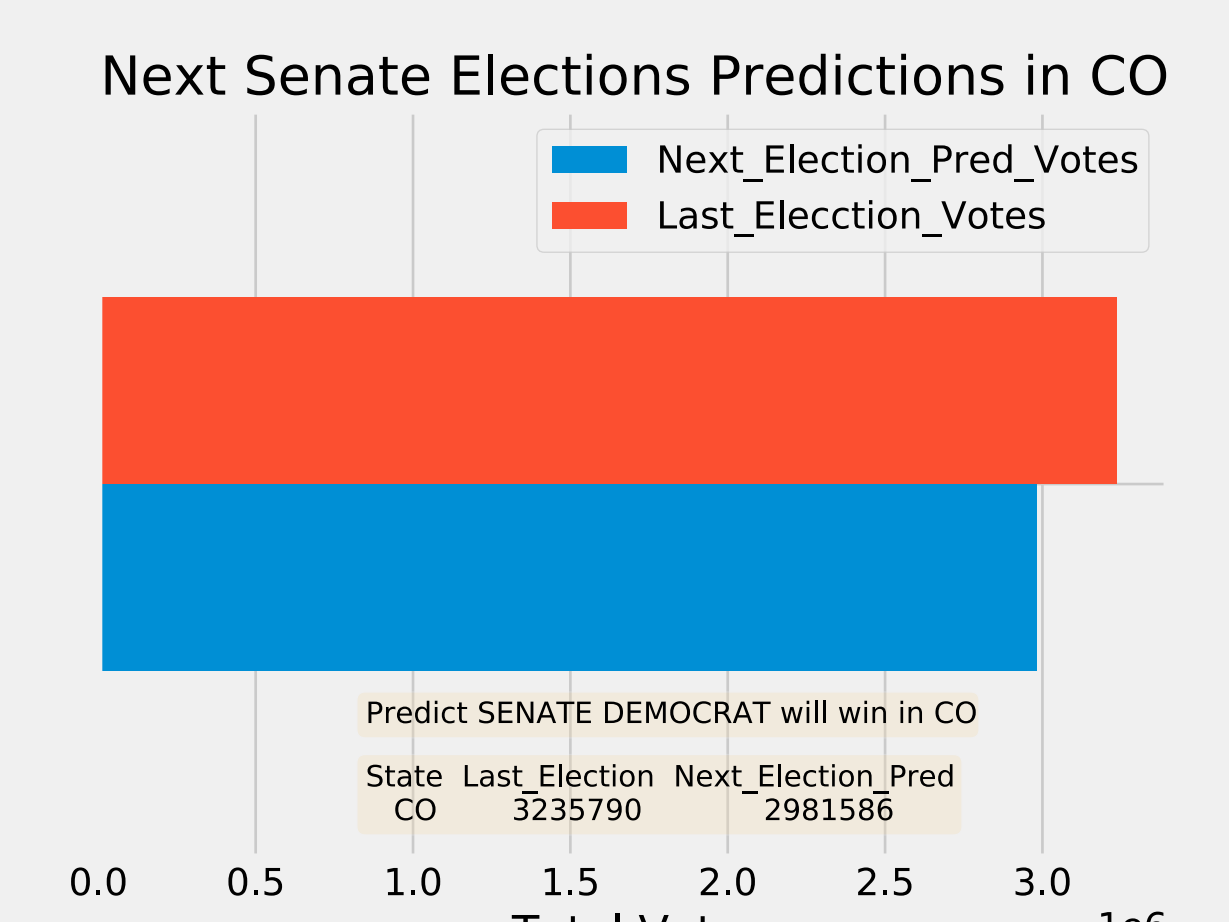
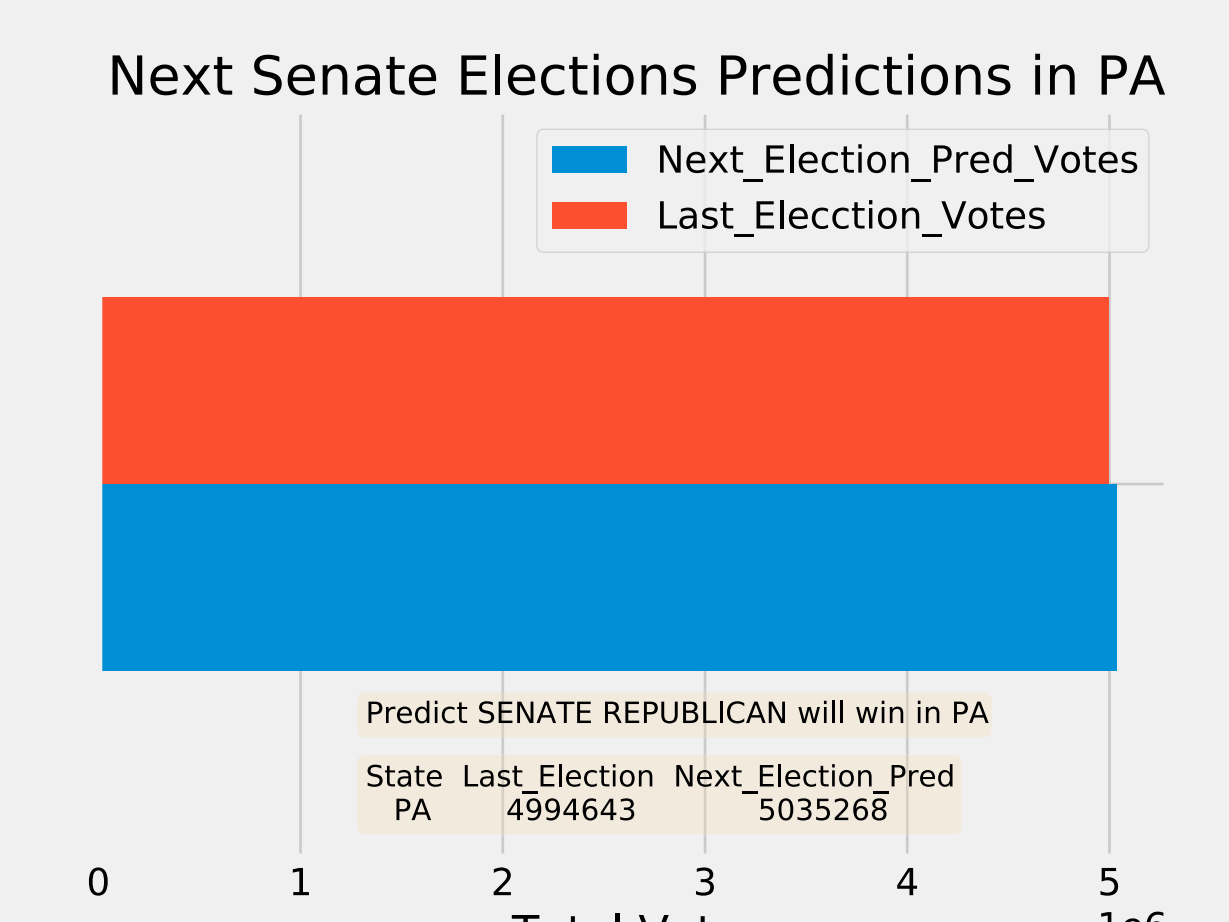
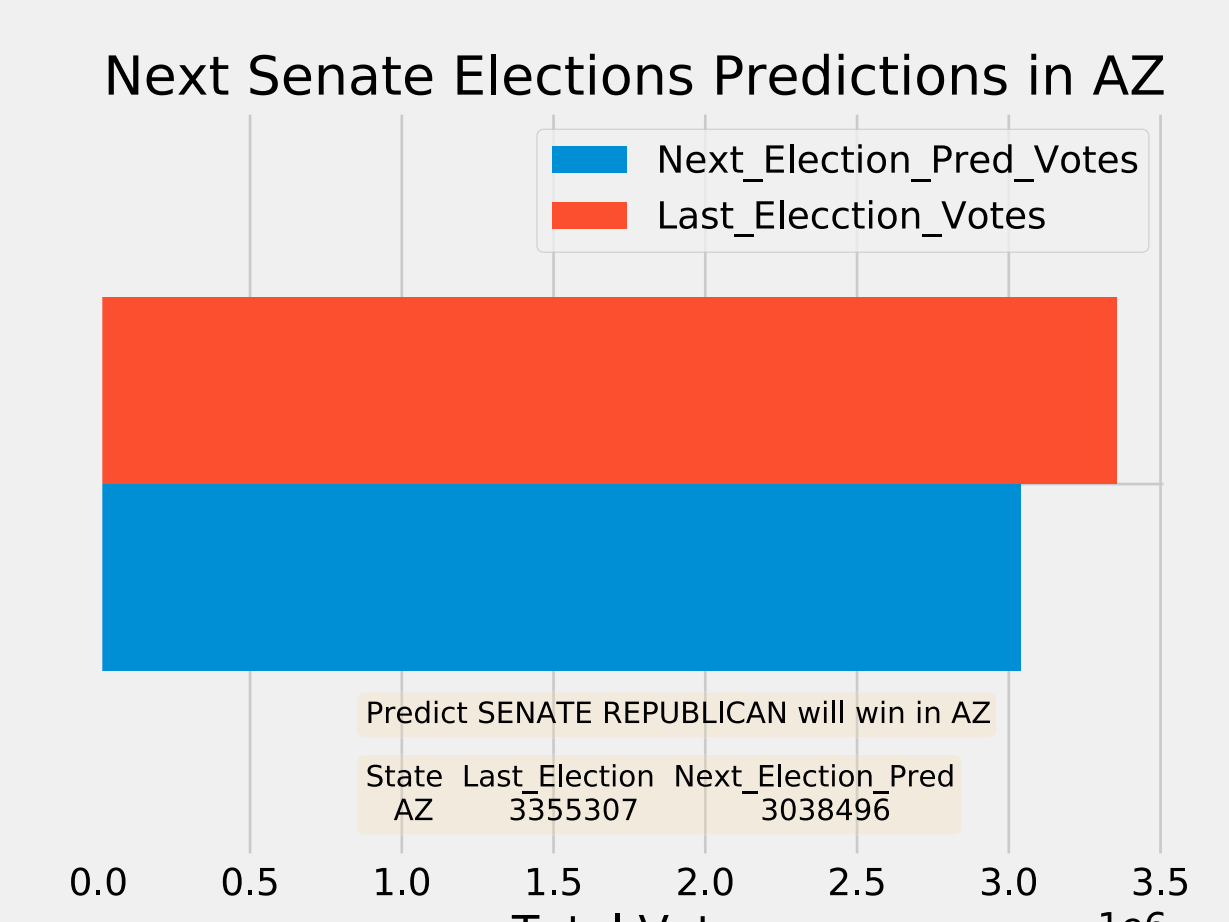

Next Senate Elections Predictions in GA
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE REPUBLICAN will win in GA
State Last_Election Next_Election_Pred
GA 4483294 4482729
0
1
2
3
4
1e6

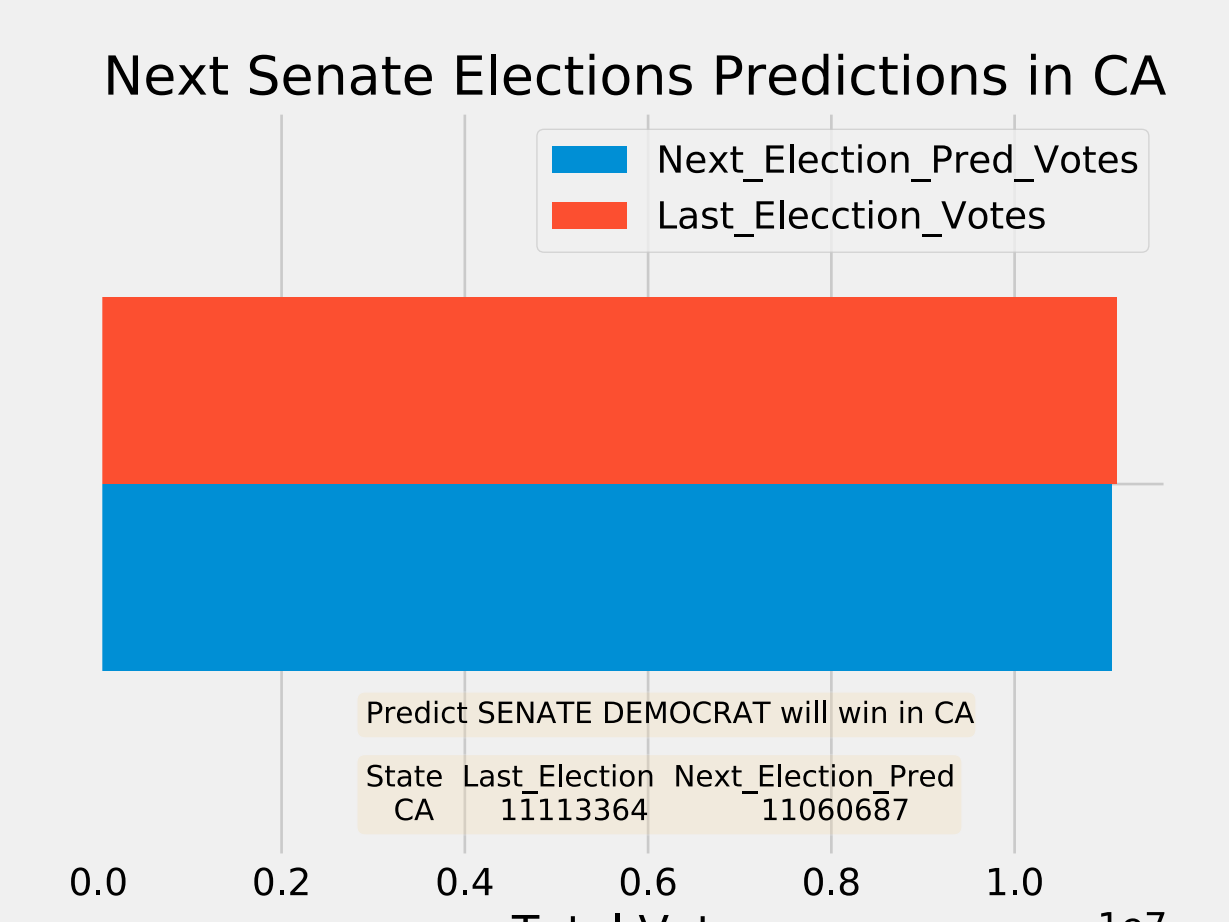

Next Senate Elections Predictions in SC
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE REPUBLICAN will win in SC
State Last_Election Next_Election_Pred
SC 2515104 2306152
0.0
0.5
1.0
1.5
2.0
2.5

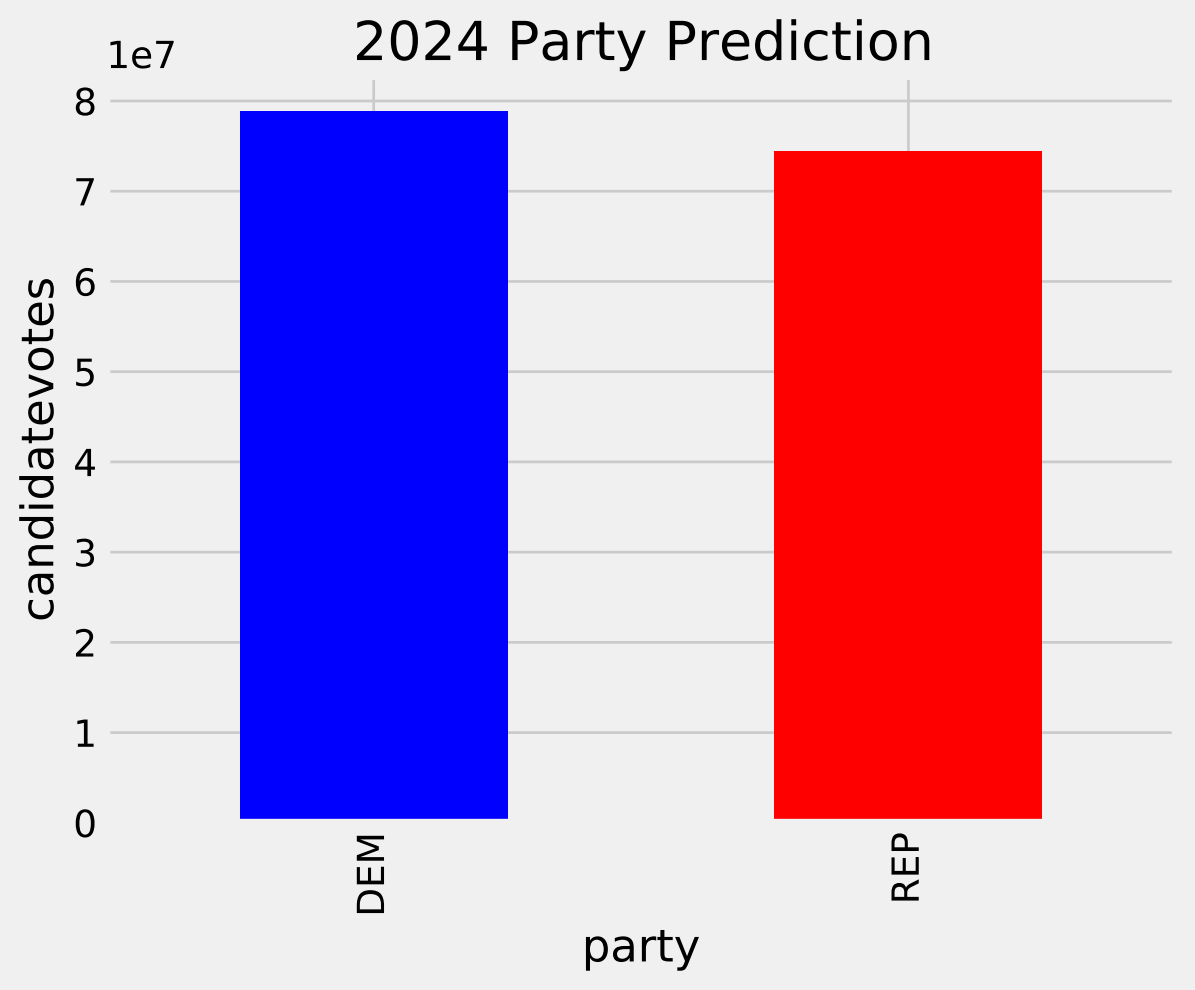

Next Senate Elections Predictions in OH
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE DEMOCRAT will win in OH
State Last_Election Next_Election_Pred
OH 4410898 4480159
0
1
2
3
4
1e6

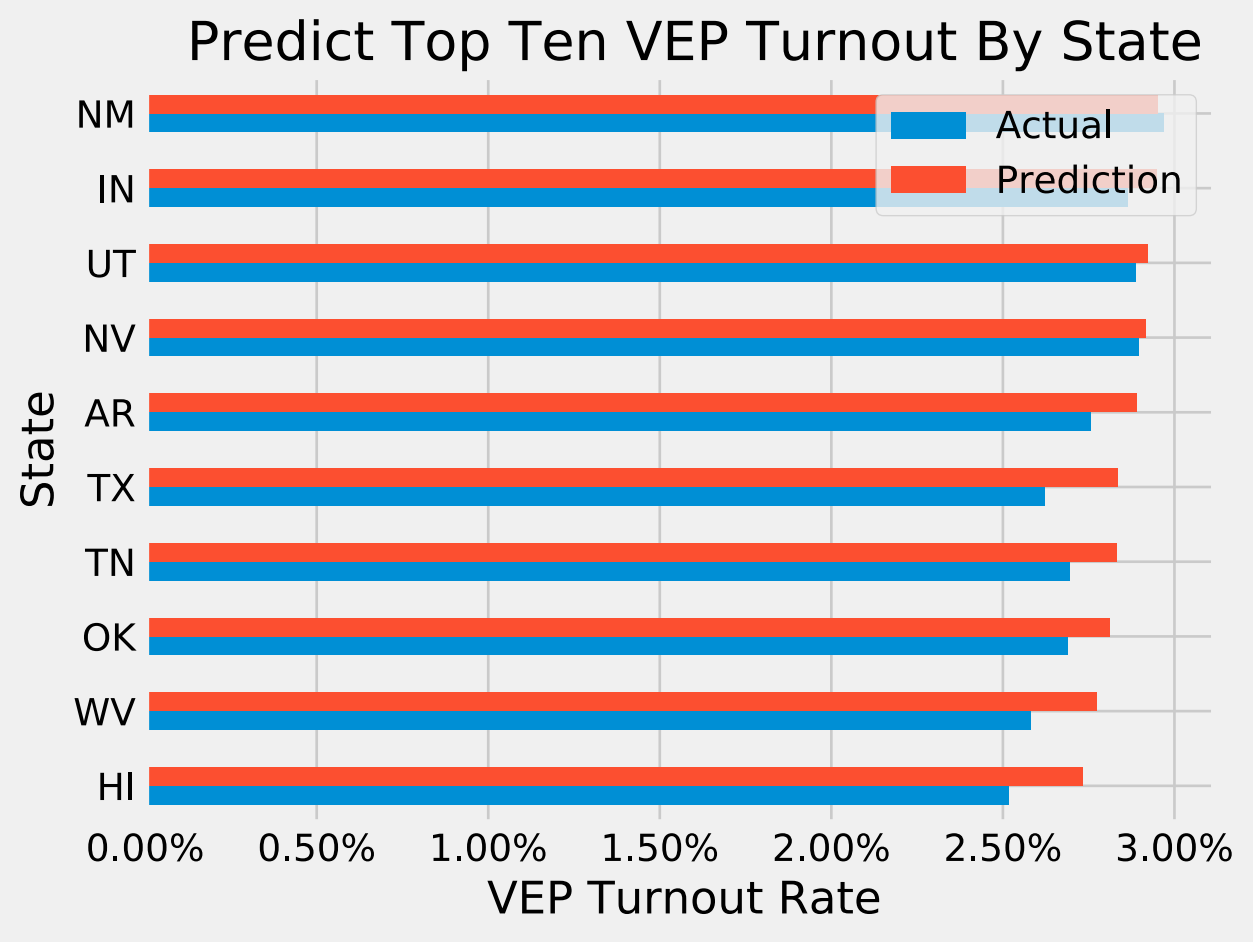

Next Senate Elections Predictions in CO
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE DEMOCRAT will win in CO
State Last_Election Next_Election_Pred
CO 3235790 2981586
0.0
0.5
1.0
1.5
2.0
2.5
3.0
1e6

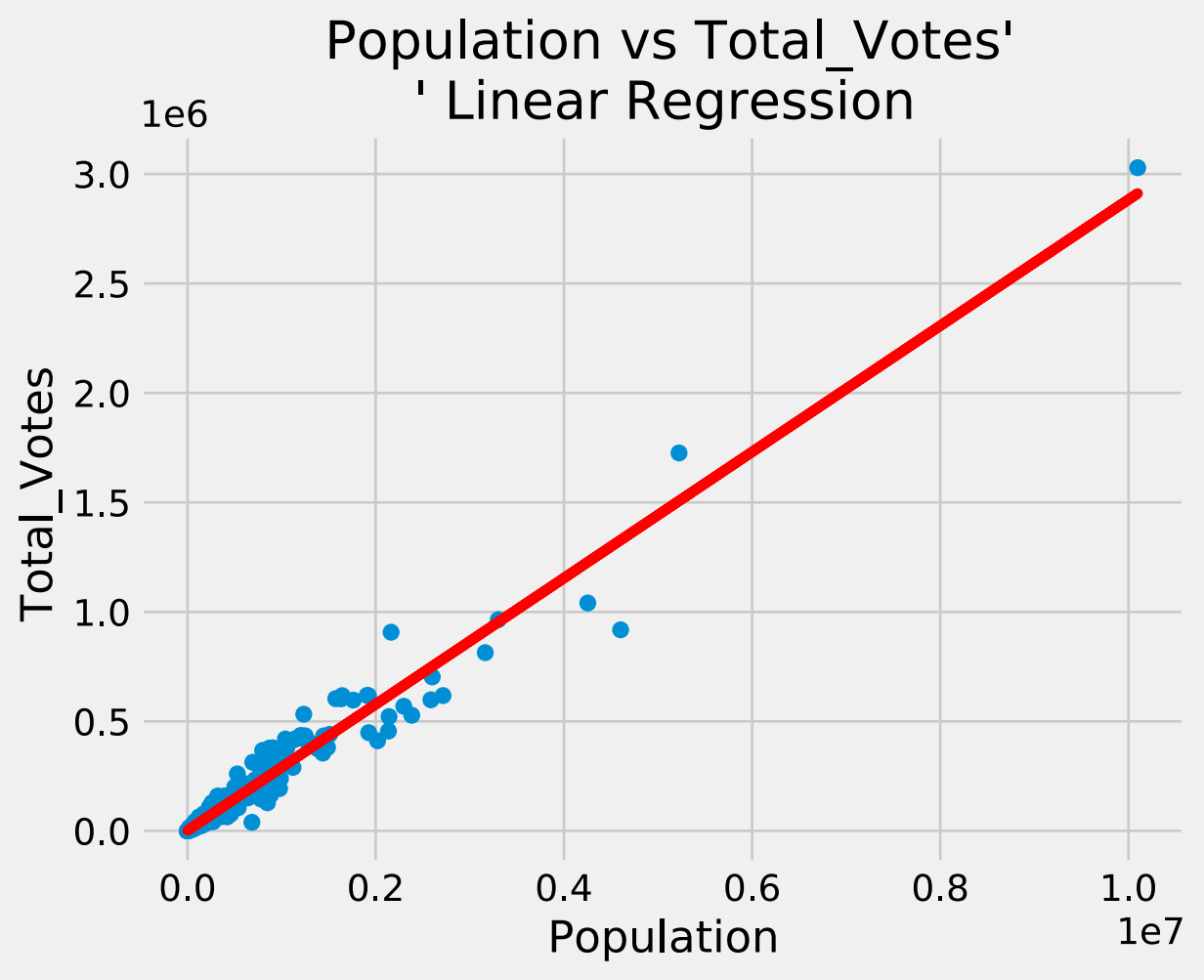

Next Senate Elections Predictions in PA
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE REPUBLICAN will win in PA
State Last_Election Next_Election_Pred
PA 4994643 5035268
0
1
2
3
4
5

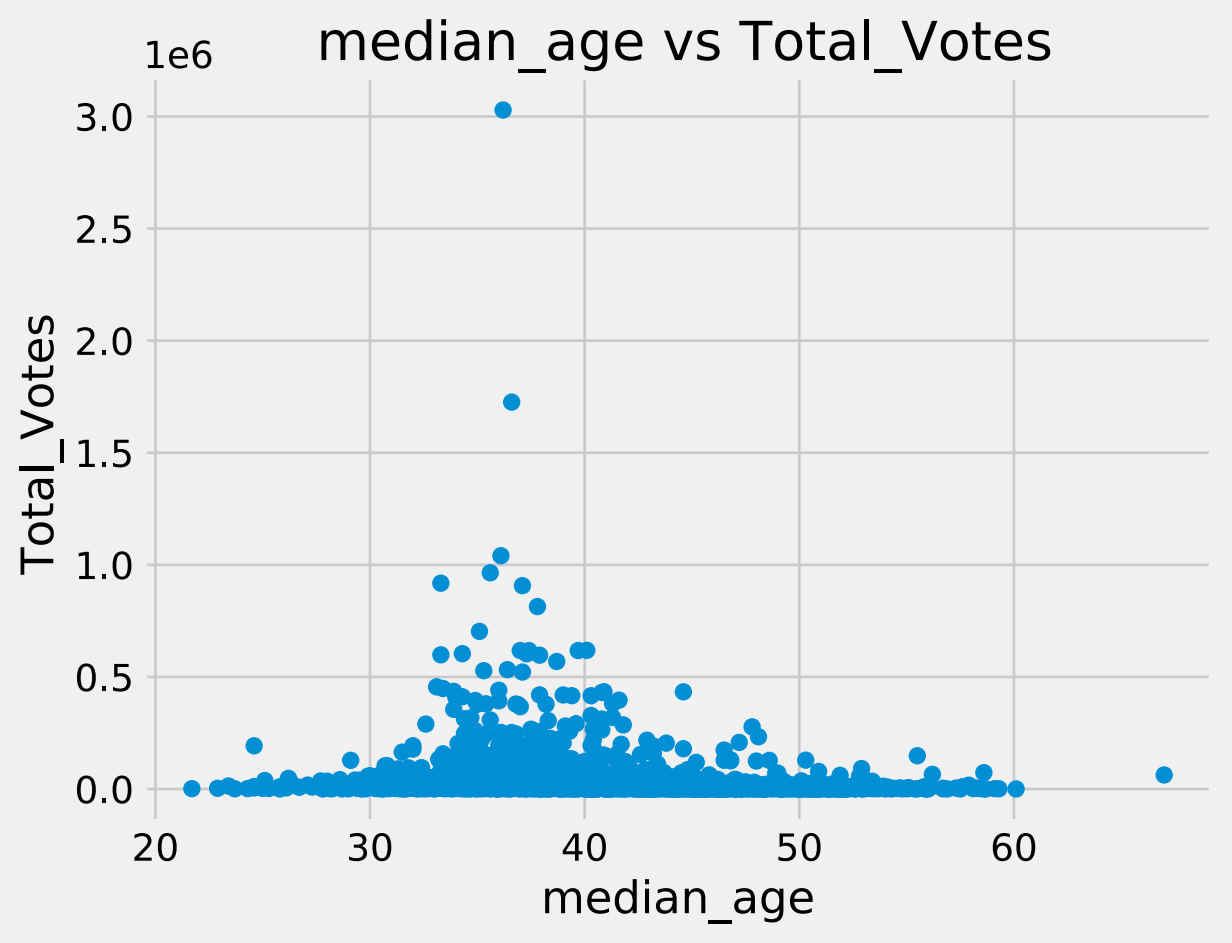

Next Senate Elections Predictions in AZ
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE REPUBLICAN will win in AZ
State Last_Election Next_Election_Pred
AZ 3355307 3038496
0.0
0.5
1.0
1.5
2.0
2.5
3.0
3.5


Next Senate Elections Predictions in CA
Next_Election_Pred_Votes
Last_Elecction_Votes
Predict SENATE DEMOCRAT will win in CA
State Last_Election Next_Election_Pred
CA 11113364 11060687
0.0
0.2
0.4
0.6
0.8
1.0

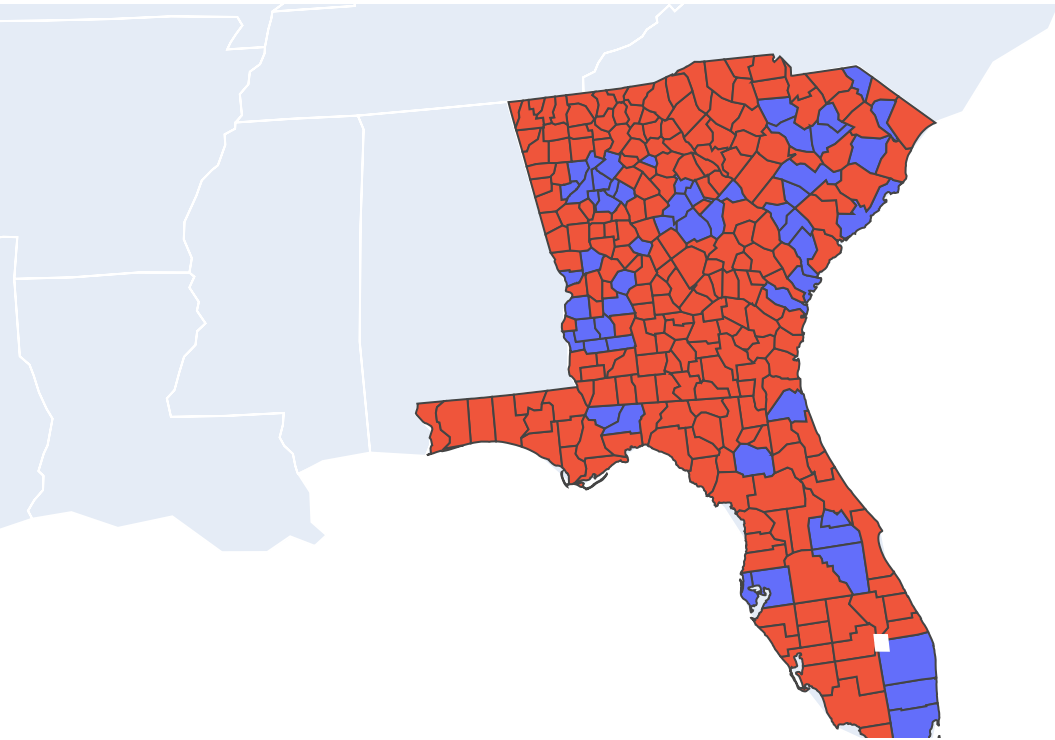

2024 Party Prediction
1e7
8
7
6
5
4
3
2
1
0
candidatevotes
DEM
REP
party


Predict Top Ten VEP Turnout By State
Actual
Prediction
NM
IN
UT
NV
AR
TX
TN
OK
WV
HI
State
0.00%
0.50%
1.00%
1.50%
2.00%
2.50%
3.00%
VEP Turnout Rate

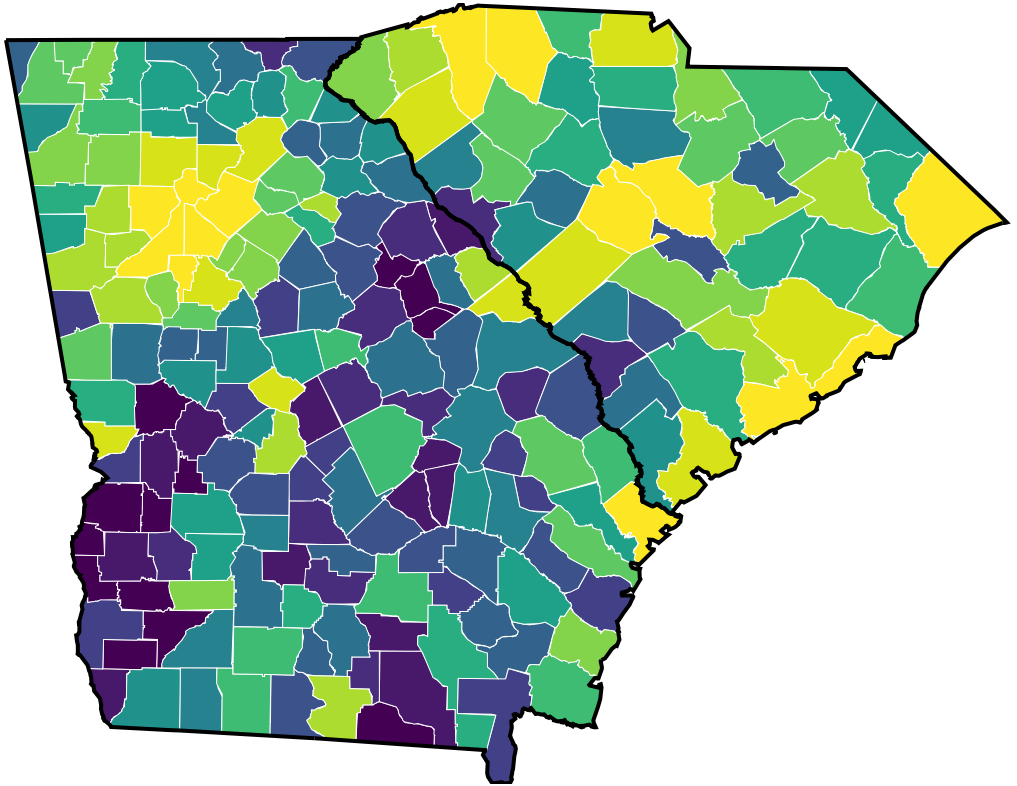

Population vs Total_Votes'
' Linear Regression
1e6
3.0
2.5
2.0
1.5
1.0
0.5
0.0
Total_Votes
0.0
0.2
0.4
0.6
0.8
1.0
1e7
Population


median_age vs Total_Votes
1e6
3.0
2.5
2.0
1.5
1.0
0.5
0.0
Total_Votes
20
30
40
50
60
median_age

Georgia vs South Carolina & Florida' ' 2020 swing states total_votes

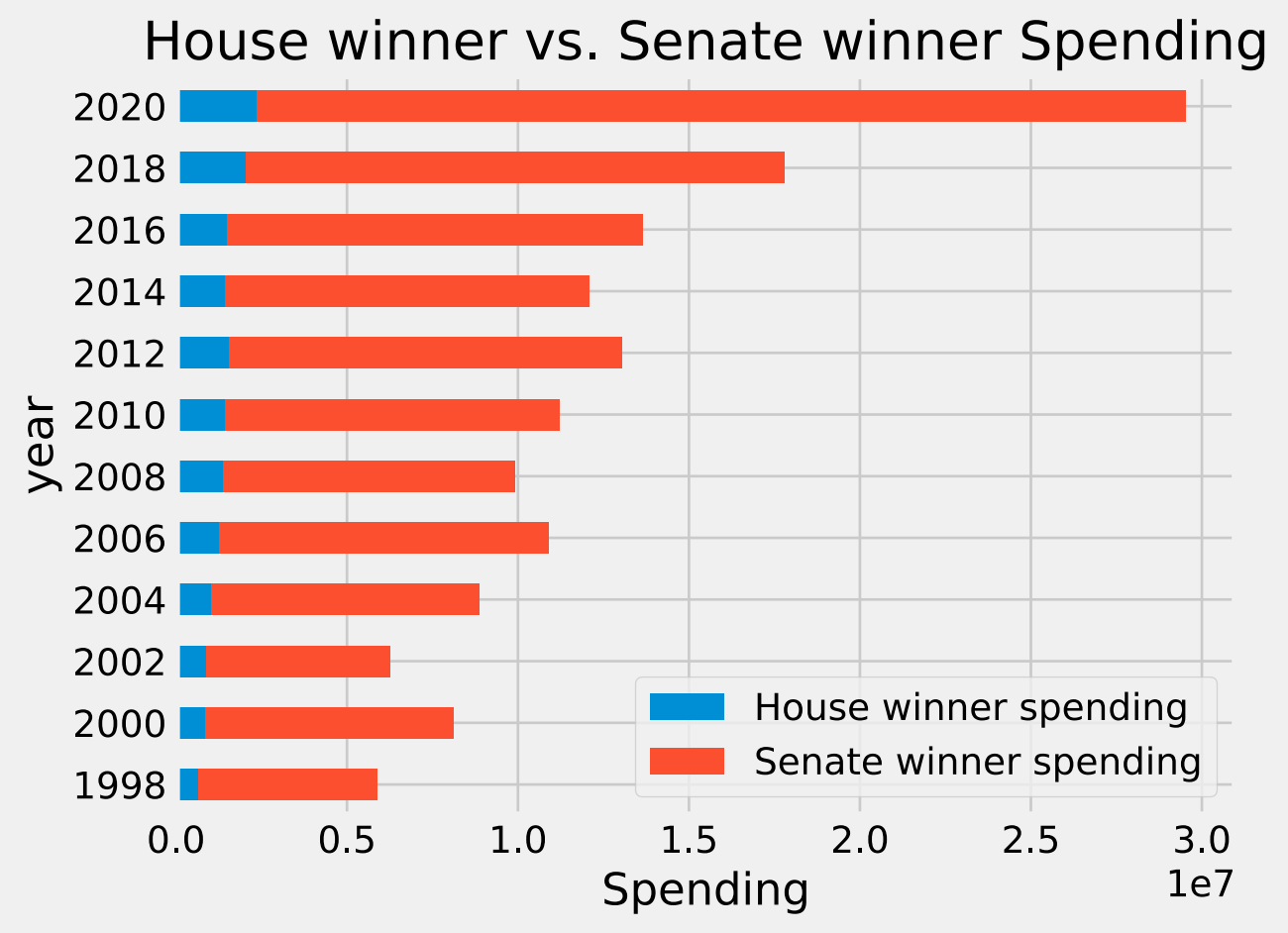


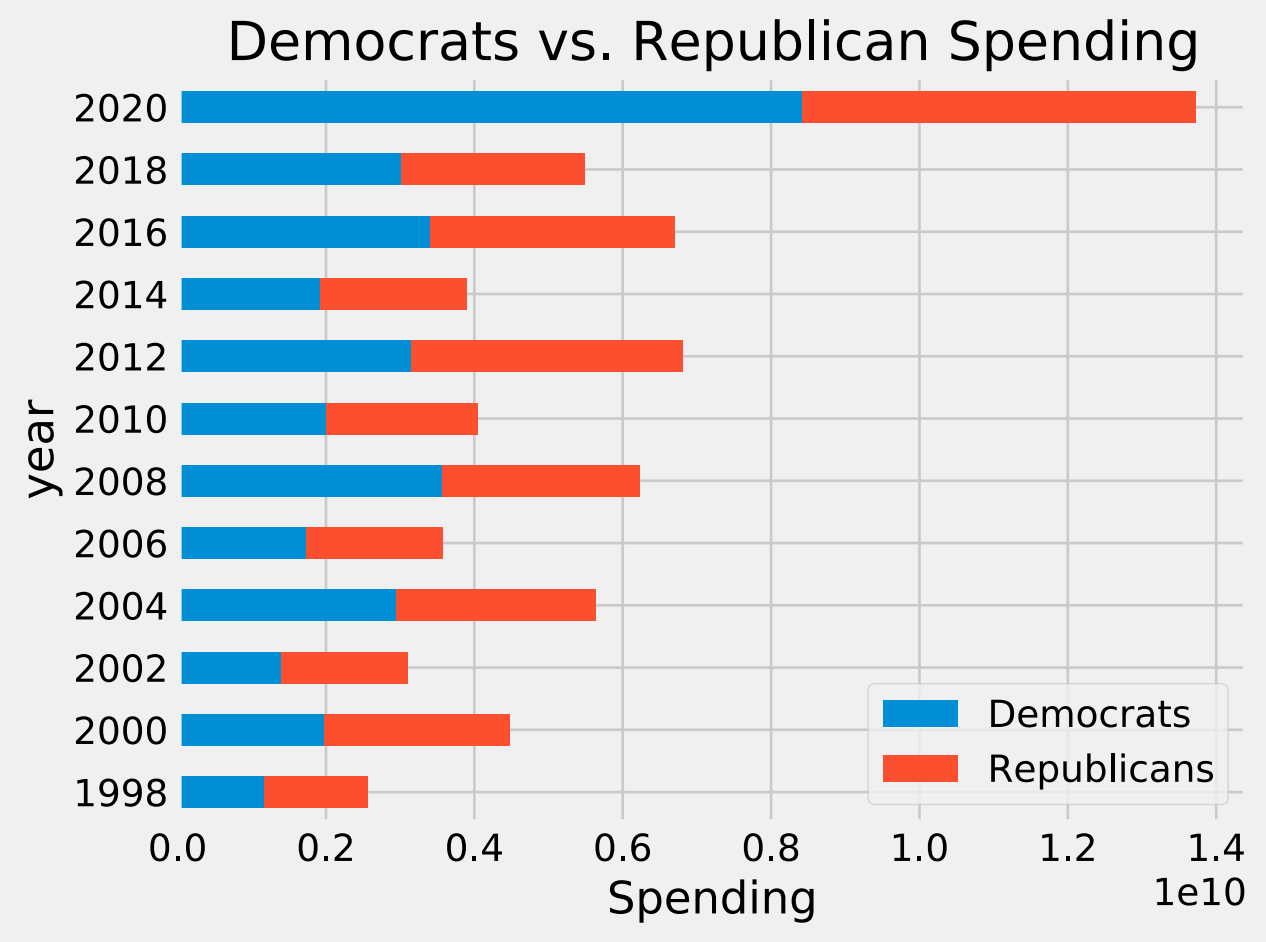
Georgia vs South Carolina Population

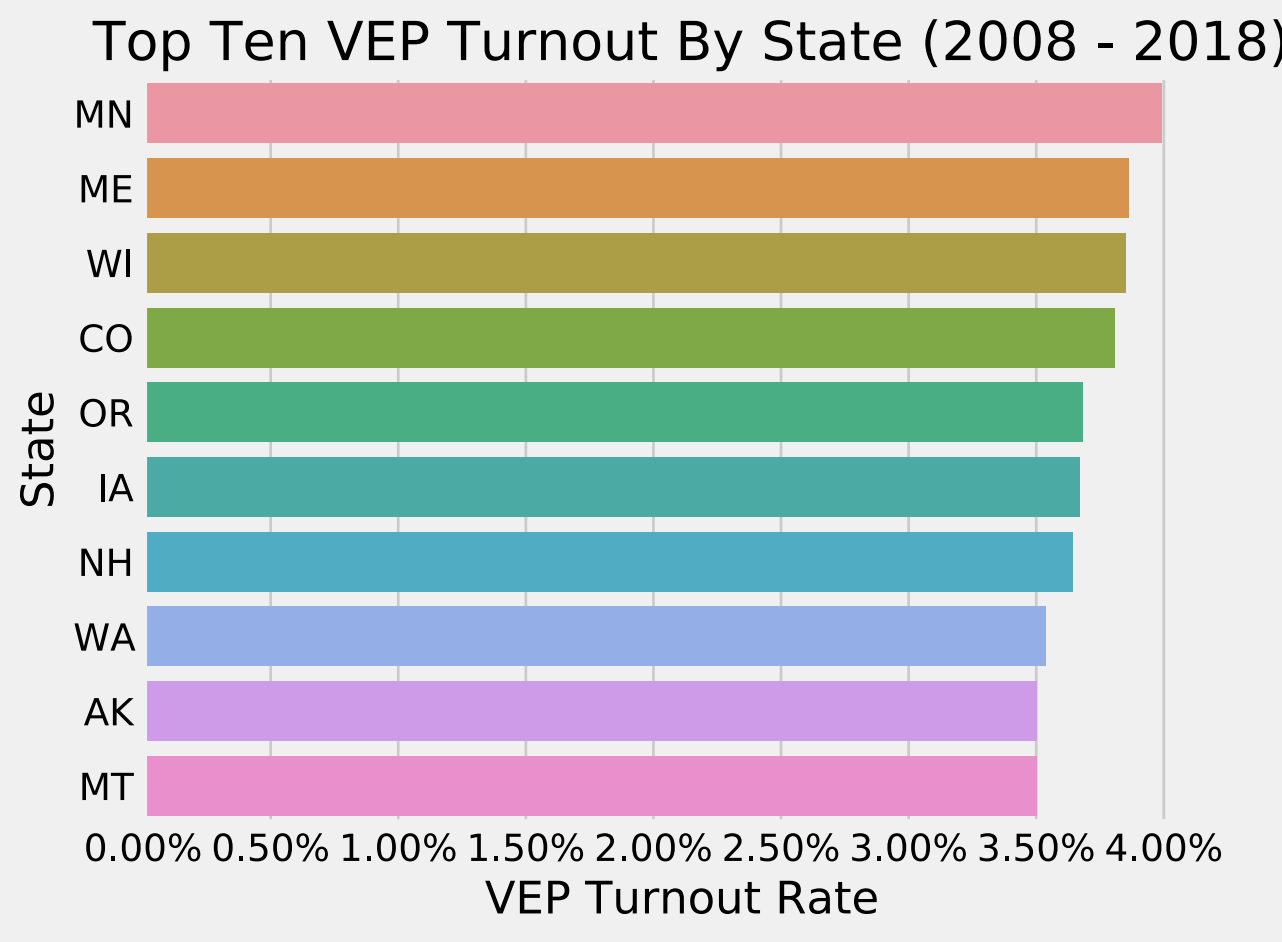


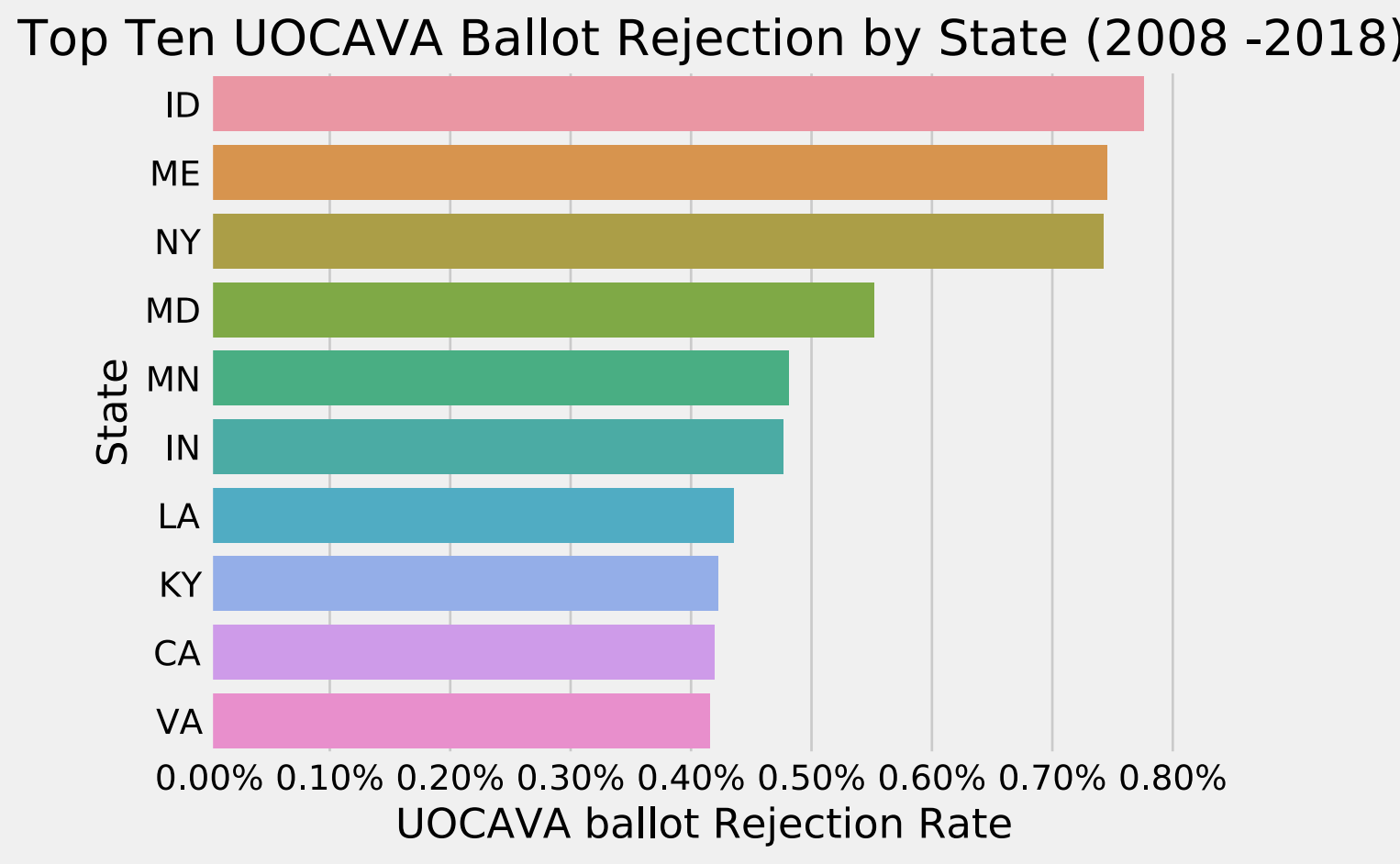
Population per county

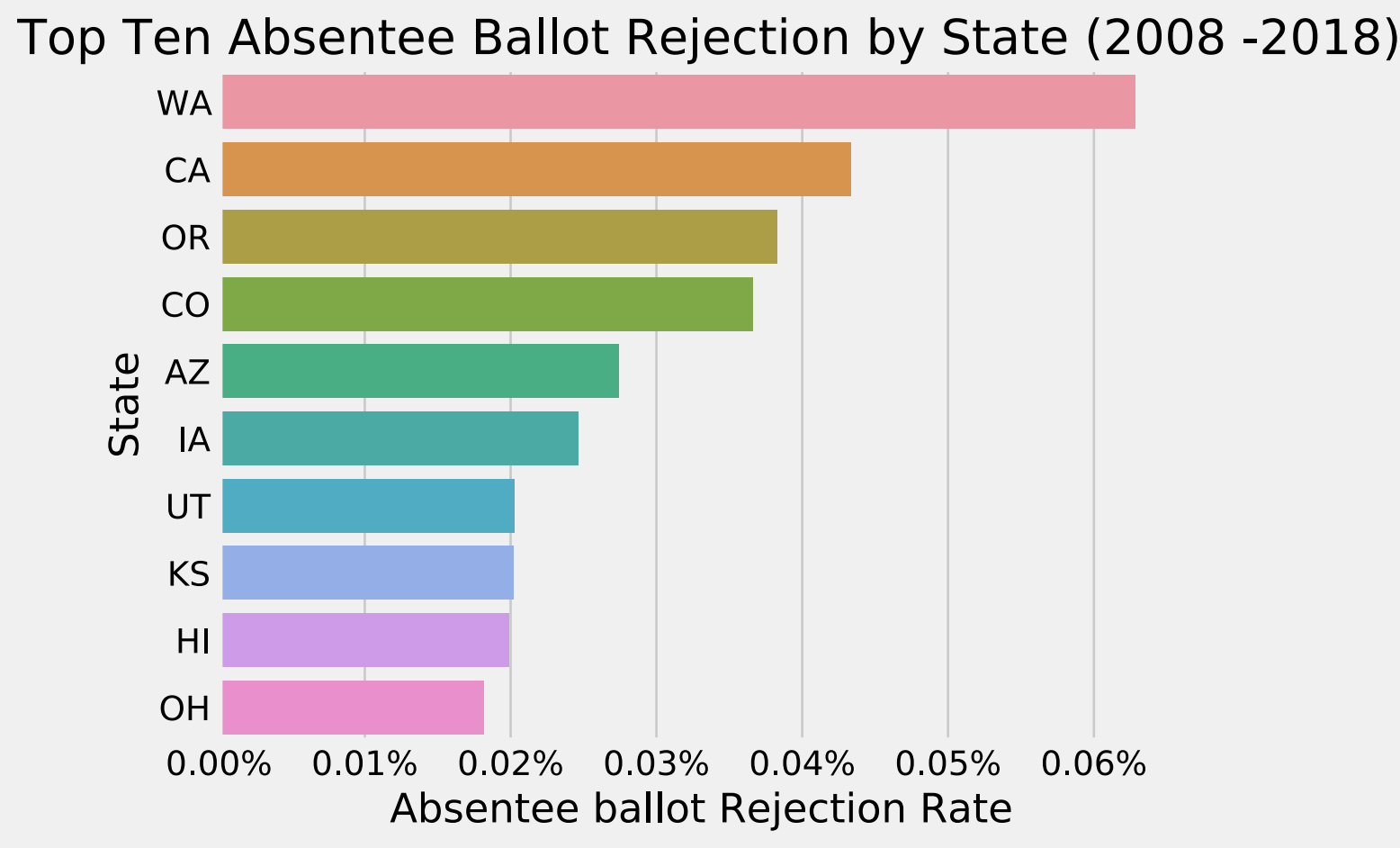
977773
842046
707442
707089
467605
393830
365162
288745
282285
276434
270406
265888
234635
221315
209053
202587
198413
189781
189355
187928

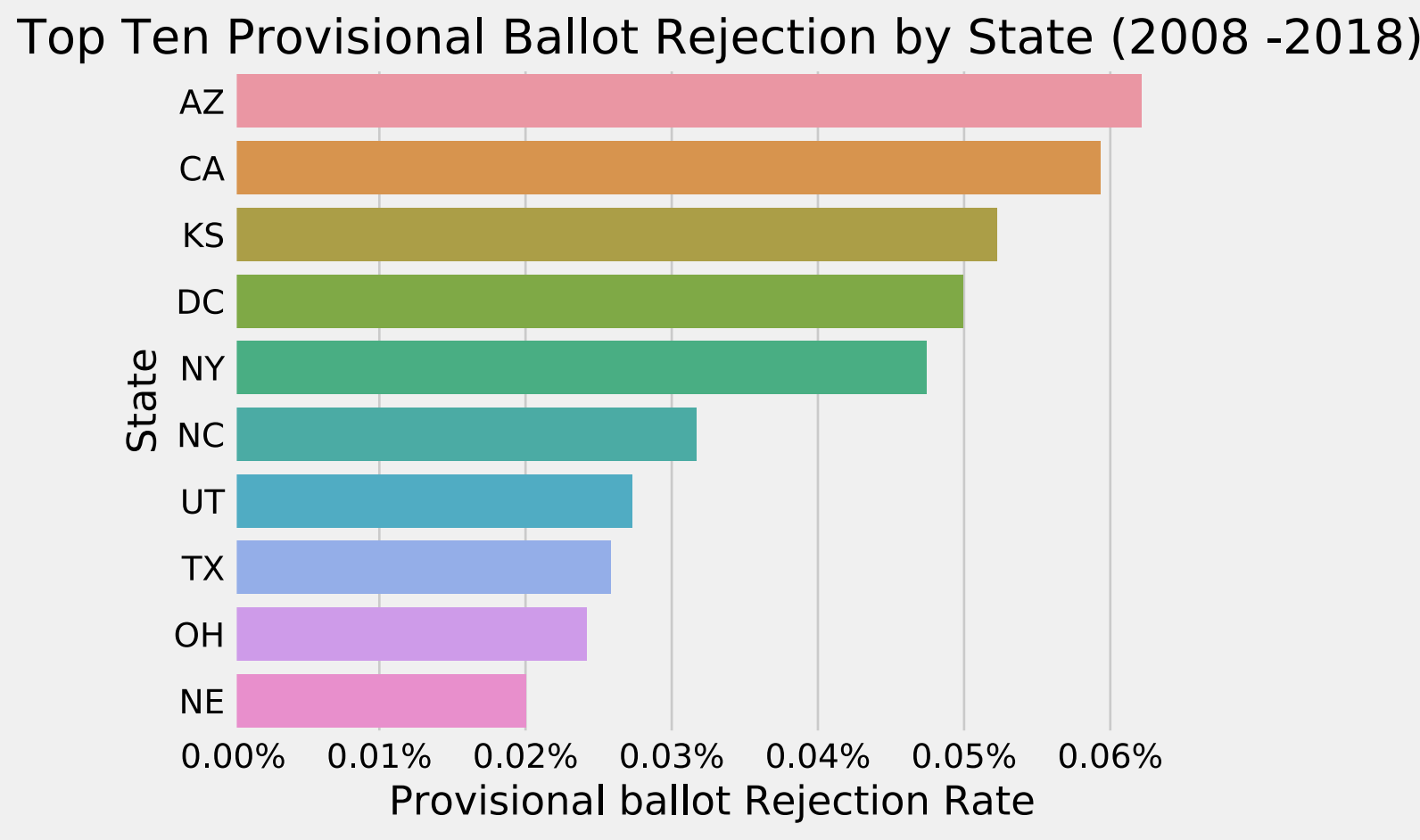

House winner vs. Senate winner Spending
year
2020
2018
2016
2014
2012
2010
2008
2006
2004
2002
2000
1998
House winner spending
Senate winner spending
0.0
0.5
1.0
1.5
2.0
2.5
3.0
Spending
1e7

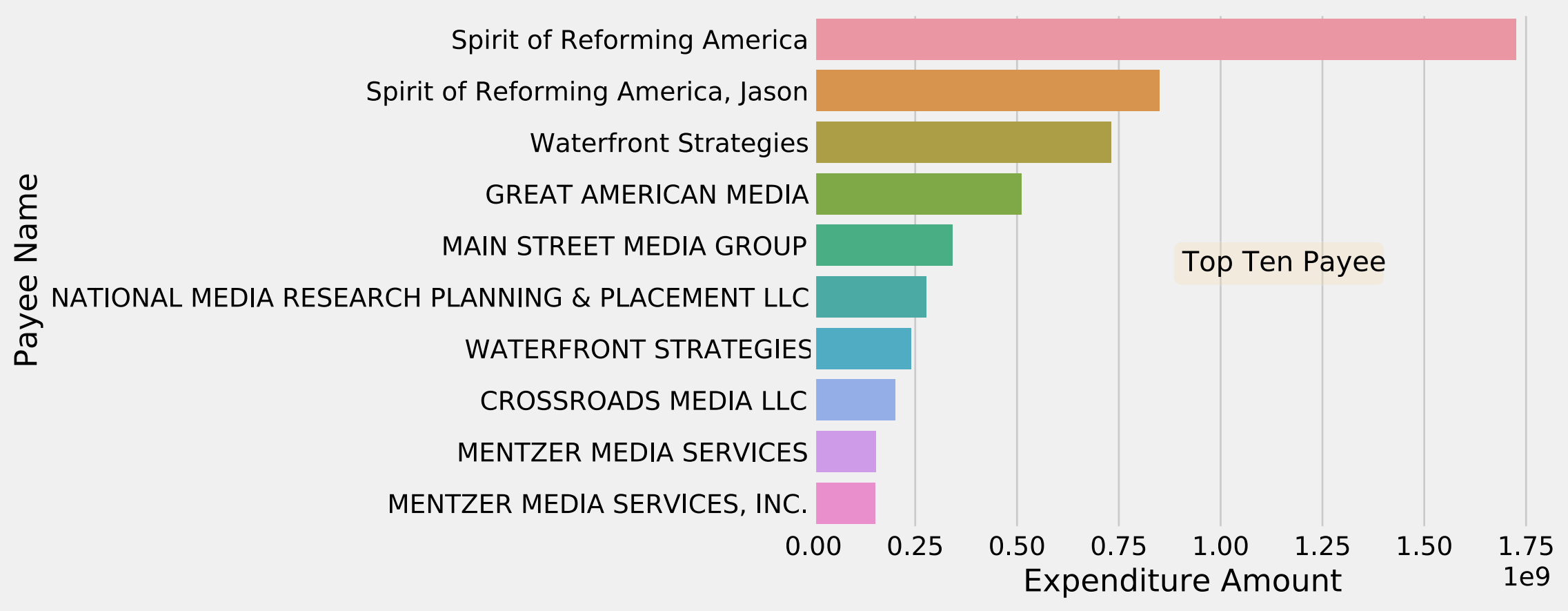

Democrats vs. Republican Spending
year
2020
2018
2016
2014
2012
2010
2008
2006
2004
2002
2000
1998
0.0
0.2
0.4
0.6
0.8
1.0
1.2
1.4
1e10
Spending
Democrats
Republicans

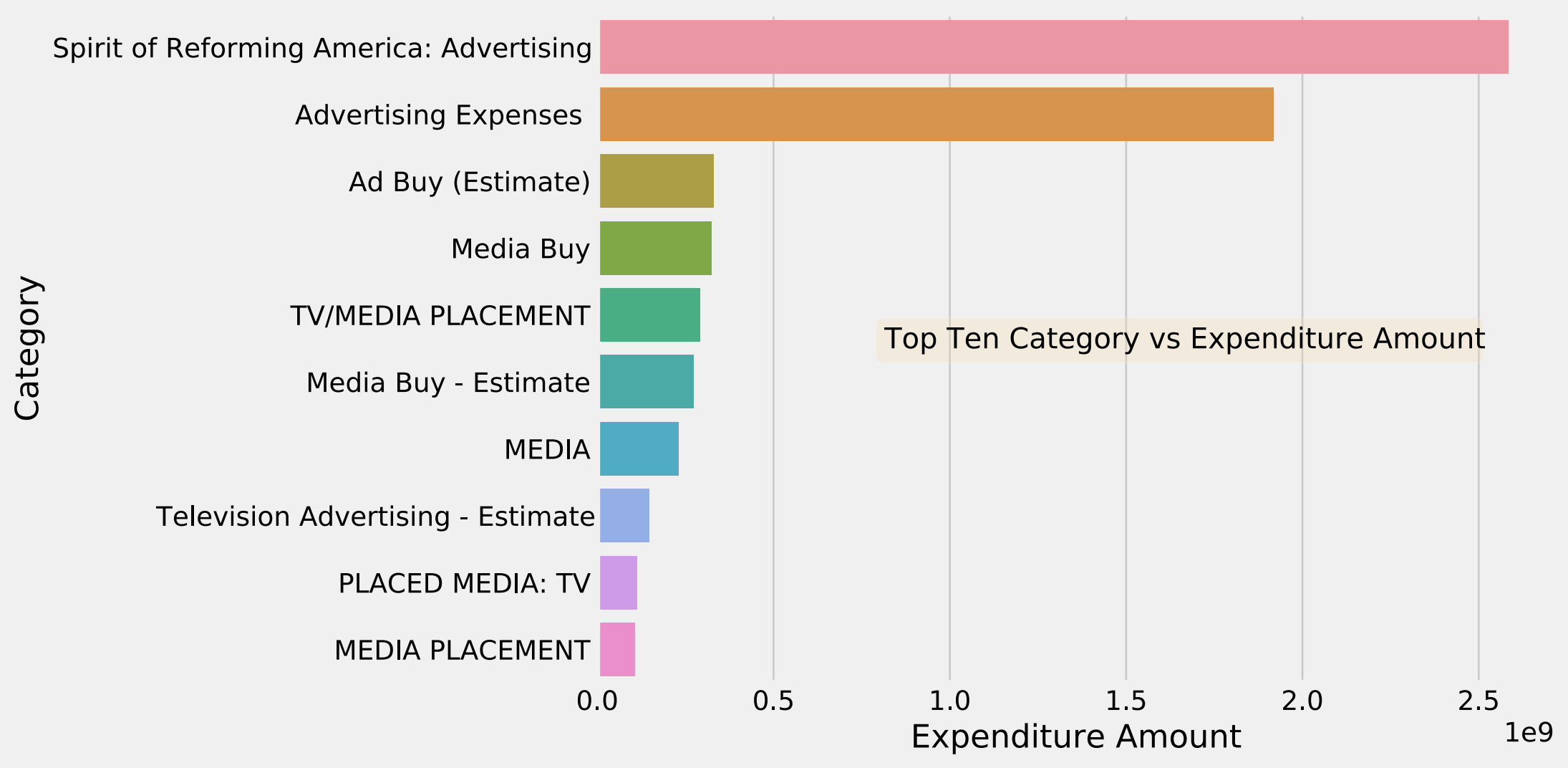

Top Ten VEP Turnout By State (2008 - 2018)
MN
ME
WI
CO
OR
IA
NH
WA
AK
MT
State
0.00%
0.50%
1.00%
1.50%
2.00%
2.50%
3.00%
3.50%
4.00%
VEP Turnout Rate

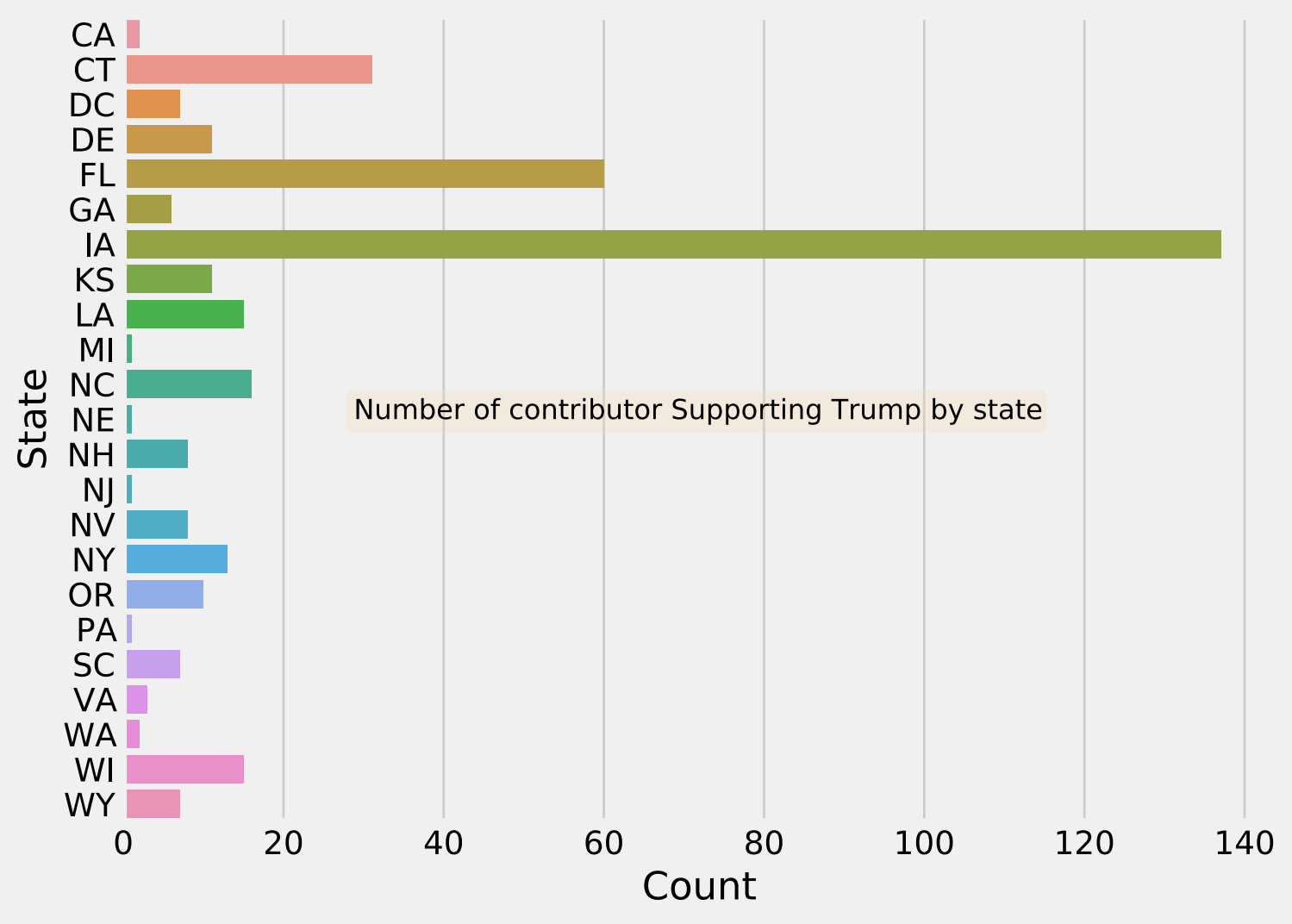

Top Ten UOCAVA Ballot Rejection by State (2008 -2018)
State
ID
ME
NY
MD
MN
IN
LA
KY
CA
VA
0.00% 0.10% 0.20% 0.30% 0.40% 0.50% 0.60% 0.70% 0.80%
UOCAVA ballot Rejection Rate

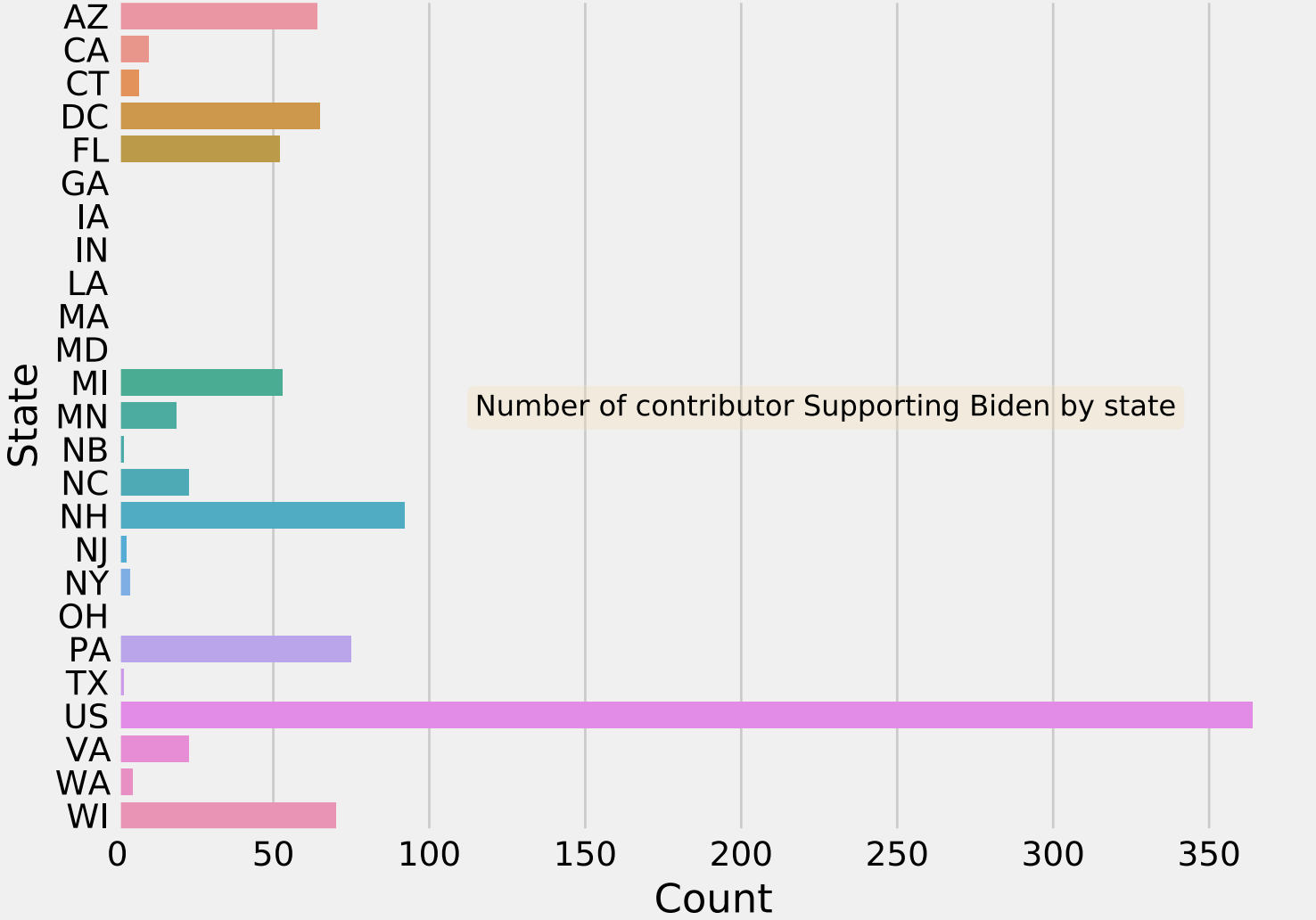

Top Ten Absentee Ballot Rejection by State (2008 -2018)
State
WA
CA
OR
CO
AZ
IA
UT
KS
HI
OH
0.00%
0.01%
0.02%
0.03%
0.04%
0.05%
0.06%
Absentee ballot Rejection Rate

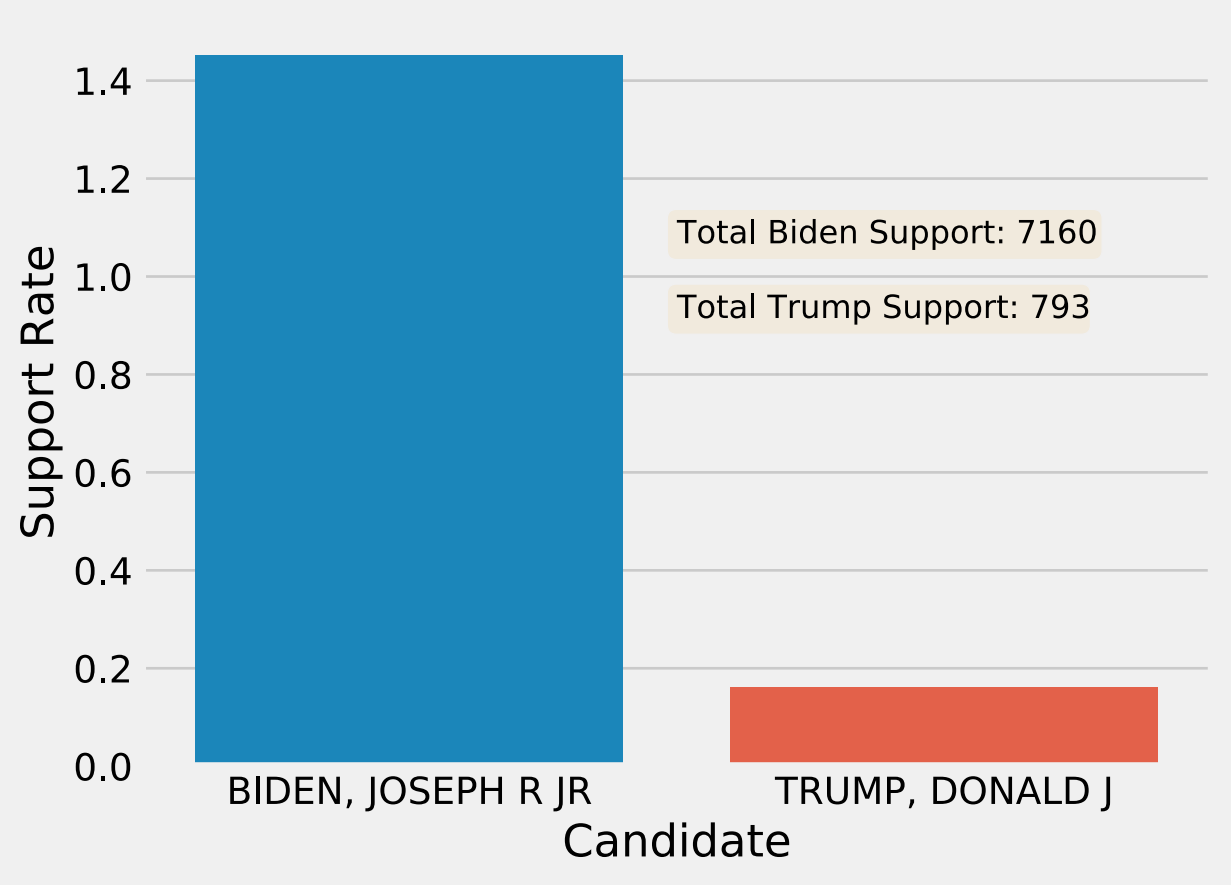

Top Ten Provisional Ballot Rejection by State (2008 -2018)
AZ
CA
KS
DC
NY
NC
UT
TX
OH
NE
State
0.00%
0.01%
0.02%
0.03%
0.04%
0.05%
0.06%
Provisional ballot Rejection Rate


Spirit of Reforming America
Spirit of Reforming America, Jason
Waterfront Strategies
GREAT AMERICAN MEDIA
MAIN STREET MEDIA GROUP
NATIONAL MEDIA RESEARCH PLANNING & PLACEMENT LLC
WATERFRONT STRATEGIES
CROSSROADS MEDIA LLC
MENTZER MEDIA SERVICES
MENTZER MEDIA SERVICES, INC.
Payee Name
Top Ten Payee
0.00
0.25
0.50
0.75
1.00
1.25
1.50
1.75
1e9
Expenditure Amount


Spirit of Reforming America: Advertising
Advertising Expenses
Ad Buy (Estimate)
Media Buy
TV/MEDIA PLACEMENT
Media Buy - Estimate
MEDIA
Television Advertising - Estimate
PLACED MEDIA: TV
MEDIA PLACEMENT
Category
Top Ten Category vs Expenditure Amount
0.0
0.5
1.0
1.5
2.0
2.5
1e9
Expenditure Amount


CA
CT
DC
DE
FL
GA
IA
KS
LA
MI
NC
NE
NH
NJ
NV
NY
OR
PA
SC
VA
WA
WI
WY
State
Number of contributor Supporting Trump by state
0
20
40
60
80
100
120
140
Count


Number of contributor Supporting Biden by state
State
AZ
CA
CT
DC
FL
GA
IA
IN
LA
MA
MD
MI
MN
NB
NC
NH
NJ
NY
OH
PA
TX
US
VA
WA
WI
0
50
100
150
200
250
300
350
Count


1.4
1.2
1.0
0.8
0.6
0.4
0.2
0.0
Support Rate
Total Biden Support: 7160
Total Trump Support: 793
BIDEN, JOSEPH R JR
TRUMP, DONALD J
Candidate

y = df['totalvotes'].values

```
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
dense (Dense)                (None, 5344)              14284512
_________________________________________________________________
dense_1 (Dense)              (None, 1)                 5345
=================================================================
Total params: 14,289,857
Trainable params: 14,289,857
Non-trainable params: 0
_________________________________________________________________
None
Train on 2721 samples
Epoch 1/100
2721/2721 [==============================] - 10s 4ms/sample - loss: 9381745978502.3516 - mse: 9381748932608.0000
Epoch 2/100
2721/2721 [==============================] - 9s 3ms/sample - loss: 9378059751868.0723 - mse: 9378063187968.0000
Epoch 3/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 9369342295058.4414 - mse: 9369342181376.0000
Epoch 4/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 9355608455385.5195 - mse: 9355607932928.0000
Epoch 5/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 9338035377987.0820 - mse: 9338035896320.0000
Epoch 95/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 3701527765344.2471 - mse: 3701528330240.0000
Epoch 96/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 3675185211342.7236 - mse: 3675185479680.0000
Epoch 97/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 3649374098189.2661 - mse: 3649372946432.0000
Epoch 98/100
2721/2721 [==============================] - 9s 3ms/sample - loss: 3623577882863.3477 - mse: 3623577976832.0000
Epoch 99/100
2721/2721 [==============================] - 9s 3ms/sample - loss: 3599386996906.1021 - mse: 3599386542080.0000
Epoch 100/100
2721/2721 [==============================] - 8s 3ms/sample - loss: 3575027308213.6743 - mse: 3575027597312.0000
r2_score of y_train: 0.1872076104671837
r2_score of y_test: 0.18492524187504644
neural network model with 1 hidden layer is done!
```

Training the model with 1 hidden layer
Total Time: 15 mins

y = df['totalvotes'].values

```
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
dense_2 (Dense)              (None, 5344)              14284512
_________________________________________________________________
dense_3 (Dense)              (None, 5344)              28563680
_________________________________________________________________
dense_4 (Dense)              (None, 1)                 5345
=================================================================
Total params: 42,853,537
Trainable params: 42,853,537
Non-trainable params: 0
_________________________________________________________________
None
Train on 2721 samples
Epoch 1/100
2721/2721 [==============================] - 25s 9ms/sample - loss: 9183323850049.3867 - mse: 9183323750400.0000
Epoch 2/100
2721/2721 [==============================] - 25s 9ms/sample - loss: 6133273795376.6406 - mse: 6133274640384.0000
Epoch 3/100
2721/2721 [==============================] - 25s 9ms/sample - loss: 3453539358362.4844 - mse: 3453539581952.0000
Epoch 4/100
2721/2721 [==============================] - 25s 9ms/sample - loss: 2582551219638.4268 - mse: 2582551199744.0000
Epoch 5/100
Epoch 95/100
2721/2721 [==============================] - 29s 11ms/sample - loss: 8880102119.6090 - mse: 8880103424.0000
Epoch 96/100
2721/2721 [==============================] - 29s 11ms/sample - loss: 8369913704.3381 - mse: 8369913856.0000
Epoch 97/100
2721/2721 [==============================] - 29s 11ms/sample - loss: 8420966648.1793 - mse: 8420965376.0000
Epoch 98/100
2721/2721 [==============================] - 29s 11ms/sample - loss: 7633665933.8773 - mse: 7633666560.0000
Epoch 99/100
2721/2721 [==============================] - 30s 11ms/sample - loss: 7245674765.5362 - mse: 7245673472.0000
Epoch 100/100
2721/2721 [==============================] - 30s 11ms/sample - loss: 7042875558.2683 - mse: 7042875392.0000
r2_score of y_train: 0.998542961527269
r2_score of y_test: 0.986836137393499
neural network model with 2 hidden layers is done!
```

Training the model with 2 hidden layers
Total Time: 45 mins

```
Model: "sequential"
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
dense (Dense)                (None, 9665)              93421890
_________________________________________________________________
dense_1 (Dense)              (None, 9665)              93421890
_________________________________________________________________
dense_2 (Dense)              (None, 1)                 9666
=================================================================
Total params: 186,853,446
Trainable params: 186,853,446
Non-trainable params: 0
_________________________________________________________________
```

[18]

```
# train the model for 100 epochs
model = nn.fit(X_train_scaled, y_train, epochs=50)
```

```
Epoch 36/50
30032/30032 [==============================] - 2663s 89ms/sample - loss: 69013763941.7944 - mse: 69013782528.0000
Epoch 37/50
30032/30032 [==============================] - 2996s 100ms/sample - loss: 27105227519.1055 - mse: 27105234944.0000
Epoch 38/50
30032/30032 [==============================] - 4423s 147ms/sample - loss: 70524384642.1082 - mse: 70524387328.0000
Epoch 39/50
30032/30032 [==============================] - 2772s 92ms/sample - loss: 34133394901.8860 - mse: 34133399552.0000
Epoch 40/50
30032/30032 [==============================] - 2771s 92ms/sample - loss: 40945962369.1734 - mse: 40945971200.0000
Epoch 41/50
30032/30032 [==============================] - 2766s 92ms/sample - loss: 31298373064.5173 - mse: 31298387968.0000
Epoch 42/50
30032/30032 [==============================] - 2780s 93ms/sample - loss: 27320213137.5509 - mse: 27320238080.0000
Epoch 43/50
30032/30032 [==============================] - 2759s 92ms/sample - loss: 46198547755.1031 - mse: 46198554624.0000
Epoch 44/50
30032/30032 [==============================] - 2761s 92ms/sample - loss: 48161741135.9622 - mse: 48161751040.0000
Epoch 45/50
30032/30032 [==============================] - 3713s 124ms/sample - loss: 38139335662.2384 - mse: 38139387904.0000
Epoch 46/50
30032/30032 [==============================] - 3621s 121ms/sample - loss: 32038363352.3756 - mse: 32038371328.0000
Epoch 47/50
30032/30032 [==============================] - 3863s 129ms/sample - loss: 30613974870.4190 - mse: 30613972992.0000
Epoch 48/50
30032/30032 [==============================] - 2746s 91ms/sample - loss: 57388354544.3636 - mse: 57388351488.0000
Epoch 49/50
30032/30032 [==============================] - 2637s 88ms/sample - loss: 22141545709.2789 - mse: 22141538304.0000
Epoch 50/50
30032/30032 [==============================] - 2627s 87ms/sample - loss: 27240833076.6209 - mse: 27240806400.0000
```

[19]

```
# predict values for the train and test sets
y_train_pred = nn.predict(X_train_scaled)
y_test_pred = nn.predict(X_test_scaled)
```

[20]

```
# score the training predictions with r2_score()
r2_score(y_train, y_train_pred)
```

```
0.9991990876693556
```

y = df['expenditure_amount'].values
Total time: 2 days

[21]

```
# score the test predictions with r2_score()
r2_score(y_test, y_test_pred)
```

```
0.8145582287092288
```