

Sanders Data Targeting Strategy

April 21, 2019

1 Bernie Sanders Campaign Plan

1.0.1 Field Targeting: The Path to 270

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1.0.3 Date: April 25, 2019

Description: This notebook will walk through a brief clustering analysis of 2020 polling data to provide projections and clusters of states that we expect to behave similarly in the general election. State clusters are used to determine targets for advertisement and field operations. In-house mapping software is used to provide visuals of our expectations for the election. Target states are then analyzed further by breaking down data for individual counties to determine which to target.

Import software needed for this analysis

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.cluster import DBSCAN
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
import statsmodels.api as sm
import matplotlib.style
import matplotlib as mpl
import geopandas as gpd
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.colors
pd.set_option('display.max_rows', 500)
pd.set_option('display.max_columns', 500)
pd.set_option('display.width', 1000)
mpl.style.use('classic')
```

2 State Analysis: Path to 270

Load Polling Data

```
In [2]: data = pd.read_csv('polling_data.csv')
        data.head()
```

```
Out[2]:
```

	Unnamed: 0	2012_D	2012_R	2012_I	2016_D	2016_R	2016_I	2020_D	2020_R	2020_I	2024_D	2024_R	2024_I
0	AL	38.40%	60.50%	1.10%	34.40%	62.10%	3.60%	35.20%	61.80%	3.10%	38.40%	60.50%	1.10%
1	AK	41.20%	55.30%	3.50%	36.60%	51.30%	12.20%	37.50%	52.10%	10.40%	41.20%	55.30%	3.50%
2	AZ	44.60%	53.70%	1.80%	45.10%	48.70%	6.20%	49.70%	45.00%	5.30%	44.60%	53.70%	1.80%
3	AR	36.90%	60.60%	2.60%	33.70%	60.60%	5.80%	34.30%	60.60%	5.10%	36.90%	60.60%	2.60%
4	CA	60.20%	37.10%	2.60%	61.70%	31.60%	6.70%	46.40%	32.70%	20.90%	60.20%	37.10%	2.60%

Data Processing

```
In [3]: for column in ['2012_D', '2012_R', '2012_I', '2016_D', '2016_R', '2016_I', '2020_D', '2020_R', '2020_I', '2024_D', '2024_R', '2024_I']:
        data[column] = data[column].apply(lambda x: float(x.split('%')[0]))
        data.head()
```

```
Out[3]:
```

	Unnamed: 0	2012_D	2012_R	2012_I	2016_D	2016_R	2016_I	2020_D	2020_R	2020_I	2024_D	2024_R	2024_I
0	AL	38.4	60.5	1.1	34.4	62.1	3.6	35.2	61.8	3.1	38.4	60.5	1.1
1	AK	41.2	55.3	3.5	36.6	51.3	12.2	37.5	52.1	10.4	41.2	55.3	3.5
2	AZ	44.6	53.7	1.8	45.1	48.7	6.2	49.7	45.0	5.3	44.6	53.7	1.8
3	AR	36.9	60.6	2.6	33.7	60.6	5.8	34.3	60.6	5.1	36.9	60.6	2.6
4	CA	60.2	37.1	2.6	61.7	31.6	6.7	46.4	32.7	20.9	60.2	37.1	2.6

Generate 2024 Projections using quadratic regression to account for the Blue Wave

```
In [4]: # add 2024 columns for better predictions
        data['2024_D'] = [0] * len(data)
        data['2024_R'] = [0] * len(data)
        data['2024_I'] = [0] * len(data)

        for index, row in data.iterrows():

            indices = np.array([2012.0, 2016.0, 2020.0])
            test_index = np.array(2024.0)
            dems_train = np.array(row[['2012_D', '2016_D', '2020_D']]).reshape(1, -1)
            reps_train = np.array(row[['2012_R', '2016_R', '2020_R']]).reshape(1, -1)
            inds_train = np.array(row[['2012_I', '2016_I', '2020_I']]).reshape(1, -1)

            model_1 = sm.OLS(dems_train[0], indices).fit()
            data.at[index, '2024_D'] = model_1.predict(test_index)

            model_2 = sm.OLS(reps_train[0], indices).fit()
            data.at[index, '2024_R'] = model_2.predict(test_index)
```

```

model_3 = sm.OLS(inds_train[0], indices).fit()
data.at[index, '2024_I'] = model_3.predict(test_index)

data.head()

```

Out[4]:

	Unnamed: 0	2012_D	2012_R	2012_I	2016_D	2016_R	2016_I	2020_D	2020_R	2020_I
0	AL	38.4	60.5	1.1	34.4	62.1	3.6	35.2	61.8	3.1
1	AK	41.2	55.3	3.5	36.6	51.3	12.2	37.5	52.1	10.4
2	AZ	44.6	53.7	1.8	45.1	48.7	6.2	49.7	45.0	5.3
3	AR	36.9	60.6	2.6	33.7	60.6	5.8	34.3	60.6	5.1
4	CA	60.2	37.1	2.6	61.7	31.6	6.7	46.4	32.7	20.9

Elbow Plot for Rigorous Optimal Number of Clusters

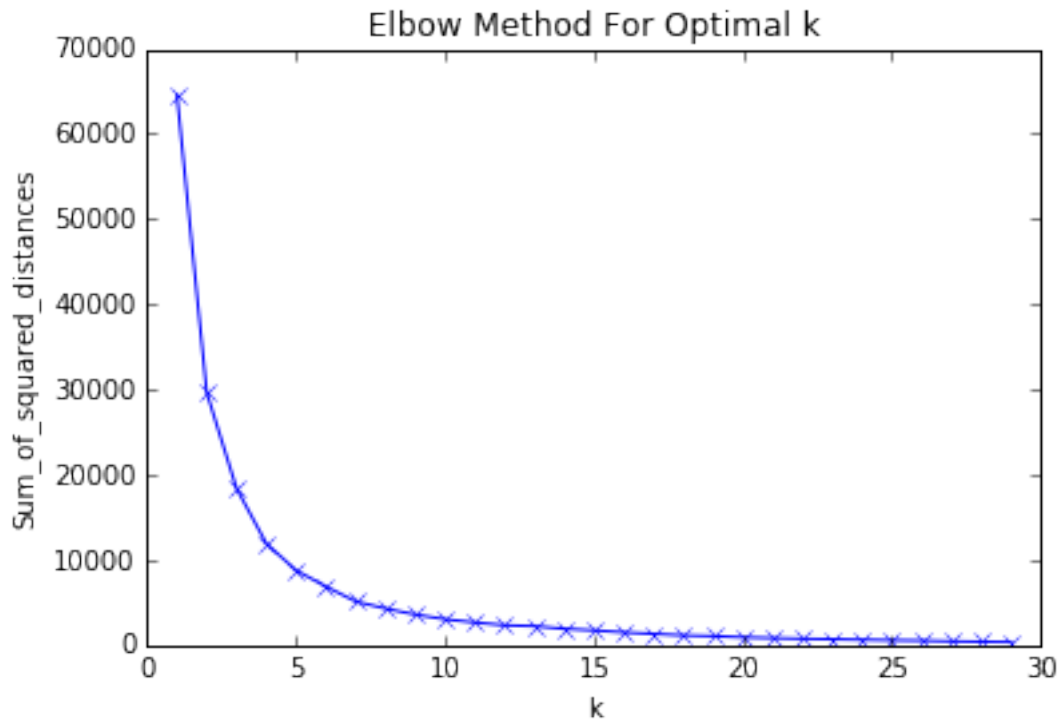
```

In [5]: # elbow plot for best number of clusters

cols = ['2012_D', '2012_R', '2012_I', '2016_D', '2016_R', '2016_I', '2020_D', '2020_R']
Sum_of_squared_distances = []
K = range(1,30)
for k in K:
    km = KMeans(n_clusters=k, random_state=0).fit(data[cols])
    Sum_of_squared_distances.append(km.inertia_)

plt.plot(K, Sum_of_squared_distances, 'bx-')
plt.xlabel('k')
plt.ylabel('Sum_of_squared_distances')
plt.title('Elbow Method For Optimal k')
plt.show()

```



Perform and Save Results for Optimal K=7

```
In [6]: kmeans = KMeans(n_clusters=7, random_state=0).fit(data[cols])
```

```
labels = kmeans.labels_
data['clust_labels'] = labels
```

Print Results

```
In [7]: cluster_data = []
```

```
for label in list(set(list(data['clust_labels']))):
    data_clust = data[data['clust_labels'] == label]
    individual = {}
    print("Cluster Number: " + str(label))
    print("States: " + str(list(data_clust['Unnamed: 0'])))
    print("Total electoral votes: " + str(sum(data_clust['Votes'])))
    print("2020 Average Dem Support: " + str(np.mean(data_clust['2020_D'])))
    print("2020 Average Rep Support: " + str(np.mean(data_clust['2020_R'])))
    individual['num'] = label
    individual['votes'] = sum(data_clust['Votes'])
    individual['dems'] = np.mean(data_clust['2020_D'])
    individual['reps'] = np.mean(data_clust['2020_R'])
```

```

individual['states'] = str(list(data_clust['Unnamed: 0']))
cluster_data.append(individual)
print()
print('-----')
print()

```

Cluster Number: 0

States: ['AL', 'AK', 'AR', 'IN', 'KS', 'KY', 'LA', 'MS', 'MO', 'MT', 'NE 1', 'SC', 'SD', 'TN',
Total electoral votes: 132
2020 Average Dem Support: 37.459999999999994
2020 Average Rep Support: 57.58

Cluster Number: 1

States: ['CT', 'DE', 'HI', 'IL', 'Maine 1', 'MD', 'MA', 'NJ ', 'OR', 'RI']
Total electoral votes: 81
2020 Average Dem Support: 56.070000000000001
2020 Average Rep Support: 37.74

Cluster Number: 2

States: ['AZ', 'CO', 'FL', 'GA', 'IA', 'Maine 2', 'MI', 'MN', 'NE 2', 'NV', 'NH', 'NC', 'OH',
Total electoral votes: 185
2020 Average Dem Support: 46.756249999999994
2020 Average Rep Support: 47.293749999999996

Cluster Number: 3

States: ['DC']
Total electoral votes: 3
2020 Average Dem Support: 90.9
2020 Average Rep Support: 4.7

Cluster Number: 4

States: ['CA', 'NM', 'NY', 'VT', 'WA']
Total electoral votes: 104
2020 Average Dem Support: 45.58
2020 Average Rep Support: 32.42

Cluster Number: 5

States: ['UT']

```
Total electoral votes: 6
2020 Average Dem Support: 24.9
2020 Average Rep Support: 45.0
```

```
Cluster Number: 6
States: ['ID', 'NE 3', 'ND', 'OK', 'WV', 'WY']
Total electoral votes: 23
2020 Average Dem Support: 26.733333333333333
2020 Average Rep Support: 66.11666666666667
```

Analyze individually states in the 'Battleground' Category

```
In [8]: import copy
```

```
battleground_states = copy.deepcopy(data[data.clust_labels==2])
kmeans = KMeans(n_clusters=4, random_state=0).fit(battleground_states[cols])

border_labels = kmeans.labels_
battleground_states['border_clusters'] = border_labels
```

Perform KMeans Clustering for k=16 and Print Results

```
In [9]: kmeans = KMeans(n_clusters=16, random_state=0).fit(data[cols])
```

```
labels = kmeans.labels_
data['clust_labels_take2'] = labels

cluster_data_2 = []

for label in list(set(list(data['clust_labels_take2']))):
    data_clust = data[data['clust_labels_take2'] == label]
    individual = {}
    print("Cluster Number: " + str(label))
    print("States: " + str(list(data_clust['Unnamed: 0'])))
    print("Total electoral votes: " + str(sum(data_clust['Votes'])))
    print("2020 Average Dem Support: " + str(np.mean(data_clust['2020_D'])))
    print("2020 Average Rep Support: " + str(np.mean(data_clust['2020_R'])))
    individual['num'] = label
    individual['votes'] = sum(data_clust['Votes'])
    individual['dems'] = np.mean(data_clust['2020_D'])
    individual['reps'] = np.mean(data_clust['2020_R'])
    individual['states'] = str(list(data_clust['Unnamed: 0']))
```

```
cluster_data_2.append(individual)
print()
print('-----')
print()
```

Cluster Number: 0
States: ['AZ', 'FL', 'GA', 'NE 2', 'NC']
Total electoral votes: 72
2020 Average Dem Support: 47.059999999999995
2020 Average Rep Support: 48.720000000000006

Cluster Number: 1
States: ['DC']
Total electoral votes: 3
2020 Average Dem Support: 90.9
2020 Average Rep Support: 4.7

Cluster Number: 2
States: ['MD', 'MA']
Total electoral votes: 21
2020 Average Dem Support: 60.45
2020 Average Rep Support: 34.05

Cluster Number: 3
States: ['NE 3', 'WY']
Total electoral votes: 4
2020 Average Dem Support: 22.200000000000003
2020 Average Rep Support: 70.75

Cluster Number: 4
States: ['AL', 'AR', 'KY', 'SD', 'TN']
Total electoral votes: 37
2020 Average Dem Support: 34.44
2020 Average Rep Support: 61.160000000000004

Cluster Number: 5
States: ['CO', 'MI', 'MN', 'NV', 'PA', 'VA', 'WI']
Total electoral votes: 84

2020 Average Dem Support: 48.528571428571425
2020 Average Rep Support: 45.942857142857136

Cluster Number: 6
States: ['UT']
Total electoral votes: 6
2020 Average Dem Support: 24.9
2020 Average Rep Support: 45.0

Cluster Number: 7
States: ['CA', 'NY', 'VT']
Total electoral votes: 87
2020 Average Dem Support: 46.79999999999999
2020 Average Rep Support: 31.266666666666667

Cluster Number: 8
States: ['AK', 'KS', 'MT', 'NE 1']
Total electoral votes: 13
2020 Average Dem Support: 36.825
2020 Average Rep Support: 55.45

Cluster Number: 9
States: ['NH', 'NM', 'WA']
Total electoral votes: 21
2020 Average Dem Support: 42.79999999999999
2020 Average Rep Support: 36.266666666666666

Cluster Number: 10
States: ['CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'OR', 'RI']
Total electoral votes: 56
2020 Average Dem Support: 53.699999999999996
2020 Average Rep Support: 39.957142857142856

Cluster Number: 11
States: ['OK', 'WV']
Total electoral votes: 12

2020 Average Dem Support: 29.0
2020 Average Rep Support: 66.44999999999999

Cluster Number: 12
States: ['IN', 'LA', 'MS', 'MO', 'SC', 'TX']
Total electoral votes: 82
2020 Average Dem Support: 40.4
2020 Average Rep Support: 56.016666666666666

Cluster Number: 13
States: ['HI']
Total electoral votes: 4
2020 Average Dem Support: 63.9
2020 Average Rep Support: 29.6

Cluster Number: 14
States: ['ID', 'ND']
Total electoral votes: 7
2020 Average Dem Support: 29.0
2020 Average Rep Support: 61.15

Cluster Number: 15
States: ['IA', 'Maine 2', 'OH']
Total electoral votes: 25
2020 Average Dem Support: 44.066666666666666
2020 Average Rep Support: 50.333333333333336

Rank Clustering by Democratic Support (D-R)

In [10]: *# make a column for ranked clusters by support*

```
dem_supports = []  
for label in list(set(list(data['clust_labels_take2']))):  
    data_clust = data[data['clust_labels_take2'] == label]  
    dem_support = np.mean(data_clust['2020_D']) - np.mean(data_clust['2020_R'])  
    dem_supports.append({'index': label, 'support': dem_support})
```

```

new_support = sorted(dem_supports, reverse=True, key=lambda k: k['support'])

ranked_cluster_list = []
for i in range(len(data)):
    cluster = list(data.clust_labels_take2)[i]
    for j in range(len(new_support)):
        if new_support[j]['index'] == cluster:
            ranked_cluster_list.append(j)

data['ranked_cluster'] = ranked_cluster_list

```

Tipping Point Analysis

In [11]: # GO THROUGH AND SHOW TIPPING POINT

```

print("TIPPING POINT ANALYSIS")
print("RANKING CLUSTERS BY DEMOCRATIC FAVORABILITY")
print()
print()
print()

max_clust = np.max(np.array(list(data.ranked_cluster)))
total_states = []
total_votes = 0

for i in range(max_clust):
    data_clust = data[data['ranked_cluster'] == i]

    print("Step " + str(i) + ":")
    print("States Won: " + str(list(data_clust['Unnamed: 0'])))
    total_states.extend(list(data_clust['Unnamed: 0']))
    print("New States Dem Pref: " + str(np.mean(data_clust['2020_D'])) + ", New States")
    print("Total States: " + str(total_states))
    print("Electoral Votes Won: " + str(sum(data_clust['Votes'])))
    total_votes = total_votes + sum(data_clust['Votes'])
    print("Total Electoral Votes: " + str(total_votes))
    print()
    print('-----')
    print()

```

TIPPING POINT ANALYSIS
RANKING CLUSTERS BY DEMOCRATIC FAVORABILITY

Step 0:

States Won: ['DC']
New States Dem Pref: 90.9, New States Rep Pref: 4.7
Total States: ['DC']
Electoral Votes Won: 3
Total Electoral Votes: 3

Step 1:
States Won: ['HI']
New States Dem Pref: 63.9, New States Rep Pref: 29.6
Total States: ['DC', 'HI']
Electoral Votes Won: 4
Total Electoral Votes: 7

Step 2:
States Won: ['MD', 'MA']
New States Dem Pref: 60.45, New States Rep Pref: 34.05
Total States: ['DC', 'HI', 'MD', 'MA']
Electoral Votes Won: 21
Total Electoral Votes: 28

Step 3:
States Won: ['CA', 'NY', 'VT']
New States Dem Pref: 46.79999999999999, New States Rep Pref: 31.266666666666667
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT']
Electoral Votes Won: 87
Total Electoral Votes: 115

Step 4:
States Won: ['CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'OR', 'RI']
New States Dem Pref: 53.699999999999996, New States Rep Pref: 39.957142857142856
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'OR', 'RI']
Electoral Votes Won: 56
Total Electoral Votes: 171

Step 5:
States Won: ['NH', 'NM', 'WA']
New States Dem Pref: 42.79999999999999, New States Rep Pref: 36.266666666666666
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'OR', 'RI', 'NH', 'NM', 'WA']

Electoral Votes Won: 21
Total Electoral Votes: 192

Step 6:

States Won: ['CO', 'MI', 'MN', 'NV', 'PA', 'VA', 'WI']
New States Dem Pref: 48.528571428571425, New States Rep Pref: 45.942857142857136
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O']
Electoral Votes Won: 84
Total Electoral Votes: 276

Step 7:

States Won: ['AZ', 'FL', 'GA', 'NE 2', 'NC']
New States Dem Pref: 47.059999999999995, New States Rep Pref: 48.720000000000006
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O']
Electoral Votes Won: 72
Total Electoral Votes: 348

Step 8:

States Won: ['IA', 'Maine 2', 'OH']
New States Dem Pref: 44.066666666666666, New States Rep Pref: 50.333333333333336
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O']
Electoral Votes Won: 25
Total Electoral Votes: 373

Step 9:

States Won: ['IN', 'LA', 'MS', 'MO', 'SC', 'TX']
New States Dem Pref: 40.4, New States Rep Pref: 56.016666666666666
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O']
Electoral Votes Won: 82
Total Electoral Votes: 455

Step 10:

States Won: ['AK', 'KS', 'MT', 'NE 1']
New States Dem Pref: 36.825, New States Rep Pref: 55.45
Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O']
Electoral Votes Won: 13
Total Electoral Votes: 468

Step 11:

States Won: ['UT']

New States Dem Pref: 24.9, New States Rep Pref: 45.0

Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O

Electoral Votes Won: 6

Total Electoral Votes: 474

Step 12:

States Won: ['AL', 'AR', 'KY', 'SD', 'TN']

New States Dem Pref: 34.44, New States Rep Pref: 61.160000000000004

Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O

Electoral Votes Won: 37

Total Electoral Votes: 511

Step 13:

States Won: ['ID', 'ND']

New States Dem Pref: 29.0, New States Rep Pref: 61.15

Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O

Electoral Votes Won: 7

Total Electoral Votes: 518

Step 14:

States Won: ['OK', 'WV']

New States Dem Pref: 29.0, New States Rep Pref: 66.44999999999999

Total States: ['DC', 'HI', 'MD', 'MA', 'CA', 'NY', 'VT', 'CT', 'DE', 'IL', 'Maine 1', 'NJ ', 'O

Electoral Votes Won: 12

Total Electoral Votes: 530

2.0.1 Visualize Results

Load State Boundary Data for Mapping

```
In [12]: us_map = gpd.read_file('states_21basic/states.shp')
        us_map
```

```
Out[12]:
```

	STATE_NAME	DRAWSEQ	STATE_FIPS	SUB_REGION	STATE_ABBR	
0	Hawaii	1	15	Pacific	HI	(POLYGON

1	Washington	2	53	Pacific	WA	(POLYGON
2	Montana	3	30	Mountain	MT	POLYGON
3	Maine	4	23	New England	ME	(POLYGON
4	North Dakota	5	38	West North Central	ND	POLYGON
5	South Dakota	6	46	West North Central	SD	POLYGON
6	Wyoming	7	56	Mountain	WY	POLYGON
7	Wisconsin	8	55	East North Central	WI	(POLYGON
8	Idaho	9	16	Mountain	ID	POLYGON
9	Vermont	10	50	New England	VT	POLYGON
10	Minnesota	11	27	West North Central	MN	POLYGON
11	Oregon	12	41	Pacific	OR	POLYGON
12	New Hampshire	13	33	New England	NH	POLYGON
13	Iowa	14	19	West North Central	IA	POLYGON
14	Massachusetts	15	25	New England	MA	(POLYGON
15	Nebraska	16	31	West North Central	NE	POLYGON
16	New York	17	36	Middle Atlantic	NY	(POLYGON
17	Pennsylvania	18	42	Middle Atlantic	PA	POLYGON
18	Connecticut	19	09	New England	CT	POLYGON
19	Rhode Island	20	44	New England	RI	(POLYGON
20	New Jersey	21	34	Middle Atlantic	NJ	POLYGON
21	Indiana	22	18	East North Central	IN	POLYGON
22	Nevada	23	32	Mountain	NV	POLYGON
23	Utah	24	49	Mountain	UT	POLYGON
24	California	25	06	Pacific	CA	(POLYGON
25	Ohio	26	39	East North Central	OH	POLYGON
26	Illinois	27	17	East North Central	IL	POLYGON
27	District of Columbia	28	11	South Atlantic	DC	POLYGON
28	Delaware	29	10	South Atlantic	DE	POLYGON
29	West Virginia	30	54	South Atlantic	WV	POLYGON
30	Maryland	31	24	South Atlantic	MD	(POLYGON
31	Colorado	32	08	Mountain	CO	POLYGON
32	Kentucky	33	21	East South Central	KY	(POLYGON
33	Kansas	34	20	West North Central	KS	POLYGON
34	Virginia	35	51	South Atlantic	VA	(POLYGON
35	Missouri	36	29	West North Central	MO	POLYGON
36	Arizona	37	04	Mountain	AZ	POLYGON
37	Oklahoma	38	40	West South Central	OK	POLYGON
38	North Carolina	39	37	South Atlantic	NC	(POLYGON
39	Tennessee	40	47	East South Central	TN	POLYGON
40	Texas	41	48	West South Central	TX	(POLYGON
41	New Mexico	42	35	Mountain	NM	POLYGON
42	Alabama	43	01	East South Central	AL	POLYGON
43	Mississippi	44	28	East South Central	MS	POLYGON
44	Georgia	45	13	South Atlantic	GA	(POLYGON
45	South Carolina	46	45	South Atlantic	SC	(POLYGON
46	Arkansas	47	05	West South Central	AR	POLYGON
47	Louisiana	48	22	West South Central	LA	(POLYGON
48	Florida	49	12	South Atlantic	FL	(POLYGON

49	Michigan	50	26	East North Central	MI	(POLYGON
50	Alaska	51	02	Pacific	AK	(POLYGON

Merge State Boundary Data with Cluster Data

```
In [13]: clusters = []
```

```
for index, row in us_map.iterrows():
    state = row['STATE_ABBR']
    cluster = list(data[data.State == state].ranked_cluster)
    if len(cluster) > 1:
        cluster = [-1]

    clusters.append(int(cluster[0]))
```

```
us_map['cluster'] = clusters
us_map
```

```
Out[13]:
```

	STATE_NAME	DRAWSEQ	STATE_FIPS	SUB_REGION	STATE_ABBR	
0	Hawaii	1	15	Pacific	HI	(POLYGON
1	Washington	2	53	Pacific	WA	(POLYGON
2	Montana	3	30	Mountain	MT	POLYGON
3	Maine	4	23	New England	ME	(POLYGON
4	North Dakota	5	38	West North Central	ND	POLYGON
5	South Dakota	6	46	West North Central	SD	POLYGON
6	Wyoming	7	56	Mountain	WY	POLYGON
7	Wisconsin	8	55	East North Central	WI	(POLYGON
8	Idaho	9	16	Mountain	ID	POLYGON
9	Vermont	10	50	New England	VT	POLYGON
10	Minnesota	11	27	West North Central	MN	POLYGON
11	Oregon	12	41	Pacific	OR	POLYGON
12	New Hampshire	13	33	New England	NH	POLYGON
13	Iowa	14	19	West North Central	IA	POLYGON
14	Massachusetts	15	25	New England	MA	(POLYGON
15	Nebraska	16	31	West North Central	NE	POLYGON
16	New York	17	36	Middle Atlantic	NY	(POLYGON
17	Pennsylvania	18	42	Middle Atlantic	PA	POLYGON
18	Connecticut	19	09	New England	CT	POLYGON
19	Rhode Island	20	44	New England	RI	(POLYGON
20	New Jersey	21	34	Middle Atlantic	NJ	POLYGON
21	Indiana	22	18	East North Central	IN	POLYGON
22	Nevada	23	32	Mountain	NV	POLYGON
23	Utah	24	49	Mountain	UT	POLYGON
24	California	25	06	Pacific	CA	(POLYGON
25	Ohio	26	39	East North Central	OH	POLYGON
26	Illinois	27	17	East North Central	IL	POLYGON
27	District of Columbia	28	11	South Atlantic	DC	POLYGON
28	Delaware	29	10	South Atlantic	DE	POLYGON

29	West Virginia	30	54	South Atlantic	WV	POLYGON
30	Maryland	31	24	South Atlantic	MD	(POLYGON
31	Colorado	32	08	Mountain	CO	POLYGON
32	Kentucky	33	21	East South Central	KY	(POLYGON
33	Kansas	34	20	West North Central	KS	POLYGON
34	Virginia	35	51	South Atlantic	VA	(POLYGON
35	Missouri	36	29	West North Central	MO	POLYGON
36	Arizona	37	04	Mountain	AZ	POLYGON
37	Oklahoma	38	40	West South Central	OK	POLYGON
38	North Carolina	39	37	South Atlantic	NC	(POLYGON
39	Tennessee	40	47	East South Central	TN	POLYGON
40	Texas	41	48	West South Central	TX	(POLYGON
41	New Mexico	42	35	Mountain	NM	POLYGON
42	Alabama	43	01	East South Central	AL	POLYGON
43	Mississippi	44	28	East South Central	MS	POLYGON
44	Georgia	45	13	South Atlantic	GA	(POLYGON
45	South Carolina	46	45	South Atlantic	SC	(POLYGON
46	Arkansas	47	05	West South Central	AR	POLYGON
47	Louisiana	48	22	West South Central	LA	(POLYGON
48	Florida	49	12	South Atlantic	FL	(POLYGON
49	Michigan	50	26	East North Central	MI	(POLYGON
50	Alaska	51	02	Pacific	AK	(POLYGON

Important Mapping Settings

```
In [14]: fig_size = plt.rcParams["figure.figsize"]
print("Current size:" + str(fig_size))
fig_size[0] = 14
fig_size[1] = 12
plt.rcParams["figure.figsize"] = fig_size
print("Current size:" + str(fig_size))

cmap_politics = matplotlib.colors.LinearSegmentedColormap.from_list("", ["navy", "darkk
```

Current size:[6.0, 4.0]

Current size:[14, 12]

Plot Lower 48 States + DC

```
In [15]: import geoplot

states = list(range(1,50))
states.remove(3)
states.remove(15)

fig = plt.figure(1)
ax1 = plt.subplot(111)
```



```

fig.set_figheight(8)
fig.set_figwidth(14)

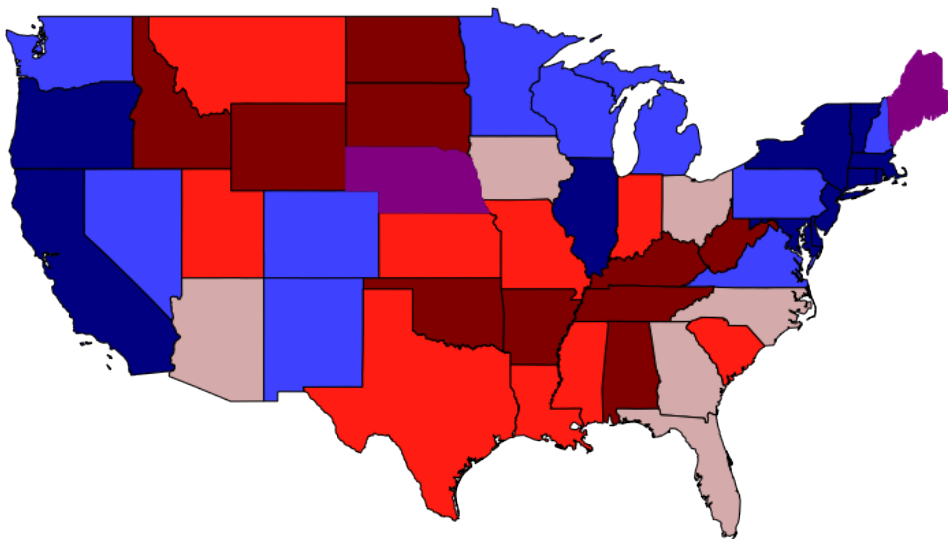
p1 = geoplot.choropleth(us_map.loc[states, :], hue='cluster', cmap=cmap_politics, ax=ax1)
p2 = geoplot.choropleth(us_map[3:4], hue='cluster', ax=ax1, color="purple", edgecolor="black")
p3 = geoplot.choropleth(us_map[15:16], hue='cluster', ax=ax1, color="purple", edgecolor="black")
plt.xlim(-125,-60)
plt.ylim(22,50)
plt.axis('off')
plt.show()

```

```

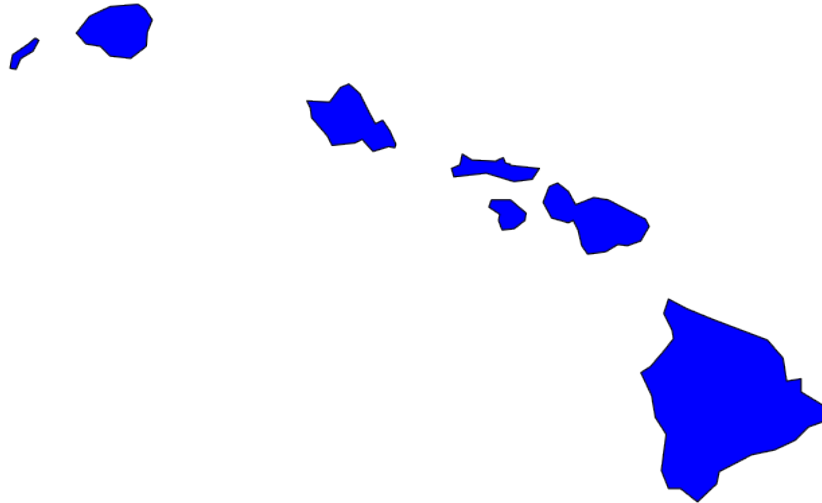
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/pysal/__init__.py:65: VisibleDeprecationWarning:
  ), VisibleDeprecationWarning)
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/scipy/stats/stats.py:1713: FutureWarning:
  return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/pysal/esda/mapclassify.py:702: RuntimeWarning:
  gadf = 1 - self.adcm / adam
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/matplotlib/patches.py:83: UserWarning:
  warnings.warn("Setting the 'color' property will override")
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/matplotlib/patches.py:83: UserWarning:
  warnings.warn("Setting the 'color' property will override")
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/matplotlib/patches.py:83: UserWarning:
  warnings.warn("Setting the 'color' property will override")

```



Hawaii

```
In [16]: us_map.loc[[0], :].plot()  
plt.axis('off')  
plt.show()
```



Alaska

```
In [17]: us_map.loc[[50], :].plot(color = 'red')  
plt.axis('off')  
plt.show()
```



3 County Analysis: Winning Key Votes

We want to target young, diverse, urban communities in these key states to drive high turnout in these communities and capitalize on 2018 gains in these challenging state to flip the state for 2020.

Load Census Data on Age and Race to build County Database

```
In [18]: age = pd.read_csv('demographics/age.csv')
age['County'] = age.Geography.apply(lambda x: x.split(' County')[0])
age['State'] = age.Geography.apply(lambda x: x.split('County, ')[1])
print(len(age))
age.head()
```

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```
Out [18]:
```

	Id	Id2	Geography	Estimate; Total: Margin of Error; Total: Margin of Error
0	0500000US04001	4001	Apache County, Arizona	71602
1	0500000US04003	4003	Cochise County, Arizona	126516
2	0500000US04005	4005	Coconino County, Arizona	138639
3	0500000US04007	4007	Gila County, Arizona	53145
4	0500000US04009	4009	Graham County, Arizona	37700

	Margin of Error; Male: - 40 to 44 years	Estimate; Male: - 45 to 49 years	Margin of Error; Male: - 50 to 54 years
0	189	2000	
1	320	3549	
2	343	3547	
3	168	1395	
4	173	1276	

	Estimate; Female: - 5 to 9 years	Margin of Error; Female: - 5 to 9 years	Estimate; Female: - 10 to 14 years
0	2722	190	
1	3818	318	
2	4232	325	
3	1457	153	
4	1433	198	

	Estimate; Female: - 55 to 59 years	Margin of Error; Female: - 55 to 59 years	Estimate; Female: - 60 to 64 years
0	2351	157	
1	4176	374	
2	4559	250	
3	2203	292	
4	812	112	

```
In [19]: race = pd.read_csv('demographics/race.csv')
age['County'] = age.Geography.apply(lambda x: x.split(' County')[0])
```

```
age['State'] = age.Geography.apply(lambda x: x.split('County, ')[1])
race.head()
```

```
Out[19]:
```

	Id	Id2	Geography	Estimate; Total: Margin of Error; T
0	0500000US04001	4001	Apache County, Arizona	71602
1	0500000US04003	4003	Cochise County, Arizona	126516
2	0500000US04005	4005	Coconino County, Arizona	138639
3	0500000US04007	4007	Gila County, Arizona	53145
4	0500000US04009	4009	Graham County, Arizona	37700


```
Estimate; Total: - Two or more races: - Two races excluding Some other race, and t
```

	Estimate; Total: - Two or more races: - Two races excluding Some other race, and t
0	1437
1	4753
2	4361
3	1382
4	743

```
In [20]: counties = pd.read_csv('county_data/county_fips.csv')
counties.head()
```

```
Out[20]:
```

	msa	pmsa	county	state	county_fips	state_fips
0	NaN	NaN	Aleutians East	AK	13	2
1	NaN	NaN	Aleutians West	AK	16	2
2	380.0	NaN	Anchorage	AK	20	2
3	NaN	NaN	Bethel	AK	50	2
4	NaN	NaN	Bristol Bay	AK	60	2

Load County Voting Dataset

```
In [21]: voting = pd.read_csv('county_pres.csv')
voting.head()
```

```
Out[21]:
```

	year	state	state_po	county	FIPS	office	candidate	party	car
0	2000	Alabama	AL	Autauga	1001.0	President	Al Gore	democrat	
1	2000	Alabama	AL	Autauga	1001.0	President	George W. Bush	republican	
2	2000	Alabama	AL	Autauga	1001.0	President	Ralph Nader	green	
3	2000	Alabama	o	Autauga	1001.0	President	Other	NaN	
4	2000	Alabama	AL	Baldwin	1003.0	President	Al Gore	democrat	

```
In [22]: states = ['Arizona', 'Georgia', 'Florida', 'North Carolina']
voting = voting[voting['state'].isin(states)]
counties = set(list(voting.county))
```

```
county_list = []
d_2016 = []
r_2016 = []
d_2012 = []
r_2012 = []
geography = []
```

```

for state in states:
    state_data = voting[voting['state'] == state]
    counties = set(list(state_data.county))
    for county in counties:
        county_list.append(county)
        county_data = voting[voting.county == county]
        data_2016 = county_data[county_data.year == 2016]
        data_2012 = county_data[county_data.year == 2012]

        dem_2016 = data_2016[data_2016.party == 'democrat']
        dem_2012 = data_2012[data_2012.party == 'democrat']
        rep_2016 = data_2016[data_2016.party == 'republican']
        rep_2012 = data_2012[data_2012.party == 'republican']

        d_2016.append(list(dem_2016.candidatevotes)[0]*1.0/list(dem_2016.totalvotes))
        d_2012.append(list(dem_2012.candidatevotes)[0]*1.0/list(dem_2012.totalvotes))
        r_2016.append(list(rep_2016.candidatevotes)[0]*1.0/list(rep_2016.totalvotes))
        r_2012.append(list(rep_2012.candidatevotes)[0]*1.0/list(rep_2012.totalvotes))
        geography.append(county + " County, " + str(state))

county_voting = pd.DataFrame({'County': list(county_list), 'Geography': geography, '2016_r': d_2016, '2012_r': d_2012, '2016_d': r_2016, '2012_d': r_2012})
county_voting['2016_r'].apply(lambda x: x*100.0)
county_voting['2012_r'].apply(lambda x: x*100.0)
county_voting['2016_d'].apply(lambda x: x*100.0)
county_voting['2012_d'].apply(lambda x: x*100.0)
#county_voting['Geography'] = county_voting['County'].apply(lambda x: x + " County")
county_voting.to_csv("County_voting.csv")
county_voting.head()

```

```

Out [22]:
   County  Geography  2016_r  2012_r  2016_d  2012_d
0  Greenlee  Greenlee County, Arizona  0.364407  0.535306  0.210324  0.440484
1   Cochise  Cochise County, Arizona  0.561671  0.601943  0.348895  0.378467
2     Gila    Gila County, Arizona  0.386462  0.625000  0.190833  0.357534
3   Graham  Graham County, Arizona  0.395203  0.681232  0.162563  0.304429
4   Apache  Apache County, Arizona  0.297892  0.319174  0.617584  0.663378

```

Merge Datasets

```

In [23]: demographic_data = pd.merge(age, race, on='Geography')
print(len(demographic_data))
demographic_data.head()

```

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```

Out [23]:
   Id_x  Id2_x  Geography  Estimate; Total:_x  Margin of Error
0  0500000US04001  4001  Apache County, Arizona  71602
1  0500000US04003  4003  Cochise County, Arizona  126516

```

2	0500000US04005	4005	Coconino County, Arizona	138639
3	0500000US04007	4007	Gila County, Arizona	53145
4	0500000US04009	4009	Graham County, Arizona	37700

	Margin of Error; Male: - 40 to 44 years	Estimate; Male: - 45 to 49 years	Margin of Error; Male: - 50 to 54 years
0	189	2000	189
1	320	3549	320
2	343	3547	343
3	168	1395	168
4	173	1276	173

	Estimate; Female: - 5 to 9 years	Margin of Error; Female: - 5 to 9 years	Estimate; Female: - 10 to 14 years
0	2722	190	2722
1	3818	318	3818
2	4232	325	4232
3	1457	153	1457
4	1433	198	1433

	Estimate; Female: - 55 to 59 years	Margin of Error; Female: - 55 to 59 years	Estimate; Female: - 60 to 64 years
0	2351	157	2351
1	4176	374	4176
2	4559	250	4559
3	2203	292	2203
4	812	112	812

	Margin of Error; Total: - Black or African American alone	Estimate; Total: - American Indian or Alaska Native alone
0	134	134
1	413	413
2	241	241
3	112	112
4	95	95

```
In [24]: demographic_data = pd.merge(demographic_data, county_voting, on='Geography')
```

Analyze list of available features for these counties and Make new features that are representative of our target audience

Selected features: 'Geography', 'County', 'State', 'percent_male_0_17', 'percent_male_18_24', 'percent_male_25_34', 'percent_male_35_44', 'percent_male_45_59', 'percent_male_60_79', 'percent_male_80_up', 'percent_female_0_17', 'percent_female_18_24', 'percent_female_25_34', 'percent_female_35_44', 'percent_female_45_59', 'percent_female_60_79', 'percent_female_80_up', 'percent_white', 'percent_black', 'percent_native', 'percent_asian', 'percent_hawaiian', 'percent_other', 'percent_two', 'percent_two_other', 'percent_three'

```
In [25]: #for col in demographic_data.columns:
#         print(col)
```

```
clustering_cols = ['Geography', 'County_x', 'State', 'percent_male_0_17', 'percent_male_18_24', 'percent_male_25_34', 'percent_male_35_44', 'percent_male_45_59', 'percent_male_60_79', 'percent_male_80_up', 'percent_female_0_17', 'percent_female_18_24', 'percent_female_25_34', 'percent_female_35_44', 'percent_female_45_59', 'percent_female_60_79', 'percent_female_80_up', 'percent_white', 'percent_black', 'percent_native', 'percent_asian', 'percent_hawaiian', 'percent_other', 'percent_two', 'percent_two_other', 'percent_three']
```

```

demographic_data['percent_male_0_17'] = demographic_data[['Estimate; Male: - Under 5 y
demographic_data['percent_male_18_24'] = demographic_data[['Estimate; Male: - 18 and
demographic_data['percent_male_25_34'] = demographic_data[['Estimate; Male: - 25 to 29
demographic_data['percent_male_35_44'] = demographic_data[['Estimate; Male: - 35 to 39
demographic_data['percent_male_45_59'] = demographic_data[['Estimate; Male: - 45 to 49
demographic_data['percent_male_60_79'] = demographic_data[['Estimate; Male: - 60 and 6
demographic_data['percent_male_80_up'] = demographic_data[['Estimate; Male: - 80 to 84
demographic_data['percent_female_0_17'] = demographic_data[['Estimate; Female: - Under
demographic_data['percent_female_18_24'] = demographic_data[['Estimate; Female: - 18 a
demographic_data['percent_female_25_34'] = demographic_data[['Estimate; Female: - 25 t
demographic_data['percent_female_35_44'] = demographic_data[['Estimate; Female: - 35 t
demographic_data['percent_female_45_59'] = demographic_data[['Estimate; Female: - 45 t
demographic_data['percent_female_60_79'] = demographic_data[['Estimate; Female: - 60 a
demographic_data['percent_female_80_up'] = demographic_data[['Estimate; Female: - 80 t
demographic_data['percent_white'] = demographic_data['Estimate; Total: - White alone']
demographic_data['percent_black'] = demographic_data['Estimate; Total: - Black or Afr
demographic_data['percent_native'] = demographic_data['Estimate; Total: - American Ind
demographic_data['percent_asian'] = demographic_data['Estimate; Total: - Asian alone']
demographic_data['percent_hawaiian'] = demographic_data['Estimate; Total: - Native Haw
demographic_data['percent_other'] = demographic_data['Estimate; Total: - Some other ra
demographic_data['percent_two'] = demographic_data['Estimate; Total: - Two or more ra
demographic_data['percent_two_other'] = demographic_data['Estimate; Total: - Two or mo
demographic_data['percent_three'] = demographic_data['Estimate; Total: - Two or more
demographic_data['percent_male'] = demographic_data['Estimate; Male:'].div(demographi
demographic_data['percent_female'] = demographic_data['Estimate; Female:'].div(demogr

demographic_data[clustering_cols].head()

```

```

Out [25]:

```

	Geography	County_x	State	percent_male_0_17	percent_male_18_24
0	Apache County, Arizona	Apache	Arizona	0.139298	0.052526
1	Cochise County, Arizona	Cochise	Arizona	0.110381	0.047630
2	Coconino County, Arizona	Coconino	Arizona	0.109082	0.093949
3	Gila County, Arizona	Gila	Arizona	0.104318	0.035714
4	Graham County, Arizona	Graham	Arizona	0.139178	0.059761

Split by State

```

In [26]: az_data = demographic_data[demographic_data.State == 'Arizona']
nc_data = demographic_data[demographic_data.State == 'North Carolina']
ga_data = demographic_data[demographic_data.State == 'Georgia']
fl_data = demographic_data[demographic_data.State == 'Florida']
print("Number of counties in GA: " + str(len(ga_data)))
print("Number of counties in FL: " + str(len(fl_data)))
print("Number of counties in AZ: " + str(len(az_data)))
print("Number of counties in NC: " + str(len(nc_data)))

```

```

Number of counties in GA: 159
Number of counties in FL: 66

```

Number of counties in AZ: 15
Number of counties in NC: 100

3.0.1 Arizona County Analysis

In [27]: az_data.head()

```
Out[27]:
```

	Id_x	Id2_x	Geography	Estimate; Total:_x	Margin of Error
0	0500000US04001	4001	Apache County, Arizona	71602	
1	0500000US04003	4003	Cochise County, Arizona	126516	
2	0500000US04005	4005	Coconino County, Arizona	138639	
3	0500000US04007	4007	Gila County, Arizona	53145	
4	0500000US04009	4009	Graham County, Arizona	37700	

	Margin of Error; Male: - 40 to 44 years	Estimate; Male: - 45 to 49 years	Margin of Error
0	189		2000
1	320		3549
2	343		3547
3	168		1395
4	173		1276

	Estimate; Female: - 5 to 9 years	Margin of Error; Female: - 5 to 9 years	Estimate
0	2722		190
1	3818		318
2	4232		325
3	1457		153
4	1433		198

	Estimate; Female: - 55 to 59 years	Margin of Error; Female: - 55 to 59 years	Estimate
0	2351		157
1	4176		374
2	4559		250
3	2203		292
4	812		112

	Margin of Error; Total: - Black or African American alone	Estimate; Total: - American Indian or Alaska Native alone
0	134	
1	413	
2	241	
3	112	
4	95	

	2012_d	percent_male_0_17	percent_male_18_24	percent_male_25_34	percent_male_35_44
0	0.663378	0.139298	0.052526	0.063727	0.052526
1	0.378467	0.110381	0.047630	0.071564	0.052526
2	0.565560	0.109082	0.093949	0.070211	0.052526
3	0.357534	0.104318	0.035714	0.048979	0.052526

4	0.304429	0.139178	0.059761	0.089284	0.0
---	----------	----------	----------	----------	-----

```
In [28]: #clustering_cols_analysis = ['2012_d', '2012_r', '2016_d', '2016_r', 'percent_male_0_18']
clustering_cols_analysis = ['2012_d', '2012_r', '2016_d', '2016_r', 'percent_male_18_24']
kmeans = KMeans(n_clusters=3, random_state=0).fit(az_data[clustering_cols_analysis])

labels = kmeans.labels_
az_data['az_clusters'] = labels
#az_data.head()
```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:6: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>

```
In [29]: cluster_data = []
```

```
print("Expecting Cococino, Navajo, Apache, Maricopa, Pima, Santa Cruz")
print()
print()

for label in list(set(list(az_data['az_clusters']))):
    data_clust = az_data[az_data['az_clusters'] == label]
    individual = {}
    print("Cluster Number: " + str(label))
    print("Counties: " + str(list(data_clust['County_x'])))
    print("Percent 18-24: " + str(np.average(list(data_clust['percent_male_18_24']))*100.0))
    print("Percent White: " + str(np.average(list(data_clust['percent_white']))*100.0))
    print("Percent Black: " + str(np.average(list(data_clust['percent_black']))*100.0))
    print("Percent Native American: " + str(np.average(list(data_clust['percent_native_american']))*100.0))
    print("Percent Other: " + str(np.average(list(data_clust['percent_other']))*100.0))
    print("Percent Male: " + str(np.average(list(data_clust['percent_male']))*100.0))
    print("Percent Female: " + str(np.average(list(data_clust['percent_female']))*100.0))
    print("Percent Dem: " + str(np.average(list(data_clust['2016_d']))*100.0) + "%")
    print()
    print('-----')
    print()
```

Expecting Cococino, Navajo, Apache, Maricopa, Pima, Santa Cruz

```
Cluster Number: 0
Counties: ['Cochise', 'Gila', 'Graham', 'Greenlee', 'La Paz', 'Mohave', 'Pinal', 'Yavapai']
Percent 18-24: 8.076219342642531%
Percent White: 83.77142934028477%
Percent Black: 1.9192058677049266%
```

Percent Native American: 7.016827393991859%
Percent Other: 3.3682159500233304%
Percent Male: 51.2587880487463%
Percent Female: 48.7412119512537%
Percent Dem: 25.953447081957837%

Cluster Number: 1
Counties: ['Apache', 'Coconino', 'Navajo']
Percent 18-24: 13.27111416362613%
Percent White: 44.74338973261921%
Percent Black: 0.9215862648670751%
Percent Native American: 48.0126883673408%
Percent Other: 2.262869136228706%
Percent Male: 49.56516151279517%
Percent Female: 50.434838487204836%
Percent Dem: 52.26916603956886%

Cluster Number: 2
Counties: ['Maricopa', 'Pima', 'Santa Cruz', 'Yuma']
Percent 18-24: 10.738596695770477%
Percent White: 78.54421001048408%
Percent Black: 2.8527495520351063%
Percent Native American: 1.7751143151573008%
Percent Other: 11.614904330566288%
Percent Male: 49.49658973016402%
Percent Female: 50.50341026983598%
Percent Dem: 53.917193321978495%

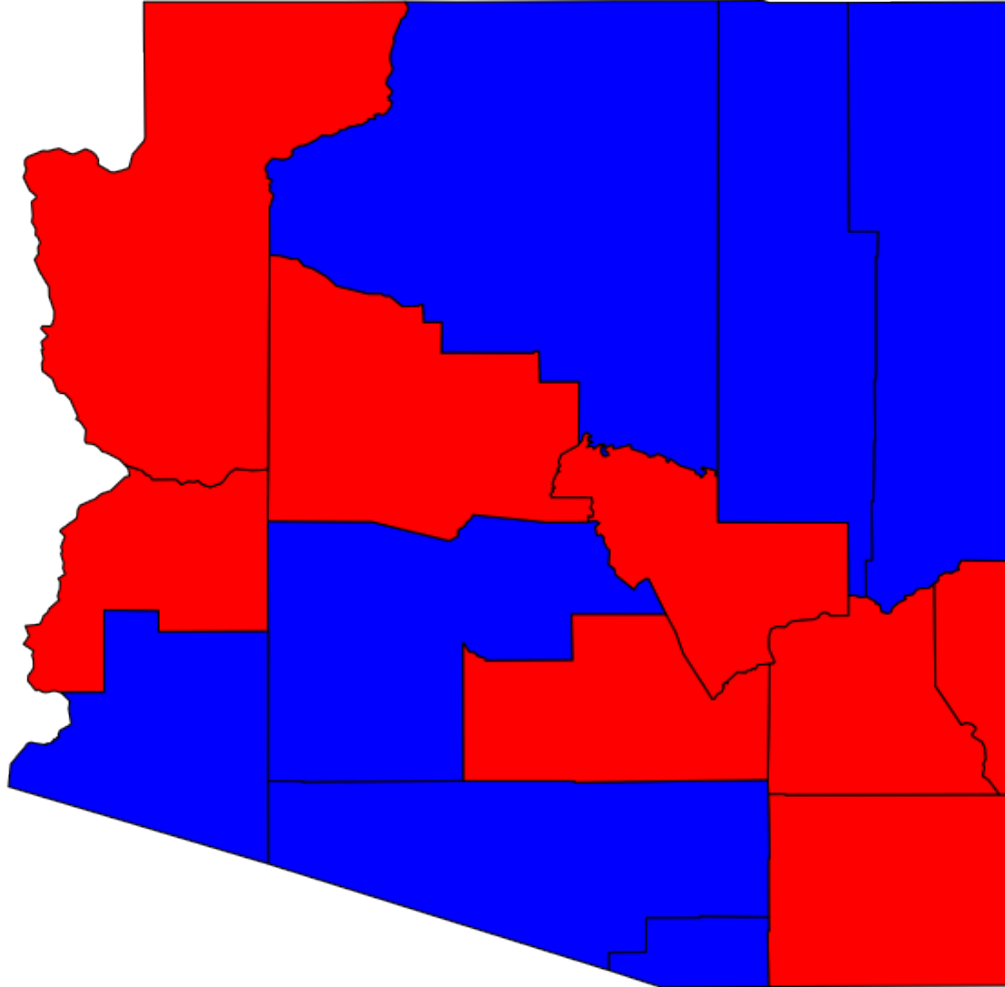
```
In [30]: binary = []
         for index, row in az_data.iterrows():
             if row.az_clusters == 0:
                 binary.append(2)
             else:
                 binary.append(1)

         az_data['binary_clusters'] = binary
```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>

```
In [31]: county = gpd.read_file('azcounty/azcounties.shp')
county.rename(columns={'NAME': 'County_x'}, inplace=True)
county = pd.merge(az_data[['County_x', 'binary_clusters']], county, on='County_x')
cmap_politics_2 = matplotlib.colors.LinearSegmentedColormap.from_list("", ["blue", "red"])
county = gpd.GeoDataFrame(county)
county.plot(column="binary_clusters", cmap=cmap_politics_2)
plt.axis('off')
plt.show()
```



3.0.2 Florida County Analysis

```
In [32]: clustering_cols_analysis = ['2012_d', '2012_r', '2016_d', '2016_r', 'percent_male_18_24']
        kmeans = KMeans(n_clusters=3, random_state=0).fit(fl_data[clustering_cols_analysis])

        labels = kmeans.labels_
        fl_data['fl_clusters'] = labels
        #az_data.head()
```

```
/Users/joshkuppersmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:5: SettingW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html
"""
```

```
In [33]: cluster_data = []
        for label in list(set(list(fl_data['fl_clusters']))):
            data_clust = fl_data[fl_data['fl_clusters'] == label]
            individual = {}
            print("Cluster Number: " + str(label))
            print("Counties: " + str(list(data_clust['County_x'])))
            print("Percent 18-24: " + str(np.average(list(data_clust['percent_male_18_24'])) + 1
            print("Percent White: " + str(np.average(list(data_clust['percent_white']))*100.0)
            print("Percent Black: " + str(np.average(list(data_clust['percent_black']))*100.0)
            print("Percent Native American: " + str(np.average(list(data_clust['percent_native
            print("Percent Other: " + str(np.average(list(data_clust['percent_other']))*100.0)
            print("Percent Male: " + str(np.average(list(data_clust['percent_male']))*100.0)
            print("Percent Female: " + str(np.average(list(data_clust['percent_female']))*100
            print("Percent Dem: " + str(np.average(list(data_clust['2016_d']))*100.0) + "%")
            print()
            print('-----')
            print()
```

```
Cluster Number: 0
Counties: ['Baker', 'Bay', 'Bradford', 'Calhoun', 'Clay', 'Columbia', 'Dixie', 'Franklin', 'Gi
Percent 18-24: 8.254400265935418%
Percent White: 82.19533669628824%
Percent Black: 12.674482342540038%
Percent Native American: 0.48237761753699887%
Percent Other: 1.3346839003247681%
Percent Male: 53.829176761967034%
Percent Female: 46.170823238032966%
Percent Dem: 22.841300984641023%
```

```
-----
```

```
Cluster Number: 1
Counties: ['Brevard', 'Charlotte', 'Citrus', 'Collier', 'Escambia', 'Flagler', 'Glades', 'Hami
Percent 18-24: 7.644140415005304%
Percent White: 83.01945966908968%
Percent Black: 10.667609197060838%
Percent Native American: 0.5535211411889396%
Percent Other: 1.8657831497436514%
Percent Male: 49.778534002622585%
Percent Female: 50.22146599737742%
```

Percent Dem: 37.27347267867142%

Cluster Number: 2

Counties: ['Alachua', 'Broward', 'Duval', 'Gadsden', 'Hillsborough', 'Jefferson', 'Leon', 'Mad.

Percent 18-24: 10.779946009433996%

Percent White: 65.1442296519164%

Percent Black: 26.490241984654073%

Percent Native American: 0.26724388438947577%

Percent Other: 2.7944530976374997%

Percent Male: 49.238901513275145%

Percent Female: 50.761098486724855%

Percent Dem: 56.14042315179603%

```
In [34]: binary = []
        for index, row in fl_data.iterrows():
            if row.fl_clusters == 2:
                binary.append(1)
            else:
                binary.append(2)

        fl_data['binary_clusters'] = binary
```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingW

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>

```
In [35]: def strip(string):
        punctuations = '!"()-[]{};:'"\<>./?@#$$%^&*~'' '
        no_punct = ""
        for char in string:
            if char not in punctuations:
                no_punct = no_punct + char
        return no_punct
```

```
In [36]: county = gpd.read_file('scuov/scuo.shp')
        county = county[county.STATE == 'FL']
        #print(len(county[county.STATE == 'FL']))
        #print(len(fl_data))
        county.head()
```

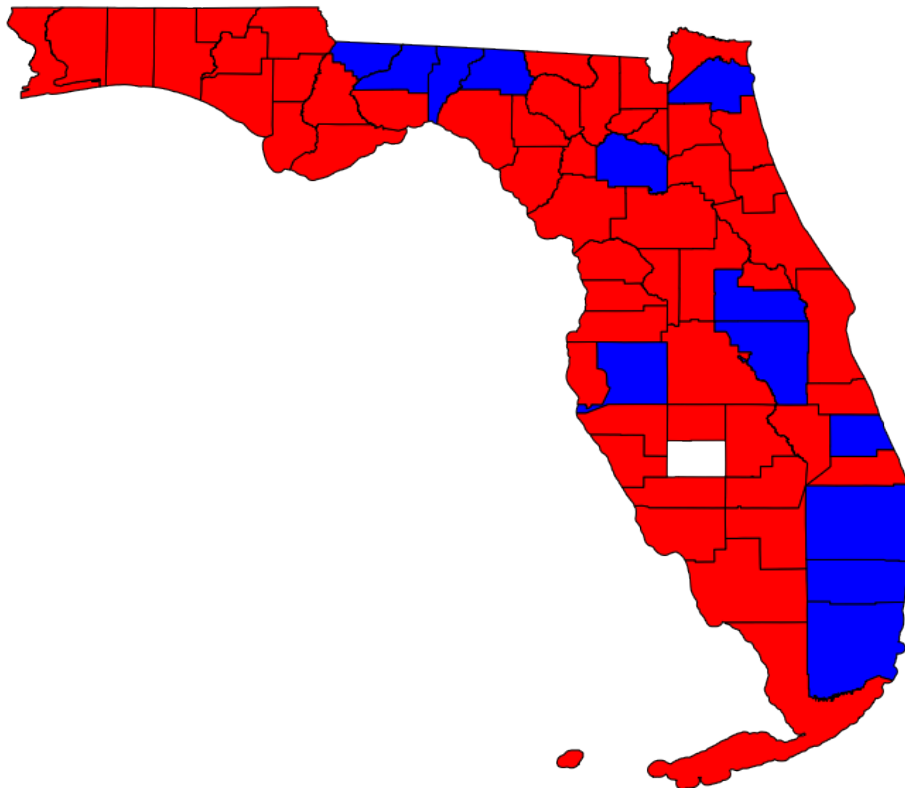
```

county.rename(columns={'NAME': 'County_x'}, inplace=True)
county['county_lower'] = county['County_x'].str.lower().apply(lambda x: strip(x))
fl_data['county_lower'] = fl_data['County_x'].str.lower().apply(lambda x: strip(x))
county = pd.merge(fl_data[['county_lower', 'binary_clusters']], county, on='county_lo
#cmap_politics_2 = matplotlib.colors.LinearSegmentedColormap.from_list("", ["blue", "r
county = gpd.GeoDataFrame(county)
county.plot(column="binary_clusters", cmap=cmap_politics_2)
plt.axis('off')
plt.show()

```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>



3.0.3 Georgia County Analysis

```
In [37]: clustering_cols_analysis = ['2012_d', '2012_r', '2016_d', '2016_r', 'percent_male_18_24']
        kmeans = KMeans(n_clusters=3, random_state=0).fit(ga_data[clustering_cols_analysis])

        labels = kmeans.labels_
        ga_data['ga_clusters'] = labels
```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:5: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>
"""

```
In [38]: for label in list(set(list(ga_data['ga_clusters']))):
        data_clust = ga_data[ga_data['ga_clusters'] == label]
        individual = {}
        print("Cluster Number: " + str(label))
        print("Counties: " + str(list(data_clust['County_x'])))
        print("Percent 18-24: " + str(np.average(list(data_clust['percent_male_18_24']))) + "%")
        print("Percent White: " + str(np.average(list(data_clust['percent_white']))) * 100.0 + "%")
        print("Percent Black: " + str(np.average(list(data_clust['percent_black']))) * 100.0 + "%")
        print("Percent Native American: " + str(np.average(list(data_clust['percent_native_american']))) * 100.0 + "%")
        print("Percent Other: " + str(np.average(list(data_clust['percent_other']))) * 100.0 + "%")
        print("Percent Male: " + str(np.average(list(data_clust['percent_male']))) * 100.0 + "%")
        print("Percent Female: " + str(np.average(list(data_clust['percent_female']))) * 100.0 + "%")
        print("Percent Dem: " + str(np.average(list(data_clust['2016_d'])) * 100.0) + "%")
        print()
        print('-----')
        print()
```

```
Cluster Number: 0
Counties: ['Baldwin', 'Bibb', 'Burke', 'Chatham', 'Clarke', 'Clayton', 'DeKalb', 'Dooly', 'Dougherty', 'Fulton', 'Gwinnett', 'Hall', 'Harris', 'Henry', 'Houston', 'Iberia', 'Jefferson', 'Jones', 'Lamar', 'Liberty', 'Madison', 'Macon', 'Marion', 'McIntosh', 'Meriwether', 'Miller', 'Mitchell', 'Monroe', 'Murray', 'Newton', 'Oconee', 'Oglethorpe', 'Peach', 'Pickens', 'Polk', 'Pulaski', 'Quitman', 'Randolph', 'Richmond', 'Rockdale', 'Seminole', 'Spalding', 'Stewart', 'Sumter', 'Taliaferro', 'Telford', 'Thomas', 'Troup', 'Turner', 'Twiggs', 'Union', 'Upson', 'Walker', 'Walton', 'Ware', 'Washington', 'Way', 'Webster', 'Wilcox', 'Worth']
Percent 18-24: 10.31077530305337%
Percent White: 42.48359794672628%
Percent Black: 51.7109685099099%
Percent Native American: 0.15900394987587643%
Percent Other: 2.0658484417585576%
Percent Male: 49.4800963534584%
Percent Female: 50.5199036465416%
Percent Dem: 57.17616733282734%
```

```
Cluster Number: 1
Counties: ['Appling', 'Bacon', 'Banks', 'Barrow', 'Bartow', 'Berrien', 'Brantley', 'Bryan', 'Camden', 'Candler', 'Charlton', 'Chatham', 'Clarke', 'Clayton', 'Cobb', 'Coffee', 'Colquhoun', 'Crawford', 'Cris', 'Cuthbert', 'Dade', 'DeKalb', 'Dooly', 'Dougherty', 'Duval', 'Fayette', 'Floyd', 'Franklin', 'Fulton', 'Gadsden', 'Gaston', 'Gilmer', 'Glynn', 'Graham', 'Greene', 'Hall', 'Harris', 'Henry', 'Houston', 'Iberia', 'Jefferson', 'Jones', 'Lamar', 'Liberty', 'Madison', 'Macon', 'Marion', 'McIntosh', 'Meriwether', 'Miller', 'Mitchell', 'Monroe', 'Murray', 'Newton', 'Oconee', 'Oglethorpe', 'Peach', 'Pickens', 'Polk', 'Pulaski', 'Quitman', 'Randolph', 'Richmond', 'Rockdale', 'Seminole', 'Spalding', 'Stewart', 'Sumter', 'Taliaferro', 'Telford', 'Thomas', 'Troup', 'Turner', 'Twiggs', 'Union', 'Upson', 'Walker', 'Walton', 'Ware', 'Washington', 'Way', 'Webster', 'Wilcox', 'Worth']
Percent 18-24: 8.502606707818076%
```

Counties: ['Atkinson', 'Baker', 'Ben Hill', 'Bleckley', 'Brooks', 'Bulloch', 'Butts', 'Calhoun

```
/Users/joshkupper-smith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

33

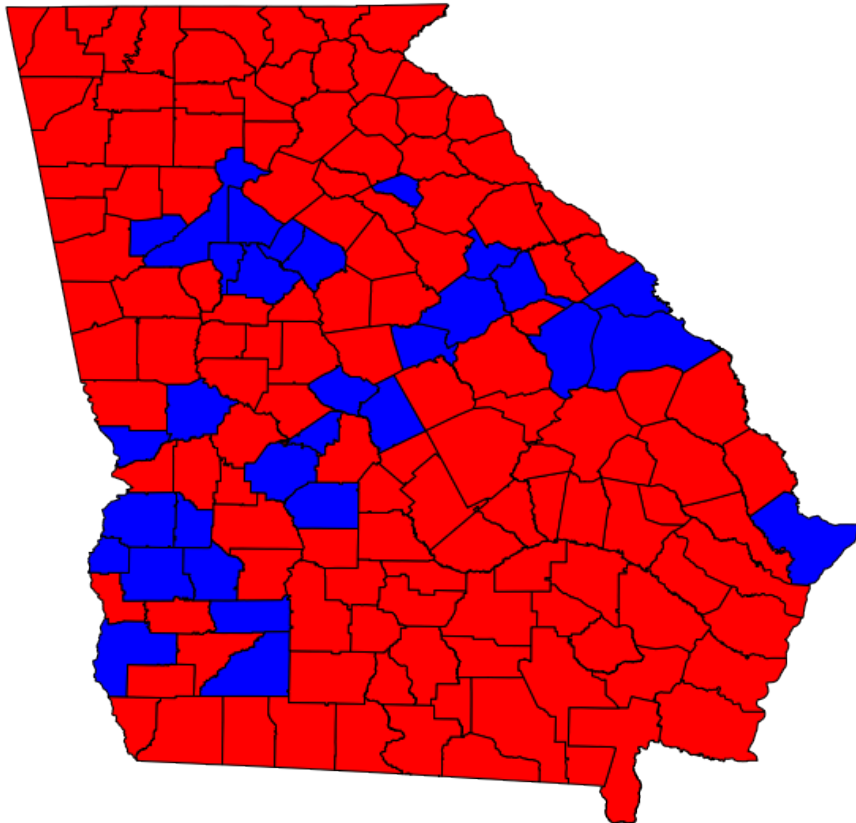
```

county['county_lower'] = county['County_x'].str.lower().apply(lambda x: strip(x))
ga_data['county_lower'] = ga_data['County_x'].str.lower().apply(lambda x: strip(x))
county = pd.merge(ga_data[['county_lower', 'binary_clusters']], county, on='county_lo
#cmap_politics_2 = matplotlib.colors.LinearSegmentedColormap.from_list("", ["blue", "r
county = gpd.GeoDataFrame(county)
county.plot(column="binary_clusters", cmap=cmap_politics_2)
plt.axis('off')
plt.show()

```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>



```
In [41]: clustering_cols_analysis = ['2012_d', '2012_r', '2016_d', '2016_r', 'percent_male_18_24']
        kmeans = KMeans(n_clusters=3, random_state=0).fit(nc_data[clustering_cols_analysis])

        labels = kmeans.labels_
        nc_data['nc_clusters'] = labels

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:5: SettingWithCopyError:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html
"""
```

```
In [42]: for label in list(set(list(nc_data['nc_clusters']))):
    data_clust = nc_data[nc_data['nc_clusters'] == label]
    individual = {}
    print("Cluster Number: " + str(label))
    print("Counties: " + str(list(data_clust['County_x'])))
    print("Percent 18-24: " + str(np.average(list(data_clust['percent_male_18_24']))+
    print("Percent White: " + str(np.average(list(data_clust['percent_white']))*100.0))
    print("Percent Black: " + str(np.average(list(data_clust['percent_black']))*100.0))
    print("Percent Native American: " + str(np.average(list(data_clust['percent_native
    print("Percent Other: " + str(np.average(list(data_clust['percent_other']))*100.0))
    print("Percent Male: " + str(np.average(list(data_clust['percent_male']))*100.0))
    print("Percent Female: " + str(np.average(list(data_clust['percent_female']))*100.0))
    print("Percent Dem: " + str(np.average(list(data_clust['2016_d']))*100.0) + "%")
    print()
    print('-----')
    print()
```

```
Cluster Number: 0
Counties: ['Alexander', 'Alleghany', 'Ashe', 'Avery', 'Brunswick', 'Caldwell', 'Camden', 'Cartersville']
Percent 18-24: 8.304250966050086%
Percent White: 86.4926336957563%
Percent Black: 7.846066385145869%
Percent Native American: 0.8393503494104979%
Percent Other: 2.0434490689251716%
Percent Male: 49.30048360411178%
Percent Female: 50.699516395888224%
Percent Dem: 28.321563270359295%
```

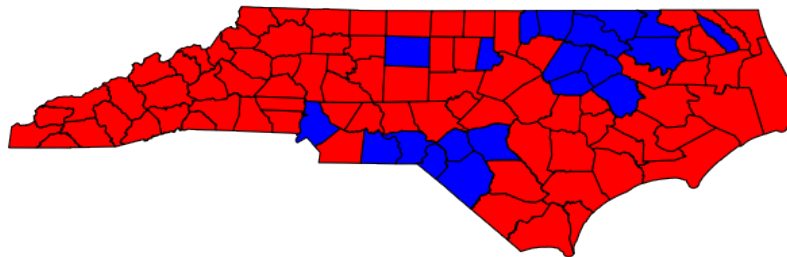
```

#print(len(fl_data))
county.head()
county.rename(columns={'NAME': 'County_x'}, inplace=True)
county['county_lower'] = county['County_x'].str.lower().apply(lambda x: strip(x))
nc_data['county_lower'] = nc_data['County_x'].str.lower().apply(lambda x: strip(x))
county = pd.merge(nc_data[['county_lower', 'binary_clusters']], county, on='county_lo
#cmap_politics_2 = matplotlib.colors.LinearSegmentedColormap.from_list("", ["blue", "r
county = gpd.GeoDataFrame(county)
county.plot(column="binary_clusters", cmap=cmap_politics_2)
plt.axis('off')
plt.show()

```

/Users/joshkupperSmith/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:8: SettingW
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.htm>



In []: