

In [1]:

```
#Add the required libraries
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.formula.api as smf
from sklearn import linear_model
import scipy.stats as st
import numpy as np
import plotly.figure_factory as ff
```

In [2]:

```
# Task 1
data_a = pd.read_csv("election_train.csv")
data_b = pd.read_csv("demographics_train.csv")

data_a_tidy = pd.pivot_table(data_a, index = ['Year', 'State', 'County', 'Office'], columns
= 'Party', values = 'Votes', aggfunc='sum').reset_index()

data_a_tidy.head()
```

Out[2]:

	Party	Year	State	County	Office	Democratic	Republican
0		2018	AZ	Apache County	US Senator	16298.0	7810.0
1		2018	AZ	Cochise County	US Senator	17383.0	26929.0
2		2018	AZ	Coconino County	US Senator	34240.0	19249.0
3		2018	AZ	Gila County	US Senator	7643.0	12180.0
4		2018	AZ	Graham County	US Senator	3368.0	6870.0

In [3]:

# Task 2

```

data_a_tidy['County'] = data_a_tidy['County'].str.replace(' County', '')
data_a_tidy['County'] = data_a_tidy['County'].str.lower()
data_b['County'] = data_b['County'].str.lower()

change_state = {'Arizona' : 'AZ', 'Connecticut' : 'CT', 'Delaware' : 'DE', 'Florida' :
'FL', 'Hawaii' : 'HI', 'Indiana' : 'IN', 'Maine' : 'ME', 'Maryland' : 'MD', 'Massachuse
tts' : 'MA', 'Michigan' : 'MI', 'Minnesota' : 'MN', 'Montana' : 'MT', 'Nebraska' : 'NE'
, 'Nevada' : 'NV', 'New Jersey' : 'NJ', 'New Mexico' : 'NM', 'New York' : 'NY', 'North
Dakota' : 'ND', 'Ohio' : 'OH', 'Pennsylvania' : 'PA', 'Rhode Island' : 'RI', 'Tennesse
e' : 'TN', 'Texas' : 'TX', 'Utah' : 'UT', 'Vermont' : 'VT', 'Virginia' : 'VA', 'Washing
ton' : 'WA', 'West Virginia' : 'WV', 'Wisconsin' : 'WI', 'Wyoming' : 'WY'}
data_b['State'] = data_b['State'].map(change_state)

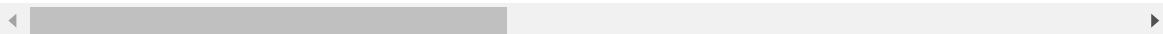
data_merge = pd.merge(data_a_tidy, data_b, how="inner", on=['County', 'State'], sort=True
)
data_merge.head()

```

Out[3]:

	Year	State	County	Office	Democratic	Republican	FIPS	Total Population	Citizen Voting- Age Population	F Hi or
0	2018	IN	adams	US Senator	3146.0	7511.0	18001	34813	0	93.
1	2018	ND	adams	US Senator	364.0	796.0	38001	2348	0	93.
2	2018	NE	adams	US Senator	3334.0	6487.0	31001	31536	0	87.
3	2018	OH	adams	US Senator	2635.0	6000.0	39001	28111	0	96.
4	2018	PA	adams	US Senator	14880.0	23419.0	42001	101759	78370	89.

5 rows × 21 columns



In [4]:

```
# Task 3
print(data_merge.info())
print(data_merge.Year.unique())
print(data_merge.Office.unique())
data_merge = data_merge.drop(columns=['Office', 'Year'])

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1200 entries, 0 to 1199
Data columns (total 21 columns):
Year                1200 non-null int64
State               1200 non-null object
County             1200 non-null object
Office             1200 non-null object
Democratic         1200 non-null float64
Republican         1200 non-null float64
FIPS               1200 non-null int64
Total Population   1200 non-null int64
Citizen Voting-Age Population 1200 non-null int64
Percent White, not Hispanic or Latino 1200 non-null float64
Percent Black, not Hispanic or Latino 1200 non-null float64
Percent Hispanic or Latino 1200 non-null float64
Percent Foreign Born 1200 non-null float64
Percent Female     1200 non-null float64
Percent Age 29 and Under 1200 non-null float64
Percent Age 65 and Older 1200 non-null float64
Median Household Income 1200 non-null int64
Percent Unemployed 1200 non-null float64
Percent Less than High School Degree 1200 non-null float64
Percent Less than Bachelor's Degree 1200 non-null float64
Percent Rural      1200 non-null float64
dtypes: float64(13), int64(5), object(3)
memory usage: 206.2+ KB
None
[2018]
['US Senator']
```

In [5]:

```
#Task 4  
print(data_merge.isnull()) # Checking to see if the values are NULL  
print(data_merge.isin([0]).sum()) # Printing out the missing values  
print(data_merge.isin([0]).sum().sum()) # Printing out all the missing values
```

	State	County	Democratic	Republican	FIPS	Total Population	\
0	False	False	False	False	False	False	
1	False	False	False	False	False	False	
2	False	False	False	False	False	False	
3	False	False	False	False	False	False	
4	False	False	False	False	False	False	
5	False	False	False	False	False	False	
6	False	False	False	False	False	False	
7	False	False	False	False	False	False	
8	False	False	False	False	False	False	
9	False	False	False	False	False	False	
10	False	False	False	False	False	False	
11	False	False	False	False	False	False	
12	False	False	False	False	False	False	
13	False	False	False	False	False	False	
14	False	False	False	False	False	False	
15	False	False	False	False	False	False	
16	False	False	False	False	False	False	
17	False	False	False	False	False	False	
18	False	False	False	False	False	False	
19	False	False	False	False	False	False	
20	False	False	False	False	False	False	
21	False	False	False	False	False	False	
22	False	False	False	False	False	False	
23	False	False	False	False	False	False	
24	False	False	False	False	False	False	
25	False	False	False	False	False	False	
26	False	False	False	False	False	False	
27	False	False	False	False	False	False	
28	False	False	False	False	False	False	
29	False	False	False	False	False	False	
...	...	...	...	...	...	...	
1170	False	False	False	False	False	False	
1171	False	False	False	False	False	False	
1172	False	False	False	False	False	False	
1173	False	False	False	False	False	False	
1174	False	False	False	False	False	False	
1175	False	False	False	False	False	False	
1176	False	False	False	False	False	False	
1177	False	False	False	False	False	False	
1178	False	False	False	False	False	False	
1179	False	False	False	False	False	False	
1180	False	False	False	False	False	False	
1181	False	False	False	False	False	False	
1182	False	False	False	False	False	False	
1183	False	False	False	False	False	False	
1184	False	False	False	False	False	False	
1185	False	False	False	False	False	False	
1186	False	False	False	False	False	False	
1187	False	False	False	False	False	False	
1188	False	False	False	False	False	False	
1189	False	False	False	False	False	False	
1190	False	False	False	False	False	False	
1191	False	False	False	False	False	False	
1192	False	False	False	False	False	False	
1193	False	False	False	False	False	False	
1194	False	False	False	False	False	False	
1195	False	False	False	False	False	False	
1196	False	False	False	False	False	False	
1197	False	False	False	False	False	False	
1198	False	False	False	False	False	False	

1199 False False False False False False

Citizen Voting-Age Population Percent White, not Hispanic or Latino

\		
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False
5	False	False
6	False	False
7	False	False
8	False	False
9	False	False
10	False	False
11	False	False
12	False	False
13	False	False
14	False	False
15	False	False
16	False	False
17	False	False
18	False	False
19	False	False
20	False	False
21	False	False
22	False	False
23	False	False
24	False	False
25	False	False
26	False	False
27	False	False
28	False	False
29	False	False
...	...	...
1170	False	False
1171	False	False
1172	False	False
1173	False	False
1174	False	False
1175	False	False
1176	False	False
1177	False	False
1178	False	False
1179	False	False
1180	False	False
1181	False	False
1182	False	False
1183	False	False
1184	False	False
1185	False	False
1186	False	False
1187	False	False
1188	False	False
1189	False	False
1190	False	False
1191	False	False
1192	False	False
1193	False	False
1194	False	False
1195	False	False

1196	False	False
1197	False	False
1198	False	False
1199	False	False

	Percent Black, not Hispanic or Latino	Percent Hispanic or Latino \
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False
5	False	False
6	False	False
7	False	False
8	False	False
9	False	False
10	False	False
11	False	False
12	False	False
13	False	False
14	False	False
15	False	False
16	False	False
17	False	False
18	False	False
19	False	False
20	False	False
21	False	False
22	False	False
23	False	False
24	False	False
25	False	False
26	False	False
27	False	False
28	False	False
29	False	False
...	...	...
1170	False	False
1171	False	False
1172	False	False
1173	False	False
1174	False	False
1175	False	False
1176	False	False
1177	False	False
1178	False	False
1179	False	False
1180	False	False
1181	False	False
1182	False	False
1183	False	False
1184	False	False
1185	False	False
1186	False	False
1187	False	False
1188	False	False
1189	False	False
1190	False	False
1191	False	False
1192	False	False
1193	False	False

1194	False	False
1195	False	False
1196	False	False
1197	False	False
1198	False	False
1199	False	False

	Percent Foreign Born	Percent Female	Percent Age 29 and Under	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	
5	False	False	False	
6	False	False	False	
7	False	False	False	
8	False	False	False	
9	False	False	False	
10	False	False	False	
11	False	False	False	
12	False	False	False	
13	False	False	False	
14	False	False	False	
15	False	False	False	
16	False	False	False	
17	False	False	False	
18	False	False	False	
19	False	False	False	
20	False	False	False	
21	False	False	False	
22	False	False	False	
23	False	False	False	
24	False	False	False	
25	False	False	False	
26	False	False	False	
27	False	False	False	
28	False	False	False	
29	False	False	False	
...	...	...	...	
1170	False	False	False	
1171	False	False	False	
1172	False	False	False	
1173	False	False	False	
1174	False	False	False	
1175	False	False	False	
1176	False	False	False	
1177	False	False	False	
1178	False	False	False	
1179	False	False	False	
1180	False	False	False	
1181	False	False	False	
1182	False	False	False	
1183	False	False	False	
1184	False	False	False	
1185	False	False	False	
1186	False	False	False	
1187	False	False	False	
1188	False	False	False	
1189	False	False	False	
1190	False	False	False	
1191	False	False	False	



1192	False	False	False
1193	False	False	False
1194	False	False	False
1195	False	False	False
1196	False	False	False
1197	False	False	False
1198	False	False	False
1199	False	False	False

	Percent Age 65 and Older	Median Household Income	Percent Unemploye
d \			
0	False	False	Fals
e			
1	False	False	Fals
e			
2	False	False	Fals
e			
3	False	False	Fals
e			
4	False	False	Fals
e			
5	False	False	Fals
e			
6	False	False	Fals
e			
7	False	False	Fals
e			
8	False	False	Fals
e			
9	False	False	Fals
e			
10	False	False	Fals
e			
11	False	False	Fals
e			
12	False	False	Fals
e			
13	False	False	Fals
e			
14	False	False	Fals
e			
15	False	False	Fals
e			
16	False	False	Fals
e			
17	False	False	Fals
e			
18	False	False	Fals
e			
19	False	False	Fals
e			
20	False	False	Fals
e			
21	False	False	Fals
e			
22	False	False	Fals
e			
23	False	False	Fals
e			
24	False	False	Fals
e			

25 e	False	False	Fals
26 e	False	False	Fals
27 e	False	False	Fals
28 e	False	False	Fals
29 e	False	False	Fals
...	...	...	
...			
1170 e	False	False	Fals
1171 e	False	False	Fals
1172 e	False	False	Fals
1173 e	False	False	Fals
1174 e	False	False	Fals
1175 e	False	False	Fals
1176 e	False	False	Fals
1177 e	False	False	Fals
1178 e	False	False	Fals
1179 e	False	False	Fals
1180 e	False	False	Fals
1181 e	False	False	Fals
1182 e	False	False	Fals
1183 e	False	False	Fals
1184 e	False	False	Fals
1185 e	False	False	Fals
1186 e	False	False	Fals
1187 e	False	False	Fals
1188 e	False	False	Fals
1189 e	False	False	Fals
1190 e	False	False	Fals
1191 e	False	False	Fals
1192 e	False	False	Fals
1193 e	False	False	Fals
1194	False	False	Fals

e			
1195	False	False	Fals
e			
1196	False	False	Fals
e			
1197	False	False	Fals
e			
1198	False	False	Fals
e			
1199	False	False	Fals
e			

Percent Less than High School Degree \

0	False
1	False
2	False
3	False
4	False
5	False
6	False
7	False
8	False
9	False
10	False
11	False
12	False
13	False
14	False
15	False
16	False
17	False
18	False
19	False
20	False
21	False
22	False
23	False
24	False
25	False
26	False
27	False
28	False
29	False
...	...
1170	False
1171	False
1172	False
1173	False
1174	False
1175	False
1176	False
1177	False
1178	False
1179	False
1180	False
1181	False
1182	False
1183	False
1184	False
1185	False
1186	False

1187	False
1188	False
1189	False
1190	False
1191	False
1192	False
1193	False
1194	False
1195	False
1196	False
1197	False
1198	False
1199	False

	Percent Less than Bachelor's Degree	Percent Rural
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False
5	False	False
6	False	False
7	False	False
8	False	False
9	False	False
10	False	False
11	False	False
12	False	False
13	False	False
14	False	False
15	False	False
16	False	False
17	False	False
18	False	False
19	False	False
20	False	False
21	False	False
22	False	False
23	False	False
24	False	False
25	False	False
26	False	False
27	False	False
28	False	False
29	False	False
...	...	...
1170	False	False
1171	False	False
1172	False	False
1173	False	False
1174	False	False
1175	False	False
1176	False	False
1177	False	False
1178	False	False
1179	False	False
1180	False	False
1181	False	False
1182	False	False
1183	False	False
1184	False	False

1185	False	False
1186	False	False
1187	False	False
1188	False	False
1189	False	False
1190	False	False
1191	False	False
1192	False	False
1193	False	False
1194	False	False
1195	False	False
1196	False	False
1197	False	False
1198	False	False
1199	False	False

[1200 rows x 19 columns]

State	0
County	0
Democratic	5
Republican	5
FIPS	0
Total Population	0
Citizen Voting-Age Population	680
Percent White, not Hispanic or Latino	0
Percent Black, not Hispanic or Latino	45
Percent Hispanic or Latino	5
Percent Foreign Born	3
Percent Female	0
Percent Age 29 and Under	0
Percent Age 65 and Older	0
Median Household Income	0
Percent Unemployed	3
Percent Less than High School Degree	0
Percent Less than Bachelor's Degree	0
Percent Rural	19

dtype: int64

765



In [6]:

```
#Task 4-2
```

```
# Dropping the column due to a lot of missing values
```

```
data_merge = data_merge.drop(columns='Citizen Voting-Age Population')
```

```
# Replaces all the 0s with NaN
```

```
data_merge = data_merge.replace(0, np.nan)
```

```
data_merge = data_merge.interpolate(method='linear')
```

```
# Printing out to show that they are no more missing values
```

```
print(data_merge.isin([0]).sum())
```

```
State          0
County         0
Democratic     0
Republican     0
FIPS           0
Total Population 0
Percent White, not Hispanic or Latino 0
Percent Black, not Hispanic or Latino 0
Percent Hispanic or Latino 0
Percent Foreign Born 0
Percent Female 0
Percent Age 29 and Under 0
Percent Age 65 and Older 0
Median Household Income 0
Percent Unemployed 0
Percent Less than High School Degree 0
Percent Less than Bachelor's Degree 0
Percent Rural 0
dtype: int64
```

In [7]:

```
#Task 5
def func(row) :
    if row['Democratic'] > row['Republican']:
        val = 1
    else:
        val = 0
    return val

data_merge['Party'] = data_merge.apply(func, axis = 1)
print(data_merge)
```

	State	County	Democratic	Republican	FIPS	\
0	IN	adams	3146.0	7511.0	18001	
1	ND	adams	364.0	796.0	38001	
2	NE	adams	3334.0	6487.0	31001	
3	OH	adams	2635.0	6000.0	39001	
4	PA	adams	14880.0	23419.0	42001	
5	WA	adams	1365.0	2867.0	53001	
6	WI	adams	4537.0	4854.0	55001	
7	VT	addison	11964.0	4439.0	50001	
8	MN	aitkin	4118.0	3808.0	27001	
9	FL	alachua	74493.0	40599.0	12001	
10	WY	albany	7576.0	6366.0	56001	
11	VA	albemarle	35701.0	16371.0	51003	
12	MI	alcona	1915.0	3541.0	26001	
13	VA	alexandria city	53307.0	10734.0	51510	
14	MI	alger	1911.0	2162.0	26003	
15	NY	allegany	4844.0	8305.0	36003	
16	VA	alleghany	1952.0	3433.0	51005	
17	PA	allegheny	355907.0	176351.0	42003	
18	IN	allen	55903.0	65927.0	18003	
19	MI	alpena	5412.0	7380.0	26007	
20	VA	amelia	1938.0	3823.0	51007	
21	VA	amherst	4541.0	7779.0	51009	
22	TX	anderson	3307.0	11335.0	48001	
23	ME	androscoggin	22150.0	18931.0	23001	
24	TX	angelina	7130.0	19166.0	48005	
25	MD	anne arundel	122910.0	92401.0	24003	
26	MN	anoka	87756.0	65707.0	27003	
27	NE	antelope	396.0	2061.0	31003	
28	MI	antrim	4953.0	7629.0	26009	
29	AZ	apache	16298.0	7810.0	4001	
...	...	...	...	...	...	
1170	MN	wilkin	1251.0	1243.0	27167	
1171	OH	williams	5272.0	7927.0	39171	
1172	VA	williamsburg city	4530.0	1547.0	51830	
1173	TN	williamson	42611.0	62039.0	47187	
1174	TN	wilson	19447.0	32810.0	47189	
1175	TX	wilson	4567.0	13025.0	48493	
1176	VA	winchester city	5242.0	3657.0	51840	
1177	CT	windham	20490.0	19032.0	9015	
1178	VT	windham	14386.0	3673.0	50025	
1179	VT	windsor	17354.0	6402.0	50027	
1180	TX	winkler	321.0	1123.0	48495	
1181	WI	winnebago	40185.0	35282.0	55139	
1182	VA	wise	2860.0	7991.0	51195	
1183	TX	wood	2635.0	13987.0	48499	
1184	WI	wood	15992.0	16899.0	55141	
1185	WV	wood	14189.0	13696.0	54107	
1186	MD	worcester	9840.0	12886.0	24047	
1187	MN	wright	26821.0	30572.0	27171	
1188	NY	wyoming	3766.0	7553.0	36121	
1189	PA	wyoming	3868.0	6582.0	42131	
1190	WV	wyoming	2607.0	3096.0	54109	
1191	VA	wythe	2879.0	7669.0	51197	
1192	WA	yakima	29476.0	40958.0	53077	
1193	AZ	yavapai	40160.0	65308.0	4025	
1194	TX	yoakum	335.0	1558.0	48501	
1195	ME	york	51387.0	32849.0	23031	
1196	NE	york	1281.0	3659.0	31185	
1197	PA	york	69272.0	95814.0	42133	
1198	TX	young	821.0	5543.0	48503	



1199	TX	zapata	1392.0	821.0	48505
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	Total Population	Percent White, not Hispanic or Latino	\
0	34813	93.740844	
1	2348	93.100511	
2	31536	87.338280	
3	28111	96.673900	
4	101759	89.762085	
5	19100	35.670157	
6	20294	90.637627	
7	36926	93.107837	
8	15722	94.122885	
9	256581	62.460198	
10	37836	83.269902	
11	104287	77.424799	
12	10461	96.138037	
13	151473	52.036337	
14	9396	84.514687	
15	47700	94.582809	
16	15919	91.940449	
17	1230360	79.368234	
18	365565	74.923748	
19	28929	96.121539	
20	12793	72.203549	
21	31999	75.611738	
22	57772	60.029080	
23	107376	91.319289	
24	87657	61.672200	
25	559737	70.303375	
26	341249	83.336508	
27	6421	95.421274	
28	23215	95.179841	
29	72346	18.571863	
...	...	...	
1170	6479	94.320111	
1171	37270	93.273410	
1172	14988	68.714972	
1173	205645	85.651973	
1174	125616	86.270857	
1175	46444	58.087159	
1176	27349	67.629529	
1177	117078	83.782606	
1178	43609	93.687083	
1179	55894	95.128279	
1180	7723	37.822090	
1181	169487	89.548461	
1182	40074	91.944902	
1183	43198	83.499236	
1184	73621	93.080779	
1185	86262	95.478890	
1186	51441	80.047044	
1187	129922	93.106633	
1188	41239	89.885788	
1189	27975	95.874888	
1190	22537	97.475263	
1191	29171	94.052312	
1192	247681	44.996588	
1193	218586	81.159361	
1194	8316	34.271284	
1195	200536	94.761539	
1196	13842	91.619708	

1197	440604	84.725513
1198	18275	79.102599
1199	14335	5.866760

	Percent Black, not Hispanic or Latino	Percent Hispanic or Latino \
0	0.709505	4.403527
1	0.894378	0.851789
2	0.821284	8.907281
3	0.355733	0.522927
4	1.352215	6.604821
5	0.308901	61.884817
6	3.129004	3.813935
7	0.974923	2.071711
8	0.368910	1.183056
9	19.504562	8.994820
10	1.345280	9.229305
11	9.319474	5.592260
12	0.248542	1.386101
13	21.068441	16.741598
14	7.322265	1.351639
15	1.419287	1.551363
16	5.207614	1.432251
17	12.825027	1.890097
18	11.330953	7.060304
19	0.508141	1.296277
20	23.528492	1.188150
21	17.844308	2.184443
22	21.197120	16.961504
23	1.551557	1.733162
24	14.582977	21.110693
25	15.559093	7.039377
26	5.091297	4.057741
27	0.264756	3.145927
28	0.323067	1.955632
29	0.486551	5.947806
...	...	...
1170	0.725421	1.265627
1171	1.078616	4.043467
1172	14.738457	7.412597
1173	4.229619	4.625933
1174	6.652019	3.677875
1175	1.332788	38.754199
1176	9.514059	16.486892
1177	1.866277	10.877364
1178	0.807173	2.086725
1179	0.713851	1.452750
1180	2.563771	58.474686
1181	1.996613	3.841003
1182	5.140490	1.230224
1183	5.037270	9.400898
1184	0.709037	2.665000
1185	1.123322	1.032900
1186	13.603935	3.355300
1187	1.139145	2.695463
1188	5.160164	3.205703
1189	0.868633	1.780161
1190	0.678884	0.257355
1191	3.383497	1.100408
1192	0.740872	47.678667
1193	0.518331	14.054880
1194	0.679040	63.552189

1195	0.839749	1.556828
1196	0.809132	4.681404
1197	5.254605	6.649735
1198	1.258550	17.668947
1199	0.013952	94.119288

	Percent Foreign Born	Percent Female	Percent Age 29 and Under \
0	0.901962	50.389222	45.313532
1	1.916525	49.446337	32.282794
2	5.717275	49.242136	41.365424
3	0.640319	50.297037	37.319910
4	3.978027	50.636307	35.814031
5	23.801047	49.130890	51.513089
6	2.774219	46.018528	25.091160
7	4.387153	50.408926	36.359205
8	0.985880	49.014120	26.039944
9	10.144555	51.670623	48.551530
10	6.081510	47.756105	55.447193
11	9.917823	52.107166	39.463212
12	1.720677	49.555492	21.202562
13	27.777888	51.770943	34.734243
14	0.925926	45.008514	28.522776
15	2.301887	49.425577	41.396226
16	1.300333	51.077329	31.660280
17	5.620225	51.800367	36.392438
18	6.367404	51.219619	42.487109
19	1.161464	50.713817	31.684469
20	0.242320	48.080982	34.565778
21	1.712554	52.042251	34.785462
22	6.536038	38.757530	34.862909
23	2.986701	51.044926	36.995232
24	8.041571	51.213252	41.628164
25	8.270491	50.540700	38.635109
26	7.303758	49.959707	38.896524
27	2.071328	49.898770	34.605202
28	2.015938	50.273530	29.450786
29	1.719515	50.598513	45.854643
...	...	...	...
1170	0.478469	49.714462	34.526933
1171	1.333512	50.531258	36.576335
1172	9.067254	56.418468	58.853750
1173	6.826813	51.159522	39.357145
1174	4.134028	51.183766	37.068526
1175	4.484971	49.812678	38.215916
1176	11.645764	51.168233	41.602984
1177	5.269991	50.422795	37.389604
1178	3.107157	50.883992	32.426793
1179	3.594304	50.976849	30.704548
1180	13.712288	48.737537	47.028357
1181	3.318839	49.604394	39.787122
1182	1.170335	48.330588	37.363378
1183	3.433029	50.407426	31.811658
1184	2.326782	50.712433	34.645006
1185	1.001600	51.756277	34.726763
1186	4.593612	51.402578	30.036352
1187	2.880959	49.683656	41.620357
1188	1.804117	45.830403	35.194840
1189	1.276139	49.773012	35.145666
1190	0.257355	50.547988	33.877623
1191	0.901580	51.204964	32.508313
1192	18.119274	49.976381	46.587748

1193	6.456955	51.092476	28.717301
1194	25.637326	49.158249	46.284271
1195	3.068277	51.174851	33.230442
1196	1.719405	50.881376	38.289265
1197	3.599150	50.565133	36.839883
1198	5.389877	50.768810	36.990424
1199	29.180328	49.877921	50.931287

	Percent Age 65 and Older	Median Household Income	Percent Unemploye
d \			
0	14.457243	47572	5.46263
2			
1	22.274276	53295	1.97784
8			
2	16.504947	51201	5.38038
9			
3	16.395717	34709	9.90799
7			
4	18.255879	61927	5.52495
5			
5	10.376963	47554	6.06894
9			
6	27.007983	43554	9.48949
7			
7	16.974490	61020	4.96146
4			
8	30.021626	44524	6.98638
2			
9	12.430772	44702	7.02015
4			
10	9.549107	43043	4.57917
4			
11	16.454592	70342	3.24457
5			
12	34.470892	38160	10.29729
7			
13	10.017627	89200	3.98333
8			
14	23.446147	41270	7.61904
8			
15	16.792453	44085	8.08327
7			
16	23.902255	45538	4.56098
6			
17	17.403687	54357	6.40696
1			
18	13.112032	49574	7.04808
0			
19	21.625359	39832	7.84705
3			
20	18.166185	58269	4.03071
0			
21	18.866215	47002	5.52695
9			
22	13.707332	42146	4.34379
0			
23	15.848048	48728	5.81262
6			
24	15.066680	44185	8.31150
6			
25	13.405046	91918	5.53196

7			
26	11.835346	73579	4.88194
9			
27	22.192805	46381	1.76361
4			
28	25.410295	48825	7.81578
4			
29	13.322091	32460	15.80743
3			
...	...	...	
...			
1170	18.598549	52963	1.97523
6			
1171	17.480547	45044	5.79099
9			
1172	15.158794	50091	9.07999
4			
1173	11.547570	100140	3.74988
2			
1174	14.487008	63426	5.37716
1			
1175	14.555163	68737	6.20907
7			
1176	15.134009	46466	5.12364
9			
1177	14.571482	60689	8.37552
4			
1178	19.172648	50917	6.04568
2			
1179	20.707053	54763	4.54163
4			
1180	11.070827	54261	4.13873
5			
1181	14.738003	53501	4.42067
8			
1182	15.768329	36352	10.37265
5			
1183	26.297514	47814	9.31908
0			
1184	18.608821	49926	5.66963
9			
1185	18.559737	43944	7.62545
8			
1186	25.475788	57227	7.71983
0			
1187	11.175936	75705	4.38956
7			
1188	15.512015	53612	7.13675
9			
1189	18.373548	53397	7.03230
9			
1190	17.846208	35469	12.07795
8			
1191	19.649652	42888	7.54525
0			
1192	12.748253	45700	8.34621
1			
1193	28.272625	46638	8.52598
6			
1194	11.423761	56655	9.13654
6			

1195	17.911996	59132	5.11030
3			
1196	18.841208	55156	1.43207
9			
1197	15.841663	59853	6.24087
4			
1198	19.387141	46978	4.04804
3			
1199	11.726543	36976	12.63616
6			

## Percent Less than High School Degree \

0	15.757807
1	4.042806
2	9.088224
3	21.768031
4	12.207748
5	33.934880
6	12.695857
7	7.484449
8	8.652110
9	7.468059
10	4.210167
11	8.612515
12	11.412535
13	8.620480
14	10.614217
15	10.952741
16	15.537543
17	6.132273
18	10.798136
19	9.607565
20	18.747199
21	17.252723
22	19.827442
23	10.560153
24	20.102557
25	8.088706
26	6.524509
27	9.038290
28	9.311741
29	21.758252
...	...
1170	7.664076
1171	11.389654
1172	5.315660
1173	4.418839
1174	10.211165
1175	15.086262
1176	16.514624
1177	11.925239
1178	8.547223
1179	7.552595
1180	31.932409
1181	7.893714
1182	24.758635
1183	14.869281
1184	7.658332
1185	10.370080
1186	10.366169
1187	5.768514

1188	11.891069
1189	8.353634
1190	22.638690
1191	17.787306
1192	27.519935
1193	9.830672
1194	28.571429
1195	7.620950
1196	5.744431
1197	11.463427
1198	18.350979
1199	49.673777

	Percent Less than Bachelor's Degree	Percent Rural	Party
0	84.390221	53.674354	0
1	72.473246	100.000000	0
2	77.436983	22.484377	0
3	89.113368	89.026270	0
4	78.133265	53.675782	0
5	86.212165	40.159120	0
6	87.335109	100.000000	0
7	64.225999	78.430787	1
8	84.000985	100.000000	1
9	58.485040	21.193437	1
10	52.136794	11.939723	1
11	48.466284	45.022734	1
12	85.547240	98.921587	0
13	37.911181	83.983239	1
14	81.904762	69.044891	0
15	79.519853	78.739836	0
16	83.711604	52.393846	0
17	60.930419	2.486292	1
18	72.630294	11.879695	0
19	83.224586	51.827826	0
20	85.925594	100.000000	0
21	81.120249	63.663339	0
22	88.231639	67.063533	0
23	79.296169	43.366883	1
24	84.339812	43.082366	0
25	60.634981	5.304135	1
26	71.774386	13.500623	1
27	81.433660	100.000000	0
28	73.314054	100.000000	0
29	88.941063	74.061076	1
...	...	...	...
1170	78.670121	50.136861	1
1171	85.800768	63.628394	0
1172	43.290517	41.507289	1
1173	43.419470	19.386184	0
1174	71.139261	38.467274	0
1175	80.703618	85.861410	0
1176	68.728921	67.811223	1
1177	76.143316	49.761036	1
1178	64.734860	68.245232	1
1179	64.822447	75.588495	1
1180	89.688042	18.129395	0
1181	73.113141	13.431021	1
1182	85.575341	56.670366	0
1183	82.862361	74.199314	0
1184	79.617467	36.682765	0
1185	79.738786	26.774461	1

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1186	70.064222	35.526878	0
1187	73.239335	32.507618	0
1188	85.003383	64.082552	0
1189	81.399716	83.455934	0
1190	91.787560	88.884687	0
1191	83.409509	75.330939	0
1192	84.726624	23.519206	0
1193	74.458362	33.197178	0
1194	82.183908	37.327072	0
1195	69.934920	56.843419	1
1196	71.650858	44.141969	0
1197	76.884656	24.720442	0
1198	81.408406	33.568733	0
1199	91.681305	23.534028	1

[1200 rows x 19 columns]



In [8]:

```
#Task 6
data_merge.groupby('Party', as_index = False)['Total Population'].mean()
democratic = data_merge[data_merge['Party'] == 1]
republican = data_merge[data_merge['Party'] == 0]

print(democratic.mean())
print(republican.mean())

#required measures
[statistic, pvalue] = st.ttest_ind(democratic['Total Population'],republican['Total Pop
ulation'], equal_var= False)

print([statistic,pvalue])
statistic, pvalue/2
```

Democratic	71193.172308
Republican	41322.861538
FIPS	37130.873846
Total Population	300998.316923
Percent White, not Hispanic or Latino	69.683766
Percent Black, not Hispanic or Latino	9.269702
Percent Hispanic or Latino	12.587391
Percent Foreign Born	7.986330
Percent Female	50.385433
Percent Age 29 and Under	38.726959
Percent Age 65 and Older	16.194826
Median Household Income	53798.732308
Percent Unemployed	6.908426
Percent Less than High School Degree	11.883760
Percent Less than Bachelor's Degree	71.968225
Percent Rural	39.118070
Party	1.000000
dtype: float64	
Democratic	7969.218286
Republican	12685.142286
FIPS	38755.305143
Total Population	53974.214857
Percent White, not Hispanic or Latino	82.597026
Percent Black, not Hispanic or Latino	4.454970
Percent Hispanic or Latino	9.847969
Percent Foreign Born	4.017041
Percent Female	49.617156
Percent Age 29 and Under	36.020984
Percent Age 65 and Older	18.814997
Median Household Income	48724.150857
Percent Unemployed	6.425490
Percent Less than High School Degree	14.029195
Percent Less than Bachelor's Degree	81.103128
Percent Rural	63.431458
Party	0.000000
dtype: float64	
[8.001207114045041, 2.0965719353509958e-14]	

Out[8]:

(8.001207114045041, 1.0482859676754979e-14)

In [9]:

#Task 7

```
[statistic, pvalue] = st.ttest_ind(democratic['Median Household Income'],republican['Median Household Income'], equal_var= False)
```

```
print(democratic.mean())
```

```
print(republican.mean())
```

```
print([statistic,pvalue])
```

```
statistic, pvalue/2
```

Democratic	71193.172308
Republican	41322.861538
FIPS	37130.873846
Total Population	300998.316923
Percent White, not Hispanic or Latino	69.683766
Percent Black, not Hispanic or Latino	9.269702
Percent Hispanic or Latino	12.587391
Percent Foreign Born	7.986330
Percent Female	50.385433
Percent Age 29 and Under	38.726959
Percent Age 65 and Older	16.194826
Median Household Income	53798.732308
Percent Unemployed	6.908426
Percent Less than High School Degree	11.883760
Percent Less than Bachelor's Degree	71.968225
Percent Rural	39.118070
Party	1.000000
dtype: float64	
Democratic	7969.218286
Republican	12685.142286
FIPS	38755.305143
Total Population	53974.214857
Percent White, not Hispanic or Latino	82.597026
Percent Black, not Hispanic or Latino	4.454970
Percent Hispanic or Latino	9.847969
Percent Foreign Born	4.017041
Percent Female	49.617156
Percent Age 29 and Under	36.020984
Percent Age 65 and Older	18.814997
Median Household Income	48724.150857
Percent Unemployed	6.425490
Percent Less than High School Degree	14.029195
Percent Less than Bachelor's Degree	81.103128
Percent Rural	63.431458
Party	0.000000
dtype: float64	
[5.507012409466501, 6.173239891230373e-08]	

Out[9]:

```
(5.507012409466501, 3.0866199456151866e-08)
```

In [10]:

```
#Task 8-1
data_merge['Party'].value_counts(normalize=True)
```

Out[10]:

```
0    0.729167
1    0.270833
Name: Party, dtype: float64
```

In [11]:

```
#Task 8-2
#Printing out the averages for each party
print(data_merge[data_merge['Party']==0].mean()) #Republican
print(" ")
print(data_merge[data_merge['Party']==1].mean()) #Democratic
```

```
Democratic                7969.218286
Republican                12685.142286
FIPS                      38755.305143
Total Population          53974.214857
Percent White, not Hispanic or Latino    82.597026
Percent Black, not Hispanic or Latino    4.454970
Percent Hispanic or Latino               9.847969
Percent Foreign Born                   4.017041
Percent Female                       49.617156
Percent Age 29 and Under               36.020984
Percent Age 65 and Older               18.814997
Median Household Income             48724.150857
Percent Unemployed                   6.425490
Percent Less than High School Degree    14.029195
Percent Less than Bachelor's Degree     81.103128
Percent Rural                       63.431458
Party                               0.000000
dtype: float64
```

```
Democratic                71193.172308
Republican                41322.861538
FIPS                      37130.873846
Total Population          300998.316923
Percent White, not Hispanic or Latino    69.683766
Percent Black, not Hispanic or Latino    9.269702
Percent Hispanic or Latino               12.587391
Percent Foreign Born                   7.986330
Percent Female                       50.385433
Percent Age 29 and Under               38.726959
Percent Age 65 and Older               16.194826
Median Household Income             53798.732308
Percent Unemployed                   6.908426
Percent Less than High School Degree    11.883760
Percent Less than Bachelor's Degree     71.968225
Percent Rural                       39.118070
Party                               1.000000
dtype: float64
```

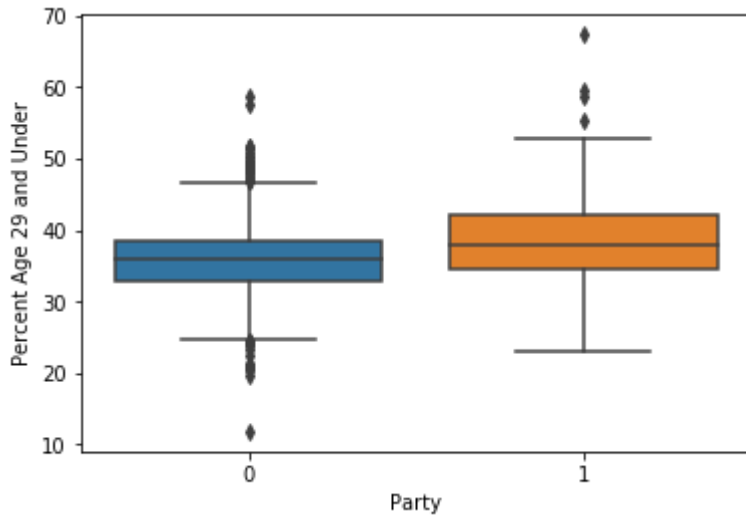
In [12]:

```
#Task 8-3
```

```
sns.boxplot(x='Party', y='Percent Age 29 and Under', data=data_merge)
```

Out[12]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24bfffef60>



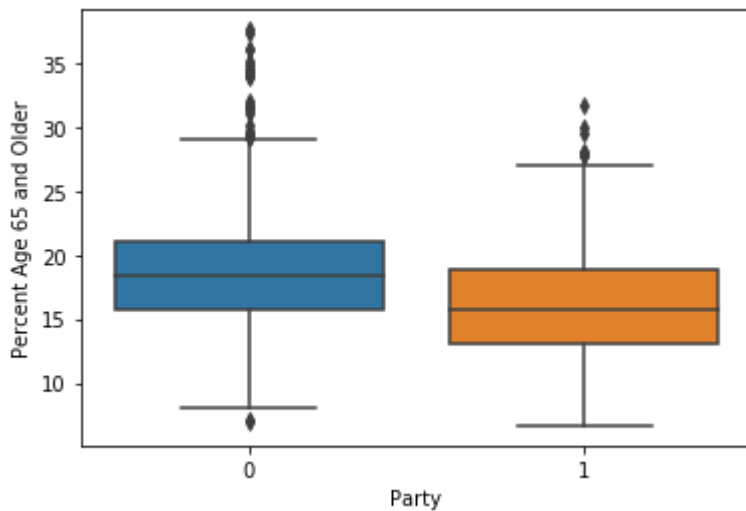
In [13]:

```
#Task 8-4
```

```
sns.boxplot(x='Party', y='Percent Age 65 and Older', data=data_merge)
```

Out[13]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c0e6d68>



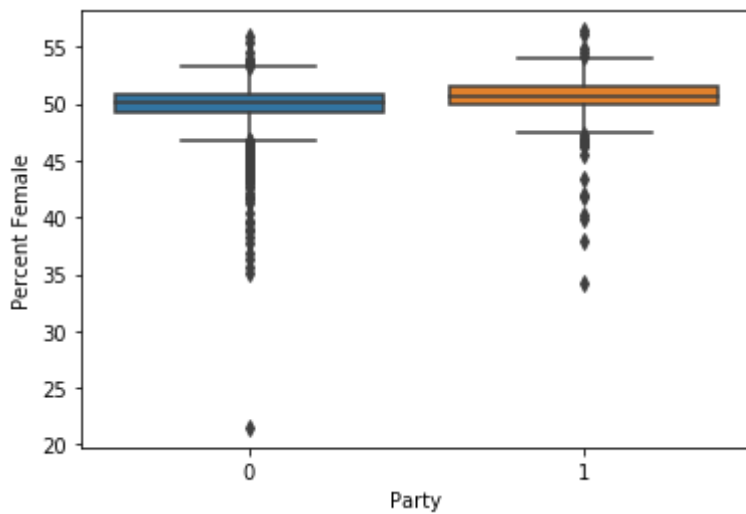
In [14]:

```
#Task 8-5
```

```
sns.boxplot(x='Party', y='Percent Female', data=data_merge)
```

Out[14]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c170e10>



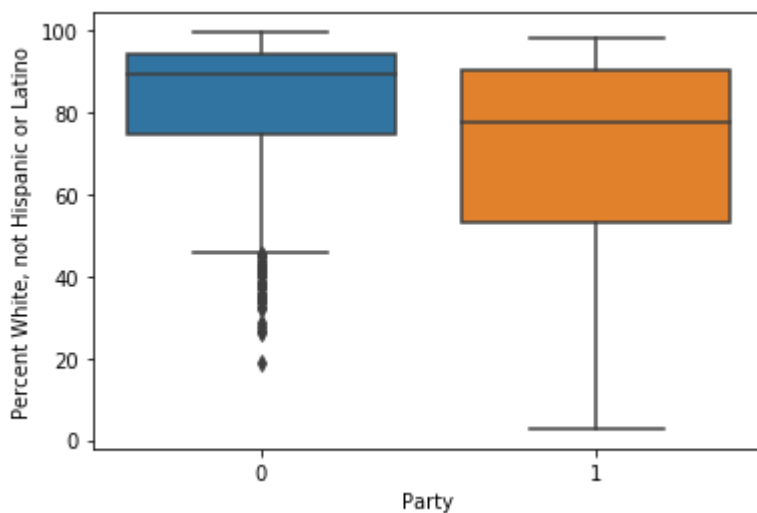
In [15]:

```
#Task 8-6
```

```
sns.boxplot(x='Party', y='Percent White, not Hispanic or Latino', data=data_merge)
```

Out[15]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c1e1828>



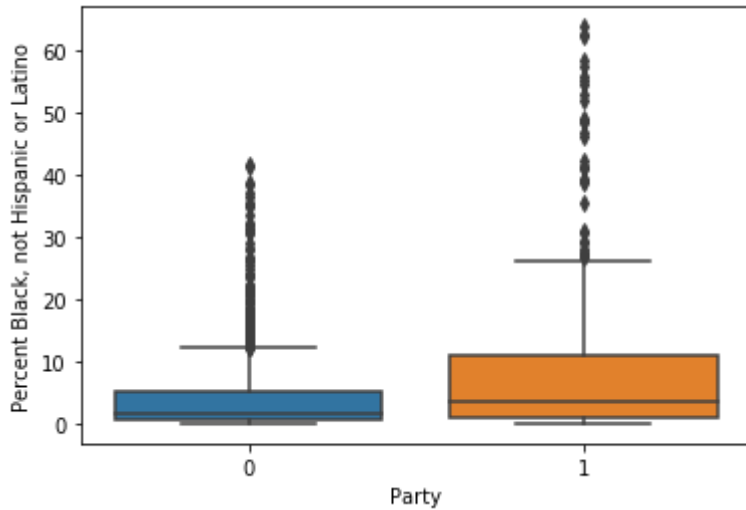
In [16]:

```
#Task 8-7
```

```
sns.boxplot(x='Party', y='Percent Black, not Hispanic or Latino', data=data_merge)
```

Out[16]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c2649b0>



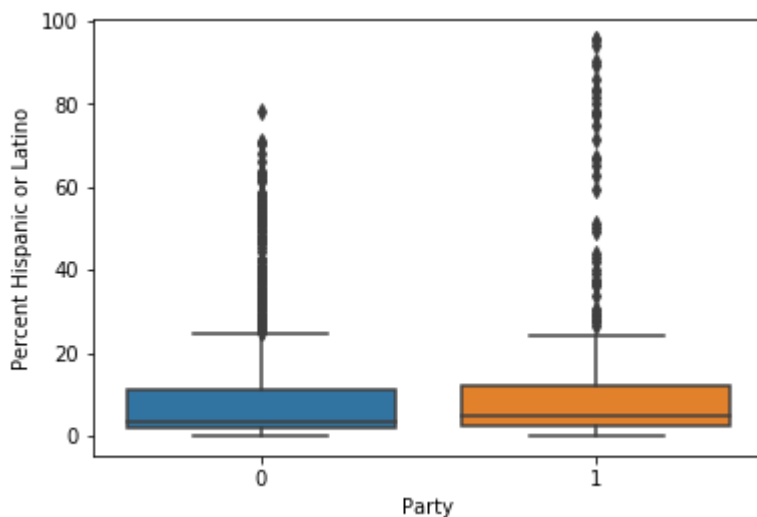
In [17]:

```
#Task 8-8
```

```
sns.boxplot(x='Party', y='Percent Hispanic or Latino', data=data_merge)
```

Out[17]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c2be4e0>



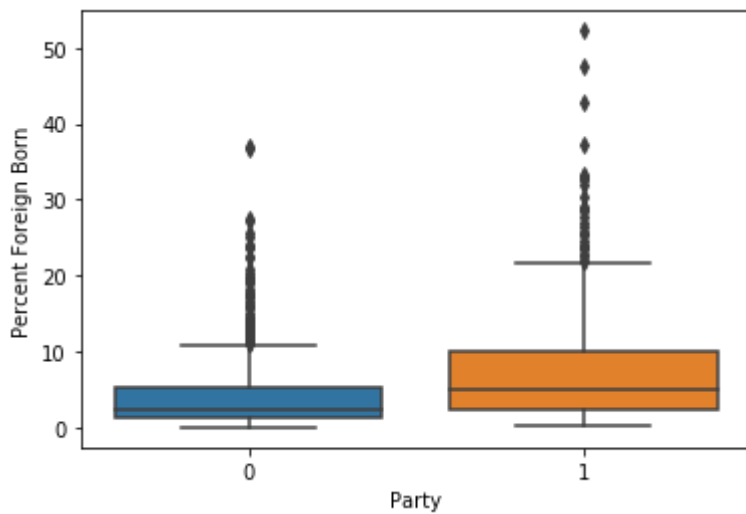
In [18]:

```
#Task 8-9
```

```
sns.boxplot(x='Party', y='Percent Foreign Born', data=data_merge)
```

Out[18]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c351630>



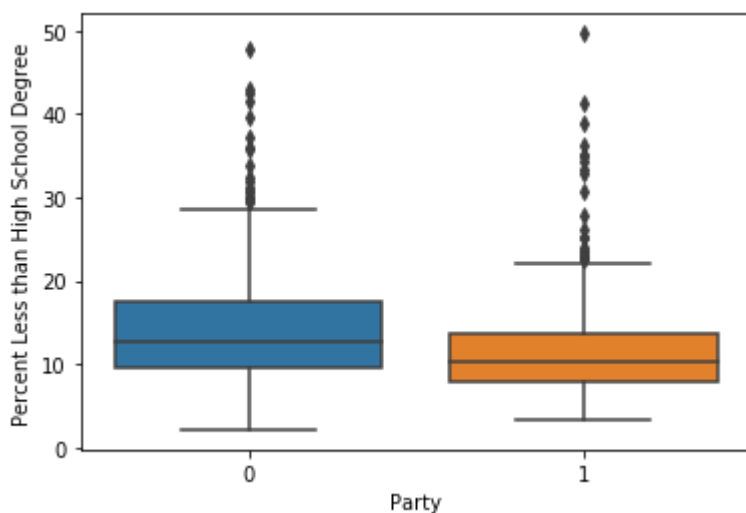
In [19]:

```
#Task 8-10
```

```
sns.boxplot(x='Party', y='Percent Less than High School Degree', data=data_merge)
```

Out[19]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c3cd278>



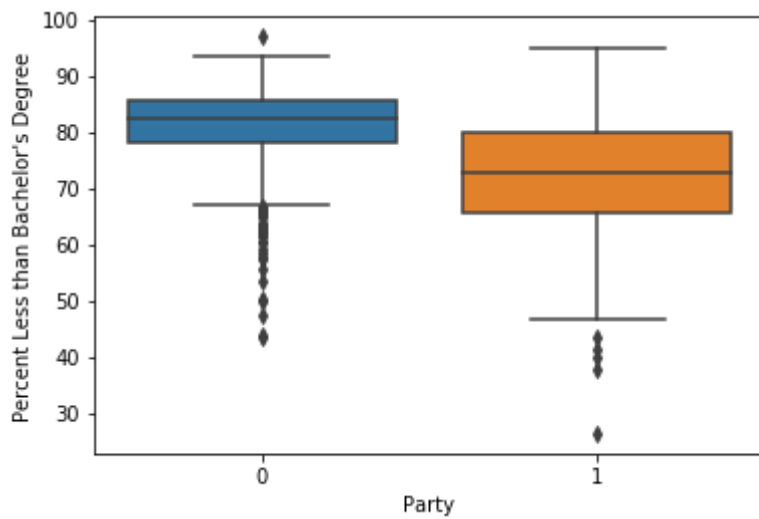
In [20]:

```
#Task 8-11
```

```
sns.boxplot(x='Party', y="Percent Less than Bachelor's Degree", data=data_merge)
```

Out[20]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1b24c435400>



In [21]:

```
#Task 9
```



In [22]:

```
#Task 10-1 Democratic Counties
```

```
fips = democratic['FIPS']
```

```
values = range(len(fips))
```

```
fig = ff.create_choropleth(fips=fips, values=values)
```

```
fig.layout.template = None
```

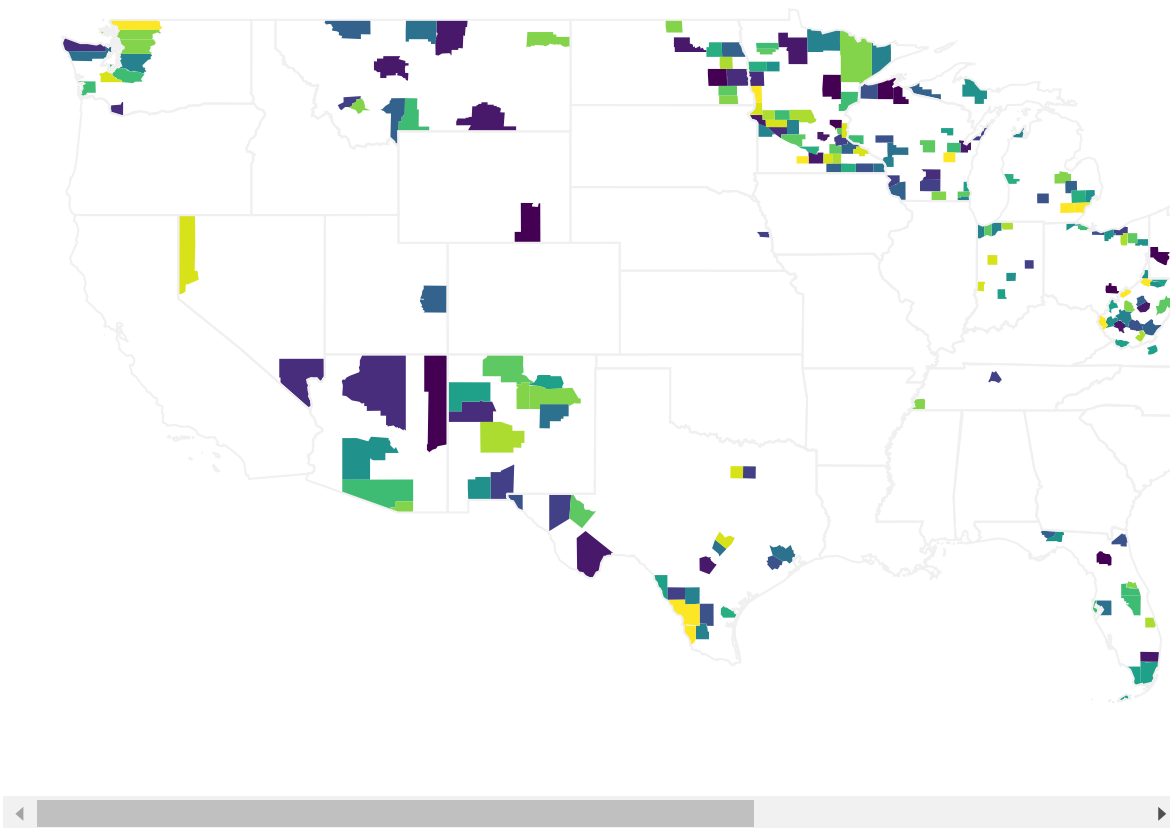
```
fig.show()
```

C:\Users\Mehul\Anaconda3\lib\site-packages\pandas\core\frame.py:6692: FutureWarning:

Sorting because non-concatenation axis is not aligned. A future version of pandas will change to not sort by default.

To accept the future behavior, pass 'sort=False'.

To retain the current behavior and silence the warning, pass 'sort=True'.



In [23]:

```
#Task 10-2-Republican Counties
```

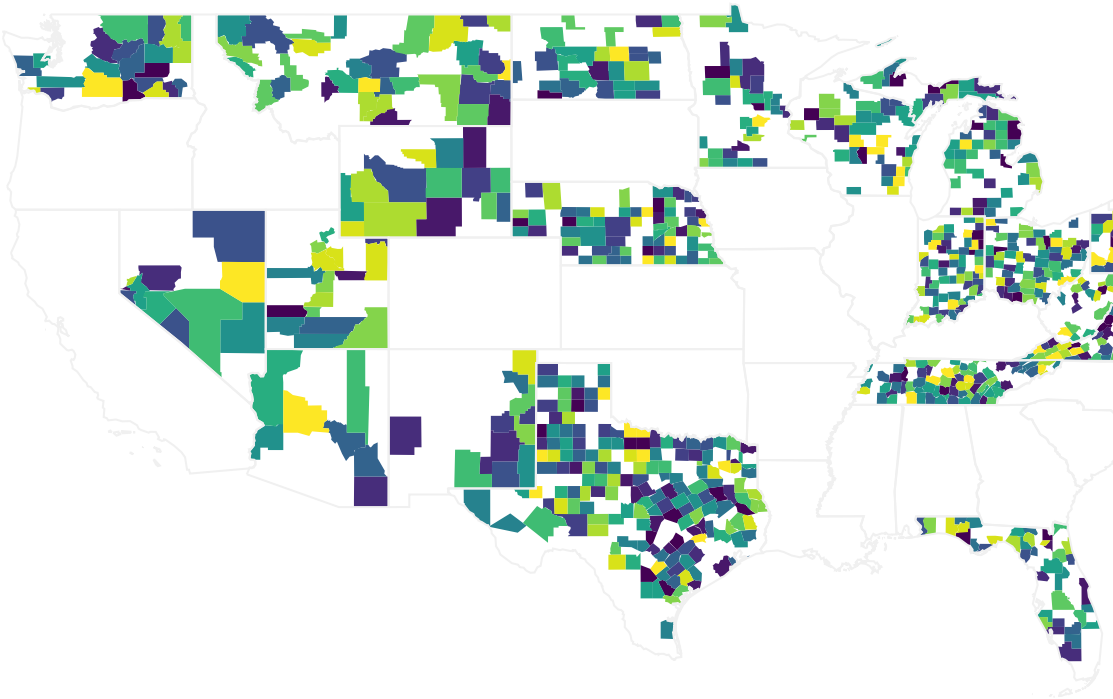
```
fips = republican['FIPS']
```

```
values = range(len(fips))
```

```
fig = ff.create_choropleth(fips=fips, values=values)
```

```
fig.layout.template = None
```

```
fig.show()
```



In [25]:

*#Task 10-3*

```
fips = data_merge['FIPS'].tolist()
values = data_merge['Party'].tolist()
```

```
colorscale = ['#f54242', '#4287f5']
```

```
fig = ff.create_choropleth(fips=fips, values=values, scope=['usa', 'HI'],
                           title_text = 'Democratic vs Republican Counties',
                           colorscale=colorscale,
                           state_outline={'color': 'rgb(0,0,0)', 'width': 1},
                           county_outline={'color': 'rgb(0,0,0)', 'width': 0.5},
                           legend_title='Party')
```

```
fig.layout.template = None
fig.show()
```

## Democratic vs Republican Counti

