```
In [1]: import pandas as pd
    import matplotlib.pyplot as plt
    import seaborn as sns
    import numpy as np

In [2]: %%javascript
    IPython.OutputArea.prototype._should_scroll = function(lines) {
        return false;
    }

In [3]: from IPython.display import set_matplotlib_formats
    set_matplotlib_formats('retina')

In [4]: pd.set_option('display.max_columns', None)
```

In [5]: top_df = pd.read_csv("/Users/adriana/Google Drive/_Learning/_DS4A/Assignments/0_Final_Project/2020_top_thir
ty.csv")
poli_df = pd.read_csv("/Users/adriana/Google Drive/_Learning/_DS4A/Assignments/0_Final_Project/top_thirty_p
oli_df.csv")
hc_df = pd.read_csv("/Users/adriana/Google Drive/_Learning/_DS4A/Assignments/0_Final_Project/top_thirty_hc_
df.csv")

In [6]: top_df.head()

Out[6]:

	region	city	state	popsize	pop_est	white_nonhi	black	asian	hisp_lat
0	northeast	New York City	New York	1M+	8336817	0.32	0.243	0.141	0.291
1	west	Los Angeles	California	1M+	3979576	0.29	0.089	0.116	0.485
2	midwest	Chicago	Illinois	1M+	2693976	0.33	0.296	0.066	0.288
3	southwest	Houston	Texas	1M+	2320268	0.24	0.226	0.068	0.450
4	west	Phoenix	Arizona	1M+	1680992	0.43	0.071	0.038	0.426

```
In [7]:
           poli_df.head()
Out[7]:
                                 city
                                                                  pop_est white_nonhi black asian hisp_lat mayor_party
                  region
                                               state
                                                                                                                             gov_party
                                                                                                                                           leg_maj
            0 southwest Albuquerque
                                         New Mexico 500k-999.999
                                                                    560513
                                                                                         0.033 0.029
                                                                                                         0.492
                                                                                   0.39
                                                                                                                   Democrat
                                                                                                                              Democrat
                                                                                                                                          Democrat
                                             Georgia 500k-999,999
                                                                    506811
                                                                                         0.510 0.044
                                                                                                         0.043
                                                                                                                   Democrat Republican
               southeast
                               Atlanta
                                                                                   0.38
                                                                                                                                          Democrat
              southwest
                               Austin
                                                     500k-999,999
                                                                    978908
                                                                                   0.48
                                                                                         0.078 0.076
                                                                                                         0.339
                                                                                                                   Democrat Republican
                                                                                                                                        Republican
                northeast
                            Baltimore
                                                     500k-999,999
                                                                                         0.624 0.026
                                                                                                         0.053
                                                                                                                   Democrat Republican
                                            Maryland
                                                                    593490
                                                                                   0.28
                                                                                                                                          Democrat
               northeast
                                      Massachusetts 500k-999,999
                                                                    692600
                                                                                   0.45 0.252 0.097
                                                                                                         0.198
                                                                                                                   Democrat Republican
                               Boston
                                                                                                                                          Democrat
In [8]:
           hc_df.head()
Out[8]:
                  region
                                  city
                                           state popsize
                                                          pop est white nonhi black asian hisp lat hc demo
               northeast New York City New York
                                                                                                            90.0
            0
                                                     1M+
                                                           8336817
                                                                           0.32
                                                                                 0.243
                                                                                       0.141
                                                                                                 0.291
                           Los Angeles California
                                                     1M+
                                                          3979576
                                                                           0.29
                                                                                 0.089
                                                                                       0.116
                                                                                                 0.485
                                                                                                           118.0
                    west
```

0.33

0.24

0.43

0.296

0.226

0.071

0.066

0.068

0.038

0.288

0.450

0.426

18.0

13.0

111.0

Merge & Clean 2019 Hate Crimes and 2018-2021 Political Parties (Congress, Governor, & Mayor)

2

midwest

west

southwest

Chicago

Houston

Phoenix

Illinois

Texas

Arizona

1M+

1M+

2693976

2320268

1M+ 1680992

```
In [9]: # merge political and hate crime dfs to top_df
top_df = pd.merge(poli_df, hc_df, on = "city", how = "left")
top_df.head()
```

Out[9]:

Democrat
Democrat
epublican
Democrat
Democrat
E

In [10]: top_df.columns

In [11]: # clean columns for top_df

Out[11]:

	region	city	state	popsize	pop_est	white_nonhi	black	asian	hisp_lat	mayor_party	gov_party	cong_maj	hc_demo
0	southwest	Albuquerque	New Mexico	500k-999,999	560513	0.39	0.033	0.029	0.492	Democrat	Democrat	Democrat	25.0
1	southeast	Atlanta	Georgia	500k-999,999	506811	0.38	0.510	0.044	0.043	Democrat	Republican	Democrat	NaN
2	southwest	Austin	Texas	500k-999,999	978908	0.48	0.078	0.076	0.339	Democrat	Republican	Republican	5.0
3	northeast	Baltimore	Maryland	500k-999,999	593490	0.28	0.624	0.026	0.053	Democrat	Republican	Democrat	NaN
4	northeast	Boston	Massachusetts	500k-999.999	692600	0.45	0.252	0.097	0.198	Democrat	Republican	Democrat	113.0

In [12]: # sort top_df by population size estimate top_df = top_df.sort_values(by=['pop_est'], ascending = False) top_df.head()

Out[12]:

	region	city	state	popsize	pop_est	white_nonhi	black	asian	hisp_lat	mayor_party	gov_party	cong_maj	hc_demo
26	northeast	New York City	New York	1M+	8336817	0.32	0.243	0.141	0.291	Democrat	Democrat	Democrat	90.0
20	west	Los Angeles	California	1M+	3979576	0.29	0.089	0.116	0.485	Democrat	Democrat	Democrat	118.0
6	midwest	Chicago	Illinois	1M+	2693976	0.33	0.296	0.066	0.288	Democrat	Democrat	Democrat	18.0
15	southwest	Houston	Texas	1M+	2320268	0.24	0.226	0.068	0.450	Democrat	Republican	Republican	13.0
30	west	Phoenix	Arizona	1M+	1680992	0.43	0.071	0.038	0.426	Democrat	Republican	Democrat	111.0

```
In [13]: # Create new columns for total population size estimate of black + asian + hisp/lat;
# and columns for each demographic's population size estimate

top_df["minor_pop_est"] = (top_df["black"] + top_df["asian"] + top_df["hisp_lat"]) * top_df["pop_est"]
    top_df["white_pop_est"] = top_df["white_nonhi"] * top_df["pop_est"]
    top_df["black_pop_est"] = top_df["black"] * top_df["pop_est"]
    top_df["asian_pop_est"] = top_df["asian"] * top_df["pop_est"]
    top_df["hisp_lat_pop_est"] = top_df["hisp_lat"] * top_df["pop_est"]
```

Out[13]:

	region	city	state	popsize	pop_est	white_nonhi	black	asian	hisp_lat	mayor_party	gov_party	cong_maj	hc_demo	minor_pop_
26	northeast	New York City	New York	1M+	8336817	0.32	0.243	0.141	0.291	Democrat	Democrat	Democrat	90.0	5627351.4
20	west	Los Angeles	California	1M+	3979576	0.29	0.089	0.116	0.485	Democrat	Democrat	Democrat	118.0	2745907.
6	midwest	Chicago	Illinois	1M+	2693976	0.33	0.296	0.066	0.288	Democrat	Democrat	Democrat	18.0	1751084.
15	southwest	Houston	Texas	1M+	2320268	0.24	0.226	0.068	0.450	Democrat	Republican	Republican	13.0	1726279.
30	west	Phoenix	Arizona	1M+	1680992	0.43	0.071	0.038	0.426	Democrat	Republican	Democrat	111.0	899330.

In [14]: # save master df for resiliency

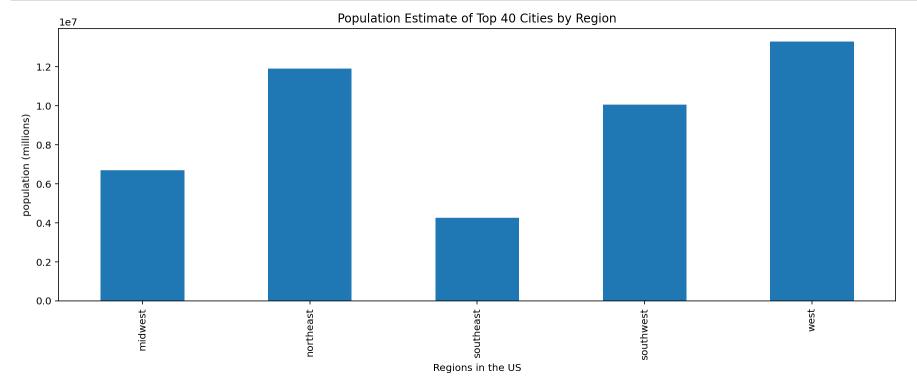
top df.head()

top_df.to_csv("/Users/adriana/Google Drive/_Learning/_DS4A/Assignments/0_Final_Project/2020_top_thirty_fina
l.csv", index = False)

Population and Minority Demos

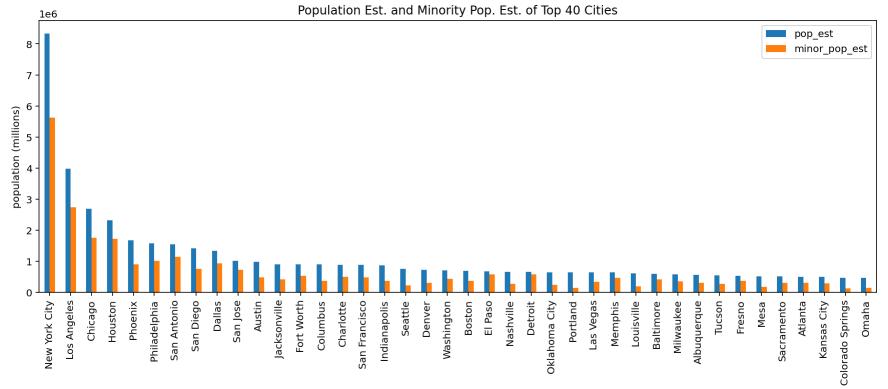
```
In [15]: # Population Estimate of Top 40 Cities by Region

pop_region_plot = top_df.groupby("region")["pop_est"].sum()
pop_region_plot.plot.bar(figsize = (15,5))
plt.xlabel("Regions in the US")
plt.ylabel("population (millions)")
plt.title("Population Estimate of Top 40 Cities by Region");
```

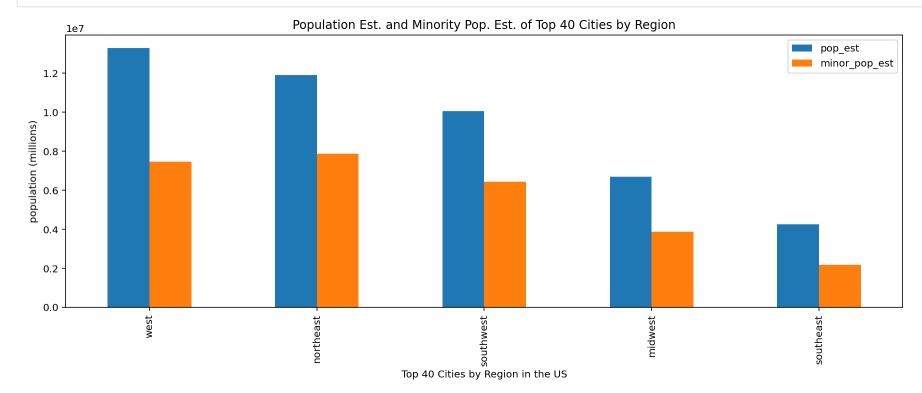


```
In [16]: # Population Est. and Minority Pop. Est. of Top 40 Cities

pop_city_plot = top_df[["city", "pop_est", "minor_pop_est"]].sort_values(by=['pop_est'], ascending = False)
pop_city_plot.plot.bar(x = "city", figsize = (15,5))
plt.xlabel("Top 40 Cities in the US")
plt.ylabel("population (millions)")
plt.title("Population Est. and Minority Pop. Est. of Top 40 Cities");
```

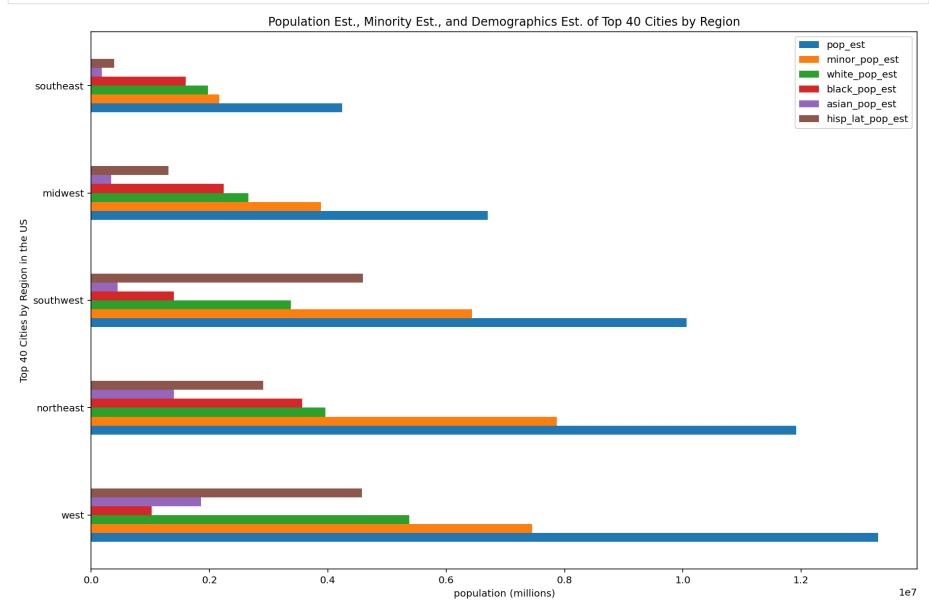


Top 40 Cities in the US



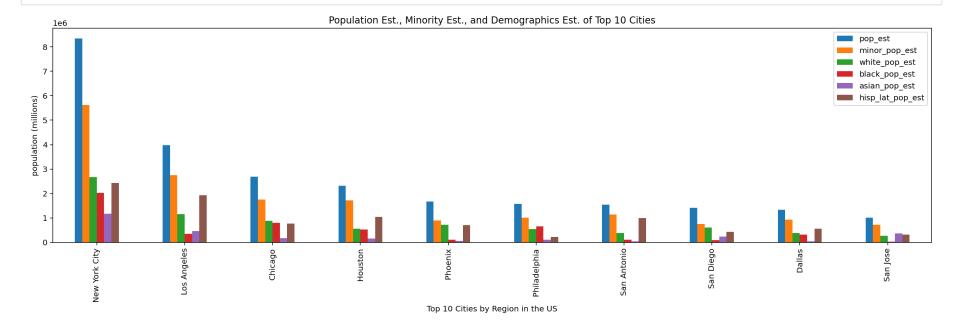
In [18]: # Population Est., Minority Est., and Demographics Est. of Top 40 Cities by Region

pop_demo_plot = top_df.groupby("region")[["pop_est", "minor_pop_est", "white_pop_est","black_pop_est","asia n_pop_est", "hisp_lat_pop_est"]].sum().sort_values(by=['pop_est'], ascending = False)
pop_demo_plot.plot.barh(figsize = (15,10))
plt.xlabel("population (millions)")
plt.ylabel("Top 40 Cities by Region in the US")
plt.title("Population Est., Minority Est., and Demographics Est. of Top 40 Cities by Region");



```
In [19]: # Population Est., Minority Est., and Demographics Est. of Top 10 Cities

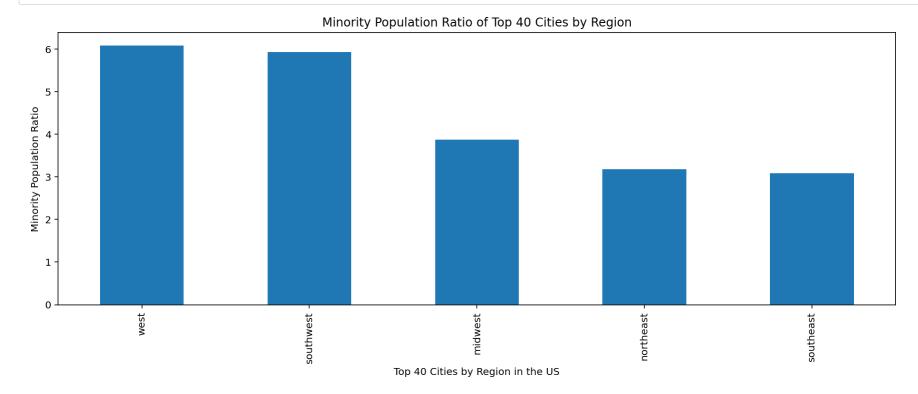
pop_demo_plot = top_df.groupby("city")[["pop_est", "minor_pop_est", "white_pop_est","black_pop_est","asian_pop_est", "hisp_lat_pop_est"]].sum().sort_values(by=['pop_est'], ascending = False).head(10)
pop_demo_plot.plot.bar(figsize = (20,5))
plt.xlabel("Top 10 Cities by Region in the US")
plt.ylabel("population (millions)")
plt.title("Population Est., Minority Est., and Demographics Est. of Top 10 Cities");
```



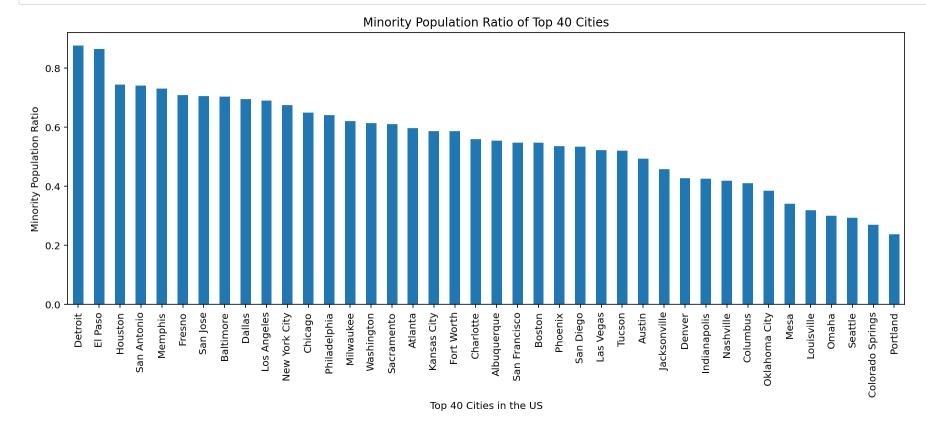
```
In [20]: # Create new features for each total minorities
# and each demographic's percentage of the total population, for each city

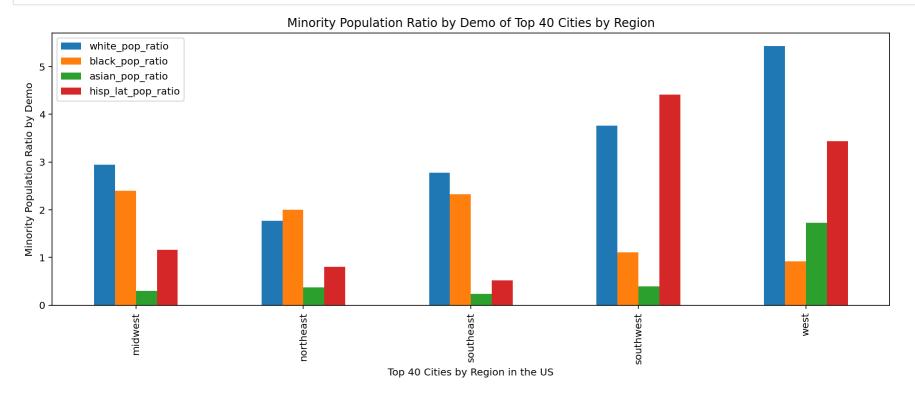
top_df["minor_pop_ratio"] = top_df["minor_pop_est"]/top_df["pop_est"]
top_df["white_pop_ratio"] = top_df["white_pop_est"]/top_df["pop_est"]
top_df["black_pop_ratio"] = top_df["black_pop_est"]/top_df["pop_est"]
top_df["asian_pop_ratio"] = top_df["asian_pop_est"]/top_df["pop_est"]
top_df["hisp_lat_pop_ratio"] = top_df["hisp_lat_pop_est"]/top_df["pop_est"]
```

In [22]: # Minority Population Ratio of Top 40 Cities by Region ratio_region_plot = top_df.groupby("region")["minor_pop_ratio"].sum().sort_values(ascending = False) ratio_region_plot.plot.bar(figsize = (15,5)) plt.xlabel("Top 40 Cities by Region in the US") plt.ylabel("Minority Population Ratio") plt.title("Minority Population Ratio of Top 40 Cities by Region");



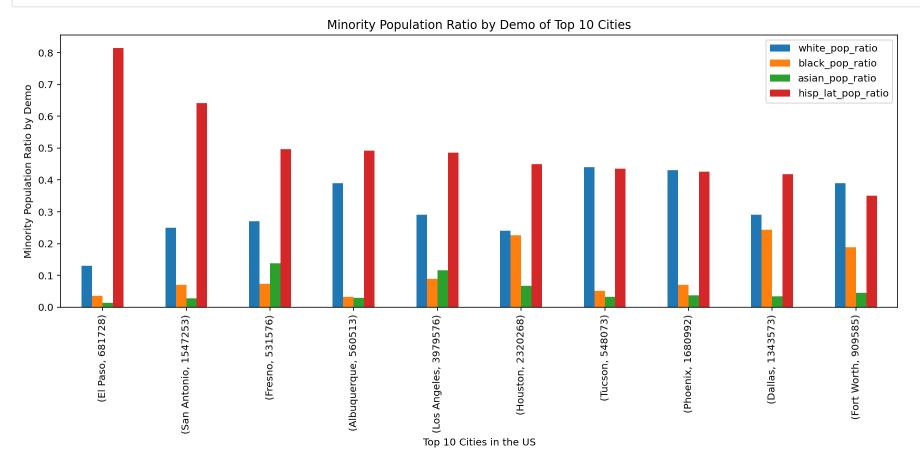
In [23]: # Minority Population Ratio of Top 40 Cities ratio_city_plot = top_df.groupby("city")["minor_pop_ratio"].sum().sort_values(ascending = False) ratio_city_plot.plot.bar(figsize = (15,5)) plt.xlabel("Top 40 Cities in the US") plt.ylabel("Minority Population Ratio") plt.title("Minority Population Ratio of Top 40 Cities");





```
In [25]: # Minority Population Ratio by Demo of Top 10 Cities

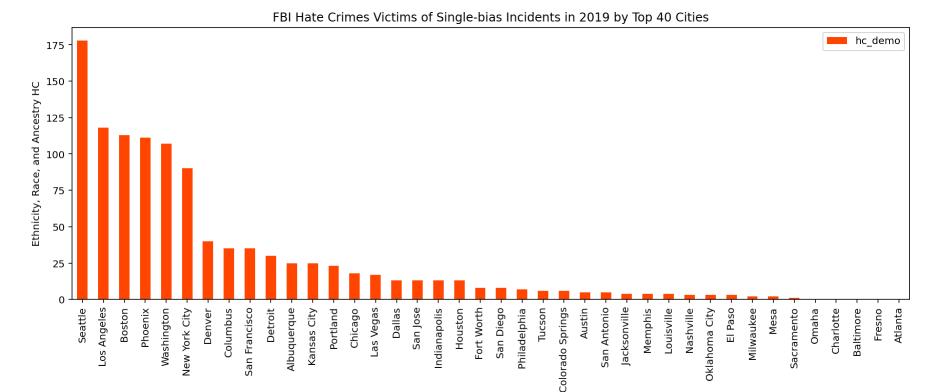
dem_ratio_city_plot = top_df.groupby(["city","pop_est"])[["white_pop_ratio", "black_pop_ratio", "asian_pop_ratio", "hisp_lat_pop_ratio"]].sum().sort_values(by = "hisp_lat_pop_ratio", ascending = False).head(10)
    dem_ratio_city_plot.plot.bar(figsize = (15,5))
    plt.xlabel("Top 10 Cities in the US")
    plt.ylabel("Minority Population Ratio by Demo")
    plt.title("Minority Population Ratio by Demo of Top 10 Cities");
```



Hate Crime

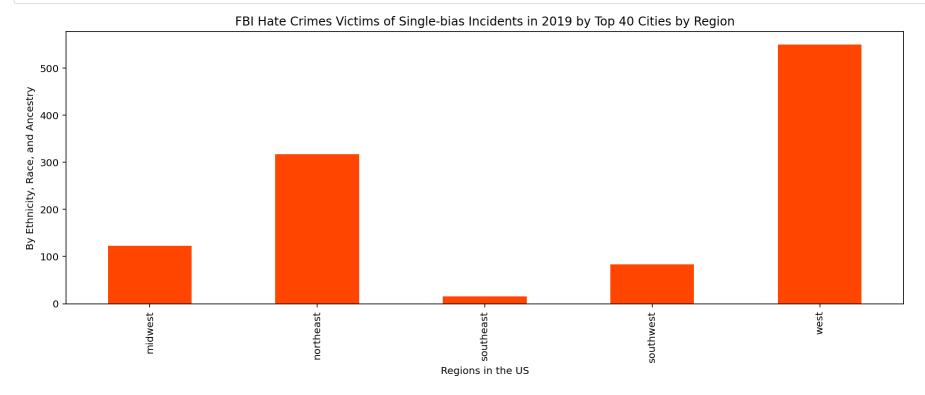
In [26]: # FBI Hate Crimes Victims of Single-bias Incidents in 2019 by Top 40 Cities

hc_plot = top_df[["city", "hc_demo"]].sort_values(by = "hc_demo", ascending = False)
hc_plot.plot.bar(x = "city", y = "hc_demo", color = "orangered", figsize = (15,5))
plt.xlabel("Top 40 Cities")
plt.ylabel("Ethnicity, Race, and Ancestry HC")
plt.title("FBI Hate Crimes Victims of Single-bias Incidents in 2019 by Top 40 Cities");



Top 40 Cities

```
In [27]: # FBI Hate Crimes Victims of Single-bias Incidents in 2019 by Top 40 Cities by Region
hc_region_plot = top_df.groupby("region")["hc_demo"].sum()
hc_region_plot.plot.bar(color = "orangered", figsize = (15,5))
plt.xlabel("Regions in the US")
plt.ylabel("By Ethnicity, Race, and Ancestry")
plt.title("FBI Hate Crimes Victims of Single-bias Incidents in 2019 by Top 40 Cities by Region");
```



Hate Crime and Population Correlations

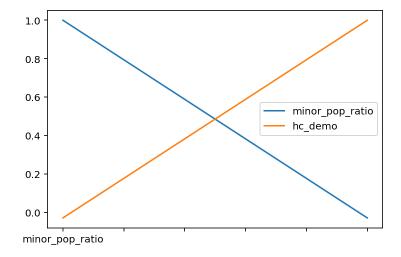
```
In [28]: # Correlation of total minority population percentage to count of victims of hate crimes (race, ethnicity, ancestry)

minor_hc_corr = top_df[["minor_pop_ratio", "hc_demo"]].corr()
minor_hc_corr
```

Out[28]:

	minor_pop_ratio	hc_demo
minor_pop_ratio	1.000000	-0.028896
hc_demo	-0.028896	1.000000

In [29]: minor_hc_corr.plot.line();

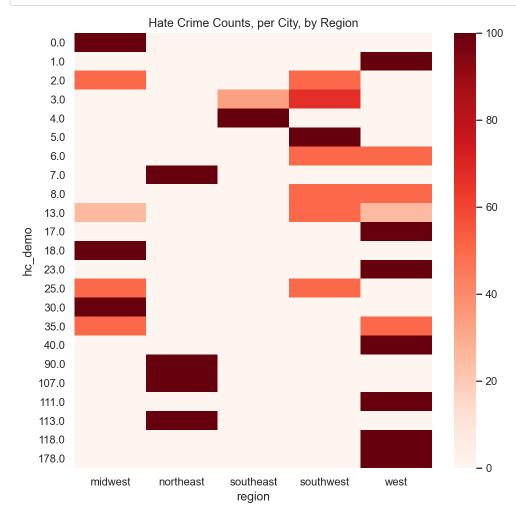


Out[30]:

region	midwest	northeast	southeast	southwest	west
hc_demo					
0.0	100.0	0.0	0.000000	0.000000	0.0
1.0	0.0	0.0	0.000000	0.000000	100.0
2.0	50.0	0.0	0.000000	50.000000	0.0
3.0	0.0	0.0	33.333333	66.666667	0.0
4.0	0.0	0.0	100.000000	0.000000	0.0

```
In [31]: # heatmap of crosstab

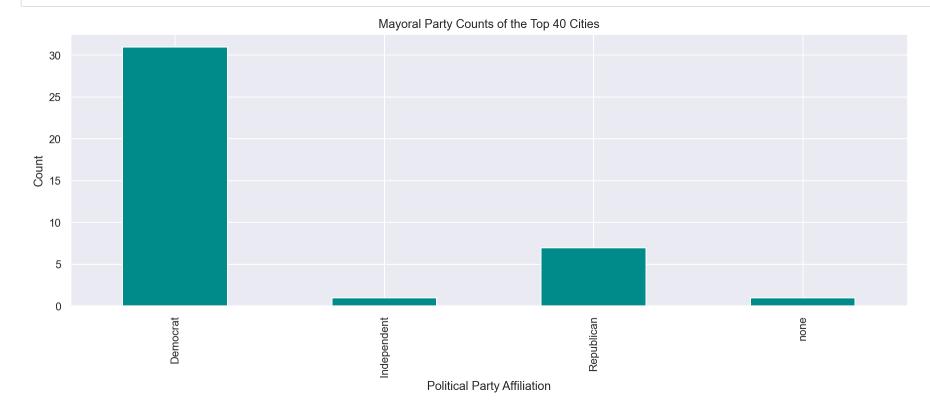
sns.set(rc = {'figure.figsize':(8, 8)})
ax = sns.heatmap(hc_norm, cmap = "Reds")
ax.set_title("Hate Crime Counts, per City, by Region");
```



Mayor Political Party

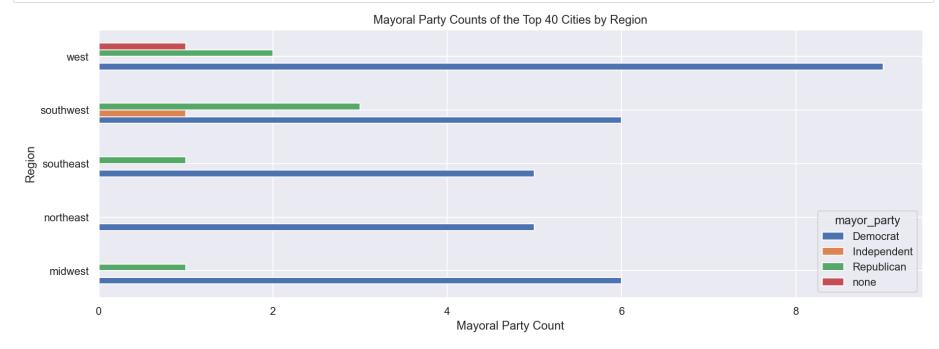
In [32]: # Mayoral Party Counts of the Top 40 Cities

mayor_plot = top_df.groupby("mayor_party")["city"].count()
mayor_plot.plot.bar(color = "darkcyan", figsize = (15,5))
plt.xlabel("Political Party Affiliation")
plt.ylabel("Count")
plt.title("Mayoral Party Counts of the Top 40 Cities");



```
In [33]: # Mayoral Party Counts of the Top 40 Cities by Region

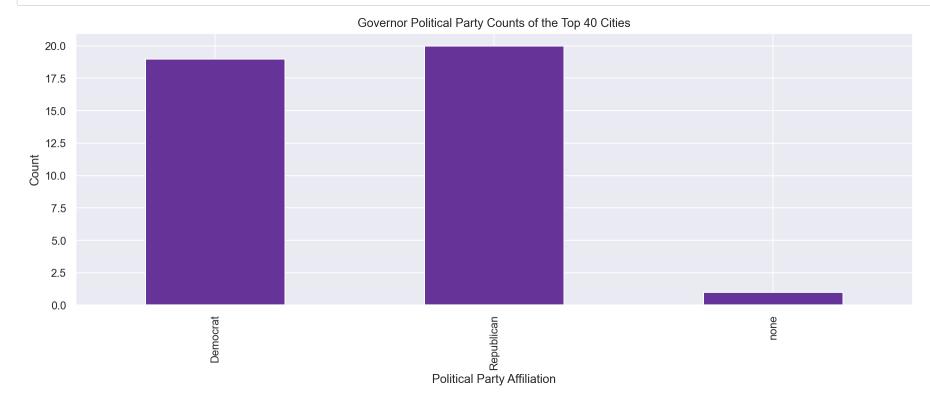
mayor_region_plot = top_df.groupby("region")["mayor_party"].value_counts().unstack()
mayor_region_plot.plot.barh(figsize = (15,5))
plt.xlabel("Mayoral Party Count")
plt.ylabel("Region")
plt.title("Mayoral Party Counts of the Top 40 Cities by Region");
```



Governor Political Party

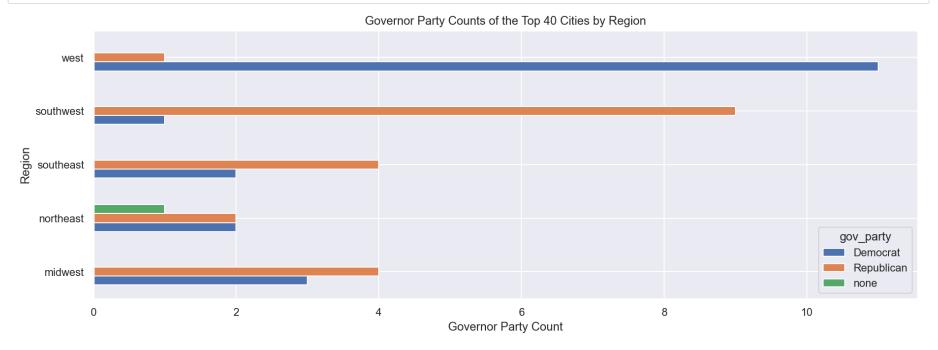
In [34]: # Governor Political Party Counts of the Top 40 Cities

gov_plot = top_df.groupby("gov_party")["city"].count()
gov_plot.plot.bar(color = "rebeccapurple", figsize = (15,5))
plt.xlabel("Political Party Affiliation")
plt.ylabel("Count")
plt.title("Governor Political Party Counts of the Top 40 Cities");



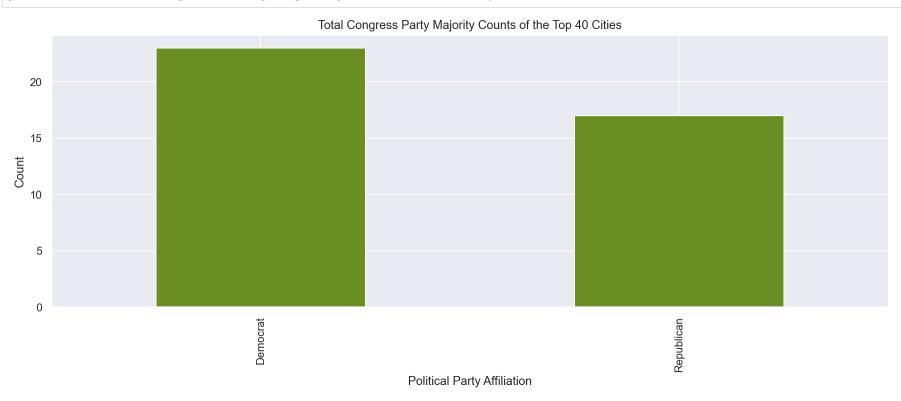
```
In [35]: # Governor Party Counts of the Top 40 Cities by Region

gov_region_plot = top_df.groupby("region")["gov_party"].value_counts().unstack()
    gov_region_plot.plot.barh(figsize = (15,5))
    plt.xlabel("Governor Party Count")
    plt.ylabel("Region")
    plt.title("Governor Party Counts of the Top 40 Cities by Region");
```



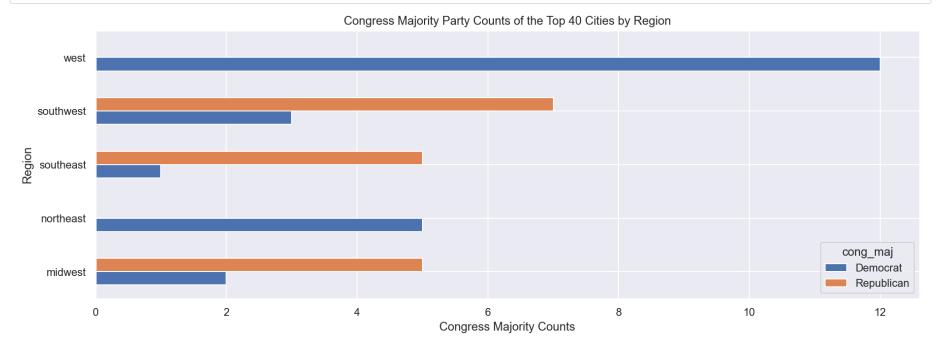
Congress Majority Party

In [36]: # Total Congress Party Majority Counts of the Top 40 Cities cong_plot = top_df.groupby("cong_maj")["city"].count() cong_plot.plot.bar(color = "olivedrab", figsize = (15,5)) plt.xlabel("Political Party Affiliation") plt.ylabel("Count") plt.title("Total Congress Party Majority Counts of the Top 40 Cities");



```
In [37]: # Congress Majority Party Counts of the Top 40 Cities by Region

cong_region_plot = top_df.groupby("region")["cong_maj"].value_counts().unstack()
cong_region_plot.plot.barh(figsize = (15,5))
plt.xlabel("Congress Majority Counts")
plt.ylabel("Region")
plt.title("Congress Majority Party Counts of the Top 40 Cities by Region");
```

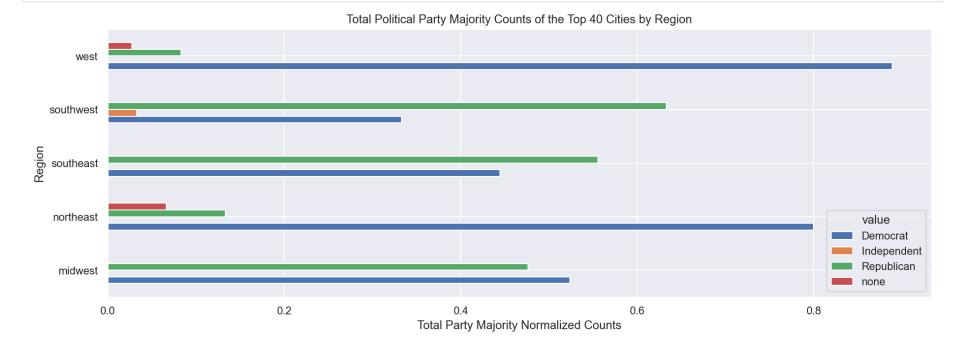


Political Sums

To-Do:

• mayor_state = how frequently mayoral party and state majority party differs

In [38]: # Total Party Majority Counts of the Top 40 Cities by Region political = top_df[["region", "city", "mayor_party", "gov_party", "cong_maj"]] political = political.melt(id_vars = ["region","city"], value_vars = ["mayor_party", "gov_party", "cong_maj"]) political_gb = political.groupby("region")["value"].value_counts(normalize=True).unstack() political_gb.plot.barh(figsize = (15,5)) plt.xlabel("Total Party Majority Normalized Counts") plt.ylabel("Region") plt.title("Total Political Party Majority Counts of the Top 40 Cities by Region");



```
In [39]: mayor_count = top_df.groupby("mayor_party")["city"].count()
    mayor_region_count = top_df.groupby("region")["mayor_party"].value_counts().unstack()
    gov_count = top_df.groupby("gov_party")["city"].count()
    gov_region_count = top_df.groupby("region")["gov_party"].value_counts().unstack()
    cong_count = top_df.groupby("cong_maj")["city"].count()
    cong_region_count = top_df.groupby("region")["cong_maj"].value_counts().unstack()
    political_total_norm = political.groupby("region")["value"].value_counts(normalize=True).unstack()
```

```
In [40]: mayor_region_count
Out[40]:
```

mayor_party	Democrat	Independent	Republican	none
region				
midwest	6.0	NaN	1.0	NaN
northeast	5.0	NaN	NaN	NaN
southeast	5.0	NaN	1.0	NaN
southwest	6.0	1.0	3.0	NaN
west	9.0	NaN	2.0	1.0

Sources

Census Data (2019-2020): https://www.census.gov/quickfacts/fact/table (https://www.census.gov/quickfacts/fact/table)

Hate Crime Data (2019 - race, ethnicity, ancestry): https://ucr.fbi.gov/hate-crime/2019 (https://ucr.fbi.gov/hate-crime/2019)

Political Data

Mayoral (2021): https://ballotpedia.org/Party_affiliation_of_the_mayors_of_the_100_largest_cities)

/Party_affiliation_of_the_mayors_of_the_100_largest_cities)

Governor (2021): https://ballotpedia.org/Partisan_composition_of_governors (https://ballotpedia.org/Partisan_composition_of_governors)

Senate (2020): https://dataverse.harvard.edu/dataset.xhtml?persistentld=doi:10.7910/DVN/PEJ5QU (https://dataverse.harvard.edu/dataset.xhtml?persistentld=doi:10.7910/DVN/PEJ5QU (https://dataverse.harvard.edu/dataset.xhtml?persistentld=doi:10.7910/DVN/PEJ5QU (https://dataverse.harvard.edu/dataset.xhtml?persistentld=doi:10.7910/DVN/PEJ5QU (https://dataverse.harvard.edu/dataset.xhtml?persistentld=doi:10.7910/DVN/PEJ5QU (https://dataverse.harvard.edu/dataset.xhtml (<a href="https://dataverse.harvard.edu/dataverse.harvard.edu/dataverse.harvard.edu/dataverse.harvard.edu/dataverse.harvard.edu

House of Representatives (2018): https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/IG0UN2 (https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/IG0UN2 (https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/IG0UN2 (https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/IG0UN2)

In []:	
---------	--