

# EDA Template

December 9, 2024

## 1 STOR 320: Introduction to Data Science

### 1.1 EDA Group PLACE\_GROUP\_NUMBER\_HERE (Ex: EDA Group 12)

### 1.2 Part 1: Data cleaning, merging, and visualization (6 points)

```
[2]: # At the start of your notebook
from IPython.display import display, HTML
import warnings
warnings.filterwarnings('ignore')

# Hide code cells
display(HTML("""
<style>
.jp-CodeCell {
    display: none !important;
}
.jp-MarkdownCell {
    display: block !important;
}
</style>
"""))

# Hide prompts
display(HTML("""
<style>
.prompt {
    display: none !important;
}
</style>
"""))
```

<IPython.core.display.HTML object>

<IPython.core.display.HTML object>

```
[61]: import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score, precision_recall_curve,
    confusion_matrix, classification_report, roc_curve
from sklearn.model_selection import train_test_split
```

```
[62]: eviction = pd.read_csv('eviction_2020_2024.csv')
eviction.head()
```

```
[62]:
```

	city	type	GEOID	racial_majority	month	\
0	Albuquerque, NM	Census Tract	35001000107	White	01/2020	
1	Albuquerque, NM	Census Tract	35001000107	White	02/2020	
2	Albuquerque, NM	Census Tract	35001000107	White	03/2020	
3	Albuquerque, NM	Census Tract	35001000107	White	04/2020	
4	Albuquerque, NM	Census Tract	35001000107	White	05/2020	

	filings_2020	filings_avg	last_updated	xcourtcofips	
0	8	5.333333	2024-08-10	NaN	
1	14	9.000000	2024-08-10	NaN	
2	10	5.666667	2024-08-10	NaN	
3	5	8.333333	2024-08-10	NaN	
4	0	6.666667	2024-08-10	NaN	

```
[63]: PIT_2023 = pd.read_excel('PITCOC.xlsb', engine='pyxlsb', sheet_name="2023")
PIT_2023.head()
```

```
[63]:
```

	CoC Number	CoC Name	\
0	AK-500	Anchorage CoC	
1	AK-501	Alaska Balance of State CoC	
2	AL-500	Birmingham/Jefferson, St. Clair, Shelby Counti...	
3	AL-501	Mobile City & County/Baldwin County CoC	
4	AL-502	Florence/Northwest Alabama CoC	

	CoC Category	Count Types	Overall Homeless	\
0	Other Largely Urban CoC	Sheltered and Unsheltered Count	1760.0	
1	Largely Rural CoC	Sheltered and Unsheltered Count	854.0	
2	Largely Suburban CoC	Sheltered and Unsheltered Count	847.0	
3	Other Largely Urban CoC	Sheltered and Unsheltered Count	670.0	
4	Largely Rural CoC	Sheltered and Unsheltered Count	195.0	

	Overall Homeless - Under 18	Overall Homeless - Age 18 to 24	\
0	185.0	161.0	
1	176.0	66.0	
2	67.0	42.0	
3	110.0	19.0	
4	63.0	9.0	

	Overall Homeless - Age 25 to 34	Overall Homeless - Age 35 to 44 \
0	377.0	419.0
1	124.0	190.0
2	127.0	182.0
3	78.0	156.0
4	42.0	36.0

	Overall Homeless - Age 45 to 54 ... \
0	315.0 ...
1	144.0 ...
2	180.0 ...
3	120.0 ...
4	23.0 ...

	Overall Homeless Parenting Youth Age 18-24 \
0	10.0
1	8.0
2	2.0
3	2.0
4	4.0

	Sheltered ES Homeless Parenting Youth Age 18-24 \
0	5.0
1	8.0
2	1.0
3	2.0
4	2.0

	Sheltered TH Homeless Parenting Youth Age 18-24 \
0	5.0
1	0.0
2	1.0
3	0.0
4	0.0

	Sheltered Total Homeless Parenting Youth Age 18-24 \
0	10.0
1	8.0
2	2.0
3	2.0
4	2.0

	Unsheltered Homeless Parenting Youth Age 18-24 \
0	0.0
1	0.0
2	0.0

```

3                                0.0
4                                2.0

Overall Homeless Children of Parenting Youth \
0                                10.0
1                                7.0
2                                2.0
3                                2.0
4                                4.0

Sheltered ES Homeless Children of Parenting Youth \
0                                5.0
1                                7.0
2                                1.0
3                                2.0
4                                2.0

Sheltered TH Homeless Children of Parenting Youth \
0                                5.0
1                                0.0
2                                1.0
3                                0.0
4                                0.0

Sheltered Total Homeless Children of Parenting Youth \
0                                10.0
1                                7.0
2                                2.0
3                                2.0
4                                2.0

Unsheltered Homeless Children of Parenting Youth
0                                0.0
1                                0.0
2                                0.0
3                                0.0
4                                2.0

```

[5 rows x 645 columns]

```
[64]: eviction.head()
```

```

[64]:      city      type      GEOID racial_majority  month \
0  Albuquerque, NM  Census Tract  35001000107      White  01/2020
1  Albuquerque, NM  Census Tract  35001000107      White  02/2020
2  Albuquerque, NM  Census Tract  35001000107      White  03/2020
3  Albuquerque, NM  Census Tract  35001000107      White  04/2020

```

4 Albuquerque, NM Census Tract 35001000107 White 05/2020

	filings_2020	filings_avg	last_updated	xcourtcofips
0	8	5.333333	2024-08-10	NaN
1	14	9.000000	2024-08-10	NaN
2	10	5.666667	2024-08-10	NaN
3	5	8.333333	2024-08-10	NaN
4	0	6.666667	2024-08-10	NaN

```
[65]: ## column names check (1 point)
```

```
[66]: eviction.columns
```

```
[66]: Index(['city', 'type', 'GEOID', 'racial_majority', 'month', 'filings_2020',  
        'filings_avg', 'last_updated', 'xcourtcofips'],  
        dtype='object')
```

```
[67]: # remove GEOID, last_updated, and xcourtcofips because they're not useful for_  
      ↪merging or relevant  
eviction = eviction.drop(["GEOID", "last_updated", "xcourtcofips"], axis=1)  
eviction.columns
```

```
[67]: Index(['city', 'type', 'racial_majority', 'month', 'filings_2020',  
        'filings_avg'],  
        dtype='object')
```

```
[68]: len(PIT_2023.columns) # len because there's a lot
```

```
[68]: 645
```

```
[69]: new_cols = PIT_2023.columns[0:25] # columns after 25 talk about sheltered and_  
      ↪unsheltered  
PIT_sliced = PIT_2023[new_cols][0:386] # truncating rows  
PIT_sliced = PIT_sliced.drop("Count Types", axis=1)  
PIT_sliced.head()
```

```
[69]:
```

	CoC Number	CoC Name \
0	AK-500	Anchorage CoC
1	AK-501	Alaska Balance of State CoC
2	AL-500	Birmingham/Jefferson, St. Clair, Shelby Counti...
3	AL-501	Mobile City & County/Baldwin County CoC
4	AL-502	Florence/Northwest Alabama CoC

	CoC Category	Overall Homeless	Overall Homeless - Under 18 \
0	Other Largely Urban CoC	1760.0	185.0
1	Largely Rural CoC	854.0	176.0
2	Largely Suburban CoC	847.0	67.0
3	Other Largely Urban CoC	670.0	110.0

4	Largely Rural CoC	195.0	63.0
---	-------------------	-------	------

	Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34	\
0	161.0	377.0	
1	66.0	124.0	
2	42.0	127.0	
3	19.0	78.0	
4	9.0	42.0	

	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54	\
0	419.0	315.0	
1	190.0	144.0	
2	182.0	180.0	
3	156.0	120.0	
4	36.0	23.0	

	Overall Homeless - Age 55 to 64	...	\
0	223.0	...	
1	123.0	...	
2	187.0	...	
3	140.0	...	
4	16.0	...	

	Overall Homeless - Gender that is not Singularly Female or Male	\
0	3.0	
1	2.0	
2	1.0	
3	2.0	
4	0.0	

	Overall Homeless - Gender Questioning	\
0	2.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	

	Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x)	\
0	1638.0	
1	717.0	
2	822.0	
3	655.0	
4	186.0	

	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White	\
0	122.0	480.0	
1	137.0	316.0	

2	25.0	295.0
3	15.0	281.0
4	9.0	130.0

	Overall Homeless - Black, African American, or African \
0	156.0
1	25.0
2	518.0
3	351.0
4	48.0

	Overall Homeless - Asian or Asian American \
0	28.0
1	10.0
2	1.0
3	2.0
4	0.0

	Overall Homeless - American Indian, Alaska Native, or Indigenous \
0	755.0
1	396.0
2	14.0
3	11.0
4	2.0

	Overall Homeless - Native Hawaiian or Other Pacific Islander \
0	83.0
1	16.0
2	2.0
3	6.0
4	0.0

	Overall Homeless - Multiple Races
0	258.0
1	91.0
2	17.0
3	19.0
4	15.0

[5 rows x 24 columns]

```
[70]: ## missing data check (1 point)
```

```
[71]: eviction_cols = eviction.columns
      for col in eviction_cols:
          slice = eviction[str(col)]
          if slice.hasnans:
```

```
print(f"{str(col)} " + f"({len(slice.unique())})" + ": " + str(slice.
↪unique()))
```

```
racial_majority (5): ['White' 'Other' 'Latinx' nan 'Black']
filings_avg (2775): [ 5.33333333  9.          5.66666667 ... 79.25          59.75
69.25          ]
```

```
[72]: # we can drop filings_avg since we are calculating the average for the year
↪later
# we can drop racial_majority since we can supplement it with census data if
↪need be
eviction = eviction.drop(["filings_avg", "racial_majority"], axis=1)
eviction
```

```
[72]:
```

	city	type	month	filings_2020
0	Albuquerque, NM	Census Tract	01/2020	8
1	Albuquerque, NM	Census Tract	02/2020	14
2	Albuquerque, NM	Census Tract	03/2020	10
3	Albuquerque, NM	Census Tract	04/2020	5
4	Albuquerque, NM	Census Tract	05/2020	0
...	...	...	...	...
598451	Wilmington, DE	Census Tract	04/2024	1
598452	Wilmington, DE	Census Tract	05/2024	9
598453	Wilmington, DE	Census Tract	06/2024	3
598454	Wilmington, DE	Census Tract	07/2024	5
598455	Wilmington, DE	Census Tract	08/2024	1

[598456 rows x 4 columns]

```
[73]: # replacing NaN with zero because values are not categorical
PIT_sliced = PIT_sliced.fillna(0)
PIT_sliced.head()
```

```
[73]:
```

	CoC Number	CoC Name \
0	AK-500	Anchorage CoC
1	AK-501	Alaska Balance of State CoC
2	AL-500	Birmingham/Jefferson, St. Clair, Shelby Counti...
3	AL-501	Mobile City & County/Baldwin County CoC
4	AL-502	Florence/Northwest Alabama CoC

	CoC Category	Overall Homeless	Overall Homeless - Under 18 \
0	Other Largely Urban CoC	1760.0	185.0
1	Largely Rural CoC	854.0	176.0
2	Largely Suburban CoC	847.0	67.0
3	Other Largely Urban CoC	670.0	110.0
4	Largely Rural CoC	195.0	63.0

Overall Homeless - Age 18 to 24   Overall Homeless - Age 25 to 34 \



0	161.0	377.0
1	66.0	124.0
2	42.0	127.0
3	19.0	78.0
4	9.0	42.0

	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54 \
0	419.0	315.0
1	190.0	144.0
2	182.0	180.0
3	156.0	120.0
4	36.0	23.0

	Overall Homeless - Age 55 to 64 ... \
0	223.0 ...
1	123.0 ...
2	187.0 ...
3	140.0 ...
4	16.0 ...

	Overall Homeless - Gender that is not Singularly Female or Male \
0	3.0
1	2.0
2	1.0
3	2.0
4	0.0

	Overall Homeless - Gender Questioning \
0	2.0
1	0.0
2	0.0
3	0.0
4	0.0

	Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \
0	1638.0
1	717.0
2	822.0
3	655.0
4	186.0

	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White \
0	122.0	480.0
1	137.0	316.0
2	25.0	295.0
3	15.0	281.0
4	9.0	130.0

	Overall Homeless - Black, African American, or African \
0	156.0
1	25.0
2	518.0
3	351.0
4	48.0

	Overall Homeless - Asian or Asian American \
0	28.0
1	10.0
2	1.0
3	2.0
4	0.0

	Overall Homeless - American Indian, Alaska Native, or Indigenous \
0	755.0
1	396.0
2	14.0
3	11.0
4	2.0

	Overall Homeless - Native Hawaiian or Other Pacific Islander \
0	83.0
1	16.0
2	2.0
3	6.0
4	0.0

	Overall Homeless - Multiple Races
0	258.0
1	91.0
2	17.0
3	19.0
4	15.0

[5 rows x 24 columns]

```
[74]: ## Outlier check (1 point)
```

```
[75]: PIT_sliced.describe()
```

```
[75]: Overall Homeless Overall Homeless - Under 18 \
```

count	386.000000	386.000000
mean	3383.958549	578.341969
std	33685.071349	5823.125602
min	55.000000	1.000000

25%	337.000000	53.000000
50%	656.500000	109.500000
75%	1549.250000	226.750000
max	653104.000000	111620.000000

	Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34 \
count	386.000000	386.000000
mean	245.782383	575.139896
std	2462.595873	5768.223300
min	2.000000	5.000000
25%	21.000000	52.250000
50%	44.000000	107.000000
75%	102.000000	244.500000
max	47436.000000	111002.000000

	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54 \
count	386.000000	386.000000
mean	618.119171	504.088083
std	6160.189569	5011.779805
min	0.000000	0.000000
25%	60.000000	48.000000
50%	120.500000	99.500000
75%	285.500000	233.000000
max	119297.000000	97289.000000

	Overall Homeless - Age 55 to 64	Overall Homeless - Over 64 \
count	386.000000	386.000000
mean	469.958549	191.735751
std	4672.930049	1905.210525
min	0.000000	0.000000
25%	41.250000	16.000000
50%	89.500000	36.000000
75%	216.750000	88.500000
max	90702.000000	37005.000000

	Overall Homeless - Female	Overall Homeless - Male ... \
count	386.000000	386.000000 ...
mean	1295.383420	2047.461140 ...
std	12896.424039	20384.594635 ...
min	20.000000	23.000000 ...
25%	134.250000	190.500000 ...
50%	255.000000	389.000000 ...
75%	569.500000	890.500000 ...
max	250009.000000	395160.000000 ...

	Overall Homeless - Gender that is not Singularly Female or Male \
count	386.000000

mean	16.005181
std	160.203799
min	0.000000
25%	0.000000
50%	2.000000
75%	6.000000
max	3089.000000

Overall Homeless - Gender Questioning \	
count	386.000000
mean	3.932642
std	39.295757
min	0.000000
25%	0.000000
50%	0.000000
75%	1.000000
max	759.000000

Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \	
count	386.000000
mean	2454.756477
std	24261.167747
min	17.000000
25%	292.000000
50%	565.500000
75%	1197.000000
max	473768.000000

Overall Homeless - Hispanic/Latin(o)(a)(x)		Overall Homeless - White \
count	386.000000	386.000000
mean	929.202073	1683.181347
std	9548.451413	16621.001323
min	0.000000	9.000000
25%	23.250000	156.750000
50%	68.000000	344.000000
75%	221.750000	821.000000
max	179336.000000	324854.000000

Overall Homeless - Black, African American, or African \	
count	386.000000
mean	1262.300518
std	12835.350265
min	0.000000
25%	70.000000
50%	177.500000
75%	436.000000
max	243624.000000

```

Overall Homeless - Asian or Asian American \
count      386.000000
mean       59.968912
std        622.043861
min         0.000000
25%         1.000000
50%         4.000000
75%        11.000000
max       11574.000000

Overall Homeless - American Indian, Alaska Native, or Indigenous \
count      386.000000
mean       119.772021
std       1188.048176
min         0.000000
25%         2.000000
50%        10.000000
75%        35.750000
max      23116.000000

Overall Homeless - Native Hawaiian or Other Pacific Islander \
count      386.000000
mean       55.502591
std       561.061296
min         0.000000
25%         0.000000
50%         2.000000
75%         9.000000
max      10712.000000

Overall Homeless - Multiple Races
count      386.000000
mean       203.233161
std      2022.758577
min         0.000000
25%        13.000000
50%        28.000000
75%       75.750000
max     39224.000000

```

[8 rows x 21 columns]

```

[76]: # the max seems quite high for overall homeless
max = PIT_sliced["Overall Homeless"].idxmax()
PIT_sliced.loc[[max]]

```

```

[76]:      CoC Number CoC Name CoC Category Overall Homeless \
385          0      Total          0          653104.0

      Overall Homeless - Under 18 Overall Homeless - Age 18 to 24 \
385                  111620.0                  47436.0

      Overall Homeless - Age 25 to 34 Overall Homeless - Age 35 to 44 \
385                  111002.0                  119297.0

      Overall Homeless - Age 45 to 54 Overall Homeless - Age 55 to 64 ... \
385                  97289.0                  90702.0 ...

      Overall Homeless - Gender that is not Singularly Female or Male \
385                              3089.0

      Overall Homeless - Gender Questioning \
385                              759.0

      Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \
385                              473768.0

      Overall Homeless - Hispanic/Latin(o)(a)(x) Overall Homeless - White \
385                  179336.0                  324854.0

      Overall Homeless - Black, African American, or African \
385                              243624.0

      Overall Homeless - Asian or Asian American \
385                              11574.0

      Overall Homeless - American Indian, Alaska Native, or Indigenous \
385                              23116.0

      Overall Homeless - Native Hawaiian or Other Pacific Islander \
385                              10712.0

      Overall Homeless - Multiple Races
385                              39224.0

[1 rows x 24 columns]

```

```

[77]: # these are totals, so it's okay to remove
      PIT_sliced = PIT_sliced.drop(max)
      PIT_sliced.describe()

```

```

[77]:      Overall Homeless Overall Homeless - Under 18 \
count          385.000000          385.000000

```

mean	1696.374026	289.922078
std	5955.731326	1343.006959
min	55.000000	1.000000
25%	337.000000	53.000000
50%	653.000000	109.000000
75%	1532.000000	226.000000
max	88025.000000	25200.000000

	Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34 \
count	385.00000	385.000000
mean	123.21039	288.316883
std	515.60255	1233.336451
min	2.00000	5.000000
25%	21.00000	52.000000
50%	44.00000	107.000000
75%	102.00000	240.000000
max	9130.00000	19577.000000

	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54 \
count	385.000000	385.000000
mean	309.862338	252.698701
std	1128.114723	852.013239
min	0.000000	0.000000
25%	60.000000	48.000000
50%	120.000000	99.000000
75%	284.000000	230.000000
max	16597.000000	13475.000000

	Overall Homeless - Age 55 to 64	Overall Homeless - Over 64 \
count	385.000000	385.000000
mean	235.589610	96.116883
std	797.125492	317.653276
min	0.000000	0.000000
25%	41.000000	16.000000
50%	89.000000	36.000000
75%	216.000000	87.000000
max	12841.000000	4721.000000

	Overall Homeless - Female	Overall Homeless - Male ... \
count	385.000000	385.000000 ...
mean	649.374026	1026.389610 ...
std	2289.696403	3623.145956 ...
min	20.000000	23.000000 ...
25%	134.000000	190.000000 ...
50%	254.000000	389.000000 ...
75%	568.000000	880.000000 ...
max	37788.000000	49650.000000 ...

	Overall Homeless - Gender that is not Singularly Female or Male \	
count	385.000000	
mean	8.023377	
std	32.806995	
min	0.000000	
25%	0.000000	
50%	2.000000	
75%	6.000000	
max	570.000000	
	Overall Homeless - Gender Questioning \	
count	385.000000	
mean	1.971429	
std	7.721164	
min	0.000000	
25%	0.000000	
50%	0.000000	
75%	1.000000	
max	110.000000	
	Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \	
count	385.000000	
mean	1230.566234	
std	3186.678681	
min	17.000000	
25%	292.000000	
50%	565.000000	
75%	1194.000000	
max	41029.000000	
	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White \
count	385.000000	385.000000
mean	465.807792	843.776623
std	2882.200067	2071.985499
min	0.000000	9.000000
25%	23.000000	156.000000
50%	68.000000	342.000000
75%	221.000000	809.000000
max	46996.000000	28624.000000
	Overall Homeless - Black, African American, or African \	
count	385.000000	
mean	632.789610	
std	3436.753939	
min	0.000000	
25%	70.000000	



50%	174.000000
75%	436.000000
max	59292.000000

Overall Homeless - Asian or Asian American \	
count	385.000000
mean	30.062338
std	204.468891
min	0.000000
25%	1.000000
50%	4.000000
75%	11.000000
max	3773.000000

Overall Homeless - American Indian, Alaska Native, or Indigenous \	
count	385.000000
mean	60.041558
std	185.485240
min	0.000000
25%	2.000000
50%	10.000000
75%	35.000000
max	2700.000000

Overall Homeless - Native Hawaiian or Other Pacific Islander \	
count	385.000000
mean	27.823377
std	138.238208
min	0.000000
25%	0.000000
50%	2.000000
75%	9.000000
max	1683.000000

Overall Homeless - Multiple Races	
count	385.000000
mean	101.880519
std	356.021559
min	0.000000
25%	13.000000
50%	28.000000
75%	75.000000
max	4888.000000

[8 rows x 21 columns]

```
[78]: eviction.describe()
```

```
[78]:      filings_2020
count  598456.000000
mean    4.744833
std     86.956711
min     0.000000
25%     0.000000
50%     1.000000
75%     3.000000
max    14556.000000
```

```
[79]: max = eviction["filings_2020"].idxmax()
eviction.loc[[max]] # this makes sense, no need to remove
```

```
[79]:      city      type  month  filings_2020
460830  New York, NY  Zip Code  01/2020      14556
```

```
[80]: ## data merging (1 point)
```

```
[81]: # remove data from eviction that is not from 2023
eviction_2023 = eviction[eviction['month'].str.contains('2023')]
eviction_2023["month"].unique()
```

```
[81]: array(['01/2023', '02/2023', '03/2023', '04/2023', '05/2023', '06/2023',
        '07/2023', '08/2023', '09/2023', '10/2023', '11/2023', '12/2023'],
        dtype=object)
```

```
[82]: eviction_city = eviction_2023.groupby("city").mean(numeric_only=True)
eviction_city = eviction_city.reset_index()
eviction_city['state'] = eviction_city['city'].str[-2:]
eviction_city["state"][10] = "FL" # ft. lauderdale was mislabeled
eviction_city["state"][20] = "FL" # miami was mislabeled
eviction_city["state"][26] = "FL" # palm beach was mislabeled
eviction_city
```

```
/var/folders/nw/5zcrqdxs7c57b12ptv8284p80000gn/T/ipykernel_1500/372597069.py:4:
SettingWithCopyWarning:
```

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

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
/var/folders/nw/5zcrqdxs7c57b12ptv8284p80000gn/T/ipykernel_1500/372597069.py:5:
SettingWithCopyWarning:
```

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

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
/var/folders/nw/5zcrqdxs7c57b12ptv8284p80000gn/T/ipykernel_1500/372597069.py:6:  
SettingWithCopyWarning:
```

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

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
[82]:
```

	city	filings_2020	state
0	Albuquerque, NM	4.560734	NM
1	Austin, TX	8.560688	TX
2	Boston, MA	2.673019	MA
3	Bridgeport, CT	1.642178	CT
4	Charleston, SC	8.527500	SC
5	Cincinnati, OH	4.695301	OH
6	Cleveland, OH	3.372671	OH
7	Columbus, OH	6.018744	OH
8	Dallas, TX	4.892802	TX
9	Fort Worth, TX	5.956454	TX
10	Ft. Lauderdale	3.701754	FL
11	Gainesville, FL	3.346045	FL
12	Greenville, SC	10.136425	SC
13	Hartford, CT	2.340749	CT
14	Houston, TX	5.739814	TX
15	Indianapolis, IN	8.988517	IN
16	Jacksonville, FL	5.681818	FL
17	Kansas City, MO	3.308114	MO
18	Las Vegas, NV	8.877954	NV
19	Memphis, TN	10.232667	TN
20	Miami	2.313795	FL
21	Milwaukee, WI	3.814631	WI
22	Minneapolis-Saint Paul, MN	2.767266	MN
23	Nashville, TN	19.700893	TN
24	New Orleans, LA	2.615489	LA
25	New York, NY	36.728682	NY
26	Palm Beach	2.003788	FL
27	Philadelphia, PA	2.681133	PA
28	Phoenix, AZ	35.612973	AZ
29	Pittsburgh, PA	7.492009	PA
30	Providence, RI	3.785959	RI
31	Richmond, VA	42.004167	VA

32	South Bend, IN	2.735944	IN
33	St Louis, MO	3.816227	MO
34	Tampa, FL	2.764184	FL
35	Wilmington, DE	4.833904	DE

```
[83]: PIT_sliced_state = PIT_sliced[PIT_sliced['CoC Name'].str.contains('State')]
PIT_sliced_state['state'] = PIT_sliced_state['CoC Number'].str[0:2]
PIT_sliced_state.head()
```

```
/var/folders/nw/5zcrqdxs7c57b12ptv8284p80000gn/T/ipykernel_1500/2207987685.py:2:
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: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
[83]:
```

	CoC Number	CoC Name	CoC Category	\
1	AK-501	Alaska Balance of State CoC	Largely Rural CoC	
9	AL-507	Alabama Balance of State CoC	Largely Rural CoC	
12	AR-503	Arkansas Balance of State CoC	Largely Rural CoC	
15	AZ-500	Arizona Balance of State CoC	Largely Rural CoC	
62	CO-500	Colorado Balance of State CoC	Largely Rural CoC	

	Overall Homeless	Overall Homeless - Under 18	\
1	854.0	176.0	
9	283.0	77.0	
12	871.0	91.0	
15	2386.0	295.0	
62	2201.0	203.0	

	Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34	\
1	66.0	124.0	
9	25.0	53.0	
12	66.0	161.0	
15	117.0	317.0	
62	120.0	394.0	

	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54	\
1	190.0	144.0	
9	60.0	38.0	
12	196.0	190.0	
15	447.0	473.0	
62	495.0	460.0	

	Overall Homeless - Age 55 to 64 ... \
1	123.0 ...
9	13.0 ...
12	133.0 ...
15	457.0 ...
62	380.0 ...

	Overall Homeless - Gender Questioning \
1	0.0
9	4.0
12	3.0
15	5.0
62	2.0

	Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \
1	717.0
9	258.0
12	832.0
15	1798.0
62	1577.0

	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White \
1	137.0	316.0
9	25.0	91.0
12	39.0	619.0
15	588.0	1662.0
62	624.0	1660.0

	Overall Homeless - Black, African American, or African \
1	25.0
9	170.0
12	233.0
15	142.0
62	90.0

	Overall Homeless - Asian or Asian American \
1	10.0
9	0.0
12	2.0
15	14.0
62	8.0

	Overall Homeless - American Indian, Alaska Native, or Indigenous \
1	396.0
9	1.0
12	3.0
15	266.0

62 217.0

```
Overall Homeless - Native Hawaiian or Other Pacific Islander \
1 16.0
9 0.0
12 5.0
15 16.0
62 10.0
```

```
Overall Homeless - Multiple Races state
1 91.0 AK
9 21.0 AL
12 9.0 AR
15 286.0 AZ
62 216.0 CO
```

[5 rows x 25 columns]

```
[84]: merge = pd.merge(PIT_sliced_state, eviction_city, on = "state")
merge.head()
```

```
[84]: CoC Number CoC Name CoC Category \
0 AZ-500 Arizona Balance of State CoC Largely Rural CoC
1 CT-505 Connecticut Balance of State CoC Largely Suburban CoC
2 CT-505 Connecticut Balance of State CoC Largely Suburban CoC
3 DE-500 Delaware Statewide CoC Largely Suburban CoC
4 IN-502 Indiana Balance of State CoC Largely Rural CoC
```

```
Overall Homeless Overall Homeless - Under 18 \
0 2386.0 295.0
1 2418.0 454.0
2 2418.0 454.0
3 1245.0 335.0
4 4398.0 923.0
```

```
Overall Homeless - Age 18 to 24 Overall Homeless - Age 25 to 34 \
0 117.0 317.0
1 190.0 378.0
2 190.0 378.0
3 66.0 195.0
4 238.0 709.0
```

```
Overall Homeless - Age 35 to 44 Overall Homeless - Age 45 to 54 \
0 447.0 473.0
1 430.0 393.0
2 430.0 393.0
3 189.0 193.0
```

4	848.0	789.0
Overall Homeless - Age 55 to 64 ... \		
0	457.0 ...	
1	431.0 ...	
2	431.0 ...	
3	211.0 ...	
4	704.0 ...	
Overall Homeless - Hispanic/Latin(o)(a)(x) Overall Homeless - White \		
0	588.0	1662.0
1	730.0	1381.0
2	730.0	1381.0
3	107.0	385.0
4	221.0	3018.0
Overall Homeless - Black, African American, or African \		
0	142.0	
1	832.0	
2	832.0	
3	773.0	
4	1113.0	
Overall Homeless - Asian or Asian American \		
0	14.0	
1	12.0	
2	12.0	
3	4.0	
4	19.0	
Overall Homeless - American Indian, Alaska Native, or Indigenous \		
0	266.0	
1	33.0	
2	33.0	
3	3.0	
4	20.0	
Overall Homeless - Native Hawaiian or Other Pacific Islander \		
0	16.0	
1	6.0	
2	6.0	
3	2.0	
4	30.0	
Overall Homeless - Multiple Races state city filings_2020		
0	286.0 AZ	Phoenix, AZ 35.612973
1	154.0 CT	Bridgeport, CT 1.642178

2	154.0	CT	Hartford, CT	2.340749
3	78.0	DE	Wilmington, DE	4.833904
4	198.0	IN	Indianapolis, IN	8.988517

[5 rows x 27 columns]

[85]: PIT\_sliced

[85]:	CoC Number	CoC Name \
0	AK-500	Anchorage CoC
1	AK-501	Alaska Balance of State CoC
2	AL-500	Birmingham/Jefferson, St. Clair, Shelby Counti...
3	AL-501	Mobile City & County/Baldwin County CoC
4	AL-502	Florence/Northwest Alabama CoC
..	...	...
380	WV-500	Wheeling, Weirton Area CoC
381	WV-501	Huntington/Cabell, Wayne Counties CoC
382	WV-503	Charleston/Kanawha, Putnam, Boone, Clay Counti...
383	WV-508	West Virginia Balance of State CoC
384	WY-500	Wyoming Statewide CoC

	CoC Category	Overall Homeless	Overall Homeless - Under 18 \
0	Other Largely Urban CoC	1760.0	185.0
1	Largely Rural CoC	854.0	176.0
2	Largely Suburban CoC	847.0	67.0
3	Other Largely Urban CoC	670.0	110.0
4	Largely Rural CoC	195.0	63.0
..	...	...	...
380	Largely Rural CoC	113.0	37.0
381	Largely Rural CoC	244.0	18.0
382	Largely Suburban CoC	293.0	20.0
383	Largely Rural CoC	766.0	50.0
384	Largely Rural CoC	532.0	37.0

	Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34 \
0	161.0	377.0
1	66.0	124.0
2	42.0	127.0
3	19.0	78.0
4	9.0	42.0
..	...	...
380	3.0	23.0
381	19.0	51.0
382	40.0	53.0
383	74.0	168.0
384	62.0	73.0



	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54 \
0	419.0	315.0
1	190.0	144.0
2	182.0	180.0
3	156.0	120.0
4	36.0	23.0
..	...	...
380	24.0	16.0
381	72.0	53.0
382	66.0	72.0
383	209.0	148.0
384	116.0	126.0

	Overall Homeless - Age 55 to 64 ... \
0	223.0 ...
1	123.0 ...
2	187.0 ...
3	140.0 ...
4	16.0 ...
..	... ..
380	9.0 ...
381	25.0 ...
382	28.0 ...
383	78.0 ...
384	75.0 ...

	Overall Homeless - Gender that is not Singularly Female or Male \
0	3.0
1	2.0
2	1.0
3	2.0
4	0.0
..	...
380	0.0
381	0.0
382	0.0
383	5.0
384	1.0

	Overall Homeless - Gender Questioning \
0	2.0
1	0.0
2	0.0
3	0.0
4	0.0
..	...
380	0.0

381	0.0
382	0.0
383	4.0
384	1.0

Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \	
0	1638.0
1	717.0
2	822.0
3	655.0
4	186.0
..	...
380	113.0
381	238.0
382	292.0
383	750.0
384	462.0

Overall Homeless - Hispanic/Latin(o)(a)(x)		Overall Homeless - White \
0	122.0	480.0
1	137.0	316.0
2	25.0	295.0
3	15.0	281.0
4	9.0	130.0
..	...	...
380	0.0	83.0
381	6.0	206.0
382	1.0	230.0
383	16.0	691.0
384	70.0	419.0

Overall Homeless - Black, African American, or African \	
0	156.0
1	25.0
2	518.0
3	351.0
4	48.0
..	...
380	17.0
381	19.0
382	42.0
383	59.0
384	47.0

Overall Homeless - Asian or Asian American \	
0	28.0
1	10.0

2	1.0
3	2.0
4	0.0
..	...
380	0.0
381	2.0
382	0.0
383	0.0
384	14.0

	Overall Homeless - American Indian, Alaska Native, or Indigenous \
0	755.0
1	396.0
2	14.0
3	11.0
4	2.0
..	...
380	4.0
381	3.0
382	0.0
383	4.0
384	22.0

	Overall Homeless - Native Hawaiian or Other Pacific Islander \
0	83.0
1	16.0
2	2.0
3	6.0
4	0.0
..	...
380	0.0
381	1.0
382	0.0
383	1.0
384	2.0

	Overall Homeless - Multiple Races
0	258.0
1	91.0
2	17.0
3	19.0
4	15.0
..	...
380	9.0
381	13.0
382	21.0
383	11.0

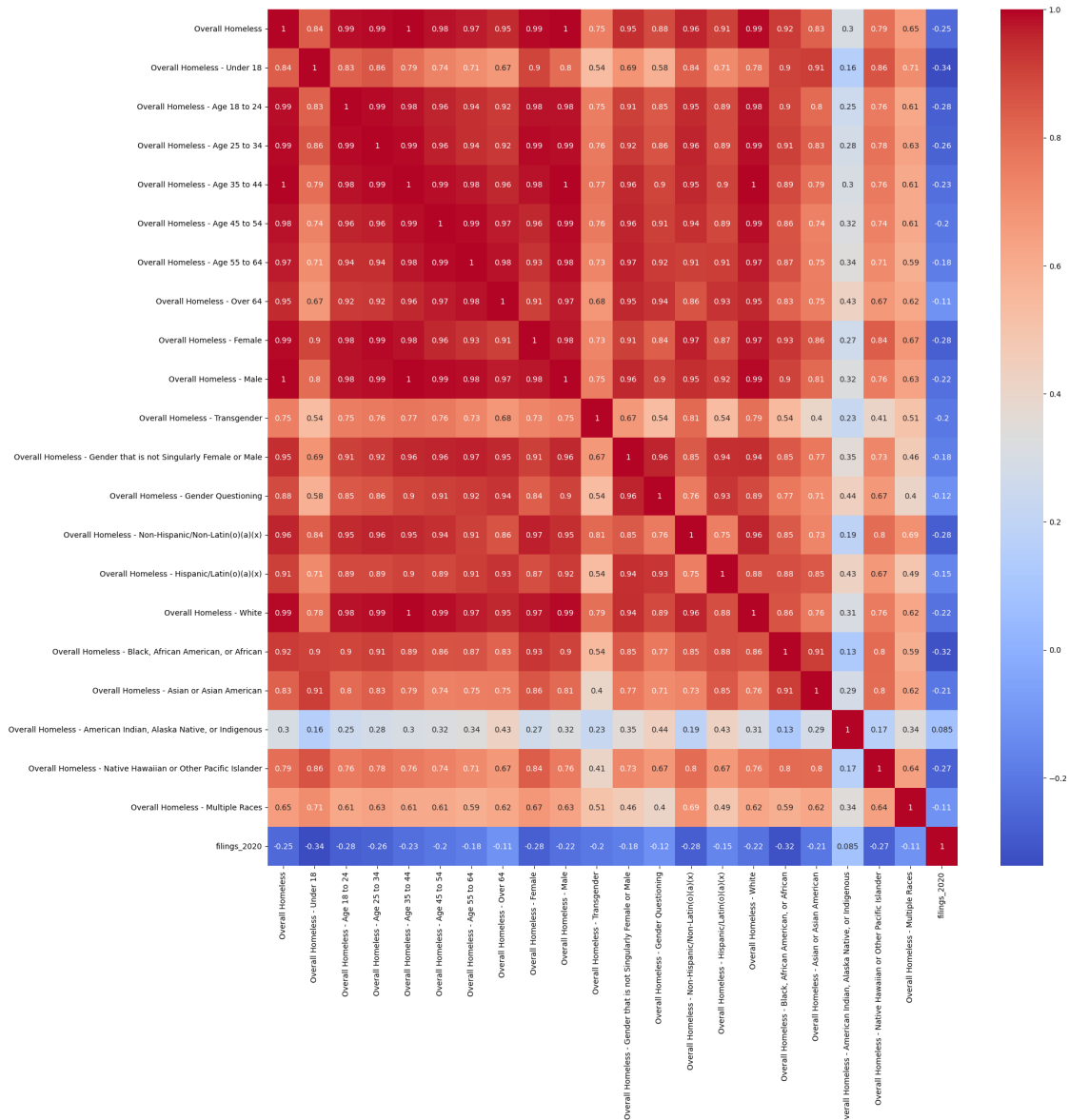
[385 rows x 24 columns]

[86]: `## data transformation, normalization, and cleaning (1 point)`

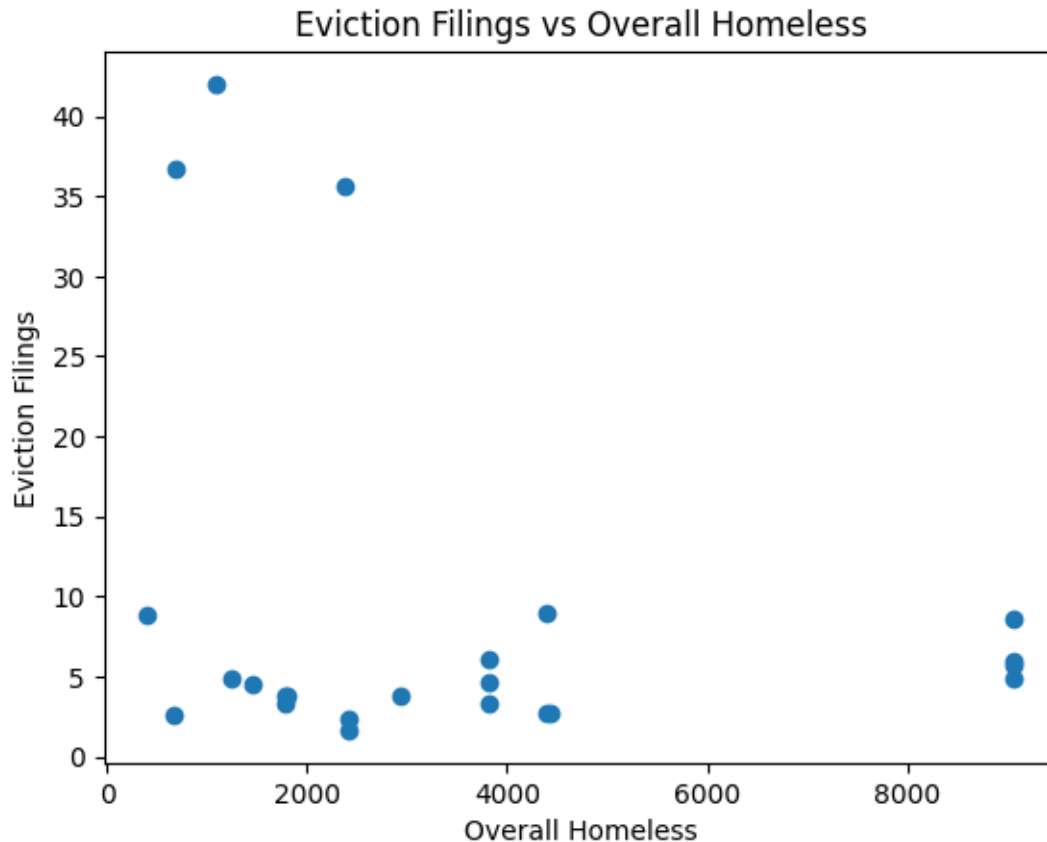
[87]: `# sets up data for corrpplots and purely quantitative analysis`  
`merge_numeric = merge.select_dtypes(include=[np.number])`

[88]: `## Exploratory data visualization (1 point)`

[89]: `plt.figure(figsize=(20,20))`  
`sns.heatmap(data=merge_numeric.corr(), annot=True, cmap='coolwarm')`  
`plt.show()`



```
[90]: import matplotlib.pyplot as plt
plt.scatter(x=merge["Overall Homeless"], y=merge["filings_2020"])
plt.ylabel("Eviction Filings")
plt.xlabel("Overall Homeless")
plt.title("Eviction Filings vs Overall Homeless")
plt.show()
```



High multicollinearity & low n, unfit for regression or advanced models.

### 1.3 Part 2: Answer questions from the proposals (8 points)

Each plot should be followed by a paragraph of explanation and observation.

#### 1.3.1 Creator: Ivy Nangalia

**Question:** What demographics of people are more likely to be homeless today?

```
[91]: means = PIT_sliced.describe()[1:2]
race_cols = means.columns[15:22]
```

```
total_sum = means[race_cols].sum(axis=1)
race_percent = means[race_cols].div(total_sum, axis=0)
race_percent.reset_index()
```

```
[91]: index Overall Homeless - White \
0 mean 0.4974

Overall Homeless - Black, African American, or African \
0 0.373025

Overall Homeless - Asian or Asian American \
0 0.017722

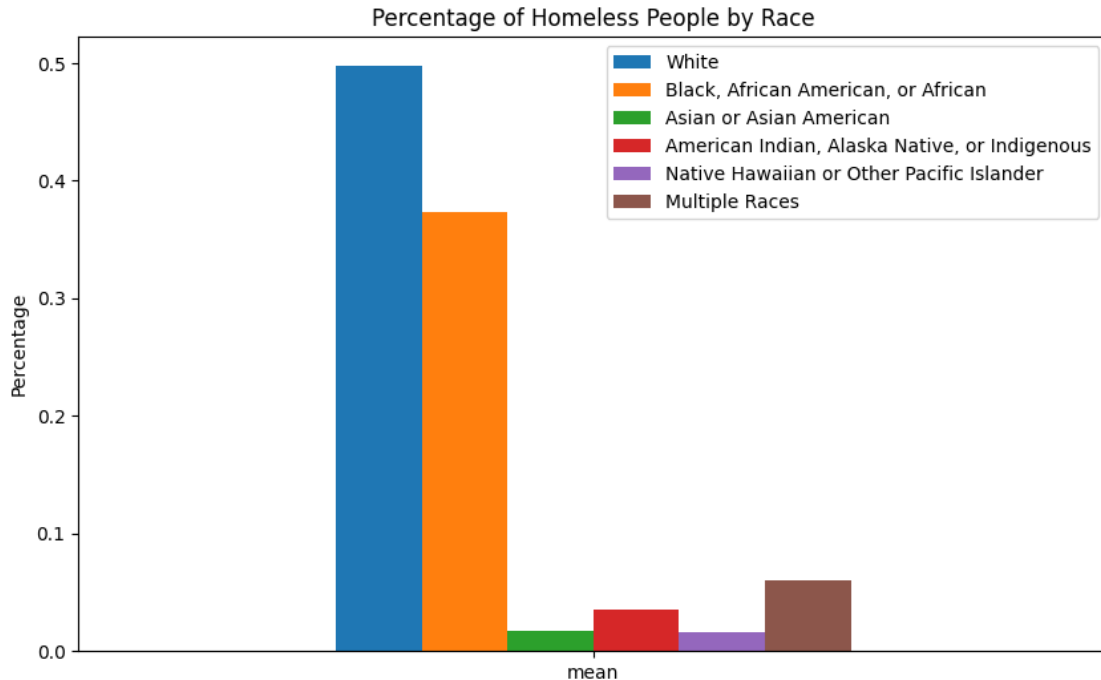
Overall Homeless - American Indian, Alaska Native, or Indigenous \
0 0.035394

Overall Homeless - Native Hawaiian or Other Pacific Islander \
0 0.016402

Overall Homeless - Multiple Races
0 0.060058
```

```
[92]: # renaming columns
race_percent = race_percent.rename(columns={'Overall Homeless - White': 'White',
                                             'Overall Homeless - Black, African_
↳American, or African': 'Black, African American, or African',
                                             'Overall Homeless - Asian or Asian_
↳American': 'Asian or Asian American',
                                             'Overall Homeless - American_
↳Indian, Alaska Native, or Indigenous': 'American Indian, Alaska Native, or_
↳Indigenous',
                                             'Overall Homeless - Native Hawaiian_
↳or Other Pacific Islander': 'Native Hawaiian or Other Pacific Islander',
                                             'Overall Homeless - Multiple Races':
↳ 'Multiple Races'})
```

```
[93]: race_percent.plot(kind='bar', figsize=(10, 6))
plt.ylabel('Percentage')
plt.title('Percentage of Homeless People by Race')
plt.xticks(rotation=0)
plt.show()
```



**Answer:** It seems that White people have the highest percentage of the homeless population (49.7%) followed closely by Black people (37.3%). White people making up the largest share of the homeless population makes sense since the majority of the American population is White. However, only 13.7% of the population is Black, which implies some disproportionate factors affecting Black people and their housing security. This makes sense considering the long history of racism and racist policies enacted in the United States. Moreover, I'd argue that White people are less likely to be homeless considering that they make up 75.3% of the population but only 49.7% of the homeless population, implying some systemic factors that improve the housing security of White people as compared to others.

Note: the population data is from the US Census.

### 1.3.2 Interpreter 2: Ximing Sun

**Question:** What trends do we see in racial segregation?

**Answer:**

```
[94]: racial = merge.iloc[:, 16:26]
      racial
```

```
[94]: Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \
0      1798.0
1      1688.0
2      1688.0
3      1138.0
```

4	4177.0
5	4177.0
6	662.0
7	2906.0
8	1725.0
9	1725.0
10	620.0
11	349.0
12	604.0
13	3655.0
14	3655.0
15	3655.0
16	1404.0
17	5676.0
18	5676.0
19	5676.0
20	5676.0
21	1029.0
22	2674.0

	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White \
0	588.0	1662.0
1	730.0	1381.0
2	730.0	1381.0
3	107.0	385.0
4	221.0	3018.0
5	221.0	3018.0
6	16.0	273.0
7	1526.0	2153.0
8	67.0	1347.0
9	67.0	1347.0
10	828.0	900.0
11	61.0	361.0
12	83.0	553.0
13	168.0	3027.0
14	168.0	3027.0
15	168.0	3027.0
16	406.0	1140.0
17	3389.0	6533.0
18	3389.0	6533.0
19	3389.0	6533.0
20	3389.0	6533.0
21	59.0	653.0
22	266.0	1933.0

	Overall Homeless - Black, African American, or African \
0	142.0



1	832.0
2	832.0
3	773.0
4	1113.0
5	1113.0
6	385.0
7	1917.0
8	328.0
9	328.0
10	136.0
11	16.0
12	87.0
13	580.0
14	580.0
15	580.0
16	436.0
17	2095.0
18	2095.0
19	2095.0
20	2095.0
21	366.0
22	603.0

Overall Homeless - Asian or Asian American \

0	14.0
1	12.0
2	12.0
3	4.0
4	19.0
5	19.0
6	1.0
7	78.0
8	5.0
9	5.0
10	4.0
11	3.0
12	1.0
13	7.0
14	7.0
15	7.0
16	12.0
17	59.0
18	59.0
19	59.0
20	59.0
21	6.0
22	35.0

	Overall Homeless - American Indian, Alaska Native, or Indigenous \
0	266.0
1	33.0
2	33.0
3	3.0
4	20.0
5	20.0
6	4.0
7	19.0
8	14.0
9	14.0
10	339.0
11	17.0
12	3.0
13	34.0
14	34.0
15	34.0
16	33.0
17	142.0
18	142.0
19	142.0
20	142.0
21	13.0
22	186.0

	Overall Homeless - Native Hawaiian or Other Pacific Islander \
0	16.0
1	6.0
2	6.0
3	2.0
4	30.0
5	30.0
6	1.0
7	40.0
8	23.0
9	23.0
10	6.0
11	0.0
12	1.0
13	11.0
14	11.0
15	11.0
16	4.0
17	33.0
18	33.0
19	33.0

```

20                                     33.0
21                                     0.0
22                                     10.0

```

```

Overall Homeless - Multiple Races state      city
0      286.0    AZ      Phoenix, AZ
1      154.0    CT      Bridgeport, CT
2      154.0    CT      Hartford, CT
3       78.0    DE      Wilmington, DE
4      198.0    IN      Indianapolis, IN
5      198.0    IN      South Bend, IN
6       14.0    LA      New Orleans, LA
7      225.0    MA      Boston, MA
8       75.0    MO      Kansas City, MO
9       75.0    MO      St Louis, MO
10     63.0    NM      Albuquerque, NM
11     13.0    NV      Las Vegas, NV
12     42.0    NY      New York, NY
13     164.0    OH      Cincinnati, OH
14     164.0    OH      Cleveland, OH
15     164.0    OH      Columbus, OH
16     185.0    RI      Providence, RI
17     203.0    TX      Austin, TX
18     203.0    TX      Dallas, TX
19     203.0    TX      Fort Worth, TX
20     203.0    TX      Houston, TX
21      50.0    VA      Richmond, VA
22     173.0    WI      Milwaukee, WI

```

```
[95]: ## Add more cells if your group has more than two interpreters
```

```
[96]: racial_categories = [
    'Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x)',
    'Overall Homeless - Hispanic/Latin(o)(a)(x)',
    'Overall Homeless - White',
    'Overall Homeless - Black, African American, or African',
    'Overall Homeless - Asian or Asian American',
    'Overall Homeless - American Indian, Alaska Native, or Indigenous',
    'Overall Homeless - Native Hawaiian or Other Pacific Islander',
    'Overall Homeless - Multiple Races'
]

fig, ax = plt.subplots(figsize=(12, 8))
for category in racial_categories:
    ax.bar(racial['city'], racial[category], label=category,
           bottom=racial[racial_categories].cumsum(axis=1)[category] - racial[category])

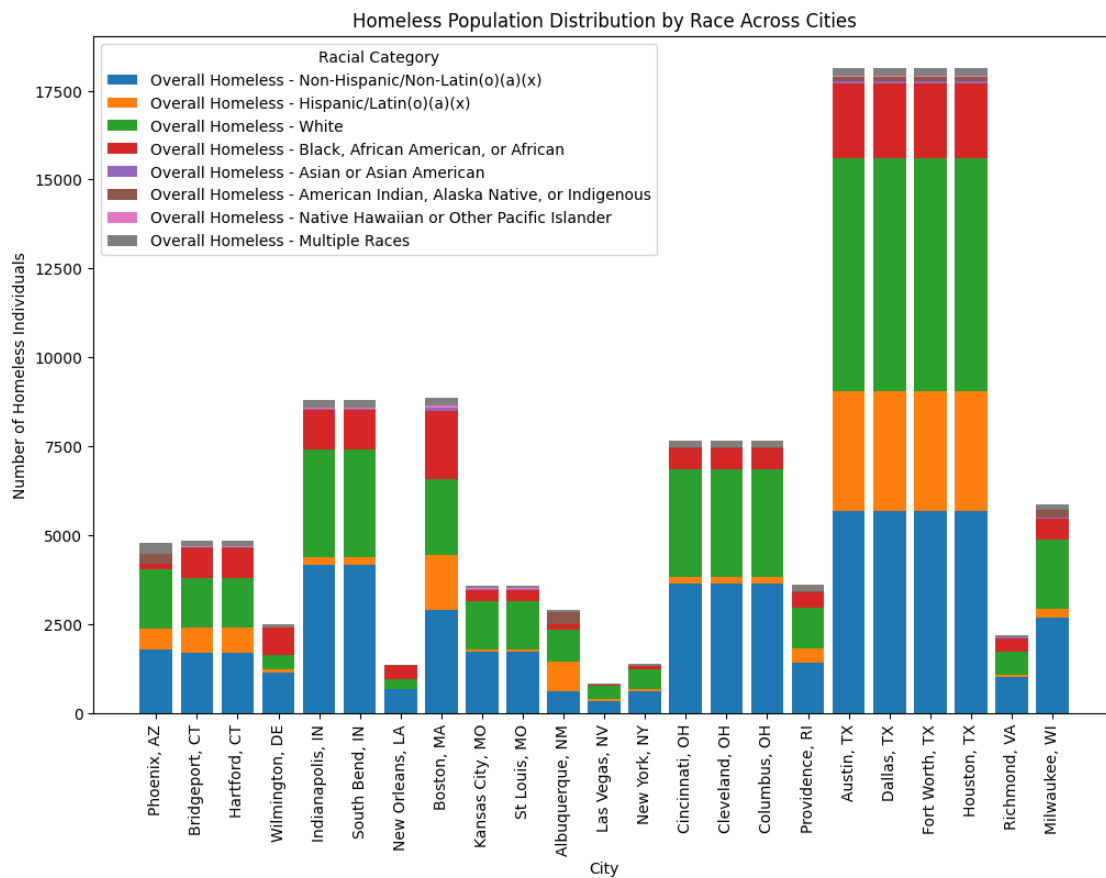
```

```

ax.set_xlabel("City")
ax.set_ylabel("Number of Homeless Individuals")
ax.set_title("Homeless Population Distribution by Race Across Cities")
ax.legend(title="Racial Category")
plt.xticks(rotation=90)

plt.show()

```



```

[97]: PIT_sliced["state"] = PIT_sliced["CoC Number"].apply(lambda x: x[:2])

```

```

racial_categories = [
    'Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x)',
    'Overall Homeless - Hispanic/Latin(o)(a)(x)',
    'Overall Homeless - White',
    'Overall Homeless - Black, African American, or African',
    'Overall Homeless - Asian or Asian American',
    'Overall Homeless - American Indian, Alaska Native, or Indigenous',
    'Overall Homeless - Native Hawaiian or Other Pacific Islander',
    'Overall Homeless - Multiple Races',
]

```

```

    "state"
]

race_country = PIT_sliced.iloc[:, 16:26]
race_country_state = race_country.groupby("state").mean()

#state validation:
states = ["AK", "AL", "AR", "AZ", "CA", "CO", "CT", "DC",
          "DE", "FL", "GA", "HI", "IA", "ID", "IL", "IN", "KS",
          "KY", "LA", "MA", "MD", "ME", "MI", "MN", "MO", "MS", "MT", "NC",
          "ND", "NE", "NH", "NJ", "NM", "NV", "NY", "OH", "OK", "OR", "PA",
          ↪ "RI",
          "SC", "SD", "TN", "TX", "UT", "VA", "VT", "WA", "WI", "WV", "WY"]
for state in race_country_state.index:
    if state not in states:
        race_country_state.drop(state, inplace=True)

len(race_country_state)
race_country_state["total"] = race_country_state.sum(axis=1)

race_country_state_pct = race_country_state.drop("total", axis=1).
    ↪div(race_country_state["total"], axis=0) * 100

hisp_pct = pd.DataFrame(race_country_state_pct.iloc[:, 1])

import plotly.express as px

state_names = {
    "AK": "Alaska", "AL": "Alabama", "AR": "Arkansas", "AZ": "Arizona",
    "CA": "California", "CO": "Colorado", "CT": "Connecticut", "DC": "District
    ↪of Columbia",
    "DE": "Delaware", "FL": "Florida", "GA": "Georgia", "HI": "Hawaii",
    "IA": "Iowa", "ID": "Idaho", "IL": "Illinois", "IN": "Indiana",
    "KS": "Kansas", "KY": "Kentucky", "LA": "Louisiana", "MA": "Massachusetts",
    "MD": "Maryland", "ME": "Maine", "MI": "Michigan", "MN": "Minnesota",
    "MO": "Missouri", "MS": "Mississippi", "MT": "Montana", "NC": "North
    ↪Carolina",
    "ND": "North Dakota", "NE": "Nebraska", "NH": "New Hampshire", "NJ": "New
    ↪Jersey",
    "NM": "New Mexico", "NV": "Nevada", "NY": "New York", "OH": "Ohio",
    "OK": "Oklahoma", "OR": "Oregon", "PA": "Pennsylvania", "RI": "Rhode
    ↪Island",
    "SC": "South Carolina", "SD": "South Dakota", "TN": "Tennessee", "TX":
    ↪ "Texas",
    "UT": "Utah", "VA": "Virginia", "VT": "Vermont", "WA": "Washington",

```

```

    "WI": "Wisconsin", "WV": "West Virginia", "WY": "Wyoming"
}

hisp_data = hisp_pct.reset_index()
hisp_data["state_name"] = hisp_data["state"].map(state_names)

fig = px.choropleth(
    hisp_data,
    locations="state",
    locationmode="USA-states",
    color="Overall Homeless - Hispanic/Latin(o)(a)(x)",
    color_continuous_scale="Viridis",
    range_color=[0, 30],
    title="Percentage of Hispanic/Latin(o)(a)(x) Homeless Population by State",
    labels={"Overall Homeless - Hispanic/Latin(o)(a)(x)": "%"}
)

fig.update_layout(
    title_x=0.5,
    geo_scope="usa",
    width=1200,
    height=800
)

fig.show()

```

[142]:

```

[98]: # get a random sample of 40 CoC numbers and their corresponding homeless racial_
      ↪ demographics, and plot them

np.random.seed(89)
racial = PIT_sliced.copy().sample(40)

racial_categories = [
    'Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x)',
    'Overall Homeless - Hispanic/Latin(o)(a)(x)',
    'Overall Homeless - White',
    'Overall Homeless - Black, African American, or African',
    'Overall Homeless - Asian or Asian American',
    'Overall Homeless - American Indian, Alaska Native, or Indigenous',
    'Overall Homeless - Native Hawaiian or Other Pacific Islander',
    'Overall Homeless - Multiple Races'
]

fig, ax = plt.subplots(figsize=(12, 8))
for category in racial_categories:

```

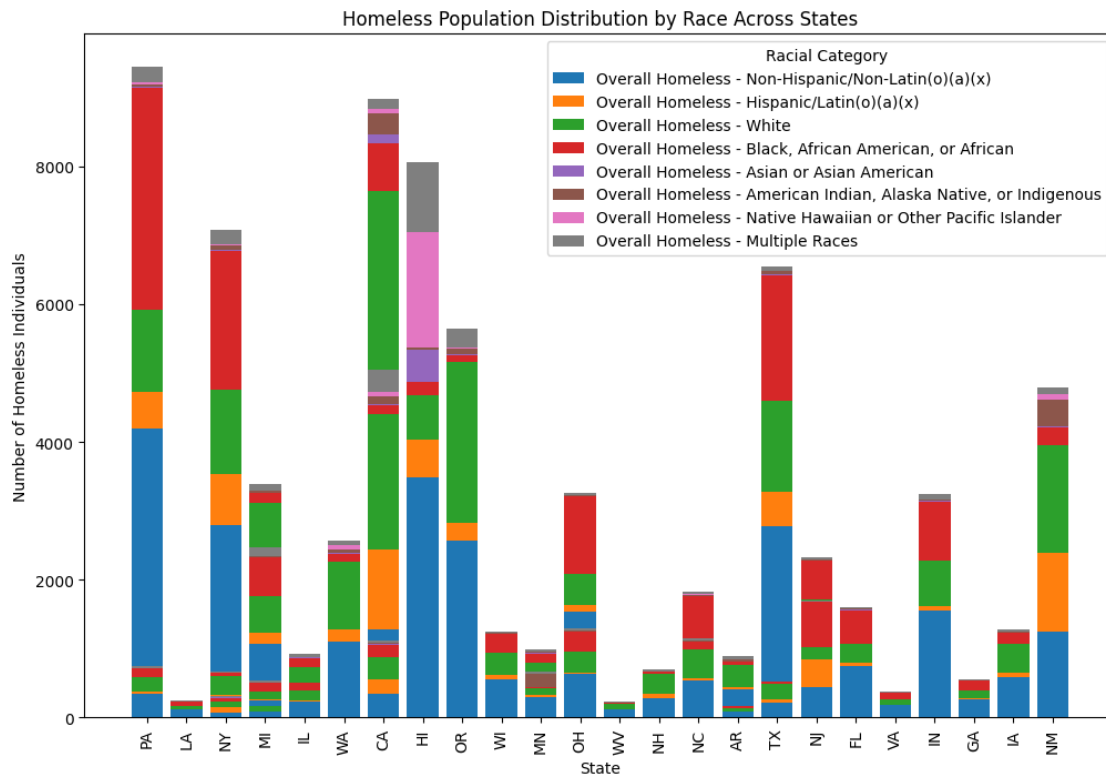
```

ax.bar(racial["state"], racial[category], label=category,
      bottom=racial[racial_categories].cumsum(axis=1)[category] - racial[category])

ax.set_xlabel("State")
ax.set_ylabel("Number of Homeless Individuals")
ax.set_title("Homeless Population Distribution by Race Across States")
ax.legend(title="Racial Category")
plt.xticks(rotation=90)

plt.show()

```



```
[99]: racial.head(15)
```

```

[99]:   CoC Number      CoC Name \
303    PA-500      Philadelphia CoC
147    LA-507  Alexandria/Central Louisiana CoC
269    NY-523  Glens Falls, Saratoga Springs/Saratoga, Washin...
189    MI-518      Livingston County CoC
129    IL-517  Aurora, Elgin/Kane County CoC
374    WA-504  Everett/Snohomish County CoC
58     CA-611  Oxnard, San Buenaventura/Ventura County CoC
108    HI-501  Honolulu City and County CoC

```

295	OR-500	Eugene, Springfield/Lane County CoC
124	IL-512	Bloomington/Central Illinois CoC
379	WI-503	Madison/Dane County CoC
194	MN-502	Rochester/Southeast Minnesota CoC
280	OH-502	Cleveland/Cuyahoga County CoC
380	WV-500	Wheeling, Weirton Area CoC
232	NH-502	Nashua/Hillsborough County CoC

	CoC Category	Overall Homeless	Overall Homeless - Under 18 \
303	Major City CoC	4725.0	825.0
147	Largely Rural CoC	122.0	15.0
269	Largely Suburban CoC	332.0	30.0
189	Largely Suburban CoC	88.0	36.0
129	Largely Suburban CoC	461.0	81.0
374	Largely Suburban CoC	1285.0	202.0
58	Other Largely Urban CoC	2441.0	141.0
108	Largely Suburban CoC	4028.0	583.0
295	Other Largely Urban CoC	2824.0	230.0
124	Largely Rural CoC	256.0	67.0
379	Other Largely Urban CoC	624.0	148.0
194	Largely Rural CoC	496.0	144.0
280	Largely Suburban CoC	1629.0	281.0
380	Largely Rural CoC	113.0	37.0
232	Largely Suburban CoC	348.0	99.0

	Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34 \
303	409.0	778.0
147	3.0	15.0
269	27.0	79.0
189	12.0	7.0
129	32.0	70.0
374	70.0	235.0
58	100.0	390.0
108	328.0	492.0
295	199.0	442.0
124	32.0	38.0
379	33.0	105.0
194	39.0	76.0
280	95.0	278.0
380	3.0	23.0
232	15.0	51.0

	Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54 \
303	968.0	753.0
147	39.0	24.0
269	81.0	48.0
189	20.0	8.0



129	79.0	81.0
374	297.0	202.0
58	513.0	509.0
108	681.0	727.0
295	564.0	589.0
124	32.0	30.0
379	120.0	82.0
194	96.0	63.0
280	309.0	248.0
380	24.0	16.0
232	69.0	40.0

Overall Homeless - Age 55 to 64 ... \

303	768.0	...
147	22.0	...
269	44.0	...
189	4.0	...
129	77.0	...
374	211.0	...
58	498.0	...
108	895.0	...
295	552.0	...
124	24.0	...
379	83.0	...
194	51.0	...
280	310.0	...
380	9.0	...
232	43.0	...

Overall Homeless - Gender Questioning \

303	8.0
147	0.0
269	0.0
189	1.0
129	0.0
374	2.0
58	0.0
108	2.0
295	17.0
124	0.0
379	1.0
194	0.0
280	0.0
380	0.0
232	0.0

Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \

303	4199.0
147	119.0
269	307.0
189	88.0
129	269.0
374	1106.0
58	1275.0
108	3479.0
295	2563.0
124	228.0
379	561.0
194	446.0
280	1532.0
380	113.0
232	280.0

	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White \
303	526.0	1197.0
147	3.0	45.0
269	25.0	267.0
189	0.0	88.0
129	192.0	278.0
374	179.0	981.0
58	1166.0	1969.0
108	549.0	656.0
295	261.0	2338.0
124	28.0	133.0
379	63.0	314.0
194	50.0	293.0
280	97.0	453.0
380	0.0	83.0
232	68.0	285.0

	Overall Homeless - Black, African American, or African \
303	3214.0
147	72.0
269	51.0
189	0.0
129	129.0
374	112.0
58	188.0
108	181.0
295	96.0
124	116.0
379	275.0
194	141.0
280	1133.0

380	17.0
232	34.0

Overall Homeless - Asian or Asian American \	
303	26.0
147	1.0
269	3.0
189	0.0
129	5.0
374	18.0
58	9.0
108	472.0
295	9.0
124	0.0
379	8.0
194	4.0
280	5.0
380	0.0
232	2.0

Overall Homeless - American Indian, Alaska Native, or Indigenous \	
303	32.0
147	0.0
269	2.0
189	0.0
129	10.0
374	38.0
58	57.0
108	31.0
295	85.0
124	0.0
379	9.0
194	21.0
280	2.0
380	4.0
232	1.0

Overall Homeless - Native Hawaiian or Other Pacific Islander \	
303	20.0
147	0.0
269	0.0
189	0.0
129	0.0
374	65.0
58	57.0
108	1683.0
295	14.0

124	1.0
379	1.0
194	1.0
280	4.0
380	0.0
232	4.0

	Overall Homeless - Multiple Races	state
303	236.0	PA
147	4.0	LA
269	9.0	NY
189	0.0	MI
129	39.0	IL
374	71.0	WA
58	161.0	CA
108	1005.0	HI
295	282.0	OR
124	6.0	IL
379	17.0	WI
194	36.0	MN
280	32.0	OH
380	9.0	WV
232	22.0	NH

[15 rows x 25 columns]

```
[100]: totals = race_country_state["total"]
```

```
[101]: len(race_country_state)
race_country_state["total"] = race_country_state.sum(axis=1)

race_country_state_pct = race_country_state.drop("total", axis=1).
    ↪div(race_country_state["total"], axis=0) * 100

hisp_pct = pd.DataFrame(race_country_state_pct.iloc[:, 1])
hisp_pct
#race_country_state_pct
```

```
[101]: Overall Homeless - Hispanic/Latin(o)(a)(x)
state
AK          2.477047
AL          0.877724
AR          1.293599
AZ          6.894009
CA          9.227035
CO          6.523998
CT          7.728027
```

DC	2.453271
DE	2.148594
FL	4.238197
GA	1.386855
HI	3.338422
IA	2.318130
ID	5.265448
IL	7.499791
IN	1.200765
KS	3.060594
KY	1.138271
LA	0.978227
MA	8.752155
MD	1.867008
ME	1.285815
MI	1.803379
MN	3.252711
MO	1.745447
MS	0.712831
MT	2.330119
NC	1.358417
ND	2.136480
NE	3.330626
NH	2.201966
NJ	7.526306
NM	12.844872
NV	4.208978
NY	12.029312
OH	1.174688
OK	2.570998
OR	3.204746
PA	3.456515
RI	5.607735
SC	1.116457
SD	1.716069
TN	0.784048
TX	7.787559
UT	5.838080
VA	2.558793
VT	1.168437
WA	3.781745
WI	2.288624
WV	0.406073
WY	3.289474

```
[102]: import plotly.express as px
```

```

state_names = {
    "AK": "Alaska", "AL": "Alabama", "AR": "Arkansas", "AZ": "Arizona",
    "CA": "California", "CO": "Colorado", "CT": "Connecticut", "DC": "District of Columbia",
    "DE": "Delaware", "FL": "Florida", "GA": "Georgia", "HI": "Hawaii",
    "IA": "Iowa", "ID": "Idaho", "IL": "Illinois", "IN": "Indiana",
    "KS": "Kansas", "KY": "Kentucky", "LA": "Louisiana", "MA": "Massachusetts",
    "MD": "Maryland", "ME": "Maine", "MI": "Michigan", "MN": "Minnesota",
    "MO": "Missouri", "MS": "Mississippi", "MT": "Montana", "NC": "North Carolina",
    "ND": "North Dakota", "NE": "Nebraska", "NH": "New Hampshire", "NJ": "New Jersey",
    "NM": "New Mexico", "NV": "Nevada", "NY": "New York", "OH": "Ohio",
    "OK": "Oklahoma", "OR": "Oregon", "PA": "Pennsylvania", "RI": "Rhode Island",
    "SC": "South Carolina", "SD": "South Dakota", "TN": "Tennessee", "TX": "Texas",
    "UT": "Utah", "VA": "Virginia", "VT": "Vermont", "WA": "Washington",
    "WI": "Wisconsin", "WV": "West Virginia", "WY": "Wyoming"
}

hisp_data = hisp_pct.reset_index()
hisp_data["state_name"] = hisp_data["state"].map(state_names)

fig = px.choropleth(
    hisp_data,
    locations="state",
    locationmode="USA-states",
    color="Overall Homeless - Hispanic/Latin(o)(a)(x)",
    color_continuous_scale="Viridis",
    range_color=[0, 30],
    title="Percentage of Hispanic/Latin(o)(a)(x) Homeless Population by State",
    labels={"Overall Homeless - Hispanic/Latin(o)(a)(x)": "Percentage"}
)

fig.update_layout(
    title_x=0.5,
    geo_scope="usa",
    width=1200,
    height=800
)

fig.show()

```

[103]: totals

[103]: state

AK	2614.000000
AL	826.000000
AR	1043.600000
AZ	9491.333333
CA	8245.409091
CO	7219.500000
CT	3015.000000
DC	9844.000000
DE	2490.000000
FL	2278.222222
GA	2732.000000
HI	6223.000000
IA	1768.666667
ID	2298.000000
IL	1257.578947
IN	6017.000000
KS	1213.000000
KY	3177.333333
LA	905.428571
MA	3190.166667
MD	1173.000000
ME	8516.000000
MI	899.700000
MN	1678.600000
MO	1729.500000
MS	654.666667
MT	4356.000000
NC	1625.666667
ND	1568.000000
NE	1641.333333
NH	1627.333333
NJ	1283.000000
NM	3842.000000
NV	5777.333333
NY	8600.000000
OH	2530.222222
OK	1162.000000
OR	5035.500000
PA	1569.500000
RI	3620.000000
SC	2026.500000
SD	2564.000000
TN	1843.000000
TX	4977.636364
UT	2458.000000
VA	845.125000

```

VT      3295.000000
WA      9345.333333
WI      2430.500000
WV       708.000000
WY      1064.000000
Name: total, dtype: float64

```

```
[104]: totals["state"] = totals.index
```

```
[105]: race_country_state
```

```
[105]: Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x) \
state
AK      1177.500000
AL       398.500000
AR       494.800000
AZ      3437.000000
CA      2601.090909
CO      2667.750000
CT      1041.500000
DC      4439.000000
DE       1138.000000
FL       946.000000
GA      1290.222222
HI      2696.000000
IA       802.333333
ID       907.000000
IL       440.157895
IN      2864.000000
KS       532.250000
KY      1516.333333
LA       435.000000
MA      1036.666667
MD       542.700000
ME      4039.000000
MI       417.400000
MN       730.100000
MO       804.375000
MS       318.000000
MT      1975.000000
NC       768.666667
ND       717.000000
NE       711.333333
NH       742.000000
NJ      448.375000
NM       934.000000
NV      2402.333333

```



NY	2230.958333
OH	1205.666667
OK	521.250000
OR	2195.000000
PA	676.250000
RI	1404.000000
SC	968.000000
SD	1194.000000
TN	892.600000
TX	1713.545455
UT	942.000000
VA	379.312500
VT	1570.500000
WA	3965.833333
WI	1104.000000
WV	348.250000
WY	462.000000

state	Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White \
AK	129.500000	398.000000
AL	14.500000	168.625000
AR	27.000000	339.200000
AZ	1308.666667	3003.666667
CA	1521.613636	2190.568182
CO	942.000000	2413.750000
CT	466.000000	826.500000
DC	483.000000	587.000000
DE	107.000000	385.000000
FL	193.111111	614.037037
GA	75.777778	511.444444
HI	415.500000	647.500000
IA	82.000000	562.333333
ID	242.000000	951.000000
IL	188.631579	291.315789
IN	144.500000	1842.000000
KS	74.250000	419.500000
KY	72.333333	1128.666667
LA	17.714286	168.428571
MA	558.416667	768.583333
MD	43.800000	194.500000
ME	219.000000	2044.000000
MI	32.450000	213.900000
MN	109.200000	299.900000
MO	60.375000	491.000000
MS	9.333333	143.000000
MT	203.000000	1480.000000

NC	44.166667	341.166667
ND	67.000000	352.000000
NE	109.333333	528.333333
NH	71.666667	725.000000
NJ	193.125000	244.750000
NM	987.000000	1228.000000
NV	486.333333	1661.333333
NY	2069.041667	1231.333333
OH	59.444444	657.444444
OK	59.750000	330.250000
OR	322.750000	1945.375000
PA	108.500000	368.562500
RI	406.000000	1140.000000
SC	45.250000	495.000000
SD	88.000000	372.000000
TN	28.900000	571.800000
TX	775.272727	1479.727273
UT	287.000000	935.000000
VA	43.250000	167.562500
VT	77.000000	1417.500000
WA	706.833333	2841.000000
WI	111.250000	684.750000
WV	5.750000	302.500000
WY	70.000000	419.000000

Overall Homeless - Black, African American, or African \

state	
AK	90.500000
AL	223.375000
AR	146.600000
AZ	1032.333333
CA	1212.931818
CO	607.750000
CT	558.500000
DC	4091.000000
DE	773.000000
FL	462.777778
GA	796.111111
HI	123.000000
IA	221.333333
ID	25.000000
IL	293.842105
IN	981.000000
KS	116.500000
KY	377.000000
LA	269.285714
MA	721.583333

MD	352.100000
ME	2013.000000
MI	199.050000
MN	315.200000
MO	309.125000
MS	168.333333
MT	63.000000
NC	420.416667
ND	102.000000
NE	176.000000
NH	41.333333
NJ	363.375000
NM	200.000000
NV	930.000000
NY	2741.291667
OH	529.444444
OK	123.875000
OR	194.875000
PA	361.250000
RI	436.000000
SC	465.750000
SD	63.000000
TN	310.100000
TX	878.909091
UT	109.666667
VA	214.625000
VT	133.000000
WA	782.333333
WI	398.000000
WV	34.250000
WY	47.000000

Overall Homeless - Asian or Asian American \	
state	
AK	19.000000
AL	1.500000
AR	1.400000
AZ	33.666667
CA	159.363636
CO	33.750000
CT	7.000000
DC	40.000000
DE	4.000000
FL	6.888889
GA	4.777778
HI	301.500000
IA	9.333333

ID	4.000000
IL	10.368421
IN	13.500000
KS	3.000000
KY	5.333333
LA	2.142857
MA	15.083333
MD	6.400000
ME	17.000000
MI	2.550000
MN	19.300000
MO	5.000000
MS	2.333333
MT	10.000000
NC	5.000000
ND	9.000000
NE	12.333333
NH	4.666667
NJ	5.125000
NM	12.000000
NV	57.333333
NY	36.375000
OH	4.444444
OK	3.000000
OR	19.125000
PA	4.062500
RI	12.000000
SC	3.000000
SD	7.000000
TN	3.000000
TX	22.363636
UT	13.000000
VA	8.312500
VT	18.000000
WA	57.166667
WI	12.750000
WV	0.500000
WY	14.000000

Overall Homeless - American Indian, Alaska Native, or Indigenous \	
state	
AK	575.500000
AL	4.625000
AR	10.000000
AZ	325.000000
CA	195.204545
CO	207.250000

CT	19.000000
DC	73.000000
DE	3.000000
FL	12.222222
GA	6.444444
HI	25.000000
IA	28.666667
ID	80.000000
IL	7.526316
IN	19.500000
KS	18.500000
KY	12.000000
LA	4.000000
MA	10.500000
MD	9.600000
ME	34.000000
MI	5.050000
MN	96.100000
MO	11.750000
MS	2.666667
MT	461.000000
NC	12.916667
ND	276.000000
NE	44.666667
NH	9.666667
NJ	9.312500
NM	359.500000
NV	79.000000
NY	39.416667
OH	10.111111
OK	77.625000
OR	124.500000
PA	5.812500
RI	33.000000
SC	9.500000
SD	779.000000
TN	8.500000
TX	37.818182
UT	72.666667
VA	3.750000
VT	41.500000
WA	344.166667
WI	52.750000
WV	2.750000
WY	22.000000

Overall Homeless - Native Hawaiian or Other Pacific Islander \

state	
AK	49.500000
AL	1.375000
AR	6.000000
AZ	35.333333
CA	77.568182
CO	100.500000
CT	3.500000
DC	52.000000
DE	2.000000
FL	4.333333
GA	2.222222
HI	1168.000000
IA	6.000000
ID	10.500000
IL	3.263158
IN	18.000000
KS	3.250000
KY	3.666667
LA	1.857143
MA	11.166667
MD	2.900000
ME	5.000000
MI	1.000000
MN	4.500000
MO	8.500000
MS	1.666667
MT	18.000000
NC	3.000000
ND	0.000000
NE	3.666667
NH	3.333333
NJ	2.562500
NM	40.500000
NV	52.333333
NY	12.250000
OH	4.222222
OK	5.625000
OR	46.875000
PA	2.812500
RI	4.000000
SC	2.000000
SD	9.000000
TN	1.000000
TX	8.727273
UT	31.666667
VA	1.125000

VT	3.000000
WA	203.000000
WI	3.000000
WV	0.500000
WY	2.000000

	Overall Homeless - Multiple Races	total
state		
AK	174.500000	5228.000000
AL	13.500000	1652.000000
AR	18.600000	2087.200000
AZ	315.666667	18982.666667
CA	287.068182	16490.818182
CO	246.750000	14439.000000
CT	93.000000	6030.000000
DC	79.000000	19688.000000
DE	78.000000	4980.000000
FL	38.851852	4556.444444
GA	45.000000	5464.000000
HI	846.500000	12446.000000
IA	56.666667	3537.333333
ID	78.500000	4596.000000
IL	22.473684	2515.157895
IN	134.500000	12034.000000
KS	45.750000	2426.000000
KY	62.000000	6354.666667
LA	7.000000	1810.857143
MA	68.166667	6380.333333
MD	21.000000	2346.000000
ME	145.000000	17032.000000
MI	28.300000	1799.400000
MN	104.300000	3357.200000
MO	39.375000	3459.000000
MS	9.333333	1309.333333
MT	146.000000	8712.000000
NC	30.333333	3251.333333
ND	45.000000	3136.000000
NE	55.666667	3282.666667
NH	29.666667	3254.666667
NJ	16.375000	2566.000000
NM	81.000000	7684.000000
NV	108.666667	11554.666667
NY	239.333333	17200.000000
OH	59.444444	5060.444444
OK	40.625000	2324.000000
OR	187.000000	10071.000000
PA	42.250000	3139.000000

RI	185.000000	7240.000000
SC	38.000000	4053.000000
SD	52.000000	5128.000000
TN	27.100000	3686.000000
TX	61.272727	9955.272727
UT	67.000000	4916.000000
VA	27.187500	1690.250000
VT	34.500000	6590.000000
WA	445.000000	18690.666667
WI	64.000000	4861.000000
WV	13.500000	1416.000000
WY	28.000000	2128.000000

```
[106]: race_country_state['Overall Homeless - Total'] = race_country_state[
    [
        'Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x)',
        'Overall Homeless - Hispanic/Latin(o)(a)(x)',
        'Overall Homeless - White',
        'Overall Homeless - Black, African American, or African',
        'Overall Homeless - Asian or Asian American',
        'Overall Homeless - American Indian, Alaska Native, or Indigenous',
        'Overall Homeless - Native Hawaiian or Other Pacific Islander',
        'Overall Homeless - Multiple Races'
    ]
].sum(axis=1)

state_homeless_totals = race_country_state.groupby('state').
    ↪sum(numeric_only=True)['Overall Homeless - Total']
state_homeless_totals = state_homeless_totals.reset_index()

fig = px.choropleth(
    state_homeless_totals,
    locations='state',
    locationmode="USA-states",
    color='Overall Homeless - Total',
    color_continuous_scale="YlOrRd",

    scope="usa",
    labels={'Overall Homeless - Total': 'Total Homeless Population'},
    title="Total Homeless Population by State in the USA"
)

fig.update_layout(
    title_x=0.5,
    geo_scope="usa",
    width=1200,
    height=800
)
```



```
)  
  
fig.show()
```

### 1.3.3 Deliverer: FIRSTNAME LASTNAME

Question:

[ ]:

Answer:

[ ]:

[ ]:

## 1.4 Part 3: Follow-up Questions (4 points)

### 1.4.1 New Questions Based Off Initial Investigation

- Q1: WRITE\_QUESTION\_HERE
- Q2: WRITE\_QUESTION\_HERE

[ ]:

## 1.5 Summary (2 points)

GIVE A 2 PARAGRAPH SUMMARY.

PARAGRAPH 1 SHOULD DESCRIBE WHAT YOU LEARNED ABOUT YOUR DATA FROM INVESTIGATING THE INITIAL QUESTIONS. DID YOU FIND ANYTHING UNUSUAL IN YOUR DATA? DID ANYTHING SURPRISE YOU? WHICH OF THE INITIAL QUESTIONS WERE HELPFUL IN LEADING YOU TO MORE QUESTIONS?

PARAGRAPH 2 SHOULD SUMMARIZE WHAT YOU LEARNED FROM INVESTIGATING THE FOLLOW-UP QUESTIONS. WHY ARE THESE FOLLOW-UP QUESTIONS INTERESTING FOR INVESTIGATION? DESCRIBE THE TABLES/FIGURES YOU USED TO EXPLORE ANSWERS TO THESE FOLLOW-UP QUESTIONS? WHAT DID YOU LEARN FROM THE TABLES/FIGURES REGARDING THE FOLLOW-UP QUESTIONS YOU PROPOSED?

[ ]:

```
[107]: # get data from new york  
new_york_slice = PIT_sliced_state.loc[269:270]  
new_york_slice
```

```
[107]:      CoC Number      CoC Name \  
270      NY-525  New York Balance of State Continuum of Care  
  
      CoC Category  Overall Homeless  Overall Homeless - Under 18 \  
270  Largely Rural CoC              687.0              138.0
```

Overall Homeless - Age 18 to 24	Overall Homeless - Age 25 to 34	\
270	69.0	151.0
Overall Homeless - Age 35 to 44	Overall Homeless - Age 45 to 54	\
270	164.0	94.0
Overall Homeless - Age 55 to 64	...	\
270	53.0	...
Overall Homeless - Gender Questioning		\
270	0.0	
Overall Homeless - Non-Hispanic/Non-Latin(o)(a)(x)		\
270	604.0	
Overall Homeless - Hispanic/Latin(o)(a)(x)	Overall Homeless - White	\
270	83.0	553.0
Overall Homeless - Black, African American, or African		\
270	87.0	
Overall Homeless - Asian or Asian American		\
270	1.0	
Overall Homeless - American Indian, Alaska Native, or Indigenous		\
270	3.0	
Overall Homeless - Native Hawaiian or Other Pacific Islander		\
270	1.0	
Overall Homeless - Multiple Races	state	
270	42.0	NY

[1 rows x 25 columns]

```
[108]: nyc_slice = PIT_sliced[PIT_sliced["CoC Number"]=="NY-600"]
total_homeless = nyc_slice["Overall Homeless"].values[0]

# Calculate demographic percentages
race_cols = ["Overall Homeless - White",
             "Overall Homeless - Black, African American, or African",
             "Overall Homeless - Asian or Asian American",
             "Overall Homeless - American Indian, Alaska Native, or Indigenous",
             "Overall Homeless - Native Hawaiian or Other Pacific Islander",
             "Overall Homeless - Multiple Races"]

race_probs = []
```

```

for col in race_cols:
    race_probs.append(nyc_slice[col].values[0] / total_homeless)

# Create simulated population
np.random.seed(42)
n_simulated = 88025

# Simulate race based on observed proportions
simulated_race = np.random.choice(len(race_cols), size=n_simulated,
    p=race_probs)

# Create dataframe with simulated data
simulated_df = pd.DataFrame({
    "race": [race_cols[i].replace("Overall Homeless - ", "") for i in
    simulated_race]
})

# Add age based on NYC proportions
age_cols = ["Overall Homeless - Under 18",
    "Overall Homeless - Age 18 to 24",
    "Overall Homeless - Age 25 to 34",
    "Overall Homeless - Age 35 to 44",
    "Overall Homeless - Age 45 to 54",
    "Overall Homeless - Age 55 to 64",
    "Overall Homeless - Over 64"]

age_probs = []
for col in age_cols:
    age_probs.append(nyc_slice[col].values[0] / total_homeless)

simulated_df["age"] = np.random.choice(
    [col.replace("Overall Homeless - ", "").replace("Age ", "") for col in
    age_cols],
    size=n_simulated,
    p=age_probs
)

# Add gender based on NYC proportions
gender_cols = ["Overall Homeless - Female",
    "Overall Homeless - Male",
    "Overall Homeless - Transgender",
    "Overall Homeless - Gender that is not Singularly Female or Male",
    "Overall Homeless - Gender Questioning"]

gender_probs = []
for col in gender_cols:
    gender_probs.append(nyc_slice[col].values[0] / total_homeless)

```

```

simulated_df["gender"] = np.random.choice(
    [col.replace("Overall Homeless - ", "") for col in gender_cols],
    size=n_simulated,
    p=gender_probs
)

simulated_df["homeless"] = 1

```

```
[109]: race_probs
```

```

[109]: [0.2504742970746947,
        0.6735813689292814,
        0.008952002272081795,
        0.008838398182334565,
        0.0026242544731610337,
        0.05552967906844646]

```

```
[110]: simulated_df
```

```

[110]:
      race      age  gender  homeless
0  Black, African American, or African  35 to 44    Male           1
1           Multiple Races  Under 18  Female           1
2  Black, African American, or African  25 to 34    Male           1
3  Black, African American, or African  35 to 44  Female           1
4           White  Under 18    Male           1
...
88020  Black, African American, or African  25 to 34  Female           1
88021           White  Under 18    Male           1
88022      Asian or Asian American  18 to 24    Male           1
88023  Black, African American, or African  35 to 44    Male           1
88024  Black, African American, or African  Under 18  Female           1

[88025 rows x 4 columns]

```

```
[111]: n_non_homeless = 8_800_000 # Approximate NYC population
```

```

non_homeless_df = pd.DataFrame()

race_probs = {
    "White": 0.375,
    "Black, African American, or African": 0.231,
    "Asian or Asian American": 0.145,
    "Multiple Races": 0.089,
    "American Indian, Alaska Native, or Indigenous": 0.006,
}

```

```

    "Native Hawaiian or Other Pacific Islander": 0.001,
    "Hispanic or Latino": 0.29 - 0.137
}

non_homeless_df["race"] = np.random.choice(
    list(race_probs.keys()),
    size=n_non_homeless,
    p=list(race_probs.values())
)

# Age distribution based on NYC census
age_dist = {
    "Under 18": 0.21,
    "18 to 24": 0.10,
    "25 to 34": 0.17,
    "35 to 44": 0.14,
    "45 to 54": 0.13,
    "55 to 64": 0.13,
    "Over 64": 0.12
}

non_homeless_df["age"] = np.random.choice(
    list(age_dist.keys()),
    size=n_non_homeless,
    p=list(age_dist.values())
)

# Gender distribution based on NYC census
gender_dist = {
    "Female": 0.52 - (0.0012 + 0.0007 + 0.0001)/2,
    "Male": 0.48 - (0.0012 + 0.0007 + 0.0001)/2,
    "Other": 0.0012 + 0.0007 + 0.0001
}

non_homeless_df["gender"] = np.random.choice(
    list(gender_dist.keys()),
    size=n_non_homeless,
    p=list(gender_dist.values())
)

non_homeless_df["homeless"] = 0

# Combine homeless and non-homeless populations
simulated_df = pd.concat([simulated_df, non_homeless_df], ignore_index=True)
#simulated_df.drop("ethnicity", axis=1, inplace=True)

```

```

[112]: df = simulated_df.copy()

X = pd.get_dummies(df[["race", "age", "gender"]], drop_first=True)
y = df["homeless"]

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    ↪random_state=42, stratify=y)

scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

model = LogisticRegression(
    class_weight="balanced",
    max_iter=1000,
    random_state=42,
    C=0.1
)
model.fit(X_train, y_train)

y_pred_proba = model.predict_proba(X_test_scaled)[: , 1]

precisions, recalls, thresholds = precision_recall_curve(y_test, y_pred_proba)
f1_scores = 2 * (precisions * recalls) / (precisions + recalls)
optimal_threshold = thresholds[np.argmax(f1_scores[:-1])]

y_pred = (y_pred_proba >= optimal_threshold).astype(int)

print(f"Optimal threshold: {optimal_threshold:.3f}")
print("\nClassification Report with Optimal Threshold:")
print(classification_report(y_test, y_pred, zero_division=0))

roc_auc = roc_auc_score(y_test, y_pred_proba)
print(f"\nROC AUC Score: {roc_auc:.3f}")

feature_importance = pd.DataFrame({
    "feature": X.columns,
    "importance": abs(model.coef_[0])
})

```

```

})
feature_importance = feature_importance.sort_values("importance",
↪ascending=False)
print("\nTop 10 Most Important Features:")
print(feature_importance.head(10))

plt.figure(figsize=(8, 6))
sns.heatmap(confusion_matrix(y_test, y_pred),
            annot=True,
            fmt="d",
            cmap="Blues",
            xticklabels=["Not Homeless", "Homeless"],
            yticklabels=["Not Homeless", "Homeless"])
plt.title("Confusion Matrix")
plt.ylabel("True Label")
plt.xlabel("Predicted Label")
plt.show()

```

/Library/Frameworks/Python.framework/Versions/3.12/lib/python3.12/site-packages/sklearn/base.py:493: UserWarning:

X does not have valid feature names, but LogisticRegression was fitted with feature names

Optimal threshold: 1.000

Classification Report with Optimal Threshold:

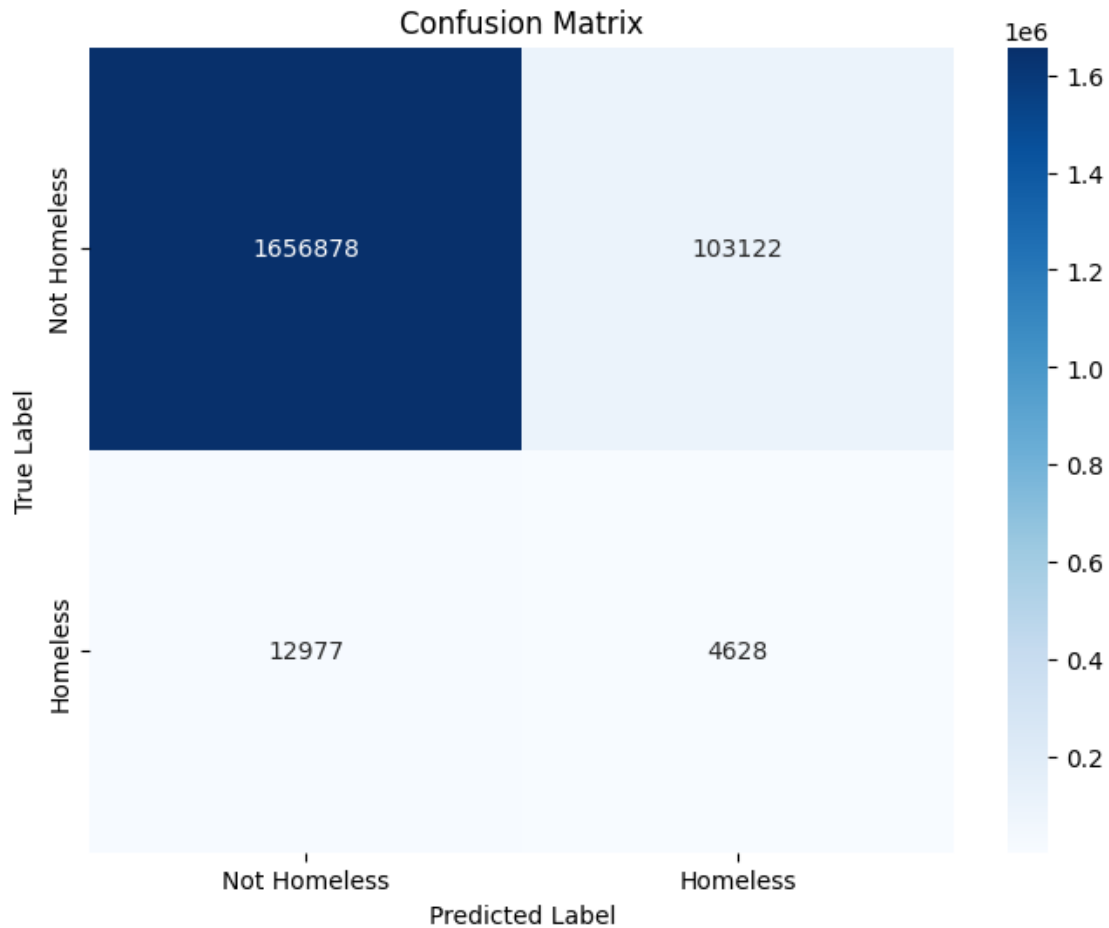
	precision	recall	f1-score	support
0	0.99	0.94	0.97	1760000
1	0.04	0.26	0.07	17605
accuracy			0.93	1777605
macro avg	0.52	0.60	0.52	1777605
weighted avg	0.98	0.93	0.96	1777605

ROC AUC Score: 0.803

Top 10 Most Important Features:

	feature	importance
2	race_Hispanic or Latino	12.341112
16	gender_Transgender	5.998717
15	gender_Other	4.039668
0	race_Asian or Asian American	3.134166
13	gender_Gender that is not Singularly Female or...	2.711221

10		age_Over 64	1.104926
12		gender_Gender Questioning	0.975387
3		race_Multiple Races	0.866681
5		race_White	0.777499
1		race_Black, African American, or African	0.697871



```
[113]: fpr, tpr, thresholds = roc_curve(y_test, y_pred)
```

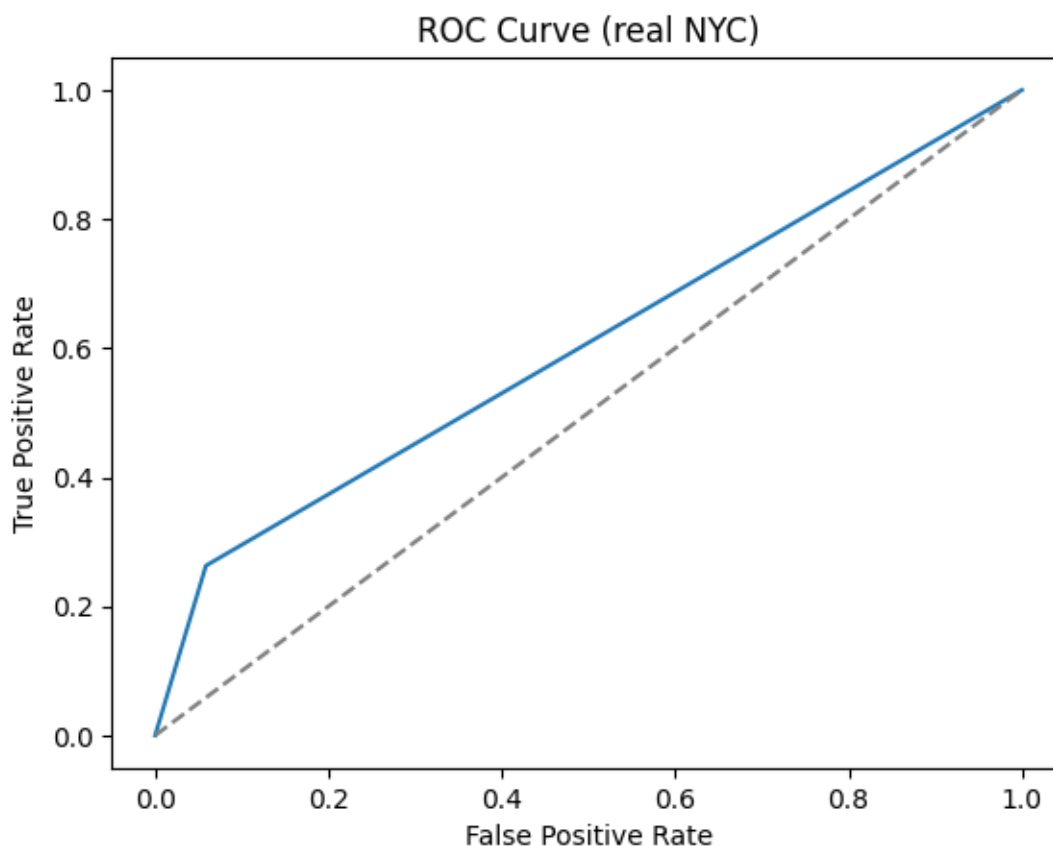
```
# use the model to make an roc curve
```

```
x = np.linspace(0, 1, 100)
plt.plot(fpr, tpr)
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('ROC Curve (real NYC)')
```

```
# random chance line
plt.plot(x,x, color="grey", linestyle="--")
```



```
plt.show()
```



```
[114]: # hypothetical scenario with same number of homeless and non homeless people

n_non_homeless = 88025

nyc_slice = PIT_sliced[PIT_sliced["CoC Number"]=="NY-600"]
total_homeless = nyc_slice["Overall Homeless"].values[0]

# Calculate demographic percentages
race_cols = ["Overall Homeless - White",
             "Overall Homeless - Black, African American, or African",
             "Overall Homeless - Asian or Asian American",
             "Overall Homeless - American Indian, Alaska Native, or Indigenous",
             "Overall Homeless - Native Hawaiian or Other Pacific Islander",
             "Overall Homeless - Multiple Races"]

race_probs = []
```

```

for col in race_cols:
    race_probs.append(nyc_slice[col].values[0] / total_homeless)

# Create simulated population
np.random.seed(42)
n_simulated = 88025

# Simulate race based on observed proportions
simulated_race = np.random.choice(len(race_cols), size=n_simulated,
    p=race_probs)

# Create dataframe with simulated data
simulated_df = pd.DataFrame({
    "race": [race_cols[i].replace("Overall Homeless - ", "") for i in
    simulated_race]
})

# Add age based on NYC proportions
age_cols = ["Overall Homeless - Under 18",
    "Overall Homeless - Age 18 to 24",
    "Overall Homeless - Age 25 to 34",
    "Overall Homeless - Age 35 to 44",
    "Overall Homeless - Age 45 to 54",
    "Overall Homeless - Age 55 to 64",
    "Overall Homeless - Over 64"]

age_probs = []
for col in age_cols:
    age_probs.append(nyc_slice[col].values[0] / total_homeless)

simulated_df["age"] = np.random.choice(
    [col.replace("Overall Homeless - ", "").replace("Age ", "") for col in
    age_cols],
    size=n_simulated,
    p=age_probs
)

# Add gender based on NYC proportions
gender_cols = ["Overall Homeless - Female",
    "Overall Homeless - Male",
    "Overall Homeless - Transgender",
    "Overall Homeless - Gender that is not Singularly Female or Male",
    "Overall Homeless - Gender Questioning"]

gender_probs = []
for col in gender_cols:
    gender_probs.append(nyc_slice[col].values[0] / total_homeless)

```

```

simulated_df["gender"] = np.random.choice(
    [col.replace("Overall Homeless - ", "") for col in gender_cols],
    size=n_simulated,
    p=gender_probs
)

simulated_df["homeless"] = 1

non_homeless_df = pd.DataFrame()

race_probs = {
    "White": 0.375,
    "Black, African American, or African": 0.231,
    "Asian or Asian American": 0.145,
    "Multiple Races": 0.089,
    "American Indian, Alaska Native, or Indigenous": 0.006,
    "Native Hawaiian or Other Pacific Islander": 0.001,
    "Hispanic or Latino": 0.29 - 0.137
}

non_homeless_df["race"] = np.random.choice(
    list(race_probs.keys()),
    size=n_non_homeless,
    p=list(race_probs.values())
)

# Age distribution based on NYC census
age_dist = {
    "Under 18": 0.21,
    "18 to 24": 0.10,
    "25 to 34": 0.17,
    "35 to 44": 0.14,
    "45 to 54": 0.13,
    "55 to 64": 0.13,
    "Over 64": 0.12
}

non_homeless_df["age"] = np.random.choice(
    list(age_dist.keys()),
    size=n_non_homeless,
    p=list(age_dist.values())
)

# Gender distribution based on NYC census

```

```

gender_dist = {
    "Female": 0.52 - (0.0012 + 0.0007 + 0.0001)/2,
    "Male": 0.48 - (0.0012 + 0.0007 + 0.0001)/2,
    "Other": 0.0012 + 0.0007 + 0.0001
}

non_homeless_df["gender"] = np.random.choice(
    list(gender_dist.keys()),
    size=n_non_homeless,
    p=list(gender_dist.values())
)

non_homeless_df["homeless"] = 0

# Combine homeless and non-homeless populations
simulated_df = pd.concat([simulated_df, non_homeless_df], ignore_index=True)
#simulated_df.drop("ethnicity", axis=1, inplace=True)

```

```

[115]: df = simulated_df.copy()

X = pd.get_dummies(df[["race", "age", "gender"]], drop_first=True)
y = df["homeless"]

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    ↪random_state=42, stratify=y)

scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

model = LogisticRegression(
    class_weight="balanced",
    max_iter=1000,
    random_state=42,
    C=0.1
)
model.fit(X_train, y_train)

y_pred_proba = model.predict_proba(X_test_scaled)[: , 1]

precisions, recalls, thresholds = precision_recall_curve(y_test, y_pred_proba)
f1_scores = 2 * (precisions * recalls) / (precisions + recalls)

```

```

optimal_threshold = thresholds[np.argmax(f1_scores[:-1])]

y_pred = (y_pred_proba >= optimal_threshold).astype(int)

print(f"Optimal threshold: {optimal_threshold:.3f}")
print("\nClassification Report with Optimal Threshold:")
print(classification_report(y_test, y_pred, zero_division=0))

roc_auc = roc_auc_score(y_test, y_pred_proba)
print(f"\nROC AUC Score: {roc_auc:.3f}")

feature_importance = pd.DataFrame({
    "feature": X.columns,
    "importance": abs(model.coef_[0])
})
feature_importance = feature_importance.sort_values("importance",
↪ascending=False)
print("\nTop 10 Most Important Features:")
print(feature_importance.head(10))

plt.figure(figsize=(8, 6))
sns.heatmap(confusion_matrix(y_test, y_pred),
            annot=True,
            fmt="d",
            cmap="Blues",
            xticklabels=["Not Homeless", "Homeless"],
            yticklabels=["Not Homeless", "Homeless"])
plt.title("Confusion Matrix")
plt.ylabel("True Label")
plt.xlabel("Predicted Label")
plt.show()

```

/Library/Frameworks/Python.framework/Versions/3.12/lib/python3.12/site-packages/sklearn/base.py:493: UserWarning:

X does not have valid feature names, but LogisticRegression was fitted with feature names

Optimal threshold: 0.473

Classification Report with Optimal Threshold:

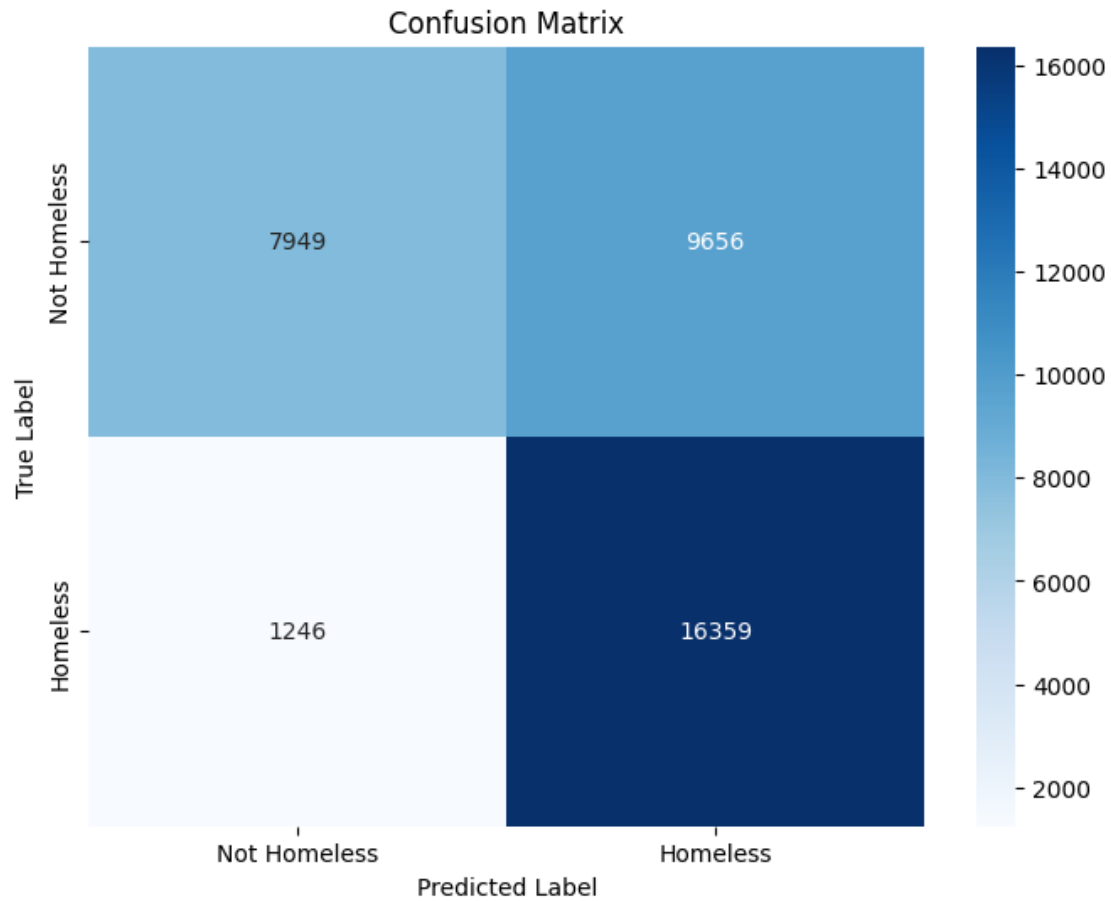
	precision	recall	f1-score	support
--	-----------	--------	----------	---------

0	0.86	0.45	0.59	17605
1	0.63	0.93	0.75	17605
accuracy			0.69	35210
macro avg	0.75	0.69	0.67	35210
weighted avg	0.75	0.69	0.67	35210

ROC AUC Score: 0.803

Top 10 Most Important Features:

	feature	importance
2	race_Hispanic or Latino	5.321921
0	race_Asian or Asian American	2.855276
16	gender_Transgender	2.612394
15	gender_Other	1.689474
10	age_Over 64	1.105533
1	race_Black, African American, or African	0.964928
13	gender_Gender that is not Singularly Female or...	0.816621
4	race_Native Hawaiian or Other Pacific Islander	0.703543
3	race_Multiple Races	0.624683
5	race_White	0.525241



```
[ ]:
```

```
[116]: fpr, tpr, thresholds = roc_curve(y_test, y_pred)
```

```
# use the model to make an roc curve

x = np.linspace(0, 1, 100)
plt.plot(fpr, tpr)
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('ROC Curve (fake NYC)')

# random chance line
plt.plot(x,x, color="grey", linestyle="--")

plt.show()
```

