

# Notebook4\_NYMA housing affordability and value

February 22, 2021

## 1 Notebook 4

Project:” Intra-Regional Migration and Transportation in New York Metro Area”

*Due to the large data our team is working with, there are a total of four notebooks submitted for this midterm (two from each team member)*

*I also outlined the notebook into the Table of Content*

## 2 Research Questions

- **Q1: On the county level, did housing costs (median housing value for owner occupied units and median rent for rental units) change between 2014 and 2018? If so, what county experience increase in housing price? And what county experience decrease in housing price?**
  - **Expected Exploration:**
    - \* We expect to explore and analyze the housing value data in New York Metro Region on the county level in 2014 and 2018. We expect to compare the housing value between the two years on the county level to find out changes in housing value. We hope to capture the geographic shift of housing value if there was any change.
    - \* We expect to create maps and charts to visually represent our findings of housing value comparison in 2014 and 2018.
- **Q2: On the county level, did housing affordability change between 2014 and 2018? If so, what county experience increase in housing affordability? And what county experience decrease in housing affordability?**
  - **Expected Exploration:**
    - \* We expect to capture changes of housing affordability in the New York Metro Area on the county level between 2014 and 2018 by exploring and analyzing the housing affordability data.
    - \* We expect to create maps and charts to visually represent our findings of housing affordability comparison in 2014 and 2018.
- **Purpose of this notebook:** I conducted data exploration and analysis of the New York Metro Area housing affordability and value data in 2014 and 2018. I used the ACS 5-year data from 2010 to 2014, and from 2014 to 2018. I created bar graphs with plotly.express to compare housing values and affordability in counties across NYMA between 2014 and 2018.

### 3 Data sources

1. 2010-2014 5-Years ACS: <https://www.census.gov/acs/www/data/data-tables-and-tools/data-profiles/2014/>
2. 2014-2018 5-Years ACS: <https://www.census.gov/acs/www/data/data-tables-and-tools/data-profiles/2018/>

### 4 Importing libraries

```
[1]: #import libraries
import pandas as pd
import geopandas as gpd
import matplotlib.pyplot as plt
import plotly.express as px
```

## 5 Data Exploration of Housing Affordability in 2014 and 2018

### 5.1 Data exploration of 2014 housing affordability data

```
[2]: ha2014 = pd.read_csv('2014ha.csv')
ha2014.head()
```

```
[2]:
```

	GEO_ID	NAME \
0	id	Geographic Area Name
1	0500000US09001	Fairfield County, Connecticut
2	0500000US09005	Litchfield County, Connecticut
3	0500000US09009	New Haven County, Connecticut
4	0500000US34003	Bergen County, New Jersey

	DP04_0023M \
0	Margin of Error!!YEAR STRUCTURE BUILT!!Total h...
1	1470
2	592
3	1379
4	1712

	DP04_0023PE \
0	Percent!!YEAR STRUCTURE BUILT!!Total housing u...
1	17.1
2	12.9
3	15.0
4	23.1

	DP04_0023PM \
0	Percent Margin of Error!!YEAR STRUCTURE BUILT!...
1	0.4

2	0.7
3	0.4
4	0.5

	DP04_0024E \
0	Estimate!!YEAR STRUCTURE BUILT!!Total housing ...
1	27396
2	5031
3	25327
4	42181

	DP04_0024M \
0	Margin of Error!!YEAR STRUCTURE BUILT!!Total h...
1	1002
2	463
3	909
4	1226

	DP04_0024PE \
0	Percent!!YEAR STRUCTURE BUILT!!Total housing u...
1	7.6
2	5.8
3	7.0
4	11.9

	DP04_0024PM \
0	Percent Margin of Error!!YEAR STRUCTURE BUILT!...
1	0.3
2	0.5
3	0.3
4	0.3

	DP04_0025E ... \
0	Estimate!!YEAR STRUCTURE BUILT!!Total housing ...
1	77682 ...
2	22131 ...
3	95576 ...
4	71674 ...

	DP04_0139PE \
0	Percent!!GROSS RENT AS A PERCENTAGE OF HOUSEHO...
1	9.1
2	10.5
3	9.6
4	8.3

	DP04_0139PM \
--	---------------

0	Percent Margin of Error!!GROSS RENT AS A PERCE...	
1		0.7
2		1.6
3		0.7
4		0.7

		DP04_0140E \
0	Estimate!!GROSS RENT AS A PERCENTAGE OF HOUSEH...	
1		45019
2		5820
3		54174
4		47940

		DP04_0140M \
0	Margin of Error!!GROSS RENT AS A PERCENTAGE OF...	
1		1326
2		489
3		1594
4		1471

		DP04_0140PE \
0	Percent!!GROSS RENT AS A PERCENTAGE OF HOUSEHO...	
1		45.6
2		38.1
3		48.1
4		43.3

		DP04_0140PM \
0	Percent Margin of Error!!GROSS RENT AS A PERCE...	
1		1.2
2		2.6
3		1.0
4		1.2

		DP04_0141E \
0	Estimate!!GROSS RENT AS A PERCENTAGE OF HOUSEH...	
1		6434
2		1813
3		7865
4		7476

		DP04_0141M \
0	Margin of Error!!GROSS RENT AS A PERCENTAGE OF...	
1		581
2		242
3		599
4		595

	DP04_0141PE \
0	Percent!!GROSS RENT AS A PERCENTAGE OF HOUSEHO...
1	(X)
2	(X)
3	(X)
4	(X)

	DP04_0141PM
0	Percent Margin of Error!!GROSS RENT AS A PERCE...
1	(X)
2	(X)
3	(X)
4	(X)

[5 rows x 566 columns]

```
[3]: #use .read_csv command to import 2014 ACS survey housing affordability data
hao2014 = pd.read_csv('hao2014.csv')
```

```
[4]: #look at the frist 5 rows of the dataset
hao2014.head()
```

```
[4]:
```

	GEO_ID	NAME	FIPS \
0	id	Geographic Area Name	FIPS
1	0500000US09001	Fairfield County, Connecticut	9001
2	0500000US09005	Litchfield County, Connecticut	9005
3	0500000US09009	New Haven County, Connecticut	9009
4	0500000US34003	Bergen County, New Jersey	34003

	DP04_0079E	DP04_0080E \
0	Estimate!!VALUE!!Owner-occupied units	ooh value<\$50,000
1	228331	4664
2	58127	1248
3	206556	5668
4	217432	4207

	DP04_0080PE	DP04_0081E \
0	%ooh value<\$50,000	ooh value \$50,000 to \$99,999
1	2	4247
2	2.1	1696
3	2.7	8206
4	1.9	1989

	DP04_0081PE	DP04_0082E \
0	%ooh value \$50,000 to \$99,999	ooh value \$100,000 to \$149,999
1	1.9	7431

2		2.9		5991
3		4		21721
4		0.9		3260

	DP04_0082PE	...	DP04_0136PE	\
0	%ooh value \$100,000 to \$149,999	...	% GRAPI 15.0 to 19.9 percent	
1		3.3	...	10.6
2		10.3	...	11
3		10.5	...	10.7
4		1.5	...	12.8

	DP04_0137E		DP04_0137PE	\
0	GRAPI 20.0 to 24.9 percent	% GRAPI 20.0 to 24.9 percent		
1		11158		11.3
2		2324		15.2
3		12362		11
4		13847		12.5

	DP04_0138E		DP04_0138PE	\
0	GRAPI 25.0 to 29.9 percent	%GRAPI 20.0 to 24.9 percent		
1		11511		11.7
2		1965		12.9
3		12281		10.9
4		12237		11

	DP04_0139E		DP04_0139PE	\
0	GRAPI 30.0 to 34.9 percent	% GRAPI 30.0 to 34.9 percent		
1		9000		9.1
2		1604		10.5
3		10813		9.6
4		9235		8.3

	DP04_0140E	DP04_0140PE	DP04_0141E
0	GRAPI > 35.0 percent	% GRAPI > 35.0 percent	GRAPI Not computed
1	45019	45.6	6434
2	5820	38.1	1813
3	54174	48.1	7865
4	47940	43.3	7476

[5 rows x 116 columns]

```
[5]: #drop the first row (row 0)
hao2014 = hao2014.drop([0])
```

```
[6]: #check datatypes and columns
hao2014.dtypes
```

```
[6]: GEO_ID      object
      NAME       object
      FIPS       object
      DP04_0079E object
      DP04_0080E object
      ...
      DP04_0139E object
      DP04_0139PE object
      DP04_0140E object
      DP04_0140PE object
      DP04_0141E object
      Length: 116, dtype: object
```

```
[7]: #convert FIPS from object to integer so that I can merge the 2014 housing_
      ↪ affordability dataframe with county boundary dataframe
hao2014['FIPS'] = hao2014['FIPS'].astype(str).astype(int)
hao2014.dtypes
```

```
[7]: GEO_ID      object
      NAME       object
      FIPS       int64
      DP04_0079E object
      DP04_0080E object
      ...
      DP04_0139E object
      DP04_0139PE object
      DP04_0140E object
      DP04_0140PE object
      DP04_0141E object
      Length: 116, dtype: object
```

## 5.2 Data exploration of 2018 housing affordability data

```
[8]: #import 2018 ACS survey housing affordability data
hao2018 = pd.read_csv('hao2018.csv')
```

```
[9]: #looking at the first 5 rows of the dataframe
hao2018.head()
```

```
[9]:
```

	GEO_ID	NAME	FIPS	\
0	id	Geographic Area Name	FIPS	
1	0500000US09001	Fairfield County, Connecticut	9001	
2	0500000US09005	Litchfield County, Connecticut	9005	
3	0500000US09009	New Haven County, Connecticut	9009	
4	0500000US34003	Bergen County, New Jersey	34003	

```

DP04_0080E
DP04_0081E \
```

0	Estimate!!VALUE!!Owner-occupied units	ooh value<\$50,000
1	229169	4667
2	57079	1214
3	204295	4853
4	217881	3109

	DP04_0081PE	DP04_0082E \
0	%ooh value<\$50,000	ooh value \$50,000 to \$99,999
1	2	4269
2	2.1	2103
3	2.4	10039
4	1.4	1961

	DP04_0082PE	DP04_0083E \
0	%ooh value \$50,000 to \$99,999	ooh value \$100,000 to \$149,999
1	1.9	7458
2	3.7	7068
3	4.9	22849
4	0.9	3082

	DP04_0083PE ...	DP04_0138PE \
0	%ooh value \$100,000 to \$149,999 ...	% GRAPI 15.0 to 19.9 percent
1	3.3 ...	10.7
2	12.4 ...	10.8
3	11.2 ...	11.1
4	1.4 ...	13.1

	DP04_0139E	DP04_0139PE \
0	GRAPI 20.0 to 24.9 percent	% GRAPI 20.0 to 24.9 percent
1	12366	11.7
2	2207	14.1
3	14309	12.3
4	14784	13

	DP04_0140E	DP04_0140PE \
0	GRAPI 25.0 to 29.9 percent	%GRAPI 20.0 to 24.9 percent
1	11064	10.5
2	2089	13.4
3	14755	12.6
4	13417	11.8

	DP04_0141E	DP04_0141PE \
0	GRAPI 30.0 to 34.9 percent	% GRAPI 30.0 to 34.9 percent
1	9455	9
2	1307	8.4
3	10224	8.8
4	9291	8.2



	DP04_0142E	DP04_0142PE	DP04_0143E
0	GRAPI > 35.0 percent	% GRAPI > 35.0 percent	GRAPI Not computed
1	48610	46.2	6071
2	6312	40.3	1263
3	52137	44.7	8834
4	45664	40.3	6981

[5 rows x 112 columns]

```
[10]: #drop the first row (row 0)
hao2018 = hao2018.drop([0])
```

```
[11]: #check datatypes
hao2018.dtypes
```

```
[11]: GEO_ID      object
NAME           object
FIPS           object
DP04_0080E     object
DP04_0081E     object
...
DP04_0141E     object
DP04_0141PE    object
DP04_0142E     object
DP04_0142PE    object
DP04_0143E     object
Length: 112, dtype: object
```

```
[12]: #convert FIPS from object to integer so that I can merge the 2018 housing_
      ↪ affordability dataframe with county boundary dataframe
hao2018['FIPS'] = hao2018['FIPS'].astype(str).astype(int)
hao2018.dtypes
```

```
[12]: GEO_ID      object
NAME           object
FIPS           int64
DP04_0080E     object
DP04_0081E     object
...
DP04_0141E     object
DP04_0141PE    object
DP04_0142E     object
DP04_0142PE    object
DP04_0143E     object
Length: 112, dtype: object
```

```
[13]: #change name of some columns to avoid repetition with hao2014 columns  
list(hao2018)
```

```
[13]: ['GEO_ID',  
      'NAME',  
      'FIPS',  
      'DP04_0080E',  
      'DP04_0081E',  
      'DP04_0081PE',  
      'DP04_0082E',  
      'DP04_0082PE',  
      'DP04_0083E',  
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      'DP04_0099E.1',  
      'DP04_0099PE.1',  
      'DP04_0100E.1',  
      'DP04_0100PE.1',  
      'DP04_0102E',  
      'DP04_0103E',
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```
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'DP04_0133E',  
'DP04_0133PE',  
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'DP04_0138E',  
'DP04_0138PE',  
'DP04_0139E',  
'DP04_0139PE',  
'DP04_0140E',  
'DP04_0140PE',  
'DP04_0141E',  
'DP04_0141PE',  
'DP04_0142E',  
'DP04_0142PE',  
'DP04_0143E']
```

```
[14]: hao2018.columns=['GEO_ID',  
    'NAME',  
    'FIPS',  
    'DP04_0080E',  
    'DP04_0081E',  
    'DP04_0081PE',  
    'DP04_0082E',  
    'DP04_0082PE',  
    'DP04_0083E',  
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    'DP04_0091E',  
    'DP04_0091PE',  
    'DP04_0092E',
```

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'DP04\_0116E\_y',  
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'DP04\_0120E\_y',  
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'DP04_0139PE',  
'DP04_0140E',  
'DP04_0140PE',  
'DP04_0141E',  
'DP04_0141PE',  
'DP04_0142E',  
'DP04_0142PE',  
'DP04_0143E']
```

### 5.3 Data analysis of the housing affordability data in 2014 and 2018

```
[15]: #import county boundary data set  
tracts=gpd.read_file('Countyborder/tl_2017_us_county.shp')  
tracts.head()
```

```
[15]: STATEFP COUNTYFP COUNTYNS GEOID      NAME      NAMELSAD LSAD CLASSFP \
0      31      039 00835841 31039      Cuming      Cuming County 06      H1
1      53      069 01513275 53069      Wahkiakum    Wahkiakum County 06      H1
2      35      011 00933054 35011      De Baca      De Baca County 06      H1
3      31      109 00835876 31109      Lancaster    Lancaster County 06      H1
4      31      129 00835886 31129      Nuckolls     Nuckolls County 06      H1
```

```
      MTFCC CSAFP CBSAFP METDIVFP FUNCSTAT      ALAND      AWATER      INTPTLAT \
0  G4020  None   None     None      A  1477641638  10701538  +41.9158651
1  G4020  None   None     None      A   680956787  61588406  +46.2946377
2  G4020  None   None     None      A  6016761648  29147345  +34.3592729
3  G4020   339  30700   None      A  2169252486  22867561  +40.7835474
4  G4020  None   None     None      A  1489645186   1718484  +40.1764918
```

```
      INTPTLON      geometry
0 -096.7885168 POLYGON ((-97.01952 42.00410, -97.01952 42.004...
1 -123.4244583 POLYGON ((-123.43639 46.23820, -123.44759 46.2...
2 -104.3686961 POLYGON ((-104.56739 33.99757, -104.56772 33.9...
3 -096.6886584 POLYGON ((-96.91060 40.95841, -96.91060 40.958...
4 -098.0468422 POLYGON ((-98.27367 40.08940, -98.27367 40.089...
```

```
[16]: #make a new column "FIPS"
      tracts['FIPS']=tracts['GEOID']
```

```
[17]: tracts.head()
```

```
[17]: STATEFP COUNTYFP COUNTYNS GEOID      NAME      NAMELSAD LSAD CLASSFP \
0      31      039 00835841 31039      Cuming      Cuming County 06      H1
1      53      069 01513275 53069      Wahkiakum    Wahkiakum County 06      H1
2      35      011 00933054 35011      De Baca      De Baca County 06      H1
3      31      109 00835876 31109      Lancaster    Lancaster County 06      H1
4      31      129 00835886 31129      Nuckolls     Nuckolls County 06      H1
```

```
      MTFCC CSAFP CBSAFP METDIVFP FUNCSTAT      ALAND      AWATER      INTPTLAT \
0  G4020  None   None     None      A  1477641638  10701538  +41.9158651
1  G4020  None   None     None      A   680956787  61588406  +46.2946377
2  G4020  None   None     None      A  6016761648  29147345  +34.3592729
3  G4020   339  30700   None      A  2169252486  22867561  +40.7835474
4  G4020  None   None     None      A  1489645186   1718484  +40.1764918
```

```
      INTPTLON      geometry      FIPS
0 -096.7885168 POLYGON ((-97.01952 42.00410, -97.01952 42.004... 31039
1 -123.4244583 POLYGON ((-123.43639 46.23820, -123.44759 46.2... 53069
2 -104.3686961 POLYGON ((-104.56739 33.99757, -104.56772 33.9... 35011
3 -096.6886584 POLYGON ((-96.91060 40.95841, -96.91060 40.958... 31109
4 -098.0468422 POLYGON ((-98.27367 40.08940, -98.27367 40.089... 31129
```

```
[18]: #check data types
      tracts.dtypes
```

```
[18]: STATEFP      object
      COUNTYFP    object
      COUNTYNS    object
      GEOID       object
      NAME        object
      NAMELSAD    object
      LSAD        object
      CLASSFP     object
      MTFCC       object
      CSAFP       object
      CBSAFP      object
      METDIVFP    object
      FUNCSTAT    object
      ALAND       int64
      AWATER      int64
      INTPTLAT    object
      INTPTLON    object
      geometry    geometry
      FIPS        object
      dtype: object
```

```
[19]: #convert FIPS data types in 'tracts' to integer to merge later with housing_
      ↪ affordability dataframe
      tracts['FIPS'] = tracts['FIPS'].astype(str).astype(int)
      tracts.dtypes
```

```
[19]: STATEFP      object
      COUNTYFP    object
      COUNTYNS    object
      GEOID       object
      NAME        object
      NAMELSAD    object
      LSAD        object
      CLASSFP     object
      MTFCC       object
      CSAFP       object
      CBSAFP      object
      METDIVFP    object
      FUNCSTAT    object
      ALAND       int64
      AWATER      int64
      INTPTLAT    object
      INTPTLON    object
      geometry    geometry
```



```
FIPS          int64
dtype: object
```

```
[20]: #merge the county boundary dataframe wiht 2018 housing affordability dataframes_
      ↪ on FIPS
tracts2018=tracts.merge(hao2018,on="FIPS")
pd.set_option('display.max_columns', None)
tracts2018.head()
```

```
[20]: STATEFP COUNTYFP COUNTYNS GEOID NAME_x NAMELSAD LSAD CLASSFP \
0      34      037 00882236 34037 Sussex Sussex County 06 H1
1      36      111 00974153 36111 Ulster Ulster County 06 H1
2      36      103 00974149 36103 Suffolk Suffolk County 06 H1
3      34      039 00882235 34039 Union Union County 06 H1
4      36      027 00974112 36027 Dutchess Dutchess County 06 H1
```

```
MTFCC CSAFP CBSAFP METDIVFP FUNCSTAT ALAND AWATER INTPTLAT \
0 G4020 408 35620 35084 A 1343552956 43234734 +41.1374609
1 G4020 408 28740 None A 2911757797 94596954 +41.9472124
2 G4020 408 35620 35004 A 2360846288 3785546967 +40.9435540
3 G4020 408 35620 35084 A 266170662 7046286 +40.6598707
4 G4020 408 35620 20524 A 2060678182 76956282 +41.7547699
```

```
INTPTLON geometry FIPS \
0 -074.6919141 POLYGON ((-74.93933 41.06918, -74.94004 41.069... 34037
1 -074.2654582 POLYGON ((-74.67402 41.97167, -74.68567 41.976... 36111
2 -072.6922184 POLYGON ((-72.86371 41.11338, -72.86351 41.113... 36103
3 -074.3086957 POLYGON ((-74.42857 40.69911, -74.42849 40.699... 34039
4 -073.7400411 POLYGON ((-73.94032 41.87275, -73.94075 41.873... 36027
```

```
GEO_ID NAME_y DP04_0080E DP04_0081E \
0 05000000US34037 Sussex County, New Jersey 44007 895
1 05000000US36111 Ulster County, New York 47815 2131
2 05000000US36103 Suffolk County, New York 390897 7197
3 05000000US34039 Union County, New Jersey 111289 2027
4 05000000US36027 Dutchess County, New York 73809 2559
```

```
DP04_0081PE DP04_0082E DP04_0082PE DP04_0083E DP04_0083PE DP04_0084E \
0 2 755 1.7 3144 7.1 6343
1 4.5 2623 5.5 5451 11.4 9511
2 1.8 4783 1.2 6753 1.7 12137
3 1.8 1195 1.1 2898 2.6 8602
4 3.5 2000 2.7 3652 4.9 8209
```

```
DP04_0084PE DP04_0085E DP04_0085PE DP04_0086E DP04_0086PE DP04_0087E \
0 14.4 15762 35.8 13236 30.1 3291
1 19.9 14861 31.1 9746 20.4 2848
```

2	3.1	65663	16.8	187835	48.1	89407
3	7.7	26144	23.5	39611	35.6	25141
4	11.1	26253	35.6	24000	32.5	6112

	DP04_0087PE	DP04_0088E	DP04_0088PE	DP04_0089E	DP04_0090PE	DP04_0091E	\
0	7.5	581	1.3	266000	44007	31635	
1	6	644	1.3	225500	47815	27977	
2	22.9	17122	4.4	386800	390897	264291	
3	22.6	5671	5.1	357300	111289	76034	
4	8.3	1024	1.4	278100	73809	49148	

	DP04_0091PE	DP04_0092E	DP04_0092PE	DP04_0093E	DP04_0093PE	DP04_0094E	\
0	71.9	12372	28.1	31635	31635	44	
1	58.5	19838	41.5	27977	27977	118	
2	67.6	126606	32.4	264291	264291	632	
3	68.3	35255	31.7	76034	76034	130	
4	66.6	24661	33.4	49148	49148	145	

	DP04_0094PE	DP04_0098E	DP04_0098PE	DP04_0099E	DP04_0099PE	DP04_0100E	\
0	0.1	846	2.7	4083	12.9	7416	
1	0.4	2022	7.2	5818	20.8	7230	
2	0.2	4121	1.6	13234	5	31662	
3	0.2	771	1	4767	6.3	10573	
4	0.3	1538	3.1	6036	12.3	11000	

	DP04_0100PE	DP04_0098E.1	DP04_0098PE.1	DP04_0099E.1	DP04_0099PE.1	\
0	23.4	7113	22.5	5312	16.8	
1	25.8	5766	20.6	3326	11.9	
2	12	51430	19.5	55037	20.8	
3	13.9	15129	19.9	16246	21.4	
4	22.4	11280	23	8250	16.8	

	DP04_0100E.1	DP04_0100PE.1	DP04_0102E	DP04_0103E	DP04_0103PE	DP04_0104E	\
0	6821	21.6	12372	191	1.5	249	
1	3697	13.2	19838	659	3.3	1374	
2	108175	40.9	126606	1474	1.2	2751	
3	28418	37.4	35255	469	1.3	573	
4	10899	22.2	24661	485	2	1323	

	DP04_0104PE	DP04_0105E	DP04_0105PE	DP04_0106E	DP04_0106PE	DP04_0107E	\
0	2	742	6	2278	18.4	3143	
1	6.9	4170	21	4904	24.7	3793	
2	2.2	5898	4.7	11971	9.5	18798	
3	1.6	997	2.8	2194	6.2	5840	
4	5.4	3572	14.5	5147	20.9	5503	

	DP04_0107PE	DP04_0108E_y	DP04_0108PE_y	DP04_0109E_y	DP04_0110E_y	\
--	-------------	--------------	---------------	--------------	--------------	---

0	25.4	5769	46.6	974	31569
1	19.1	4938	24.9	754	27852
2	14.8	85714	67.7	1194	263078
3	16.6	25182	71.4	1196	75687
4	22.3	8631	35	863	49072

	DP04_0110PE_y	DP04_0111E_y	DP04_0111PE_y	DP04_0112E_y	DP04_0112PE_y \
0	31569	10897	34.5	5128	16.2
1	27852	10046	36.1	4418	15.9
2	263078	75539	28.7	42813	16.3
3	75687	23777	31.4	12385	16.4
4	49072	16592	33.8	8185	16.7

	DP04_0113E_y	DP04_0113PE_y	DP04_0114E_y	DP04_0114PE_y	DP04_0115E_y \
0	4472	14.2	2769	8.8	8303
1	3373	12.1	2322	8.3	7693
2	33963	12.9	24782	9.4	85981
3	9218	12.2	6365	8.4	23942
4	6313	12.9	4387	8.9	13595

	DP04_0115PE_y	DP04_0116E_y	DP04_0117E_y	DP04_0120E_y	DP04_0120PE_y \
0	26.3	66	12205	7346	60.1
1	27.6	125	19623	12124	61.9
2	32.7	1213	125313	68664	54.8
3	31.6	347	34916	19381	55.6
4	27.7	76	24477	15905	65

	DP04_0121E	DP04_0121PE	DP04_0122E	DP04_0122PE	DP04_0123E	DP04_0123PE \
0	1310	10.7	784	6.4	669	5.5
1	1927	9.8	1240	6.3	845	4.3
2	12157	9.7	8235	6.6	6688	5.3
3	3424	9.8	2382	6.8	1966	5.6
4	1953	8	1400	5.7	1199	4.9

	DP04_0124E	DP04_0124PE	DP04_0125E	DP04_0126E	DP04_0127E	DP04_0127PE \
0	2096	17.2	167	8765	277	3.2
1	3487	17.8	215	20470	1321	6.5
2	29569	23.6	1293	91960	4916	5.3
3	7763	22.2	339	74937	4674	6.2
4	4020	16.4	184	31793	1772	5.6

	DP04_0128E	DP04_0128PE	DP04_0129E	DP04_0129PE	DP04_0130E	DP04_0130PE \
0	1691	19.3	3911	44.6	2029	23.1
1	7047	34.4	8306	40.6	2793	13.6
2	8369	9.1	21870	23.8	27381	29.8
3	14546	19.4	32414	43.3	15711	21
4	8178	25.7	13913	43.8	5205	16.4

	DP04_0131E	DP04_0131PE	DP04_0132E	DP04_0132PE	DP04_0133E	DP04_0133PE	\
0	649	7.4	152	1.7	56	0.6	
1	626	3.1	303	1.5	74	0.4	
2	18756	20.4	6931	7.5	3737	4.1	
3	5170	6.9	1493	2	929	1.2	
4	1907	6	538	1.7	280	0.9	

	DP04_0134E	DP04_0135E	Unnamed: 98	DP04_0137E	DP04_0137PE	DP04_0138E	\
0	1260	589	8624	787	9.1	1157	
1	1086	1254	20042	1905	9.5	1991	
2	1698	5124	90135	7435	8.2	8701	
3	1248	2761	73956	8842	12	8751	
4	1194	1745	31426	3392	10.8	4085	

	DP04_0138PE	DP04_0139E	DP04_0139PE	DP04_0140E	DP04_0140PE	DP04_0141E	\
0	13.4	1176	13.6	1018	11.8	815	
1	9.9	2457	12.3	1716	8.6	2111	
2	9.7	10047	11.1	9917	11	8180	
3	11.8	8847	12	8186	11.1	6691	
4	13	3583	11.4	3021	9.6	2754	

	DP04_0141PE	DP04_0142E	DP04_0142PE	DP04_0143E
0	9.5	3671	42.6	730
1	10.5	9862	49.2	1682
2	9.1	45855	50.9	6949
3	9	32639	44.1	3742
4	8.8	14591	46.4	2112

```
[21]: #merge tracts with hao2018 on FIPS plot it on the map
tracts2014=tracts.merge(hao2014,on="FIPS")
pd.set_option('display.max_columns', None)
tracts2014.head()
```

	STATEFP	COUNTYFP	COUNTYNS	GEOID	NAME_x	NAMELSAD	LSAD	CLASSFP	\
0	34	037	00882236	34037	Sussex	Sussex County	06	H1	
1	36	111	00974153	36111	Ulster	Ulster County	06	H1	
2	36	103	00974149	36103	Suffolk	Suffolk County	06	H1	
3	34	039	00882235	34039	Union	Union County	06	H1	
4	36	027	00974112	36027	Dutchess	Dutchess County	06	H1	

	MTFCC	CSAFP	CBSAFP	METDIVFP	FUNCSTAT	ALAND	AWATER	INTPTLAT	\
0	G4020	408	35620	35084	A	1343552956	43234734	+41.1374609	
1	G4020	408	28740	None	A	2911757797	94596954	+41.9472124	
2	G4020	408	35620	35004	A	2360846288	3785546967	+40.9435540	
3	G4020	408	35620	35084	A	266170662	7046286	+40.6598707	
4	G4020	408	35620	20524	A	2060678182	76956282	+41.7547699	

	INTPTLON		geometry	FIPS	\
0	-074.6919141	POLYGON	((-74.93933 41.06918, -74.94004 41.069...	34037	
1	-074.2654582	POLYGON	((-74.67402 41.97167, -74.68567 41.976...	36111	
2	-072.6922184	POLYGON	((-72.86371 41.11338, -72.86351 41.113...	36103	
3	-074.3086957	POLYGON	((-74.42857 40.69911, -74.42849 40.699...	34039	
4	-073.7400411	POLYGON	((-73.94032 41.87275, -73.94075 41.873...	36027	

	GEO_ID	NAME_y	DP04_0079E	DP04_0080E	\
0	0500000US34037	Sussex County, New Jersey	45642	1002	
1	0500000US36111	Ulster County, New York	48242	2494	
2	0500000US36103	Suffolk County, New York	393936	9261	
3	0500000US34039	Union County, New Jersey	111028	2298	
4	0500000US36027	Dutchess County, New York	74252	2980	

	DP04_0080PE	DP04_0081E	DP04_0081PE	DP04_0082E	DP04_0082PE	DP04_0083E	\
0	2.2	592	1.3	2519	5.5	6336	
1	5.2	2094	4.3	5002	10.4	9860	
2	2.4	5373	1.4	7398	1.9	13976	
3	2.1	1211	1.1	3365	3	8616	
4	4	1989	2.7	2964	4	7689	

	DP04_0083PE	DP04_0084E	DP04_0084PE	DP04_0085E	DP04_0085PE	DP04_0086E	\
0	13.9	16183	35.5	14291	31.3	4164	
1	20.4	15806	32.8	9247	19.2	3085	
2	3.5	71250	18.1	189896	48.2	80437	
3	7.8	25684	23.1	42557	38.3	22790	
4	10.4	26482	35.7	25104	33.8	5773	

	DP04_0086PE	DP04_0087E	DP04_0087PE	DP04_0088E	DP04_0089E	DP04_0090E	\
0	9.1	555	1.2	277600	45642	34326	
1	6.4	654	1.4	226600	48242	30144	
2	20.4	16345	4.1	376800	393936	275950	
3	20.5	4507	4.1	353300	111028	78045	
4	7.8	1271	1.7	282100	74252	52242	

	DP04_0090PE	DP04_0091E	DP04_0091PE	DP04_0092E	DP04_0092PE	DP04_0093E	\
0	75.2	11316	24.8	34326	34326	12	
1	62.5	18098	37.5	30144	30144	11	
2	70	117986	30	275950	275950	202	
3	70.3	32983	29.7	78045	78045	75	
4	70.4	22010	29.6	52242	52242	15	

	DP04_0093PE	DP04_0094E	DP04_0094PE	DP04_0095E	DP04_0095PE	DP04_0096E	\
0	0	100	0.3	124	0.4	711	
1	0	183	0.6	417	1.4	1783	
2	0.1	699	0.3	1591	0.6	4482	

3	0.1	258	0.3	307	0.4	1008
4	0	138	0.3	456	0.9	1753

	DP04_0096PE	DP04_0097E	DP04_0097PE	DP04_0098E	DP04_0098PE	DP04_0099E \
0	2.1	4030	11.7	7377	21.5	21972
1	5.9	6190	20.5	7910	26.2	13650
2	1.6	15772	5.7	34159	12.4	219045
3	1.3	4741	6.1	11418	14.6	60238
4	3.4	6053	11.6	11183	21.4	32644

	DP04_0099M	DP04_0099PE	DP04_0100E	DP04_0101E	DP04_0102E	DP04_0102PE \
0	661	64	2304	11316	39	0.3
1	667	45.3	1908	18098	26	0.1
2	2186	79.4	2776	117986	280	0.2
3	1193	77.2	2688	32983	28	0.1
4	780	62.5	2276	22010	13	0.1

	DP04_0103E	DP04_0103PE	DP04_0104E	DP04_0104PE	DP04_0105E	DP04_0105PE \
0	80	0.7	138	1.2	223	2
1	120	0.7	574	3.2	1059	5.9
2	531	0.5	1317	1.1	2219	1.9
3	227	0.7	354	1.1	449	1.4
4	123	0.6	305	1.4	876	4

	DP04_0106E	DP04_0106PE	DP04_0107E	DP04_0108E	DP04_0108PE	DP04_0109E \
0	10836	95.8	952	34237	34237	8937
1	16319	90.2	739	30025	30025	8986
2	113639	96.3	1,000+	274895	274895	64353
3	31925	96.8	1,000+	77588	77588	20583
4	20693	94	836	52110	52110	14690

	DP04_0109PE	DP04_0110E	DP04_0110PE	DP04_0111E	DP04_0111PE	DP04_0112E \
0	26.1	5934	17.3	4719	13.8	3766
1	29.9	4300	14.3	4281	14.3	2790
2	23.4	40407	14.7	38373	14	28291
3	26.5	10809	13.9	9854	12.7	7390
4	28.2	7916	15.2	6983	13.4	5548

	DP04_0112PE	DP04_0113E	DP04_0113PE	DP04_0114E	DP04_0115E	DP04_0116E \
0	11	10881	31.8	89	11261	2362
1	9.3	9668	32.2	119	17862	4134
2	10.3	103471	37.6	1055	116764	20637
3	9.5	28952	37.3	457	32638	6437
4	10.6	16973	32.6	132	21840	5339

	DP04_0116PE	DP04_0117E	DP04_0117PE	DP04_0118E	DP04_0118PE	DP04_0119E \
0	21	2249	20	1782	15.8	1252

1	23.1	3287	18.4	2912	16.3	1736
2	17.7	22002	18.8	17533	15	12323
3	19.7	5862	18	4526	13.9	3017
4	24.4	4416	20.2	3187	14.6	2406

	DP04_0119PE	DP04_0120E	DP04_0120PE	DP04_0121E	DP04_0121PE	DP04_0122E \
0	11.1	728	6.5	703	6.2	2185
1	9.7	1211	6.8	883	4.9	3699
2	10.6	8127	7	6309	5.4	29833
3	9.2	2622	8	1937	5.9	8237
4	11	1562	7.2	1103	5.1	3827

	DP04_0122PE	DP04_0123E	DP04_0124E	DP04_0125E	DP04_0125PE	DP04_0126E \
0	19.4	55	7996	90	1.1	161
1	20.7	236	19682	142	0.7	417
2	25.5	1222	97718	355	0.4	1511
3	25.2	345	72659	649	0.9	1511
4	17.5	170	31058	165	0.5	603

	DP04_0126PE	DP04_0127E	DP04_0127PE	DP04_0128E	DP04_0128PE	DP04_0129E \
0	2	174	2.2	321	4	1429
1	2.1	889	4.5	2582	13.1	5510
2	1.5	3021	3.1	4501	4.6	7362
3	2.1	2544	3.5	4232	5.8	14241
4	1.9	1348	4.3	2970	9.6	6136

	DP04_0129PM	DP04_0130E	DP04_0130PE	DP04_0131E	DP04_0131PE	DP04_0132E \
0	3	3543	44.3	2278	28.5	1219
1	2.3	7917	40.2	2225	11.3	1014
2	0.6	31087	31.8	49881	51	1519
3	1	31509	43.4	17973	24.7	1181
4	1.7	13423	43.2	6413	20.6	1124

	DP04_0133E	DP04_0134E	DP04_0135E	DP04_0135PE	DP04_0136E	DP04_0136PE \
0	610	7937	688	8.7	809	10.2
1	1464	19227	1457	7.6	2074	10.8
2	5126	95932	7779	8.1	9384	9.8
3	2001	71536	6895	9.6	8093	11.3
4	1588	30634	2884	9.4	3716	12.1

	DP04_0137E	DP04_0137PE	DP04_0138E	DP04_0138PE	DP04_0139E	DP04_0139PE \
0	915	11.5	651	8.2	600	7.6
1	2403	12.5	1833	9.5	2005	10.4
2	11194	11.7	10693	11.1	8803	9.2
3	8439	11.8	8532	11.9	6006	8.4
4	3617	11.8	3876	12.7	2251	7.3

	DP04_0140E	DP04_0140PE	DP04_0141E
0	4274	53.8	669
1	9455	49.2	1919
2	48079	50.1	6912
3	33571	46.9	3124
4	14290	46.6	2012

[22]: *#convert the data type of columns in 2014 housing affordability dataframe from object to integers for mapping*

```
tracts2014["DP04_0109E"] = tracts2014["DP04_0109E"].astype(str).astype(int)
tracts2014["DP04_0109PE"] = tracts2014["DP04_0109PE"].astype(float).astype(int)
tracts2014["DP04_0110E"] = tracts2014["DP04_0110E"].astype(str).astype(int)
tracts2014["DP04_0110PE"] = tracts2014["DP04_0110PE"].astype(float).astype(int)
tracts2014["DP04_0111E"] = tracts2014["DP04_0111E"].astype(float).astype(int)
tracts2014["DP04_0111PE"] = tracts2014["DP04_0111PE"].astype(float).astype(int)
tracts2014["DP04_0112E"] = tracts2014["DP04_0112E"].astype(float).astype(int)
tracts2014["DP04_0112PE"] = tracts2014["DP04_0112PE"].astype(float).astype(int)
tracts2014["DP04_0113E"] = tracts2014["DP04_0113E"].astype(float).astype(int)
tracts2014["DP04_0113PE"] = tracts2014["DP04_0113PE"].astype(float).astype(int)
tracts2014["DP04_0114E"] = tracts2014["DP04_0114E"].astype(float).astype(int)
```

[23]: *#convert the data type of columns in 2018 housing affordability dataframe from object to integers for mapping*

```
tracts2018["DP04_0111E_y"] = tracts2018["DP04_0111E_y"].astype(str).astype(int)
tracts2018["DP04_0111PE_y"] = tracts2018["DP04_0111PE_y"].astype(float).
    ↳astype(int)
tracts2018["DP04_0112E_y"] = tracts2018["DP04_0112E_y"].astype(str).astype(int)
tracts2018["DP04_0112PE_y"] = tracts2018["DP04_0112PE_y"].astype(float).
    ↳astype(int)
tracts2018["DP04_0113E_y"] = tracts2018["DP04_0113E_y"].astype(float).astype(int)
tracts2018["DP04_0113PE_y"] = tracts2018["DP04_0113PE_y"].astype(float).
    ↳astype(int)
tracts2018["DP04_0114E_y"] = tracts2018["DP04_0114E_y"].astype(float).astype(int)
tracts2018["DP04_0114PE_y"] = tracts2018["DP04_0114PE_y"].astype(float).
    ↳astype(int)
tracts2018["DP04_0115E_y"] = tracts2018["DP04_0115E_y"].astype(float).astype(int)
tracts2018["DP04_0115PE_y"] = tracts2018["DP04_0115PE_y"].astype(float).
    ↳astype(int)
tracts2018["DP04_0116E_y"] = tracts2018["DP04_0116E_y"].astype(float).astype(int)
```



### 5.3.1 Comparing housing affordability for owner occupied housing in 2014 and 2018 with Plotly.Express

```
[24]: #calculating the SMOCAPI<30 for 2014 by adding three SMOCAPI categories
tracts2014['SMOCAPI1430']=
↳tracts2014['DP04_0109PE']+tracts2014['DP04_0110PE']+tracts2014['DP04_0111PE']
tracts2014.head()
```

```
[24]: STATEFP COUNTYFP COUNTYNS GEOID NAME_x NAMELSAD LSAD CLASSFP \
0 34 037 00882236 34037 Sussex Sussex County 06 H1
1 36 111 00974153 36111 Ulster Ulster County 06 H1
2 36 103 00974149 36103 Suffolk Suffolk County 06 H1
3 34 039 00882235 34039 Union Union County 06 H1
4 36 027 00974112 36027 Dutchess Dutchess County 06 H1
```

```
MTFCC CSAFP CBSAFP METDIVFP FUNCSTAT ALAND AWATER INTPTLAT \
0 G4020 408 35620 35084 A 1343552956 43234734 +41.1374609
1 G4020 408 28740 None A 2911757797 94596954 +41.9472124
2 G4020 408 35620 35004 A 2360846288 3785546967 +40.9435540
3 G4020 408 35620 35084 A 266170662 7046286 +40.6598707
4 G4020 408 35620 20524 A 2060678182 76956282 +41.7547699
```

```
INTPTLON geometry FIPS \
0 -074.6919141 POLYGON ((-74.93933 41.06918, -74.94004 41.069... 34037
1 -074.2654582 POLYGON ((-74.67402 41.97167, -74.68567 41.976... 36111
2 -072.6922184 POLYGON ((-72.86371 41.11338, -72.86351 41.113... 36103
3 -074.3086957 POLYGON ((-74.42857 40.69911, -74.42849 40.699... 34039
4 -073.7400411 POLYGON ((-73.94032 41.87275, -73.94075 41.873... 36027
```

```
GEO_ID NAME_y DP04_0079E DP04_0080E \
0 0500000US34037 Sussex County, New Jersey 45642 1002
1 0500000US36111 Ulster County, New York 48242 2494
2 0500000US36103 Suffolk County, New York 393936 9261
3 0500000US34039 Union County, New Jersey 111028 2298
4 0500000US36027 Dutchess County, New York 74252 2980
```

```
DP04_0080PE DP04_0081E DP04_0081PE DP04_0082E DP04_0082PE DP04_0083E \
0 2.2 592 1.3 2519 5.5 6336
1 5.2 2094 4.3 5002 10.4 9860
2 2.4 5373 1.4 7398 1.9 13976
3 2.1 1211 1.1 3365 3 8616
4 4 1989 2.7 2964 4 7689
```

```
DP04_0083PE DP04_0084E DP04_0084PE DP04_0085E DP04_0085PE DP04_0086E \
0 13.9 16183 35.5 14291 31.3 4164
1 20.4 15806 32.8 9247 19.2 3085
2 3.5 71250 18.1 189896 48.2 80437
```

3	7.8	25684	23.1	42557	38.3	22790
4	10.4	26482	35.7	25104	33.8	5773

	DP04_0086PE	DP04_0087E	DP04_0087PE	DP04_0088E	DP04_0089E	DP04_0090E \
0	9.1	555	1.2	277600	45642	34326
1	6.4	654	1.4	226600	48242	30144
2	20.4	16345	4.1	376800	393936	275950
3	20.5	4507	4.1	353300	111028	78045
4	7.8	1271	1.7	282100	74252	52242

	DP04_0090PE	DP04_0091E	DP04_0091PE	DP04_0092E	DP04_0092PE	DP04_0093E \
0	75.2	11316	24.8	34326	34326	12
1	62.5	18098	37.5	30144	30144	11
2	70	117986	30	275950	275950	202
3	70.3	32983	29.7	78045	78045	75
4	70.4	22010	29.6	52242	52242	15

	DP04_0093PE	DP04_0094E	DP04_0094PE	DP04_0095E	DP04_0095PE	DP04_0096E \
0	0	100	0.3	124	0.4	711
1	0	183	0.6	417	1.4	1783
2	0.1	699	0.3	1591	0.6	4482
3	0.1	258	0.3	307	0.4	1008
4	0	138	0.3	456	0.9	1753

	DP04_0096PE	DP04_0097E	DP04_0097PE	DP04_0098E	DP04_0098PE	DP04_0099E \
0	2.1	4030	11.7	7377	21.5	21972
1	5.9	6190	20.5	7910	26.2	13650
2	1.6	15772	5.7	34159	12.4	219045
3	1.3	4741	6.1	11418	14.6	60238
4	3.4	6053	11.6	11183	21.4	32644

	DP04_0099M	DP04_0099PE	DP04_0100E	DP04_0101E	DP04_0102E	DP04_0102PE \
0	661	64	2304	11316	39	0.3
1	667	45.3	1908	18098	26	0.1
2	2186	79.4	2776	117986	280	0.2
3	1193	77.2	2688	32983	28	0.1
4	780	62.5	2276	22010	13	0.1

	DP04_0103E	DP04_0103PE	DP04_0104E	DP04_0104PE	DP04_0105E	DP04_0105PE \
0	80	0.7	138	1.2	223	2
1	120	0.7	574	3.2	1059	5.9
2	531	0.5	1317	1.1	2219	1.9
3	227	0.7	354	1.1	449	1.4
4	123	0.6	305	1.4	876	4

	DP04_0106E	DP04_0106PE	DP04_0107E	DP04_0108E	DP04_0108PE	DP04_0109E \
0	10836	95.8	952	34237	34237	8937

1	16319	90.2	739	30025	30025	8986
2	113639	96.3	1,000+	274895	274895	64353
3	31925	96.8	1,000+	77588	77588	20583
4	20693	94	836	52110	52110	14690

	DP04_0109PE	DP04_0110E	DP04_0110PE	DP04_0111E	DP04_0111PE	DP04_0112E \
0	26	5934	17	4719	13	3766
1	29	4300	14	4281	14	2790
2	23	40407	14	38373	14	28291
3	26	10809	13	9854	12	7390
4	28	7916	15	6983	13	5548

	DP04_0112PE	DP04_0113E	DP04_0113PE	DP04_0114E	DP04_0115E	DP04_0116E \
0	11	10881	31	89	11261	2362
1	9	9668	32	119	17862	4134
2	10	103471	37	1055	116764	20637
3	9	28952	37	457	32638	6437
4	10	16973	32	132	21840	5339

	DP04_0116PE	DP04_0117E	DP04_0117PE	DP04_0118E	DP04_0118PE	DP04_0119E \
0	21	2249	20	1782	15.8	1252
1	23.1	3287	18.4	2912	16.3	1736
2	17.7	22002	18.8	17533	15	12323
3	19.7	5862	18	4526	13.9	3017
4	24.4	4416	20.2	3187	14.6	2406

	DP04_0119PE	DP04_0120E	DP04_0120PE	DP04_0121E	DP04_0121PE	DP04_0122E \
0	11.1	728	6.5	703	6.2	2185
1	9.7	1211	6.8	883	4.9	3699
2	10.6	8127	7	6309	5.4	29833
3	9.2	2622	8	1937	5.9	8237
4	11	1562	7.2	1103	5.1	3827

	DP04_0122PE	DP04_0123E	DP04_0124E	DP04_0125E	DP04_0125PE	DP04_0126E \
0	19.4	55	7996	90	1.1	161
1	20.7	236	19682	142	0.7	417
2	25.5	1222	97718	355	0.4	1511
3	25.2	345	72659	649	0.9	1511
4	17.5	170	31058	165	0.5	603

	DP04_0126PE	DP04_0127E	DP04_0127PE	DP04_0128E	DP04_0128PE	DP04_0129E \
0	2	174	2.2	321	4	1429
1	2.1	889	4.5	2582	13.1	5510
2	1.5	3021	3.1	4501	4.6	7362
3	2.1	2544	3.5	4232	5.8	14241
4	1.9	1348	4.3	2970	9.6	6136

	DP04_0129PM	DP04_0130E	DP04_0130PE	DP04_0131E	DP04_0131PE	DP04_0132E	\
0	3	3543	44.3	2278	28.5	1219	
1	2.3	7917	40.2	2225	11.3	1014	
2	0.6	31087	31.8	49881	51	1519	
3	1	31509	43.4	17973	24.7	1181	
4	1.7	13423	43.2	6413	20.6	1124	

	DP04_0133E	DP04_0134E	DP04_0135E	DP04_0135PE	DP04_0136E	DP04_0136PE	\
0	610	7937	688	8.7	809	10.2	
1	1464	19227	1457	7.6	2074	10.8	
2	5126	95932	7779	8.1	9384	9.8	
3	2001	71536	6895	9.6	8093	11.3	
4	1588	30634	2884	9.4	3716	12.1	

	DP04_0137E	DP04_0137PE	DP04_0138E	DP04_0138PE	DP04_0139E	DP04_0139PE	\
0	915	11.5	651	8.2	600	7.6	
1	2403	12.5	1833	9.5	2005	10.4	
2	11194	11.7	10693	11.1	8803	9.2	
3	8439	11.8	8532	11.9	6006	8.4	
4	3617	11.8	3876	12.7	2251	7.3	

	DP04_0140E	DP04_0140PE	DP04_0141E	SMOCAPI1430
0	4274	53.8	669	56
1	9455	49.2	1919	57
2	48079	50.1	6912	51
3	33571	46.9	3124	51
4	14290	46.6	2012	56

```
[25]: #calculating the SMOCAPI<30 for 2018 by adding three SMOCAPI categories
tracts2018['SMOCAPI1830']=_
↳tracts2018['DP04_0111PE_y']+tracts2018['DP04_0112PE_y']+tracts2018['DP04_0113PE_y']
tracts2018.head()
```

```
[25]: STATEFP COUNTYFP COUNTYNS GEOID NAME_x NAMELSAD LSAD CLASSFP \
0 34 037 00882236 34037 Sussex Sussex County 06 H1
1 36 111 00974153 36111 Ulster Ulster County 06 H1
2 36 103 00974149 36103 Suffolk Suffolk County 06 H1
3 34 039 00882235 34039 Union Union County 06 H1
4 36 027 00974112 36027 Dutchess Dutchess County 06 H1
```

	MTFCC	CSAFP	CBSAFP	METDIVFP	FUNCSTAT	ALAND	AWATER	INTPTLAT	\
0	G4020	408	35620	35084	A	1343552956	43234734	+41.1374609	
1	G4020	408	28740	None	A	2911757797	94596954	+41.9472124	
2	G4020	408	35620	35004	A	2360846288	3785546967	+40.9435540	
3	G4020	408	35620	35084	A	266170662	7046286	+40.6598707	
4	G4020	408	35620	20524	A	2060678182	76956282	+41.7547699	

	INTPTLON		geometry	FIPS	\
0	-074.6919141	POLYGON	((-74.93933 41.06918, -74.94004 41.069...	34037	
1	-074.2654582	POLYGON	((-74.67402 41.97167, -74.68567 41.976...	36111	
2	-072.6922184	POLYGON	((-72.86371 41.11338, -72.86351 41.113...	36103	
3	-074.3086957	POLYGON	((-74.42857 40.69911, -74.42849 40.699...	34039	
4	-073.7400411	POLYGON	((-73.94032 41.87275, -73.94075 41.873...	36027	

	GEO_ID	NAME_y	DP04_0080E	DP04_0081E	\
0	0500000US34037	Sussex County, New Jersey	44007	895	
1	0500000US36111	Ulster County, New York	47815	2131	
2	0500000US36103	Suffolk County, New York	390897	7197	
3	0500000US34039	Union County, New Jersey	111289	2027	
4	0500000US36027	Dutchess County, New York	73809	2559	

	DP04_0081PE	DP04_0082E	DP04_0082PE	DP04_0083E	DP04_0083PE	DP04_0084E	\
0	2	755	1.7	3144	7.1	6343	
1	4.5	2623	5.5	5451	11.4	9511	
2	1.8	4783	1.2	6753	1.7	12137	
3	1.8	1195	1.1	2898	2.6	8602	
4	3.5	2000	2.7	3652	4.9	8209	

	DP04_0084PE	DP04_0085E	DP04_0085PE	DP04_0086E	DP04_0086PE	DP04_0087E	\
0	14.4	15762	35.8	13236	30.1	3291	
1	19.9	14861	31.1	9746	20.4	2848	
2	3.1	65663	16.8	187835	48.1	89407	
3	7.7	26144	23.5	39611	35.6	25141	
4	11.1	26253	35.6	24000	32.5	6112	

	DP04_0087PE	DP04_0088E	DP04_0088PE	DP04_0089E	DP04_0090PE	DP04_0091E	\
0	7.5	581	1.3	266000	44007	31635	
1	6	644	1.3	225500	47815	27977	
2	22.9	17122	4.4	386800	390897	264291	
3	22.6	5671	5.1	357300	111289	76034	
4	8.3	1024	1.4	278100	73809	49148	

	DP04_0091PE	DP04_0092E	DP04_0092PE	DP04_0093E	DP04_0093PE	DP04_0094E	\
0	71.9	12372	28.1	31635	31635	44	
1	58.5	19838	41.5	27977	27977	118	
2	67.6	126606	32.4	264291	264291	632	
3	68.3	35255	31.7	76034	76034	130	
4	66.6	24661	33.4	49148	49148	145	

	DP04_0094PE	DP04_0098E	DP04_0098PE	DP04_0099E	DP04_0099PE	DP04_0100E	\
0	0.1	846	2.7	4083	12.9	7416	
1	0.4	2022	7.2	5818	20.8	7230	
2	0.2	4121	1.6	13234	5	31662	
3	0.2	771	1	4767	6.3	10573	

4	0.3	1538	3.1	6036	12.3	11000
---	-----	------	-----	------	------	-------

	DP04_0100PE	DP04_0098E.1	DP04_0098PE.1	DP04_0099E.1	DP04_0099PE.1	\
0	23.4	7113	22.5	5312	16.8	
1	25.8	5766	20.6	3326	11.9	
2	12	51430	19.5	55037	20.8	
3	13.9	15129	19.9	16246	21.4	
4	22.4	11280	23	8250	16.8	

	DP04_0100E.1	DP04_0100PE.1	DP04_0102E	DP04_0103E	DP04_0103PE	DP04_0104E	\
0	6821	21.6	12372	191	1.5	249	
1	3697	13.2	19838	659	3.3	1374	
2	108175	40.9	126606	1474	1.2	2751	
3	28418	37.4	35255	469	1.3	573	
4	10899	22.2	24661	485	2	1323	

	DP04_0104PE	DP04_0105E	DP04_0105PE	DP04_0106E	DP04_0106PE	DP04_0107E	\
0	2	742	6	2278	18.4	3143	
1	6.9	4170	21	4904	24.7	3793	
2	2.2	5898	4.7	11971	9.5	18798	
3	1.6	997	2.8	2194	6.2	5840	
4	5.4	3572	14.5	5147	20.9	5503	

	DP04_0107PE	DP04_0108E_y	DP04_0108PE_y	DP04_0109E_y	DP04_0110E_y	\
0	25.4	5769	46.6	974	31569	
1	19.1	4938	24.9	754	27852	
2	14.8	85714	67.7	1194	263078	
3	16.6	25182	71.4	1196	75687	
4	22.3	8631	35	863	49072	

	DP04_0110PE_y	DP04_0111E_y	DP04_0111PE_y	DP04_0112E_y	DP04_0112PE_y	\
0	31569	10897	34	5128	16	
1	27852	10046	36	4418	15	
2	263078	75539	28	42813	16	
3	75687	23777	31	12385	16	
4	49072	16592	33	8185	16	

	DP04_0113E_y	DP04_0113PE_y	DP04_0114E_y	DP04_0114PE_y	DP04_0115E_y	\
0	4472	14	2769	8	8303	
1	3373	12	2322	8	7693	
2	33963	12	24782	9	85981	
3	9218	12	6365	8	23942	
4	6313	12	4387	8	13595	

	DP04_0115PE_y	DP04_0116E_y	DP04_0117E_y	DP04_0120E_y	DP04_0120PE_y	\
0	26	66	12205	7346	60.1	
1	27	125	19623	12124	61.9	

2	32	1213	125313	68664	54.8
3	31	347	34916	19381	55.6
4	27	76	24477	15905	65

	DP04_0121E	DP04_0121PE	DP04_0122E	DP04_0122PE	DP04_0123E	DP04_0123PE \
0	1310	10.7	784	6.4	669	5.5
1	1927	9.8	1240	6.3	845	4.3
2	12157	9.7	8235	6.6	6688	5.3
3	3424	9.8	2382	6.8	1966	5.6
4	1953	8	1400	5.7	1199	4.9

	DP04_0124E	DP04_0124PE	DP04_0125E	DP04_0126E	DP04_0127E	DP04_0127PE \
0	2096	17.2	167	8765	277	3.2
1	3487	17.8	215	20470	1321	6.5
2	29569	23.6	1293	91960	4916	5.3
3	7763	22.2	339	74937	4674	6.2
4	4020	16.4	184	31793	1772	5.6

	DP04_0128E	DP04_0128PE	DP04_0129E	DP04_0129PE	DP04_0130E	DP04_0130PE \
0	1691	19.3	3911	44.6	2029	23.1
1	7047	34.4	8306	40.6	2793	13.6
2	8369	9.1	21870	23.8	27381	29.8
3	14546	19.4	32414	43.3	15711	21
4	8178	25.7	13913	43.8	5205	16.4

	DP04_0131E	DP04_0131PE	DP04_0132E	DP04_0132PE	DP04_0133E	DP04_0133PE \
0	649	7.4	152	1.7	56	0.6
1	626	3.1	303	1.5	74	0.4
2	18756	20.4	6931	7.5	3737	4.1
3	5170	6.9	1493	2	929	1.2
4	1907	6	538	1.7	280	0.9

	DP04_0134E	DP04_0135E	Unnamed: 98	DP04_0137E	DP04_0137PE	DP04_0138E \
0	1260	589	8624	787	9.1	1157
1	1086	1254	20042	1905	9.5	1991
2	1698	5124	90135	7435	8.2	8701
3	1248	2761	73956	8842	12	8751
4	1194	1745	31426	3392	10.8	4085

	DP04_0138PE	DP04_0139E	DP04_0139PE	DP04_0140E	DP04_0140PE	DP04_0141E \
0	13.4	1176	13.6	1018	11.8	815
1	9.9	2457	12.3	1716	8.6	2111
2	9.7	10047	11.1	9917	11	8180
3	11.8	8847	12	8186	11.1	6691
4	13	3583	11.4	3021	9.6	2754

DP04\_0141PE DP04\_0142E DP04\_0142PE DP04\_0143E SMOCAPI1830

0	9.5	3671	42.6	730	64
1	10.5	9862	49.2	1682	63
2	9.1	45855	50.9	6949	56
3	9	32639	44.1	3742	59
4	8.8	14591	46.4	2112	61

```
[26]: #exclude 5 NYC counties from analysis and plotting since they are most likely
      ↪to be outliers
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36081']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36047']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36061']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36005']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36085']
#calculate housing affordability change between 2014 and 2018
tracts2018['ac']=tracts2018['SMOCAPI1830']-tracts2014['SMOCAPI1430']
#get top 5 (oadt5) and bottom 5 (oacb5) counties in NYMA experienced most and
      ↪least amount of changes in affordability
oact5=tracts2018 .sort_values(by='ac',ascending = False).head(5)
oacb5=tracts2018 .sort_values(by='ac',ascending = False).tail(5)
#create new dataframe by combining the top 5 and bottom 5 counties for plotting
ooaffordability=oact5.append(oacb5)
```

```
[27]: #look at the change of housing affordability between 2014 and 2018 on county
      ↪level
figac=px.bar(
    ooaffordability,
    x='ac',
    y='NAME_x',
    orientation='h', #change orientation of bar chart
    color='ac',
    labels={'ac': '% change', 'NAME_x': 'County Names'}, #change labels
    color_continuous_scale='Bluered' #change color of the bars
)
#update bar charts
figac.update_layout(title={
    'text': "% of affordable housing change in top five and bottom five
      ↪counties", #add title
    'y':1, #change position of the title
    'x':0.5}
)
figac
```

```
[28]: #again, exlude 5 counties in NYC from analysis
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36081']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36047']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36061']
tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36005']
```



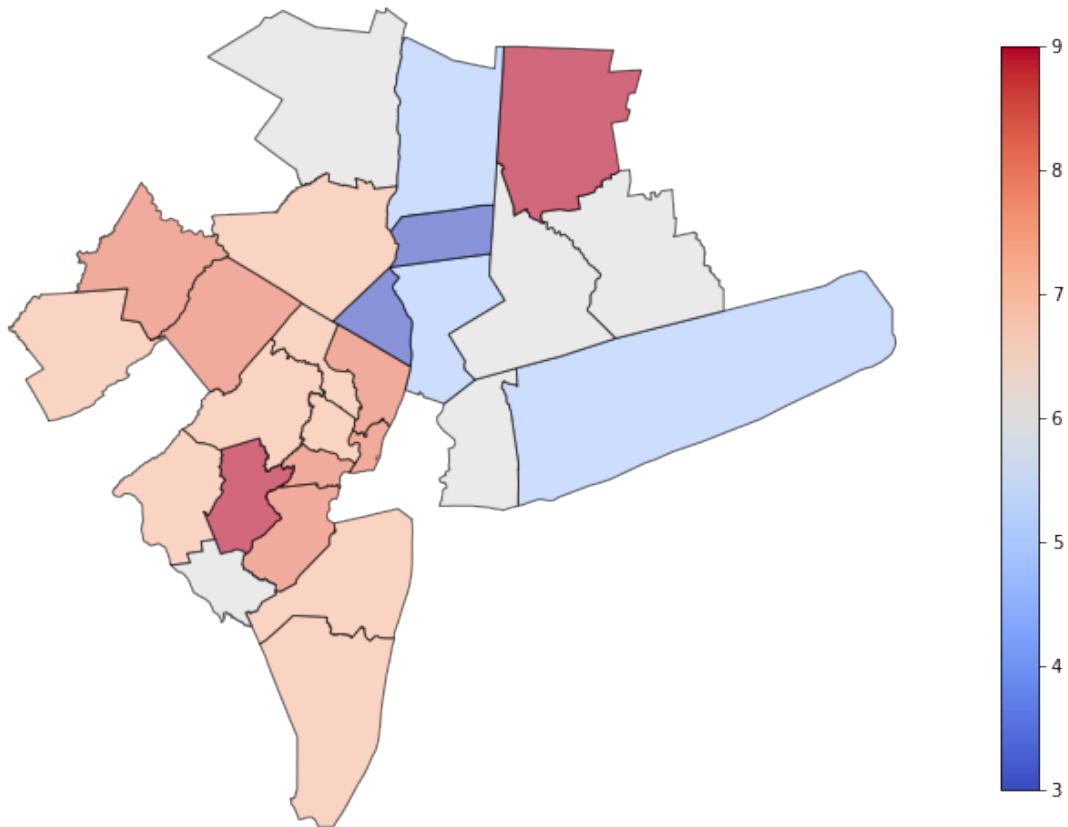
```

tracts2018=tracts2018.loc[tracts2018['GEOID'] != '36085']
#plot the chane of housing affordability on the map
tracts2018.plot(figsize=(12,10),
    cmap='coolwarm',
    column='ac',
    legend = True,
    edgecolor='Black',
    alpha=0.6,
    legend_kwds={'shrink': 0.75}) #change the size of legend
plt.axis('off');
plt.title('% of affordable housing change between 2014 and 2018',fontsize=16)

```

[28]: Text(0.5, 1.0, '% of affordable housing change between 2014 and 2018')

% of affordable housing change between 2014 and 2018



### 5.3.2 Comparing housing affordability for rental housing in 2014 and 2018

```
[45]: #convert data type from object to integer for later calculation
tracts2014["DP04_0135PE"] =tracts2014["DP04_0135PE"].astype(float).astype(int)
tracts2014["DP04_0136PE"] =tracts2014["DP04_0136PE"].astype(float).astype(int)
tracts2014["DP04_0137PE"] =tracts2014["DP04_0137PE"].astype(float).astype(int)
tracts2014["DP04_0138PE"] =tracts2014["DP04_0138PE"].astype(float).astype(int)
tracts2018["DP04_0137PE"] =tracts2018["DP04_0137PE"].astype(float).astype(int)
tracts2018["DP04_0138PE"] =tracts2018["DP04_0138PE"].astype(float).astype(int)
tracts2018["DP04_0139PE"] =tracts2018["DP04_0139PE"].astype(float).astype(int)
tracts2018["DP04_0140PE"] =tracts2018["DP04_0140PE"].astype(float).astype(int)
```

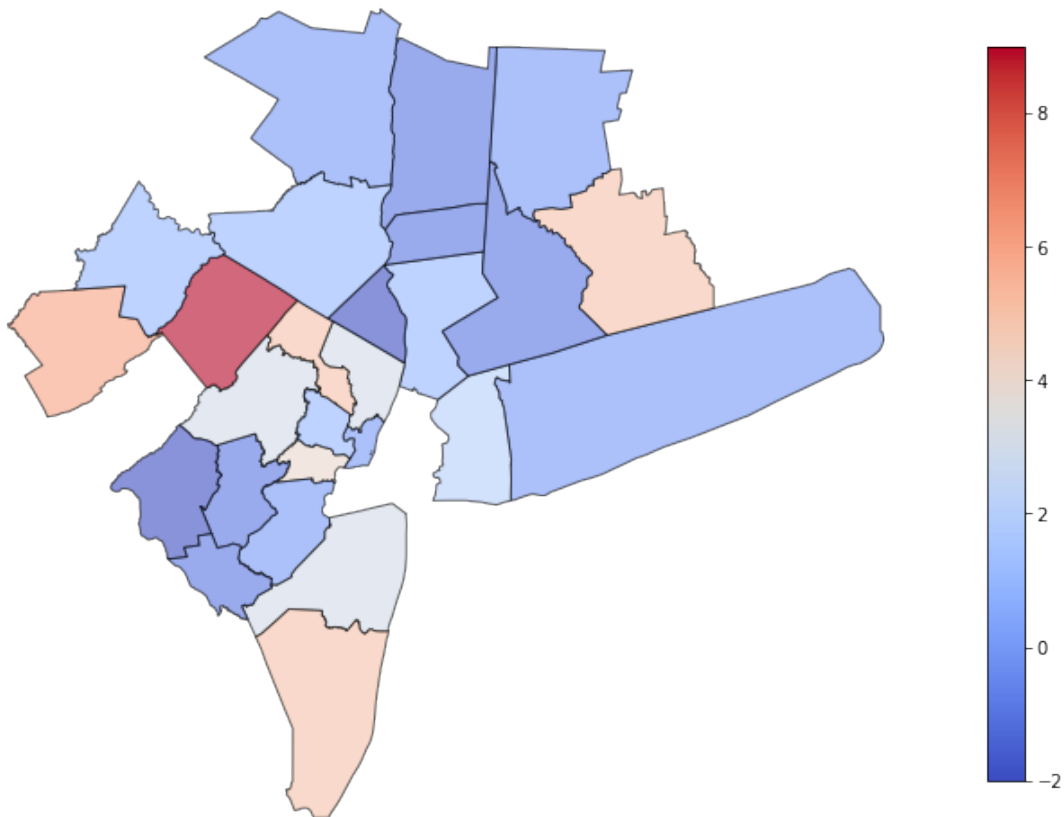
```
[49]: #calculate housing affordability change for rental housing units between 2014
      ↪and 2018
tracts2014['GRAPI1430']=
      ↪tracts2014['DP04_0135PE']+tracts2014['DP04_0136PE']+tracts2014['DP04_0137PE']+tracts2014['D
tracts2018['GRAPI1830']=
      ↪tracts2018['DP04_0137PE']+tracts2018['DP04_0138PE']+tracts2018['DP04_0139PE']+tracts2018['D
tracts2018['rac']=tracts2018['GRAPI1830']-tracts2014['GRAPI1430']
#get top 5 (ract5) and bottom 5 (racb5) counties in NYMA experienced most and
      ↪least amount of changes in affordability
ract5=tracts2018 .sort_values(by='rac',ascending = False).head(5)
racb5=tracts2018 .sort_values(by='rac',ascending = False).tail(5)
#create new dataframe by combining the top 5 and bottom 5 counties for plotting
raffordability=ract5.append(racb5)
```

```
[53]: #look at the change of rental housing affordability between 2014 and 2018 on
      ↪county level
figrac=px.bar(
    raffordability,
    x='rac',
    y='NAME_x',
    orientation='h', #change orientation of bar chart
    color='rac',
    labels={'rac': '% change', 'NAME_x': 'County Names'}, #change labels
    color_continuous_scale='Bluered' #change color of the bars
)
#update bar charts
figrac.update_layout(title={
    'text': "% of affordable rental housing change in top five and bottom
      ↪five counties", #add title
    'y':1, #change position of the title
    'x':0.5}
)
figrac
```

```
[54]: #plot rental housing affordaibility change on the map
tracts2018.plot(figsize=(12,10),
               cmap='coolwarm',
               column='rac',
               legend = True,
               edgecolor='Black',
               alpha=0.6,
               legend_kwds={'shrink': 0.75}) #change the size of legend
plt.axis('off');
plt.title('% of affordable rental housing change between 2014 and_
↪2018',fontsize=16)
```

```
[54]: Text(0.5, 1.0, '% of affordable rental housing change between 2014 and 2018')
```

% of affordable rental housing change between 2014 and 2018



## 5.4 Conclusion

From the data exploration and analysis of housing affordability of owner occupied housing units with mortgage in NYMA, there is an overall increase in housing affordability in NYMA counties,

ranging from two to eight percent. The highest increase is in counties further away from New York County (NYC), but also in Hudson County, which is located across from New York County.

Next step of the affordability analysis will be focused on the relationship between migration, housing affordability, and transit density on the county level.

## 6 Data exploration and analysis of median housing value of owner-occupied housing and median rent in 2014 and 2018

### 6.1 Data exploration of median housing value

```
[29]: #change the data type of the columns under analysis from object to integer
tracts2014["DP04_0088E"] = tracts2014["DP04_0088E"].astype(str).astype(int)

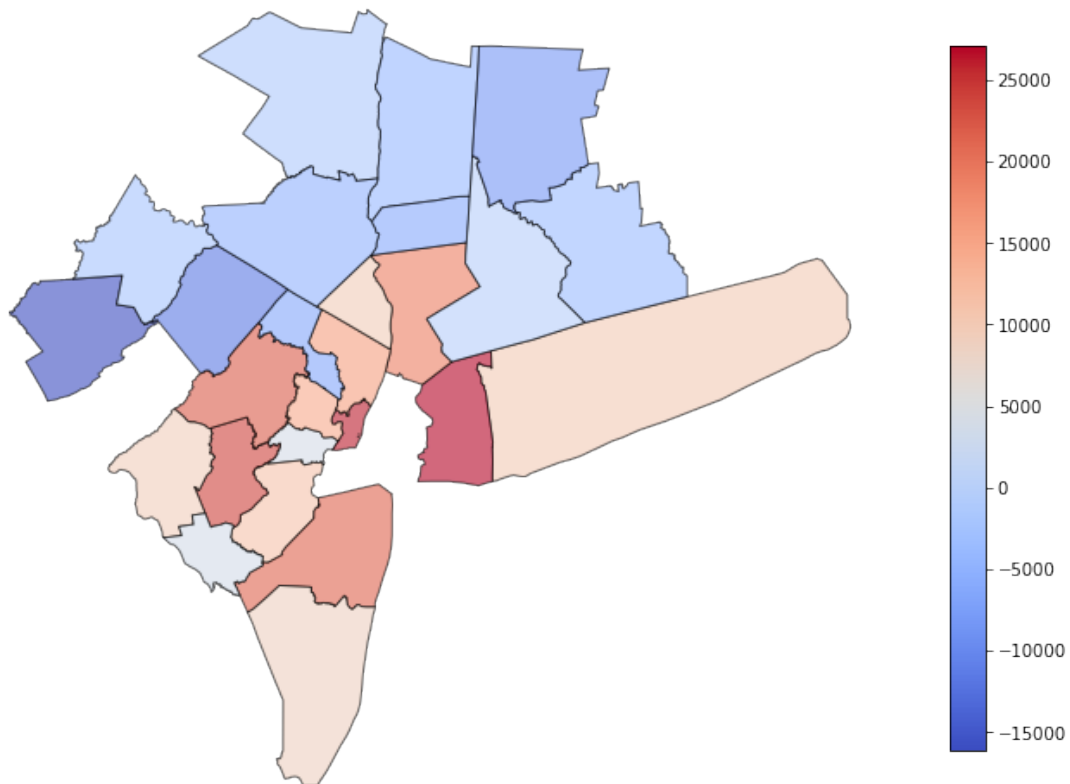
[30]: tracts2018["DP04_0089E"] = tracts2018["DP04_0089E"].astype(str).astype(int)

[31]: #calculate the change of owner occupied median housing cost between 2014 and
      →2018, and adding it onto tracts2018 dataframe
tracts2018['omc'] = tracts2018['DP04_0089E'] - tracts2014['DP04_0088E']

[32]: #plot the change of housing median value on the map
tracts2018.plot(figsize=(12,10),
               cmap='coolwarm',
               column='omc',
               legend = True,
               edgecolor='Black',
               alpha=0.6,
               legend_kwds={'shrink': 0.75})
plt.axis('off');
plt.title('Owner-occupied Median Value-2014 v 2018',fontsize=16)

[32]: Text(0.5, 1.0, 'Owner-occupied Median Value-2014 v 2018')
```

Owner-occupied Median Value-2014 v 2018



```
[33]: #use sort function to find the top 5 and bottom 5 counties experienced most
      ↪ increases and most descreases in median value
top5m = tracts2018 .sort_values(by='omc',ascending = False).head(5)
bot5m = tracts2018 .sort_values(by='omc',ascending = False).tail(5)
```

```
[34]: #create a new dataframe with top 5 and bottom 5
oomedian = top5m.append(bot5m)
```

```
[35]: #plot the counties with top 5 and bottom 5 changes of median housing value of
      ↪ owner-occupied units
figm=px.bar(
    oomedian,
    x='omc',
    y='NAME_x',
    orientation='h',
    color='omc',
    color_continuous_scale='Bluered',
    labels={'omc':'median value change','NAME_x':'County Names'}, #change
    ↪ labels
```

```
)
figm.update_layout(title={
    'text': "owner occupied median value: 2014 v 2018", #add title
    'y':1,#change position of the title
    'x':0.5}
)
figm
```

## 6.2 Data exploration of median rent

```
[36]: #change the data type of the columns under analysis from object to integer
tracts2014["DP04_0132E"] =tracts2014["DP04_0132E"].astype(str).astype(int)
```

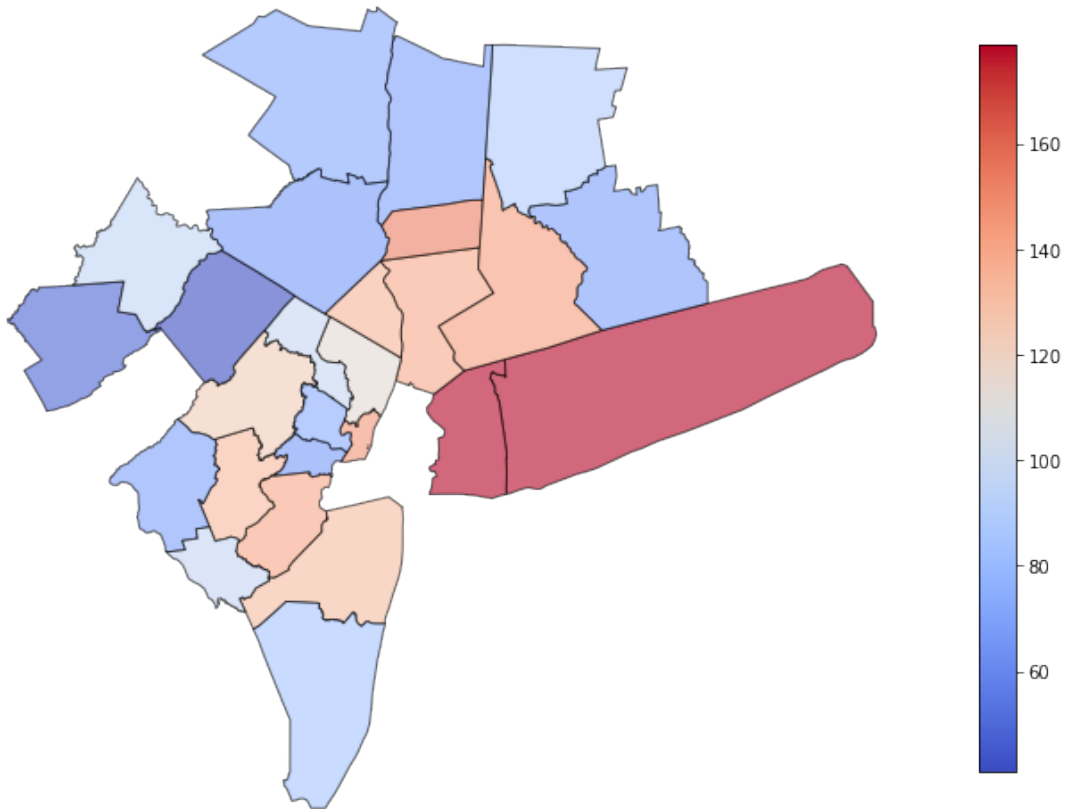
```
[37]: tracts2018["DP04_0134E"] =tracts2018["DP04_0134E"].astype(str).astype(int)
```

```
[38]: #calculate the change of median rent
tracts2018['rmc']=tracts2018['DP04_0134E']-tracts2014['DP04_0132E']
```

```
[39]: #plot the chane of housing median value on the map
tracts2018.plot(figsize=(12,10),
    cmap='coolwarm',
    column='rmc',
    legend = True,
    edgecolor='Black',
    alpha=0.6,
    legend_kwds={'shrink': 0.75})
plt.axis('off');
plt.title('Median Rent-2014 v 2018',fontsize=16)
```

```
[39]: Text(0.5, 1.0, 'Median Rent-2014 v 2018')
```

Median Rent-2014 v 2018



```
[40]: #use sort function to find the top 5 and bottom 5 counties experienced most
      ↪ increases and most decreases in median value
```

```
top5rm = tracts2018 .sort_values(by='rmc',ascending = False).head(5)
bot5rm = tracts2018 .sort_values(by='rmc',ascending = False).tail(5)
```

```
[41]: #create a new dataframe with top 5 and bottom 5
```

```
rmedian = top5rm.append(bot5rm)
```

```
[42]: figm=px.bar(
      rmedian,
      x='rmc',
      y='NAME_x',
      orientation='h',
      color='rmc',
      color_continuous_scale='Bluered',
      labels={'rmc':'median rent change','NAME_x':'County Names'}, #change
      ↪ labels
    )
```

```
figm.update_layout(title={
    'text': "owner occupied median value: 2014 v 2018", #add title
    'y':1,#change position of the title
    'x':0.5}
)
figm
```

## 7 Conclusion on housing affordability and housing value in NYMA between 2014 and 2018

In this data exploration and analysis notebook, I conducted basic data exploration of housing affordability datasets, housing value datasets and county boundary datasets. I analyzed and mapped housing affordability and housing value in 2014 and 2018 across 31 counties in New York Metro Area. In the future weeks during this quarter, I will conduct more analysis on the impact of migration on housing price and affordability in NYMA.