

hw3 with census reporter data

January 31, 2022

This is me trying again to do a map with Census data! After talking with Ryan, I've learnt that there might have been an issue with the Puerto Rico map I had downloaded. Because of that, I will now use data from Census Reporter, as it comes in geoJSON and does not need me to add another map.

```
[2]: import pandas as pd
```

```
[3]: # to read and visualize spatial data
import geopandas as gpd

# to provide basemaps
import contextily as ctx

# to give more power to your figures (plots)
import matplotlib.pyplot as plt
```

```
/opt/conda/lib/python3.9/site-packages/geopandas/_compat.py:106: UserWarning:
The Shapely GEOS version (3.9.1-CAPI-1.14.2) is incompatible with the GEOS
version PyGEOS was compiled with (3.10.1-CAPI-1.16.0). Conversions between both
will be slow.
```

```
warnings.warn(
```

```
[4]: # load a data file
# note the relative filepath! where is this file located?

gdf = gpd.read_file('rent_sanjuan.geojson')
```

```
[5]: gdf.shape
```

```
[5]: (2595, 57)
```

```
[6]: # first five rows, but not all columns?
gdf.head()
```

```
[6]:
```

	geoid	name	B25063001	\
0	04000US72	Puerto Rico	380029.0	
1	15000US720019563001	Block Group 1, Adjuntas, PR	167.0	
2	15000US720019563002	Block Group 2, Adjuntas, PR	223.0	

```

3 15000US720019564001 Block Group 1, Adjuntas, PR      239.0
4 15000US720019564002 Block Group 2, Adjuntas, PR      154.0

```

```

      B25063001, Error  B25063002  B25063002, Error  B25063003  B25063003, Error  \
0          4032.0    232258.0          3022.0    14006.0          757.0
1           54.0      89.0           62.0      0.0           14.0
2           89.0      94.0           67.0      0.0           14.0
3           94.0     149.0           75.0      0.0           14.0
4           77.0      16.0           27.0      0.0           14.0

```

```

      B25063004  B25063004, Error  ...  B25063023, Error  B25063024  \
0    11090.0          786.0  ...          259.0    435.0
1         0.0          14.0  ...          14.0     0.0
2         0.0          14.0  ...          14.0     0.0
3         0.0          14.0  ...          14.0     0.0
4         0.0          14.0  ...          14.0     0.0

```

```

      B25063024, Error  B25063025  B25063025, Error  B25063026  B25063026, Error  \
0         163.0     186.0          107.0    171.0          96.0
1         14.0      0.0           14.0      0.0          14.0
2         14.0      0.0           14.0      0.0          14.0
3         14.0      0.0           14.0      0.0          14.0
4         14.0      0.0           14.0      0.0          14.0

```

```

      B25063027  B25063027, Error  \
0  147771.0          2724.0
1     78.0           48.0
2    129.0           56.0
3     90.0           58.0
4    138.0           76.0

```

```

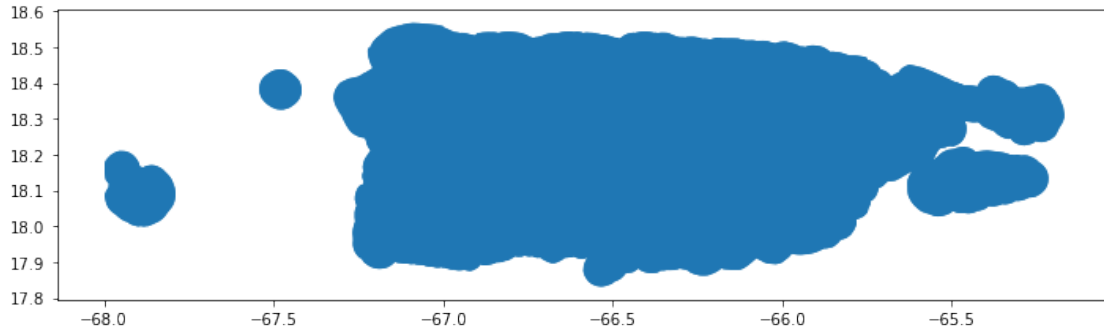
                                geometry
0 MULTIPOLYGON (((-67.20794 18.44228, -67.20844 ...
1 MULTIPOLYGON (((-66.75808 18.17672, -66.75776 ...
2 MULTIPOLYGON (((-66.79443 18.24988, -66.79381 ...
3 MULTIPOLYGON (((-66.83326 18.22802, -66.83275 ...
4 MULTIPOLYGON (((-66.83723 18.19298, -66.83685 ...

```

[5 rows x 57 columns]

```
[8]: #Okay, I'm going to plot it now
      gdf.plot(figsize=(12,12))
```

[8]: <AxesSubplot:>



#That's so cool! Note: I have also followed Yoh's advice here of using Puerto Rico completely, as opposed of just San Juan

```
[10]: # This command will let me see all the data
pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)
```

```
[11]: gdf.sample()
```

```
[11]:
```

	geoid	name	B25063001	\
1558	15000US721055204001	Block Group 1, Naranjito, PR	207.0	
	B25063001, Error	B25063002	B25063002, Error	B25063003 \
1558	117.0	110.0	110.0	0.0
	B25063003, Error	B25063004	B25063004, Error	B25063005 \
1558	14.0	0.0	14.0	0.0
	B25063005, Error	B25063006	B25063006, Error	B25063007 \
1558	14.0	0.0	14.0	0.0
	B25063007, Error	B25063008	B25063008, Error	B25063009 \
1558	14.0	0.0	14.0	44.0
	B25063009, Error	B25063010	B25063010, Error	B25063011 \
1558	68.0	0.0	14.0	52.0
	B25063011, Error	B25063012	B25063012, Error	B25063013 \
1558	84.0	0.0	14.0	0.0
	B25063013, Error	B25063014	B25063014, Error	B25063015 \
1558	14.0	0.0	14.0	0.0
	B25063015, Error	B25063016	B25063016, Error	B25063017 \
1558	14.0	14.0	24.0	0.0

```

1558      B25063017, Error  B25063018  B25063018, Error  B25063019  \
          14.0           0.0           14.0           0.0

1558      B25063019, Error  B25063020  B25063020, Error  B25063021  \
          14.0           0.0           14.0           0.0

1558      B25063021, Error  B25063022  B25063022, Error  B25063023  \
          14.0           0.0           14.0           0.0

1558      B25063023, Error  B25063024  B25063024, Error  B25063025  \
          14.0           0.0           14.0           0.0

1558      B25063025, Error  B25063026  B25063026, Error  B25063027  \
          14.0           0.0           14.0           97.0

1558      B25063027, Error                                geometry
          78.0  MULTIPOLYGON (((-66.23937 18.29248, -66.23926 ...

```

```
[12]: #data types
      gdf.info()
```

```

<class 'geopandas.geodataframe.GeoDataFrame'>
RangeIndex: 2595 entries, 0 to 2594
Data columns (total 57 columns):
#   Column                Non-Null Count  Dtype
---  -
0   geoid                  2595 non-null   object
1   name                   2595 non-null   object
2   B25063001              2595 non-null   float64
3   B25063001, Error      2595 non-null   float64
4   B25063002              2595 non-null   float64
5   B25063002, Error      2595 non-null   float64
6   B25063003              2595 non-null   float64
7   B25063003, Error      2595 non-null   float64
8   B25063004              2595 non-null   float64
9   B25063004, Error      2595 non-null   float64
10  B25063005              2595 non-null   float64
11  B25063005, Error      2595 non-null   float64
12  B25063006              2595 non-null   float64
13  B25063006, Error      2595 non-null   float64
14  B25063007              2595 non-null   float64
15  B25063007, Error      2595 non-null   float64
16  B25063008              2595 non-null   float64
17  B25063008, Error      2595 non-null   float64
18  B25063009              2595 non-null   float64
19  B25063009, Error      2595 non-null   float64

```

```

20 B25063010          2595 non-null   float64
21 B25063010, Error  2595 non-null   float64
22 B25063011          2595 non-null   float64
23 B25063011, Error  2595 non-null   float64
24 B25063012          2595 non-null   float64
25 B25063012, Error  2595 non-null   float64
26 B25063013          2595 non-null   float64
27 B25063013, Error  2595 non-null   float64
28 B25063014          2595 non-null   float64
29 B25063014, Error  2595 non-null   float64
30 B25063015          2595 non-null   float64
31 B25063015, Error  2595 non-null   float64
32 B25063016          2595 non-null   float64
33 B25063016, Error  2595 non-null   float64
34 B25063017          2595 non-null   float64
35 B25063017, Error  2595 non-null   float64
36 B25063018          2595 non-null   float64
37 B25063018, Error  2595 non-null   float64
38 B25063019          2595 non-null   float64
39 B25063019, Error  2595 non-null   float64
40 B25063020          2595 non-null   float64
41 B25063020, Error  2595 non-null   float64
42 B25063021          2595 non-null   float64
43 B25063021, Error  2595 non-null   float64
44 B25063022          2595 non-null   float64
45 B25063022, Error  2595 non-null   float64
46 B25063023          2595 non-null   float64
47 B25063023, Error  2595 non-null   float64
48 B25063024          2595 non-null   float64
49 B25063024, Error  2595 non-null   float64
50 B25063025          2595 non-null   float64
51 B25063025, Error  2595 non-null   float64
52 B25063026          2595 non-null   float64
53 B25063026, Error  2595 non-null   float64
54 B25063027          2595 non-null   float64
55 B25063027, Error  2595 non-null   float64
56 geometry          2595 non-null   geometry
dtypes: float64(54), geometry(1), object(2)
memory usage: 1.1+ MB

```

```
[13]: gdf.geoid.head()
```

```

[13]: 0          04000US72
1      15000US720019563001
2      15000US720019563002
3      15000US720019564001
4      15000US720019564002

```

Name: geoid, dtype: object

```
[14]: # check the data again
      gdf.head()
```

```
[14]:
```

	geoid		name	B25063001	\
0	04000US72		Puerto Rico	380029.0	
1	15000US720019563001	Block Group 1, Adjuntas, PR		167.0	
2	15000US720019563002	Block Group 2, Adjuntas, PR		223.0	
3	15000US720019564001	Block Group 1, Adjuntas, PR		239.0	
4	15000US720019564002	Block Group 2, Adjuntas, PR		154.0	

	B25063001, Error	B25063002	B25063002, Error	B25063003	B25063003, Error	\
0	4032.0	232258.0	3022.0	14006.0	757.0	
1	54.0	89.0	62.0	0.0	14.0	
2	89.0	94.0	67.0	0.0	14.0	
3	94.0	149.0	75.0	0.0	14.0	
4	77.0	16.0	27.0	0.0	14.0	

	B25063004	B25063004, Error	B25063005	B25063005, Error	B25063006	\
0	11090.0	786.0	10321.0	737.0	9963.0	
1	0.0	14.0	0.0	14.0	0.0	
2	0.0	14.0	12.0	19.0	0.0	
3	0.0	14.0	0.0	14.0	0.0	
4	0.0	14.0	0.0	14.0	0.0	

	B25063006, Error	B25063007	B25063007, Error	B25063008	B25063008, Error	\
0	650.0	10456.0	671.0	13538.0	734.0	
1	14.0	0.0	14.0	47.0	53.0	
2	14.0	0.0	14.0	21.0	33.0	
3	14.0	0.0	14.0	14.0	23.0	
4	14.0	0.0	14.0	0.0	14.0	

	B25063009	B25063009, Error	B25063010	B25063010, Error	B25063011	\
0	16835.0	1012.0	19097.0	988.0	19252.0	
1	10.0	16.0	0.0	14.0	17.0	
2	15.0	26.0	18.0	30.0	16.0	
3	27.0	44.0	19.0	30.0	33.0	
4	16.0	27.0	0.0	14.0	0.0	

	B25063011, Error	B25063012	B25063012, Error	B25063013	B25063013, Error	\
0	1117.0	16928.0	1016.0	15255.0	863.0	
1	28.0	0.0	14.0	15.0	24.0	
2	25.0	0.0	14.0	0.0	14.0	
3	39.0	13.0	22.0	16.0	26.0	
4	14.0	0.0	14.0	0.0	14.0	

	B25063014	B25063014, Error	B25063015	B25063015, Error	B25063016	\
0	12223.0	646.0	9815.0	761.0	9003.0	
1	0.0	14.0	0.0	14.0	0.0	
2	12.0	19.0	0.0	14.0	0.0	
3	27.0	32.0	0.0	14.0	0.0	
4	0.0	14.0	0.0	14.0	0.0	

	B25063016, Error	B25063017	B25063017, Error	B25063018	B25063018, Error	\
0	710.0	7813.0	697.0	11592.0	811.0	
1	14.0	0.0	14.0	0.0	14.0	
2	14.0	0.0	14.0	0.0	14.0	
3	14.0	0.0	14.0	0.0	14.0	
4	14.0	0.0	14.0	0.0	14.0	

	B25063019	B25063019, Error	B25063020	B25063020, Error	B25063021	\
0	7684.0	678.0	9298.0	768.0	3711.0	
1	0.0	14.0	0.0	14.0	0.0	
2	0.0	14.0	0.0	14.0	0.0	
3	0.0	14.0	0.0	14.0	0.0	
4	0.0	14.0	0.0	14.0	0.0	

	B25063021, Error	B25063022	B25063022, Error	B25063023	B25063023, Error	\
0	494.0	2640.0	433.0	946.0	259.0	
1	14.0	0.0	14.0	0.0	14.0	
2	14.0	0.0	14.0	0.0	14.0	
3	14.0	0.0	14.0	0.0	14.0	
4	14.0	0.0	14.0	0.0	14.0	

	B25063024	B25063024, Error	B25063025	B25063025, Error	B25063026	\
0	435.0	163.0	186.0	107.0	171.0	
1	0.0	14.0	0.0	14.0	0.0	
2	0.0	14.0	0.0	14.0	0.0	
3	0.0	14.0	0.0	14.0	0.0	
4	0.0	14.0	0.0	14.0	0.0	

	B25063026, Error	B25063027	B25063027, Error	\
0	96.0	147771.0	2724.0	
1	14.0	78.0	48.0	
2	14.0	129.0	56.0	
3	14.0	90.0	58.0	
4	14.0	138.0	76.0	

geometry

0	MULTIPOLYGON	(((-67.20794 18.44228, -67.20844 ...
1	MULTIPOLYGON	(((-66.75808 18.17672, -66.75776 ...
2	MULTIPOLYGON	(((-66.79443 18.24988, -66.79381 ...
3	MULTIPOLYGON	(((-66.83326 18.22802, -66.83275 ...

4 MULTIPOLYGON (((-66.83723 18.19298, -66.83685 ...

```
[15]: #Now, we're going to drop things that we don't need
# drop the row with index 0 (i.e. the first row)
gdf = gdf.drop([0])
```

```
[16]: #checking!
gdf.head()
```

[16]:

	geoid		name	B25063001	\
1	15000US720019563001	Block Group 1, Adjuntas, PR		167.0	
2	15000US720019563002	Block Group 2, Adjuntas, PR		223.0	
3	15000US720019564001	Block Group 1, Adjuntas, PR		239.0	
4	15000US720019564002	Block Group 2, Adjuntas, PR		154.0	
5	15000US720019565001	Block Group 1, Adjuntas, PR		198.0	

	B25063001, Error	B25063002	B25063002, Error	B25063003	B25063003, Error	\
1	54.0	89.0	62.0	0.0	14.0	
2	89.0	94.0	67.0	0.0	14.0	
3	94.0	149.0	75.0	0.0	14.0	
4	77.0	16.0	27.0	0.0	14.0	
5	93.0	88.0	63.0	0.0	14.0	

	B25063004	B25063004, Error	B25063005	B25063005, Error	B25063006	\
1	0.0	14.0	0.0	14.0	0.0	
2	0.0	14.0	12.0	19.0	0.0	
3	0.0	14.0	0.0	14.0	0.0	
4	0.0	14.0	0.0	14.0	0.0	
5	0.0	14.0	0.0	14.0	0.0	

	B25063006, Error	B25063007	B25063007, Error	B25063008	B25063008, Error	\
1	14.0	0.0	14.0	47.0	53.0	
2	14.0	0.0	14.0	21.0	33.0	
3	14.0	0.0	14.0	14.0	23.0	
4	14.0	0.0	14.0	0.0	14.0	
5	14.0	28.0	32.0	29.0	46.0	

	B25063009	B25063009, Error	B25063010	B25063010, Error	B25063011	\
1	10.0	16.0	0.0	14.0	17.0	
2	15.0	26.0	18.0	30.0	16.0	
3	27.0	44.0	19.0	30.0	33.0	
4	16.0	27.0	0.0	14.0	0.0	
5	0.0	14.0	15.0	24.0	0.0	

	B25063011, Error	B25063012	B25063012, Error	B25063013	B25063013, Error	\
1	28.0	0.0	14.0	15.0	24.0	
2	25.0	0.0	14.0	0.0	14.0	

3	39.0	13.0	22.0	16.0	26.0
4	14.0	0.0	14.0	0.0	14.0
5	14.0	16.0	28.0	0.0	14.0

	B25063014	B25063014, Error	B25063015	B25063015, Error	B25063016 \
1	0.0	14.0	0.0	14.0	0.0
2	12.0	19.0	0.0	14.0	0.0
3	27.0	32.0	0.0	14.0	0.0
4	0.0	14.0	0.0	14.0	0.0
5	0.0	14.0	0.0	14.0	0.0

	B25063016, Error	B25063017	B25063017, Error	B25063018	B25063018, Error \
1	14.0	0.0	14.0	0.0	14.0
2	14.0	0.0	14.0	0.0	14.0
3	14.0	0.0	14.0	0.0	14.0
4	14.0	0.0	14.0	0.0	14.0
5	14.0	0.0	14.0	0.0	14.0

	B25063019	B25063019, Error	B25063020	B25063020, Error	B25063021 \
1	0.0	14.0	0.0	14.0	0.0
2	0.0	14.0	0.0	14.0	0.0
3	0.0	14.0	0.0	14.0	0.0
4	0.0	14.0	0.0	14.0	0.0
5	0.0	14.0	0.0	14.0	0.0

	B25063021, Error	B25063022	B25063022, Error	B25063023	B25063023, Error \
1	14.0	0.0	14.0	0.0	14.0
2	14.0	0.0	14.0	0.0	14.0
3	14.0	0.0	14.0	0.0	14.0
4	14.0	0.0	14.0	0.0	14.0
5	14.0	0.0	14.0	0.0	14.0

	B25063024	B25063024, Error	B25063025	B25063025, Error	B25063026 \
1	0.0	14.0	0.0	14.0	0.0
2	0.0	14.0	0.0	14.0	0.0
3	0.0	14.0	0.0	14.0	0.0
4	0.0	14.0	0.0	14.0	0.0
5	0.0	14.0	0.0	14.0	0.0

	B25063026, Error	B25063027	B25063027, Error \
1	14.0	78.0	48.0
2	14.0	129.0	56.0
3	14.0	90.0	58.0
4	14.0	138.0	76.0
5	14.0	110.0	69.0

geometry

```
1 MULTIPOLYGON (((-66.75808 18.17672, -66.75776 ...
2 MULTIPOLYGON (((-66.79443 18.24988, -66.79381 ...
3 MULTIPOLYGON (((-66.83326 18.22802, -66.83275 ...
4 MULTIPOLYGON (((-66.83723 18.19298, -66.83685 ...
5 MULTIPOLYGON (((-66.77916 18.15809, -66.77876 ...
```

```
[17]: #good! Now let's take out the things I don't need
      # check to see if it has been deleted
      list(gdf)
```

```
[17]: ['geoid',
      'name',
      'B25063001',
      'B25063001, Error',
      'B25063002',
      'B25063002, Error',
      'B25063003',
      'B25063003, Error',
      'B25063004',
      'B25063004, Error',
      'B25063005',
      'B25063005, Error',
      'B25063006',
      'B25063006, Error',
      'B25063007',
      'B25063007, Error',
      'B25063008',
      'B25063008, Error',
      'B25063009',
      'B25063009, Error',
      'B25063010',
      'B25063010, Error',
      'B25063011',
      'B25063011, Error',
      'B25063012',
      'B25063012, Error',
      'B25063013',
      'B25063013, Error',
      'B25063014',
      'B25063014, Error',
      'B25063015',
      'B25063015, Error',
      'B25063016',
      'B25063016, Error',
      'B25063017',
      'B25063017, Error',
      'B25063018',
```

```
'B25063018, Error',  
'B25063019',  
'B25063019, Error',  
'B25063020',  
'B25063020, Error',  
'B25063021',  
'B25063021, Error',  
'B25063022',  
'B25063022, Error',  
'B25063023',  
'B25063023, Error',  
'B25063024',  
'B25063024, Error',  
'B25063025',  
'B25063025, Error',  
'B25063026',  
'B25063026, Error',  
'B25063027',  
'B25063027, Error',  
'geometry']
```

```
[18]: columns_to_keep = ['geoid',  
    'name',  
    'B25063001',  
    'B25063002',  
    'B25063003',  
    'B25063004',  
    'B25063005',  
    'B25063006',  
    'B25063007',  
    'B25063008',  
    'B25063009',  
    'B25063010',  
    'B25063011',  
    'B25063012',  
    'B25063013',  
    'B25063014',  
    'B25063015',  
    'B25063016',  
    'B25063017',  
    'B25063018',  
    'B25063019',  
    'B25063020',  
    'B25063021',  
    'B25063022',  
    'B25063023',  
    'B25063024',
```

```
'B25063025',  
'B25063026',  
'B25063027',  
'geometry']
```

```
[19]: # redefine gdf with only columns to keep  
gdf = gdf[columns_to_keep]
```

```
[21]: # check the slimmed down gdf  
gdf.head()
```

```
[21]:
```

	geoid	name	B25063001	B25063002	\
1	15000US720019563001	Block Group 1, Adjuntas, PR	167.0	89.0	
2	15000US720019563002	Block Group 2, Adjuntas, PR	223.0	94.0	
3	15000US720019564001	Block Group 1, Adjuntas, PR	239.0	149.0	
4	15000US720019564002	Block Group 2, Adjuntas, PR	154.0	16.0	
5	15000US720019565001	Block Group 1, Adjuntas, PR	198.0	88.0	

	B25063003	B25063004	B25063005	B25063006	B25063007	B25063008	\
1	0.0	0.0	0.0	0.0	0.0	47.0	
2	0.0	0.0	12.0	0.0	0.0	21.0	
3	0.0	0.0	0.0	0.0	0.0	14.0	
4	0.0	0.0	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	28.0	29.0	

	B25063009	B25063010	B25063011	B25063012	B25063013	B25063014	\
1	10.0	0.0	17.0	0.0	15.0	0.0	
2	15.0	18.0	16.0	0.0	0.0	12.0	
3	27.0	19.0	33.0	13.0	16.0	27.0	
4	16.0	0.0	0.0	0.0	0.0	0.0	
5	0.0	15.0	0.0	16.0	0.0	0.0	

	B25063015	B25063016	B25063017	B25063018	B25063019	B25063020	\
1	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	0.0	0.0	

	B25063021	B25063022	B25063023	B25063024	B25063025	B25063026	\
1	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	0.0	0.0	

	B25063027	geometry
--	-----------	----------

```
1      78.0  MULTIPOLYGON (((-66.75808 18.17672, -66.75776 ...
2     129.0  MULTIPOLYGON (((-66.79443 18.24988, -66.79381 ...
3      90.0  MULTIPOLYGON (((-66.83326 18.22802, -66.83275 ...
4     138.0  MULTIPOLYGON (((-66.83723 18.19298, -66.83685 ...
5     110.0  MULTIPOLYGON (((-66.77916 18.15809, -66.77876 ...
```

```
[22]: list(gdf)
```

```
[22]: ['geoid',
      'name',
      'B25063001',
      'B25063002',
      'B25063003',
      'B25063004',
      'B25063005',
      'B25063006',
      'B25063007',
      'B25063008',
      'B25063009',
      'B25063010',
      'B25063011',
      'B25063012',
      'B25063013',
      'B25063014',
      'B25063015',
      'B25063016',
      'B25063017',
      'B25063018',
      'B25063019',
      'B25063020',
      'B25063021',
      'B25063022',
      'B25063023',
      'B25063024',
      'B25063025',
      'B25063026',
      'B25063027',
      'geometry']
```

```
[24]: gdf.columns = ['geoid',
      'name',
      'Total',
      'With cash rent',
      'Less than $100',
      '$100 to $149',
      '$150 to $199',
      '200 to $249',
```

```
'$250 to $299',
'$300 to $349',
'$350 to $399',
'$400 to $449',
'$450 to $499',
'$500 to $549',
'$550 to $599',
'$600 to $649',
'$650 to $699',
'$700 to $749',
'$750 to $799',
'$800 to $899',
'$900 to $999',
'$1,000 to $1,249',
'$1,250 to $1,499',
'$1,500 to $1,999',
'$2,000 to $2,499',
'$2,500 to $2,999',
'$3,000 to $3,499',
'$3,500 or more',
'No cash rent',
      'geometry']
```

```
[25]: gdf.head()
```

```
[25]:
```

	geoid	name	Total	With cash rent	\
1	15000US720019563001	Block Group 1, Adjuntas, PR	167.0	89.0	
2	15000US720019563002	Block Group 2, Adjuntas, PR	223.0	94.0	
3	15000US720019564001	Block Group 1, Adjuntas, PR	239.0	149.0	
4	15000US720019564002	Block Group 2, Adjuntas, PR	154.0	16.0	
5	15000US720019565001	Block Group 1, Adjuntas, PR	198.0	88.0	

	Less than \$100	\$100 to \$149	\$150 to \$199	200 to \$249	\$250 to \$299	\
1	0.0	0.0	0.0	0.0	0.0	
2	0.0	0.0	12.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	28.0	

	\$300 to \$349	\$350 to \$399	\$400 to \$449	\$450 to \$499	\$500 to \$549	\
1	47.0	10.0	0.0	17.0	0.0	
2	21.0	15.0	18.0	16.0	0.0	
3	14.0	27.0	19.0	33.0	13.0	
4	0.0	16.0	0.0	0.0	0.0	
5	29.0	0.0	15.0	0.0	16.0	

	\$550 to \$599	\$600 to \$649	\$650 to \$699	\$700 to \$749	\$750 to \$799	\

1	15.0	0.0	0.0	0.0	0.0
2	0.0	12.0	0.0	0.0	0.0
3	16.0	27.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0
5	0.0	0.0	0.0	0.0	0.0

	\$800 to \$899	\$900 to \$999	\$1,000 to \$1,249	\$1,250 to \$1,499	\
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	

	\$1,500 to \$1,999	\$2,000 to \$2,499	\$2,500 to \$2,999	\$3,000 to \$3,499	\
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	

	\$3,500 or more	No cash rent	\
1	0.0	78.0	
2	0.0	129.0	
3	0.0	90.0	
4	0.0	138.0	
5	0.0	110.0	

```

                                geometry
1 MULTIPOLYGON (((-66.75808 18.17672, -66.75776 ...
2 MULTIPOLYGON (((-66.79443 18.24988, -66.79381 ...
3 MULTIPOLYGON (((-66.83326 18.22802, -66.83275 ...
4 MULTIPOLYGON (((-66.83723 18.19298, -66.83685 ...
5 MULTIPOLYGON (((-66.77916 18.15809, -66.77876 ...

```

```
[26]: # get a random record
random_tract = gdf.sample()
random_tract
```

```
[26]:
          geoid          name  Total  With cash rent  \
2027  15000US721270055001  Block Group 1, San Juan, PR  195.0      154.0

      Less than $100  $100 to $149  $150 to $199  200 to $249  $250 to $299  \
2027              0.0              0.0              0.0              0.0      35.0

      $300 to $349  $350 to $399  $400 to $449  $450 to $499  $500 to $549  \
2027              0.0              0.0              0.0          39.0          29.0
```

```

    $550 to $599  $600 to $649  $650 to $699  $700 to $749  $750 to $799  \
2027             0.0           0.0           0.0           51.0           0.0

    $800 to $899  $900 to $999  $1,000 to $1,249  $1,250 to $1,499  \
2027             0.0           0.0           0.0           0.0

    $1,500 to $1,999  $2,000 to $2,499  $2,500 to $2,999  $3,000 to $3,499  \
2027             0.0           0.0           0.0           0.0

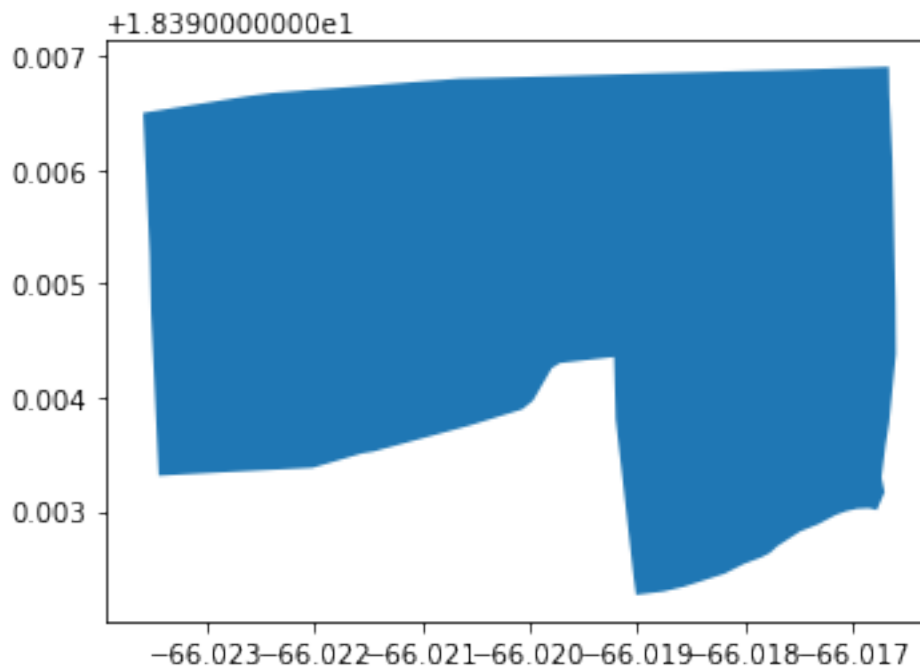
    $3,500 or more  No cash rent  \
2027             0.0           41.0

                                geometry
2027 MULTIPOLYGON (((-66.02359 18.39650, -66.02288 ...

```

```
[27]: random_tract.plot()
```

```
[27]: <AxesSubplot:>
```



```
[28]: #Okay, that looks like something. I am not sure yet what it represents, but
      ->from the gdf it looks like it worked
```

```
[29]: # example usage of iloc to get the total population of our random record
      # "for the 0th record, get the value in the Total column"
      random_tract.iloc[0]['Total']
```


[29]: 195.0

```
[30]: # print this out in plain english
print('Total population: ' + str(random_tract.iloc[0]['Total']))
```

Total population: 195.0

```
[31]: # access a single column like df['col_name']
gdf['Total'].head()
```

```
[31]: 1    167.0
      2    223.0
      3    239.0
      4    154.0
      5    198.0
      Name: Total, dtype: float64
```

```
[32]: # What is the mean?
gdf['Total'].mean()
```

[32]: 146.50308404009252

```
[33]: # What is the median?
gdf['Total'].median()
```

[33]: 126.0

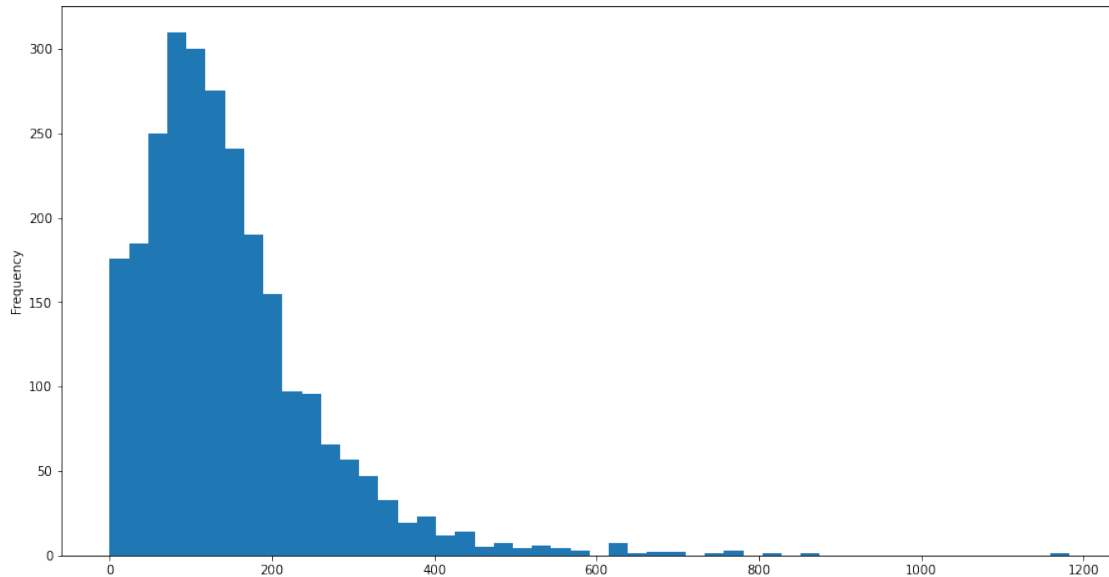
Huh, interesting

```
[34]: # get some stats
gdf['Total'].describe()
```

```
[34]: count    2594.000000
      mean     146.503084
      std     110.231525
      min       0.000000
      25%      74.000000
      50%     126.000000
      75%     192.000000
      max    1183.000000
      Name: Total, dtype: float64
```

```
[35]: # plot it as a histogram with 50 bins
gdf['Total'].plot.hist(figsize=(15,8),bins=50)
```

[35]: <AxesSubplot:ylabel='Frequency'>



#According to the US Census definition, this “gross rent” variable includes “the contract rent plus the estimated average monthly cost of utilities (electricity, gas, and water and sewer) and fuels (oil, coal, kerosene, wood, etc.”. I think the fact people stay on a cheaper side might be indicative of people around the island actually *owing* their houses. It might be worth doing this map for only San Juan, as I expect there to be more rented properties.

```
[36]: gdf_sorted = gdf.sort_values(by='Total',ascending = False)
```

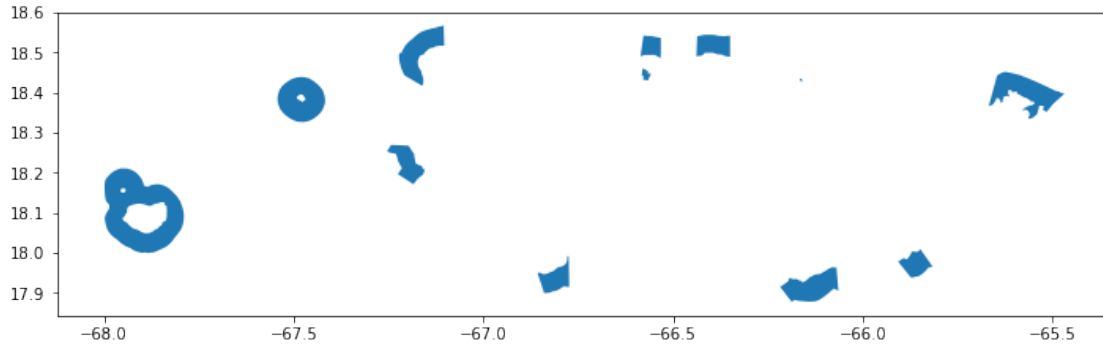
```
[37]: # display the data, but just a few columns to keep it clean
gdf_sorted[['geoid','Total']].head(10)
```

```
[37]:
```

	geoid	Total
2067	15000US721270068001	1183.0
1891	15000US721270013011	864.0
267	15000US720210301032	812.0
1892	15000US721270013021	773.0
688	15000US720310504023	769.0
1495	15000US720970821031	761.0
2184	15000US721270098005	742.0
1894	15000US721270013023	693.0
1944	15000US721270035012	688.0
2046	15000US721270061013	679.0

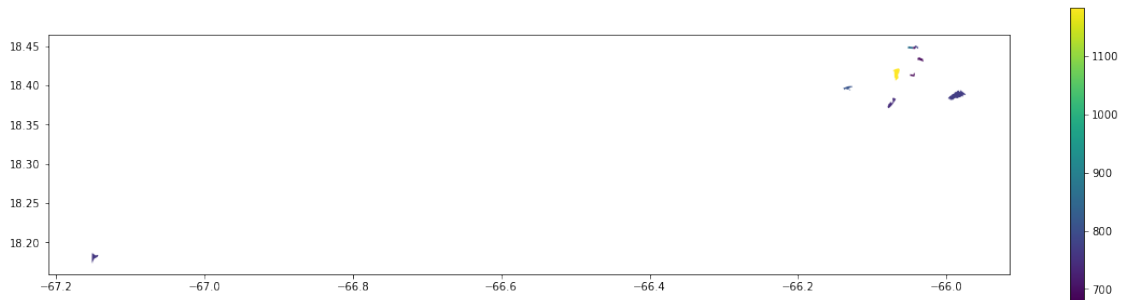
```
[38]: # plot it
gdf_sorted.tail(10).plot(figsize=(12,12))
```

```
[38]: <AxesSubplot:>
```



```
[43]: # Make it prettier
gdf_sorted.head(10).plot(figsize=(20,5),
                          column='Total',
                          legend=True)
```

[43]: <AxesSubplot:>



```
[44]: # subset the data so that we can see the data per row...
# in other words, this syntax is asking to "show me the values in my dataframe
↳ that match this filter
gdf[gdf['Total']==0]
```

```
[44]:
```

	geoid	name	Total	\
18	15000US720034302003	Block Group 3, Aguada, PR	0.0	
41	15000US720039900000	Block Group 0, Aguada, PR	0.0	
86	15000US720059900000	Block Group 0, Aguadilla, PR	0.0	
119	15000US720118101002	Block Group 2, Añasco, PR	0.0	
129	15000US720118105001	Block Group 1, Añasco, PR	0.0	
139	15000US720119900000	Block Group 0, Añasco, PR	0.0	
214	15000US720139929000	Block Group 0, Arecibo, PR	0.0	
219	15000US720152801023	Block Group 3, Arroyo, PR	0.0	
228	15000US720159915000	Block Group 0, Arroyo, PR	0.0	

240	15000US720175903004	Block Group 4, Barceloneta, PR	0.0
245	15000US720179900000	Block Group 0, Barceloneta, PR	0.0
313	15000US720210310314	Block Group 4, Bayamón, PR	0.0
403	15000US720210317043	Block Group 3, Bayamón, PR	0.0
406	15000US720210318002	Block Group 2, Bayamón, PR	0.0
455	15000US720239800001	Block Group 1, Cabo Rojo, PR	0.0
456	15000US720239900000	Block Group 0, Cabo Rojo, PR	0.0
465	15000US720252002004	Block Group 4, Caguas, PR	0.0
471	15000US720252003026	Block Group 6, Caguas, PR	0.0
485	15000US720252005001	Block Group 1, Caguas, PR	0.0
570	15000US720252028003	Block Group 3, Caguas, PR	0.0
592	15000US720279900000	Block Group 0, Camuy, PR	0.0
644	15000US720310502122	Block Group 2, Carolina, PR	0.0
645	15000US720310502123	Block Group 3, Carolina, PR	0.0
731	15000US720310509024	Block Group 4, Carolina, PR	0.0
733	15000US720310510011	Block Group 1, Carolina, PR	0.0
747	15000US720319800031	Block Group 1, Carolina, PR	0.0
748	15000US720319900000	Block Group 0, Carolina, PR	0.0
767	15000US720330205002	Block Group 2, Cataño, PR	0.0
769	15000US720339800011	Block Group 1, Cataño, PR	0.0
770	15000US720339800021	Block Group 1, Cataño, PR	0.0
772	15000US720339800041	Block Group 1, Cataño, PR	0.0
773	15000US720339800051	Block Group 1, Cataño, PR	0.0
774	15000US720339800061	Block Group 1, Cataño, PR	0.0
775	15000US720339800071	Block Group 1, Cataño, PR	0.0
776	15000US720339800081	Block Group 1, Cataño, PR	0.0
777	15000US720339902010	Block Group 0, Cataño, PR	0.0
809	15000US720371601001	Block Group 1, Ceiba, PR	0.0
810	15000US720371601002	Block Group 2, Ceiba, PR	0.0
819	15000US720379900000	Block Group 0, Ceiba, PR	0.0
833	15000US720412401012	Block Group 2, Cidra, PR	0.0
908	15000US720475306001	Block Group 1, Corozal, PR	0.0
914	15000US720499905010	Block Group 0, Culebra, PR	0.0
934	15000US720519900210	Block Group 0, Dorado, PR	0.0
962	15000US720539901030	Block Group 0, Fajardo, PR	0.0
986	15000US720559900000	Block Group 0, Guánica, PR	0.0
988	15000US720572701002	Block Group 2, Guayama, PR	0.0
1013	15000US720572707002	Block Group 2, Guayama, PR	0.0
1019	15000US720579926000	Block Group 0, Guayama, PR	0.0
1035	15000US720599900010	Block Group 0, Guayanilla, PR	0.0
1053	15000US720610403042	Block Group 2, Guaynabo, PR	0.0
1089	15000US720610406026	Block Group 6, Guaynabo, PR	0.0
1107	15000US720632102022	Block Group 2, Gurabo, PR	0.0
1133	15000US720653103003	Block Group 3, Hatillo, PR	0.0
1147	15000US720659900160	Block Group 0, Hatillo, PR	0.0
1195	15000US720699918000	Block Group 0, Humacao, PR	0.0
1222	15000US720719900000	Block Group 0, Isabel, PR	0.0

1265	15000US720759900010	Block Group 0, Juana Díaz, PR	0.0
1305	15000US720799911000	Block Group 0, Lajas, PR	0.0
1370	15000US720879900000	Block Group 0, Loíza, PR	0.0
1385	15000US720899900010	Block Group 0, Luquillo, PR	0.0
1413	15000US720919900250	Block Group 0, Manatí, PR	0.0
1429	15000US720959900000	Block Group 0, Maunabo, PR	0.0
1499	15000US720979900000	Block Group 0, Mayagüez, PR	0.0
1549	15000US721039900130	Block Group 0, Naguabo, PR	0.0
1589	15000US721092902004	Block Group 4, Patillas, PR	0.0
1596	15000US721099900000	Block Group 0, Patillas, PR	0.0
1611	15000US721119900000	Block Group 0, Peñuelas, PR	0.0
1695	15000US721130722026	Block Group 6, Ponce, PR	0.0
1739	15000US721130730081	Block Group 1, Ponce, PR	0.0
1745	15000US721139930000	Block Group 0, Ponce, PR	0.0
1759	15000US721159900000	Block Group 0, Quebradillas, PR	0.0
1770	15000US721179904000	Block Group 0, Rincón, PR	0.0
1773	15000US721191301013	Block Group 3, Río Grande, PR	0.0
1804	15000US721199927000	Block Group 0, Río Grande, PR	0.0
1805	15000US721219603001	Block Group 1, Sabana Grande, PR	0.0
1822	15000US721239526001	Block Group 1, Salinas, PR	0.0
1844	15000US721239928000	Block Group 0, Salinas, PR	0.0
1862	15000US721258406002	Block Group 2, San Germán, PR	0.0
1968	15000US721270044002	Block Group 2, San Juan, PR	0.0
2038	15000US721270058002	Block Group 2, San Juan, PR	0.0
2039	15000US721270058003	Block Group 3, San Juan, PR	0.0
2136	15000US721270086025	Block Group 5, San Juan, PR	0.0
2181	15000US721270098002	Block Group 2, San Juan, PR	0.0
2227	15000US721279801011	Block Group 1, San Juan, PR	0.0
2228	15000US721279801021	Block Group 1, San Juan, PR	0.0
2230	15000US721279801041	Block Group 1, San Juan, PR	0.0
2231	15000US721279801051	Block Group 1, San Juan, PR	0.0
2232	15000US721279801071	Block Group 1, San Juan, PR	0.0
2233	15000US721279801081	Block Group 1, San Juan, PR	0.0
2235	15000US721279803001	Block Group 1, San Juan, PR	0.0
2236	15000US721279900000	Block Group 0, San Juan, PR	0.0
2298	15000US721339900000	Block Group 0, Santa Isabel, PR	0.0
2327	15000US721355106022	Block Group 2, Toa Alta, PR	0.0
2365	15000US721371214003	Block Group 3, Toa Baja, PR	0.0
2369	15000US721371217023	Block Group 3, Toa Baja, PR	0.0
2392	15000US721379900000	Block Group 0, Toa Baja, PR	0.0
2395	15000US721390601023	Block Group 3, Trujillo Alto, PR	0.0
2480	15000US721439900000	Block Group 0, Vega Alta, PR	0.0
2487	15000US721455602042	Block Group 2, Vega Baja, PR	0.0
2505	15000US721455607013	Block Group 3, Vega Baja, PR	0.0
2506	15000US721455607014	Block Group 4, Vega Baja, PR	0.0
2518	15000US721459900000	Block Group 0, Vega Baja, PR	0.0
2525	15000US721479900000	Block Group 0, Vieques, PR	0.0

2566 15000US721519900000 Block Group 0, Yabucoa, PR 0.0

	With cash rent	Less than \$100	\$100 to \$149	\$150 to \$199	200 to \$249	\
18	0.0	0.0	0.0	0.0	0.0	0.0
41	0.0	0.0	0.0	0.0	0.0	0.0
86	0.0	0.0	0.0	0.0	0.0	0.0
119	0.0	0.0	0.0	0.0	0.0	0.0
129	0.0	0.0	0.0	0.0	0.0	0.0
139	0.0	0.0	0.0	0.0	0.0	0.0
214	0.0	0.0	0.0	0.0	0.0	0.0
219	0.0	0.0	0.0	0.0	0.0	0.0
228	0.0	0.0	0.0	0.0	0.0	0.0
240	0.0	0.0	0.0	0.0	0.0	0.0
245	0.0	0.0	0.0	0.0	0.0	0.0
313	0.0	0.0	0.0	0.0	0.0	0.0
403	0.0	0.0	0.0	0.0	0.0	0.0
406	0.0	0.0	0.0	0.0	0.0	0.0
455	0.0	0.0	0.0	0.0	0.0	0.0
456	0.0	0.0	0.0	0.0	0.0	0.0
465	0.0	0.0	0.0	0.0	0.0	0.0
471	0.0	0.0	0.0	0.0	0.0	0.0
485	0.0	0.0	0.0	0.0	0.0	0.0
570	0.0	0.0	0.0	0.0	0.0	0.0
592	0.0	0.0	0.0	0.0	0.0	0.0
644	0.0	0.0	0.0	0.0	0.0	0.0
645	0.0	0.0	0.0	0.0	0.0	0.0
731	0.0	0.0	0.0	0.0	0.0	0.0
733	0.0	0.0	0.0	0.0	0.0	0.0
747	0.0	0.0	0.0	0.0	0.0	0.0
748	0.0	0.0	0.0	0.0	0.0	0.0
767	0.0	0.0	0.0	0.0	0.0	0.0
769	0.0	0.0	0.0	0.0	0.0	0.0
770	0.0	0.0	0.0	0.0	0.0	0.0
772	0.0	0.0	0.0	0.0	0.0	0.0
773	0.0	0.0	0.0	0.0	0.0	0.0
774	0.0	0.0	0.0	0.0	0.0	0.0
775	0.0	0.0	0.0	0.0	0.0	0.0
776	0.0	0.0	0.0	0.0	0.0	0.0
777	0.0	0.0	0.0	0.0	0.0	0.0
809	0.0	0.0	0.0	0.0	0.0	0.0
810	0.0	0.0	0.0	0.0	0.0	0.0
819	0.0	0.0	0.0	0.0	0.0	0.0
833	0.0	0.0	0.0	0.0	0.0	0.0
908	0.0	0.0	0.0	0.0	0.0	0.0
914	0.0	0.0	0.0	0.0	0.0	0.0
934	0.0	0.0	0.0	0.0	0.0	0.0
962	0.0	0.0	0.0	0.0	0.0	0.0

986	0.0	0.0	0.0	0.0	0.0
988	0.0	0.0	0.0	0.0	0.0
1013	0.0	0.0	0.0	0.0	0.0
1019	0.0	0.0	0.0	0.0	0.0
1035	0.0	0.0	0.0	0.0	0.0
1053	0.0	0.0	0.0	0.0	0.0
1089	0.0	0.0	0.0	0.0	0.0
1107	0.0	0.0	0.0	0.0	0.0
1133	0.0	0.0	0.0	0.0	0.0
1147	0.0	0.0	0.0	0.0	0.0
1195	0.0	0.0	0.0	0.0	0.0
1222	0.0	0.0	0.0	0.0	0.0
1265	0.0	0.0	0.0	0.0	0.0
1305	0.0	0.0	0.0	0.0	0.0
1370	0.0	0.0	0.0	0.0	0.0
1385	0.0	0.0	0.0	0.0	0.0
1413	0.0	0.0	0.0	0.0	0.0
1429	0.0	0.0	0.0	0.0	0.0
1499	0.0	0.0	0.0	0.0	0.0
1549	0.0	0.0	0.0	0.0	0.0
1589	0.0	0.0	0.0	0.0	0.0
1596	0.0	0.0	0.0	0.0	0.0
1611	0.0	0.0	0.0	0.0	0.0
1695	0.0	0.0	0.0	0.0	0.0
1739	0.0	0.0	0.0	0.0	0.0
1745	0.0	0.0	0.0	0.0	0.0
1759	0.0	0.0	0.0	0.0	0.0
1770	0.0	0.0	0.0	0.0	0.0
1773	0.0	0.0	0.0	0.0	0.0
1804	0.0	0.0	0.0	0.0	0.0
1805	0.0	0.0	0.0	0.0	0.0
1822	0.0	0.0	0.0	0.0	0.0
1844	0.0	0.0	0.0	0.0	0.0
1862	0.0	0.0	0.0	0.0	0.0
1968	0.0	0.0	0.0	0.0	0.0
2038	0.0	0.0	0.0	0.0	0.0
2039	0.0	0.0	0.0	0.0	0.0
2136	0.0	0.0	0.0	0.0	0.0
2181	0.0	0.0	0.0	0.0	0.0
2227	0.0	0.0	0.0	0.0	0.0
2228	0.0	0.0	0.0	0.0	0.0
2230	0.0	0.0	0.0	0.0	0.0
2231	0.0	0.0	0.0	0.0	0.0
2232	0.0	0.0	0.0	0.0	0.0
2233	0.0	0.0	0.0	0.0	0.0
2235	0.0	0.0	0.0	0.0	0.0
2236	0.0	0.0	0.0	0.0	0.0

2298	0.0	0.0	0.0	0.0	0.0
2327	0.0	0.0	0.0	0.0	0.0
2365	0.0	0.0	0.0	0.0	0.0
2369	0.0	0.0	0.0	0.0	0.0
2392	0.0	0.0	0.0	0.0	0.0
2395	0.0	0.0	0.0	0.0	0.0
2480	0.0	0.0	0.0	0.0	0.0
2487	0.0	0.0	0.0	0.0	0.0
2505	0.0	0.0	0.0	0.0	0.0
2506	0.0	0.0	0.0	0.0	0.0
2518	0.0	0.0	0.0	0.0	0.0
2525	0.0	0.0	0.0	0.0	0.0
2566	0.0	0.0	0.0	0.0	0.0

	\$250 to \$299	\$300 to \$349	\$350 to \$399	\$400 to \$449	\$450 to \$499 \
18	0.0	0.0	0.0	0.0	0.0
41	0.0	0.0	0.0	0.0	0.0
86	0.0	0.0	0.0	0.0	0.0
119	0.0	0.0	0.0	0.0	0.0
129	0.0	0.0	0.0	0.0	0.0
139	0.0	0.0	0.0	0.0	0.0
214	0.0	0.0	0.0	0.0	0.0
219	0.0	0.0	0.0	0.0	0.0
228	0.0	0.0	0.0	0.0	0.0
240	0.0	0.0	0.0	0.0	0.0
245	0.0	0.0	0.0	0.0	0.0
313	0.0	0.0	0.0	0.0	0.0
403	0.0	0.0	0.0	0.0	0.0
406	0.0	0.0	0.0	0.0	0.0
455	0.0	0.0	0.0	0.0	0.0
456	0.0	0.0	0.0	0.0	0.0
465	0.0	0.0	0.0	0.0	0.0
471	0.0	0.0	0.0	0.0	0.0
485	0.0	0.0	0.0	0.0	0.0
570	0.0	0.0	0.0	0.0	0.0
592	0.0	0.0	0.0	0.0	0.0
644	0.0	0.0	0.0	0.0	0.0
645	0.0	0.0	0.0	0.0	0.0
731	0.0	0.0	0.0	0.0	0.0
733	0.0	0.0	0.0	0.0	0.0
747	0.0	0.0	0.0	0.0	0.0
748	0.0	0.0	0.0	0.0	0.0
767	0.0	0.0	0.0	0.0	0.0
769	0.0	0.0	0.0	0.0	0.0
770	0.0	0.0	0.0	0.0	0.0
772	0.0	0.0	0.0	0.0	0.0
773	0.0	0.0	0.0	0.0	0.0

774	0.0	0.0	0.0	0.0	0.0
775	0.0	0.0	0.0	0.0	0.0
776	0.0	0.0	0.0	0.0	0.0
777	0.0	0.0	0.0	0.0	0.0
809	0.0	0.0	0.0	0.0	0.0
810	0.0	0.0	0.0	0.0	0.0
819	0.0	0.0	0.0	0.0	0.0
833	0.0	0.0	0.0	0.0	0.0
908	0.0	0.0	0.0	0.0	0.0
914	0.0	0.0	0.0	0.0	0.0
934	0.0	0.0	0.0	0.0	0.0
962	0.0	0.0	0.0	0.0	0.0
986	0.0	0.0	0.0	0.0	0.0
988	0.0	0.0	0.0	0.0	0.0
1013	0.0	0.0	0.0	0.0	0.0
1019	0.0	0.0	0.0	0.0	0.0
1035	0.0	0.0	0.0	0.0	0.0
1053	0.0	0.0	0.0	0.0	0.0
1089	0.0	0.0	0.0	0.0	0.0
1107	0.0	0.0	0.0	0.0	0.0
1133	0.0	0.0	0.0	0.0	0.0
1147	0.0	0.0	0.0	0.0	0.0
1195	0.0	0.0	0.0	0.0	0.0
1222	0.0	0.0	0.0	0.0	0.0
1265	0.0	0.0	0.0	0.0	0.0
1305	0.0	0.0	0.0	0.0	0.0
1370	0.0	0.0	0.0	0.0	0.0
1385	0.0	0.0	0.0	0.0	0.0
1413	0.0	0.0	0.0	0.0	0.0
1429	0.0	0.0	0.0	0.0	0.0
1499	0.0	0.0	0.0	0.0	0.0
1549	0.0	0.0	0.0	0.0	0.0
1589	0.0	0.0	0.0	0.0	0.0
1596	0.0	0.0	0.0	0.0	0.0
1611	0.0	0.0	0.0	0.0	0.0
1695	0.0	0.0	0.0	0.0	0.0
1739	0.0	0.0	0.0	0.0	0.0
1745	0.0	0.0	0.0	0.0	0.0
1759	0.0	0.0	0.0	0.0	0.0
1770	0.0	0.0	0.0	0.0	0.0
1773	0.0	0.0	0.0	0.0	0.0
1804	0.0	0.0	0.0	0.0	0.0
1805	0.0	0.0	0.0	0.0	0.0
1822	0.0	0.0	0.0	0.0	0.0
1844	0.0	0.0	0.0	0.0	0.0
1862	0.0	0.0	0.0	0.0	0.0
1968	0.0	0.0	0.0	0.0	0.0

2038	0.0	0.0	0.0	0.0	0.0
2039	0.0	0.0	0.0	0.0	0.0
2136	0.0	0.0	0.0	0.0	0.0
2181	0.0	0.0	0.0	0.0	0.0
2227	0.0	0.0	0.0	0.0	0.0
2228	0.0	0.0	0.0	0.0	0.0
2230	0.0	0.0	0.0	0.0	0.0
2231	0.0	0.0	0.0	0.0	0.0
2232	0.0	0.0	0.0	0.0	0.0
2233	0.0	0.0	0.0	0.0	0.0
2235	0.0	0.0	0.0	0.0	0.0
2236	0.0	0.0	0.0	0.0	0.0
2298	0.0	0.0	0.0	0.0	0.0
2327	0.0	0.0	0.0	0.0	0.0
2365	0.0	0.0	0.0	0.0	0.0
2369	0.0	0.0	0.0	0.0	0.0
2392	0.0	0.0	0.0	0.0	0.0
2395	0.0	0.0	0.0	0.0	0.0
2480	0.0	0.0	0.0	0.0	0.0
2487	0.0	0.0	0.0	0.0	0.0
2505	0.0	0.0	0.0	0.0	0.0
2506	0.0	0.0	0.0	0.0	0.0
2518	0.0	0.0	0.0	0.0	0.0
2525	0.0	0.0	0.0	0.0	0.0
2566	0.0	0.0	0.0	0.0	0.0

	\$500 to \$549	\$550 to \$599	\$600 to \$649	\$650 to \$699	\$700 to \$749 \
18	0.0	0.0	0.0	0.0	0.0
41	0.0	0.0	0.0	0.0	0.0
86	0.0	0.0	0.0	0.0	0.0
119	0.0	0.0	0.0	0.0	0.0
129	0.0	0.0	0.0	0.0	0.0
139	0.0	0.0	0.0	0.0	0.0
214	0.0	0.0	0.0	0.0	0.0
219	0.0	0.0	0.0	0.0	0.0
228	0.0	0.0	0.0	0.0	0.0
240	0.0	0.0	0.0	0.0	0.0
245	0.0	0.0	0.0	0.0	0.0
313	0.0	0.0	0.0	0.0	0.0
403	0.0	0.0	0.0	0.0	0.0
406	0.0	0.0	0.0	0.0	0.0
455	0.0	0.0	0.0	0.0	0.0
456	0.0	0.0	0.0	0.0	0.0
465	0.0	0.0	0.0	0.0	0.0
471	0.0	0.0	0.0	0.0	0.0
485	0.0	0.0	0.0	0.0	0.0
570	0.0	0.0	0.0	0.0	0.0

592	0.0	0.0	0.0	0.0	0.0
644	0.0	0.0	0.0	0.0	0.0
645	0.0	0.0	0.0	0.0	0.0
731	0.0	0.0	0.0	0.0	0.0
733	0.0	0.0	0.0	0.0	0.0
747	0.0	0.0	0.0	0.0	0.0
748	0.0	0.0	0.0	0.0	0.0
767	0.0	0.0	0.0	0.0	0.0
769	0.0	0.0	0.0	0.0	0.0
770	0.0	0.0	0.0	0.0	0.0
772	0.0	0.0	0.0	0.0	0.0
773	0.0	0.0	0.0	0.0	0.0
774	0.0	0.0	0.0	0.0	0.0
775	0.0	0.0	0.0	0.0	0.0
776	0.0	0.0	0.0	0.0	0.0
777	0.0	0.0	0.0	0.0	0.0
809	0.0	0.0	0.0	0.0	0.0
810	0.0	0.0	0.0	0.0	0.0
819	0.0	0.0	0.0	0.0	0.0
833	0.0	0.0	0.0	0.0	0.0
908	0.0	0.0	0.0	0.0	0.0
914	0.0	0.0	0.0	0.0	0.0
934	0.0	0.0	0.0	0.0	0.0
962	0.0	0.0	0.0	0.0	0.0
986	0.0	0.0	0.0	0.0	0.0
988	0.0	0.0	0.0	0.0	0.0
1013	0.0	0.0	0.0	0.0	0.0
1019	0.0	0.0	0.0	0.0	0.0
1035	0.0	0.0	0.0	0.0	0.0
1053	0.0	0.0	0.0	0.0	0.0
1089	0.0	0.0	0.0	0.0	0.0
1107	0.0	0.0	0.0	0.0	0.0
1133	0.0	0.0	0.0	0.0	0.0
1147	0.0	0.0	0.0	0.0	0.0
1195	0.0	0.0	0.0	0.0	0.0
1222	0.0	0.0	0.0	0.0	0.0
1265	0.0	0.0	0.0	0.0	0.0
1305	0.0	0.0	0.0	0.0	0.0
1370	0.0	0.0	0.0	0.0	0.0
1385	0.0	0.0	0.0	0.0	0.0
1413	0.0	0.0	0.0	0.0	0.0
1429	0.0	0.0	0.0	0.0	0.0
1499	0.0	0.0	0.0	0.0	0.0
1549	0.0	0.0	0.0	0.0	0.0
1589	0.0	0.0	0.0	0.0	0.0
1596	0.0	0.0	0.0	0.0	0.0
1611	0.0	0.0	0.0	0.0	0.0

1695	0.0	0.0	0.0	0.0	0.0
1739	0.0	0.0	0.0	0.0	0.0
1745	0.0	0.0	0.0	0.0	0.0
1759	0.0	0.0	0.0	0.0	0.0
1770	0.0	0.0	0.0	0.0	0.0
1773	0.0	0.0	0.0	0.0	0.0
1804	0.0	0.0	0.0	0.0	0.0
1805	0.0	0.0	0.0	0.0	0.0
1822	0.0	0.0	0.0	0.0	0.0
1844	0.0	0.0	0.0	0.0	0.0
1862	0.0	0.0	0.0	0.0	0.0
1968	0.0	0.0	0.0	0.0	0.0
2038	0.0	0.0	0.0	0.0	0.0
2039	0.0	0.0	0.0	0.0	0.0
2136	0.0	0.0	0.0	0.0	0.0
2181	0.0	0.0	0.0	0.0	0.0
2227	0.0	0.0	0.0	0.0	0.0
2228	0.0	0.0	0.0	0.0	0.0
2230	0.0	0.0	0.0	0.0	0.0
2231	0.0	0.0	0.0	0.0	0.0
2232	0.0	0.0	0.0	0.0	0.0
2233	0.0	0.0	0.0	0.0	0.0
2235	0.0	0.0	0.0	0.0	0.0
2236	0.0	0.0	0.0	0.0	0.0
2298	0.0	0.0	0.0	0.0	0.0
2327	0.0	0.0	0.0	0.0	0.0
2365	0.0	0.0	0.0	0.0	0.0
2369	0.0	0.0	0.0	0.0	0.0
2392	0.0	0.0	0.0	0.0	0.0
2395	0.0	0.0	0.0	0.0	0.0
2480	0.0	0.0	0.0	0.0	0.0
2487	0.0	0.0	0.0	0.0	0.0
2505	0.0	0.0	0.0	0.0	0.0
2506	0.0	0.0	0.0	0.0	0.0
2518	0.0	0.0	0.0	0.0	0.0
2525	0.0	0.0	0.0	0.0	0.0
2566	0.0	0.0	0.0	0.0	0.0

	\$750 to \$799	\$800 to \$899	\$900 to \$999	\$1,000 to \$1,249	\
18	0.0	0.0	0.0	0.0	
41	0.0	0.0	0.0	0.0	
86	0.0	0.0	0.0	0.0	
119	0.0	0.0	0.0	0.0	
129	0.0	0.0	0.0	0.0	
139	0.0	0.0	0.0	0.0	
214	0.0	0.0	0.0	0.0	
219	0.0	0.0	0.0	0.0	

228	0.0	0.0	0.0	0.0
240	0.0	0.0	0.0	0.0
245	0.0	0.0	0.0	0.0
313	0.0	0.0	0.0	0.0
403	0.0	0.0	0.0	0.0
406	0.0	0.0	0.0	0.0
455	0.0	0.0	0.0	0.0
456	0.0	0.0	0.0	0.0
465	0.0	0.0	0.0	0.0
471	0.0	0.0	0.0	0.0
485	0.0	0.0	0.0	0.0
570	0.0	0.0	0.0	0.0
592	0.0	0.0	0.0	0.0
644	0.0	0.0	0.0	0.0
645	0.0	0.0	0.0	0.0
731	0.0	0.0	0.0	0.0
733	0.0	0.0	0.0	0.0
747	0.0	0.0	0.0	0.0
748	0.0	0.0	0.0	0.0
767	0.0	0.0	0.0	0.0
769	0.0	0.0	0.0	0.0
770	0.0	0.0	0.0	0.0
772	0.0	0.0	0.0	0.0
773	0.0	0.0	0.0	0.0
774	0.0	0.0	0.0	0.0
775	0.0	0.0	0.0	0.0
776	0.0	0.0	0.0	0.0
777	0.0	0.0	0.0	0.0
809	0.0	0.0	0.0	0.0
810	0.0	0.0	0.0	0.0
819	0.0	0.0	0.0	0.0
833	0.0	0.0	0.0	0.0
908	0.0	0.0	0.0	0.0
914	0.0	0.0	0.0	0.0
934	0.0	0.0	0.0	0.0
962	0.0	0.0	0.0	0.0
986	0.0	0.0	0.0	0.0
988	0.0	0.0	0.0	0.0
1013	0.0	0.0	0.0	0.0
1019	0.0	0.0	0.0	0.0
1035	0.0	0.0	0.0	0.0
1053	0.0	0.0	0.0	0.0
1089	0.0	0.0	0.0	0.0
1107	0.0	0.0	0.0	0.0
1133	0.0	0.0	0.0	0.0
1147	0.0	0.0	0.0	0.0
1195	0.0	0.0	0.0	0.0

1222	0.0	0.0	0.0	0.0
1265	0.0	0.0	0.0	0.0
1305	0.0	0.0	0.0	0.0
1370	0.0	0.0	0.0	0.0
1385	0.0	0.0	0.0	0.0
1413	0.0	0.0	0.0	0.0
1429	0.0	0.0	0.0	0.0
1499	0.0	0.0	0.0	0.0
1549	0.0	0.0	0.0	0.0
1589	0.0	0.0	0.0	0.0
1596	0.0	0.0	0.0	0.0
1611	0.0	0.0	0.0	0.0
1695	0.0	0.0	0.0	0.0
1739	0.0	0.0	0.0	0.0
1745	0.0	0.0	0.0	0.0
1759	0.0	0.0	0.0	0.0
1770	0.0	0.0	0.0	0.0
1773	0.0	0.0	0.0	0.0
1804	0.0	0.0	0.0	0.0
1805	0.0	0.0	0.0	0.0
1822	0.0	0.0	0.0	0.0
1844	0.0	0.0	0.0	0.0
1862	0.0	0.0	0.0	0.0
1968	0.0	0.0	0.0	0.0
2038	0.0	0.0	0.0	0.0
2039	0.0	0.0	0.0	0.0
2136	0.0	0.0	0.0	0.0
2181	0.0	0.0	0.0	0.0
2227	0.0	0.0	0.0	0.0
2228	0.0	0.0	0.0	0.0
2230	0.0	0.0	0.0	0.0
2231	0.0	0.0	0.0	0.0
2232	0.0	0.0	0.0	0.0
2233	0.0	0.0	0.0	0.0
2235	0.0	0.0	0.0	0.0
2236	0.0	0.0	0.0	0.0
2298	0.0	0.0	0.0	0.0
2327	0.0	0.0	0.0	0.0
2365	0.0	0.0	0.0	0.0
2369	0.0	0.0	0.0	0.0
2392	0.0	0.0	0.0	0.0
2395	0.0	0.0	0.0	0.0
2480	0.0	0.0	0.0	0.0
2487	0.0	0.0	0.0	0.0
2505	0.0	0.0	0.0	0.0
2506	0.0	0.0	0.0	0.0
2518	0.0	0.0	0.0	0.0

2525	0.0	0.0	0.0	0.0
2566	0.0	0.0	0.0	0.0

	\$1,250 to \$1,499	\$1,500 to \$1,999	\$2,000 to \$2,499	\$2,500 to \$2,999	\
18	0.0	0.0	0.0	0.0	
41	0.0	0.0	0.0	0.0	
86	0.0	0.0	0.0	0.0	
119	0.0	0.0	0.0	0.0	
129	0.0	0.0	0.0	0.0	
139	0.0	0.0	0.0	0.0	
214	0.0	0.0	0.0	0.0	
219	0.0	0.0	0.0	0.0	
228	0.0	0.0	0.0	0.0	
240	0.0	0.0	0.0	0.0	
245	0.0	0.0	0.0	0.0	
313	0.0	0.0	0.0	0.0	
403	0.0	0.0	0.0	0.0	
406	0.0	0.0	0.0	0.0	
455	0.0	0.0	0.0	0.0	
456	0.0	0.0	0.0	0.0	
465	0.0	0.0	0.0	0.0	
471	0.0	0.0	0.0	0.0	
485	0.0	0.0	0.0	0.0	
570	0.0	0.0	0.0	0.0	
592	0.0	0.0	0.0	0.0	
644	0.0	0.0	0.0	0.0	
645	0.0	0.0	0.0	0.0	
731	0.0	0.0	0.0	0.0	
733	0.0	0.0	0.0	0.0	
747	0.0	0.0	0.0	0.0	
748	0.0	0.0	0.0	0.0	
767	0.0	0.0	0.0	0.0	
769	0.0	0.0	0.0	0.0	
770	0.0	0.0	0.0	0.0	
772	0.0	0.0	0.0	0.0	
773	0.0	0.0	0.0	0.0	
774	0.0	0.0	0.0	0.0	
775	0.0	0.0	0.0	0.0	
776	0.0	0.0	0.0	0.0	
777	0.0	0.0	0.0	0.0	
809	0.0	0.0	0.0	0.0	
810	0.0	0.0	0.0	0.0	
819	0.0	0.0	0.0	0.0	
833	0.0	0.0	0.0	0.0	
908	0.0	0.0	0.0	0.0	
914	0.0	0.0	0.0	0.0	
934	0.0	0.0	0.0	0.0	

962	0.0	0.0	0.0	0.0
986	0.0	0.0	0.0	0.0
988	0.0	0.0	0.0	0.0
1013	0.0	0.0	0.0	0.0
1019	0.0	0.0	0.0	0.0
1035	0.0	0.0	0.0	0.0
1053	0.0	0.0	0.0	0.0
1089	0.0	0.0	0.0	0.0
1107	0.0	0.0	0.0	0.0
1133	0.0	0.0	0.0	0.0
1147	0.0	0.0	0.0	0.0
1195	0.0	0.0	0.0	0.0
1222	0.0	0.0	0.0	0.0
1265	0.0	0.0	0.0	0.0
1305	0.0	0.0	0.0	0.0
1370	0.0	0.0	0.0	0.0
1385	0.0	0.0	0.0	0.0
1413	0.0	0.0	0.0	0.0
1429	0.0	0.0	0.0	0.0
1499	0.0	0.0	0.0	0.0
1549	0.0	0.0	0.0	0.0
1589	0.0	0.0	0.0	0.0
1596	0.0	0.0	0.0	0.0
1611	0.0	0.0	0.0	0.0
1695	0.0	0.0	0.0	0.0
1739	0.0	0.0	0.0	0.0
1745	0.0	0.0	0.0	0.0
1759	0.0	0.0	0.0	0.0
1770	0.0	0.0	0.0	0.0
1773	0.0	0.0	0.0	0.0
1804	0.0	0.0	0.0	0.0
1805	0.0	0.0	0.0	0.0
1822	0.0	0.0	0.0	0.0
1844	0.0	0.0	0.0	0.0
1862	0.0	0.0	0.0	0.0
1968	0.0	0.0	0.0	0.0
2038	0.0	0.0	0.0	0.0
2039	0.0	0.0	0.0	0.0
2136	0.0	0.0	0.0	0.0
2181	0.0	0.0	0.0	0.0
2227	0.0	0.0	0.0	0.0
2228	0.0	0.0	0.0	0.0
2230	0.0	0.0	0.0	0.0
2231	0.0	0.0	0.0	0.0
2232	0.0	0.0	0.0	0.0
2233	0.0	0.0	0.0	0.0
2235	0.0	0.0	0.0	0.0

2236	0.0	0.0	0.0	0.0
2298	0.0	0.0	0.0	0.0
2327	0.0	0.0	0.0	0.0
2365	0.0	0.0	0.0	0.0
2369	0.0	0.0	0.0	0.0
2392	0.0	0.0	0.0	0.0
2395	0.0	0.0	0.0	0.0
2480	0.0	0.0	0.0	0.0
2487	0.0	0.0	0.0	0.0
2505	0.0	0.0	0.0	0.0
2506	0.0	0.0	0.0	0.0
2518	0.0	0.0	0.0	0.0
2525	0.0	0.0	0.0	0.0
2566	0.0	0.0	0.0	0.0

	\$3,000 to \$3,499	\$3,500 or more	No cash rent	\
18	0.0	0.0	0.0	
41	0.0	0.0	0.0	
86	0.0	0.0	0.0	
119	0.0	0.0	0.0	
129	0.0	0.0	0.0	
139	0.0	0.0	0.0	
214	0.0	0.0	0.0	
219	0.0	0.0	0.0	
228	0.0	0.0	0.0	
240	0.0	0.0	0.0	
245	0.0	0.0	0.0	
313	0.0	0.0	0.0	
403	0.0	0.0	0.0	
406	0.0	0.0	0.0	
455	0.0	0.0	0.0	
456	0.0	0.0	0.0	
465	0.0	0.0	0.0	
471	0.0	0.0	0.0	
485	0.0	0.0	0.0	
570	0.0	0.0	0.0	
592	0.0	0.0	0.0	
644	0.0	0.0	0.0	
645	0.0	0.0	0.0	
731	0.0	0.0	0.0	
733	0.0	0.0	0.0	
747	0.0	0.0	0.0	
748	0.0	0.0	0.0	
767	0.0	0.0	0.0	
769	0.0	0.0	0.0	
770	0.0	0.0	0.0	
772	0.0	0.0	0.0	

773	0.0	0.0	0.0
774	0.0	0.0	0.0
775	0.0	0.0	0.0
776	0.0	0.0	0.0
777	0.0	0.0	0.0
809	0.0	0.0	0.0
810	0.0	0.0	0.0
819	0.0	0.0	0.0
833	0.0	0.0	0.0
908	0.0	0.0	0.0
914	0.0	0.0	0.0
934	0.0	0.0	0.0
962	0.0	0.0	0.0
986	0.0	0.0	0.0
988	0.0	0.0	0.0
1013	0.0	0.0	0.0
1019	0.0	0.0	0.0
1035	0.0	0.0	0.0
1053	0.0	0.0	0.0
1089	0.0	0.0	0.0
1107	0.0	0.0	0.0
1133	0.0	0.0	0.0
1147	0.0	0.0	0.0
1195	0.0	0.0	0.0
1222	0.0	0.0	0.0
1265	0.0	0.0	0.0
1305	0.0	0.0	0.0
1370	0.0	0.0	0.0
1385	0.0	0.0	0.0
1413	0.0	0.0	0.0
1429	0.0	0.0	0.0
1499	0.0	0.0	0.0
1549	0.0	0.0	0.0
1589	0.0	0.0	0.0
1596	0.0	0.0	0.0
1611	0.0	0.0	0.0
1695	0.0	0.0	0.0
1739	0.0	0.0	0.0
1745	0.0	0.0	0.0
1759	0.0	0.0	0.0
1770	0.0	0.0	0.0
1773	0.0	0.0	0.0
1804	0.0	0.0	0.0
1805	0.0	0.0	0.0
1822	0.0	0.0	0.0
1844	0.0	0.0	0.0
1862	0.0	0.0	0.0

1968	0.0	0.0	0.0
2038	0.0	0.0	0.0
2039	0.0	0.0	0.0
2136	0.0	0.0	0.0
2181	0.0	0.0	0.0
2227	0.0	0.0	0.0
2228	0.0	0.0	0.0
2230	0.0	0.0	0.0
2231	0.0	0.0	0.0
2232	0.0	0.0	0.0
2233	0.0	0.0	0.0
2235	0.0	0.0	0.0
2236	0.0	0.0	0.0
2298	0.0	0.0	0.0
2327	0.0	0.0	0.0
2365	0.0	0.0	0.0
2369	0.0	0.0	0.0
2392	0.0	0.0	0.0
2395	0.0	0.0	0.0
2480	0.0	0.0	0.0
2487	0.0	0.0	0.0
2505	0.0	0.0	0.0
2506	0.0	0.0	0.0
2518	0.0	0.0	0.0
2525	0.0	0.0	0.0
2566	0.0	0.0	0.0

geometry

18	MULTIPOLYGON	(((-67.18630 18.37942, -67.18615 ...
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86	MULTIPOLYGON	(((-67.22272 18.48982, -67.22236 ...
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934 MULTIPOLYGON (((-66.31456 18.47891, -66.31398 ...
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1305 MULTIPOLYGON (((-67.10989 17.94235, -67.10758 ...
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1385 MULTIPOLYGON (((-65.74858 18.38707, -65.74185 ...
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1589 MULTIPOLYGON (((-66.03826 18.00077, -66.03792 ...
1596 MULTIPOLYGON (((-66.01321 17.96702, -66.01106 ...

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1968 MULTIPOLYGON (((-66.05595 18.42848, -66.05575 ...
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2136 MULTIPOLYGON (((-66.07196 18.38648, -66.07196 ...
2181 MULTIPOLYGON (((-66.08581 18.38608, -66.08576 ...
2227 MULTIPOLYGON (((-66.07828 18.41638, -66.07704 ...
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2327 MULTIPOLYGON (((-66.24766 18.37678, -66.24744 ...
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2395 MULTIPOLYGON (((-66.02526 18.37983, -66.02519 ...
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2487 MULTIPOLYGON (((-66.41847 18.47759, -66.41818 ...
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2506 MULTIPOLYGON (((-66.40035 18.43625, -66.40021 ...
2518 MULTIPOLYGON (((-66.43862 18.49332, -66.43850 ...
2525 MULTIPOLYGON (((-65.63097 18.12302, -65.63095 ...
2566 MULTIPOLYGON (((-65.84775 18.00767, -65.84587 ...

```

```

[50]: # create a new variable for census tracts with zero pop
      gdf_no_rent = gdf[gdf['Total']==0]

```

```

[51]: # how many records?
      print('There are ' + str(len(gdf_no_rent)) + ' blockgroups with no rent-paying_
      ↪in them')

```

There are 104 blockgroups with no rent-paying in them

```
[52]: # display it
      gdf_no_rent[['geoid', 'Total']]
```

```
[52]:
```

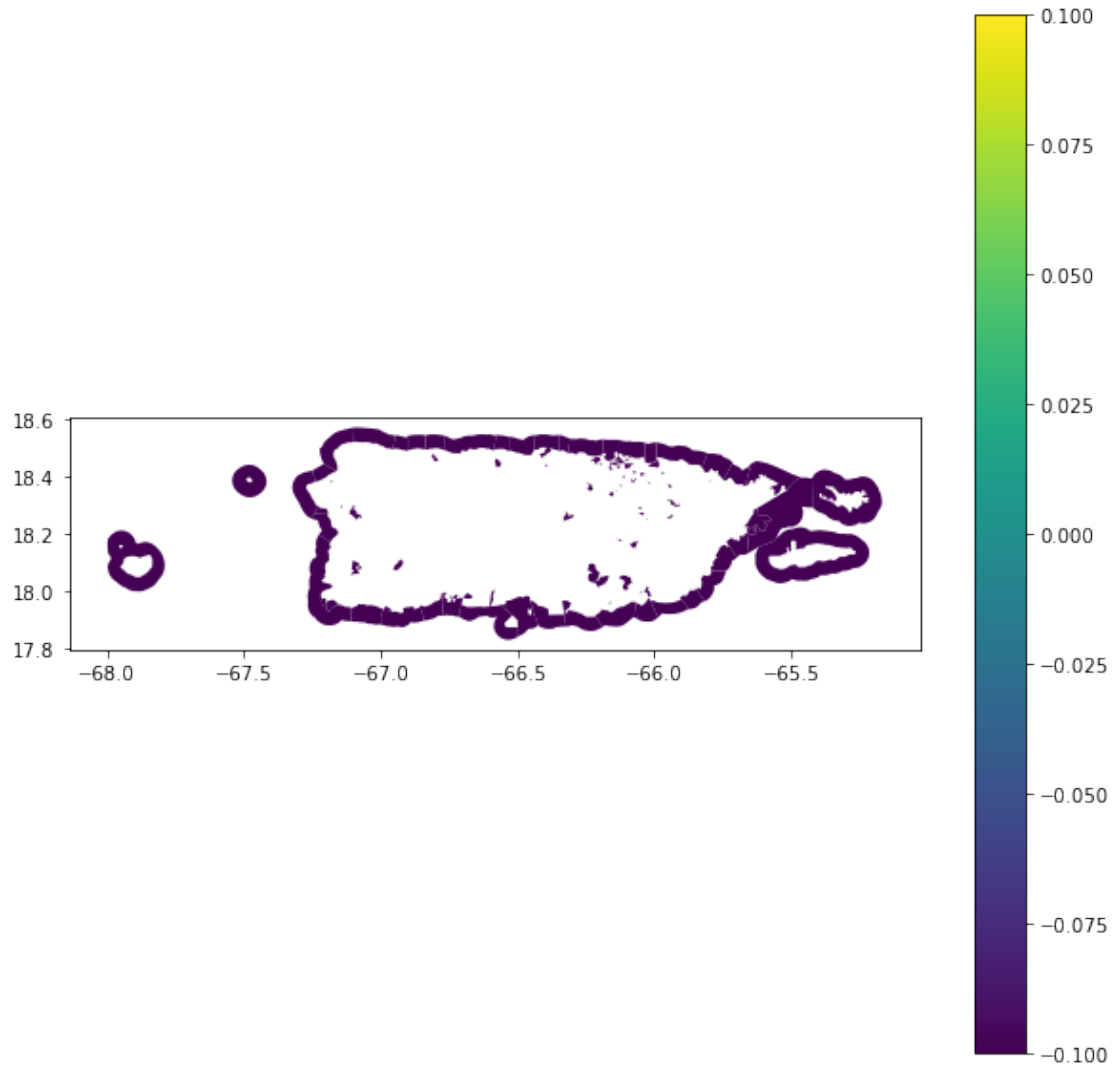
	geoid	Total
18	15000US720034302003	0.0
41	15000US720039900000	0.0
86	15000US720059900000	0.0
119	15000US720118101002	0.0
129	15000US720118105001	0.0
139	15000US720119900000	0.0
214	15000US720139929000	0.0
219	15000US720152801023	0.0
228	15000US720159915000	0.0
240	15000US720175903004	0.0
245	15000US720179900000	0.0
313	15000US720210310314	0.0
403	15000US720210317043	0.0
406	15000US720210318002	0.0
455	15000US720239800001	0.0
456	15000US720239900000	0.0
465	15000US720252002004	0.0
471	15000US720252003026	0.0
485	15000US720252005001	0.0
570	15000US720252028003	0.0
592	15000US720279900000	0.0
644	15000US720310502122	0.0
645	15000US720310502123	0.0
731	15000US720310509024	0.0
733	15000US720310510011	0.0
747	15000US720319800031	0.0
748	15000US720319900000	0.0
767	15000US720330205002	0.0
769	15000US720339800011	0.0
770	15000US720339800021	0.0
772	15000US720339800041	0.0
773	15000US720339800051	0.0
774	15000US720339800061	0.0
775	15000US720339800071	0.0
776	15000US720339800081	0.0
777	15000US720339902010	0.0
809	15000US720371601001	0.0
810	15000US720371601002	0.0
819	15000US720379900000	0.0
833	15000US720412401012	0.0
908	15000US720475306001	0.0
914	15000US720499905010	0.0
934	15000US720519900210	0.0

962	15000US720539901030	0.0
986	15000US720559900000	0.0
988	15000US720572701002	0.0
1013	15000US720572707002	0.0
1019	15000US720579926000	0.0
1035	15000US720599900010	0.0
1053	15000US720610403042	0.0
1089	15000US720610406026	0.0
1107	15000US720632102022	0.0
1133	15000US720653103003	0.0
1147	15000US720659900160	0.0
1195	15000US720699918000	0.0
1222	15000US720719900000	0.0
1265	15000US720759900010	0.0
1305	15000US720799911000	0.0
1370	15000US720879900000	0.0
1385	15000US720899900010	0.0
1413	15000US720919900250	0.0
1429	15000US720959900000	0.0
1499	15000US720979900000	0.0
1549	15000US721039900130	0.0
1589	15000US721092902004	0.0
1596	15000US721099900000	0.0
1611	15000US721119900000	0.0
1695	15000US721130722026	0.0
1739	15000US721130730081	0.0
1745	15000US721139930000	0.0
1759	15000US721159900000	0.0
1770	15000US721179904000	0.0
1773	15000US721191301013	0.0
1804	15000US721199927000	0.0
1805	15000US721219603001	0.0
1822	15000US721239526001	0.0
1844	15000US721239928000	0.0
1862	15000US721258406002	0.0
1968	15000US721270044002	0.0
2038	15000US721270058002	0.0
2039	15000US721270058003	0.0
2136	15000US721270086025	0.0
2181	15000US721270098002	0.0
2227	15000US721279801011	0.0
2228	15000US721279801021	0.0
2230	15000US721279801041	0.0
2231	15000US721279801051	0.0
2232	15000US721279801071	0.0
2233	15000US721279801081	0.0
2235	15000US721279803001	0.0

```
2236 15000US721279900000 0.0
2298 15000US721339900000 0.0
2327 15000US721355106022 0.0
2365 15000US721371214003 0.0
2369 15000US721371217023 0.0
2392 15000US721379900000 0.0
2395 15000US721390601023 0.0
2480 15000US721439900000 0.0
2487 15000US721455602042 0.0
2505 15000US721455607013 0.0
2506 15000US721455607014 0.0
2518 15000US721459900000 0.0
2525 15000US721479900000 0.0
2566 15000US721519900000 0.0
```

```
[53]: # Quiz! Plot the census tracts with zero rent-paying below (make it nice and
      ↪big)
      gdf_no_pop.plot(figsize=(10,10),
                      column='Total',
                      legend=True)
```

```
[53]: <AxesSubplot:>
```

```
[54]: # output columns  
list(gdf)
```

```
[54]: ['geoid',  
      'name',  
      'Total',  
      'With cash rent',  
      'Less than $100',  
      '$100 to $149',  
      '$150 to $199',  
      '200 to $249',  
      '$250 to $299',  
      '$300 to $349',  
      '$350 to $399',
```

```
'$400 to $449',
'$450 to $499',
'$500 to $549',
'$550 to $599',
'$600 to $649',
'$650 to $699',
'$700 to $749',
'$750 to $799',
'$800 to $899',
'$900 to $999',
'$1,000 to $1,249',
'$1,250 to $1,499',
'$1,500 to $1,999',
'$2,000 to $2,499',
'$2,500 to $2,999',
'$3,000 to $3,499',
'$3,500 or more',
'No cash rent',
'geometry']
```

```
[58]: gdf.head()
```

```
[58]:
```

	geoid	name	Total	With cash rent	\
1	15000US720019563001	Block Group 1, Adjuntas, PR	167.0	89.0	
2	15000US720019563002	Block Group 2, Adjuntas, PR	223.0	94.0	
3	15000US720019564001	Block Group 1, Adjuntas, PR	239.0	149.0	
4	15000US720019564002	Block Group 2, Adjuntas, PR	154.0	16.0	
5	15000US720019565001	Block Group 1, Adjuntas, PR	198.0	88.0	

	Less than \$100	\$100 to \$149	\$150 to \$199	200 to \$249	\$250 to \$299	\
1	0.0	0.0	0.000000	0.0	0.0	
2	0.0	0.0	5.381166	0.0	0.0	
3	0.0	0.0	0.000000	0.0	0.0	
4	0.0	0.0	0.000000	0.0	0.0	
5	0.0	0.0	0.000000	0.0	28.0	

	\$300 to \$349	\$350 to \$399	\$400 to \$449	\$450 to \$499	\$500 to \$549	\
1	47.0	10.0	0.0	17.0	0.0	
2	21.0	15.0	18.0	16.0	0.0	
3	14.0	27.0	19.0	33.0	13.0	
4	0.0	16.0	0.0	0.0	0.0	
5	29.0	0.0	15.0	0.0	16.0	

	\$550 to \$599	\$600 to \$649	\$650 to \$699	\$700 to \$749	\$750 to \$799	\
1	15.0	0.0	0.0	0.0	0.0	
2	0.0	12.0	0.0	0.0	0.0	
3	16.0	27.0	0.0	0.0	0.0	

4	0.0	0.0	0.0	0.0	0.0
5	0.0	0.0	0.0	0.0	0.0

	\$800 to \$899	\$900 to \$999	\$1,000 to \$1,249	\$1,250 to \$1,499	\
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	

	\$1,500 to \$1,999	\$2,000 to \$2,499	\$2,500 to \$2,999	\$3,000 to \$3,499	\
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	

	\$3,500 or more	No cash rent	\
1	0.0	78.0	
2	0.0	129.0	
3	0.0	90.0	
4	0.0	138.0	
5	0.0	110.0	

```

                                geometry
1 MULTIPOLYGON (((-66.75808 18.17672, -66.75776 ...
2 MULTIPOLYGON (((-66.79443 18.24988, -66.79381 ...
3 MULTIPOLYGON (((-66.83326 18.22802, -66.83275 ...
4 MULTIPOLYGON (((-66.83723 18.19298, -66.83685 ...
5 MULTIPOLYGON (((-66.77916 18.15809, -66.77876 ...

```

Okay, I'm going to try to do a map now...

```

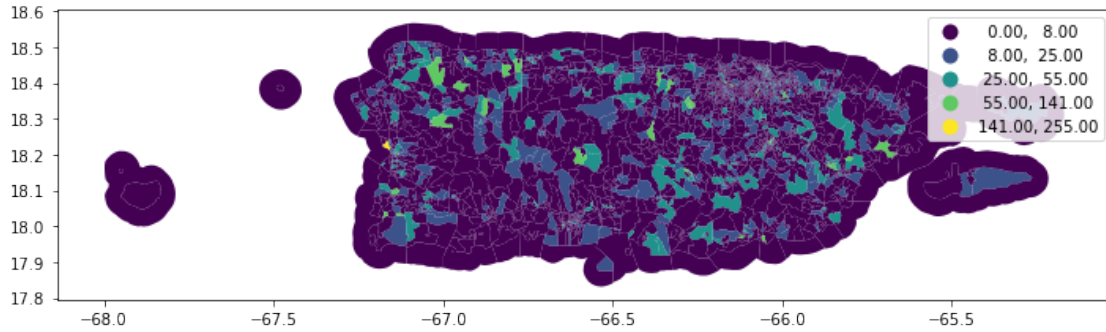
[60]: # natural breaks
      gdf.plot(figsize=(12,12),
              column='$450 to $499',
              legend=True,
              scheme='NaturalBreaks')

```

```

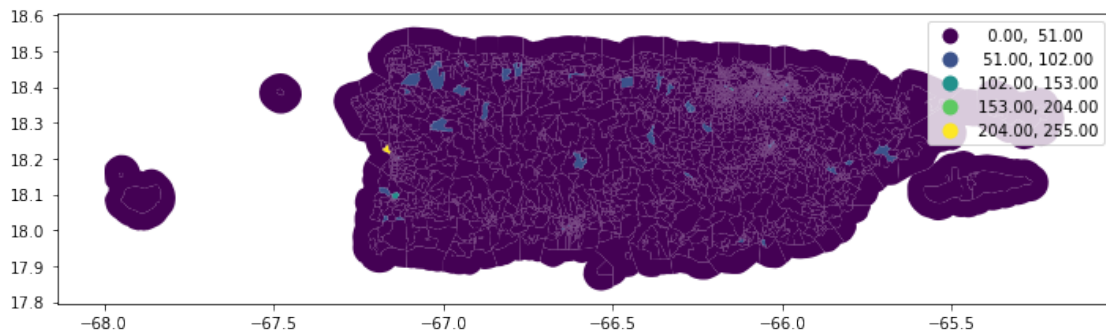
[60]: <AxesSubplot:>

```



```
[61]: # equal interval
gdf.plot(figsize=(12,12),
         column='$450 to $499',
         legend=True,
         scheme='equal_interval')
```

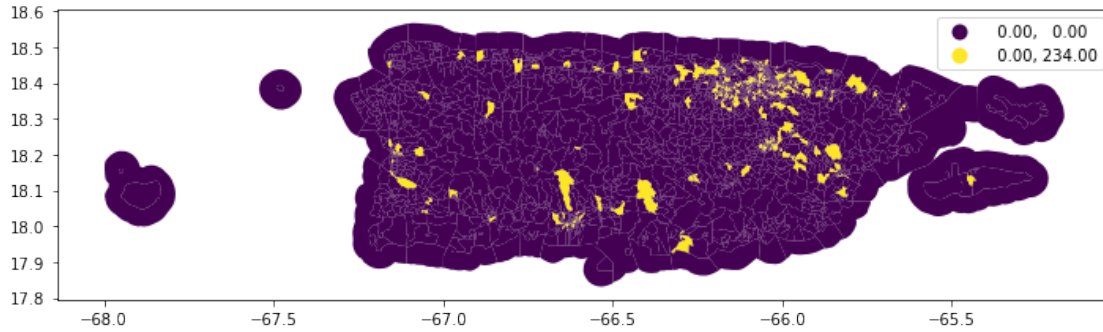
[61]: <AxesSubplot:>



```
[62]: # quantiles
gdf.plot(figsize=(12,12),
         column='$900 to $999',
         legend=True,
         scheme='quantiles')
```

```
/opt/conda/lib/python3.9/site-packages/mapclassify/classifiers.py:238:
UserWarning: Warning: Not enough unique values in array to form k classes
  Warn(
/opt/conda/lib/python3.9/site-packages/mapclassify/classifiers.py:241:
UserWarning: Warning: setting k to 2
  Warn("Warning: setting k to %d" % k_q, UserWarning)
```

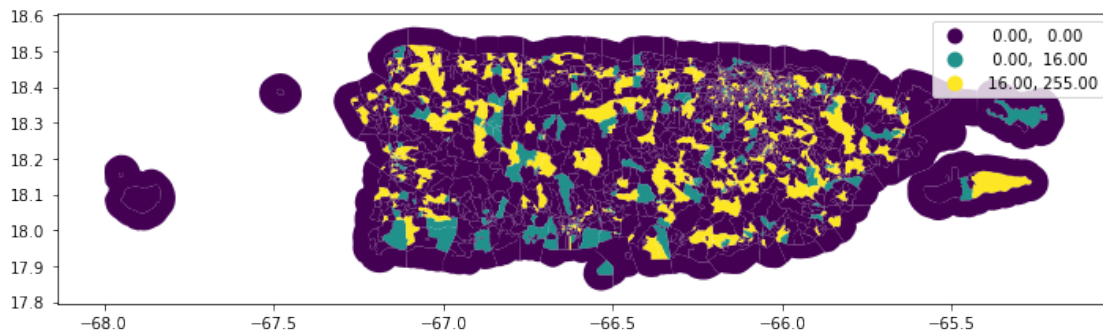
[62]: <AxesSubplot:>



```
[63]: # quantiles
gdf.plot(figsize=(12,12),
         column='$450 to $499',
         legend=True,
         scheme='quantiles')
```

```
/opt/conda/lib/python3.9/site-packages/mapclassify/classifiers.py:238:
UserWarning: Warning: Not enough unique values in array to form k classes
Warn(
/opt/conda/lib/python3.9/site-packages/mapclassify/classifiers.py:241:
UserWarning: Warning: setting k to 3
Warn("Warning: setting k to %d" % k_q, UserWarning)
```

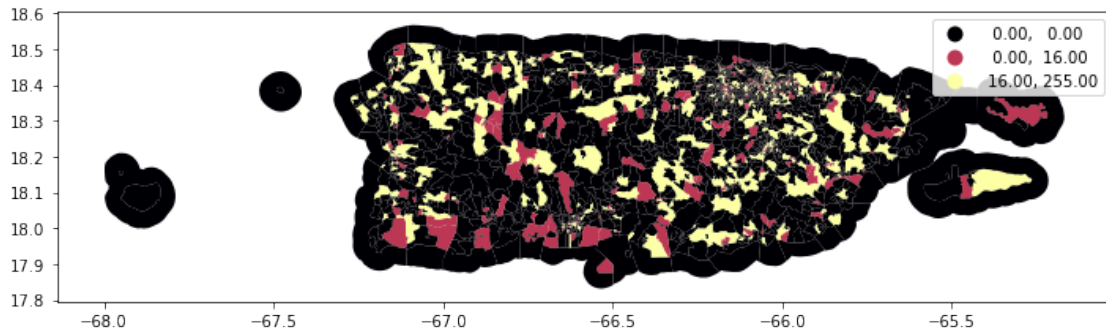
[63]: <AxesSubplot:>



```
[64]: # quantiles with different color scheme
gdf.plot(figsize=(12,12),
         column='$450 to $499',
         legend=True,
         scheme='quantiles',
         cmap='inferno'
    )
```

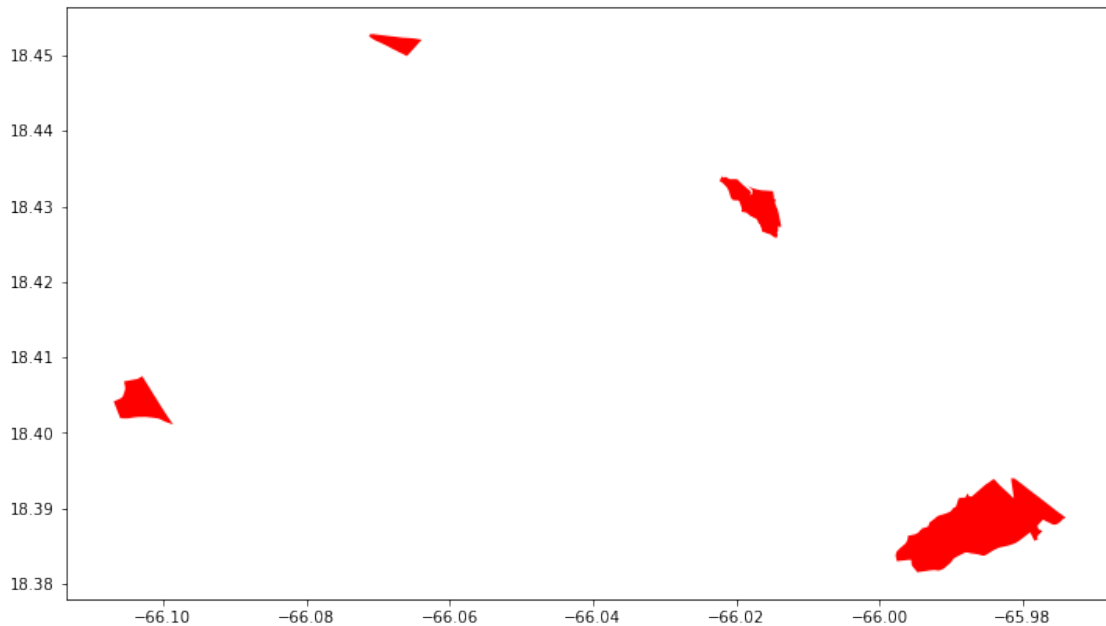
```
/opt/conda/lib/python3.9/site-packages/mapclassify/classifiers.py:238:  
UserWarning: Warning: Not enough unique values in array to form k classes  
Warn(  
/opt/conda/lib/python3.9/site-packages/mapclassify/classifiers.py:241:  
UserWarning: Warning: setting k to 3  
Warn("Warning: setting k to %d" % k_q, UserWarning)
```

[64]: <AxesSubplot:>



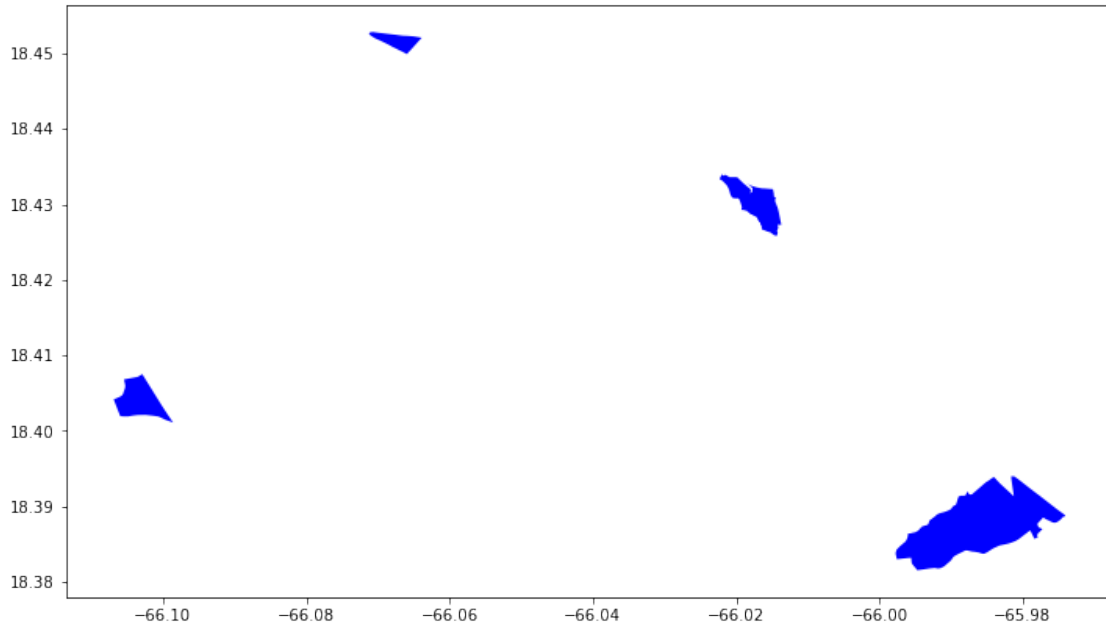
```
[65]: # plot it in red  
gdf[gdf['$900 to $999'] > 90].plot(figsize=(12,10),color="red")
```

[65]: <AxesSubplot:>



```
[69]: # plot it in red
gdf[gdf['$900 to $999'] > 90].plot(figsize=(12,10),color="blue")
```

```
[69]: <AxesSubplot:>
```



```
[70]: #Okay, I will add a basemap now
```

```
[71]: # reproject to Web Mercator
gdf_web_mercator = gdf.to_crs(epsg=3857)
```

```
[75]: # use subplots that make it easier to create multiple layered maps
fig, ax = plt.subplots(figsize=(15, 15))

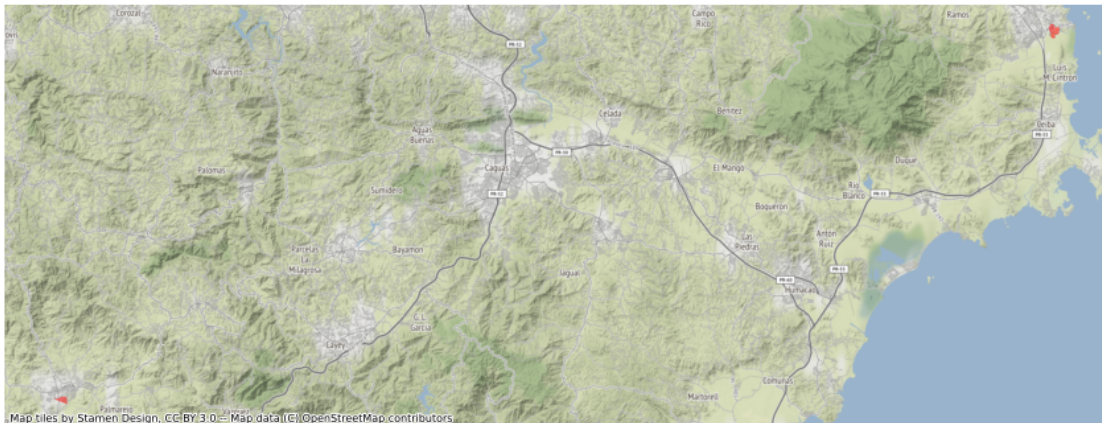
# add the layer with ax=ax in the argument
gdf_web_mercator[gdf_web_mercator['$100 to $149'] > 90].plot(
    ax=ax,
    color="red",
    alpha=0.5
)

# turn the axis off
ax.axis('off')

# set a title
ax.set_title('Block Groups with more than 90% Houses that Pay $100 to_
↪$149',fontsize=24,pad=20)
```

```
# add a basemap
ctx.add_basemap(ax)
```

Block Groups with more than 90% Houses that Pay 100to149



```
[76]: # use subplots that make it easier to create multiple layered maps
fig, ax = plt.subplots(figsize=(15, 15))

# add the layer with ax=ax in the argument
gdf_web_mercator[gdf_web_mercator['$450 to $499'] > 90].plot(
    ax=ax,
    color="red",
    alpha=0.5
)

# turn the axis off
ax.axis('off')

# set a title
ax.set_title('Block Groups with more than 90% Houses that Pay $450 to $499 in ↵
↳Gross Rent',fontsize=24,pad=20)

# add a basemap
ctx.add_basemap(ax)
```


Block Groups with more than 90% Houses that Pay 450to499 in Gross Rent



It was sooo good finally doing some maps! I think my next step will be to run this with only San Juan, perhaps looking at different variables for rent.

[]: