01_{viz}

February 4, 2019

```
In [1]: from IPython.display import Image, IFrame
```

1 Manipulating & Visualising spatial data

```
In [2]: %matplotlib inline
    import geopandas as gpd
    import contextily as ctx
    import matplotlib.pyplot as plt
    from qgrid import show_grid
```

1.1 Sources

1.1.1 "Arturo"

```
In [3]: Image('../figs/arturo.png', width=750)
Out[3]:
```



1.1.2 Inside AirBnb

```
In [4]: #Image('../figs/insideabb.png', width=750)
```

1.2 Data

```
In [5]: db = gpd.read_file('../data/demo_data.gpkg')
In [6]: db.info()
<class 'geopandas.geodataframe.GeoDataFrame'>
RangeIndex: 128 entries, 0 to 127
Data columns (total 15 columns):
neighbourhood
                              128 non-null object
neighbourhood_group
                              128 non-null object
land_use_mix
                              128 non-null float64
closeness_small_parks
                              128 non-null float64
residence_ratio
                              128 non-null float64
                              128 non-null float64
age_diversity
                              128 non-null float64
population_density
arturo_score
                              128 non-null float64
                              126 non-null float64
abb_review_scores_location
                              127 non-null float64
abb_price_usd
N. Abb
                              127 non-null float64
```

N. streets 128 non-null int64
Scored streets 128 non-null int64
pct_scored 128 non-null float64
geometry 128 non-null object

dtypes: float64(10), int64(2), object(3)

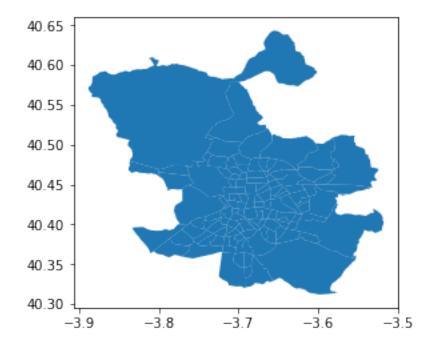
memory usage: 15.1+ KB

In [7]: show_grid(db)

 $\label{lem:qgridWidget} QgridWidget(grid_options = \{ \ 'fullWidthRows': \ True, \ 'syncColumnCellResize': \ True, \ 'forceFitColumns' \ True, \ 'syncColumnCellResize': \ True, \ 'syncColumnCellResiz$

In [8]: db.plot()

Out[8]: <matplotlib.axes._subplots.AxesSubplot at 0x7f8c6db43978>



1.3 Queries

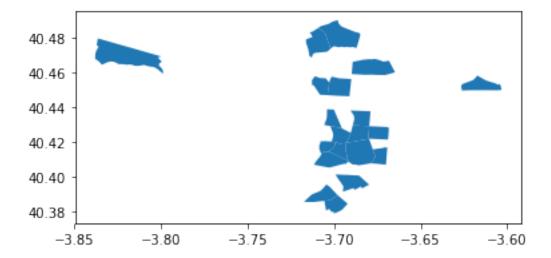
• Atomic

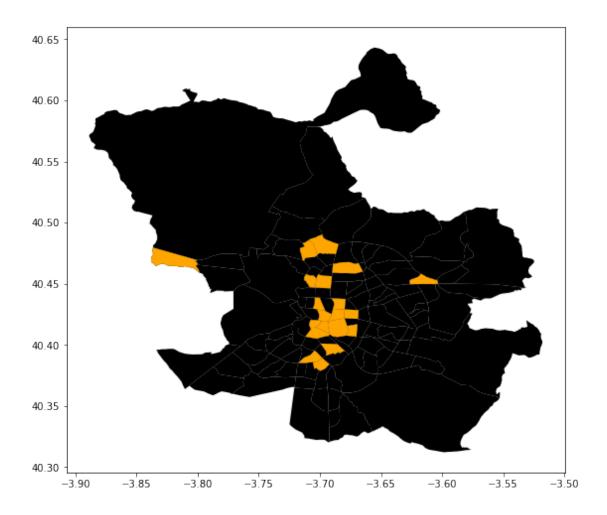
In [9]: db.set_index('neighbourhood').loc['Atocha', 'arturo_score']

Out[9]: 25.037677469271745

• Value-based

```
In [10]: db.query("neighbourhood_group == 'Centro'")
Out[10]:
           neighbourhood neighbourhood_group land_use_mix closeness_small_parks
                                                   0.000365
         0
                 Palacio
                                      Centro
                                                                         79.556113
         1
             Embajadores
                                      Centro
                                                   0.000549
                                                                        191.088707
         2
                  Cortes
                                      Centro
                                                                        184.110366
                                                   0.000346
         3
                Justicia
                                      Centro
                                                   0.000440
                                                                        193.695258
         4
             Universidad
                                      Centro
                                                   0.000481
                                                                        156.970200
         5
                                                                        122.989820
                     Sol
                                      Centro
                                                   0.000615
            residence_ratio
                             age_diversity population_density arturo_score
         0
                   0.708910
                                  0.002835
                                                    5343.068770
                                                                    28.029880
                                                   11271.551466
         1
                   0.671200
                                  0.002734
                                                                    28.549250
         2
                                  0.002756
                                                    4536.868443
                                                                    29.068393
                   0.475500
         3
                                  0.002774
                                                    6030.387988
                                                                    27.994642
                   0.691089
         4
                                                                    27.944427
                   0.738066
                                  0.002778
                                                   10863.626270
         5
                   0.583537
                                  0.002930
                                                    5061.064928
                                                                    29.524770
            abb_review_scores_location abb_price_usd N. Abb N. streets \
         0
                              9.878536
                                             94.704802 1416.0
                                                                       604
                              9.675348
                                                                       421
         1
                                           117.799054 2538.0
         2
                                                                       271
                              9.911258
                                           102.711364
                                                        880.0
         3
                              9.929070
                                           140.807287 1043.0
                                                                       301
         4
                              9.883041
                                           82.927123 1825.0
                                                                       436
         5
                              9.923518
                                           101.430099 1216.0
                                                                       272
            Scored streets pct_scored
         0
                       152
                             25.165563
                             31.828979
         1
                       134
         2
                             36.900369
                       100
         3
                        97
                             32.225914
         4
                       168
                             38.532110
         5
                        84
                             30.882353
                                                      geometry
           (POLYGON ((-3.70584 40.420297, -3.706248 40.42...
         1 (POLYGON ((-3.703837 40.414318, -3.70277 40.41...
         2 (POLYGON ((-3.697959 40.419286, -3.696453 40.4...
         3 (POLYGON ((-3.695463 40.418977, -3.696453 40.4...
         4 (POLYGON ((-3.701075 40.421339, -3.701547 40.4...
         5 (POLYGON ((-3.704747 40.42027, -3.701501 40.42...
   • Condition-based
In [11]: caros = db.query("abb_price_usd > 100")
         caros.plot()
Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x7f8c6d800f60>
```





In [13]: db.abb_price_usd.describe()

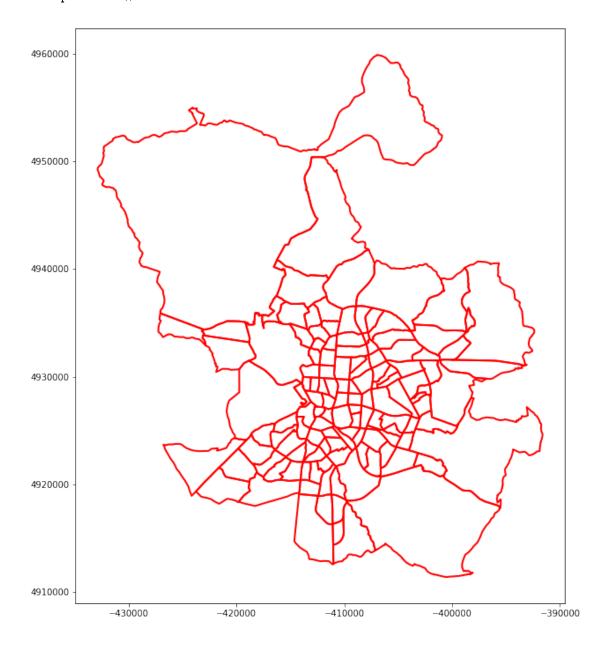
```
Out[13]: count 127.000000
mean 69.256524
std 42.574609
min 20.638889
25% 42.390351
50% 60.315789
75% 79.847826
max 319.400000
```

Name: abb_price_usd, dtype: float64

CHALLENGE

- Find the neighbourhoods where the average AirBnb price is greater than 150
- Make a map highlighting the neighbourhoods in the Hortaleza neighbourhood group
- Find the most attractive neighborhood based on the Arturo score

1.4 Basemaps (and projections)



CHALLENGE

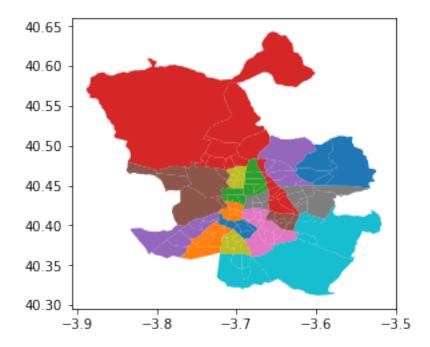
• Make a map of the Madrid neighbourhoods using the Terrain (ST_Terrain) basemap

1.5 Choropleths

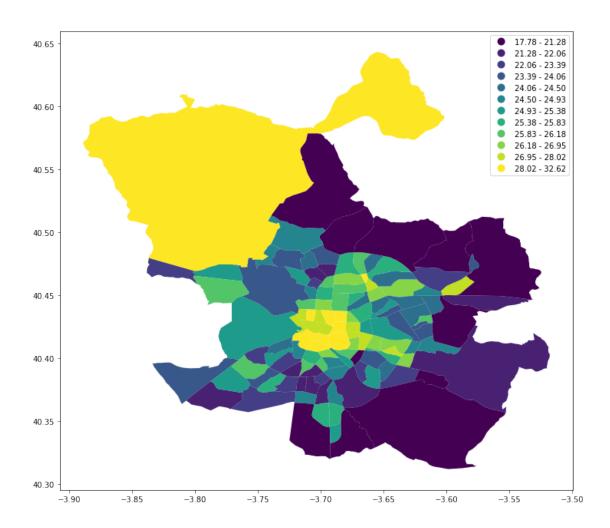
1.5.1 Categorical

In [18]: db.plot(column='neighbourhood_group')

Out[18]: <matplotlib.axes._subplots.AxesSubplot at 0x7f8c6d65bcc0>



1.5.2 Continuous



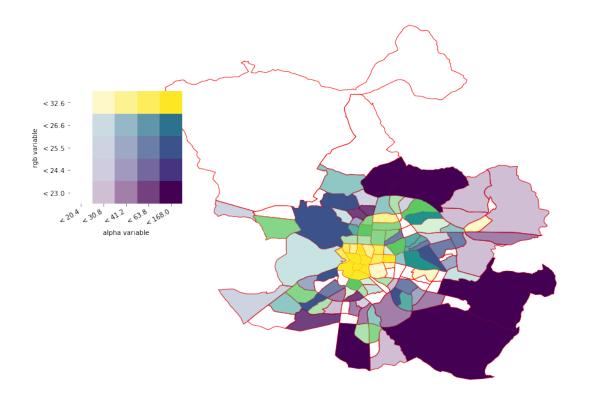
CHALLENGE

- Make a quantile choropleth of abb_review_scores_location
- Make a quantile choropleth of abb_price_usd
- Compare the three maps above. Do they display similar patterns?

1.5.3 Value by alpha

```
In [20]: from pysal.viz.splot.mapping import vba_choropleth
```

/opt/conda/lib/python3.6/site-packages/pysal/model/spvcm/abstracts.py:10: UserWarning: The `dill
from .sqlite import head_to_sql, start_sql



CHALLENGE - Make a similar VBA map for average AirBnb price (abb_price_usd) using the number of AirBnb properties (N. Abb) to tweak transparency