

# 02\_explore

February 4, 2019

```
In [1]: from IPython.display import HTML
```

## 1 *Explore* spatial data

```
In [2]: %matplotlib inline
```

```
import pandas as pd
import geopandas as gpd
from pysal.lib import weights
from pysal.explore import esda
from pysal.viz.splot.esda import plot_moran, plot_local_autocorrelation, lisa_cluster
import contextily as ctx
import matplotlib.pyplot as plt
```

```
db = gpd.read_file('../data/demo_data.gpkg')
```

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

### 1.1 Spatial weights

#### 1.1.1 Lattice example

```
In [3]: import numpy as np
        from shapely.geometry import Polygon

        # do a regular 3x3 lattice and draw it here
        w = weights.lat2W(3, 3, rook=True)
        # Get points in a grid
        l = np.arange(3)
        xs, ys = np.meshgrid(l, l)
        # Set up store
        polys = []
        # Generate polygons
        for x, y in zip(xs.flatten(), ys.flatten()):
            poly = Polygon([(x, y), (x+1, y), (x+1, y+1), (x, y+1)])
```

```

    polys.append(poly)
# Convert to GeoSeries
polys = gpd.GeoSeries(polys)
gdf = gpd.GeoDataFrame({'geometry': polys,
                        'id': ['P-%s'%str(i).zfill(2) for i in range(len(polys))]])
w.remap_ids(gdf['id'].values)
# Annotate & Visualise
ax = polys.plot(edgecolor='k', facecolor='w')
[plt.text(x, y, t,
          verticalalignment='center',
          horizontalalignment='center') for x, y, t in zip(
    [p.centroid.x for p in polys],
    [p.centroid.y for p in polys],
    [i for i in gdf['id']])]
ax.set_axis_off()

```

P-06	P-07	P-08
P-03	P-04	P-05
P-00	P-01	P-02

- Full matrix

```

In [4]: pd.DataFrame(w.full()[0],
                    index=gdf['id'],
                    columns=gdf['id'],
                    ).astype(int)

```

```

Out[4]: id    P-00 P-01 P-02 P-03 P-04 P-05 P-06 P-07 P-08
id
P-00      0     1     0     1     0     0     0     0     0

```

P-01	1	0	1	0	1	0	0	0	0
P-02	0	1	0	0	0	1	0	0	0
P-03	1	0	0	0	1	0	1	0	0
P-04	0	1	0	1	0	1	0	1	0
P-05	0	0	1	0	1	0	0	0	1
P-06	0	0	0	1	0	0	0	1	0
P-07	0	0	0	0	1	0	1	0	1
P-08	0	0	0	0	0	1	0	1	0

### 1.1.2 Real-word data

```
In [5]: w_queen = weights.Queen.from_dataframe(db)
```

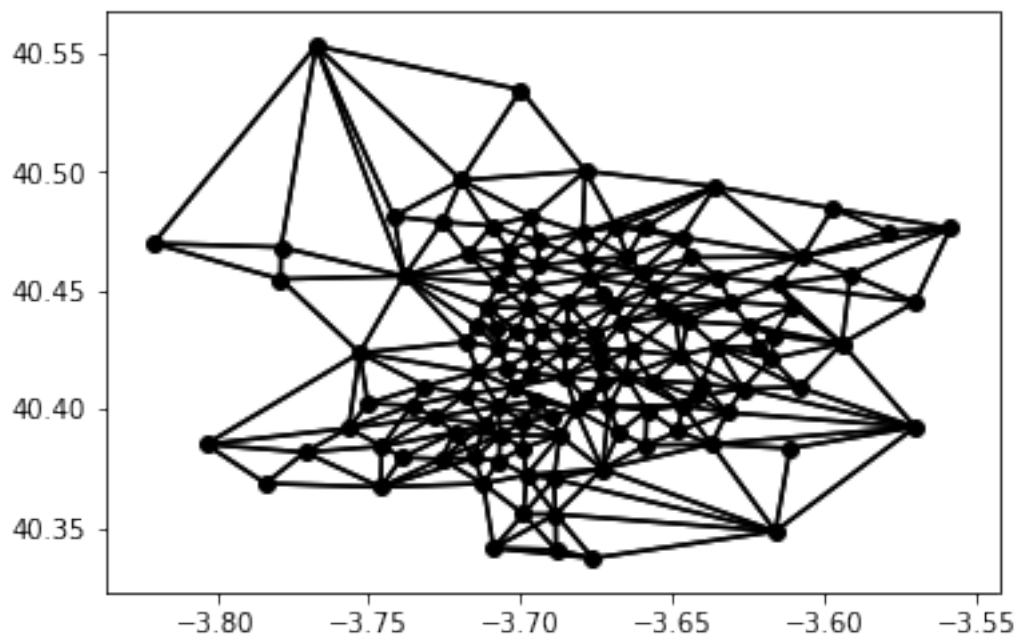
```
In [6]: w_k1 = weights.KNN.from_dataframe(db, k=1)
```

```
/opt/conda/lib/python3.6/site-packages/pysal/lib/weights/weights.py:170: UserWarning: The weights matrix is not fully connected. There are %d components" % self.n_components
warnings.warn("The weights matrix is not fully connected. There are %d components" % self.n_components)
```

- Visualising weights

```
In [7]: w_queen.plot(db)
```

```
Out[7]: (<Figure size 432x288 with 1 Axes>,
<matplotlib.axes._subplots.AxesSubplot at 0x7fa3146b01d0>)
```

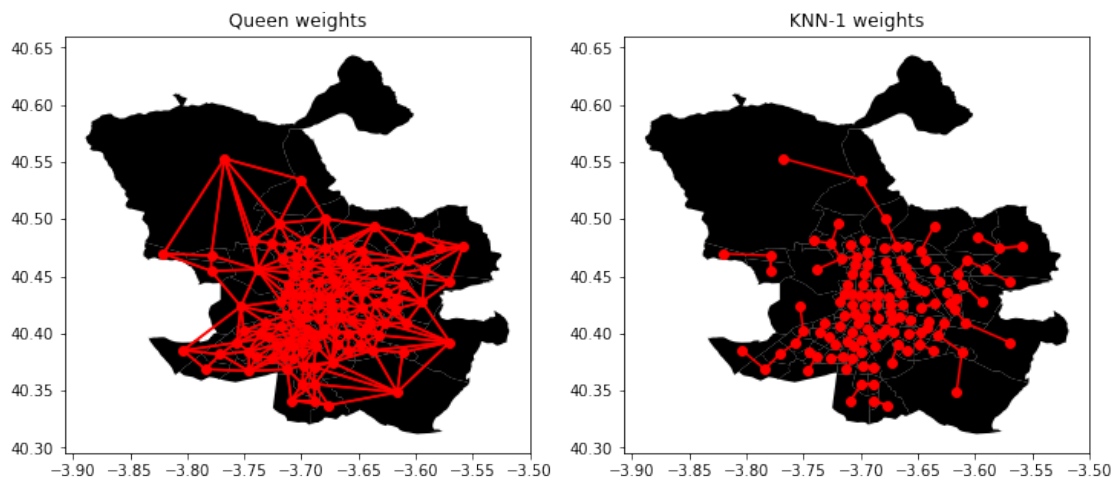


```
In [8]: f, axs = plt.subplots(1, 2, figsize=(12, 6))
```

```
db.plot(color='k', ax=axs[0])
w_queen.plot(db, ax=axs[0], color='red')
axs[0].set_title('Queen weights')
```

```
db.plot(color='k', ax=axs[1])
w_k1.plot(db, ax=axs[1], color='red')
axs[1].set_title('KNN-1 weights')
```

```
plt.show()
```



- And fancier graphs are also possible...

```
In [9]: blob = """
```

```
<blockquote class="twitter-tweet" data-lang="en">
```

```
<p lang="en" dir="ltr">
```

```
A few more mesmerizing bundled graphs from Spatial Weights Matrices and the
<a href="https://twitter.com/hashtag/PySAL?src=hash&ref_src=twsrc%5Etfw">
#PySAL</a>/<a href="https://twitter.com/datashader?ref_src=twsrc%5Etfw">@data
code I used. Tbh, I&#39;m not sure bundles are more useful at identifying the
spatial structure than traditional graphs, what do people think?
```

```
<a href="https://t.co/UkaPZlvUqL">https://t.co/UkaPZlvUqL</a>
```

```
<a href="https://t.co/3QW7z0IaCB">pic.twitter.com/3QW7z0IaCB</a></p>&mdash;
Dani Arribas-Bel (@darribas)
```

```
<a href="https://twitter.com/darribas/status/1056293955422294016?ref_src=twsrc%5Etfw">
October 27, 2018</a></blockquote>
```

```
<script async src="https://platform.twitter.com/widgets.js"
charset="utf-8"></script>
```

```
"""
```

```
HTML(blob)
```

```
Out[9]: <IPython.core.display.HTML object>
```

**CHALLENGE** - Create a spatial weights matrix using the KNN algorithm and picking the average number of neighbors under the queen criterium (hint: checkout the `mean_neighbors` attribute in `w_queen`)

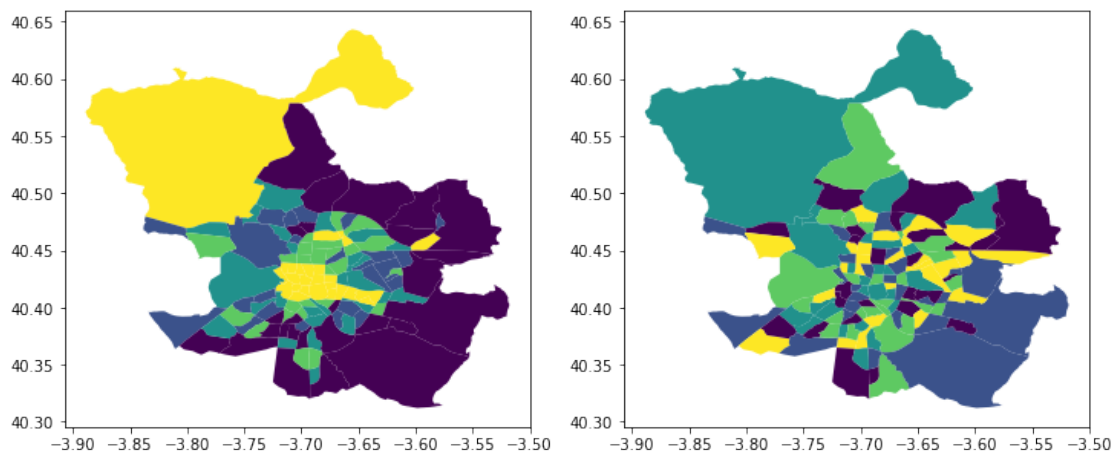
## 1.2 Global Spatial autocorrelation

- Shuffle values across space

```
In [10]: np.random.seed(1234)
         db['arturo_score_shuffled'] = db['arturo_score'].sample(frac=1).values
```

- Same values, different geographical implications...

```
In [11]: f, axs = plt.subplots(1, 2, figsize=(12, 6))
         db.plot(column='arturo_score', scheme='quantiles', ax=axs[0])
         db.plot(column='arturo_score_shuffled', scheme='quantiles', ax=axs[1])
         plt.show()
```



- Moran's I

```
In [12]: moran = esda.Moran(db['arturo_score'], w_queen)
```

```
In [13]: moran.I
```

```
Out[13]: 0.43867621491355296
```

```
In [14]: moran.p_sim
```

```
Out[14]: 0.001
```

```
In [15]: moran_shuffled = esda.Moran(db['arturo_score_shuffled'], w_queen)
```

```
In [16]: moran_shuffled.I
```

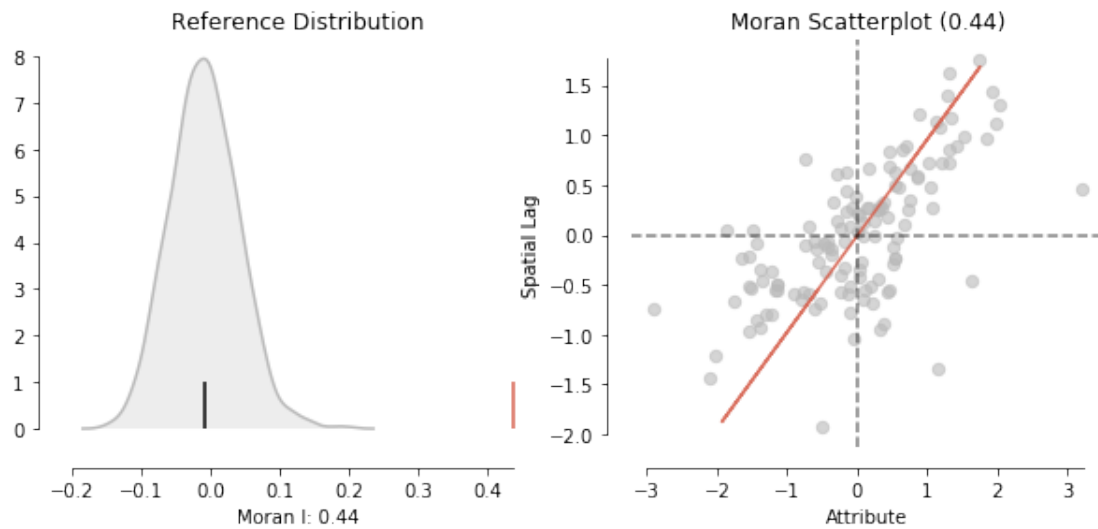
```
Out[16]: -0.06423418028115105
```

```
In [17]: moran_shuffled.p_sim
```

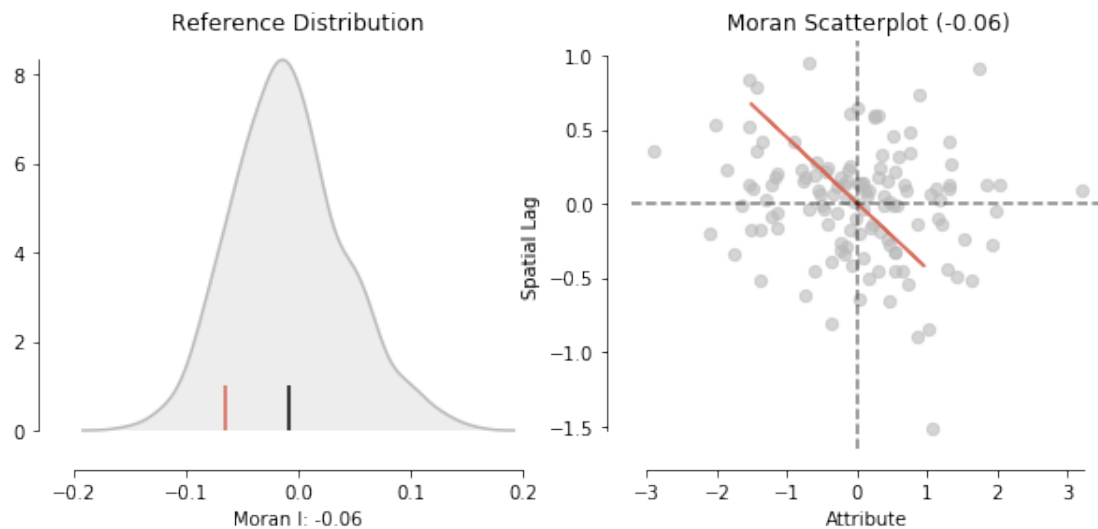
```
Out[17]: 0.124
```

- Moran Scatterplots

```
In [18]: plot_moran(moran);
```



```
In [19]: plot_moran(moran_shuffled);
```



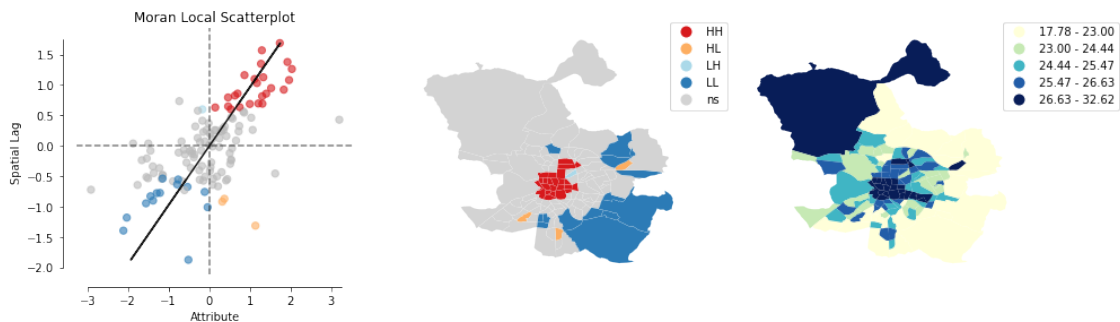
**CHALLENGE** - Compute Moran's I for the average price in AirBnb (abb\_price\_usd) and explore its scatterplot

### 1.3 Local Spatial autocorrelation

```
In [20]: lisa = esda.Moran_Local(db['arturo_score'], w_queen)
```

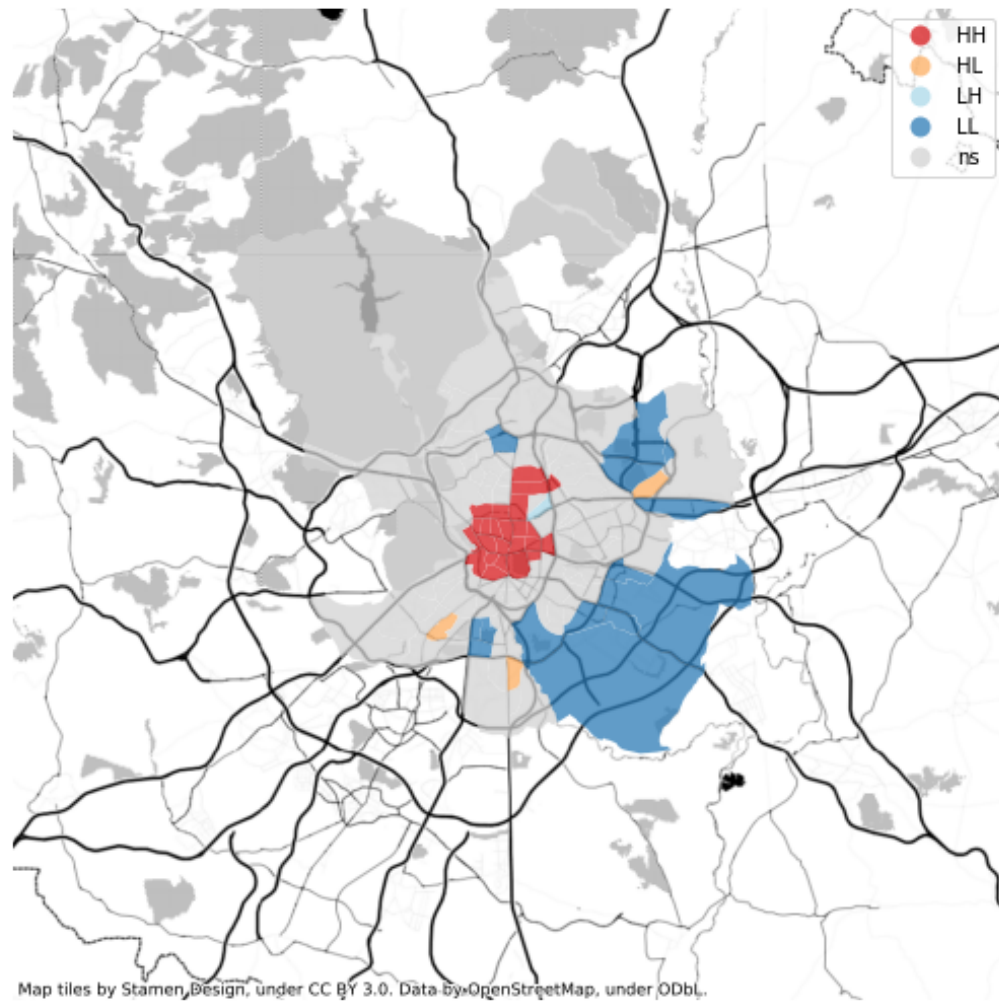
- Overall plot

```
In [21]: plot_local_autocorrelation(lisa, db, 'arturo_score');
```



- Cluster map

```
In [22]: f, ax = plt.subplots(1, figsize=(9, 9))
         lisa_cluster(lisa, db.to_crs(epsg=3857), alpha=0.75, ax=ax)
         ctx.add_basemap(ax, url=ctx.sources.ST_TONER_BACKGROUND)
         plt.show()
```



**CHALLENGE** - Create a similar LISA map for population density (population\_density)