

Proximity Analysis

Data analysis with GeoPandas

Kunal Khurana

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Techniques

measure distance between points on map

select all points within same radius

```
import folium
from folium import Marker, GeoJson
from folium.plugins import HeatMap
from shapely.geometry import MultiPolygon
```

```
import math
import pandas as pd
import geopandas as gpd
```

```
releases = gpd.read_file('data_for_all_courses\\toxic_release_pennsylvania\\toxic_release_
releases.head()
```

	YEAR	CITY	COUNTY	ST	LATITUDE	LONGITUDE	CHEMICAL
0	2016	PHILADELPHIA	PHILADELPHIA	PA	40.005901	-75.072103	FORMIC ACID
1	2016	PHILADELPHIA	PHILADELPHIA	PA	39.920120	-75.146410	ETHYLENE GLYCO
2	2016	PHILADELPHIA	PHILADELPHIA	PA	40.023880	-75.220450	CERTAIN GLYCOL
3	2016	PHILADELPHIA	PHILADELPHIA	PA	39.913540	-75.198890	LEAD COMPOUND
4	2016	PHILADELPHIA	PHILADELPHIA	PA	39.913540	-75.198890	BENZENE

```
releases.info()
```

```

<class 'geopandas.geodataframe.GeoDataFrame'>
RangeIndex: 4663 entries, 0 to 4662
Data columns (total 10 columns):
#   Column          Non-Null Count  Dtype
---  -
0   YEAR            4663 non-null   object
1   CITY            4663 non-null   object
2   COUNTY          4663 non-null   object
3   ST              4663 non-null   object
4   LATITUDE        4663 non-null   float64
5   LONGITUDE       4663 non-null   float64
6   CHEMICAL        4663 non-null   object
7   UNIT_OF_ME      4663 non-null   object
8   TOTAL_RELE      4663 non-null   float64
9   geometry        4663 non-null   geometry
dtypes: float64(3), geometry(1), object(6)
memory usage: 364.4+ KB

```

```

# air quality data
stations = gpd.read_file('data_for_all_courses\\toxic_release_pennsylvania.shx')
stations.head()

```

	geometry
0	POINT (2718560.227 256380.179)
1	POINT (2698674.606 224522.905)
2	POINT (2676833.394 261701.856)
3	POINT (2684030.004 221697.388)
4	POINT (2684030.004 221697.388)

```
stations.info()
```

```

<class 'geopandas.geodataframe.GeoDataFrame'>
RangeIndex: 4663 entries, 0 to 4662
Data columns (total 1 columns):
#   Column          Non-Null Count  Dtype
---  -
0   geometry        4663 non-null   geometry
dtypes: geometry(1)
memory usage: 36.6 KB

```

```
stations2 = gpd.read_file('data_for_all_courses\\toxic_release_pennsylvania.shp')
stations2.head()
```

	geometry
0	POINT (2718560.227 256380.179)
1	POINT (2698674.606 224522.905)
2	POINT (2676833.394 261701.856)
3	POINT (2684030.004 221697.388)
4	POINT (2684030.004 221697.388)

```
stations3 = gpd.read_file('data_for_all_courses\\toxic_release_pennsylvania.dbf')
stations3.head()
```

	YEAR	CITY	COUNTY	ST	LATITUDE	LONGITUDE	CHEMICAL
0	2016	PHILADELPHIA	PHILADELPHIA	PA	40.005901	-75.072103	FORMIC ACID
1	2016	PHILADELPHIA	PHILADELPHIA	PA	39.920120	-75.146410	ETHYLENE GLYCO
2	2016	PHILADELPHIA	PHILADELPHIA	PA	40.023880	-75.220450	CERTAIN GLYCOL
3	2016	PHILADELPHIA	PHILADELPHIA	PA	39.913540	-75.198890	LEAD COMPOUND
4	2016	PHILADELPHIA	PHILADELPHIA	PA	39.913540	-75.198890	BENZENE

```
stations3.info()
```

```
<class 'geopandas.geodataframe.GeoDataFrame'>
RangeIndex: 4663 entries, 0 to 4662
Data columns (total 10 columns):
#   Column          Non-Null Count  Dtype
---  -
0   YEAR            4663 non-null   object
1   CITY            4663 non-null   object
2   COUNTY          4663 non-null   object
3   ST              4663 non-null   object
4   LATITUDE        4663 non-null   float64
5   LONGITUDE       4663 non-null   float64
6   CHEMICAL        4663 non-null   object
7   UNIT_OF_ME      4663 non-null   object
8   TOTAL_RELE      4663 non-null   float64
9   geometry        4663 non-null   geometry
dtypes: float64(3), geometry(1), object(6)
memory usage: 364.4+ KB
```

```
# checking crs coordinates for both
print(stations3.crs)
print(releases.crs)
```

None
EPSG:2272

```
# stations3 = stations3.set_crs
stations3 = stations3.set_crs(epsg=2272, inplace=True)
```

```
# checking crs coordinates for both
print(stations3.crs)
print(releases.crs)
```

EPSG:2272
EPSG:2272

```
# select one release incident in particular
recent_release = releases.iloc[360]

# measure distances from each station
distances = stations3.geometry.distance(recent_release.geometry)
distances
```

```
0      48941.110275
1      14914.687505
2      40646.631420
3           0.000000
4           0.000000
...
4658    41735.245165
4659    40909.967527
4660     4519.771240
4661    32442.454868
4662    20534.504851
Length: 4663, dtype: float64
```

```
# mean distance
print(f'Mean distance to monitoring stations: {distances.mean()} feet')
```

Mean distance to monitoring stations: 35350.82207483399 feet

```
# print minimum
print(stations3.iloc[distances.idxmin()][['COUNTY', 'LATITUDE', 'LONGITUDE']])
```

```
COUNTY      PHILADELPHIA
LATITUDE      39.91354
LONGITUDE     -75.19889
Name: 3, dtype: object
```

Creating a buffer

to understand some points on the map that are some distance away from the reference point

use **folium.GeoJson()** to plot each polygon

```
two_mile_buffer = stations3.geometry.buffer(2*5280)
two_mile_buffer.head()
```

```
0    POLYGON ((2729120.227 256380.179, 2729069.378 ...
1    POLYGON ((2709234.606 224522.905, 2709183.756 ...
2    POLYGON ((2687393.394 261701.856, 2687342.544 ...
3    POLYGON ((2694590.004 221697.388, 2694539.155 ...
4    POLYGON ((2694590.004 221697.388, 2694539.155 ...
dtype: geometry
```

```
# create base map
m = folium.Map(location=[39.9526,-75.1652], zoom_start=11)
HeatMap(data = releases[["LATITUDE", 'LONGITUDE']], radius= 12).add_to(m)

for idx, row in stations3.iterrows():
    Marker([row['LATITUDE'], row['LONGITUDE']]).add_to(m)
```

```
# plot
GeoJson(two_mile_buffer.set_crs(epsg=2272)).add_to(m)
```

```
# show
#m
```

<folium.features.GeoJson at 0x1ccf09ef2d0>

```
# turn a group of polygons into a single polygon
```

```
my_union = two_mile_buffer.geometry.unary_union
print('Type: ', type(my_union))
```

```
# show
my_union
```

Type: <class 'shapely.geometry.multipolygon.MultiPolygon'>

