Spatial data

Manipulation and plotting

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Overview

- <u>Introduction</u>
- Spatial fundementals: shapes and coordinates
- Working with spatial data
 - Geopandas for data stucturing
 - Plotting: static and <u>interactive</u>
- Spatial procedures:
 - input-output
 - set-operations and joins
 - interpolation
- Other: assignments and exam project

Motivation

Why spatial data?

- Data supply is exploding
 - Free public big data
 - Infrastructure and buildings
 - Weather, housing market, traffic, job openings
 - Other APIs
 - Private big data
 - Smartphone: GPS sensors/wifi
- Spatial proximity > social relation?

Why spatial data? (2)

Let's try some free public data:

• Where are the buildings in Denmark? (i.e. where do people live)

In [2]: f_restaurant_buffer

Out[2]:



Why spatial data? (3)

What is spatial data?

- What are spatial objects? Do you know any?
- Spatial objects include classic geometric objects (lines, circles, squares etc.)
- Divided into four different generic shapes:
 - Point
 - LineString
 - Polygon (also approximates circles)
 - Multipolygon

What is spatial data? (2)

Example of a LineString

```
In [3]: from shapely.geometry import LineString
    line_coords = np.array([[0,3],[1,3],[2,4]])
    LineString(line_coords)
Out[3]:
```

What is spatial data? (3)

Example of a polygon

```
In [4]: from shapely.geometry import Polygon
    triangle_coords = np.array([[0,0],[2,0],[1,2]])
    Polygon(triangle_coords)
```

Out[4]:



What is spatial data? (4)

Examples of multi-polygons

Out[5]:





How do we measure spatial data?

We need to define the space we work in. What could this be?

- Observations on earth use different Coordinate References System (CRS).
- Some CRS are local and some are global.

How do we measure spatial data? (2)

The standard CRS is:

- The Global Positioning System (GPS)
 - Technically called WGS84
 - Has EPSG code 4326 (EPSG: European Petroleum Survey Group, now extinct)

How do we measure spatial data? (3)

Working with Danish spatial data you often encounter:

- The Danish mapping reference
 - Technically called ETRS89 UTM Zone 32 North
 - Has EPSG code 25832
 - This is used in Danish admin data.
 - Coordinates measured in meters: can use Euclidian distance

Applied spatial data

How do we work with spatial data?

GeoPandas:

- Collection of spatial objects
 - Powerful data structuring
 - Static plots
- Combines
 - Pandas (labelled arrays and methods)
 - Shapely (Python shape objects)
 - Fiona (low level, ultra fast for complex operations)

Folium

• Interactive plots: map overlay for OpenStreetMaps and zoom

Spatial Data Structures

Spatial shapes

We have already seen how to employ shapely to construct spatial shapes.

```
In [6]: unit_circle_points = [(np.cos(d),np.sin(d)) for d in np.linspace(0, 2*np.pi, 6)]
Polygon(unit_circle_points)
Out[6]:
```

What if we have more than one spatial object?

We could store spatial objects in lists but that is not smart. Why?

GeoSeries

A smart container for spatial data is the GeoSeries:

- 1d array with labels (like Pandas series)
- useful spatial tools

POLYGON ((20.59024743010491 41.85540416113361,...

Name: geometry, dtype: object

GeoDataFrame

There is also the GeoDataFrame which has a dedicated column for geometry:

In [8]: world_map.head()

Out[8]:

	pop_est	continent	name	iso_a3	gdp_md_est	geometry
0	28400000.0	Asia	Afghanistan	AFG	22270.0	POLYGON ((61.21081709172574 35.65007233330923,
1	12799293.0	Africa	Angola	AGO	110300.0	(POLYGON ((16.32652835456705 - 5.87747039146621
2	3639453.0	Europe	Albania	ALB	21810.0	POLYGON ((20.59024743010491 41.85540416113361,
3	4798491.0	Asia	United Arab Emirates	ARE	184300.0	POLYGON ((51.57951867046327 24.24549713795111,
4	40913584.0	South America	Argentina	ARG	573900.0	(POLYGON ((-65.5000000000003 -55.199999999999

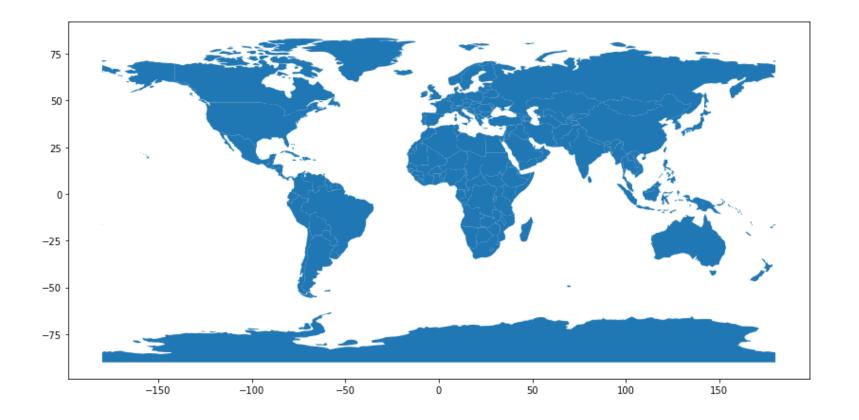
Map making

Static plots

Let's try to plot the world's countries. Easy as hell with GeoPandas:

```
In [9]: world_map.plot(figsize=(14,7))
```

Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x190e2a070b8>

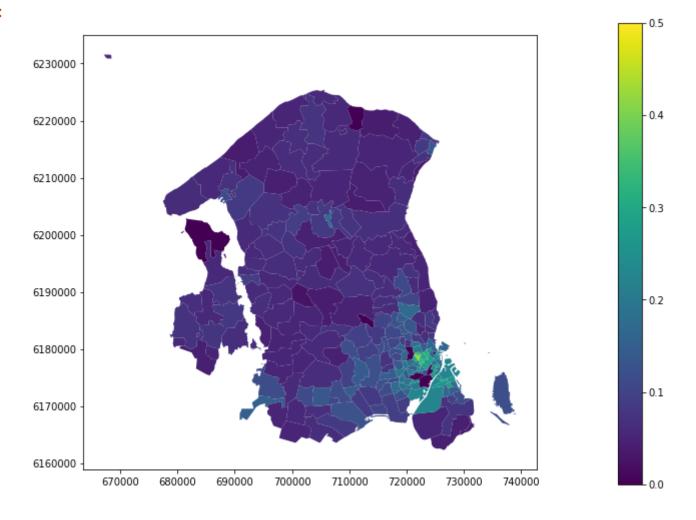


Static plots (2)

Now let's try to plot the share of young around in the Capital Region of Denmark (excluding Bornholm):

In [10]: | f_cph_young

Out[10]:

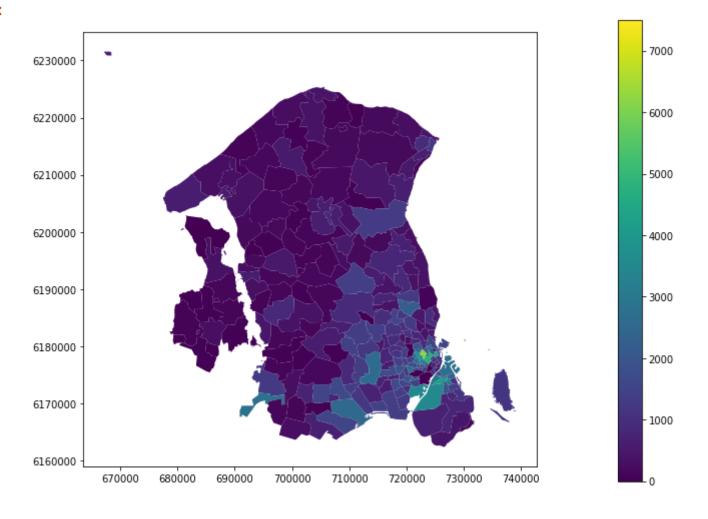


Static plots (3)

Never plot sum or count !!!!

In [11]: f_cph_young_count

Out[11]:



Folium: Plots with interaction

Make a map of many points for Danish supermarkets.

In [12]:	m_dk_supermarket
Out[12]:	+

Loading and storing spatial data

There are two standard formats:

- GeoJSON (.geojson):
 - This is getting more common, especially in more modern applications.
- Shapefiles (.shp)
 - The de-facto standard format. This is a collection of files where the central one is the . shp file.

Example - load house price data:

```
In [13]: gdf = gpd.read_file('house_prices.geojson')
```

Spatial operations

When having two or more spatial objects it is possible to make interactions between them.

Basic operations

- Shortest distance between shapes (e.g. Euclidian distance between two points)
 - What about in spherical coordinates? (using Haversine formula is one solution)

Spatial operations (2)

Advanced operations

- Set like operations ~ make new shapes:
 - Intersection (A and B)
 - Union (A or B)
 - Difference (A not B)
- We can ask whether two shapes
 - Intersect or not
 - Touch or not

Spatial operations (3)

Basic 'set-like' operations in shapely.

```
In [14]: rect_coords = [[0, 0], [0, 2], [1, 2], [1, 0]]

A = Polygon(triangle_coords)
B = Polygon(rect_coords)

# A | B # union
# A & B # intersection
# A - B # set difference
```

Spatial operations (4)

Example of spatial join through intersection with GeoPandas' sjoin.

```
In [15]:
```

supermarket_cph = gpd.sjoin(gdf_supermarket, kommuner_cph) supermarket_cph.head(3)

Out[15]:

	id	lat	lon	tags	type	geometry	nan	e index_right	komkode	komnavı
8	272860527	55.655972	12.485416	{'name': 'Netto', 'shop': 'supermarket'}	node	POINT (719255.6405412797 6173300.325984053)	Netto	167	0101	Københavi
9	272860703	55.643196	12.501363	{'name': 'Netto', 'shop': 'supermarket'}	node	POINT (720330.1542702764 6171929.991743754)	Netto	167	0101	Københavi
15	282680168	55.647999	12.526752	{'name': 'SuperBrugsen', 'opening_hours': 'Mo	node	POINT (721899.9097495852 6172545.122327959)	SuperBrugse	n 167	0101	Københavi

Spatial operations (5)

Check output of spatial join

In [16]: centroid = supermarket_cph.centroid.to_crs(epsg=4326)
 for lat,lon in zip(centroid.y, centroid.x):
 marker_cluster_cph.add_child(folium.Marker(location=[lat, lon]))
 marker_cluster_cph.add_to(m_dk_supermarket_cph)
 m_dk_supermarket_cph

Out[16]:



31

206

28

Leaflet (http://leafletjs.com)

Spatial interpolation

Why spatial interpolation?

- Often we are interested in
 - Data for neighborhoods not just points
 - Having measure available everywhere
- Spatial data is often sparse and pointwise
 - What about areas in between?
 - Which points to use?

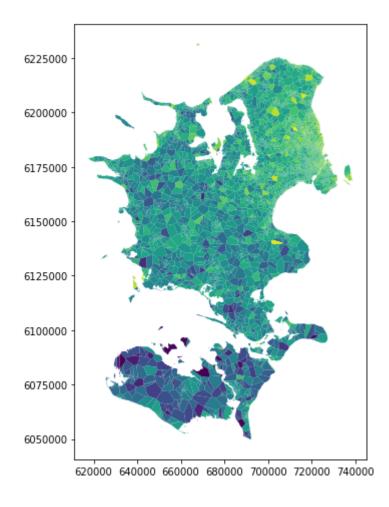
Why spatial interpolation? (2)

Let's plot 2012 house sale prices.

• We notice some rural regions have little coverage

In [17]: | f_price_pointcloud

Out[17]:



What is spatial interpolation?

What is the simplest possible spatial interpolation pattern?

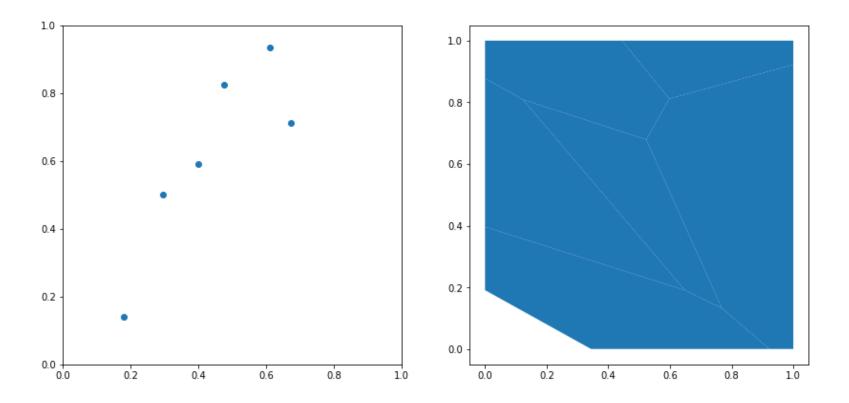
- The nearest neighbor?
- The nearest neighbor plot is known a the Voronoi plot

What is spatial interpolation? (2)

Let's see what a Voronoi plot looks like

In [18]: from shapely.geometry import box, Point f, ax = plt.subplots(1, 2, figsize=(14, 7)) coords = pd.DataFrame(np.random.rand(6,2)) gpd.GeoSeries(coords.apply(Point,axis=1)).plot(ax=ax[0]) ax[0].set_xlim(0,1) ax[0].set_ylim(0,1) voronoi_wrapper(coords, box(0,0,1,1)).plot(ax=ax[1])

Out[18]: <matplotlib.axes._subplots.AxesSubplot at 0x190eabb8550>



What is spatial interpolation? (3)

The Voronoi plot for Danish house sale prices in 2012:

What is spatial interpolation? (4)

The nearest neighbor method (exercise)

- Find the k nearest points
 - Regression: Take mean of k-nearest
 - Classification: Take mode of k-nearest
- The measurement space can be physical e.g. surface distance i 2d
- A machine learning model
 - non-parametric approach
 - requires L2 normalization

Summary

We have learned about

- Spatial shapes and coordinate systems
- Perform operations on single shapes and collections of shapes
- Make interactive maps
- Interpolate spatial data

Group work

Assignments

• individual hand-in

Projects

- data
- groups

Further topics

- Spatial autocorrelation
- Interpolation with kriging: assesing uncertainty