

Geographic Information Systems: GeoPandas

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Introduction to Programming for Public Policy

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Geographic Information Systems: Maps

People love maps – emotional response to ‘seeing yourself.’

- ▶ This is a huge field: there are entire classes at Harris and around the University for GIS, spatial statistics, etc.
- ▶ But huge bang for the buck at the entry level.
 - ▶ Easy to make compelling graphics.
 - ▶ Many datasets represent a spatial area or point at a specific time.
⇒ Great potential for joins!

What We'll Cover

1. Making simple maps with GeoPandas (pandas+).
 - ▶ Finding and importing shapefiles and geojson (like `read_csv()`).
 - ▶ Projections (briefly).
2. Attribute and spatial joins.
 - ▶ Geocoding APIs: geopy/nominatim, Census, etc.
 - ▶ Making a map with real data!
3. Making a simple web (!) map with GeoPandas
 - ▶ Largely revisiting old material, with new functions.

Shapefiles

- ▶ Three forms of geographic objects: points (schools, crimes), lines (roads, rivers), and polygons (lots, census tracts, regions, lakes, etc.).
- ▶ Many, many sources for geographic data: [Chicago](#), [US Census](#), [USGS](#), [Global Administrative Areas](#) (GADM), etc.
- ▶ Much of this is provided in 'ESRI Shapefiles' (Environmental Systems Research Institute, major GIS company) or in geojson. Modern databases (postgres) are helpful for assembling large datasets.
 - ▶ Shapefiles come zipped with a lot of other files. The shp file is the 'master' file, and references the others. That's what you import.
 - ▶ Let's browse: [census shapefiles](#).
- ▶ Addresses may be geocoded and coordinates are also points!

Loading a Shapefile with GeoPandas

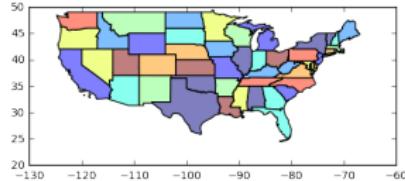
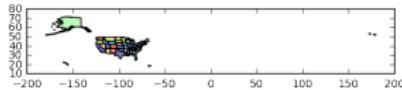
- ▶ **GeoPandas** simply adds a geometry series to a pandas DataFrame.
- ▶ It is tightly integrated with many other geographic programs, among them **fiona** for reading geojson/shapefiles and **shapely** for geometric operations (intersections, etc.).
- ▶ Really easy to import! Both shapefiles and geojson:

```
import geopandas as gpd  
gdf = gpd.read_file("myfile.shp")  
gdf.plot() # WOW!!!!
```

- ▶ All of the 'standard' dataframe operations (slicing, indexing, merging) are still available.

Making a Better Map

- ▶ Let's restrict ourselves to the contiguous 48 states.
- ▶ Make a mask to get rid of Alaska and Hawaii (STATEFP 2 and 15), and the territories (STATEFP ≥ 57).
 - ▶ Alternative: translate, rotate, and scale them with [shapely](#).
- ▶ We also need a better projection: `gpd.to_crs(epsg=2163)`.



Coordinate Reference Systems (CRS)

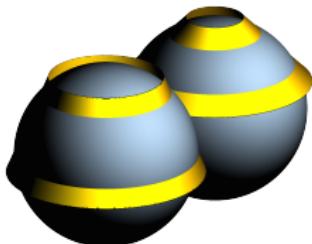
- ▶ To make maps, we need a description of the shape of Earth (an ellipsoid) and an origin/center. This is called a **datum**.
- ▶ We also need a **projection** from 3D to 2D.
- ▶ These are standardized in **EPSG codes**:
 - 4269** By default, GeoPandas uses a plate carée projection: a mapping of longitude and latitude lines to horizontal and vertical lines (gross).
 - 3857** Most online maps use web Mercator, which is conformal (preserves shapes/angles) but much-maligned.
 - 2163** Albers Equal Area is a good conic projection for the US. ✓
- ▶ Inappropriate projections make maps look stupid.



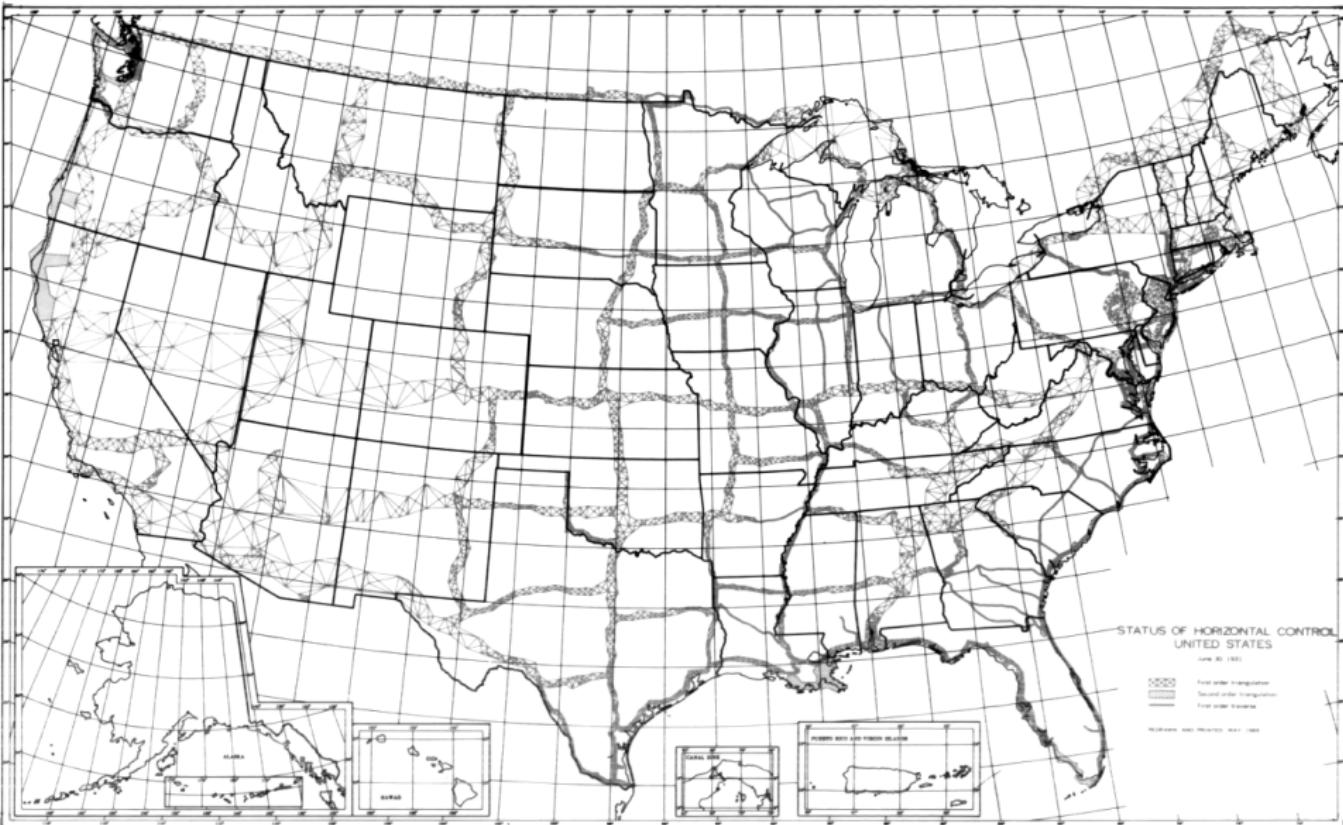
Center of the World



~ Goode Homolosine



Albers Equal Area!!



Joins

Several distinct goals:

1. Visualize a dataset as a map (join it to a shapefile).
2. Attribute join on two datasets with matching geometries.
 - ▶ Don't care about the geometry, just use it!
3. Datasets with different geometries (e.g., points and polygons).
 - ▶ Use a spatial join; may not care about map!

Attribute Join

- ▶ Attribute joins are the joins we've already been doing with pandas.
- ▶ Prepare them for the join by matching the indices (state codes).

Two examples:

1. Single mothers in the United States.
 - ▶ Join the state shapes to data from the ([census API](#)).
2. Voting returns in Pennsylvania from the election return [site](#)

Choropleth Maps: Shaded Areas

Easy to make basic, beautiful maps!

```
gdf.plot(column = "Percent Mothers Unmarried",
          scheme = "quantiles", k = 10,
          cmap = "rainbow", alpha = 0.4,
          edgecolor = "black", linewidth = 0.5,
          legend = True, figsize = (12, 8))
```

- ▶ The built-in method also allows for `equal_intervals` (linear, default), `quantiles`, and `fisher_jenks`.
 - ▶ Fisher Jenks defines categories by minimizing the in-group variance , and maximizing the between-group variance.
 - ▶ Most lay-people will only understand equal intervals!!
 - ▶ Without a scheme, geopandas will make a smooth, equal-interval coloring but a bad legend.
 - ▶ Can also use `vmin` and `vmax` for a smooth colormap.
- ▶ There are many, many colormaps.

Point to Polygon: Spatial Joins

- ▶ Spatial joins (`sjoin`) use properties of two geometries – instead of equality of attributes – to align rows.
 - ▶ One geometry 'contains', is 'within', or 'intersects' another:

```
gpd.sjoin(pt_df, poly_df, how='left', op='within')
```

- ▶ For example: aggregate crimes (points) by community area (polygon).
- ▶ The geometry from the left DataFrame is preserved.

Building a GeoDataFrame from Scratch

- ▶ We also need to be able to create a GeoDataFrame from scratch.
- ▶ A GeoDataFrame, is just a DataFrame with a ‘GeoSeries.’
- ▶ The GeoSeries is just a list of points, which we can construct as:

```
from shapely.geometry import Point  
pt = Point(x, y)
```

- ▶ Create the GeoDataFrame by setting the geometry and CRS (4269):

```
gpd.GeoDataFrame(crime_df, crs = {'init':  
'epsg:4269'}, geometry=geometry)
```

**Example: associate murders to
census tracts and community areas.**

- ▶ Folium creates a powerful javascript map on OpenStreetMap.
- ▶ Nice interface, easily embedded in other sites:

- ▶ `<iframe src="map.html" width=800px height=500px></iframe>`

```
import folium

m = folium.Map([40, -98], tiles='cartodbpositron',
               zoom_start=4, max_zoom=14, min_zoom=4)

ft = "Percent Mothers Unmarried"
colormap = folium.LinearColormap(("orange", "white", "purple"),
                                  vmin = geo_merge[ft].min(),
                                  vmax = geo_merge[ft].max(),
                                  caption = ft)
colormap.add_to(m)

folium.GeoJson(geo_merge,
               style_function = lambda feature: {
                   'fillColor': colormap(feature['properties'][ft]),
                   "color" : "black", "weight" : 1, "fillOpacity" : 0.4
               }).add_to(m)

m.save("mothers.html")
```

Other Folium Features

- ▶ You can plot a collection of points with GeoJson, but you can get somewhat more control with

```
folium.Marker([41.7855052, -87.5971531],  
             popup='Harris School').add_to(map)
```

- ▶ See also e.g., CircleMarker, RegularPolygonMarker, etc.
- ▶ Full documentation [here](#).
- ▶ We'll come back to this after our last example.

Geocoding

- ▶ Often, we have latitudes and longitudes (ready to be wrapped as points), but addresses.
- ▶ Geocoding is the process of turning addresses into coordinates.
 - ▶ We have already done this with the google API.
- ▶ Many geocoding services (Census, Texas A&M) also provide census tracts, counties, etc. \implies Huge time saver!

Built-In Geocoding

- ▶ geopy plugs into the OpenStreetMap ‘Nominatim’ API, as well as google bing, yahoo, and openmapquest.
 - ▶ Some of the others require API keys for large numbers of requests.

```
from geopy.geocoders import Nominatim  
Nominatim().geocode("1155 E. 60th St, Chicago 60637")
```

- ▶ GeoPandas provides tools to do many geocodes in one pass:

```
gpd.tools.geocode(["London", "Paris",  
                   "New York", "Hong Kong"])
```

Census API (For Interest)

- ▶ The Census geocoding API matches tracts in geography endpoint
 - ▶ Also standard location mode.
- ▶ Capable of up to 1000 addresses at a time in batch mode:

```
curl -F addressFile=@short.csv\  
      -F benchmark=Public_AR_Current\  
      -F vintage=ACS2015_Current \  
      https://geocoding.geo.census.gov/geocoder/geographies/addressbatch
```

Merges and Spatial Operations as Geocoding

Spatial operations (`intersects`, `within`, and `contains`) are effectively geocodes:

```
geo_df[geo_df.contains(pt)]["NAME"]
```

Second Folium Example

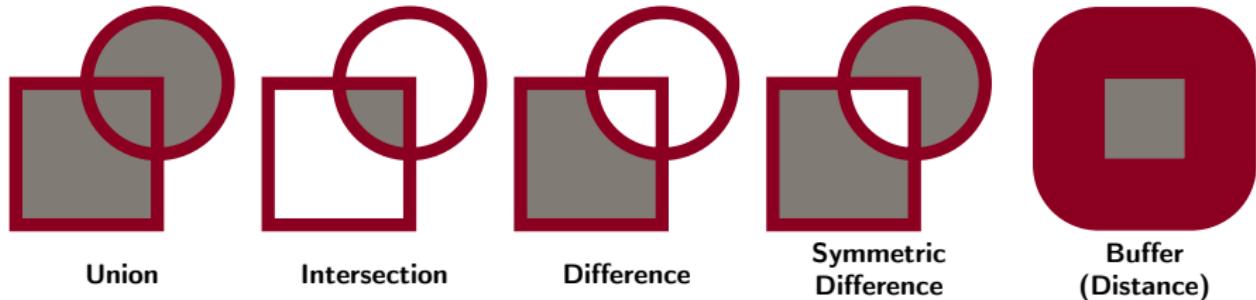
- ▶ Make a map of places represented in this class (points and countries).
- ▶ Curl these shapefiles for the world:

http://thematicmapping.org/downloads/TM_WORLD_BORDERS_SIMPL-0.3.zip

Spatial Operations: Beyond Joins and Aggregations

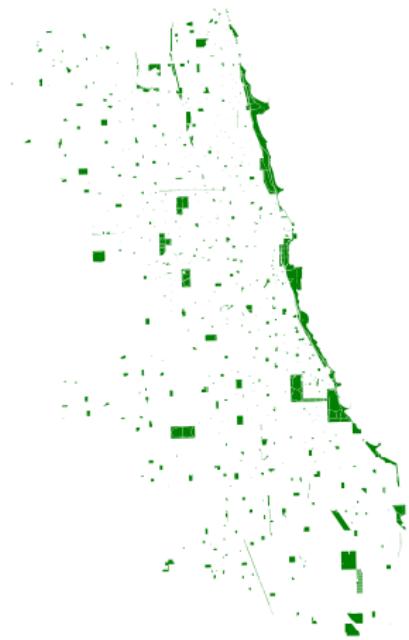
With spatial operations, we can create variables that express the distances to a point of service/care (trauma center), or the amount of a resource (e.g., park-land or grocery store square-footage) available within a radius.

- ▶ Non-trivial questions: how close do you have to be to a park to enjoy it, or to a grocery store to use it?
- ▶ Geopandas provides these operations through **shapely**.
- ▶ To reduce an entire dataset to one shape: `gdf.unary_union`.

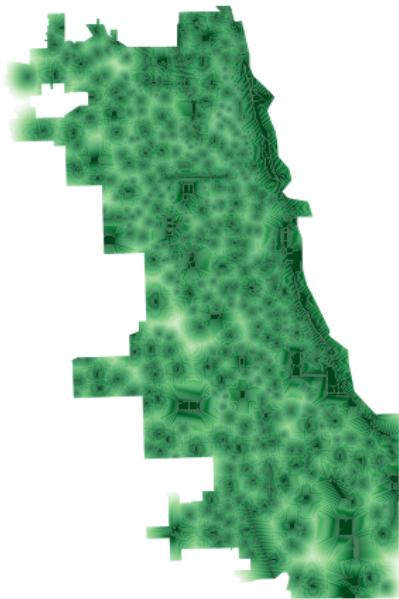


Access to Parks

City Parks



Distance to Area-Weighted Park



Quantiles of Area, < 250m from Census Tract

