

# Geographic Information Systems

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Introduction to Programming for Public Policy

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

- ▶ GIS is a huge field. There are other classes at Harris for this.
  - ▶ A lot of work on 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!

## 1. Making simple maps with GeoPandas (pandas+).

- ▶ Finding and importing shapefiles and geojson (like `read_csv()`).
- ▶ Projections (briefly).

## 2. Attribute and spatial joins.

- ▶ Using the census geolocation API (APIs).
- ▶ Making a map with real data!

## 3. Making a simple web (!) map with GeoPandas

- ▶ Largely revisiting old material.

# 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: [data.cityofchicago.org](http://data.cityofchicago.org), the [US Census](#), [USGS](#), etc.
- ▶ Much of this is provided in 'ESRI Shapefiles' (Environmental Systems Research Institute, major GIS company) or in geojson.
  - ▶ Shapefiles come zipped with a lot of other files, it's the shp you want.
  - ▶ Let's browse: [census shapefiles](#).
- ▶ Addresses may be geocoded and coordinates are also points!

# Loading a Shapefile with GeoPandas

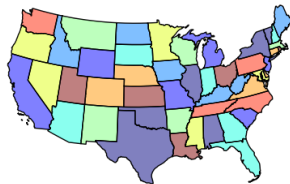
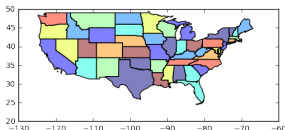
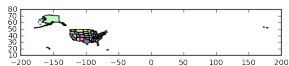
- ▶ **GeoPandas** simply adds a geometry series to a pandas DataFrame.
- ▶ It is tightly integrated with many other geographic programs, among them **fiona** for reading geojson/shapefiles and **shapely** for geometric operations (intersections, etc.).
- ▶ Really easy to import! Both shapefiles and geojson:

```
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 Slightly 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  $\geq 57$ ).
- ▶ We can also use a better projection: `gpd.to_crs(epsg=2163)`.



# Coordinate Reference Systems (CRS)

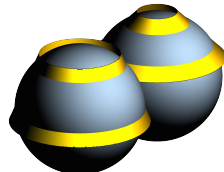
- ▶ To make maps, we need a description of the shape of Earth (an ellipsoid) and an origin/center. This is called a **datum**.
- ▶ We also need a **projection** from 3D to 2D.
- ▶ These are standardized in **EPSG codes**:
  - 4269** By default, GeoPandas uses a Plate-Carée projection: a mapping of longitude 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!!

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 choropleth maps!

---

```
gdf.plot(column = "Percent Mothers Unmarried",  
         scheme = "quantiles", k = 5,  
         cmap = "rainbow", legend = True,  
         alpha = 0.4, linewidth = 0.5,  
         figsize = (12, 8))
```

---

- ▶ The built-in method also allows for quantiles (default), equal\_intervals (linear), 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!!
- ▶ There are many, many [colormaps](#).
- ▶ The defaults only allow 9 breaks; see `Adanced.ipynb` for a gradient.

# Point to Polygon: Spatial Joins

- ▶ GeoPandas provides many powerful geometric operations.
- ▶ Spatial joins (`sjoin`) use the geometry as you might expect, :  
`gpd.sjoin(pt_df, poly_df, how = 'left', op = 'within')`
  - ▶ You can also do 'contains', 'within', or 'intersects'.
- ▶ This joins rows with locations (points) in one dataframe, to regions (polygons) in the other.

# Building a GeoDataFrame from Scratch

- ▶ We also need to be able to create a GeoDataFrame from scratch.
- ▶ A GeoDataFrame, as we've said, is just a DataFrame with a geometry.
- ▶ So we need to build the GeoSeries...
- ▶ In this case, the GeoSeries consists of a list of points, which we can construct as

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

- ▶ Then create a GeoDataFrame, by setting the geometry and crs:

```
gpd.GeoDataFrame(crime_df, crs = tract_df.crs,  
geometry=geometry)
```

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

- ▶ Folium creates a powerful javascript map on OpenStreetMap.
- ▶ Really nice interface, easily embedded in other sites:
  - ▶ `<iframe src="map.html" width=800px height=500px></iframe>`

```
import folium
```

```
m = folium.Map([39.828175, -98.5795], tiles='cartodbpositron',  
               zoom_start=4, max_zoom=14, min_zoom=4,)
```

```
ft = "Percent Mothers Unmarried"
```

```
cmap = folium.colormap.linear.YlOrRd.scale(merged[ft].min(), merged[ft].max())
```

```
folium.GeoJson(merged, style_function=lambda feature: {  
    'fillColor': cmap(feature['properties'][ft]),  
    'fillOpacity': 0.6,  
    'weight': 2, 'color': 'black'  
}).add_to(m)
```

```
cmap.caption = 'Percent Children Born to Single Mothers'  
cmap.add_to(m)
```

```
m.save("us_single_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.

- ▶ Often, we have latitudes and longitudes (ready to be wrapped as points), but addresses.
- ▶ Geocoding is the process of turning addresses into coordinates.
- ▶ Many geocoding services can additionally provide census tracts, counties, etc.  $\implies$  Huge time saver!



- ▶ geopy plugs into the OpenStreetMap 'Nominatim' API.
- ▶ Super easy to use!!

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

- ▶ 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 layers=9 \  
-F vintage=ACS2015_Current \  
-F benchmark=Public_AR1_Current \  
https://geocoding.geo.census.gov/geocoder/geographies/addressbatch
```

# Geocoding in GeoPandas

- ▶ GeoPandas has geopy built-in, with google, bing, yahoo, openmapquest, or nominatim.
  - ▶ Nominatim is great, but has a 1-second request delay.
  - ▶ Some of the others require API keys for large numbers of requests.

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

- ▶ Of course you can also 'geocode' areas with `contains()`:

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

## Second Folium Example

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

[http://thematicmapping.org/downloads/TM\\_WORLD\\_BORDERS\\_SIMPL-0.3.zip](http://thematicmapping.org/downloads/TM_WORLD_BORDERS_SIMPL-0.3.zip)