Plotting with GeoJSON

VISUALIZING GEOSPATIAL DATA IN PYTHON

Mary van Valkenburg
Data Science Program Manager,
Nashville Software School

datacamp

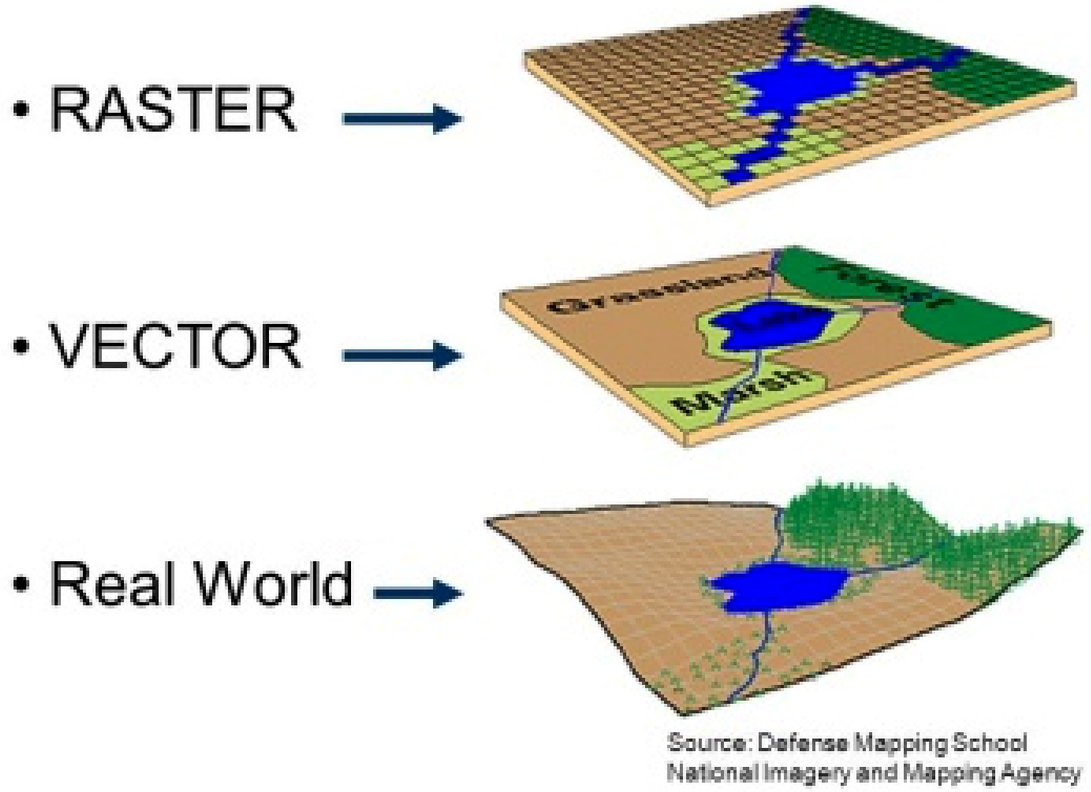
Neighborhoods GeoJSON

```
{ "type": "FeatureCollection",
  "features": [
    { "type":"Feature",
      "properties":{
    "name":"Historic Buena Vista"
    },"geometry":{
        "type":"MultiPolygon","coordinates":[[[[-86.79511056795417,36.17575....
```

```
neighborhoods = gpd.read_file('./data/neighborhood_boundaries.geojson')
neighborhoods.head(1)
```

```
                name                                           geometry
Historic Buena Vista    (POLYGON ((-86.79511056795417 36.17575964963348...)))
```

Geopandas dependencies

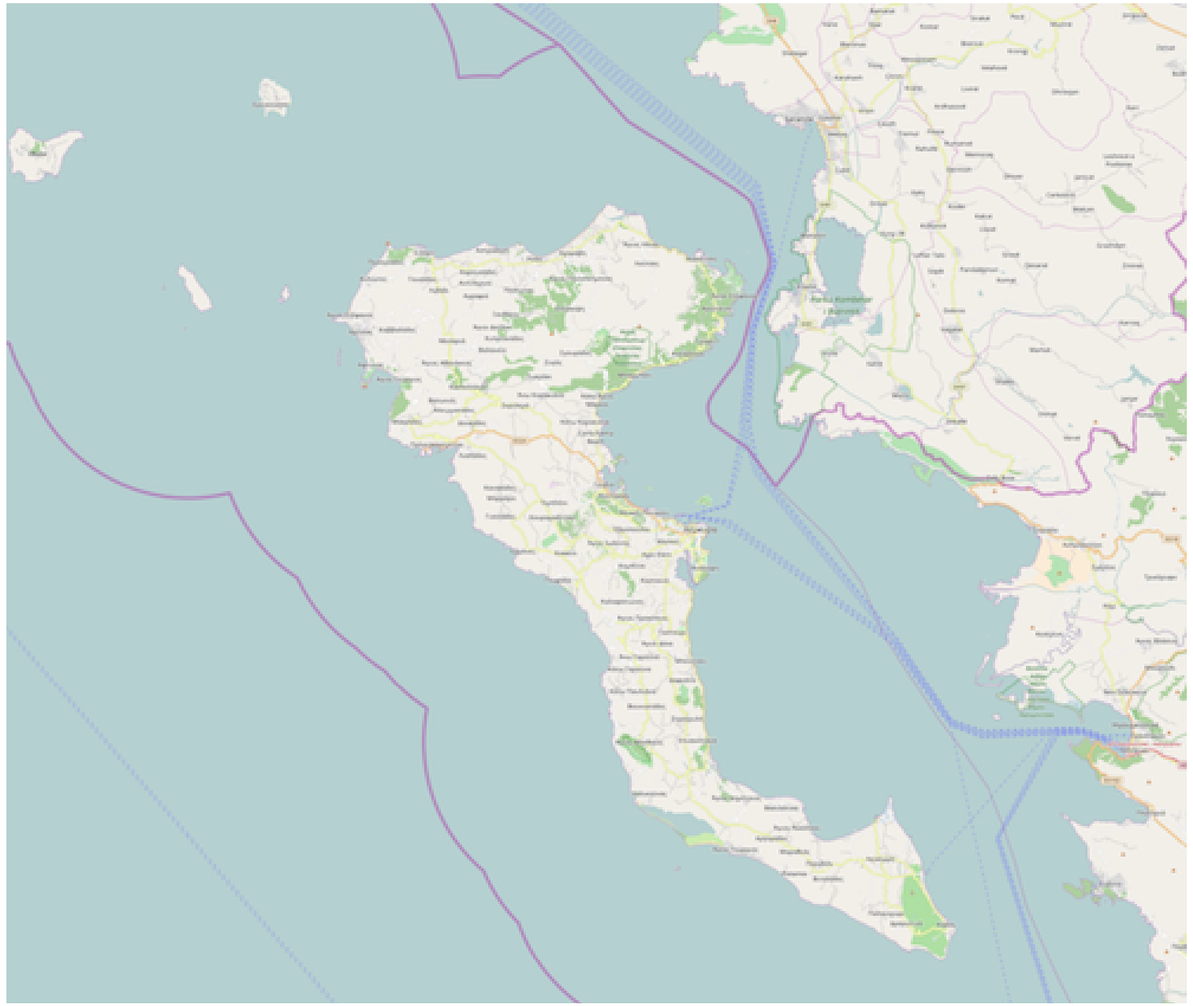


- Fiona
  - provides an python API for OGR
- GDAL/OGR
  - GDAL for translating raster data
  - OGR for translating vector data

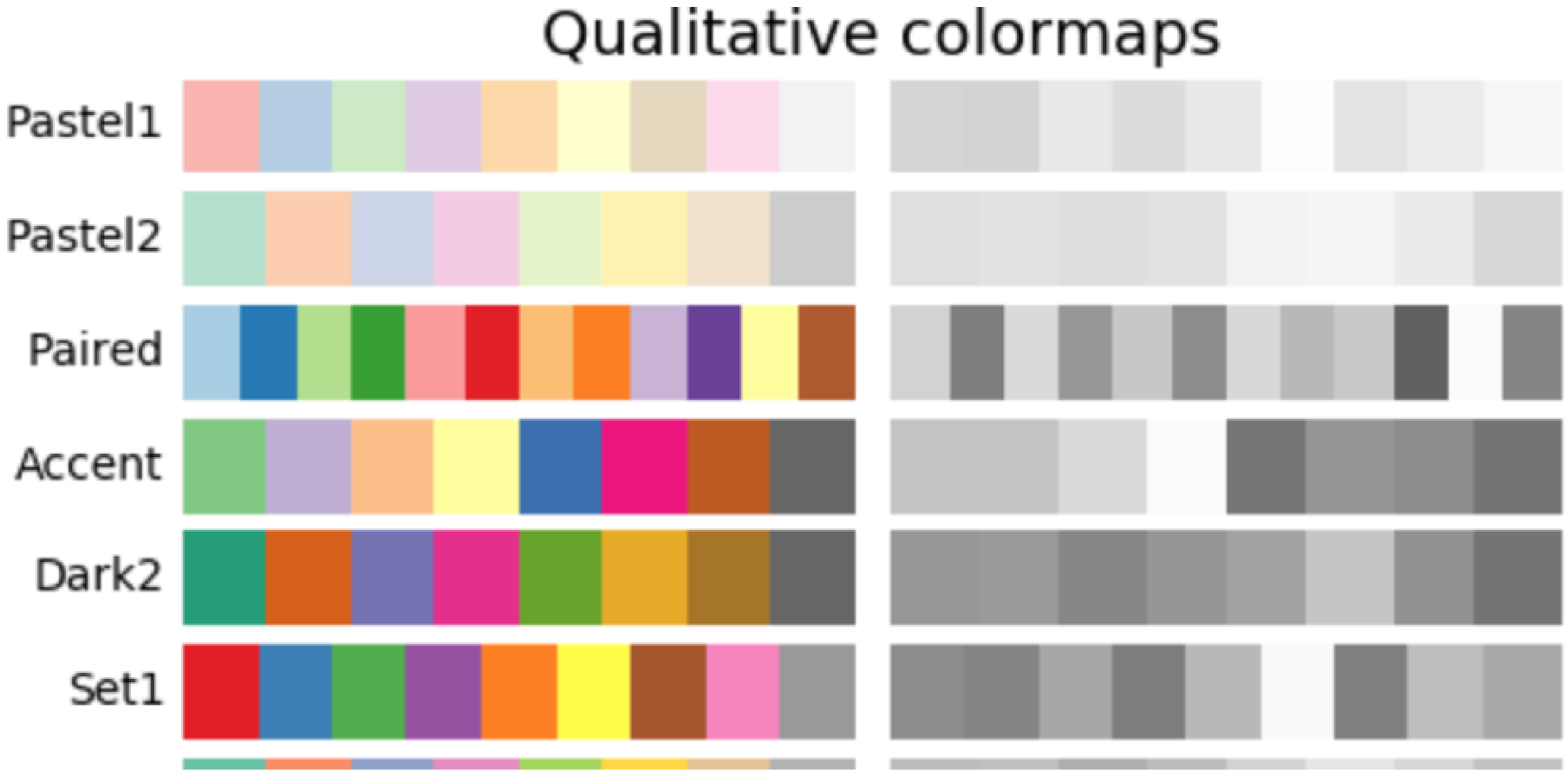
Comparing raster and vector graphics

raster image of Corfu

vector image of Corfu



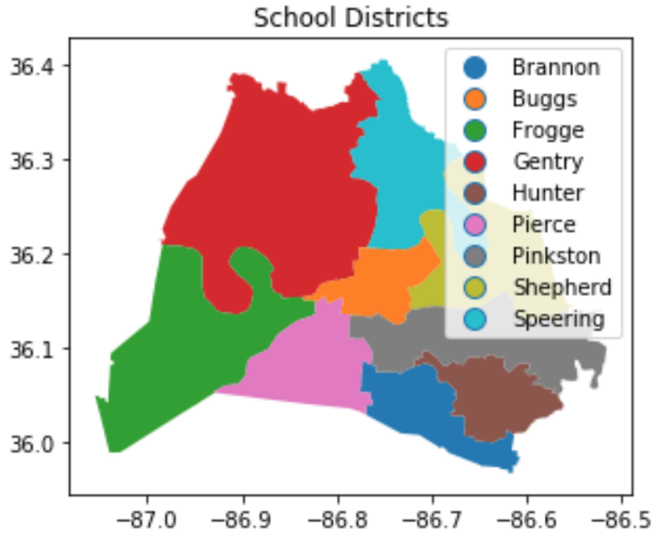
Colormaps


Qualitative colormaps
Pastel1
Pastel2
Paired
Accent
Dark2
Set1

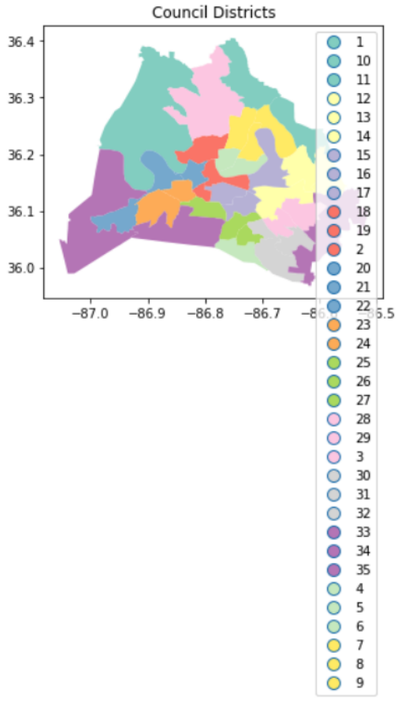
Plotting with color

```
school_districts.head(3)
```

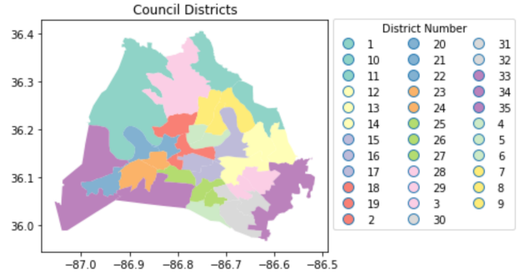
```
first_name     last_name     position     district      geometry
Sharon         Gentry        Member        1     (POLYGON ((-86.771 36.383)...))
Jill           Speering      Vice-Chair    3     (POLYGON ((-86.753 36.404)...))
Jo Ann         Brannon       Member        2     (POLYGON ((-86.766 36.083)...))
```



```
council_dists.plot(
                    column='district',
                    cmap='Set3',
                    legend=True)
plt.title('Council Districts')
plt.show();
```




```
leg_kwds={'title':'District Number',
          'loc': 'upper left',
          'bbox_to_anchor':(1, 1.03),
          'ncol':3}
council_dists.plot(column='district',
                   cmap='Set3',
                   legend=True,
                legend_kwds=leg_kwds)
plt.title('Council Districts')
plt.show();
```



Let's practice!

VISUALIZING GEOSPATIAL DATA IN PYTHON

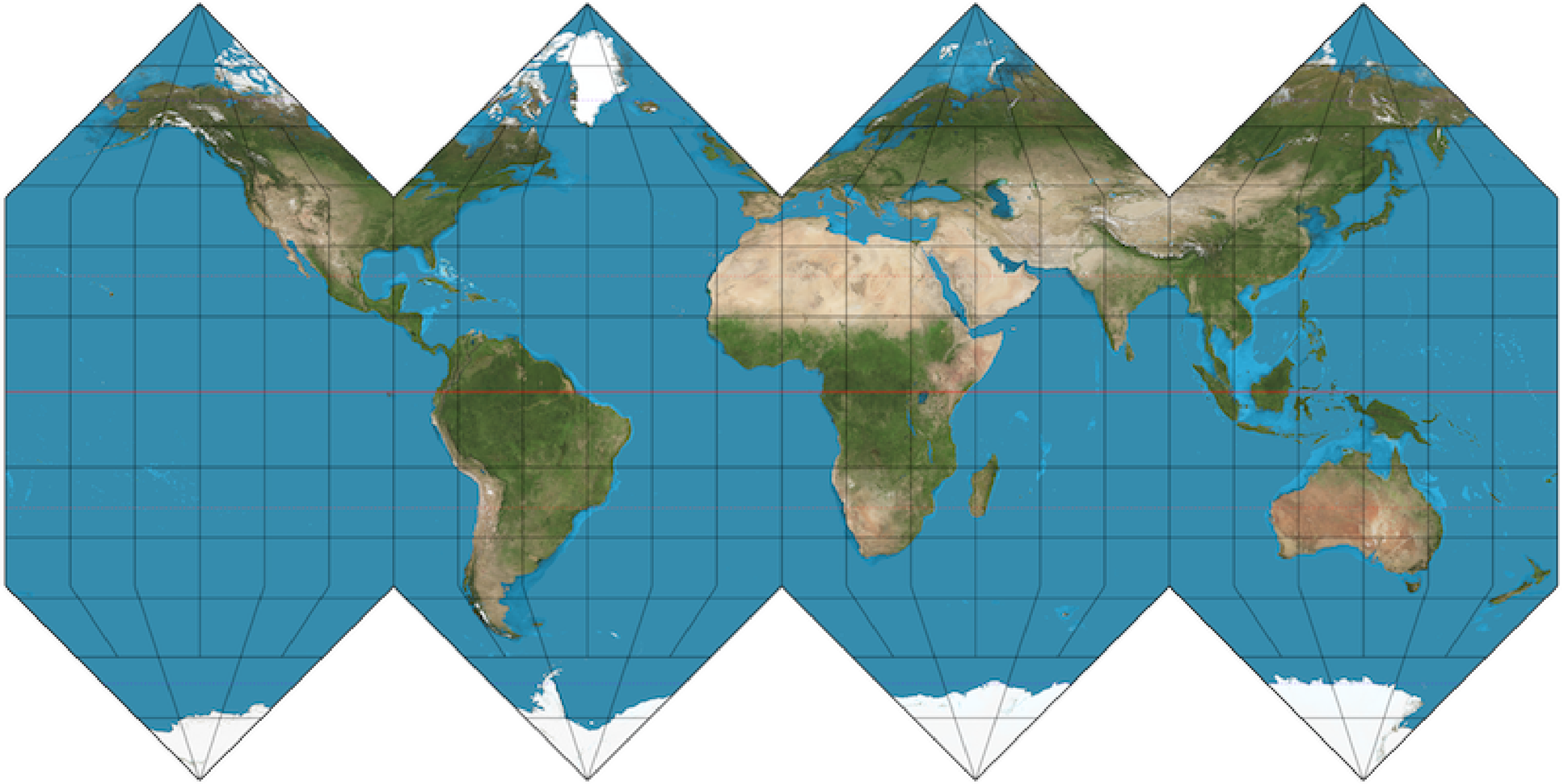
Projections and Coordinate Reference Systems

VISUALIZING GEOSPATIAL DATA IN PYTHON

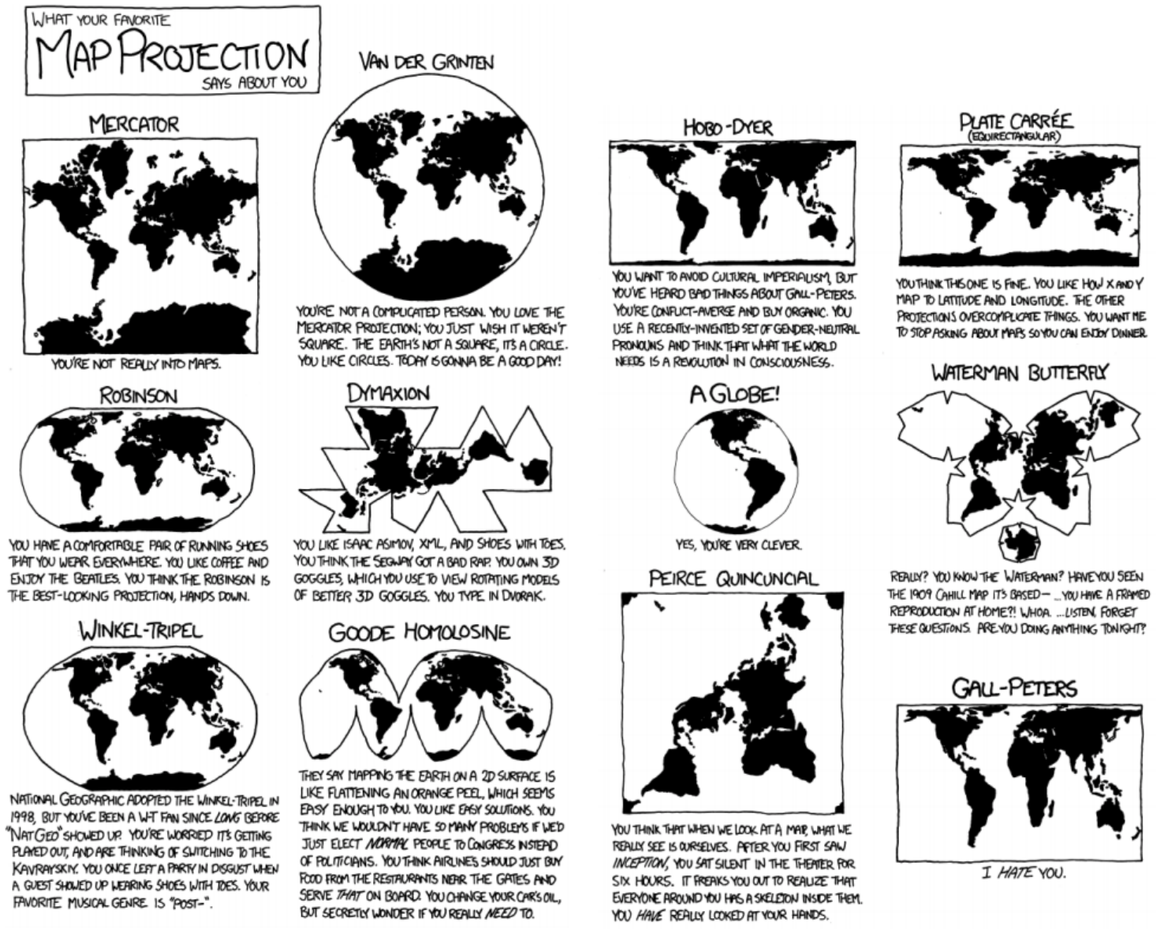
Mary van Valkenburg
Data Science Program Manager,
Nashville Software School

datacamp

Projections



Many approaches to map projection



*What's that? You think I don't like the Peters map because I'm uncomfortable with having my cultural assumptions challenged? Are you sure you're not... ::puts on sunglasses:: ... **projecting**?* http://xkcd.com/977/ - http://bit.ly/explainxkcd-977

Coordinate Reference Systems

EPSG:4326

- used by Google Earth
- units are decimal degrees

EPSG:3857

- used by Google Maps, Bing Maps, Open Street Maps
- units are meters

```
School Name                 Latitude     Longitude
A. Z. Kelley Elementary     36.021       -86.658
Alex Green Elementary       36.252       -86.832
Amqui Elementary            36.273       -86.703
```

```
# create a point geometry column
from shapely.geometry import Point
schools['geometry'] = schools.apply(lambda x: Point((x.Longitude, x.Latitude)),
                                                    axis = 1)

schools.head(3)
```

```
School Name                 Latitude     Longitude     geometry
A. Z. Kelley Elementary     36.021    -86.658       POINT (-86.658 36.021)
Alex Green Elementary       36.252    -86.832       POINT (-86.832 36.252)
Amqui Elementary            36.273    -86.703       POINT (-86.703 36.273)
```

Creating a GeoDataFrame from a DataFrame

```
import geopandas as gpd

schools_crs = {'init':  'epsg:4326'}
schools_geo = gpd.GeoDataFrame(schools,
                               crs = schools_crs,
                               geometry = schools.geometry)
```

```
schools_geo.head(3)
```

```
School Name                Latitude     Longitude     geometry
A. Z. Kelley Elementary    36.021    -86.658       POINT (-86.658 36.021)
Alex Green Elementary      36.252    -86.832       POINT (-86.832 36.252)
Amqui Elementary           36.273    -86.703       POINT (-86.703 36.273)
```

Changing from one CRS to another

```
schools_geo.head(2)
```

```
School Name                Latitude     Longitude     geometry
A. Z. Kelley Elementary    36.021    -86.658       POINT (-86.658 36.021)
Alex Green Elementary      36.252    -86.832       POINT (-86.832 36.252)
```

```
# convert geometry from decimal degrees to meters
schools_geo.geometry = schools_geo.geometry.to_crs(epsg = 3857)
schools_geo.head(2)
```

```
School Name                Latitude    Longitude    geometry
A. Z. Kelley Elementary    36.021    -86.658    POINT (-9646818.8 4303623.8)
Alex Green Elementary      36.252    -86.832    POINT (-9666119.5 4335484.4)
```

Let's practice!

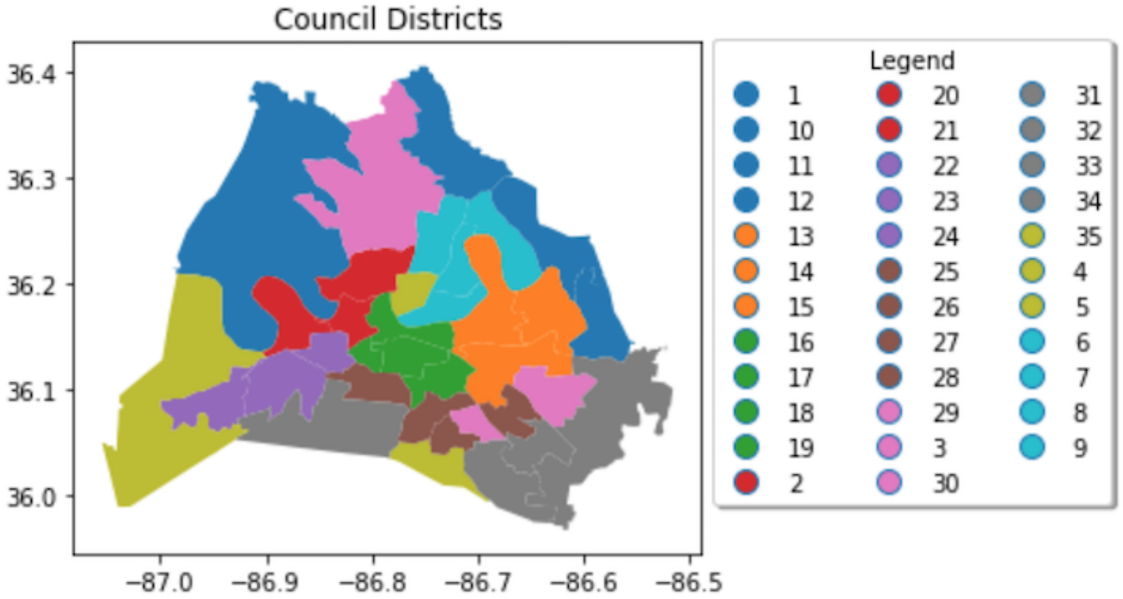
VISUALIZING GEOSPATIAL DATA IN PYTHON

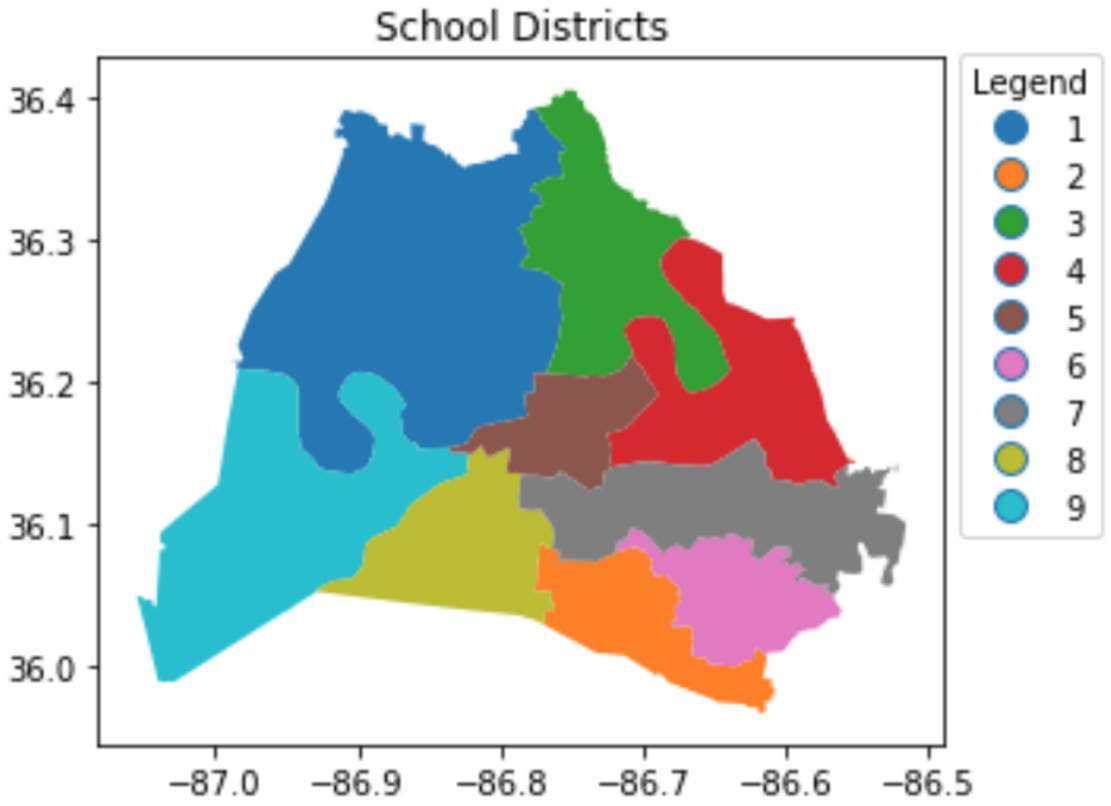
Spatial joins

VISUALIZING GEOSPATIAL DATA IN PYTHON

Mary van Valkenburg
Data Science Program Manager,
Nashville Software School

Council districts and school districts





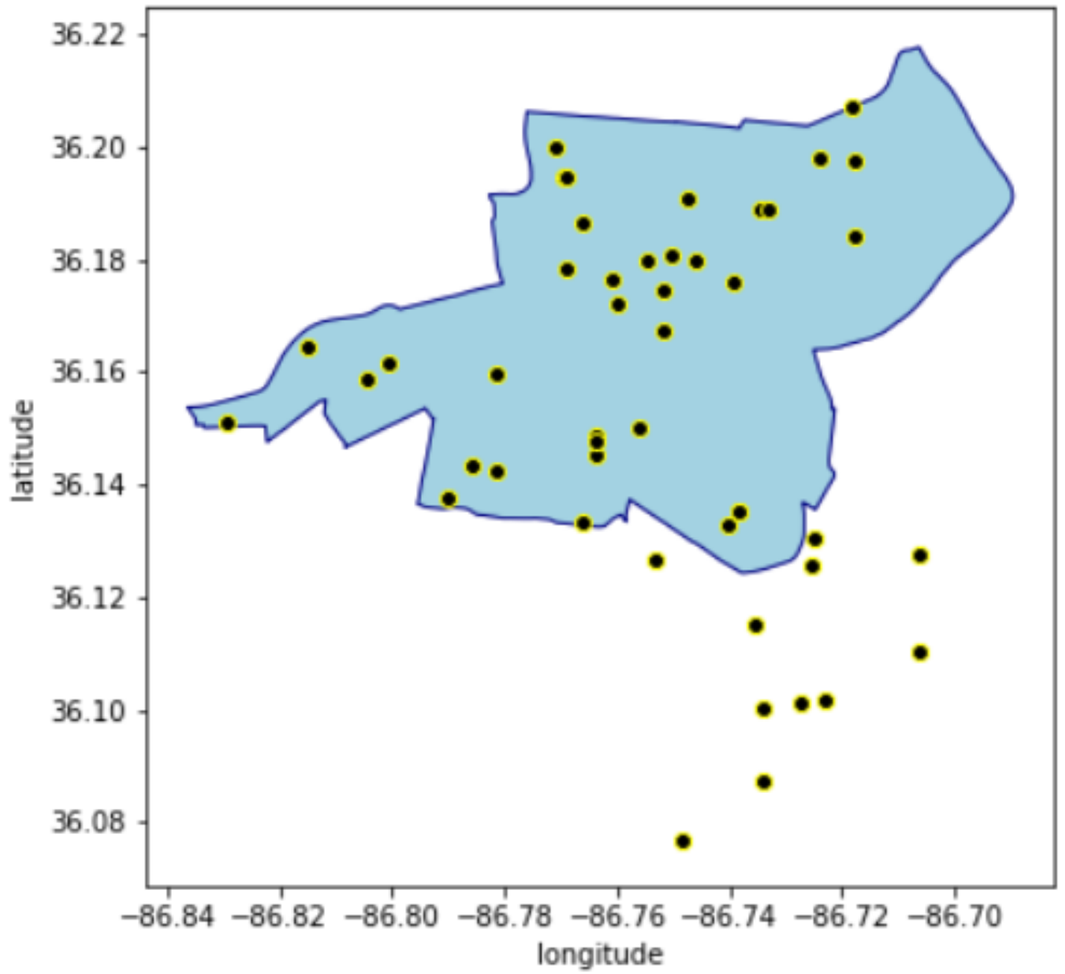
The .sjoin() op argument

```
import geopandas as gpd


gpd.sjoin(blue_region_gdf, black_point_gdf, op = <operation>)
```

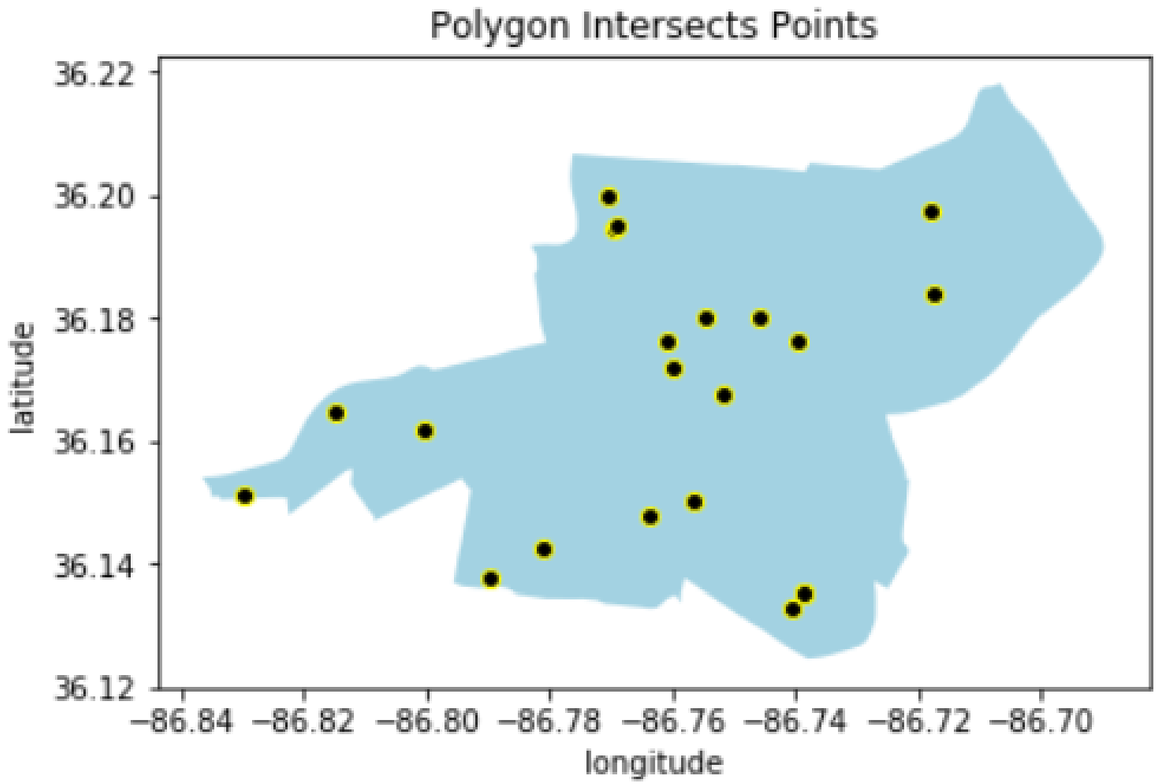
operation can be ***intersects***, ***contains***, or ***within***

Using .sjoin()


36.22
36.20
36.18
36.16
36.14
36.12
36.10
36.08
latitude
−86.84
−86.82
−86.80
−86.78
−86.76
−86.74
−86.72
−86.70
longitude

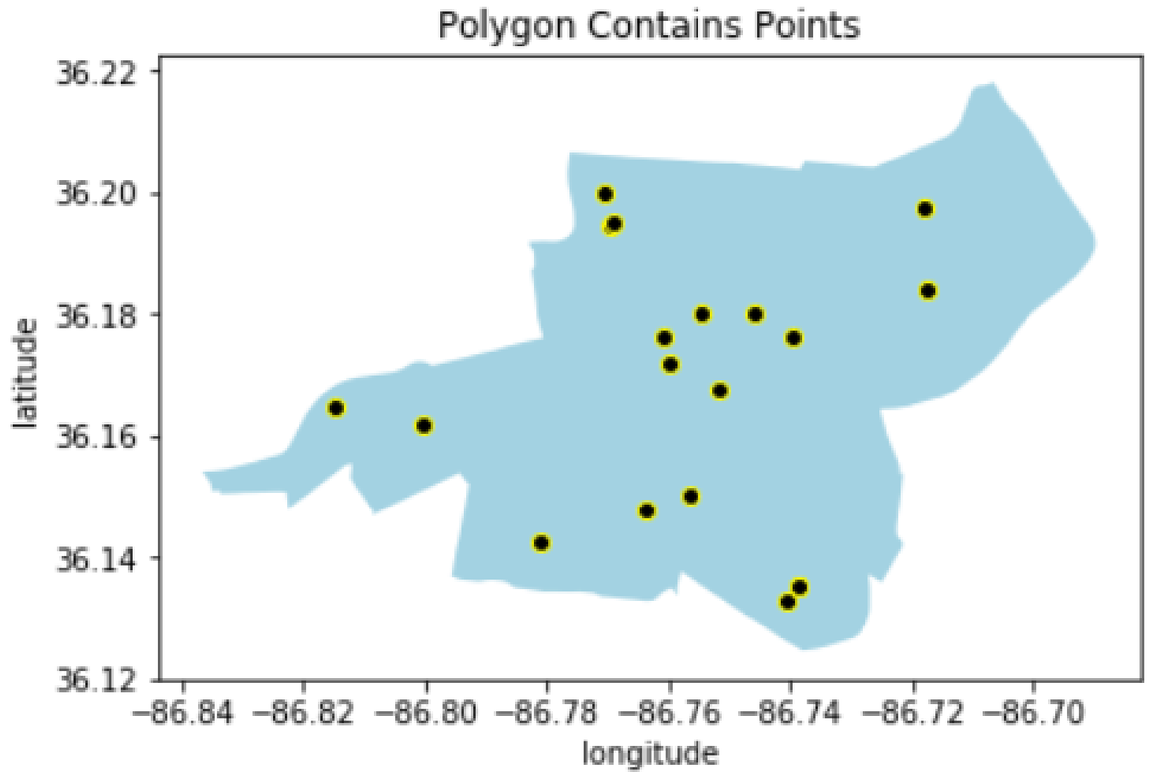
op = 'intersects'

```
gpd.sjoin(blue_region_gdf, black_point_gdf, op = 'intersects')
```



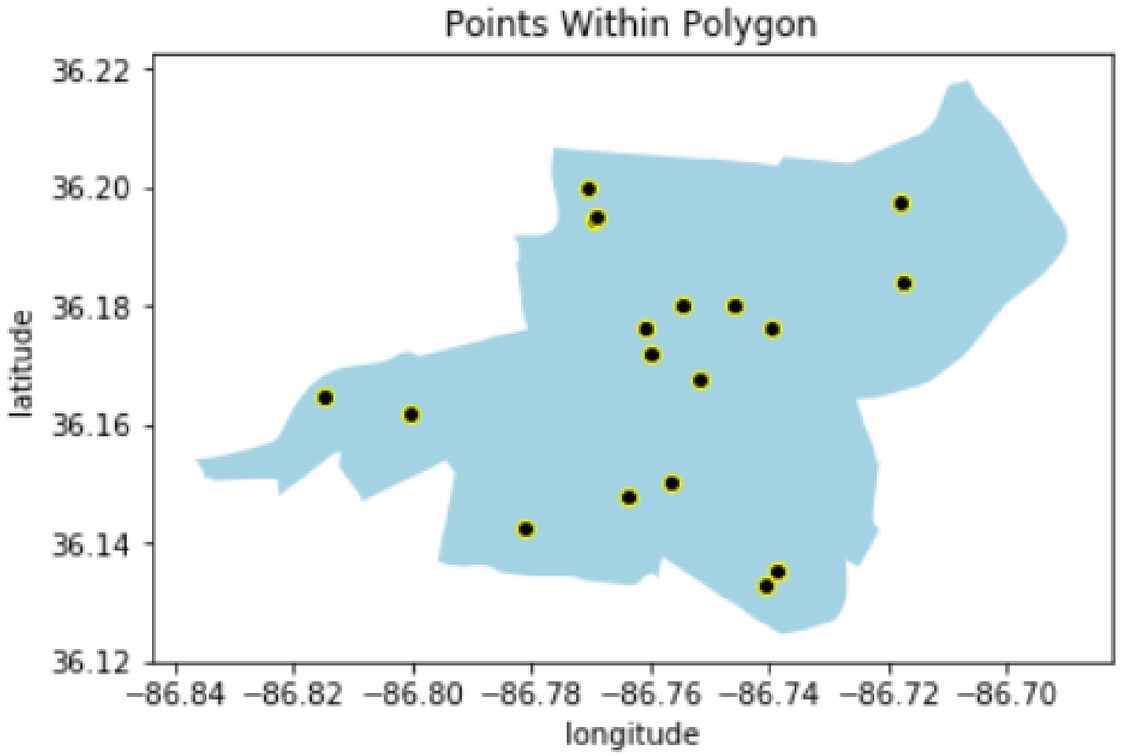
op = 'contains'

```
gpd.sjoin(blue_region_gdf, black_point_gdf, op = 'contains')
```



op = 'within'

```
gpd.sjoin(black_point_gdf, blue_region_gdf, op = 'within')
```



The sjoin.() op argument - within

```
# find council districts within school districts

within_gdf =gpd.sjoin(council_districts, school_districts, op='within')
print('council districts within school districts: ', within_gdf.shape[0])
```

```
council districts within school districts:  11
```

The sjoin.() op argument - contains

```
# find school districts that contain council districts

contains_gdf=pd.sjoin(school_districts, council_districts, op='contains')
print('school districts contain council districts: ', contains_gdf.shape[0])
```

```
school districts contain council districts:  11
```

The sjoin.() op argument - intersects

```
# find council districts that intersect with school districts


intersect_gdf=gpd.sjoin(council_districts, school_districts, op='intersects')
print('council districts intersect school districts: ', intersect.shape[0])
```

```
council districts intersect school districts:  100
```

Columns in a spatially joined GeoDataFrame

```
within_gdf=gpd.sjoin(council_districts, school_districts, op = 'within')
within_gdf.head()
```

	first_name_left	last_name_left	district_left	index_right
0	Nick	Leonardo	1	0
1	DeCosta	Hastings	2	0
2	Nancy	VanReece	8	1
3	Bill	Pridemore	9	1
9	Doug	Pardue	10	1

```
# Aggregate council districts by school district - first rename district_left and district_right
within_gdf.district_left = council_district
within_gdf.district_right = school_district
within_gdf[['council_district', 'school_district']
          ].groupby('school_district'
                    ).agg('count'
                         ).sort_values('council_district', ascending = False)
```

```
                 council_district
school_district
       3                        3
       1                        2
       9                        2
       2                        1
       5                        1
       6                        1
       8                        1
```

Let's Practice!

VISUALIZING GEOSPATIAL DATA IN PYTHON