What is a choropleth?

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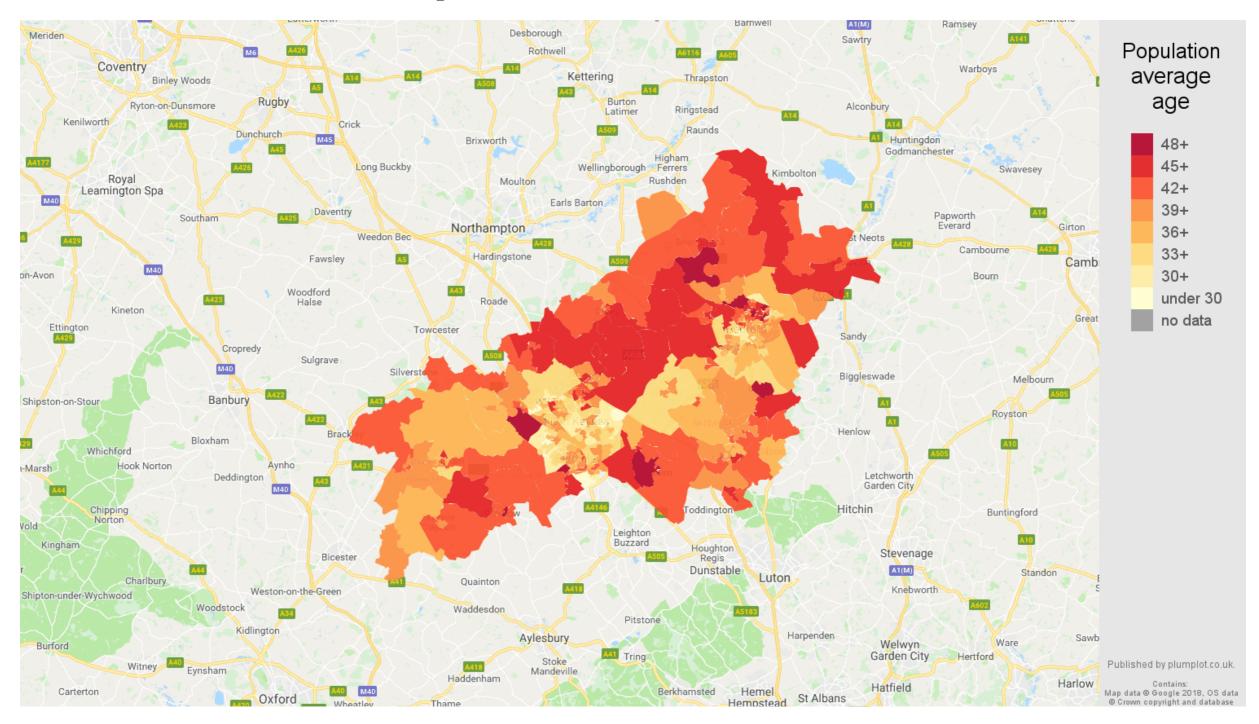


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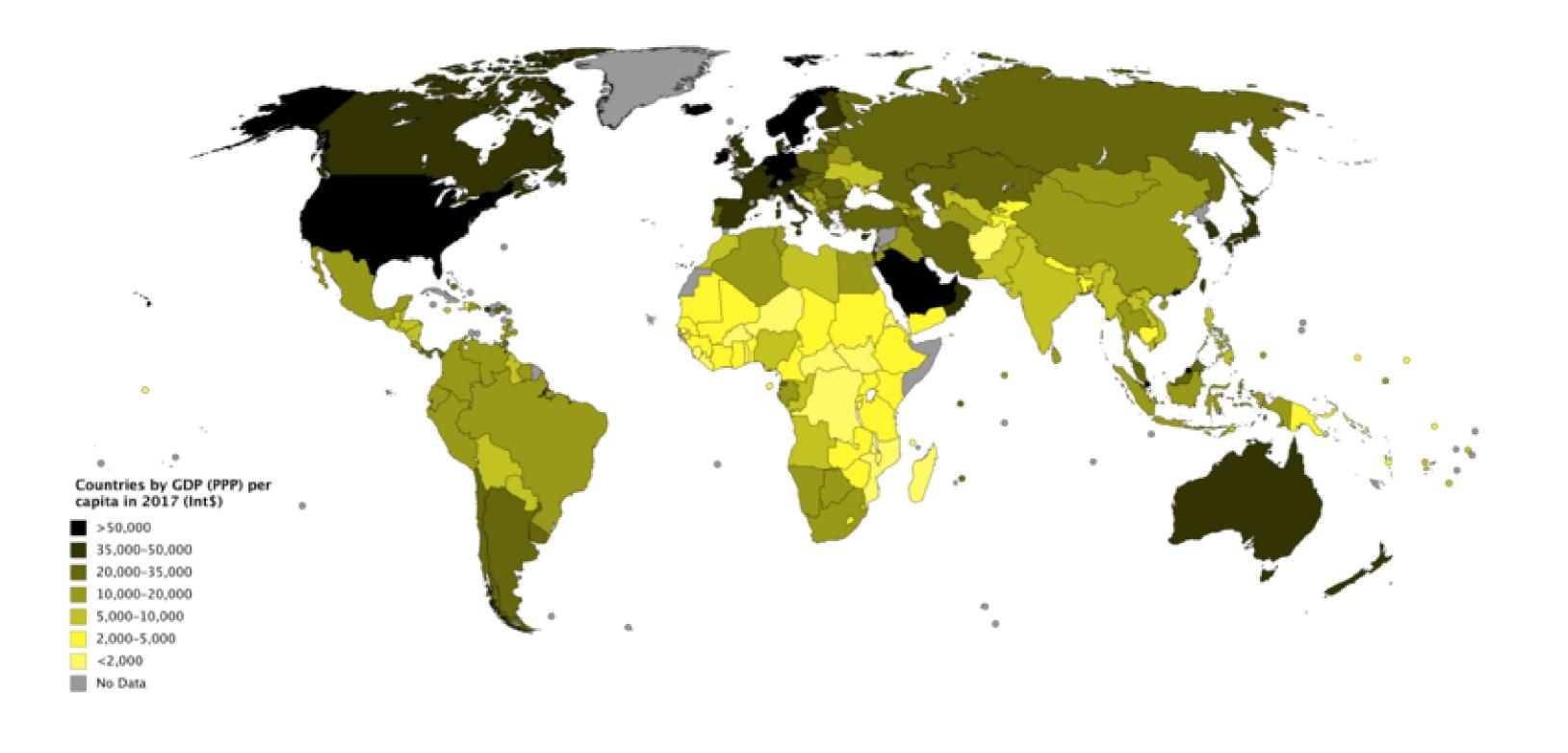
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Definition of a choropleth









Density

```
schools_in_districts.head(2)
```

```
district geometry name lat lng
1 (POLYGON ((-86.77 36.38... Nashville Prep 36.16 -86.85
1 (POLYGON ((-86.77 36.38... Rocketship Prep 36.17 -86.79
```



Get counts

```
school_counts = schools_in_districts.groupby(['district']).size()
print(school_counts)
```

```
district
    30
    11
3
    19
    18
5
    36
6
    21
    13
    10
8
    12
dtype: int64
```

Add counts

```
school_counts_df = school_counts.to_frame()
school_counts_df = school_counts_df.reset_index()
school_counts_df.columns = ['district', 'school_count']
districts_with_counts = pd.merge(school_districts, school_counts_df,
                                on = 'district')
districts_with_counts.head(2)
district geometry
                           school_count
        (POLYGON ((-86.77 36.38... 30
```

(POLYGON ((-86.75 36.40... 19

Divide counts by areas

```
districts_with_counts['area'] = districts_with_counts.area

districts_with_counts['school_density'] = districts_with_counts.apply(
    lambda row: row.school_count/row.area, axis = 1)

districts_with_counts.head(2)
```

```
district geometry school_count area school_density
1 (POLYGON ((-86.77 36.38... 30 0.036641 818.745403
3 (POLYGON ((-86.75 36.40... 19 0.014205 1337.594495
```

Let's Practice!

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Choropleths with geopandas

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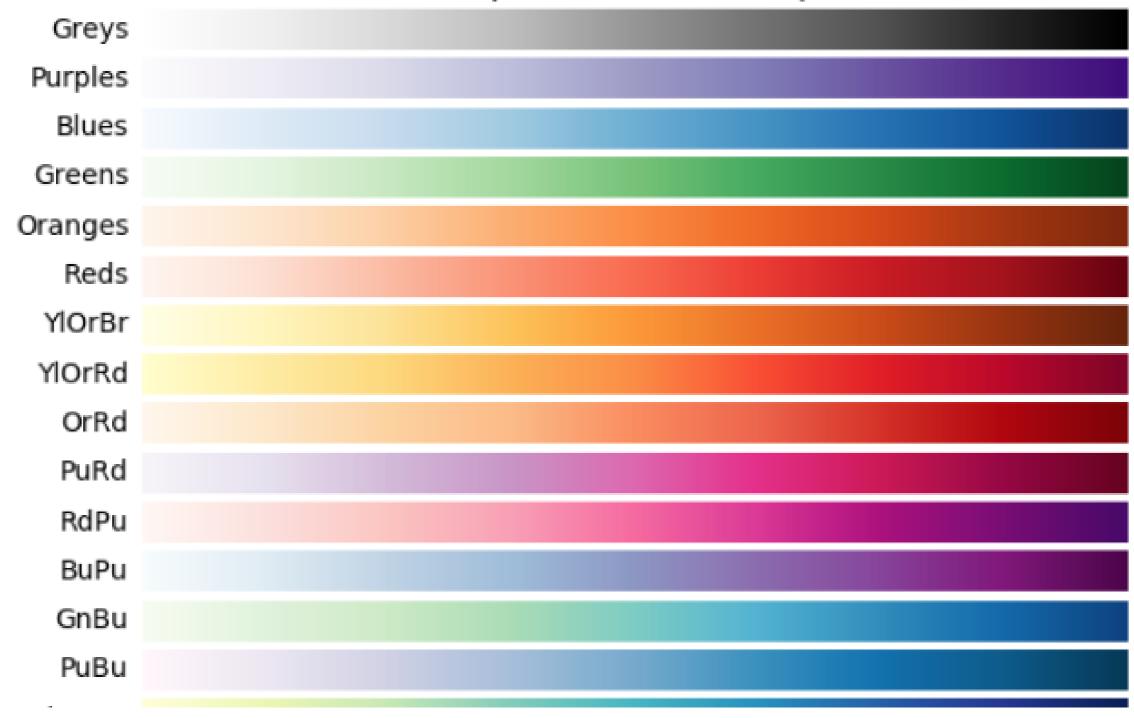


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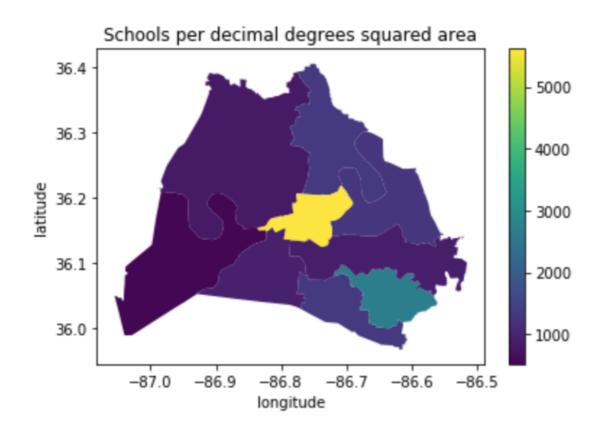
Sequential colormaps





Choropleth with GeoDataFrame.plot()

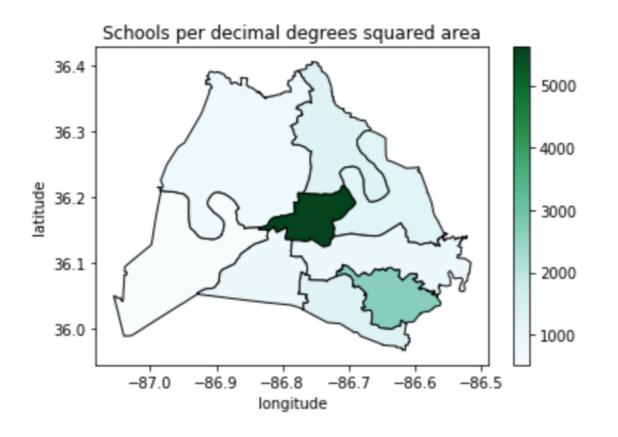
```
districts_with_counts.plot(column = 'school_density', legend = True)
plt.title('Schools per decimal degrees squared area')
plt.xlabel('longitude')
plt.ylabel('latitude');
```





Choropleth with GeoDataFrame.plot()

```
districts_with_counts.plot(column = 'school_density', cmap = 'BuGn', edgecolor = 'black', legend = True)
plt.title('Schools per decimal degrees squared area')
plt.xlabel('longitude')
plt.ylabel('latitude');
```





Area in Kilometers Squared

```
# starting CRS
print(school_districts.crs)
```

epsg:4326

```
# convert to EPSG 3857
school_districts = school_districts.to_crs(epsg = 3857)
print(school_districts.crs)
```

epsg:3857

Area in Kilometers Squared

```
# define a variable for m^2 to km^2
sqm_to_sqkm = 10**6

school_districts['area'] = school_districts.area / sqm_to_sqkm
school_districts.head(2)
```

```
district geometry area

1 (POLYGON ((-965.055 4353528.766... 563.134380

3 (POLYGON ((-965.823 4356392.677... 218.369949
```



```
# change crs back to 4326
school_districts = school_districts.to_crs(epsg = 4326)
print(school_districts.crs)
```

epsg:4326

```
print(school_districts.head(2))
```

```
district geometry area

1 (POLYGON ((-86.771 36.383... 563.134380

3 (POLYGON ((-86.753 36.404... 218.369949
```

```
# spatial join to get districts that contain schools
schools_in_districts = gpd.sjoin(school_districts, schools_geo, predicate = 'contains')
```



```
# aggregate to get counts
school_counts = schools_in_districts.groupby(['district']).size()
# convert school_counts to a df
school_counts_df = school_counts.to_frame()
school_counts_df = school_counts_df.reset_index(level=0)
school_counts_df.columns = ['district', 'school_count']
# merge
districts_with_counts = pd.merge(school_districts,
                                school_counts_df, on = 'district')
districts_with_counts.head(1)
district geometry
                                                    school_count
                                        area
          (POLYGON ((-86.771 36.383.. 563.134380
                                                       30
```



Calculating school density

plt.title('Schools per kilometers squared')

plt.xlabel('longitude')

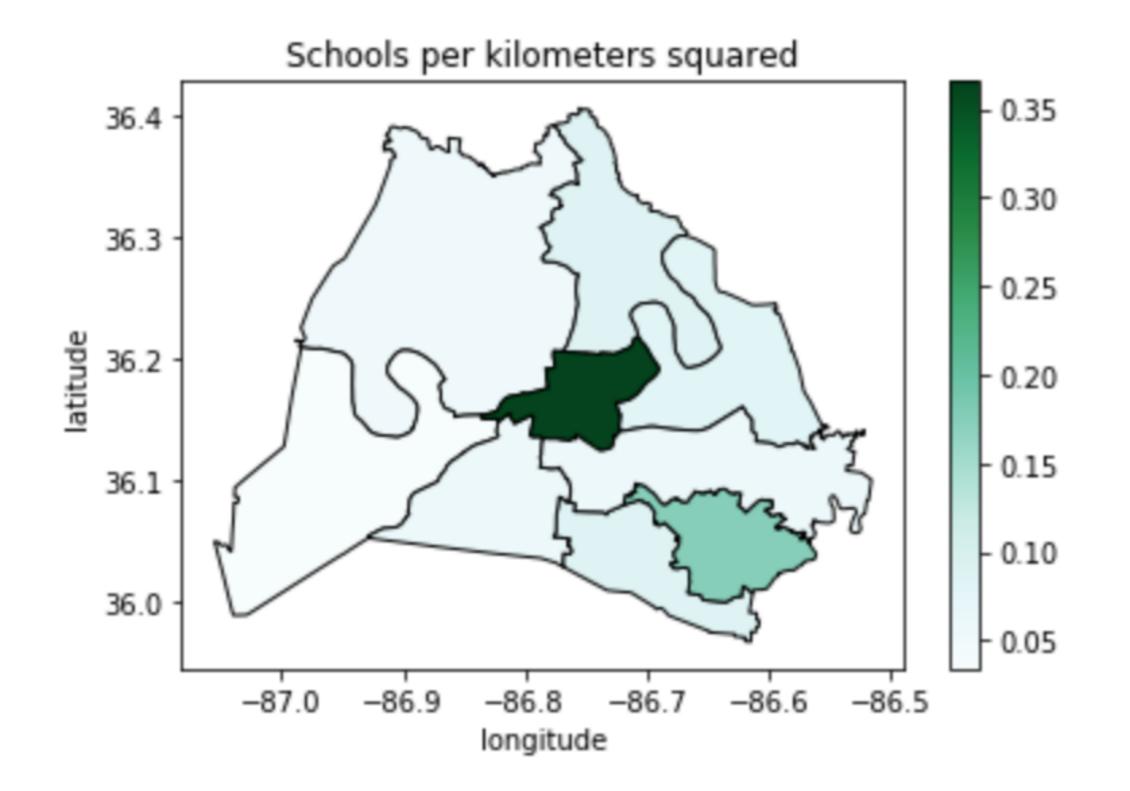
plt.ylabel('latitude')

```
# create school_density
districts_with_counts['school_density'] = districts_with_counts.apply(
    lambda row: row.school_count/row.area, axis = 1)

# plot it
districts_with_counts.plot(column = 'school_density', cmap = 'BuGn',
```

edgecolor = 'black', legend = True)

plt.show();



Let's practice!

VISUALIZING GEOSPATIAL DATA IN PYTHON



Choropleths with folium

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folium.Map choropleth

```
# Construct a map object for Nashville
nashville = [36.1636,-86.7823]
m = folium.Map(location=nashville, zoom_start=10)
# Create a choropleth using the folium Choropleth class
folium.Choropleth(...)
```

Arguments of the folium choropleth

- geo_data the source data for the polygons (geojson file or a GeoDataFrame)
- name the name of the geometry column (or geojson property) for the polygons
- data the source DataFrame or Series for the normalized data
- columns a list of columns: one that corresponds to the polygons and one that has the value to plot

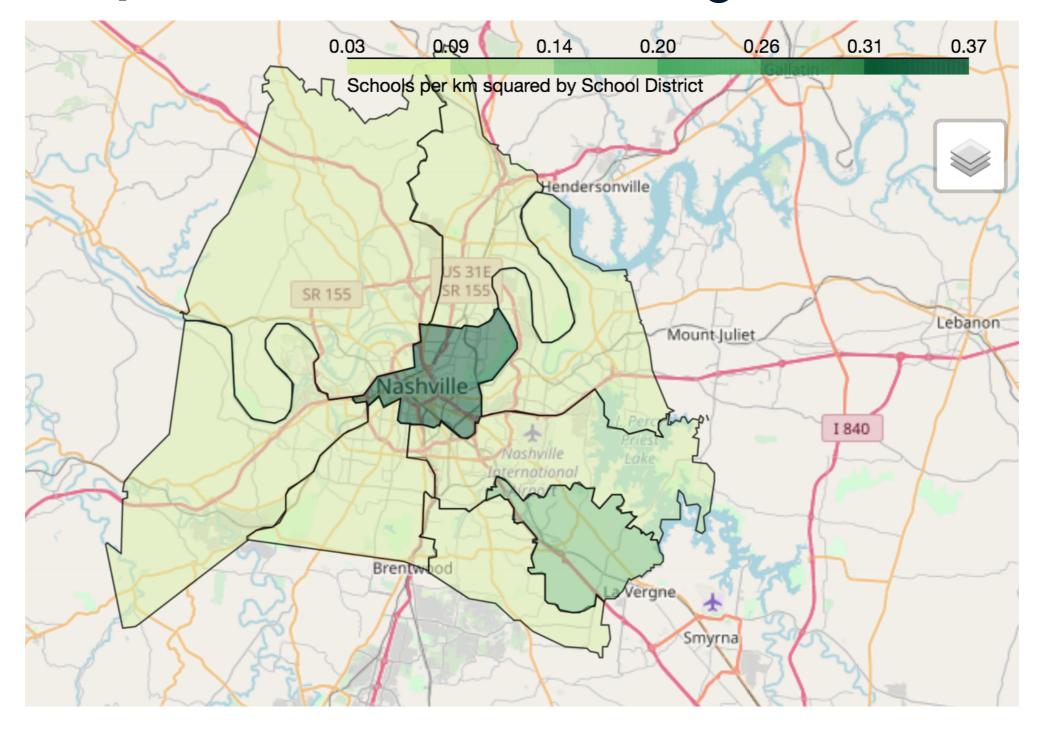
Additional arguments of the folium choropleth

- key_on a GeoJSON variable to bind the data to (always starts with feature)
- fill_color polygon fill color (defaults to blue)
- fill_opacity range between 0 (transparent) and 1 (completely opaque)
- line_color color of polygon border lines (defaults to black)
- line_opacity range between 0 (transparent) and 1 (completely opaque)
- legend_name creates a title for the legend

```
# Center point and map for Nashville
nashville = [36.1636, -86.7823]
m = folium.Map(location=nashville, zoom_start=10)
# Define a choropleth layer for the map
folium.Choropleth(
    geo_data=districts_with_counts,
    name='qeometry',
    data=districts_with_counts,
    columns=['district', 'school_density'],
    key_on='feature.properties.district',
    fill_color='YlGn',
    fill_opacity=0.75,
    line_opacity=0.5,
    legend_name='Schools per km squared by School District'
   ).add_to(m)
folium.LayerControl().add_to_map()
display(m)
```

Q datacamp

Folium choropleth of school density





Let's Practice!

VISUALIZING GEOSPATIAL DATA IN PYTHON



Congratulations!

VISUALIZING GEOSPATIAL DATA IN PYTHON



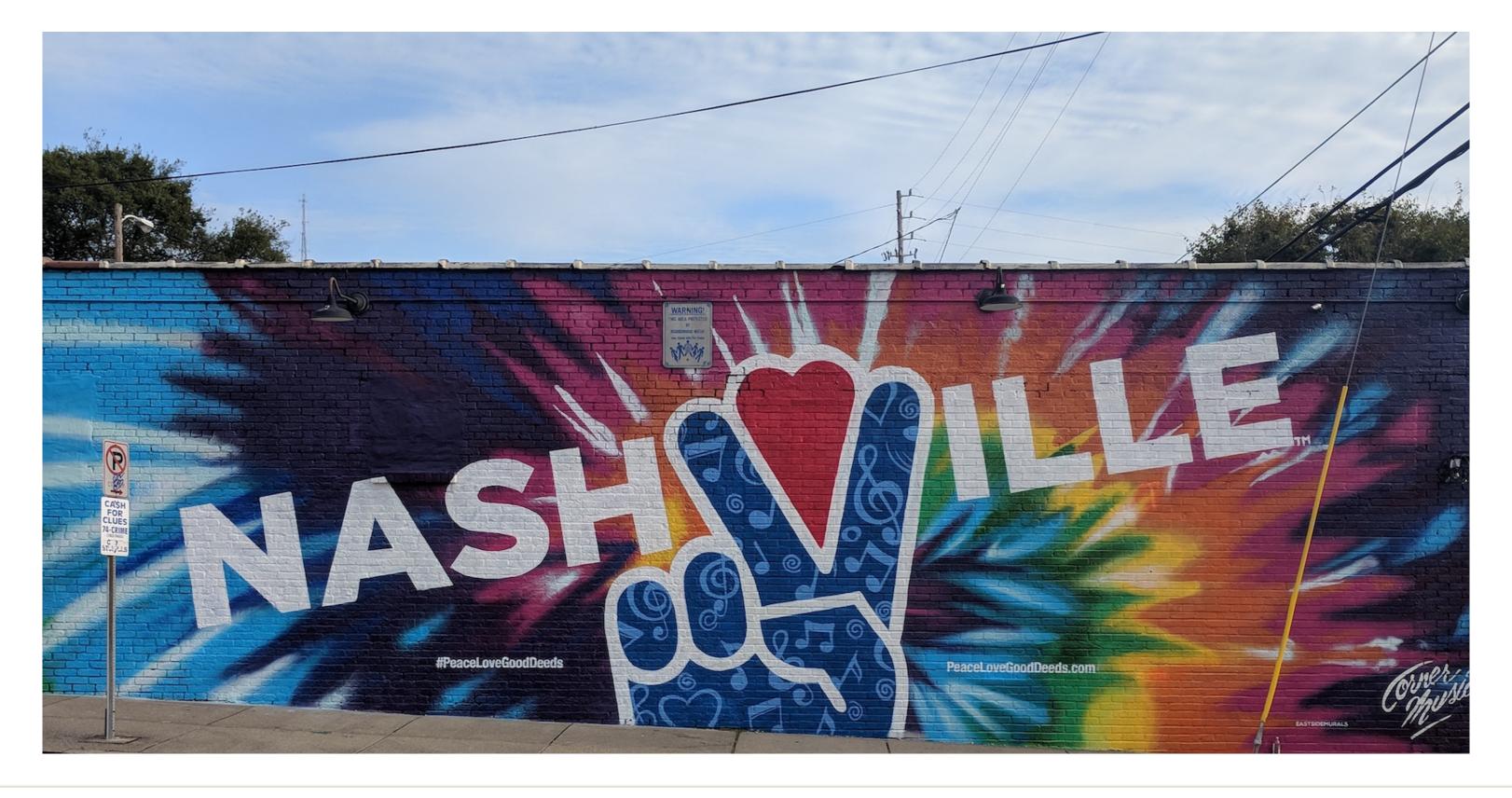
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Skills list

- how to work with shapefiles and GeoJSON
- how to work with geometries
- how to use geopandas, shapely, and folium to extract meaning from geospatial data
- how to create beautiful and informative geospatial visualizations



Goodbye

VISUALIZING GEOSPATIAL DATA IN PYTHON

