

Project4

May 2, 2022

1 Project 4

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1.1 Part 1: Getting Some Data

The dataset that I chose for this project comes from <https://data.baltimorecity.gov/datasets/baltimore::part1-crime-data/>, containing data on “Part 1” crimes within Baltimore, MD dating all the way back to 1963. Part 1 crimes are major crimes including homicide, shooting, robbery, aggravated assault, etc.

```
[134]: import folium
import requests
import pandas as pd

data = pd.read_csv('part1CrimeData.csv', low_memory = False)
data
```

```
[134]:
```

	X	Y	RowID	CrimeDateTime	CrimeCode	\
0	1.424850e+06	569777.328405	1	2022/04/23 04:12:45+00	9S	
1	1.442033e+06	611635.789459	2	2022/04/21 11:34:27+00	9S	
2	1.421723e+06	592309.927233	3	2022/04/21 00:55:03+00	9S	
3	1.408272e+06	586720.112426	4	2022/04/20 16:01:00+00	6D	
4	1.434514e+06	592840.753893	5	2022/04/20 15:29:00+00	6F	
...	
518413	1.400822e+06	604541.096449	518414	1973/07/01 23:00:00+00	2A	
518414	1.409359e+06	598671.099707	518415	1970/06/15 00:01:00+00	2A	
518415	1.415457e+06	616506.081314	518416	1969/07/20 21:00:00+00	2A	
518416	1.394549e+06	593520.411508	518417	1966/01/01 01:00:00+00	2A	
518417	1.396465e+06	604781.500706	518418	1963/10/30 00:00:00+00	2A	

	Location	Description	Inside_Outside	Weapon	Post	\
0	4100 6TH ST	SHOOTING	Outside	FIREARM	913	
1	5900 BELAIR RD	SHOOTING	Outside	FIREARM	425	
2	300 SAINT PAUL ST	SHOOTING	Outside	FIREARM	111	
3	2700 WILKENS AVE	LARCENY FROM AUTO	NaN	NaN	834	

4	3500 E FAIRMOUNT AVE	LARCENY	NaN	NaN	223
...
518413	4000 SPRINGDALE AVE	RAPE	I	OTHER	621
518414	2400 ST STEPHENS CT	RAPE	I	OTHER	731
518415	5400 ROLAND AVE	RAPE	NaN	OTHER	534
518416	900 STAMFORD RD	RAPE	I	OTHER	823
518417	3100 FERNDAL AVE	RAPE	I	OTHER	622

	District	Neighborhood	Latitude	Longitude	\
0	SOUTHERN	BROOKLYN	39.2305	-76.6028	
1	NORTHEAST	CEDMONT	39.3452	-76.5414	
2	CENTRAL	DOWNTOWN	39.2924	-76.6135	
3	SOUTHWEST	MILLHILL	39.2772	-76.6611	
4	SOUTHEAST	BALTIMORE HIGHLANDS	39.2937	-76.5683	
...
518413	NORTHWEST	CENTRAL FOREST PARK	39.3262	-76.6872	
518414	WESTERN	MONDAWMIN	39.3100	-76.6571	
518415	NORTHERN	ROLAND PARK	39.3589	-76.6353	
518416	SOUTHWEST	WEST HILLS	39.2960	-76.7095	
518417	NORTHWEST	HOWARD PARK	39.3269	-76.7026	

	GeoLocation	Premise	VRIName	Total_Incidents
0	(39.2305,-76.6028)	PUBLIC AREA	NaN	1
1	(39.3452,-76.5414)	STREET	NaN	1
2	(39.2924,-76.6135)	HOSPITAL	NaN	1
3	(39.2772,-76.6611)	NaN	NaN	1
4	(39.2937,-76.5683)	NaN	NaN	1
...
518413	(39.3262,-76.6872)	ROW/TOWNHOUSE-OCC	NaN	1
518414	(39.31,-76.6571)	ROW/TOWNHOUSE-OCC	NaN	1
518415	(39.3589,-76.6353)	NaN	NaN	1
518416	(39.296,-76.7095)	ROW/TOWNHOUSE-OCC	NaN	1
518417	(39.3269,-76.7026)	ROW/TOWNHOUSE-OCC	NaN	1

[518418 rows x 18 columns]

```
[135]: print("# of GeoLocation values: ", data[pd.
        ↪notnull(data["GeoLocation"])]["GeoLocation"].count())
print("# of Latitude values: ", data[pd.notnull(data["Latitude"])]["Latitude"].
        ↪count())
print("# of Longitude values: ", data[pd.
        ↪notnull(data["Longitude"])]["Longitude"].count())
```

```
# of GeoLocation values: 518418
# of Latitude values: 517530
# of Longitude values: 517530
```

```
[136]: null_data = data[data['Latitude'].isnull()]
null_data
```

```
[136]:      X    Y  RowID      CrimeDateTime CrimeCode  \
846   NaN NaN    847   2022/04/10 14:00:00+00      2A
5203  NaN NaN   5204   2022/02/21 08:21:00+00      4B
6716  NaN NaN   6717   2022/02/05 11:00:00+00      6D
10574 NaN NaN  10575   2021/12/26 10:49:00+00      6D
16053 NaN NaN  16054   2021/11/06 15:47:00+00     3AJ0
...   ..   ..   ...   ...   ...   ...
517247 NaN NaN  517248  2011/01/09 00:00:00+00      6E
517788 NaN NaN  517789  2011/01/04 11:27:00+00      6G
517816 NaN NaN  517817  2011/01/04 11:27:00+00      6G
517841 NaN NaN  517842  2011/01/04 19:30:00+00      4E
517890 NaN NaN  517891  2011/01/04 19:30:00+00      4E
```

```
      Location      Description Inside_Outside  \
846      3500 Carriage Hill Court      RAPE      NaN
5203  5000 BALTIMORE NATIONAL PIKE  AGG. ASSAULT  Outside
6716      NaN  LARCENY FROM AUTO      NaN
10574      NaN  LARCENY FROM AUTO      NaN
16053      NaN  ROBBERY - CARJACKING      NaN
...   ...   ...   ...   ...
517247      2800 OLD FRANKLIN ST  LARCENY      0
517788      100 VILLAGE SQUARE  LARCENY      I
517816      100 VILLAGE SQUARE  LARCENY      I
517841      1100 MONDAWMIN MA  COMMON ASSAULT  I
517890      1100 MONDAWMIN MA  COMMON ASSAULT  I
```

```
      Weapon Post District Neighborhood Latitude  \
846      NaN NaN NaN NaN NaN
5203  KNIFE_CUTTING_INSTRUMENT NaN NaN NaN NaN
6716      NaN NaN NaN NaN NaN
10574      NaN NaN NaN NaN NaN
16053      OTHER NaN NaN NaN NaN
...   ...   ...   ...   ...
517247      NaN NaN NaN NaN NaN
517788      NaN NaN NaN NaN NaN
517816      NaN NaN NaN NaN NaN
517841      NaN NaN NaN NaN NaN
517890      NaN NaN NaN NaN NaN
```

```
      Longitude GeoLocation      Premise VRIName Total_Incidents
846      NaN      (,)      NaN NaN      1
5203      NaN      (,)  OTHER/RESIDENTIAL NaN      1
6716      NaN      (,)      NaN NaN      1
10574      NaN      (,)      NaN NaN      1
```

16053	NaN	(,)		NaN	NaN	1
...	
517247	NaN	(,)	STREET		NaN	1
517788	NaN	(,)	BARBER/BEAUTY SHOP		NaN	1
517816	NaN	(,)	BARBER/BEAUTY SHOP		NaN	1
517841	NaN	(,)	SHOPPING MALLS/CNTR		NaN	1
517890	NaN	(,)	SHOPPING MALLS/CNTR		NaN	1

[888 rows x 18 columns]

```
[137]: data = data.dropna(subset = ['Latitude', 'Longitude'])
data
```

```
[137]:
```

	X	Y	RowID	CrimeDateTime	CrimeCode	\
0	1.424850e+06	569777.328405	1	2022/04/23 04:12:45+00	9S	
1	1.442033e+06	611635.789459	2	2022/04/21 11:34:27+00	9S	
2	1.421723e+06	592309.927233	3	2022/04/21 00:55:03+00	9S	
3	1.408272e+06	586720.112426	4	2022/04/20 16:01:00+00	6D	
4	1.434514e+06	592840.753893	5	2022/04/20 15:29:00+00	6F	
...	
518413	1.400822e+06	604541.096449	518414	1973/07/01 23:00:00+00	2A	
518414	1.409359e+06	598671.099707	518415	1970/06/15 00:01:00+00	2A	
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518416	1.394549e+06	593520.411508	518417	1966/01/01 01:00:00+00	2A	
518417	1.396465e+06	604781.500706	518418	1963/10/30 00:00:00+00	2A	

	Location	Description	Inside_Outside	Weapon	Post	\
0	4100 6TH ST	SHOOTING	Outside	FIREARM	913	
1	5900 BELAIR RD	SHOOTING	Outside	FIREARM	425	
2	300 SAINT PAUL ST	SHOOTING	Outside	FIREARM	111	
3	2700 WILKENS AVE	LARCENY FROM AUTO	NaN	NaN	834	
4	3500 E FAIRMOUNT AVE	LARCENY	NaN	NaN	223	
...	
518413	4000 SPRINGDALE AVE	RAPE	I	OTHER	621	
518414	2400 ST STEPHENS CT	RAPE	I	OTHER	731	
518415	5400 ROLAND AVE	RAPE	NaN	OTHER	534	
518416	900 STAMFORD RD	RAPE	I	OTHER	823	
518417	3100 FERNDAL AVE	RAPE	I	OTHER	622	

	District	Neighborhood	Latitude	Longitude	\
0	SOUTHERN	BROOKLYN	39.2305	-76.6028	
1	NORTHEAST	CEDMONT	39.3452	-76.5414	
2	CENTRAL	DOWNTOWN	39.2924	-76.6135	
3	SOUTHWEST	MILLHILL	39.2772	-76.6611	
4	SOUTHEAST	BALTIMORE HIGHLANDS	39.2937	-76.5683	
...	
518413	NORTHWEST	CENTRAL FOREST PARK	39.3262	-76.6872	

518414	WESTERN	MONDAWMIN	39.3100	-76.6571
518415	NORTHERN	ROLAND PARK	39.3589	-76.6353
518416	SOUTHWEST	WEST HILLS	39.2960	-76.7095
518417	NORTHWEST	HOWARD PARK	39.3269	-76.7026

	GeoLocation	Premise	VRIName	Total_Incidents
0	(39.2305,-76.6028)	PUBLIC AREA	NaN	1
1	(39.3452,-76.5414)	STREET	NaN	1
2	(39.2924,-76.6135)	HOSPITAL	NaN	1
3	(39.2772,-76.6611)	NaN	NaN	1
4	(39.2937,-76.5683)	NaN	NaN	1
...
518413	(39.3262,-76.6872)	ROW/TOWNHOUSE-OCC	NaN	1
518414	(39.31,-76.6571)	ROW/TOWNHOUSE-OCC	NaN	1
518415	(39.3589,-76.6353)	NaN	NaN	1
518416	(39.296,-76.7095)	ROW/TOWNHOUSE-OCC	NaN	1
518417	(39.3269,-76.7026)	ROW/TOWNHOUSE-OCC	NaN	1

[517530 rows x 18 columns]

I downloaded the .csv file, uploaded it to the Jupyter notebook, and read it using pandas. Just based on the 'GeoLocation' column, it seems like there is a location for every row. However if we look at the 'Latitude' and 'Longitude' columns, we can see there are 88 missing values that are being marked as (,) in the 'GeoLocation' column. It doesn't make sense to impute location's since they are not dependent on the data, so I am removing rows with missing locations.

1.2 Part 2: Making a Map

```
[138]: map_osm = folium.Map(location=[39.29, -76.61], zoom_start=13)
map_osm
```

```
[138]: <folium.folium.Map at 0xffff289aac10>
```

Since my dataset is based in Baltimore, I passed the coordinates of the center of Baltimore into the folium.Map function.

1.3 Part 3

```
[139]: data.columns
```

```
[139]: Index(['X', 'Y', 'RowID', 'CrimeDateTime', 'CrimeCode', 'Location',
        'Description', 'Inside_Outside', 'Weapon', 'Post', 'District',
        'Neighborhood', 'Latitude', 'Longitude', 'GeoLocation', 'Premise',
        'VRIName', 'Total_Incidents'],
        dtype='object')
```

```
[140]: crime_description = data['Description'].unique()
crime_description
```

```
[140]: array(['SHOOTING', 'LARCENY FROM AUTO', 'LARCENY', 'AUTO THEFT',
        'AGG. ASSAULT', 'COMMON ASSAULT', 'ROBBERY - CARJACKING',
        'ROBBERY - COMMERCIAL', 'ROBBERY - STREET', 'ROBBERY - RESIDENCE',
        'HOMICIDE', 'BURGLARY', 'ARSON', 'RAPE'], dtype=object)
```

```
[141]: start_date = '2022/04/17 00:00:00+00'
end_date = '2022/04/23 23:59:59+00'

mask = (data['CrimeDateTime'] >= start_date) & (data['CrimeDateTime'] <=
    ↪end_date)
data_subset = data.loc[mask]
data_subset
```

```
[141]:
```

	X	Y	RowID	CrimeDateTime	CrimeCode	\
0	1.424850e+06	569777.328405	1	2022/04/23 04:12:45+00	9S	
1	1.442033e+06	611635.789459	2	2022/04/21 11:34:27+00	9S	
2	1.421723e+06	592309.927233	3	2022/04/21 00:55:03+00	9S	
3	1.408272e+06	586720.112426	4	2022/04/20 16:01:00+00	6D	
4	1.434514e+06	592840.753893	5	2022/04/20 15:29:00+00	6F	
..	
271	1.409879e+06	581007.669603	272	2022/04/17 02:00:00+00	6D	
272	1.396770e+06	615236.193327	273	2022/04/17 21:30:00+00	3NF	
273	1.441232e+06	590651.552545	274	2022/04/17 22:40:00+00	4E	
274	1.419009e+06	577874.917843	275	2022/04/17 10:14:00+00	4E	
275	1.416162e+06	595856.495217	276	2022/04/17 20:45:39+00	9S	

	Location	Description	Inside_Outside	\
0	4100 6TH ST	SHOOTING	Outside	
1	5900 BELAIR RD	SHOOTING	Outside	
2	300 SAINT PAUL ST	SHOOTING	Outside	
3	2700 WILKENS AVE	LARCENY FROM AUTO	NaN	
4	3500 E FAIRMOUNT AVE	LARCENY	NaN	
..	
271	2800 DELMONT AVE	LARCENY FROM AUTO	NaN	
272	6400 REISTERSTOWN RD	ROBBERY - STREET	NaN	
273	6000 EASTERN AVE	COMMON ASSAULT	NaN	
274	2400 SEABURY RD	COMMON ASSAULT	NaN	
275	1500 PENNSYLVANIA AVE	SHOOTING	Inside	

	Weapon Post	District	Neighborhood	Latitude	\
0	FIREARM 913	SOUTHERN	BROOKLYN	39.2305	
1	FIREARM 425	NORTHEAST	CEDMONT	39.3452	
2	FIREARM 111	CENTRAL	DOWNTOWN	39.2924	
3	NaN 834	SOUTHWEST	MILLHILL	39.2772	

4		NaN	223	SOUTHEAST	BALTIMORE HIGHLANDS	39.2937
..	
271		NaN	831	SOUTHWEST	MORRELL PARK	39.2615
272	AUTOMATIC_HANDGUN		631	NORTHWEST	GLEN	39.3556
273	PERSONAL_WEAPONS		232	SOUTHEAST	BAYVIEW	39.2876
274	PERSONAL_WEAPONS		922	SOUTHERN	CHERRY HILL	39.2528
275	FIREARM		123	CENTRAL	UPTON	39.3022

	Longitude	GeoLocation	Premise	VRIName	Total_Incidents
0	-76.6028	(39.2305,-76.6028)	PUBLIC AREA	NaN	1
1	-76.5414	(39.3452,-76.5414)	STREET	NaN	1
2	-76.6135	(39.2924,-76.6135)	HOSPITAL	NaN	1
3	-76.6611	(39.2772,-76.6611)	NaN	NaN	1
4	-76.5683	(39.2937,-76.5683)	NaN	NaN	1
..
271	-76.6555	(39.2615,-76.6555)	NaN	NaN	1
272	-76.7014	(39.3556,-76.7014)	NaN	NaN	1
273	-76.5446	(39.2876,-76.5446)	NaN	NaN	1
274	-76.6233	(39.2528,-76.6233)	NaN	NaN	1
275	-76.6331	(39.3022,-76.6331)	DWELLING	Central	1

[276 rows x 18 columns]

```
[142]: def find_color(case):
    if case == "SHOOTING": return "red"
    elif case == "LARCENY FROM AUTO": return "blue"
    elif case == "LARCENY": return "blue"
    elif case == "AUTO THEFT": return "blue"
    elif case == "AGG. ASSAULT": return "green"
    elif case == "COMMON ASSAULT": return "green"
    elif case == "ROBBERY - CARJACKING": return "darkpurple"
    elif case == "ROBBERY - COMMERCIAL": return "darkpurple"
    elif case == "ROBBERY - STREET": return "darkpurple"
    elif case == "ROBBERY - RESIDENCE": return "darkpurple"
    elif case == "HOMICIDE": return "lightgray"
    elif case == "BURGLARY": return "orange"
    elif case == "ARSON": return "black"
    elif case == "RAPE": return "pink"

    for index, location_info in data_subset.iterrows():
        folium.Marker(
            [location_info["Latitude"],location_info["Longitude"]],
            popup=location_info["Description"],
            icon = folium.Icon(color = find_color(location_info["Description"]))
        ).add_to(map_osm)
```

```

[143]: def add_legend(folium_map, title, colors, labels):
    if len(colors) != len(labels):
        raise ValueError("colors and labels must have the same length.")

    color_by_label = dict(zip(labels, colors))

    legend_categories = ""
    for label, color in color_by_label.items():
        legend_categories += f"<li><span style='background:{color}'></span>{label}</li>"

    legend_html = f"""
<div id='maplegend' class='maplegend'>
  <div class='legend-title'>{title}</div>
  <div class='legend-scale'>
    <ul class='legend-labels'>
      {legend_categories}
    </ul>
  </div>
</div>
"""

    script = f"""
<script type="text/javascript">
  var oneTimeExecution = (function() {{
    var executed = false;
    return function() {{
      if (!executed) {{
        var checkExist = setInterval(function() {{
          if ((document.
↳getElementsByClassName('leaflet-top leaflet-right').length) || (!executed))↳
↳{{
          document.
↳getElementsByClassName('leaflet-top leaflet-right')[0].style.display = "flex"
          document.
↳getElementsByClassName('leaflet-top leaflet-right')[0].style.flexDirection =↳
↳"column"
          document.
↳getElementsByClassName('leaflet-top leaflet-right')[0].innerHTML +=↳
↳`{legend_html}`;
          clearInterval(checkExist);
          executed = true;
        }}
      }}, 100);
    }}
  }});
  })();
  oneTimeExecution()

```



```

    </script>
    """

    CSS = """

    <style type='text/css'>
        .maplegend {
            z-index:9999;
            float:right;
            background-color: rgba(255, 255, 255, 1);
            border-radius: 5px;
            border: 2px solid #bbb;
            padding: 10px;
            font-size:12px;
            positon: relative;
        }
        .maplegend .legend-title {
            text-align: left;
            margin-bottom: 5px;
            font-weight: bold;
            font-size: 90%;
        }
        .maplegend .legend-scale ul {
            margin: 0;
            margin-bottom: 5px;
            padding: 0;
            float: left;
            list-style: none;
        }
        .maplegend .legend-scale ul li {
            font-size: 80%;
            list-style: none;
            margin-left: 0;
            line-height: 18px;
            margin-bottom: 2px;
        }
        .maplegend ul.legend-labels li span {
            display: block;
            float: left;
            height: 16px;
            width: 30px;
            margin-right: 5px;
            margin-left: 0;
            border: 0px solid #ccc;
        }
        .maplegend .legend-source {

```

```

        font-size: 80%;
        color: #777;
        clear: both;
    }
    .maplegend a {
        color: #777;
    }
</style>
"""

folium_map.get_root().header.add_child(folium.Element(script + css))

return folium_map

```

```

[144]: map_osm = add_legend(map_osm, 'Legend',
                             colors = ['#d73e29', '#37aadd', '#71b025',
↪      '#5b3a6b', '#a3a3a3', '#f2952f', '#000000', '#ff8de7'],
                             labels = ['Shooting', 'Larceny', 'Assault',
↪      'Robbery', 'Homicide', 'Burglary', 'Arson', 'Rape'])
map_osm

```

```

[144]: <folium.folium.Map at 0xffff289aac10>

```

Looking at the columns of the dataset, I decided to categorize the data based on the type of crime. I decided to group crimes into groups since there were so many types. For example, the four types of robbery became one robbery. I then color-coded the groups and made the markers, where clicking on the marker gives the specific type of crime (not just the group). However because there are so many entries in the dataset, I had to limit the markers to just the last week of crime, or else the points took over the map. Then, since folium only has functionality for creating a numerical legend, I used HTML to create a legend just for ease when viewing the map.

This map shows the Part 1 crime that has occurred in Baltimore over the span of the last week. My initial goal with this dataset was to see if different types of crime are more common in different parts of Baltimore, which doesn't seem to be distinguishable from this map (especially when just showing a week's worth of data). However, we are able to see where crime is most concentrated (Central Baltimore) as well as which types of crime are most common (Assault, Larceny). It is interesting to see how relatively little crime occurs in the outskirts of Baltimore compared to the center. Additionally, despite seeing things like shootings on the news in Baltimore constantly, it is nothing compared to the 100s of cases of assault and larceny that occur during the week. Also, the sheer amount of crime that occurs during the week is absurd, and I was surprised that there were 276 cases of crime in just one week. It would be interesting to see how a normal week like this compares to amount of crime in the holiday season or other times of the year, and if the time of year affects the type of crime as well.