

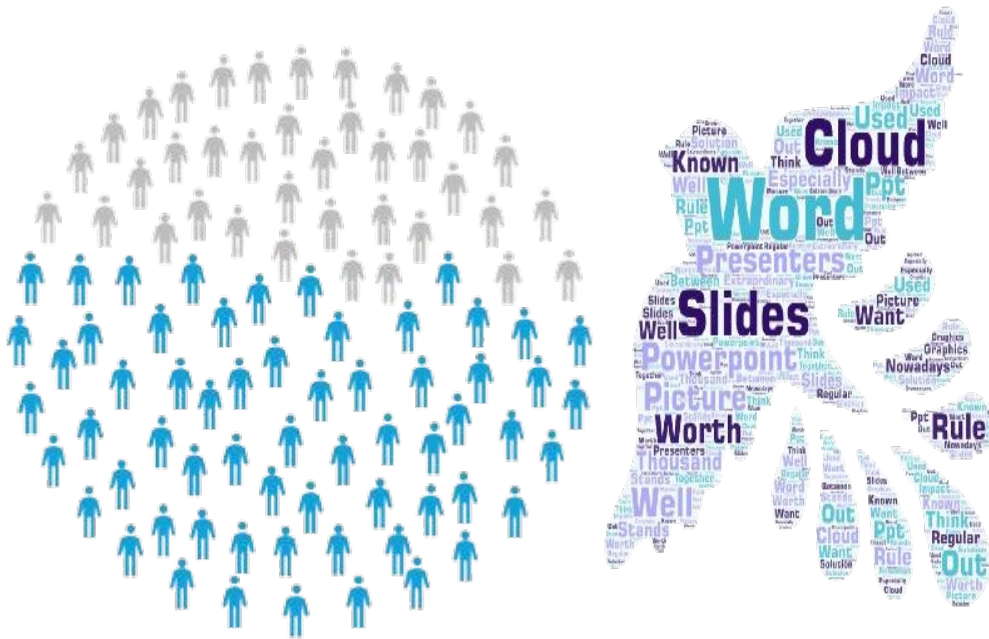
# Group 3

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# Waffle Charts And Word Cloud

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Explore waffle charts with the help of an illustration

---

Identify the use cases of waffle charts

---

Explore word cloud with the help of an illustration

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Identify the use cases of Word Cloud

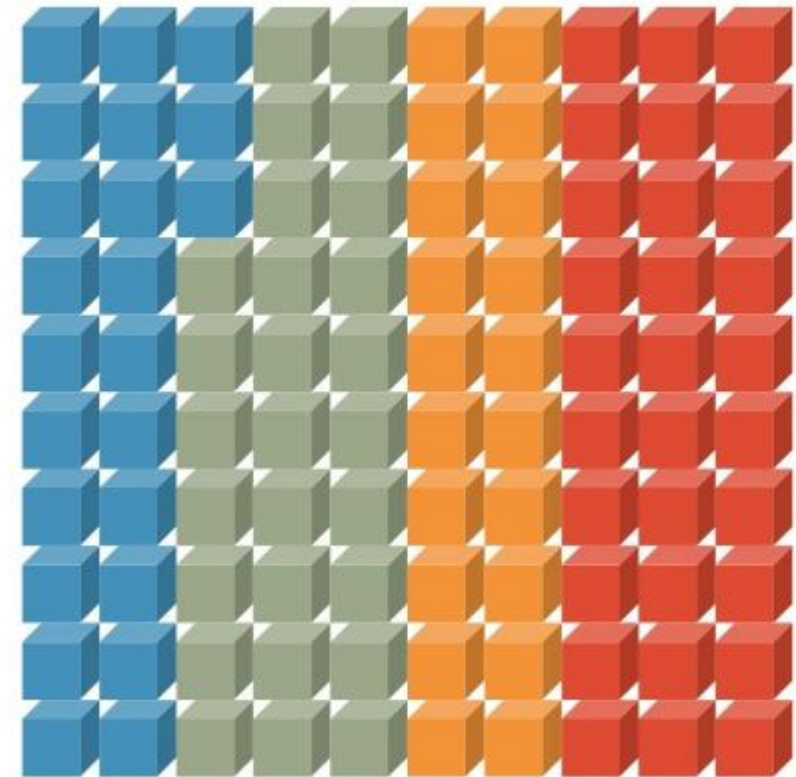
---

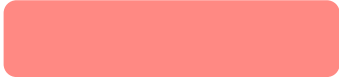
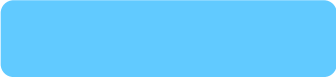


# Understanding Waffle Charts

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- WHAT ARE WAFFLE CHARTS?
- ADVANTAGES AND LIMITATIONS OF WAFFLE CHART



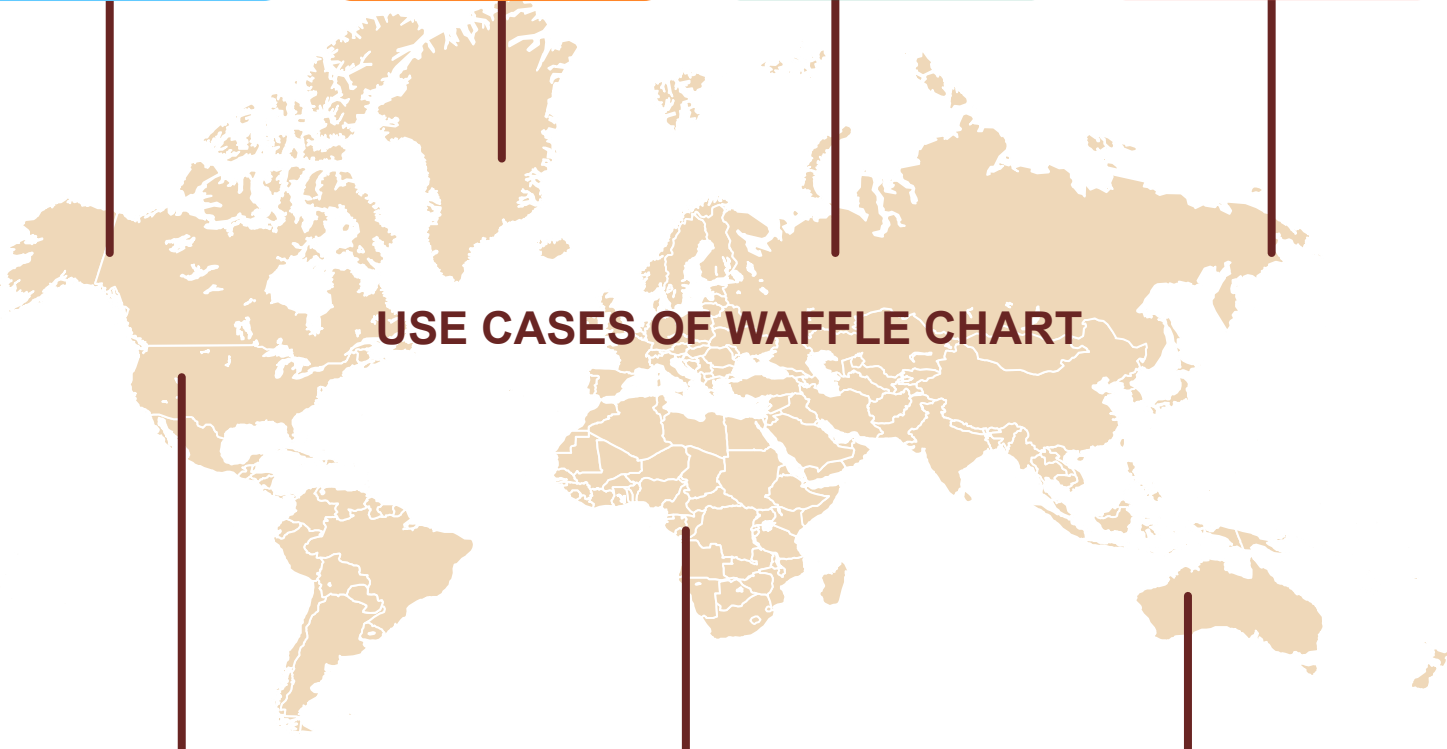


**Market Share Analysis**

**Demographic Representation**

**Project Progress Tracking**

**Budget Allocation**



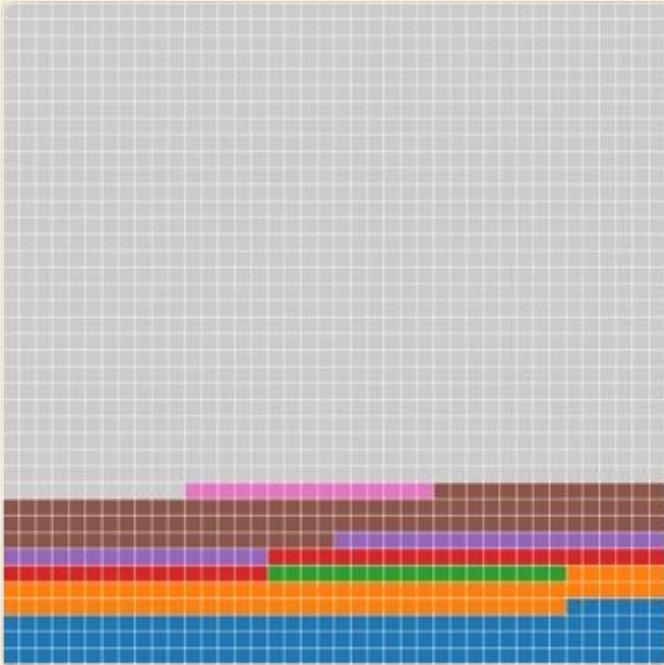
**USE CASES OF WAFFLE CHART**



**Survey Responses**

**Election Results**

**Product Sales Analysis**



# pywaffle LIBRARY

- WHAT IS pywaffle?
- KEY FEATURES OF PyWaffle
- EXAMPLE CODE
- LIMITATIONS

<https://pywaffle.readthedocs.io/en/latest/>

```
import matplotlib.pyplot as plt
from pywaffle import Waffle

# Data
data = {'Category A': 15,
        'Category B': 35,
        'Category C': 50}

# Plot
fig = plt.figure(
    FigureClass=Waffle,
    rows=5,
    columns=10,
    values=data,
    legend={'loc': 'upper left', 'bbox_to_anchor': (1, 1)},
    colors=['#2196f3', '#ff9800', '#4caf50'],
)

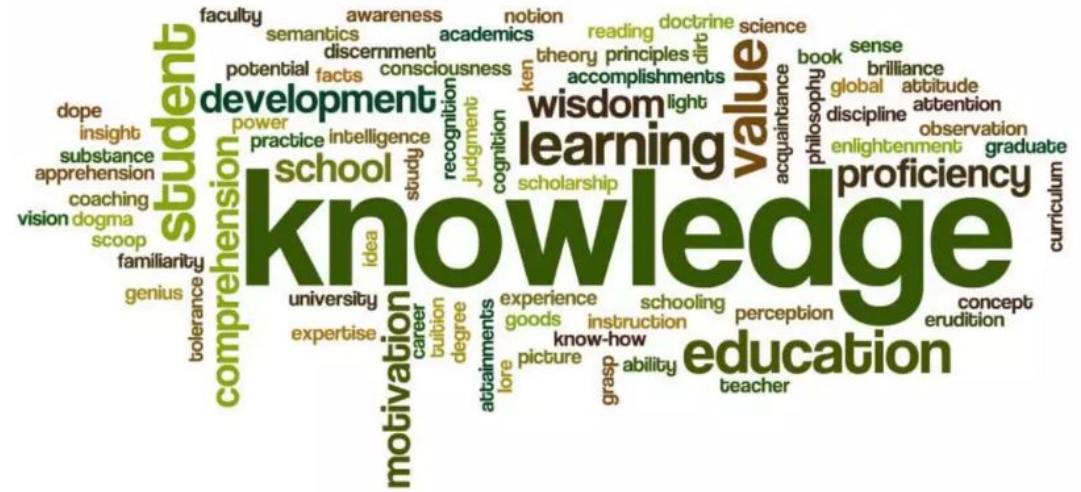
# Add a title
plt.title('Sales Distribution by Category')

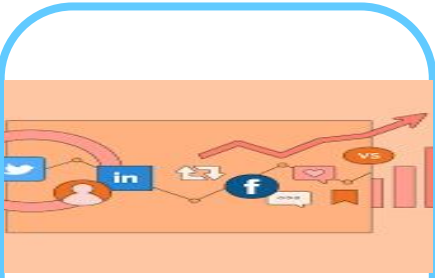
# Show the chart
plt.show()
```

# Understanding Word cloud

---

- WHAT ARE WORD CLOUD?
- ADVANTAGES AND LIMITATIONS OF WORD CLOUD
- USE CASES OF WORD CLOUD





**Social Media  
Analysis**



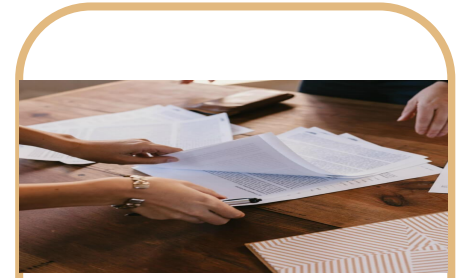
**Customer  
Feedback  
Analysis**



**Content  
Analysis**



**Market  
Research**



**Resume or  
Job  
Description  
Analysis**

# **USE CASES OF WORD CLOUD**

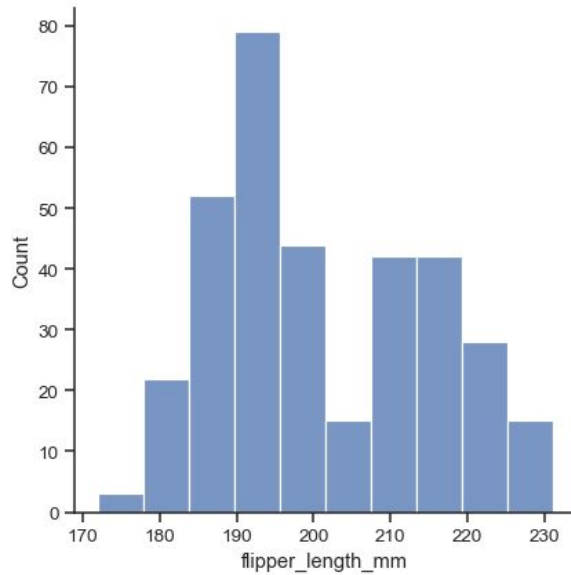


# seaborn

an overview

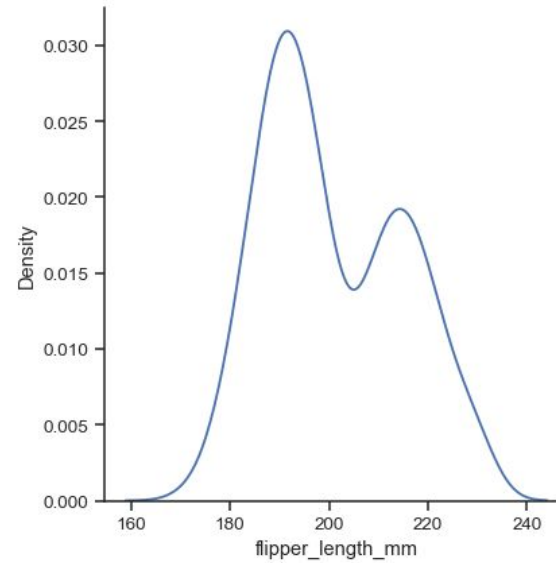


### Histogram



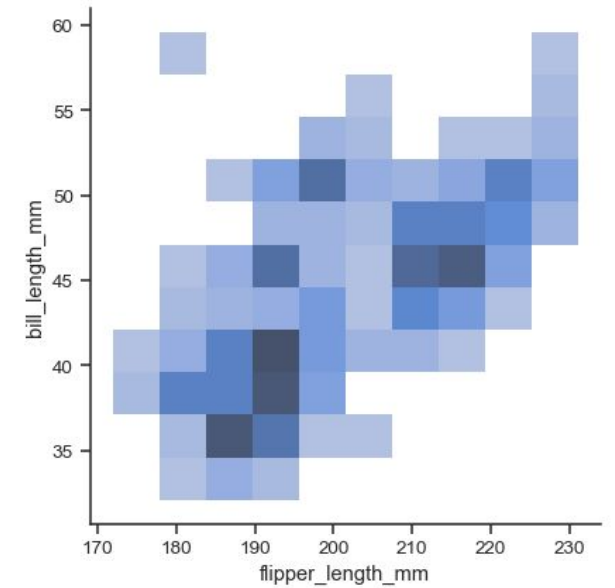
```
sns.displot(  
    data=penguins,  
    x="flipper_length_mm"  
)
```

### Kernel Density Estimation



```
sns.displot(  
    data=penguins,  
    x="flipper_length_mm",  
    kind="kde",  
)
```

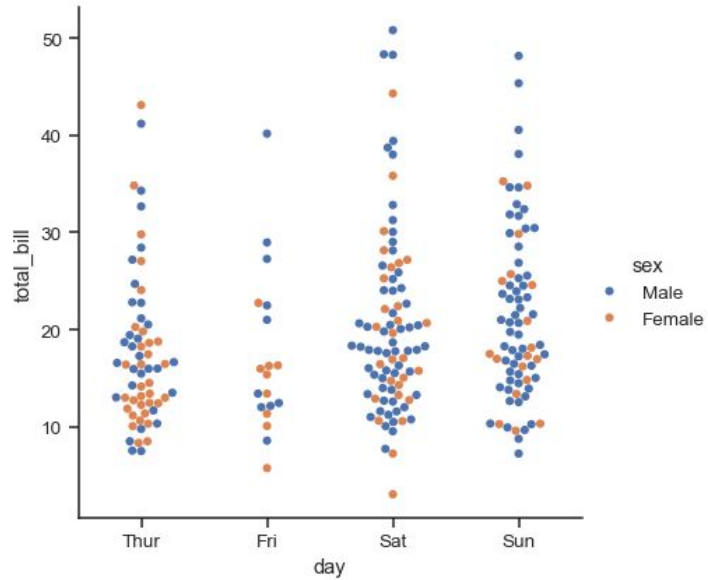
### Bivariate



```
sns.displot(  
    data=penguins,  
    x="flipper_length_mm",  
    y="bill_length_mm",  
    kind="kde"  
)
```

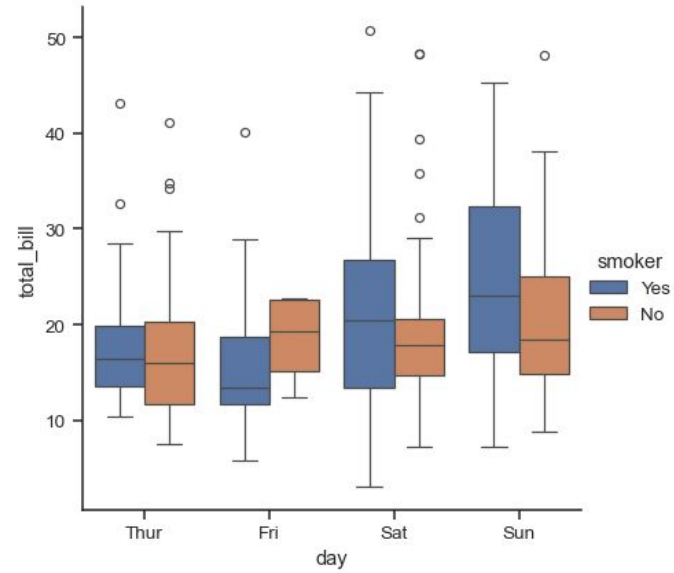
# distribution

## Categorical Scatter Plot



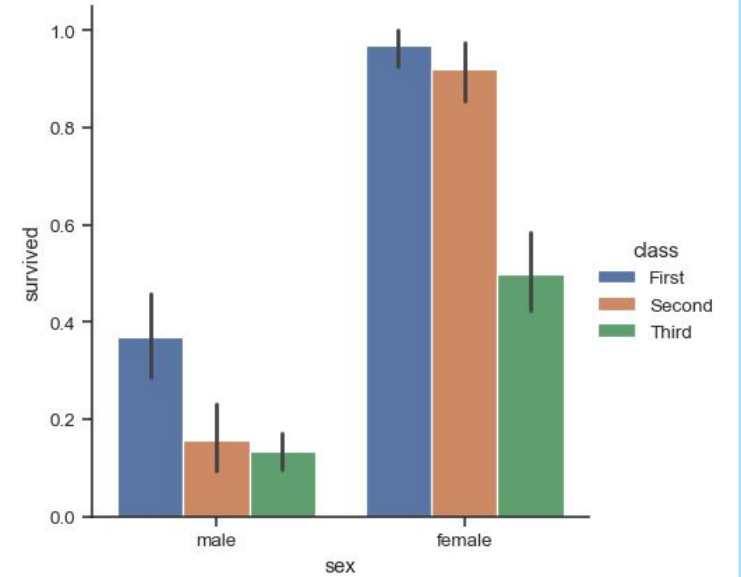
```
sns.catplot(  
    data=tips,  
    x="day",  
    y="total_bill",  
    hue="sex",  
    kind="swarm"  
)
```

## Box Plot



```
sns.catplot(  
    data=tips,  
    x="day",  
    y="total_bill",  
    hue="smoker",  
    kind="box"  
)
```

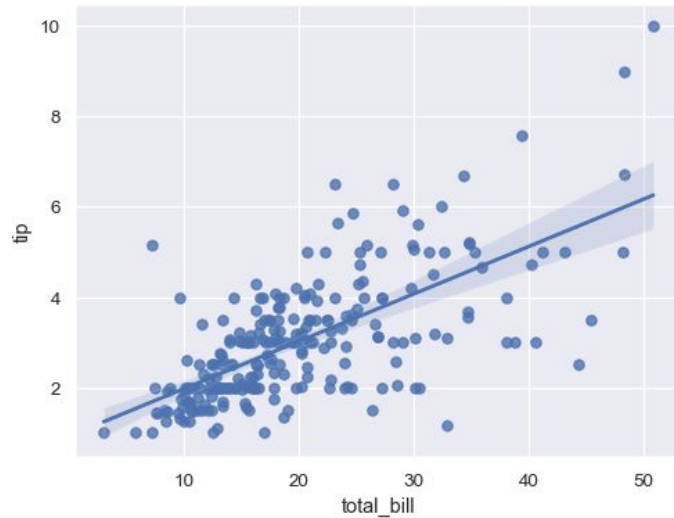
## Bar Plot



```
sns.catplot(  
    data=titanic,  
    x="sex",  
    y="survived",  
    hue="class",  
    kind="bar"  
)
```

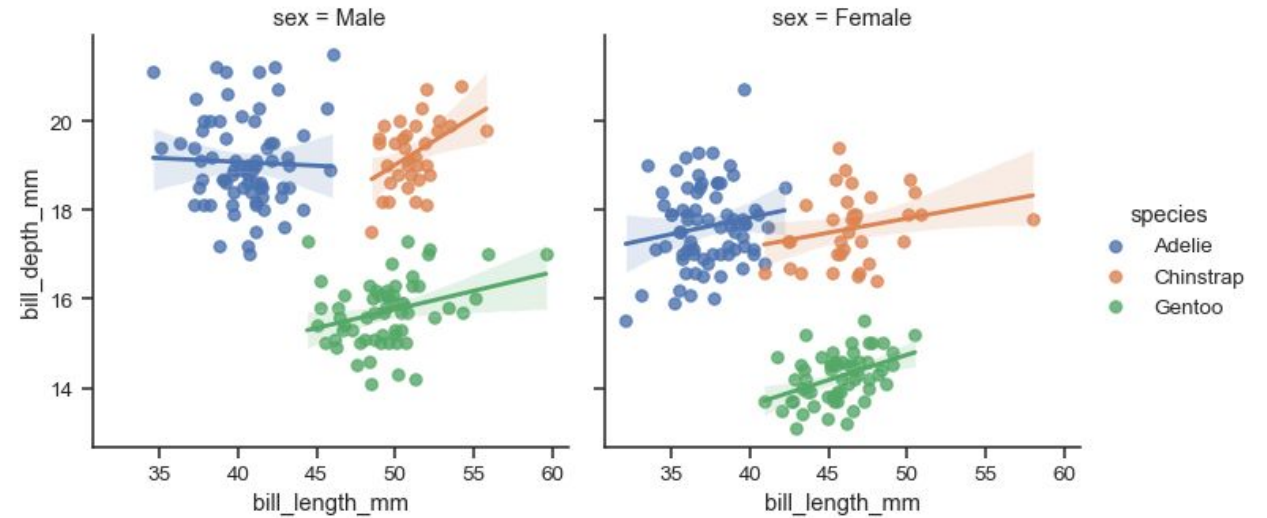
# categorical

## Linear Regression via regplot



```
sns.regplot(  
    data=tips,  
    x="total_bill",  
    y="tip"  
);
```

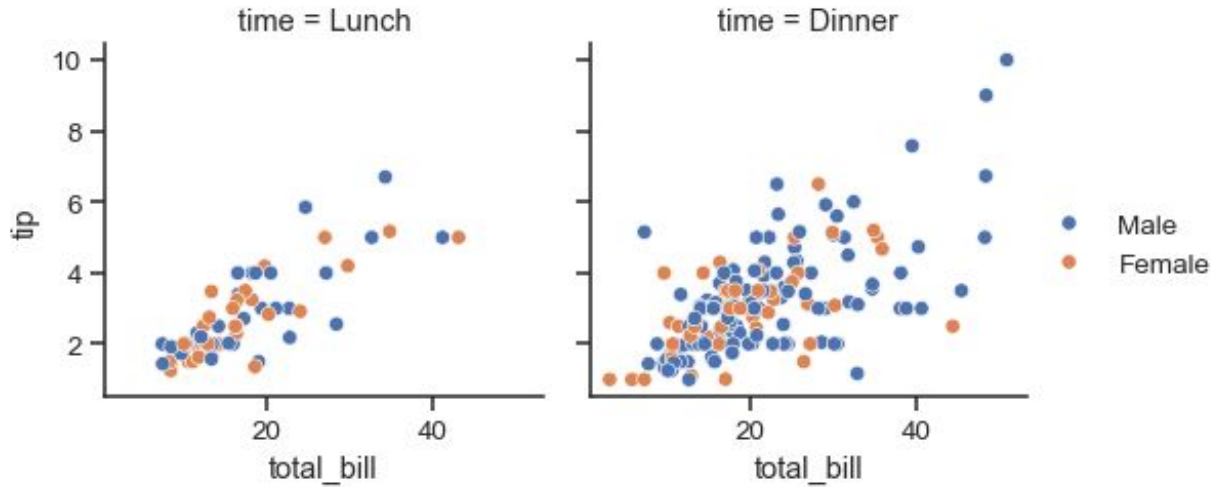
## Regression via lmplot



```
sns.lmplot(  
    data=penguins,  
    x="bill_length_mm",  
    y="bill_depth_mm",  
    hue="species",  
    col="sex",  
    height=4,  
)
```

# regression

# FacetGrid

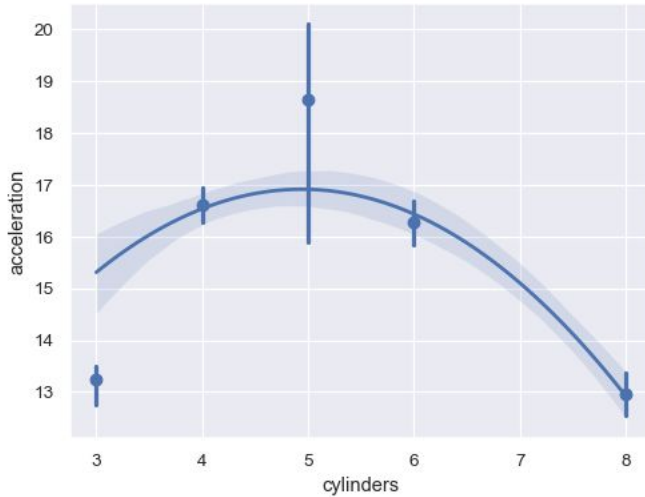


```
g = sns.FacetGrid(tips, col="time")
g.map_dataframe(
    sns.scatterplot,
    x="total_bill",
    y="tip",
    hue="sex"
)
g.add_legend()
```

The aforementioned `lmp1ot` method in the “regression” section is simply a wrapper for a quick way to render a `FacetGrid`.

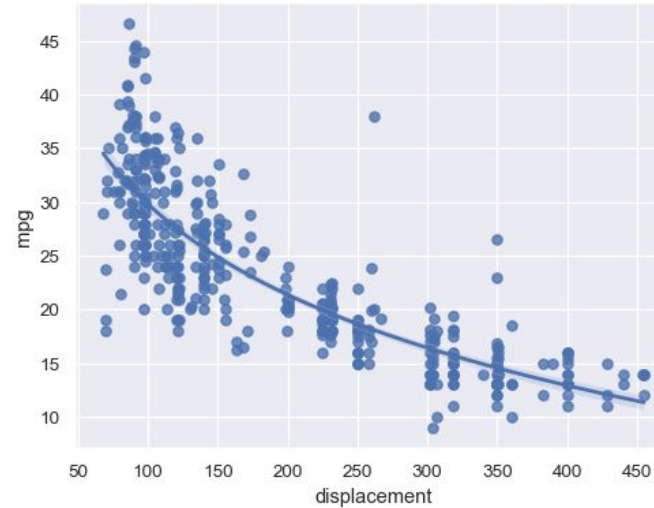


## Polynomial (Binomial)



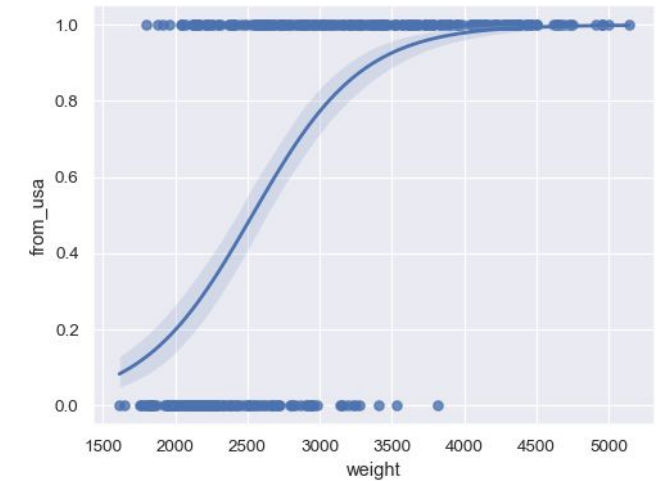
```
sns.regplot(  
    data=mpg,  
    x="cylinders",  
    y="acceleration",  
    x_estimator=np.mean,  
    order=2  
)
```

## Logarithmic



```
sns.regplot(  
    data=mpg,  
    x="displacement",  
    y="mpg",  
    logx=True  
)
```

## Logistic



```
sns.regplot(  
    x=mpg["weight"],  
    y=mpg["origin"]  
        .eq("usa")  
        .rename("from_usa"),  
    logistic=True  
)
```

# nonlinear regression



# Folium

## Folium

## Introduction

## &

## Use Cases

- Powerful Data Visualization Library in Python
- Primarily used to help people visualize geospatial data
- Create a map for any location in the world using Folium
- Parameters : Longitude and Latitude values
- Superimpose Markers and clusters for interesting Visualizations
- Street-Level & Stamen Maps can also be created

# Creating World Map

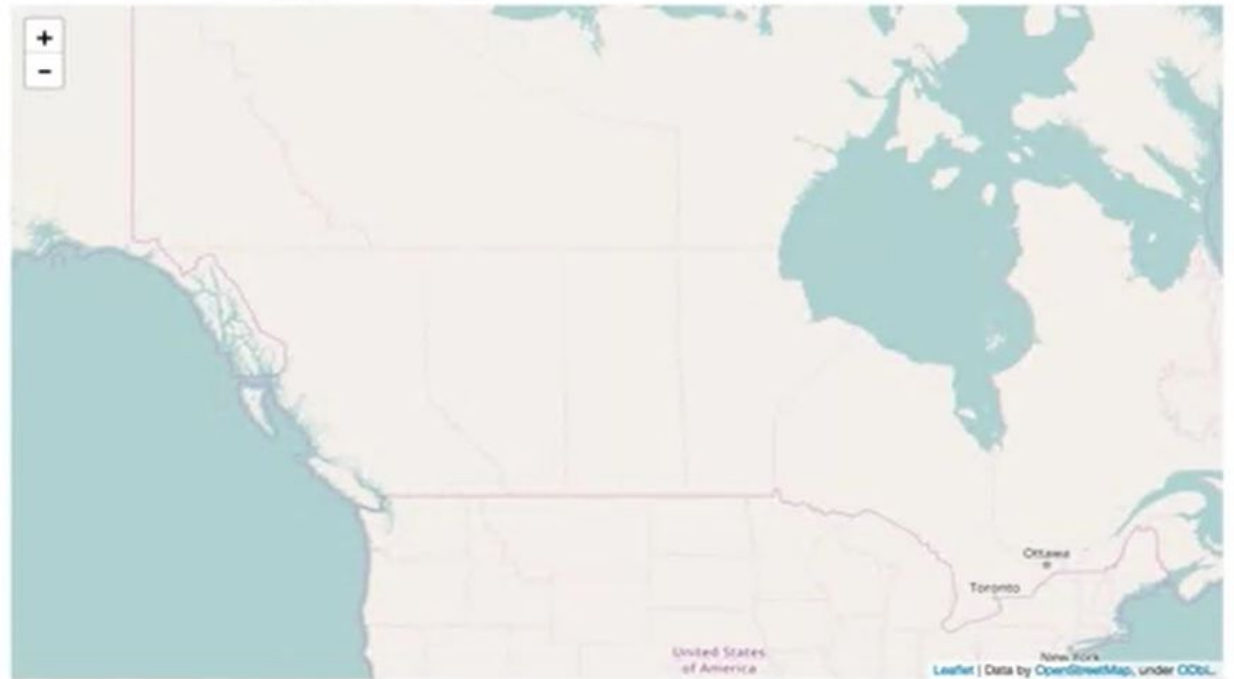
```
#import library  
import folium
```

```
# define the world map  
world_map = folium.Map()  
  
# display world map  
world_map
```



# Creating a map of Canada

```
# define the world map centered around  
# Canada with a low zoom level  
world_map = folium.Map(  
    location=[56.130, -106.35],  
    zoom_start=4  
)  
  
# display world map  
world_map
```





# Map Style - 'Stamen Toner'

This style is great for visualizing and exploring rivers and coastal zones.

```
# create a Stamen Toner map of  
# the world centered around Canada  
world_map = folium.Map(  
    location=[56.130, -106.35],  
    zoom start=4,  
    tiles='Stamen Toner'  
)  
  
# display map  
world_map
```



# Map Style - 'Stamen Terrain'

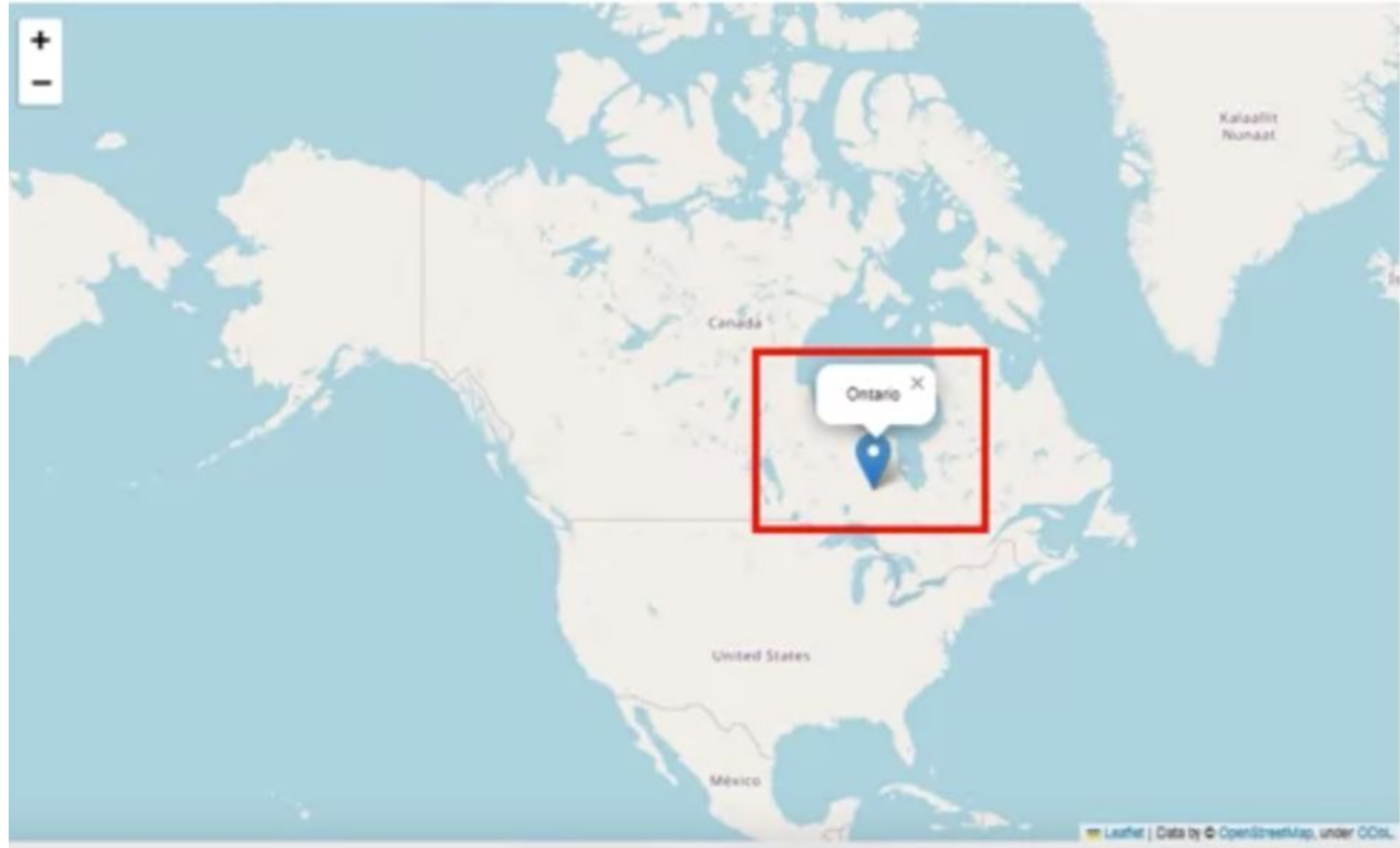
This style is excellent for visualizing hill shading and natural vegetation colors.

```
# create a Stamen Toner map of  
# the world centered around Canada  
world_map = folium.Map(  
    location=[56.130, -106.35],  
    zoom start=4,  
    tiles='Stamen Terrain'  
)  
  
# display map  
world_map
```



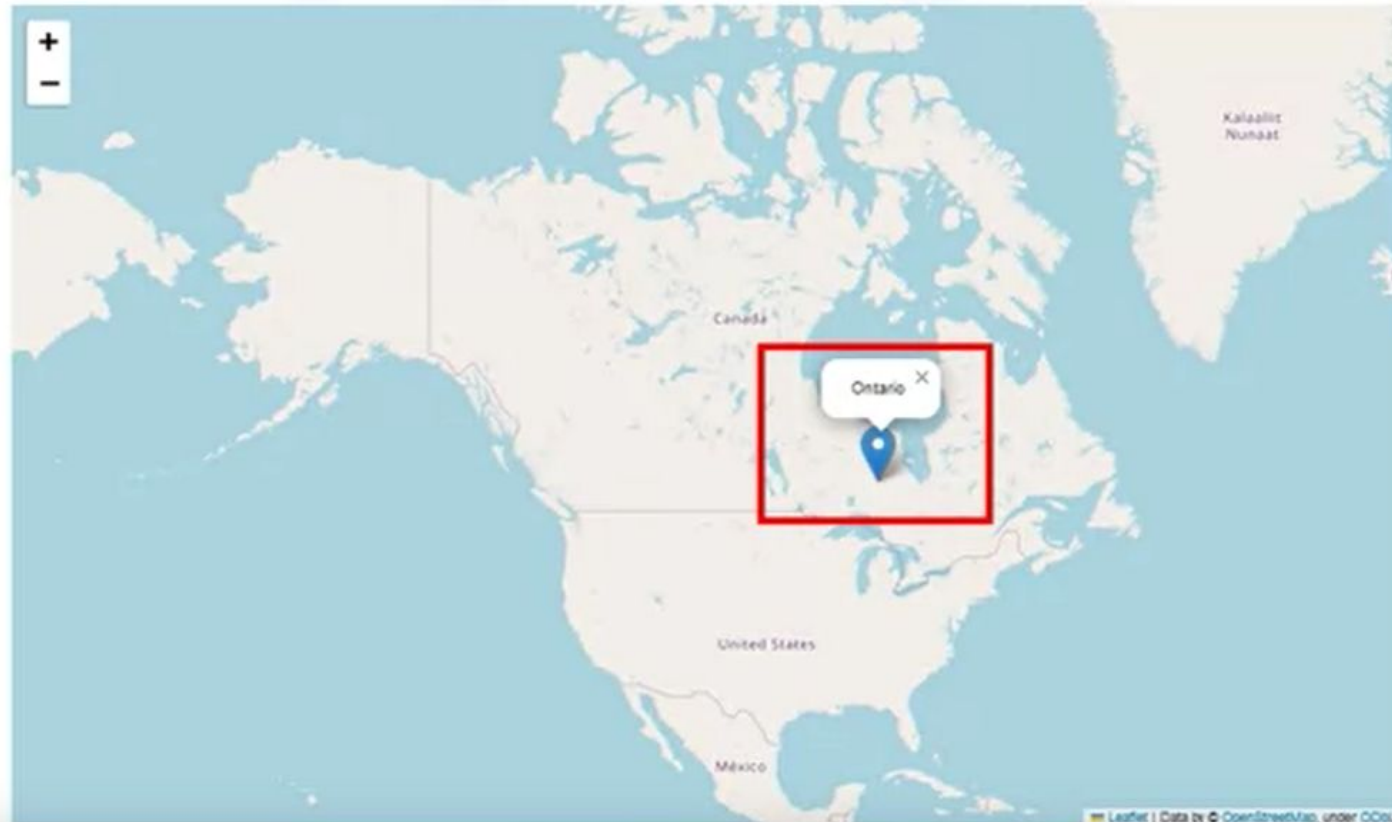
# Adding Markers & Labels to the Map

Markers play a vital role in enhancing interactivity and adding context to maps. They represent specific locations or points of interest, providing additional information when clicked.



# Adding Markers & Labels to the Map

```
# Add a marker for Ontario province  
folium.Marker(location=[51.2538, -85.3232], popup='Ontario').add_to(canada_map)
```





# Adding Markers with Feature Group

```
# generate map of Canada
canada_map = folium.Map(
    location=[56.130, -106.35],
    zoom_start=4
)

## add a red marker to Ontario

# create a feature group
ontario = folium.map.FeatureGroup()

# style the feature group
ontario.add_child(
    folium.features.CircleMarker(
        [51.25, -85.32], radius = 5,
        color = "red", fill_color = "Red"
    )
)

# add the feature group to the map
canada_map.add_child(ontario)

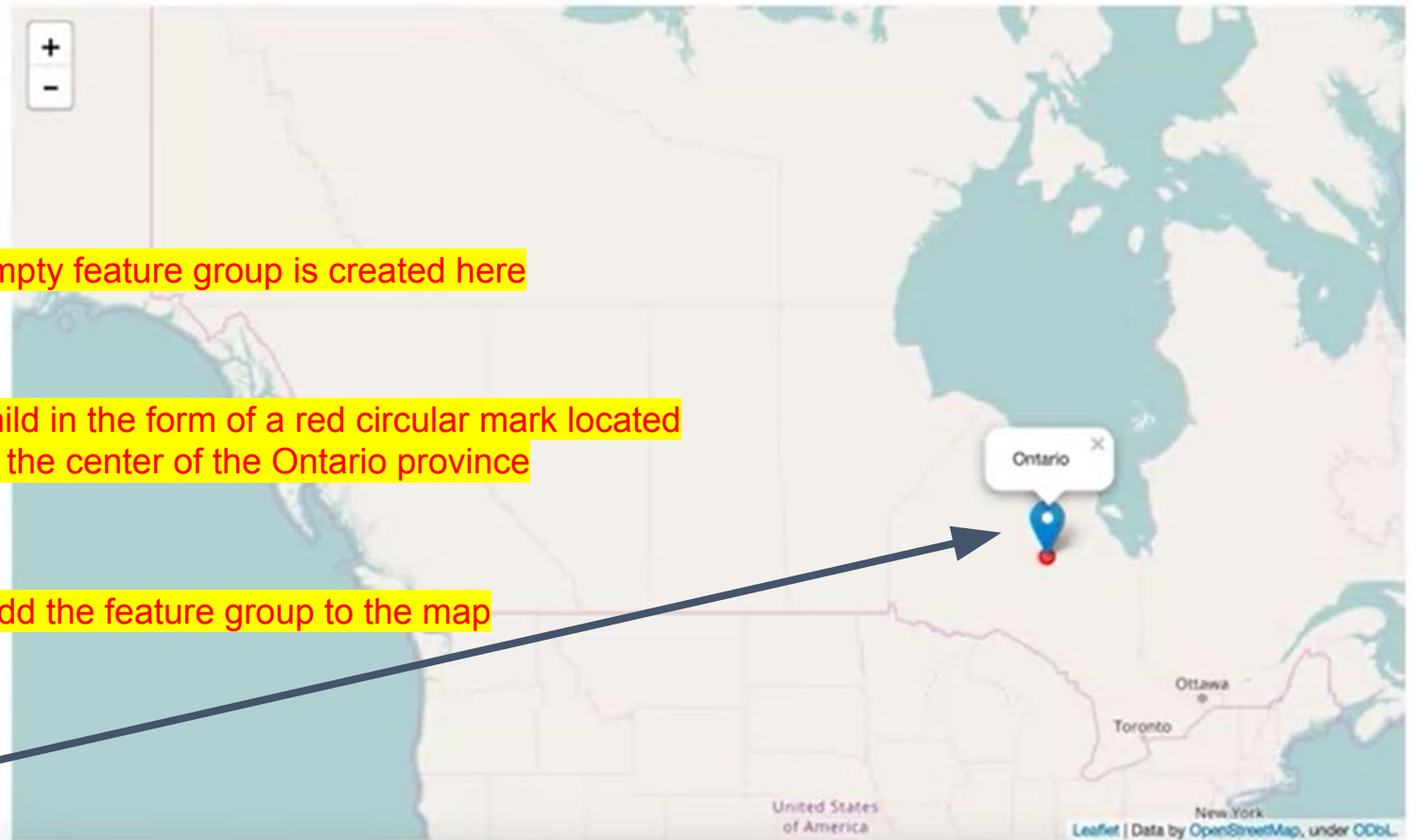
# label the marker
folium.Marker([51.25, -85.32],
    popup='Ontario').add_to(canada_map)

# display map
canada_map
```

empty feature group is created here

child in the form of a red circular mark located at the center of the Ontario province

add the feature group to the map

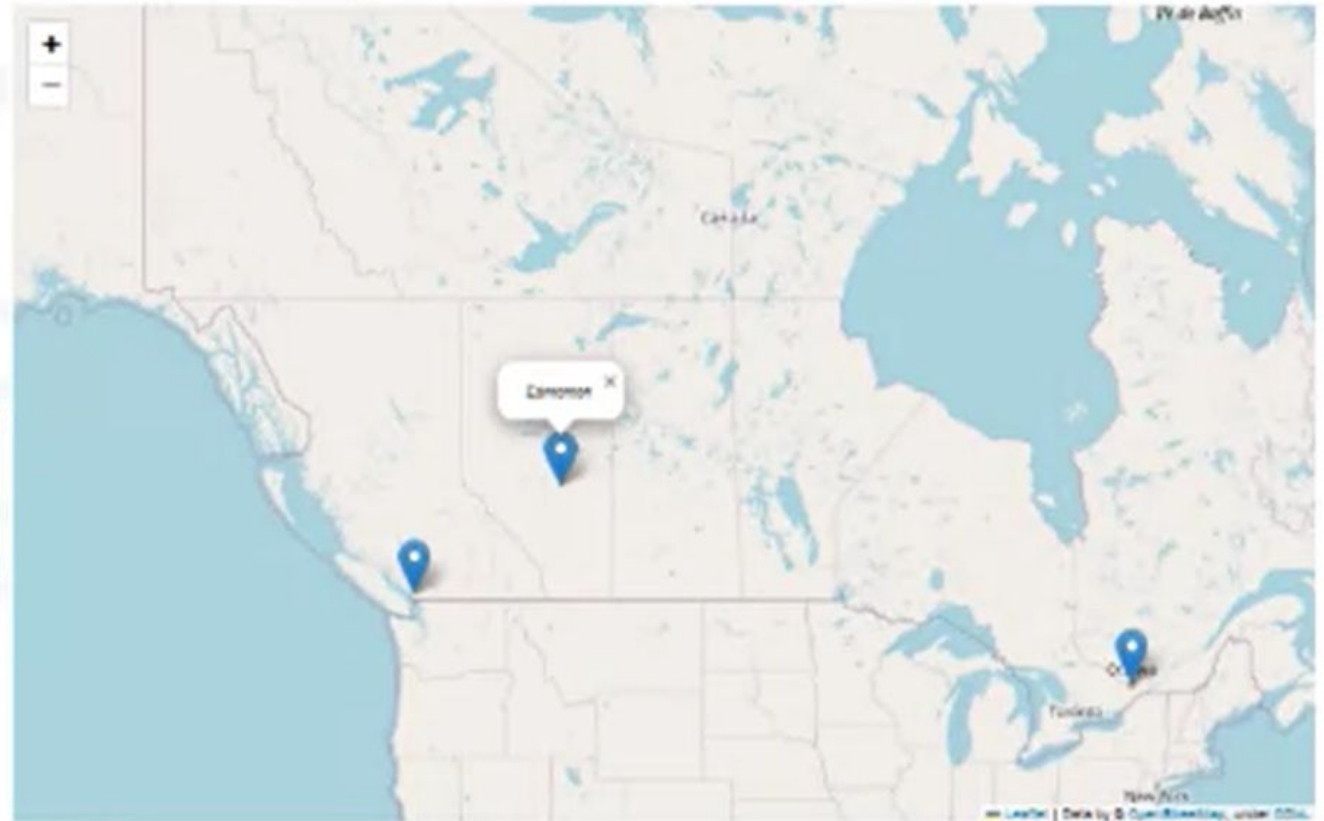


# Adding Multiple Markers to the Map

```
# Define a list of locations and their corresponding popups
locations = [
    {"location": [45.4215, -75.6989], "popup": "Ottawa"},
    {"location": [53.5461, -113.4938], "popup": "Edmonton"},
    {"location": [49.2827, -123.1207], "popup": "Vancouver"},
    # Add more locations and their popups here
]
```

```
# Add markers for each location in the list
for loc in locations:
    folium.Marker(location=loc["location"],
                  popup=loc["popup"]).add_to(map)
```

```
# Display the map with the markers
map
```

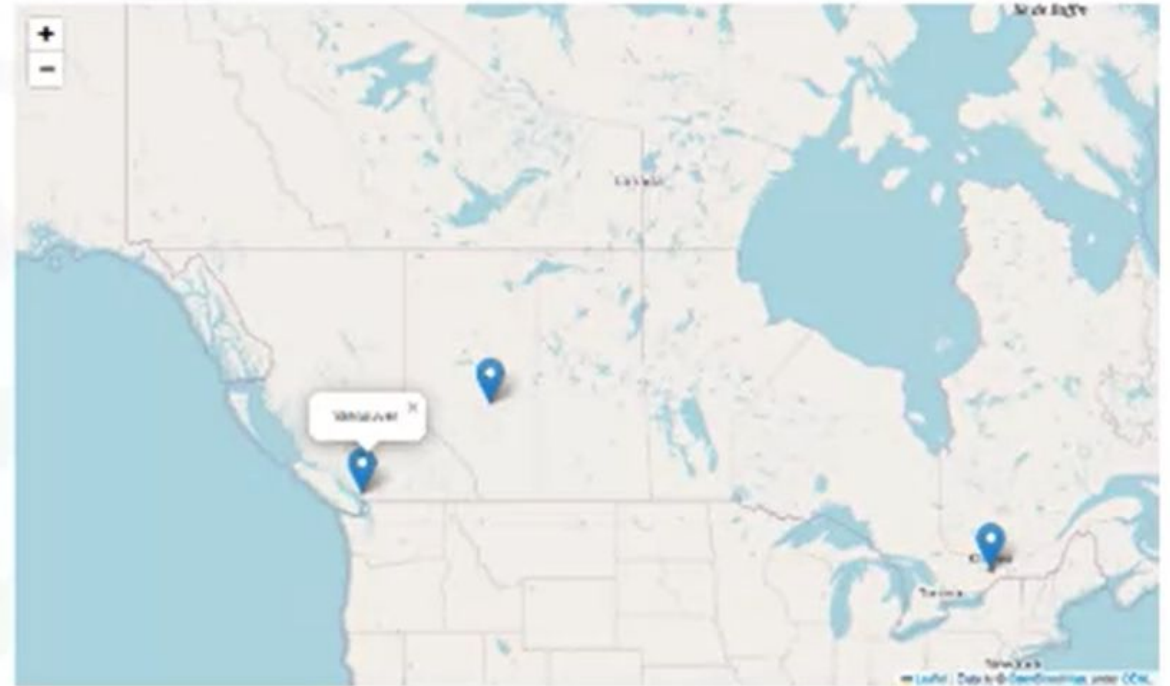


# Clustering Feature

```
#import MarkerCluster
from folium.plugins import MarkerCluster

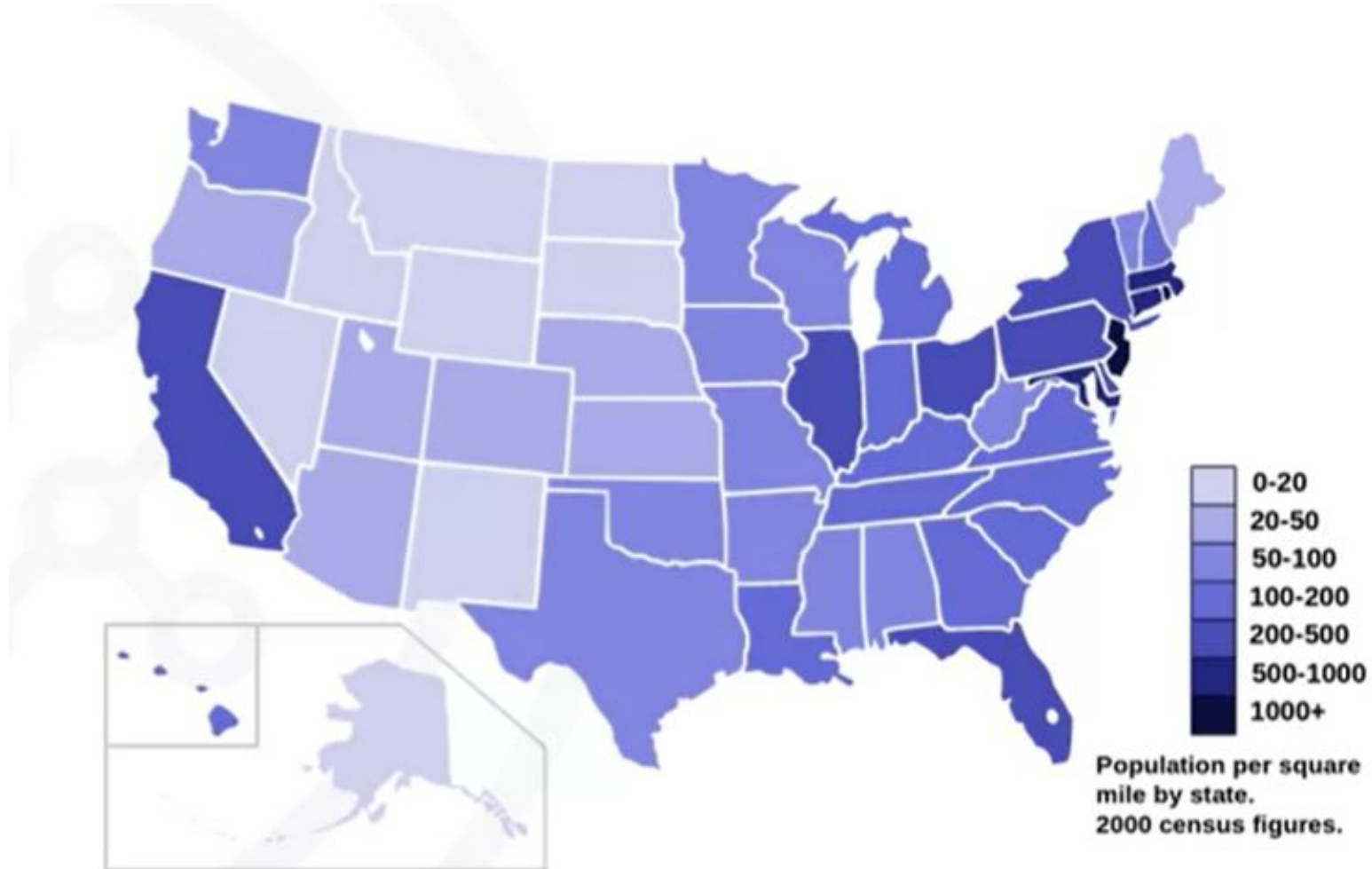
# Create a MarkerCluster object
marker_cluster = MarkerCluster().add_to(map)

# Add markers for each location in the list to the MarkerCluster
for loc in locations:
    folium.Marker(location=loc["location"],
                  popup=loc["popup"]).add_to(marker_cluster)
```



This clustering feature enhances the visual presentation by preventing overcrowding and ensuring a clear representation, primarily when numerous markers are close.

# Choropleth maps





# Choropleth maps

```
df_can = pd.read_csv('https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBMDeveloperSkillsNetwork-DV0101EN-SkillsNetwork/Data%20Files/C  
df_can
```

	Country	Continent	Region	DevName	1980	1981	1982	1983	1984	1985	...	2005	2006	2007	2008	2009	2010	2011	2012	2013	Total
0	Afghanistan	Asia	Southern Asia	Developing regions	16	39	39	47	71	340	...	3436	3009	2652	2111	1746	1758	2203	2635	2004	58639
1	Albania	Europe	Southern Europe	Developed regions	1	0	0	0	0	0	...	1223	856	702	560	716	561	539	620	603	15699
2	Algeria	Africa	Northern Africa	Developing regions	80	67	71	69	63	44	...	3626	4807	3623	4005	5393	4752	4325	3774	4331	69439
3	American Samoa	Oceania	Polynesia	Developing regions	0	1	0	0	0	0	...	0	1	0	0	0	0	0	0	0	6
4	Andorra	Europe	Southern Europe	Developed regions	0	0	0	0	0	0	...	0	1	1	0	0	0	0	1	1	15

# Choropleth maps

```
# create a plain world map
world_map = folium.Map(
    zoom_start=2,
    tiles='Mapbox Bright'
)

## geojson file
world_geo = r'world_countries.json'
```



# GeoJSON

```
{
  "type": "FeatureCollection",
  "features": [
    {
      "type": "Feature",
      "properties": {"name": "Brunei"},
      "geometry": {
        "type": "Polygon",
        "coordinates": [
          [
            [114.204017, 4.525874],
            [114.599961, 4.900011],
            [115.45071, 5.44773],
            [115.4057, 4.955228],
            [115.347461, 4.316636],
            [114.869557, 4.348314],
            [114.659596, 4.007637],
            [114.204017, 4.525874]
          ]
        ]
      },
      "id": "BRN"
    }
  ]
}
```



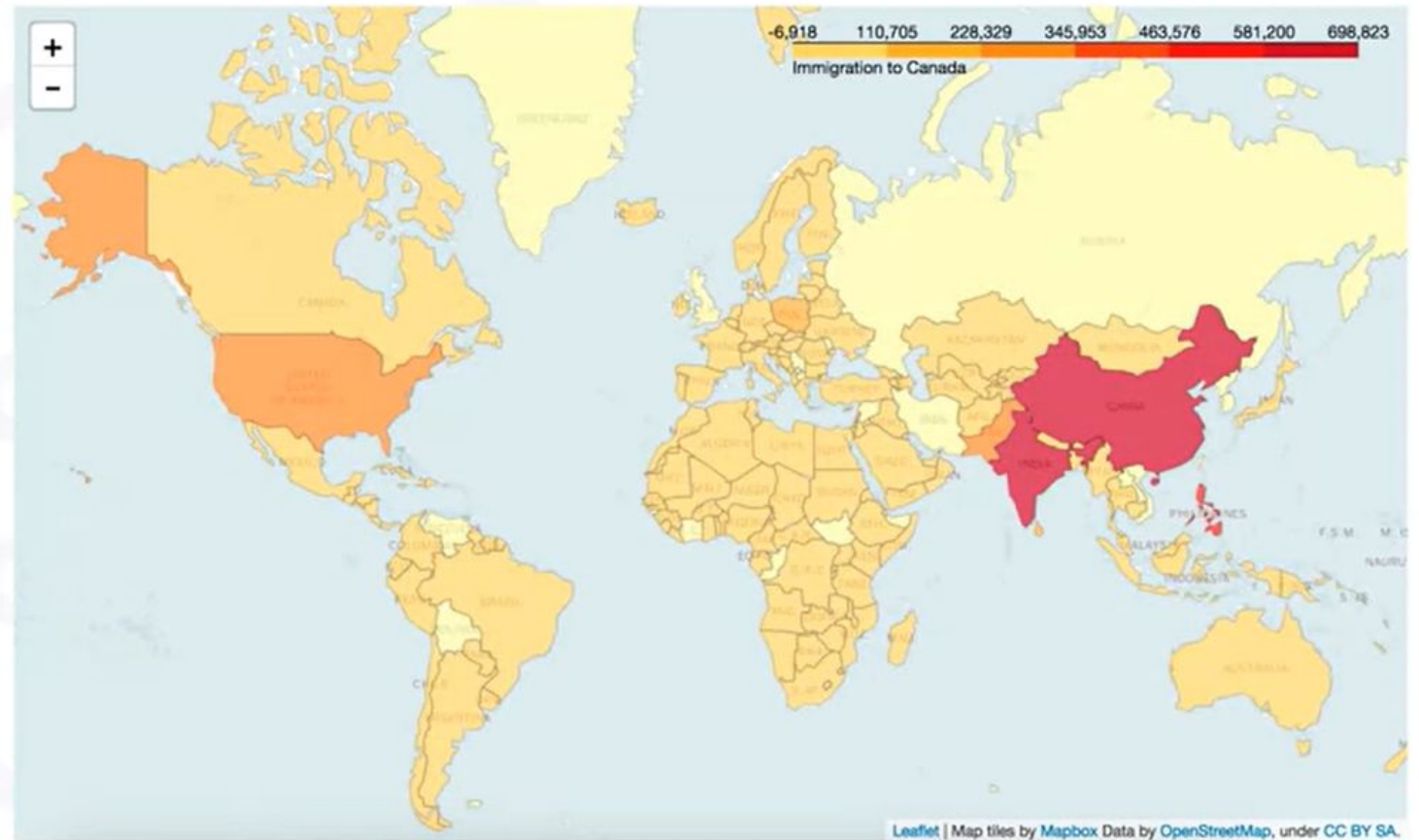
# Choropleth maps

```
# create a plain world map
world_map = folium.Map(
    zoom_start=2,
    tiles='Mapbox Bright'
)

## geojson file
world_geo = r'world_countries.json'

# generate choropleth map using the total
# population of each country to Canada from
# 1980 to 2013
world_map.choropleth(
    geo_path=world_geo,
    data=df_canada,
    columns=['Country', 'Total'],
    key_on='feature.properties.name',
    fill_color='YlOrRd',
    legend_name='Immigration to Canada'
)

# display map
world_map
```



**Thank You**