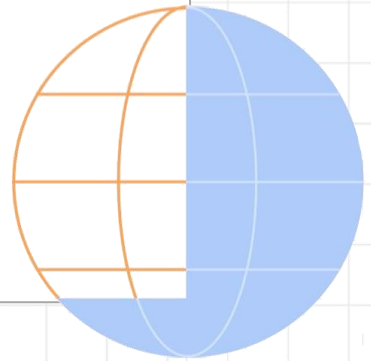
Day 02

Data Science with Python

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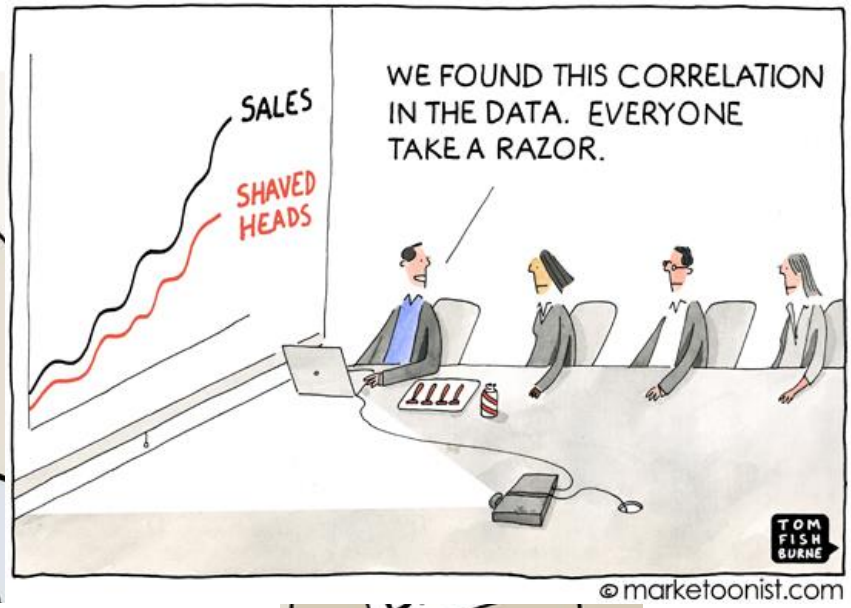
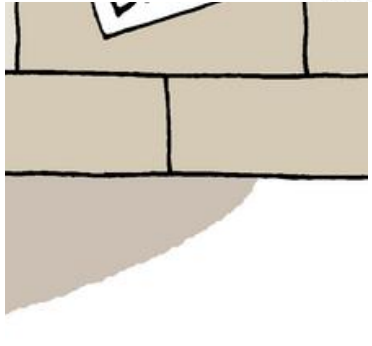
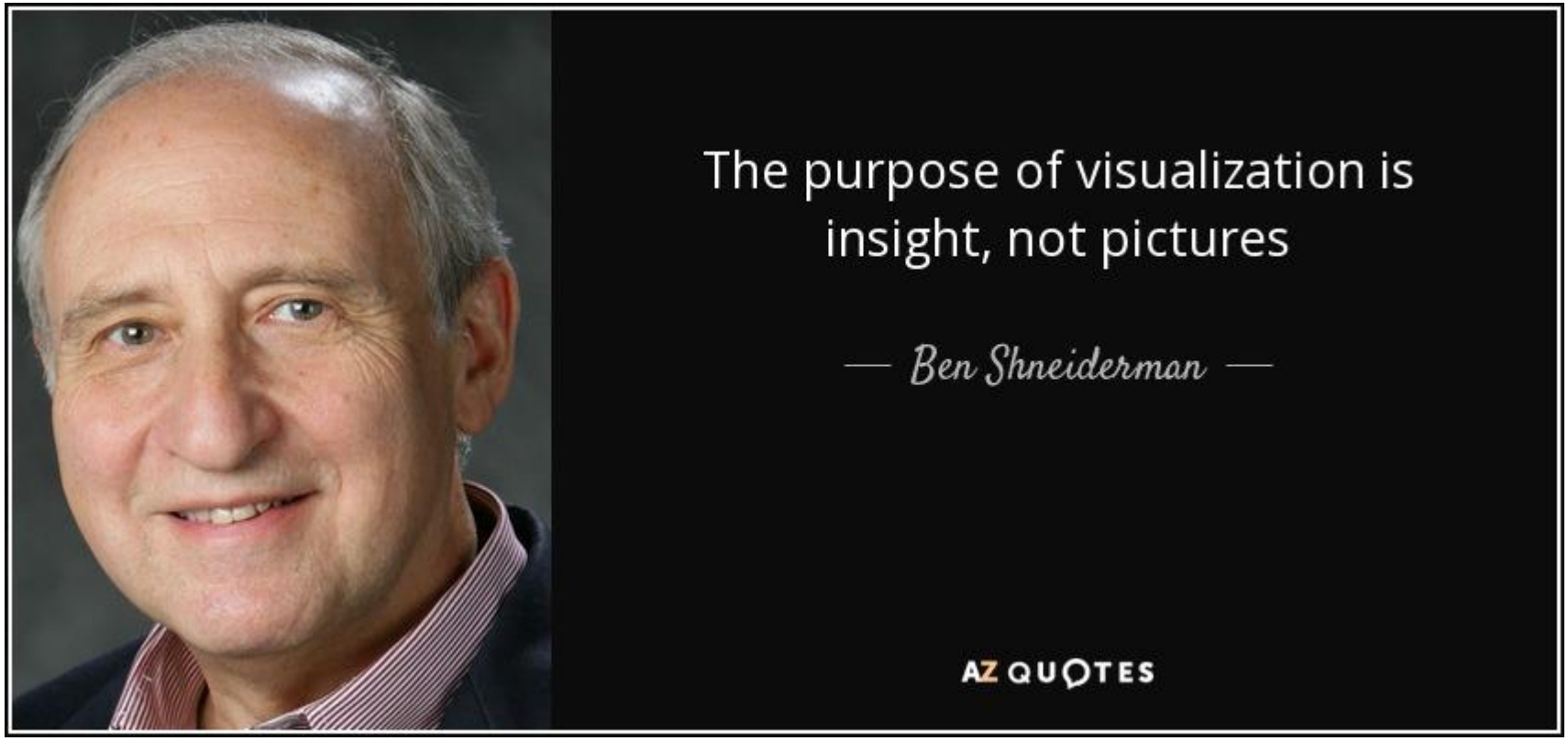

SALES
SHAVED HEADS
WE FOUND THIS CORRELATION IN THE DATA. EVERYONE TAKE A RAZOR.
TOM FISH BURNE
©marketoonist.com




The purpose of visualization is insight, not pictures
— Ben Shneiderman —
AZ QUOTES

Data

- Collection of information gathered by observations, measurements, research, or analysis.
- It may comprise facts, figures, numbers, names, or even general descriptions of things.
- Can be organized in the form of graphs, charts, or tables for ease in our study.

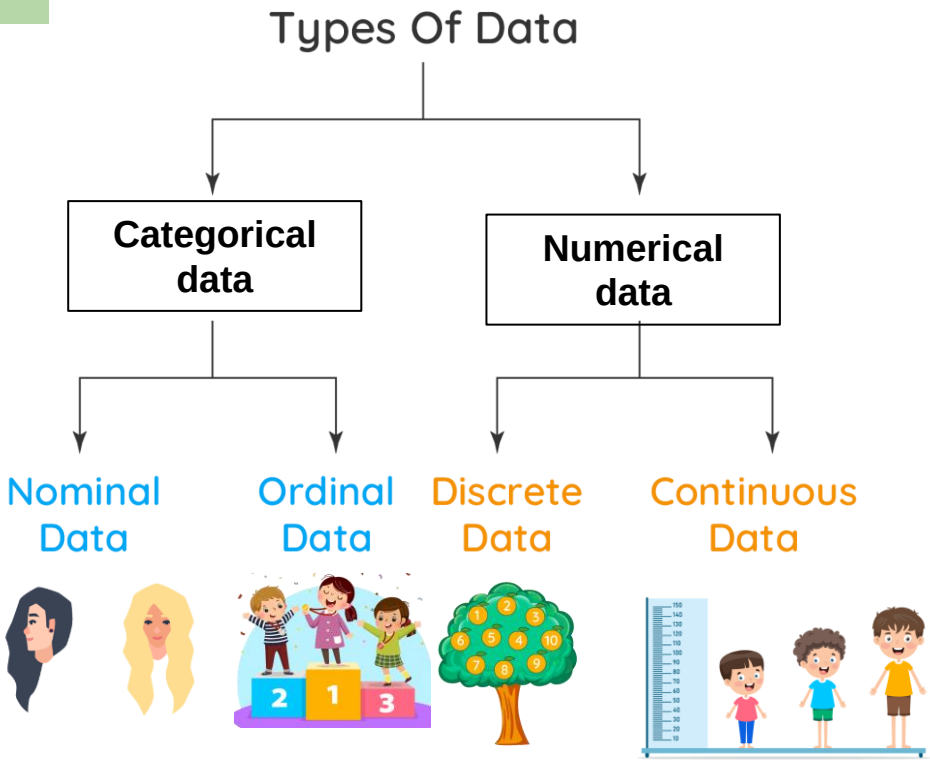
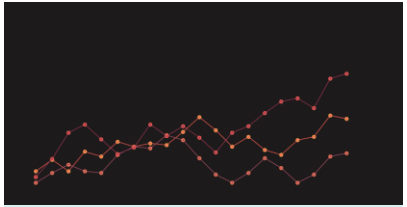


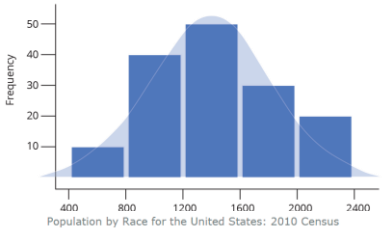
Chart types

- **Line Chart:** showing *trends or patterns over time or continuous data*.
- Ex: *Stock prices, temperature variations, sales trends.*



- **Bar Chart**: comparing *categorical data or discrete values*.
- Ex: *Comparison of sales by product category, population by country, survey results*.

- **Histogram:** visualizing *distribution of continuous or discrete data*.
- Ex: *Age distribution, exam scores distribution, frequency of occurrence.*



- **Pie Chart**: representing *parts of a whole or proportions.*
- Ex: *Market share of different products, composition of a budget, demographic distribution*.

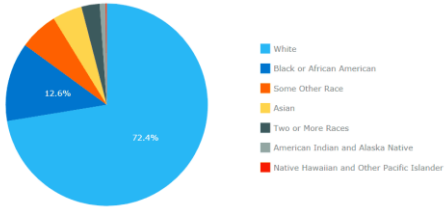
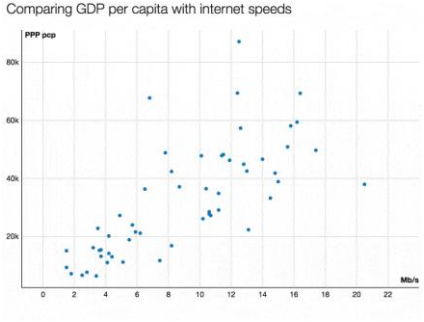
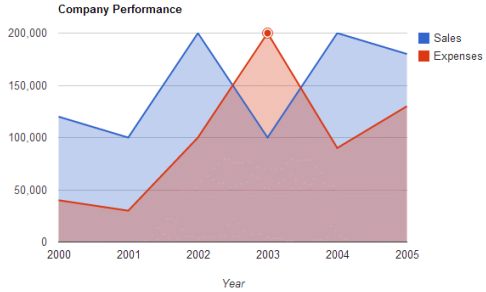


Chart types

- **Scatter Plot:** Suitable for visualizing the relationship between two *continuous variables*.
- Ex: *Correlation between height and weight, relationship between study time and exam scores.*



- **Area Chart:** Suitable for showing the *cumulative sum or proportions* over time.
- Ex: *Total sales over time, population growth over years, stacked area chart for market share.*



- **Box Plot:** Suitable for displaying the *distribution and variability of a dataset*.
- Ex: *Comparison of salaries across departments, distribution of test scores by subject.*

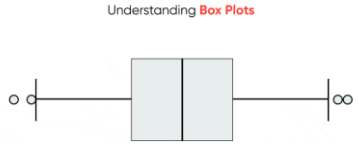
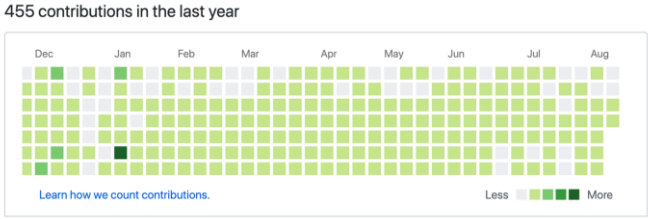
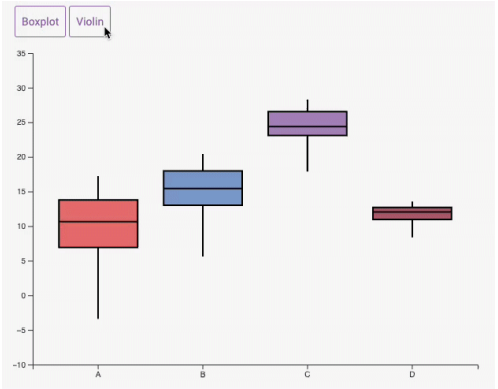


Chart types

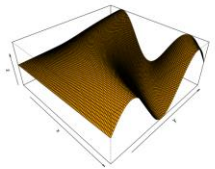
- **Heatmap:** Suitable for displaying relationships between two *categorical variables or visualizing correlation matrices*.
- Ex: *Confusion matrix in machine learning, correlation matrix of variables, visualizing a grid of data*.



- **Violin Plot:** Suitable for displaying the *distribution and density of a dataset*.
- Ex: *Comparison of income distribution by occupation, distribution of housing prices by location.*



- **3D Plots:** Suitable for visualizing *three-dimensional data or relationships*.
- Ex: *3D surface plots, 3D scatter plots, contour plots.*



Data types

- ***<u>Continuous numerical data (e.g. temperature, height, weight)</u>* -**
  - The most common chart types are ***line charts, scatter plots, and histograms***.
  - useful for showing trends over time or a continuous range of values.
  - ***Scatter plot*** > show the ***relationship between two continuous variables***.
  - ***Histogram*** > show the ***distribution of values in a dataset.***

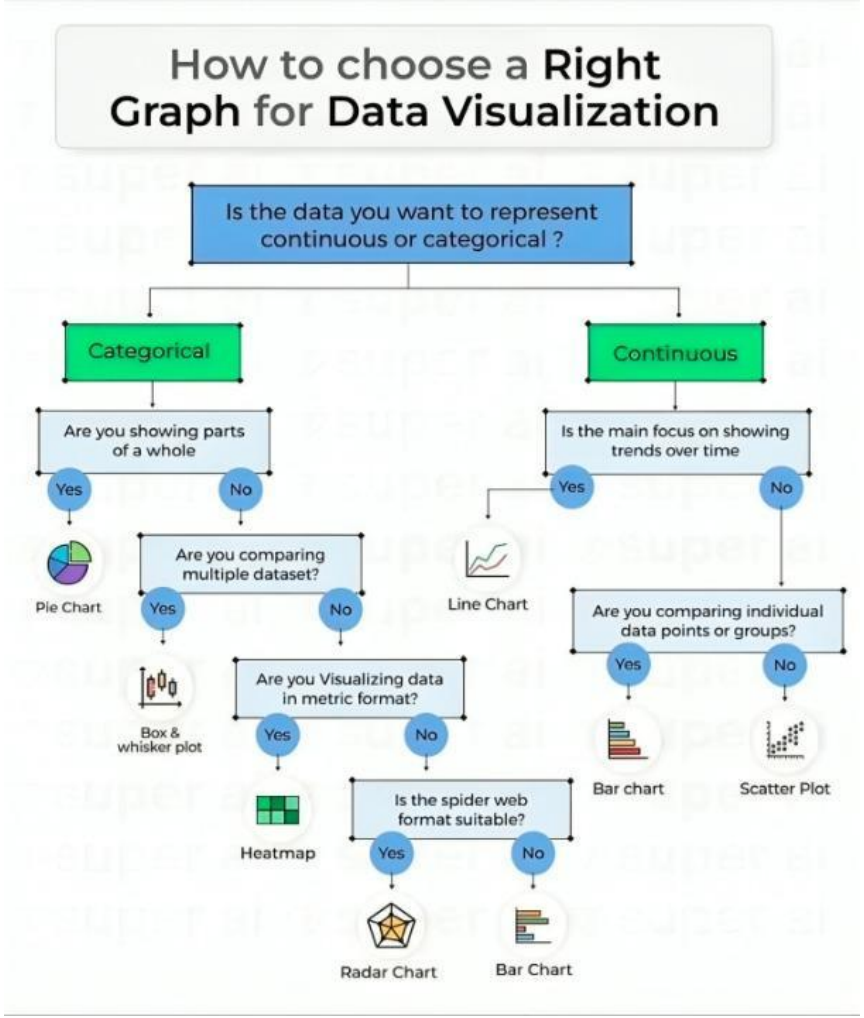
- ***<u>Categorical data</u> (e.g. colors, age groups, food cuisines, sports, genders, shapes) -***
  - Common chart types are ***bar charts and pie charts.***
  - ***Bar chart*** > compare the ***frequency of different categories***
  - ***Pie chart*** > ***show the proportion of each category in the dataset***.

Data types Cont.

- ***Time-series data (e.g. stock prices, weather patterns) -***
  - ***Line charts*** are typically the most useful for ***visualizing trends over time.***

- ***Geographic data (e.g. city populations, sales by region) -***
  - ***Choropleth*** maps are useful for showing data on a geographic map
  - ***Bubble maps*** can be used to show data points at specific geographic locations.

How to choose a Right Graph for Data Visualization



visualizations approaches

- ***Procedural visualization libraries*** - require you to explicitly specify the steps needed to create the visualization.
    - Matplotlib, ggplot2 (for R), and D3.js (for JavaScript).
- ***Declarative visualization libraries*** - allow you to describe the visualization in terms of its intended output, without specifying the steps needed to create it.
    - Plotly

Grammar of Graphics

- **Data**: Raw information that you want to visualize
- **Aesthetics**: How the data variables are mapped to visual properties such as position, size, color, and shape.
- **Geometries**: Visual marks or shapes used to represent the data. - points, lines, bars, areas, and more.
- **Scales**: How the values of data variables are mapped to the visual range.
- **Facets**: Subsets of the data and display them in separate panels or subplots.

Data Visualization with Matplotlib Library

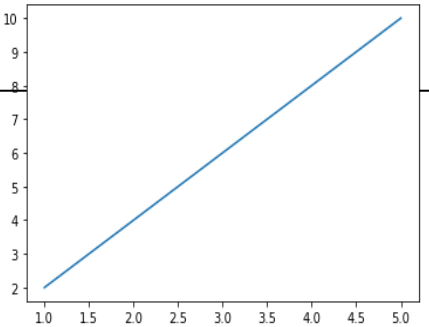
Introduction to Matplotlib

- Popular Python library used for creating static, animated, and interactive visualizations.
- It is widely used in data analysis, scientific research, and machine learning, among other fields.
- Matplotlib provides a variety of functions for creating ***line plots, scatter plots, bar charts, histograms, and more.***
- One of the most widely used data visualization libraries in Python.

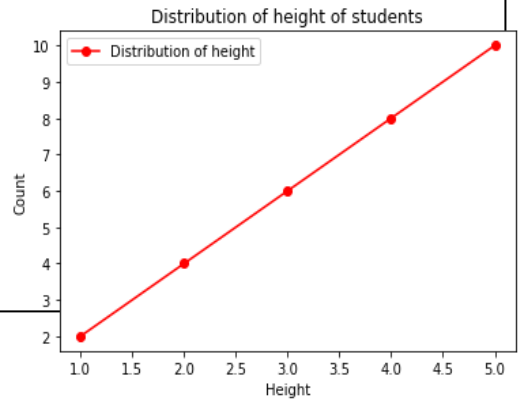
Line plots

- A line plot is a plot that displays data as a series of points connected by straight lines.

```
import matplotlib.pyplot as plt

x = [1, 2, 3, 4, 5]
y = [2, 4, 6, 8, 10]

plt.plot(x, y)

plt.show()
```




```
import matplotlib.pyplot as plt

x = [1, 2, 3, 4, 5]
y = [2, 4, 6, 8, 10]

plt.plot(x, y,'pattern',c='',label=' ')

plt.xlabel(' ')
plt.ylabel(' ')
plt.title(' ')

plt.legend()
plt.show()
```



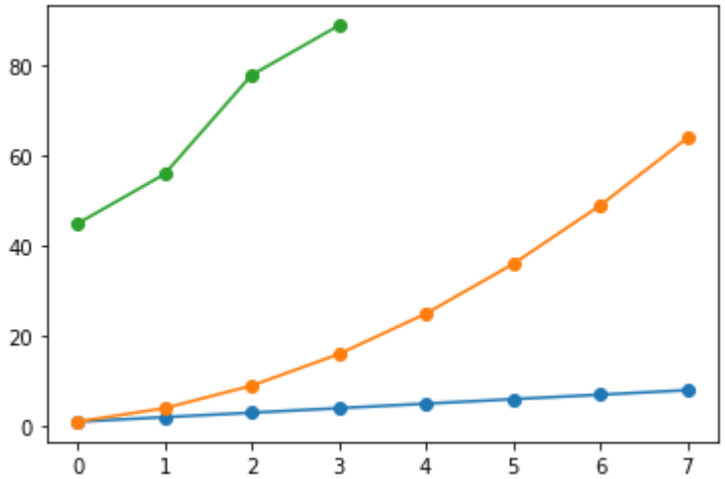
- 3rd argument - ‘pattern’

Plotting with matplotlib

- fg=plt.figure(figsize=(10,10))
- plt.gcf()
- gh=plt.gca()
- gh.axis([starting_valueX,ending_valueX,starting_valueY,ending_valueY])
- gh.get_children()

Line plots

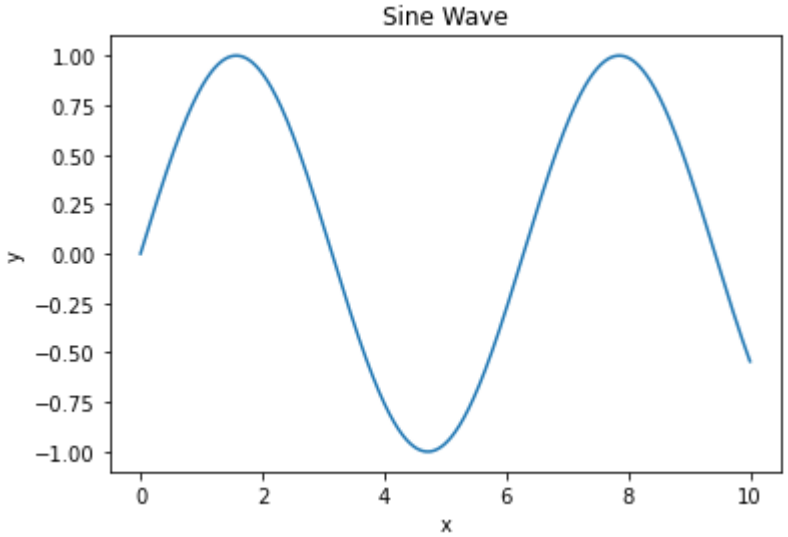
```
linear_data=np.array([1,2,3,4,5,6,7,8])

exponential_data=linear_data**2

x=[45,56,78,89]
```




```
import matplotlib.pyplot as plt
import numpy as np

linear_data=np.array([1,2,3,4,5,6,7,8])

exponential_data=linear_data**2

plt.figure()

plt.plot(linear_data,'-o',exponential_data,'-o')

plt.plot([45,56,78,89],'-o')
```

Plotting Mathematical Functions

```
import matplotlib.pyplot as plt
import numpy as np

x = np.linspace(0, 10, 1000)
y = np.sin(x)

# Plot the sine wave
plt.plot(x, y)
plt.xlabel('x')
plt.ylabel('y')
plt.title('Sine Wave')
plt.show()
```



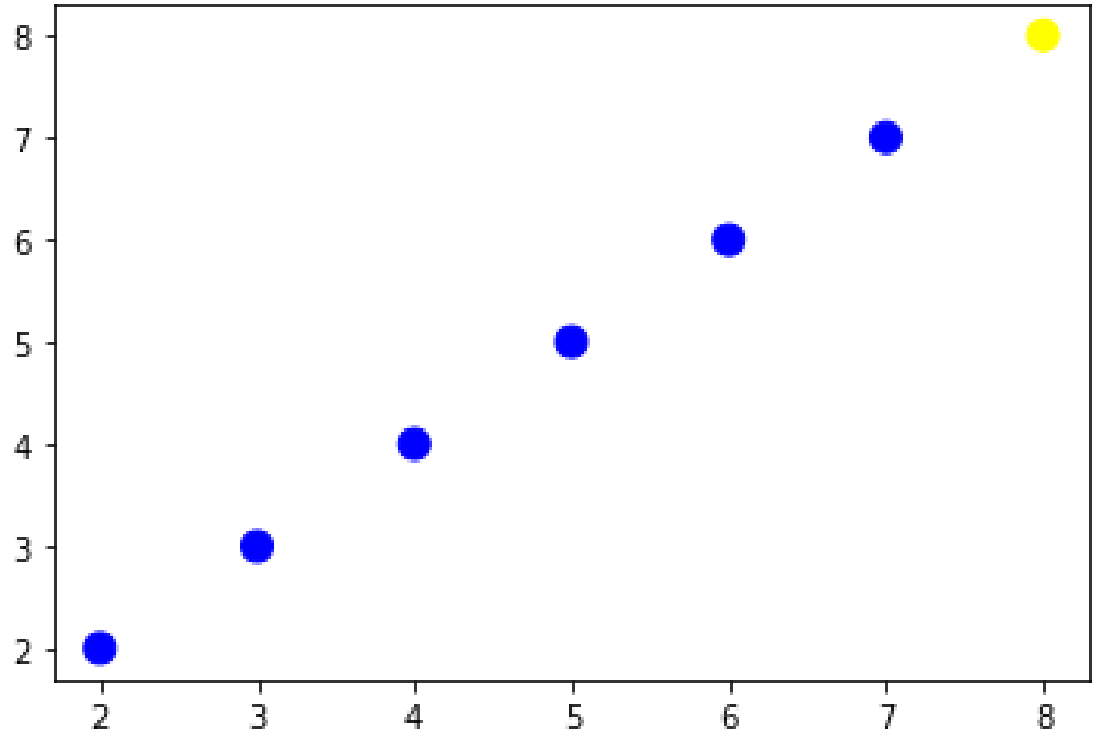
Scatterplots

Plot that displays data as a collection of points.

- plt.scatter(x, y, s=None, c=None)
- .xlabel()
- .ylabel()
- .title()

```
import matplotlib.pyplot as plt

x = np.random.randint(low=0, high=10000, size = 90)
y = np.random.randint(low=0, high=10000, size = 90)

plt.scatter(x, y)

plt.show()
```

Scatterplots

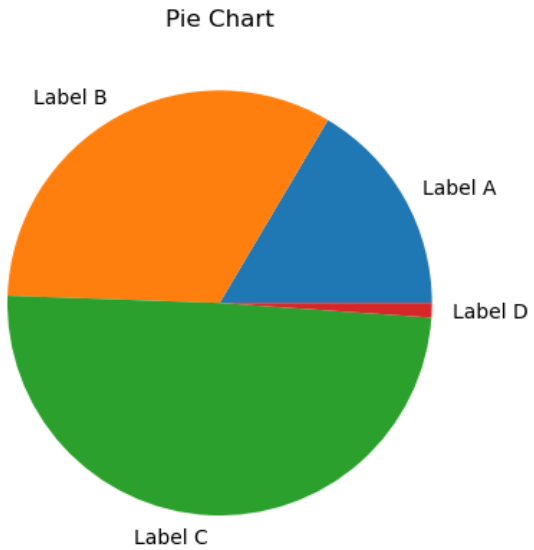



```
import numpy as np

y=np.array([2,3,4,5,6,7,8])
x=y

colors=['blue']*(len(y)-1)
colors.append('yellow')

plt.figure()

plt.scatter(x,y,s=100,c=colors)
```

Pie charts

- Pie charts are circular-shaped charts that record discrete data whereby pie represents the whole and the slices represent the parts of the whole.

```
import matplotlib.pyplot as plt

values = [15, 30, 45, 1]

labels = ['Label A', 'Label B', 'Label C', 'Label D']

# Create a pie chart
plt.pie(values, labels=labels)

plt.title('Pie Chart')

plt.show()
```

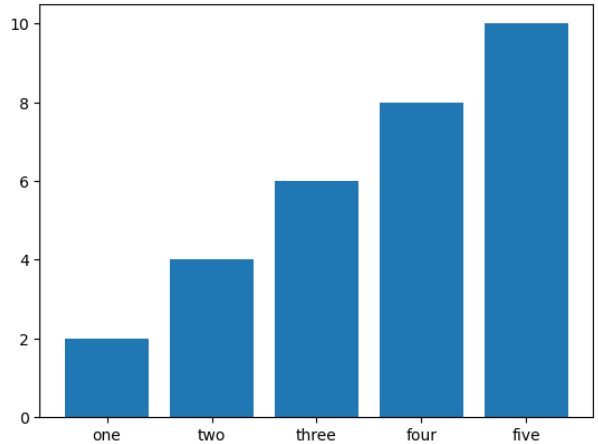


Bar charts

Plot that displays data as a series of bars, with the height of each bar representing the value of the data.

- .bar(x,y,width)

```
import matplotlib.pyplot as plt

x = ['one', 'two', 'three', 'four', 'five']
y = [2, 4, 6, 8, 10]

plt.bar(x, y)

plt.show()
```

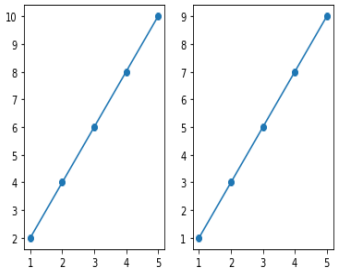


- stacked bar graph - .bar(x, ,width=,bottom= ,color=’ ’)
- horizontal stacked bar graph - .barh(x, ,height=,left= ,color=’ ’)

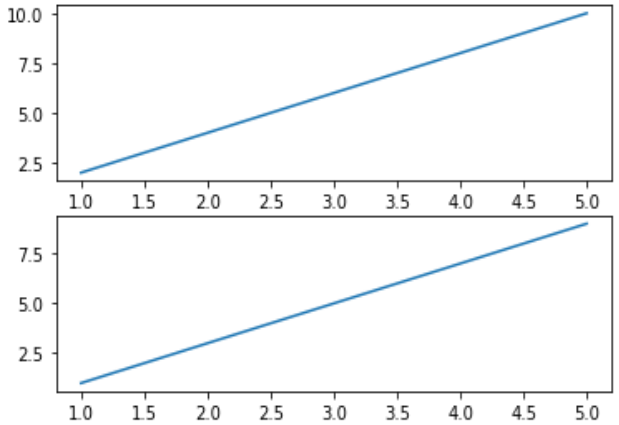
Subplots

Way to display multiple plots in a single figure. This is useful when you want to compare different data sets or display related data sets together.

- .subplot(rows,cols,plot number)
- fig, ((, ,),(, ,),(, ,))=.subplots()

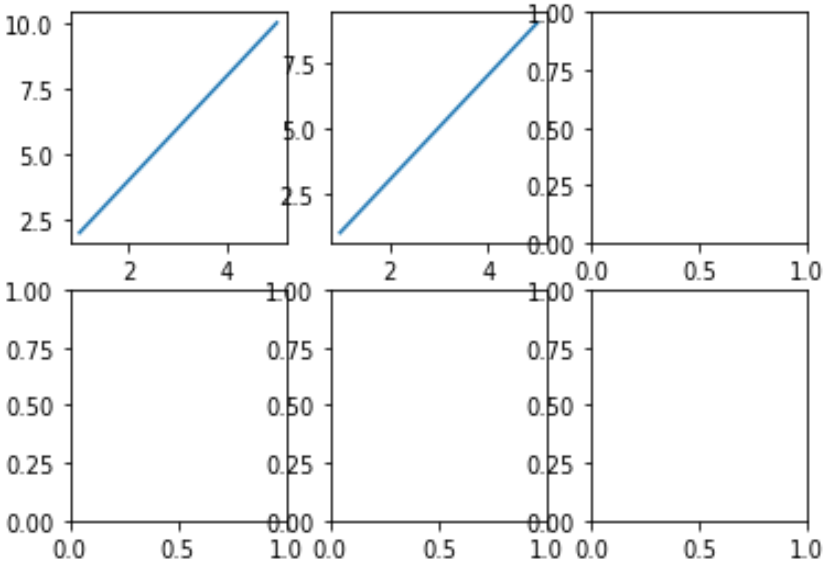


```
plt.subplot(1,2,1)

x = [1, 2, 3, 4, 5]
y1 = [2, 4, 6, 8, 10]
y2 = [1, 3, 5, 7, 9]

plt.plot(x,y1,'-o')
plt.subplot(1,2,2)
plt.plot(x,y2,'-o')
```



```
import matplotlib.pyplot as plt

x = [1, 2, 3, 4, 5]
y1 = [2, 4, 6, 8, 10]
y2 = [1, 3, 5, 7, 9]

fig, axs = plt.subplots(2, 1)

axs[0].plot(x, y1)
axs[1].plot(x, y2)

plt.show()
```

Subplots



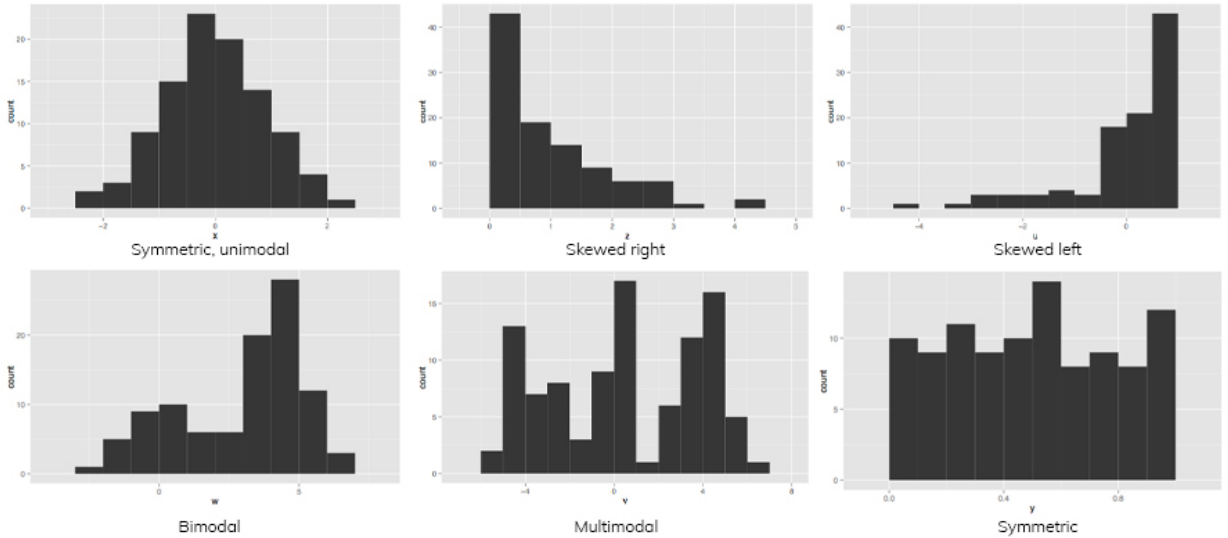

```
import matplotlib.pyplot as plt

x = [1, 2, 3, 4, 5]
y1 = [2, 4, 6, 8, 10]
y2 = [1, 3, 5, 7, 9]

fig, axs = plt.subplots(2, 3)

axs[0,0].plot(x, y1)
axs[0,1].plot(x, y2)

plt.show()
```

Subplots

```
import matplotlib.pyplot as plt
import pandas as pd

df = pd.read_csv('winequality-red.csv')

# Define the column pairs for the scatter plots
column_pairs = [('alcohol', 'pH'), ('fixed acidity', 'citric acid'), ('residual sugar', 'chlorides'), ('density', 'sulphates')]

# Create subplots
fig, axs = plt.subplots(2, 2, figsize=(10, 8))

# Iterate over the column pairs and plot each scatter plot
for i, (x_col, y_col) in enumerate(column_pairs):
    ax = axs[i // 2, i % 2]
    ax.scatter(df[x_col], df[y_col], color='steelblue', alpha=0.7)
    ax.set_xlabel(x_col)
    ax.set_ylabel(y_col)

# Adjust spacing between subplots
plt.tight_layout()

plt.show()
```

Histogram

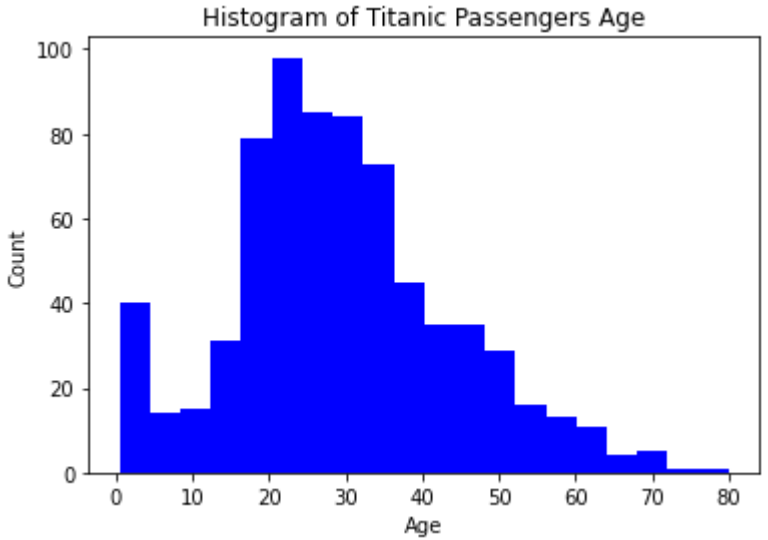
- Common way to visualize the distribution of a dataset.
- They provide a graphical representation of the ***frequency of values in a dataset***, which can help identify patterns and anomalies in the data.



Histogram

- Common way to visualize the distribution of a dataset.
- They provide a graphical representation of the ***frequency of values in a dataset***, which can help identify patterns and anomalies in the data.

```
import pandas as pd
import matplotlib.pyplot as plt

titanic_data = pd.read_csv('titanic.csv')

plt.hist(titanic_data['Age'], bins=20, color='blue')

plt.xlabel('Age')
plt.ylabel('Count')
plt.title('Histogram of Titanic Passengers Age')

plt.show()
```

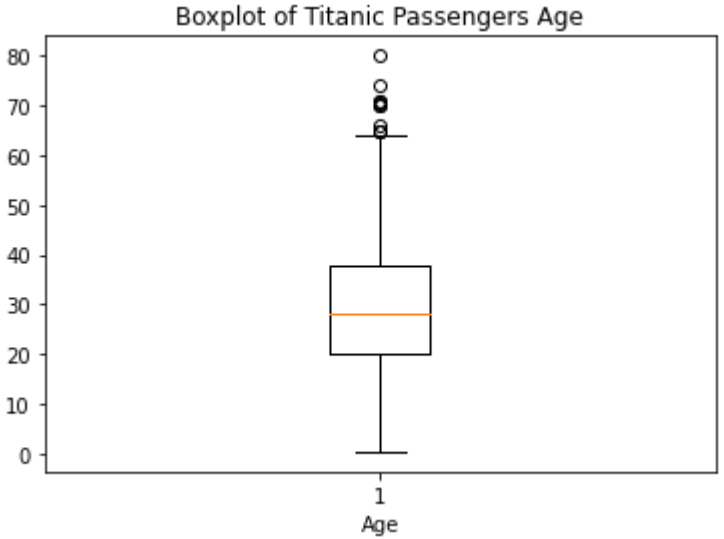


Box Plots

way of displaying the distribution of a dataset through its quartiles.

- Aggregate statistics - 5 number summary - min, max, median, 1st and 3rd quartiles
- plt.boxplot(, whis=[0,100])

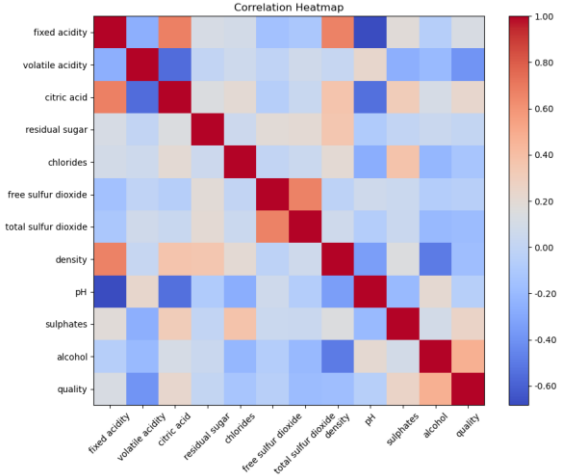
```
import pandas as pd
import matplotlib.pyplot as plt

titanic_data = pd.read_csv('titanic.csv')

plt.boxplot(titanic_data['Age'].dropna())

plt.xlabel('Age')
plt.title('Boxplot of Titanic Passengers Age')

plt.show()
```



Heatmaps

- 3 dimensions
- way of displaying data in a color-coded matrix, where different colors represent different values.

```
import matplotlib.pyplot as plt
import pandas as pd
df = pd.read_csv('winequality-red.csv')


# Calculate the correlation matrix
corr_matrix = df.corr()

# Create a heatmap
plt.figure(figsize=(10, 8))
plt.imshow(corr_matrix, cmap='coolwarm', aspect='auto')
plt.colorbar(format='%.2f')
plt.xticks(range(len(corr_matrix.columns)), corr_matrix.columns,
rotation=45)
plt.yticks(range(len(corr_matrix.columns)), corr_matrix.columns)
plt.title('Correlation Heatmap')

# Display the heatmap
plt.show()
```



Data Visualization with Pandas Library

What is Pandas?

- A popular Python library used for data manipulation and analysis
- “Pandas” comes from both "Panel Data" and "Python Data Analysis"
- Built on top of other popular Python libraries, such as NumPy and matplotlib
- Can be used for a wide range of data-related tasks, such as data cleaning, transformation, exploration, analysis, and visualization.

What is Pandas?

Types of visualizations

- Line plots
- Scatter plots
- Bar plots
- Histograms
- Box plots
- Area plots
- Pie charts
- Kernel density estimation plots
- Hexbin plots
- 3D scatter plots

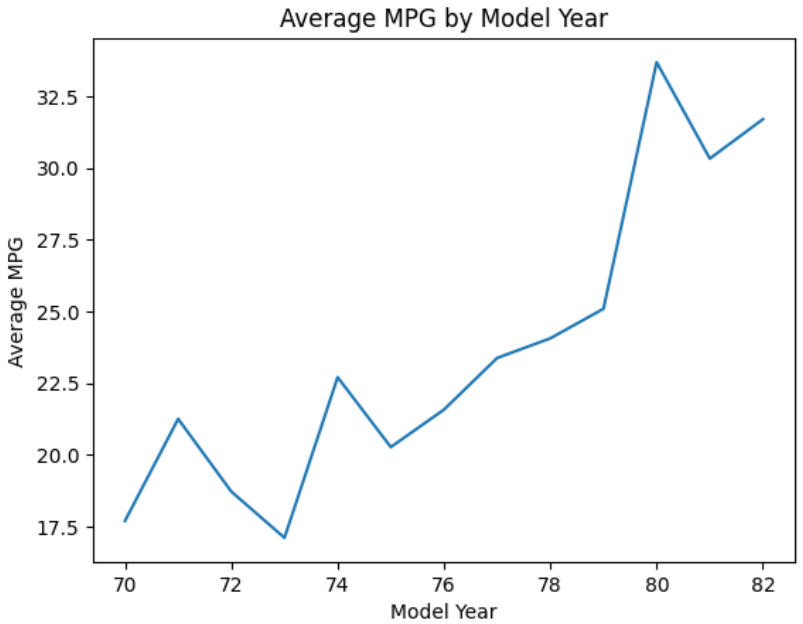
Plotting with pandas

```
import pandas as pd
import matplotlib.pyplot as plt
df = pd.read_csv('mpg.csv')
df.head()
```

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	"chevrolet chevelle malibu"
1	15.0	8	350.0	165	3693	11.5	70	1	"buick skylark 320"
2	18.0	8	318.0	150	3436	11.0	70	1	"plymouth satellite"
3	16.0	8	304.0	150	3433	12.0	70	1	"amc rebel sst"
4	17.0	8	302.0	140	3449	10.5	70	1	"ford torino"

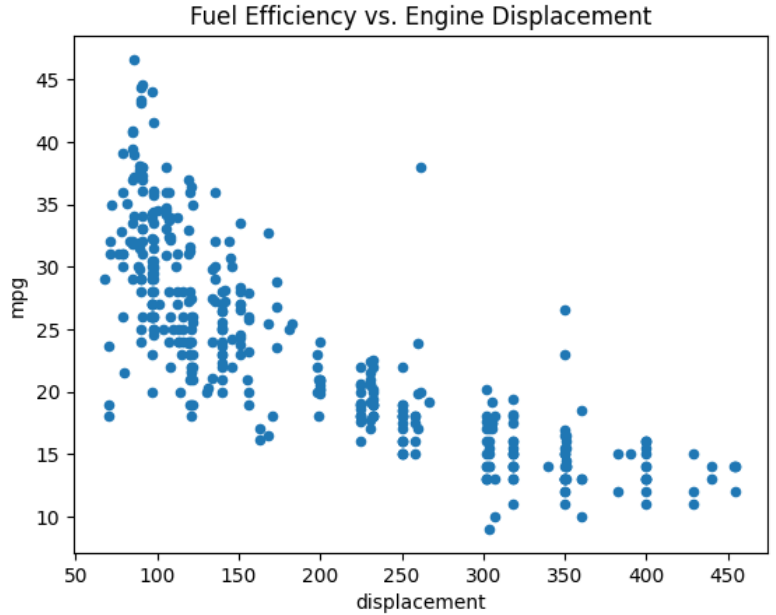
Line Plot

```
plt.figure()
df.groupby('model year')['mpg'].mean().plot(kind='line')
plt.xlabel('Model Year')
plt.ylabel('Average MPG')
plt.title('Average MPG by Model Year')
plt.show()
```



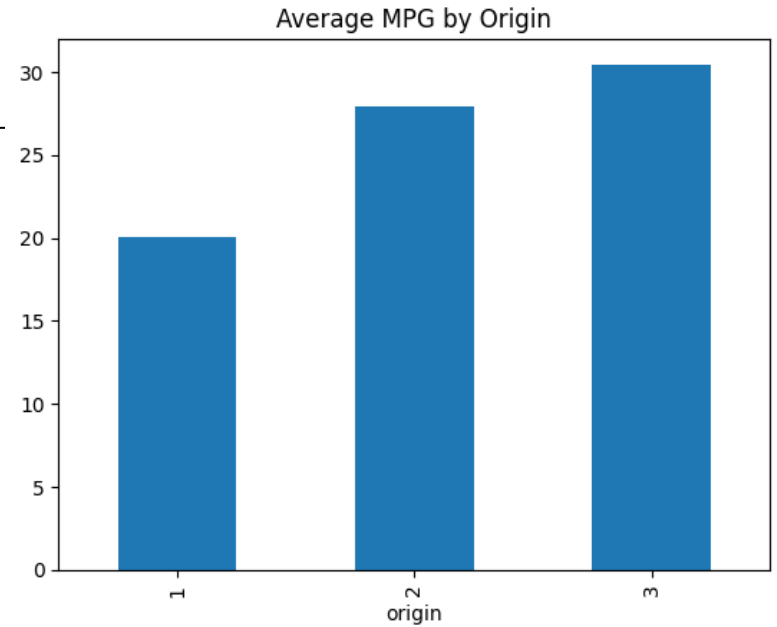
Scatter Plot

```
plt.figure()
df.plot(kind='scatter', x='displacement', y='mpg')
plt.title('Fuel Efficiency vs. Engine Displacement')
plt.savefig('scatter_plot.png')
plt.show()
```



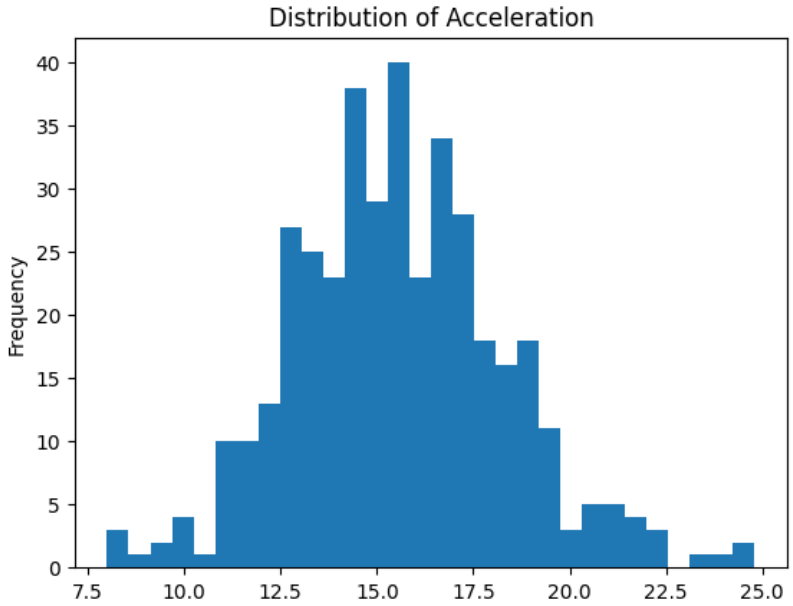
Bar Plot

```
plt.figure()
df.groupby('origin')['mpg'].mean().plot(kind='bar')
plt.title('Average MPG by Origin')
plt.show()
```



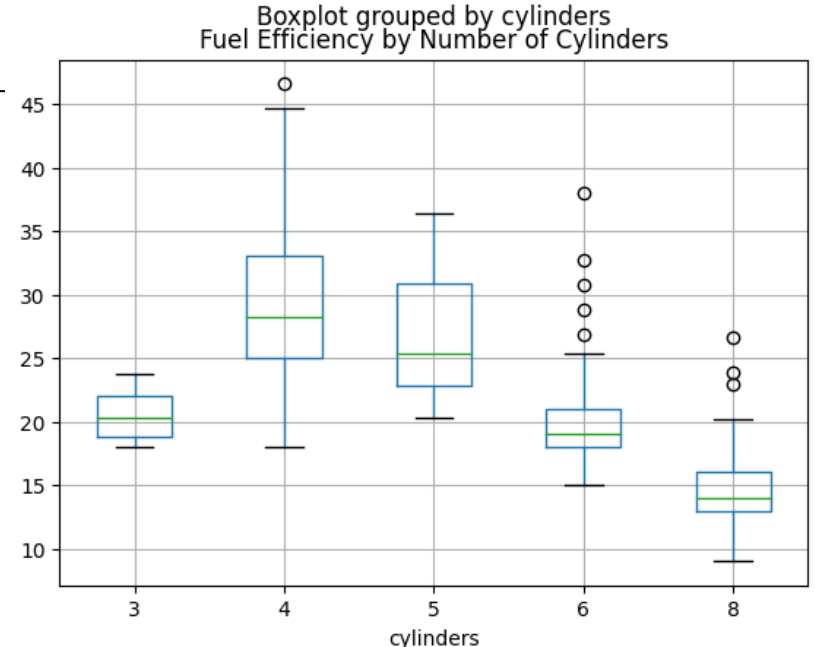
Histogram

```
plt.figure()
df['acceleration'].plot(kind='hist', bins=30)
plt.title('Distribution of Acceleration')
plt.show()
```



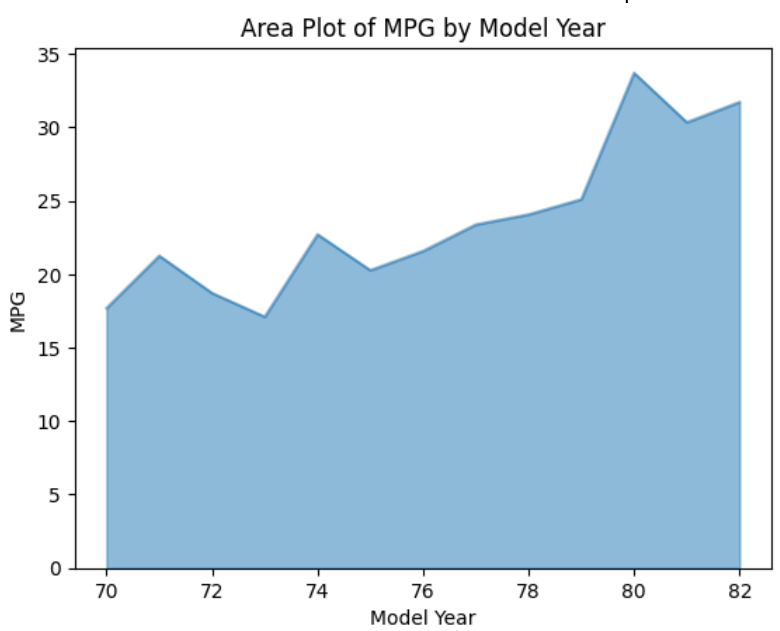
Box Plot

```
plt.figure()
df.boxplot(column='mpg', by='cylinders')
plt.title('Fuel Efficiency by Number of Cylinders')
plt.show()
```



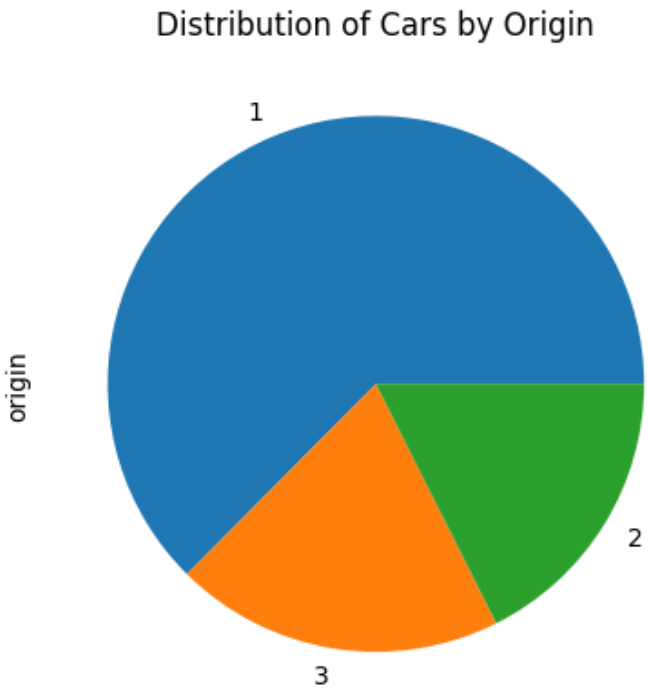
Area Plot

```
plt.figure()
df.groupby('model year')['mpg'].mean().plot(kind='area', alpha=0.5)
plt.title('Area Plot of MPG by Model Year')
plt.xlabel('Model Year')
plt.ylabel('MPG')
plt.show()
```



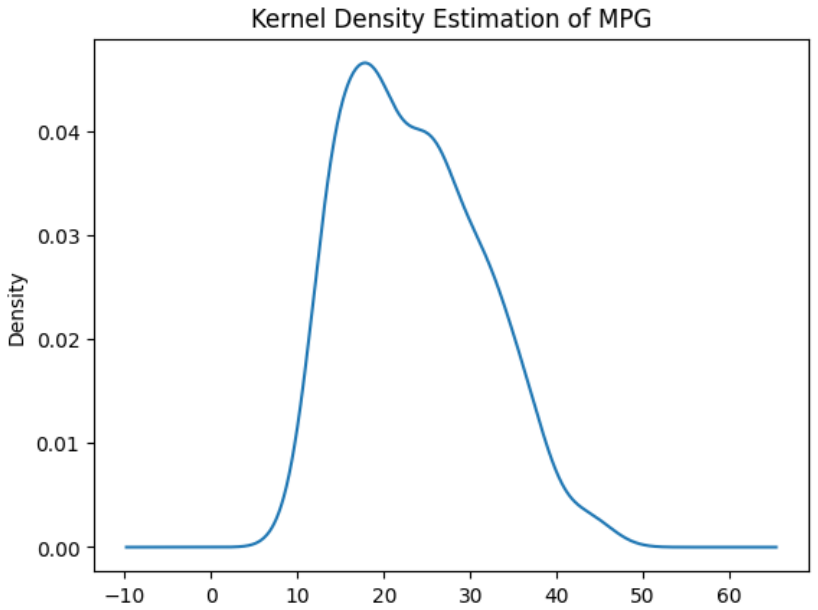
Pie Chart

```
plt.figure()
df['origin'].value_counts().plot(kind='pie')
plt.title('Distribution of Cars by Origin')
plt.show()
```



Kernel density estimation plots(KDE)

```
plt.figure()
df['mpg'].plot(kind='kde')
plt.title('Kernel Density Estimation of MPG')
plt.show()
```



Data Visualization with Seaborn Library

What is Seaborn

- A Python data visualization library based on matplotlib
- Plotting functions operate on data frames and arrays
- Graphs can be customized easily

Installation

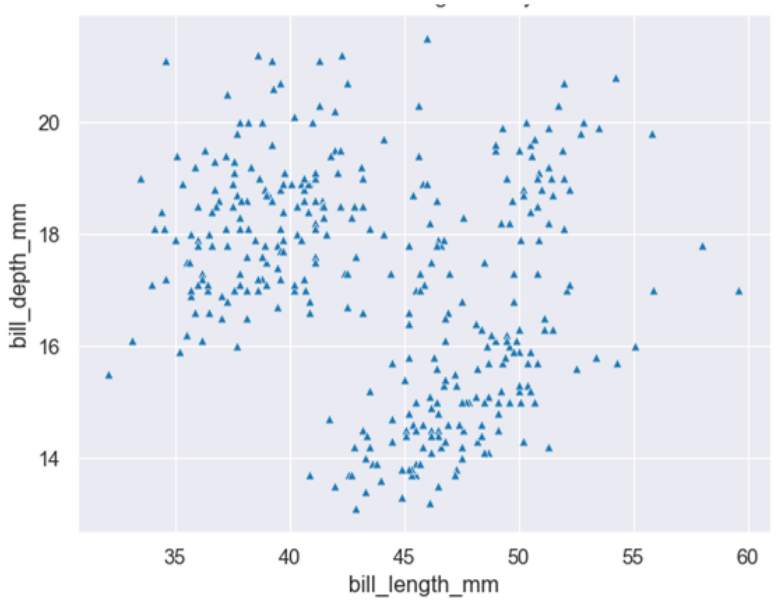
```
pip install seaborn

conda install seaborn
```

Plotting with Seaborn

```
import seaborn as sns
import pandas as pd
import matplotlib.pyplot as plt

#retreive the available datasets
sns.get_dataset_names()

sns.set_style("darkgrid")

penguins = sns.load_dataset('penguins')
penguins.head()
```

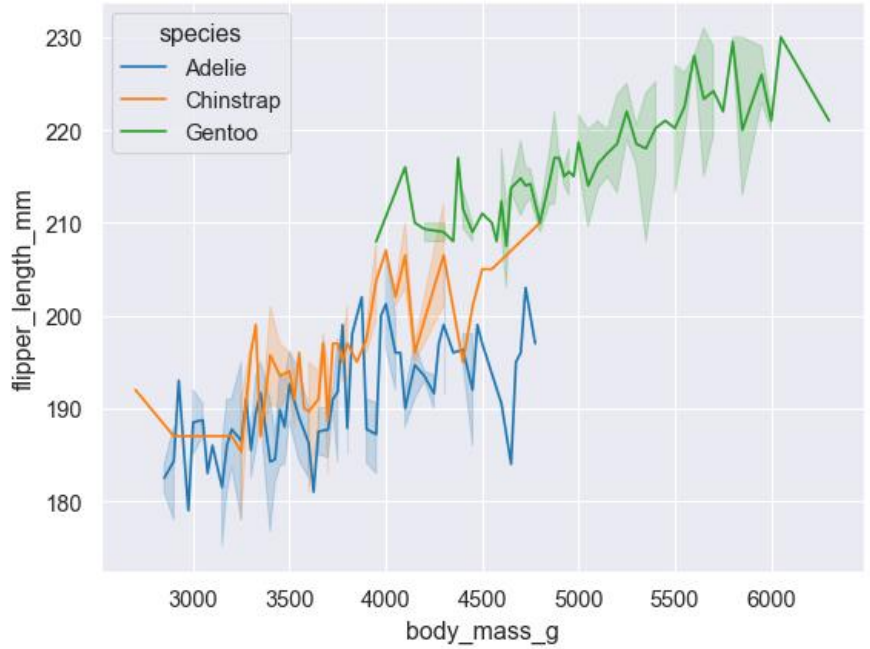
Scatter Plot

```
penguins = sns.load_dataset("penguins")
sns.scatterplot(x="bill_length_mm", y="bill_depth_mm", data=penguins, marker="^")
plt.show()
```



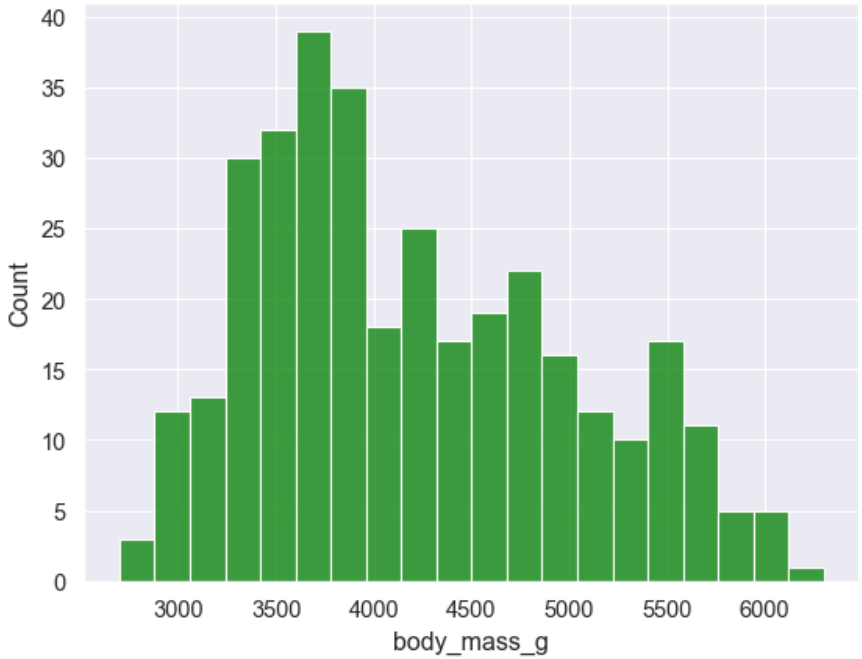
Line Plot

```
sns.lineplot(x="body_mass_g", y="flipper_length_mm", hue="species", data=penguins)
plt.show()
```



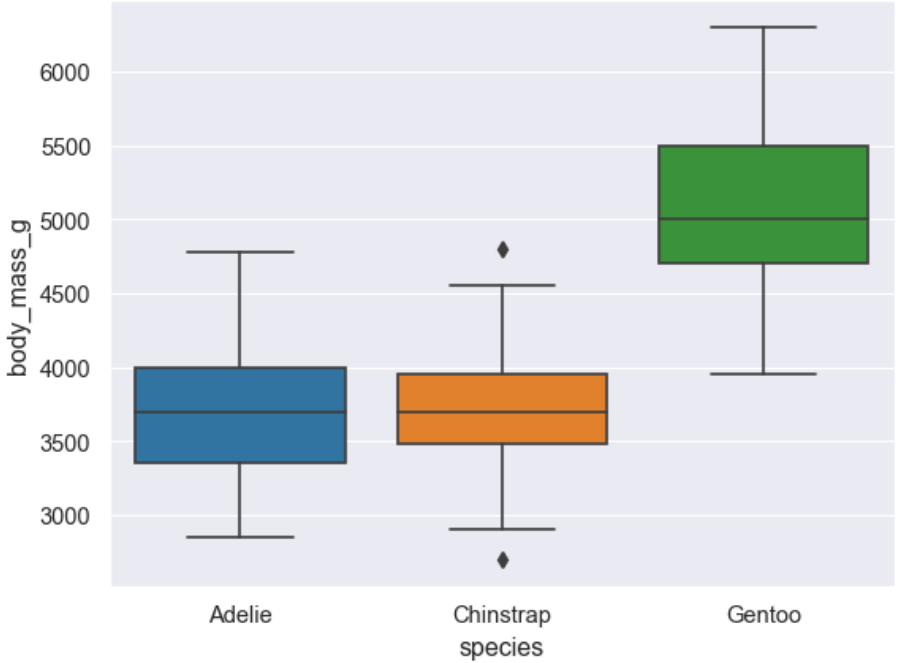
Histogram

```
sns.histplot(x="body_mass_g", data=penguins, color='green', bins = 20)
plt.show()
```



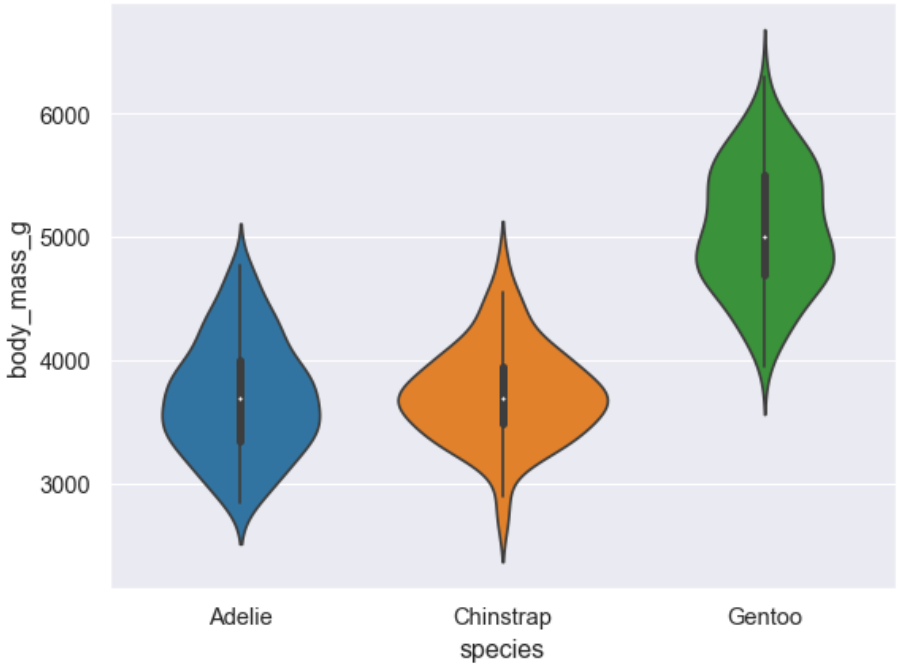
Box Plot

```
sns.boxplot(x="species", y="body_mass_g", data=penguins)
plt.show()
```



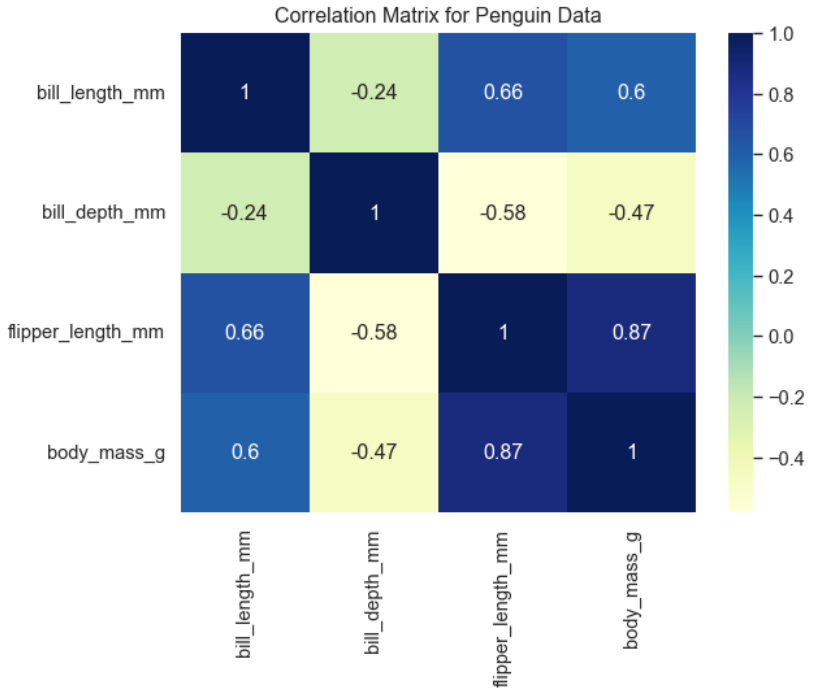
Violin Plot

```
sns.violinplot(x="species", y="body_mass_g", data=penguins)
plt.show()
```



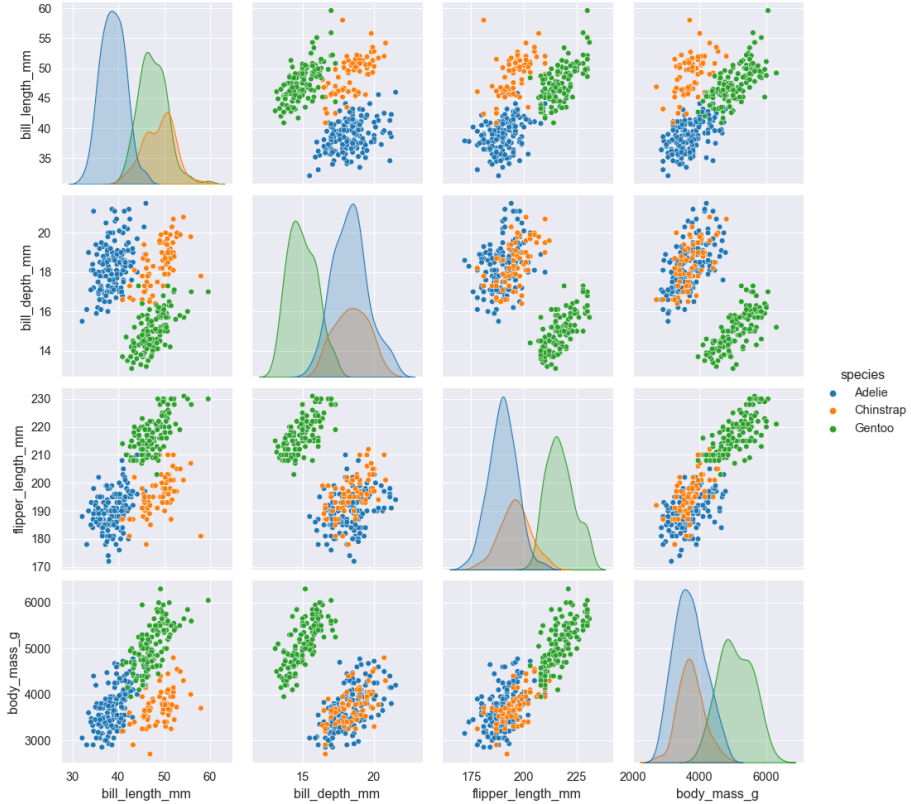
Heatmap

```
corr_matrix = penguins.corr()
sns.heatmap(corr_matrix, annot=True, cmap="
plt.show()
```



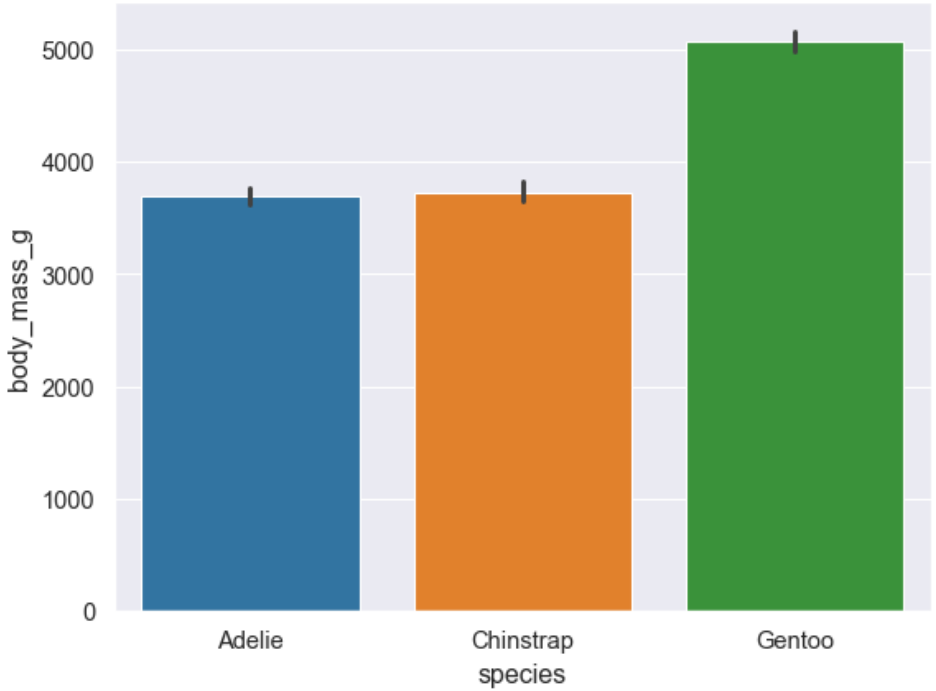
Pair Plot

```
sns.pairplot(data=penguins, hue="species")
plt.show()
```



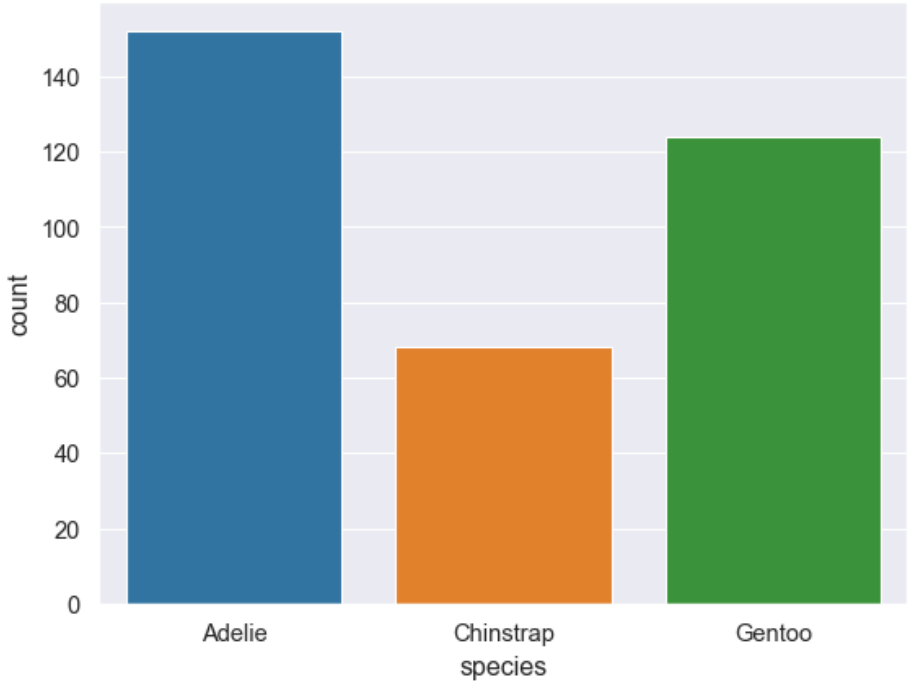
Bar Plot

```
sns.barplot(x="species", y="body_mass_g", data=penguins)
plt.show()
```



Count Plot

```
sns.countplot(x="species", data=penguins)
plt.show()
```



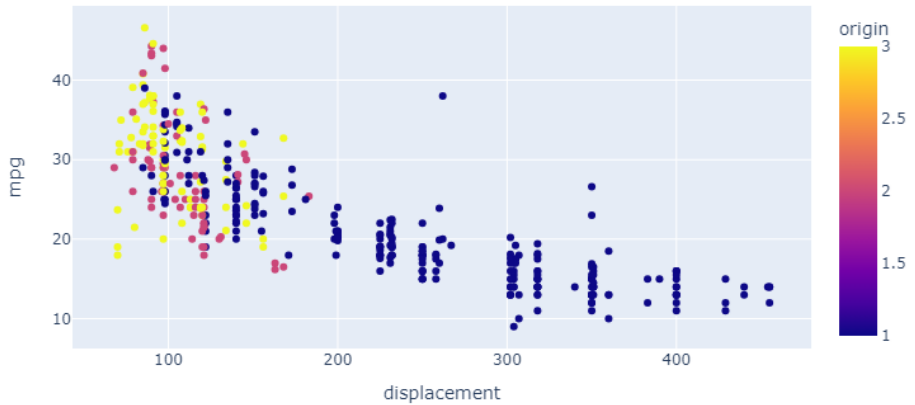
Data Visualization with Plotly Library

What is plotly?

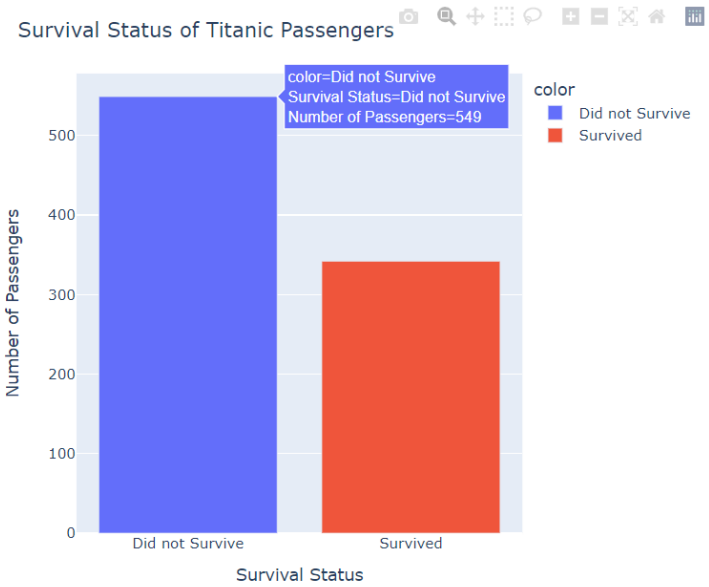
- Plotly is a Python library for creating interactive data visualizations.
- Has become a popular choice for data visualization in industry and academia.
- A wide range of chart types, including ***scatter plots, bar charts, line charts, and more.***
- Interactive features like ***hover effects, zooming, and panning.***
- Support for creating dashboards and web-based applications with Dash.
- Easy to use and customize.
- Installation - `pip install plotly`

Scatter plot

```
import pandas as pd
import plotly.express as px
df = pd.read_csv('mpg.csv')
fig = px.scatter(df, x='displacement', y='mpg', color='origin', title='Scatter
Plot')
fig.show()
```

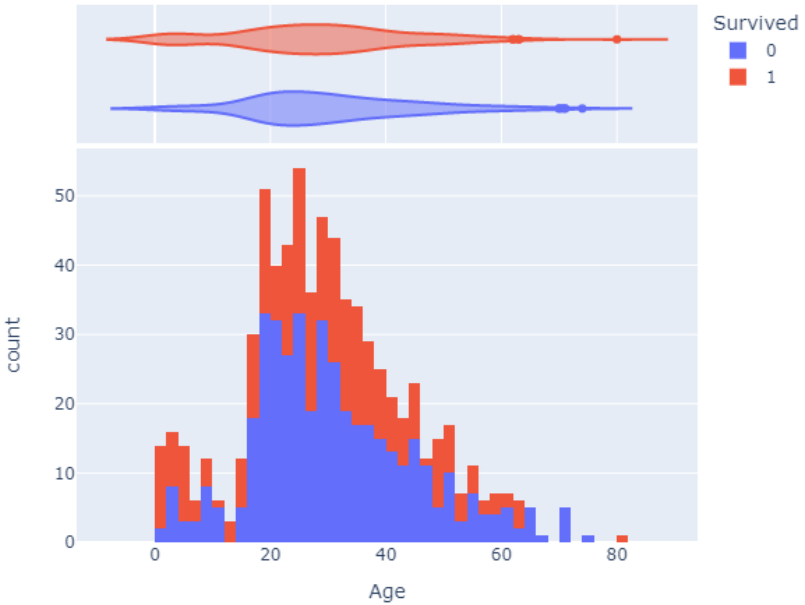


Bar chart



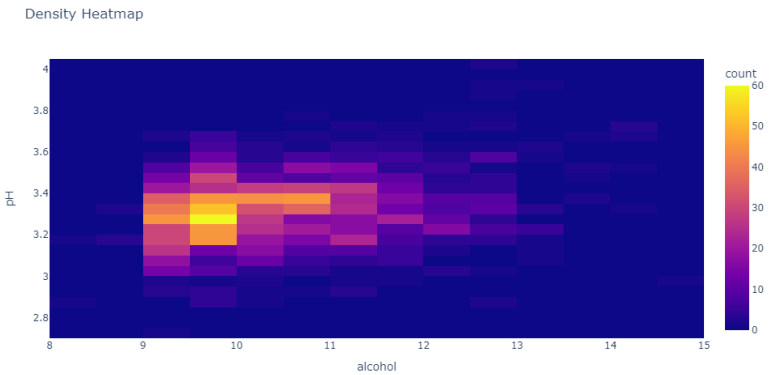

```
import plotly.express as px
import pandas as pd

titanic = pd.read_csv('titanic.csv')
survival_counts = titanic['Survived'].value_counts()

fig = px.bar(
    x=['Did not Survive', 'Survived'],
    y=survival_counts.values,
    labels={'x': 'Survival Status', 'y': 'Number of
Passengers'},
    color=['Did not Survive', 'Survived']
)

fig.update_layout(title='Survival Status of Titanic
Passengers')

fig.show()
```

Histogram



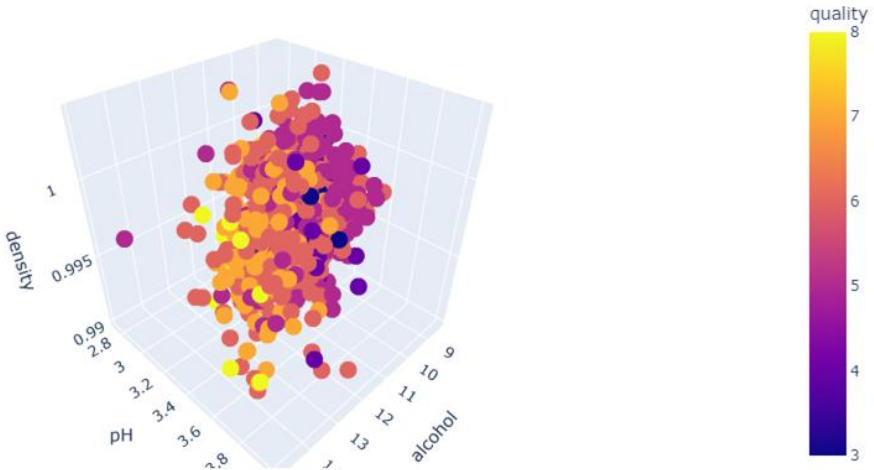

```
import plotly.express as px
df = pd.read_csv('titanic.csv')
fig = px.histogram(df, x="Age",
color="Survived", marginal="violin")

fig.show()
```

Heatmap




```
import pandas as pd
import plotly.express as px

# Load the Wine Quality dataset from a CSV file
df = pd.read_csv('winequality-red.csv')

# Create a density heatmap using Plotly Express
fig = px.density_heatmap(df, x='alcohol', y='pH',
title='Density Heatmap')
fig.show()
```

3D plots




```
import plotly.express as px
df = pd.read_csv('winequality-red.csv')

fig = px.scatter_3d(df, x='alcohol',
y='pH',z='density',color='quality')
fig.show()
```

Activity 02