#### **MATPLOTLIB**

#### **Introduction to Matplotlib**

Matplotlib is the foundational plotting library for Python, renowned for its flexibility and extensive support for 2D visualizations.

Unique Features:

- Highly customizable plots.
- "pyplot" interface emulates MATLAB-like plotting.
- Wide range of supported plot types.

Typical Use Cases:

- Data exploration
- Publication-quality figures
- Serving as a base for other libraries (e.g., Seaborn)

#### **Matplotlib Examples**

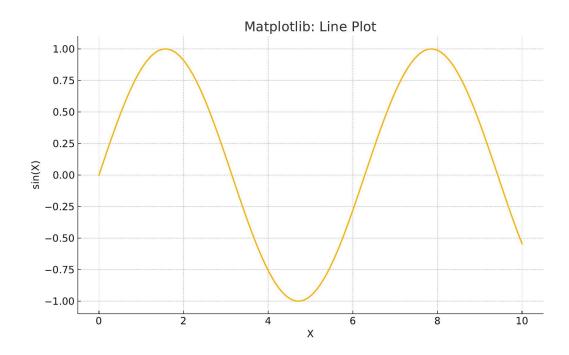
Below are some common types of graphs generated using Matplotlib, based on the example code provided.

- Line Plot
- Scatter Plot
- Bar Chart
- Histogram
- Pie Chart
- Box Plot

#### Line Plot

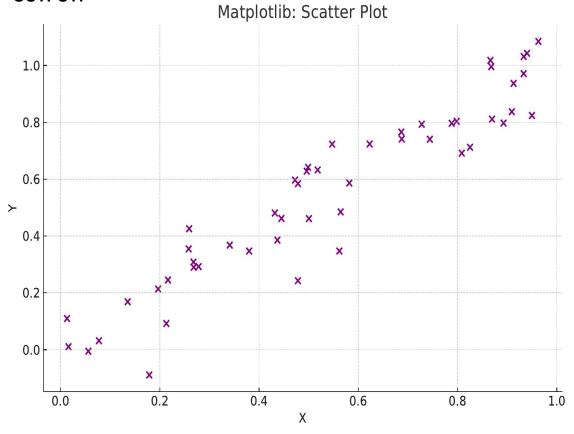
Trend over continous variables(eg:time series)

```
Line plot.py > ...
      import matplotlib.pyplot as plt
  1
      import numpy as np
  2
      x = np.linspace(0, 10, 100)
  3
      y = np.sin(x)
  4
      plt.plot(x, y)
  5
      plt.title('Matplotlib: Line Plot')
  6
      plt.xlabel('X')
 7
      plt.ylabel('sin(X)')
  8
      plt.show()
  9
 10
```



#### **Scatter Plot**

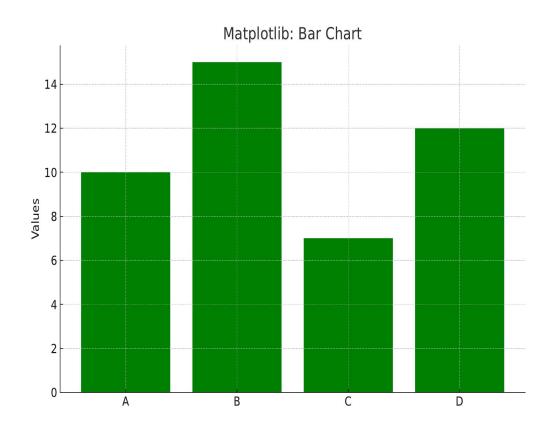
Relationship between two variables



#### **Bar Chart**

#### **Comparing Categories**

```
Bar.py > ...
1  categories = ['A', 'B', 'C', 'D']
2  values = [10, 15, 7, 12]
3  plt.bar(categories, values, color='green')
4  plt.title('Matplotlib: Bar Chart')
5  plt.ylabel('Values')
6  plt.show()
7
```



# Histogram Distribution of numerical data Histogram.py > ... data = np.random.randn(200) plt.hist(data, bins=20, color='coral', edgecolor='black') plt.title('Matplotlib: Histogram') plt.xlabel('Value') plt.ylabel('Frequency') plt.show() Matplotlib: Histogram 25 20 Frequency 12 10 -1Value

#### Pie Chart

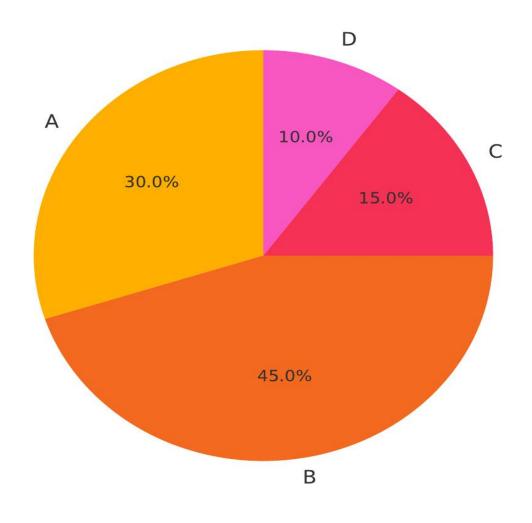
Proportions within a while

```
Pie chart.py > ...

1     sizes = [30, 45, 15, 10]
2     labels = ['A', 'B', 'C', 'D']
3     plt.pie(sizes, labels=labels, autopct='%1.1f%%', startangle=90)
4     plt.title('Matplotlib: Pie Chart')
5     plt.show()
6
```

#### **OUTPUT:**

## Matplotlib: Pie Chart



#### **BOX PLOT**

Showing distribution and outliers

```
Box plot.py > ...

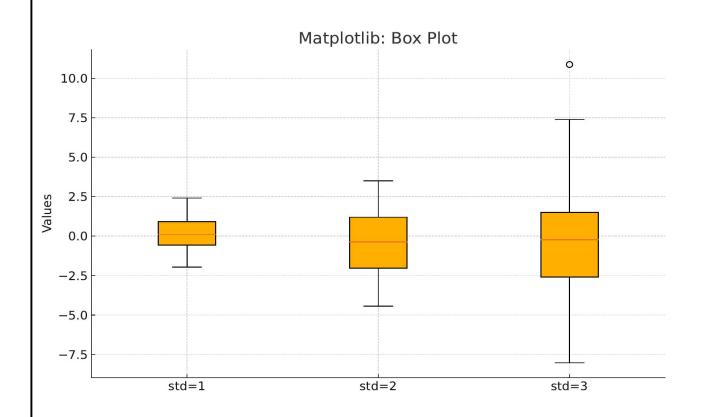
data = [np.random.normal(0, std, 100) for std in range(1, 4)]

plt.boxplot(data, vert=True, patch_artist=True, labels=['std=1', 'std=2', 'std=3'])

plt.title('Matplotlib: Box Plot')

plt.ylabel('Values')

plt.show()
```



#### **SEABORN**

#### **Introduction to Seaborn**

Seaborn is built on top of Matplotlib and is designed for statistical data visualization with a focus on attractive defaults and simplification.

#### Unique Features:

- Simplified syntax for creating complex statistical plots.
- Built-in themes, color palettes, and integration with pandas DataFrames.
- Automatically manages plot aesthetics for clarity.

#### Typical Use Cases:

- Statistical data analysis
- Exploring and visualizing relationships and distributions

#### **Seaborn Examples**

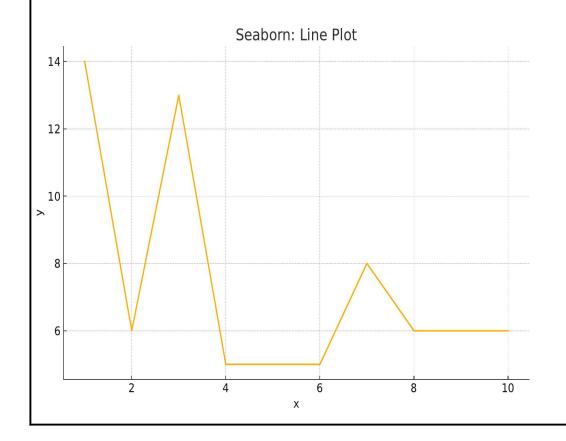
Below are various statistical plots created using Seaborn.

- Line Plot
- Grouped Scatter Plot
- Bar Plot
- Histogram with KDE
- Box Plot
- Vilion Plot

#### LINE PLOT

Time series with confidence interval

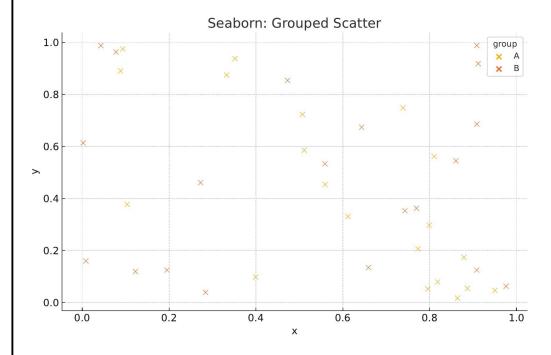
```
import pandas as pd
    x = np.arange(1, 11)
    y = np.random.randint(5, 20, 10)
    df = pd.DataFrame({'x': x, 'y': y})
    sns.lineplot(x='x', y='y', data=df)
    plt.title('Seaborn: Line Plot')
    plt.show()
```



#### **GROUPED SCATTER**

Relationship, with hue for categories

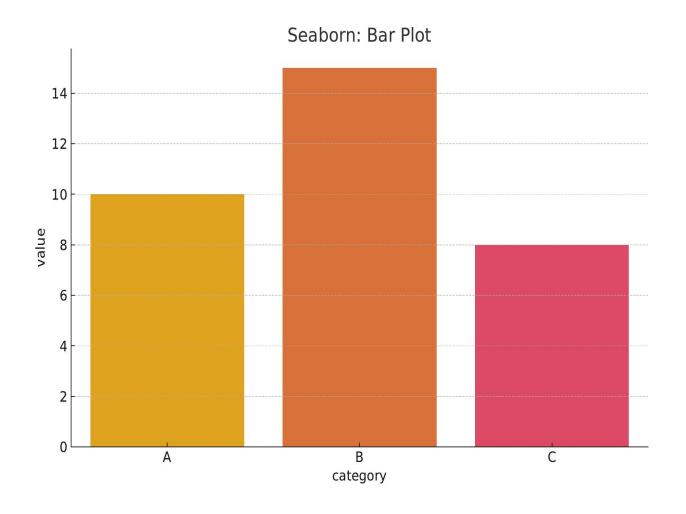
```
SCATTER PLOT.py > ...
      df = pd.DataFrame({
 1
          'x': np.random.rand(40),
 2
          'y': np.random.rand(40),
 3
          'group': ['A']*20 + ['B']*20
 4
 5
      1)
      sns.scatterplot(x='x', y='y', hue='group', data=df)
 6
 7
      plt.title('Seaborn: Grouped Scatter')
      plt.show()
 8
 9
```



#### **BAR PLOT**

Category-wise mean/CI visualization

```
BAR PLOT.py > ...
      df = pd.DataFrame({
 1
          'category': ['A', 'B', 'C'],
 2
          'value': [10, 15, 8]
 3
 4
      })
      sns.barplot(x='category', y='value', data=df)
 5
      plt.title('Seaborn: Bar Plot')
 6
      plt.show()
 7
 8
```



#### **HISTOGRAM+KDE**

Distribution with density estimation

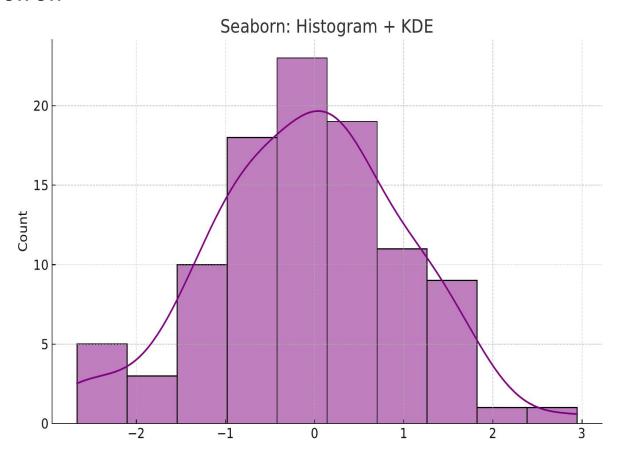
```
HISTOGRAM+KDE.py > ...

data = np.random.randn(100)

sns.histplot(data, kde=True, color='purple')

plt.title('Seaborn: Histogram + KDE')

plt.show()
```

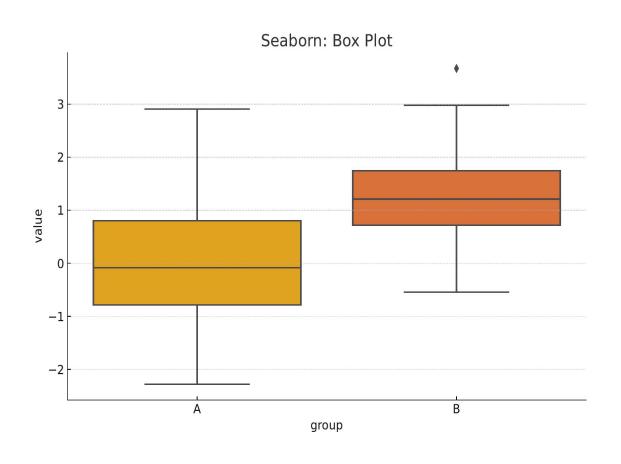


#### **BOX PLOT**

Distribution with groups, outlier detection

```
$\int \text{SEABORN BOX PLOT.py } ...

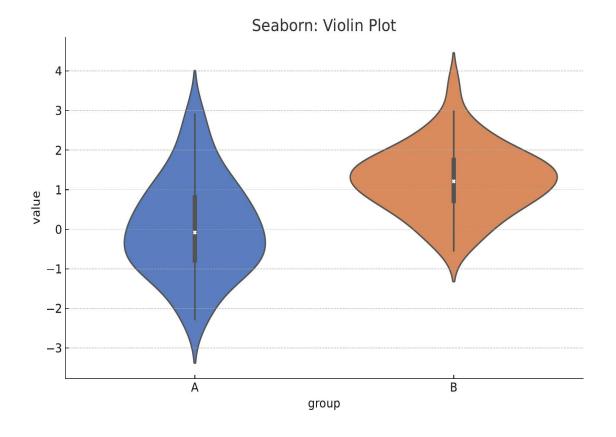
1     df = pd.DataFrame({
2          'group': np.repeat(['A', 'B'], 50),
3          'value': np.concatenate([np.random.normal(0,1,50), np.random.normal(1,1,50)])
4     })
5     sns.boxplot(x='group', y='value', data=df)
6     plt.title('Seaborn: Box Plot')
7     plt.show()
8
```



#### **VIOLIN PLOT**

Distribution shape and summary with groups

```
vilion Plot.py
sns.violinplot(x='group', y='value', data=df, palette='muted')
plt.title('Seaborn: Violin Plot')
plt.show()
```



#### Matplotlib vs Seaborn: Comparison

#### Feature Comparison:

#### Ease of Use:

- Matplotlib: Steeper learning curve, very flexible
- Seaborn: Simpler syntax for statistical plots

#### Customization:

- Matplotlib: Highly configurable (fine-tuned control)
- Seaborn: Limited, but can be adjusted via Matplotlib

#### Graph Variety:

- Matplotlib: Supports most 2D plots; 3D via mpl\_toolkits
- Seaborn: Focuses on statistical data visualization

#### Interactivity:

- Matplotlib: Basic (plugins/extensions available)
- Seaborn: Inherits Matplotlib's interactivity

#### Dataset Handling:

- Matplotlib: Manual handling, works with numpy, pandas
- Seaborn: Natively works with DataFrames; easy grouping

#### Performance:

- Matplotlib: Handles large datasets well
- Seaborn: Slightly less performant for very large data