

## 1. Introduction to Matplotlib

Matplotlib = Python's fundamental plotting library

Used for:

- Line plots
- Bar charts
- Scatter plots
- Histograms
- Custom visualizations

### Basic Structure

```
import matplotlib.pyplot as plt
```

```
plt.plot(x, y)
```

```
plt.title("My Plot")
```

```
plt.xlabel("X-axis")
```

```
plt.ylabel("Y-axis")
```

```
plt.show()
```

## 2. LINE PLOTS

Used for:

- Trends
- Time series
- Continuous data

### Example:

```
days = [1, 2, 3, 4, 5]
```

```
sales = [100, 150, 120, 180, 200]
```

```
plt.plot(days, sales)
```

```
plt.title("Daily Sales Trend")
```

```
plt.xlabel("Days")
```

```
plt.ylabel("Sales")
```

```
plt.grid(True)
```

```
plt.show()
```

**Key insight:**

Sales increased steadily, with a dip on Day 3.

### 3. BAR CHARTS

Used for comparison between categories.

**Example:**

```
products = ["A", "B", "C", "D"]  
revenue = [5000, 3000, 7000, 4000]  
plt.bar(products, revenue)  
plt.title("Product Revenue Comparison")  
plt.xlabel("Product")  
plt.ylabel("Revenue")  
plt.show()
```

**Insight:**

Product C generates the maximum revenue.

### 4. SCATTER PLOTS

Used for:

- Relationship between variables
- Trends
- Outliers

**Example:**

```
age = [20, 25, 30, 35, 40, 45]  
salary = [25, 35, 50, 60, 75, 85]  
  
plt.scatter(age, salary)  
plt.title("Age vs Salary")  
plt.xlabel("Age")  
plt.ylabel("Salary")
```

```
plt.show()
```

**Insight:**

There is a positive correlation between age and salary.

## 5. Styling & Best Practices

- Always add **titles, labels, legends**
- Use **grid lines** when needed
- Don't overload too much data
- Use color contrast appropriately
- Use consistent scales
- Include insights when presenting

## PART 2 — Seaborn Visualizations

### 1. Introduction to Seaborn

Seaborn provides statistical visualization built on Matplotlib:

- Heatmaps
- Distribution plots
- Pairplots
- Boxplots
- Violin plots
- Category visualizations

```
import seaborn as sns
```

```
sns.set_style("whitegrid")
```

### 2. Distribution Plots (distplot / histplot / kdeplot)

**Example:**

```
sns.histplot(df["Age"], kde=True)
```

```
plt.title("Age Distribution")
```

```
plt.show()
```

**Insight:**

Age follows a normal distribution with a peak around 30.

### 3. Boxplot

Shows spread, median, and outliers.

```
sns.boxplot(x=df["Fare"])  
  
plt.title("Fare Distribution - Titanic")  
  
plt.show()
```

#### Insight:

Many passengers paid low fares, with few expensive outliers.

### 4. Category Plot (countplot)

```
sns.countplot(x=df["Survived"])  
  
plt.title("Survival Count")  
  
plt.show()
```

#### Insight:

More passengers did not survive than survived.

### 5. Heatmaps

Used for correlation analysis.

```
sns.heatmap(df.corr(), annot=True, cmap="coolwarm")  
  
plt.title("Correlation Heatmap")  
  
plt.show()
```

#### Insight:

On Titanic, Fare correlates positively with Survival.

### 6. Pairplot (BONUS)

Shows relationship between all numerical columns.

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

#### Insight:

Setosa is clearly separated from the other species in petal measurements.

## PART 3 — Guided Hands-On

Students work on **Iris** and **Titanic** datasets.

## Dataset 1 — IRIS

### Visualizations to create:

1. **Scatter Plot:**  
Sepal length vs Petal length (color by species)
2. **Histogram:**  
Distribution of petal length
3. **Boxplot:**  
Petal width for each species
4. **Pairplot:**  
Compare all numeric features across species

### Expected Insights:

- Setosa shows smaller petal measurements
- Virginica tends to have the largest petals
- Species clusters are visually separable

## Dataset 2 — TITANIC

### Visualizations to create:

1. **Countplot:**  
Survivors vs Non-survivors
2. **Bar Chart:**  
Survival rate by gender
3. **Histogram:**  
Fare distribution
4. **Boxplot:**  
Age distribution among survivors
5. **Heatmap:**  
Correlation among Age, Fare, Survived, Pclass

### Expected Insights:

- Higher female survival rate
- Passengers in 1st class survived more
- Older passengers were fewer than young
- Fare positively correlates with survival

## **PART 4 — Understanding Insights from Graphs**

Teach students:

- What story does the graph tell?
- What patterns do you see?
- Are there any anomalies?
- Are two variables correlated?
- What decision would a business make from this?

## **PART 5 — Good Visualization Practices**

- ✓ Choose appropriate chart type
- ✓ Use readable colors
- ✓ Don't clutter labels
- ✓ Always label axes
- ✓ Keep titles meaningful
- ✓ Add insights — don't let the graph speak alone

## **PART 6 — Fun Task**

### **Draw Your Mood Graph**

Students create a 7-day personal dataset:

Day,Mood

Mon,7

Tue,5

Wed,8

Thu,6

Fri,9

Sat,10

Sun,8

### **Step 1 — Line Plot**

```
plt.plot(days, moods, marker='o')
```

```
plt.title("My Mood Over the Week")
```

```
plt.xlabel("Day")
```

```
plt.ylabel("Mood Level (1-10)")
```

```
plt.ylim(0, 10)
```

```
plt.grid(True)
```

```
plt.show()
```

## **Step 2 — Add Colors**

Students choose colors representing emotions:

- Blue → Calm
- Red → Stress
- Yellow → Happy

## **Step 3 — Write 2–3 Insights**

Example:

*“My mood improves significantly on Friday and peaks on Saturday. Mid-week slump seen on Tuesday.”*