

Acknowledgement

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Ministry of Economy, Trade and Industry



Overseas Employment Corporation

What you have Learnt Last Week

We were focused on following points.

- Usage of control and loop flow statement
- Performing Linear Algebra in Numpy
- Concatenation, and Stacking in NumPy
- Import and export data effortlessly between different file formats.
- Inspecting and Understanding Data
- Basics of creating, loading, and exploring DataFrames
- Understanding of 1D, and 2D NumPy arrays
- Array indexing and slicing

What you will Learn Today

We will focus on following points.

- Basics of Matplotlib's plotting library, setting up figures, and axes
- Customize your plots with various line styles, markers, and colors
- · Making data visualization more engaging and informative.
- Visualize relationships and correlations with scatter plots
- Upload code on Github
- Quiz
- Q&A Session

This project simulates an HR system that manages employee details such as Name, Age, Salary, Department, and Work Location

[Steps Involved]

- •Create DataFrames from different sources (CSV, dictionary, list, NumPy array)
- •Select and filter employee records based on conditions.
- •Modify DataFrame (add new columns, delete columns, update rows).
- •Sort employees by salary and name.
- Handle missing values in employee data.
- Merge employee records from multiple departments.
- •Save and export the processed data.

1. Creating a DataFrame from a dictionary

```
import pandas as pd
import numpy as np
# Creating a DataFrame from a dictionary
employees = {
  'Emp_ID': [101, 102, 103, 104, 105],
  'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eve'],
  'Age': [25, 30, 35, np.nan, 29],
   'Salary': [50000, 60000, 75000, 80000, np.nan],
  'Department': ['HR', 'IT', 'Finance', 'IT', 'HR'],
  'Location': ['New York', 'San Francisco', 'Chicago', 'Los Angeles', 'New York']
df = pd.DataFrame(employees)
```

2. Handling Missing Data and Filter Employees

```
#Handling Missing Data
df['Age'].fillna(df['Age'].mean(), inplace=True) # Fill missing age with mean
df.dropna(subset=['Salary'], inplace=True) # Drop rows with missing Salary
#Selecting Data (Filter employees in IT department)
it_employees = df[df['Department'] == 'IT']
print("\nIT Employees:\n", it_employees)
```

3. Sorting employees by Salary

```
#Adding a new column (Experience Level based on Salary)

df['Experience_Level'] = np.where(df['Salary'] > 70000, 'Senior', 'Junior')

# Sorting employees by Salary in descending order

df_sorted = df.sort_values(by='Salary', ascending=False)
```

4. Creating another DataFrame for merging

```
#Creating another DataFrame for merging
new_hires = pd.DataFrame({
  'Emp_ID': [106, 107],
  'Name': ['Frank', 'Grace'],
  'Age': [27, 26],
  'Salary': [55000, 62000],
  'Department': ['HR', 'Finance'],
  'Location': ['Boston', 'Chicago']
})
```

5. Exporting the final DataFrame to CSV

```
#Merging the new hires into the existing DataFrame
df_final = pd.concat([df, new_hires], ignore_index=True)

# Exporting the final DataFrame to CSV
df_final.to_csv('employee_data.csv', index=False)

print("\nFinal Employee Data:\n", df_final)
```

What is Matplotlib?

Matplotlib is a powerful data visualization library in Python

[Features]

- It provides 2D plotting capabilities and supports basic 3D plotting
- Used for creating static, animated, and interactive visualizations
- Works seamlessly with NumPy, Pandas, and Jupyter Notebook
- Inspired by MATLAB's plotting functions but built for Python

Why Use Matplotlib for Data Visualization?

Matplotlib is a powerful data visualization library in Python

[Why Matplotlib?]

- Easy to Use Simple syntax for quick plotting.
- Highly Customizable Control over colors, labels, grid, styles, and more.
- Multiple Plot Types Line, bar, scatter, histogram, pie charts, etc.
- Integration Works well with SciPy, Pandas, and Seaborn.
- Interactive & Static Plots Export graphs as images or interactive charts.

Install and Import Matplotlib

Using Pyplot Interface (quicker plots)

[Example]

import matplotlib.pyplot as plt

```
# Creating a simple line plot x = [1, 2, 3, 4, 5] y = [10, 20, 25, 30, 40] plt.plot(x, y, marker='o', linestyle='-', color='b', label="Line 1") # Line plot plt.xlabel("X-axis Label") # Label for x-axis plt.ylabel("Y-axis Label") # Label for y-axis plt.title("Simple Line Plot") # Title of the plot plt.legend() # Show legend plt.show() # Display the plot
```

Figure and Axes in Matplotlib

Using Object-Oriented Approach (complex plots)

```
import matplotlib.pyplot as plt
# Creating a figure with two subplots (1 row, 2 columns)
fig, ax = plt.subplots(1, 2, figsize=(10, 5)) # 1 row, 2 columns
# First subplot
ax[0].plot([1, 2, 3, 4], [10, 20, 25, 30], marker='o', linestyle='-', color='r')
ax[0].set_title("First Subplot")
ax[0].set xlabel("X-axis")
ax[0].set ylabel("Y-axis")
# Second subplot
ax[1].bar(["A", "B", "C", "D"], [5, 7, 3, 8], color='g') # Bar chart
ax[1].set_title("Second Subplot")
plt.tight layout() # Adjust layout to prevent overlap
plt.show()
```

Saving Plot Figures

You can save plots as image files using plt.savefig()

[Visualizing temperature trends using line plots]

```
import pandas as pd
data = pd.read_csv('temperature_data.csv')
plt.plot(data['Month'], data['AvgTemperature'])
plt.xlabel('Month')
plt.ylabel('Temperature (°C)')
plt.title('Monthly Temperature Trends')
plt.show()
# Save as PNG with high resolution
plt.savefig("plot.png", dpi=300, bbox_inches='tight')
```

Task

You can save plots as image files using plt.savefig()

```
import pandas as pd
import matplotlib.pyplot as plt
# Load the dataset
file_path = "temperature.csv"
data = pd.read csv(file path)
# Convert temperature from Fahrenheit to Celsius and
drop missing values
data["AverageTemperatureC"] =
(data["AverageTemperatureFahr"] - 32) * 5/9
data cleaned =
data.dropna(subset=["AverageTemperatureC"])
# Group by month and calculate the average
temperature
monthly_avg_temp =
data_cleaned.groupby("month")["AverageTemperatureC
"].mean()
```

```
# Plot the temperature trend
plt.figure(figsize=(10, 5))
plt.plot(monthly_avg_temp.index,
monthly_avg_temp.values, marker='o', linestyle=
color='b')
# Labels and title
plt.xlabel("Month")
plt.ylabel("Temperature (°C)")
plt.title("Monthly Temperature Trends")
# Show and save the plot
plt.savefig("temperature_plot.png", dpi=300,
bbox_inches='tight')
plt.show()
```

Customization by using various function

- 1. Line Customization: Different colors, line styles, and widths are used
- 2. Marker Customization: Different markers ('o' and 's'), sizes, and colors.
- 3. Color Customization: Uses named colors ('red', 'cyan'), hex ('#00A6D6'), and custom colormaps.
- **4. Annotations:** The plt.annotate() function highlights a peak.
- 5. Grid and Background: plt.grid() customizes grid lines, and set_facecolor() changes the background color.

1. Sample the Data

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

# Sample data
x = np.linspace(0, 10, 10)
y1 = np.sin(x)
y2 = np.cos(x)

plt.figure(figsize=(10, 6))
```

2. Sample the Data

```
# Plot with different line styles, colors, and markers
plt.plot(x, y1, color='red', linestyle='-', linewidth=2, marker='o', markersize=8,
         markerfacecolor='yellow', markeredgecolor='black', label='Sine Wave')
plt.plot(x, y2, color='#00A6D6', linestyle='--', linewidth=2, marker='s', markersize=6,
        markerfacecolor='cyan', markeredgecolor='black', label='Cosine Wave')
# Adding labels and title
plt.xlabel("X-axis", fontsize=12)
plt.ylabel("Y-axis", fontsize=12)
plt.title("Customized Plot Example", fontsize=14)
```

3. Adding a legend and annotation

4. Adding grid and change background color

```
# Adding a grid with customization
plt.grid(color='gray', linestyle=':', linewidth=0.5)
# Changing the background color
plt.gca().set_facecolor('whitesmoke')
# Show the plot
plt.show()
```

Visualizing Relationships and Correlations with Scatter Plots

A scatter plot is used to visualize the relationship between two numerical variables

[Why Scatter Plot]

- 1. Show Relationships Between Variables
- 2. Detect Patterns & Trends
- 3. Identify Outliers
- 4. Compare Multiple Datasets
- 5. Basis for Further Analysis

```
import matplotlib.pyplot as plt
import numpy as np
# Sample data
x = np.random.rand(50)
y = np.random.rand(50)
# Create scatter plot
plt.scatter(x, y)
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.title("Basic Scatter Plot")
plt.show()
```

Customizing Scatter Plots

Changing Colors Based on Categories

```
# Assigning categories randomly
categories = np.random.randint(0, 3, 50) # Three categories: 0, 1, 2
colors = ['red', 'blue', 'green']

plt.scatter(x, y, c=[colors[i] for i in categories])
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.title("Scatter Plot with Categorical Colors")
plt.show()
```

Customizing Scatter Plots

Using Different Marker Styles

```
markers = ['o', 's', 'D'] # Circle, Square, Diamond
for i, marker in enumerate(markers):
    plt.scatter(x[categories == i], y[categories == i], marker=marker, label=f'Category {i}')

plt.legend()
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.title("Scatter Plot with Different Marker Styles")
plt.show()
```

Customizing Scatter Plots

Adjusting Transparency (alpha)

```
plt.scatter(x, y, alpha=0.5, color='purple')
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.title("Scatter Plot with Transparency")
plt.show()
```

Adding Size Variations

Using Data-Driven Marker Sizes

```
sizes = np.random.rand(50) * 500 # Scale up marker sizes

plt.scatter(x, y, s=sizes, alpha=0.5, color='orange')

plt.xlabel("X-axis")

plt.ylabel("Y-axis")

plt.title("Scatter Plot with Varying Marker Sizes")

plt.show()
```

Adding Size Variations

Scaling Markers Dynamically

```
z = np.random.rand(50) * 100 # Another variable for scaling

plt.scatter(x, y, s=z, alpha=0.5, c=z, cmap='coolwarm')

plt.colorbar(label="Size Scaling Variable")

plt.xlabel("X-axis")

plt.ylabel("Y-axis")

plt.title("Scatter Plot with Dynamic Scaling and Colormap")

plt.show()
```

Colormap (cmap=coolwarm) •Low z_{values} → Dark blue. •High z_{values} → Dark red.

Colormap and Colorbar

A colormap is used in scatter plots to represent an additional dimension of data (e.g., intensity, magnitude)

```
plt.scatter(x, y, c=z, cmap='viridis', s=100, alpha=0.7)
plt.colorbar(label="Color Intensity Based on Z")
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.title("Scatter Plot with Colormap and Colorbar")
plt.show()
                                                   Colormap (cmap='viridis)
                                                   •Low z_{values} \rightarrow Dark purple/blue.
                                                   •Mid z_{\text{values}} \rightarrow \text{Green}.
                                                   •High z_{\text{values}} \rightarrow \text{Yellow}.
```

Displaying Regression Trends

used for polynomial regression, which means fitting a curve to data points

```
# Generate a linear relationship
x = np.linspace(0, 10, 50)
y = 3*x + np.random.randn(50) * 5 # y = 3x + noise
plt.scatter(x, y, color='blue', label="Data")
# Fit a linear trend line
m, b = np.polyfit(x, y, 1) # First-degree polynomial fit
plt.plot(x, m*x + b, color='red', label="Trend Line")
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.legend()
plt.title("Scatter Plot with Regression Trend Line")
plt.show()
```

Why Seaborn?

Using Seaborn for creating beautiful, informative, and statistical visualizations

[Why Seaborn?]

- 1. High-Level Interface for Statistical Graphics
- 2. Built-in Themes for Better Aesthetics
- 3. Simplifies Complex Visualizations
- 4. Integrates Well with Pandas and NumPy
- 5. Supports Multiple Plot Types
- 6. Handles Large Datasets Efficiently
- 7. Enhances Matplotlib with Additional Features

Displaying Regression Trends

Using Seaborn for creating beautiful, informative, and statistical visualizations

```
import seaborn as sns

sns.regplot(x=x, y=y, scatter_kws={'color':'blue'}, line_kws={'color':'red'})
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.title("Scatter Plot with Seaborn Regression")
plt.show()
```

3D Scatter Plots

Using mpl_toolkits.mplot3d

```
from mpl_toolkits.mplot3d import Axes3D
# Generate 3D data
x = np.random.rand(50)
y = np.random.rand(50)
z = np.random.rand(50)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(x, y, z, c=z, cmap='plasma', s=100)
ax.set xlabel("X-axis")
ax.set_ylabel("Y-axis")
ax.set_zlabel("Z-axis")
ax.set title("3D Scatter Plot")
plt.show()
```

```
    Colormap (cmap=plasma)
    Low z<sub>values</sub> → Dark purple.
    High z<sub>values</sub> → Dark orange.
```



Quiz

Everyone student should click on submit button before time ends otherwise MCQs will not be submitted

[Guidelines of MCQs]

- 1. There are 20 MCQs
- 2. Time duration will be 10 minutes
- 3. This link will be share on 6:10pm (Pakistan time)
- 4. MCQs will start from 6:15pm (Pakistan time)
- 5. This is exact time and this will not change
- 6. Everyone student should click on submit button otherwise MCQs will not be submitted after time will finish
- 7. Every student should submit Github profile and LinkedIn post link for every class. It include in your performance

Assignment

Assignment should be submit before the next class

[Assignments Requirements]

- 1. Create a post of today's lecture and post on LinkedIn.
- 2. Make sure to tag @Plus W @Pak-Japan Centre and instructors LinkedIn profile
- 3. Upload your code of assignment and lecture on GitHub and share your GitHub profile in respective your region group WhatsApp group
- 4. If you have any query regarding assignment, please share on your region WhatsApp group.
- 5. Students who already done assignment, please support other students



ありがとうございます。 Thank you.

شكريا



For the World with Diverse Individualities