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# Chapter 1 Matplotlib

# 1.1 Imports

- To use Matplotlib effectively, it is essential to import the library with standard conventions.
- The most common imports for Matplotlib are:
  - import matplotlib.pyplot as plt: This is the standard convention for accessing the Pyplot module, which provides a simple interface for creating charts and plots.
  - import numpy as np: Often used alongside Matplotlib for numerical operations and generating data for visualizations.

#### Example 1: Standard Matplotlib Import.

```
# Import Matplotlib's Pyplot module
import matplotlib.pyplot as plt

# Create a simple line plot
plt.plot([1, 2, 3, 4])
plt.title("Simple Line Chart")

plt.show()
```

#### Example 2: Using NumPy with Matplotlib.

```
# Import Matplotlib and NumPy
import matplotlib.pyplot as plt
import numpy as np

# Generate data using NumPy
x = np.linspace(0, 10, 100)
y = np.sin(x)

# Create a sine wave plot
plt.plot(x, y)
plt.title("Sine Wave")
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.show()
```

#### Example 3: Importing 3D Toolkit.

```
# Import Matplotlib's 3D plotting toolkit
from mpl_toolkits.mplot3d import Axes3D
import matplotlib.pyplot as plt
import numpy as np

# Generate data for 3D surface plot
x = np.linspace(-5, 5, 100)
y = np.linspace(-5, 5, 100)
X, Y = np.meshgrid(x, y)
Z = np.sin(np.sqrt(X**2 + Y**2))

# Create a 3D surface plot
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.plot_surface(X, Y, Z, cmap='viridis')
plt.show()
```

# 1.2 Introduction to Matplotlib

# • Matplotlib Overview:

- Used for visualising datasets in Jupyter Notebook with Pandas and Matplotlib.
- A Python library specialising in the creation of two-dimensional charts (including 3D charts).
- One of the most widely used tools for graphical representation of data.

# • Pyplot Module:

- Provides a classic Python interface for the matplotlib library.
- Requires the import of the Numpy package separately.

# • Code for Importing Pyplot:

import matplotlib.pyplot as plt

# 1.3 Customising Charts

# 1.3.1 Adding Titles and Text

- Use the text() function to add custom text annotations anywhere on the chart.
- Use the title() function to set the chart's title.
- Example code:

# 1.3.2 Adding Grids and Legends

- Use the grid() function to add gridlines to your chart.
- Use the legend() function to add a legend to the chart.
- By default, the legend is added to the upper-right corner of the chart.
- The loc keyword argument in the legend() function allows you to change the location of the legend.
- The table below lists the location codes and their corresponding positions:

Location Code	Location String
0	best
1	upper right
2	upper left
3	lower left
4	lower right
5	right
6	center left
7	center right
8	lower center
9	upper center
10	center

Table 1.1: Legend Location Codes

Listing 1.1: Adding a Grid and a Legend

# 1.3.3 Handling Date Values

- Use matplotlib.dates to handle date values in plots.
- Set major and minor locators and formatters for the x-axis to display dates
- Example code:

```
import datetime
import matplotlib.dates as mdates
# Define data
events = [datetime.date(2017, 1, 23), datetime.date(2017, 2, 3),
         datetime.date(2017, 2, 24), datetime.date(2017, 3, 15),
         datetime.date(2017, 3, 24), datetime.date(2017, 4, 8),
         datetime.date(2017, 4, 24)]
readings = [12, 22, 25, 20, 18, 15, 17, 14]
# Define locators and formatters
months = mdates.MonthLocator()
days = mdates.DayLocator()
timeFmt = mdates.DateFormatter('%Y-%m')
# Create the plot
fig, ax = plt.subplots()
plt.plot(events, readings)
ax.xaxis.set_major_locator(months)
ax.xaxis.set_major_formatter(timeFmt)
ax.xaxis.set_minor_locator(days)
plt.show()
```

# 1.4 Line Charts

# Introduction to Line Charts

- The simplest chart is a line chart.
- A line chart represents a sequence of data points connected by a line.
- Each data point consists of a pair of values (x, y).
- Use the color and linestyle keyword arguments (kwargs) to define the appearance of the stroke.

#### Color Codes in Matplotlib

Code	Colour
b	blue
g	green
r	red
c	cyan
m	magenta
У	yellow
k	black
W	white

Table 1.2: Matplotlib color codes.

### **Examples of Line Charts**

Listing 1.2: Basic Line Chart

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

x = np.arange(-2 * np.pi, 2 * np.pi, 0.01)
y = np.sin(3 * x) / x
plt.plot(x, y)
plt.show()
```

Listing 1.3: Customised Line Chart

```
x = np.arange(-2 * np.pi, 2 * np.pi, 0.01)
y = np.sin(3 * x) / x
plt.plot(x, y, color='r', linestyle='--', linewidth=5)
plt.show()
```

# Multiple Line Charts in a Single Plot

Listing 1.4: Multiple Line Charts

```
x = np.arange(-2 * np.pi, 2 * np.pi, 0.01)
y1 = np.sin(3 * x) / x
y2 = np.sin(2 * x) / x
y3 = np.sin(x) / x

plt.plot(x, y1, 'k-', linewidth=3) # Black solid line
plt.plot(x, y2, 'g-', linewidth=2) # Green solid line
plt.plot(x, y3, 'r--', linewidth=5) # Red dashed line
plt.show()
```

#### Line Charts with DataFrames

- Data visualisation is straightforward using Pandas DataFrames.
- Pass a DataFrame as an argument to the plot() function to create multiseries line charts.

Listing 1.5: Line Chart with Pandas DataFrame

# 1.5 Working with Multiple Figures and Axes

- The subplot() function is used to represent various subplots in a single figure by subdividing the figure into different drawing areas.
- It can also be used to focus the commands on a specific subplot.
- The argument passed to the **subplot()** function determines the mode of subdivision:
  - The first integer specifies how many parts the figure is split into vertically.
  - The second integer specifies how many parts the figure is split into horizontally.
  - The third integer selects the current subplot on which the commands will act.
- Below are examples of sinusoidal trends divided into subplots.

#### Example 1: Two horizontal subplots (211 and 212).

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

t = np.arange(0, 5, 0.1)
y1 = np.sin(2 * np.pi * t)
y2 = np.cos(2 * np.pi * t)

plt.subplot(211) # First subplot
plt.plot(t, y1, 'b-.') # Plot sine in blue dash-dot

plt.subplot(212) # Second subplot
plt.plot(t, y2, 'r--') # Plot cosine in red dashed line
plt.show()
```

# Example 2: Two vertical subplots (121 and 122).

```
t = np.arange(0.1, 1, 0.05)
y1 = np.sin(2 * np.pi * t)
y2 = np.cos(2 * np.pi * t)

plt.subplot(121) # First vertical subplot
plt.plot(t, y1, 'b-.') # Plot sine in blue dash-dot

plt.subplot(122) # Second vertical subplot
plt.plot(t, y2, 'r--') # Plot cosine in red dashed line
plt.show()
```

# 1.6 Histograms

- A histogram consists of adjacent rectangles erected on the *x-axis*, split into discrete intervals called **bins**.
- The hist() function allows you to represent a histogram.
- Practical example: Let's generate a population of 100 random values from 0 to 100 using the random.randint() function.

Listing 1.6: Generating Random Population

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

pop = np.random.randint(0, 100, 100)
```

The generated population is then used to create a histogram.

- To create a histogram of these samples, use the hist() function.
- The bins kwarg divides the occurrences into specific intervals (default value is 10 bins).
- Example with 20 bins and the default 10 bins are shown below.

Listing 1.7: Creating a Histogram

```
n, bins, patches = plt.hist(pop, bins=20)
plt.show()
```

# 1.7 Bar Charts

#### 1.7.1 Introduction to Bar Charts

- Bar charts are a common type of chart, similar to histograms, but the x-axis is used to reference categories.
- The bar() function is used to create a bar chart.

```
index = [0, 1, 2, 3, 4]
values = [5, 7, 3, 4, 6]
plt.bar(index, values)
plt.show()
```

#### 1.7.2 Customizing Bar Charts with Tick Labels

- The tick labels are defined by a list of strings passed to the xticks() function.
- The location of ticks is specified using the first argument of xticks().

```
index = np.arange(5)
values = [5, 7, 3, 4, 6]
plt.bar(index, values)
plt.xticks(index + 0.4, ['A', 'B', 'C', 'D', 'E'])
plt.show()
```

#### 1.7.3 Adding Error Bars and Transparency

- Use the yerr kwarg to add error bars, combined with error\_kw for customization.
- The alpha kwarg adjusts transparency of the bars.

#### 1.7.4 Horizontal Bar Charts

- Horizontal bar charts are created using the barh() function.
- The roles of axes are reversed: categories are on the y-axis, and numerical values are on the x-axis.

#### 1.7.5 Multiseries Bar Charts

- Multiseries bar charts are created by stacking bars side by side.
- Adjust the bar positions using the width parameter.

# 1.7.6 Multiseries Horizontal Bar Charts

• Similar to multiseries bar charts but oriented horizontally.

```
plt.barh(index, values, height=0.4, label='Series 1')
plt.barh(index + 0.4, values2, height=0.4, label='Series 2')
plt.yticks(index + 0.2, ['A', 'B', 'C', 'D', 'E'])
plt.legend()
plt.show()
```

#### 1.7.7 Stacked Bar Charts with Pandas DataFrame

- Use Pandas DataFrame to create stacked bar charts for better handling of multiseries data.
- Sum the series directly to stack bars vertically.

## 1.7.8 Bar Charts with Negative Values

- Represent one series with negative values using the facecolor kwarg.
- Add numerical values at the end of each bar using the text() function.

# 1.8 Pie Charts

# 1.8.1 Introduction to Pie Charts

- The pie() function is used to create pie charts in Matplotlib.
- A pie chart represents data as slices of a circle, with each slice proportional to the value it represents.
- Various kwargs are available for customization, including:
  - explode: Separates slices for emphasis.
  - startangle: Rotates the chart (default is 0).
  - autopct: Adds percentage text labels to the center of each slice.
  - shadow: Adds a shadow effect to the chart.

#### 1.8.2 Creating a Pie Chart

## 1.8.3 Pie Chart with Pandas DataFrame

- Pie charts can also be created directly from a Pandas DataFrame using the plot() function with the kind='pie' kwarg.
- The figsize kwarg ensures the pie chart is displayed circular.

# 1.9 Scatter Plots

#### 1.9.1 Introduction to Scatter Plots

- A scatter plot is used to represent data points in a two-dimensional space, with one variable plotted along the x-axis and another along the y-axis.
- The scatter() function in Matplotlib is used to create scatter plots.
- Key customizations for scatter plots include:
  - c: Specifies the colors of the points. Accepts an array of values, a single color, or a colormap.
    - \* 'blue': Standard blue shade.
    - \* 'red': Vibrant red.
    - \* 'green': Natural green.
    - \* 'orange': Bright orange.
    - \* 'purple': Rich purple.
  - marker: Specifies the shape of the markers. Common options include:
    - \* 'o': Circle (default)
    - \* 's': Square
    - \* ' : Triangleup'  $\forall$  ' : Triangledown
    - \* 'x': Cross
    - \* 'D': Diamond
  - s: Specifies the size of the markers.
  - alpha: Adjusts the transparency of the points (range: 0.0 to 1.0).

## 1.9.2 Creating a Basic Scatter Plot

```
import numpy as np
import matplotlib.pyplot as plt
# Generate random data
N = 100
x = np.random.rand(N)
y = np.random.rand(N)
colors = np.random.rand(N) # Random colors
area = np.pi * (15 * np.random.rand(N))**2 # Marker sizes
# Create scatter plot
plt.scatter(x, y, s=area, c=colors, alpha=0.7, marker='o')
plt.title('Basic Scatter Plot')
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
# Adds a color scale legend
plt.colorbar(label='Color Scale')
plt.show()
```

#### 1.9.3 Scatter Plot with Different Markers and Colors

- To differentiate data points, use the c kwarg for colors and marker kwarg for shapes.
- Below is an example with multiple marker styles.

# 1.10 Advanced Chart - Contour Plot

- Contour plots are used to represent three-dimensional data in two dimensions using contour lines or filled regions.
- The contour() function draws contour lines, while the contourf() function fills the areas between the lines with colors.
- The data for contour plots is defined as a grid, with values at each grid point determining the height (or Z-value).
- Key arguments for customization:
  - levels: Specifies the number or specific values of contour levels.
  - cmap: Defines the color map (e.g., 'viridis', 'plasma', 'coolwarm').
  - linewidths: Adjusts the thickness of the contour lines.
  - alpha: Controls the transparency of the plot.

#### Example 1: Basic Contour Plot.

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

# Generate grid data
x = np.linspace(-5, 5, 100)
y = np.linspace(-5, 5, 100)
X, Y = np.meshgrid(x, y)
Z = np.sin(np.sqrt(X**2 + Y**2))

# Create a contour plot
plt.contour(X, Y, Z, levels=10, cmap='viridis')
plt.title('Contour Plot')
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
plt.colorbar(label='Z-value')
plt.show()
```

#### Example 2: Filled Contour Plot.

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

# Generate grid data
x = np.linspace(-5, 5, 100)
y = np.linspace(-5, 5, 100)
X, Y = np.meshgrid(x, y)
Z = np.sin(np.sqrt(X**2 + Y**2))

# Create a filled contour plot
plt.contourf(X, Y, Z, levels=15, cmap='plasma', alpha=0.8)
plt.colorbar(label='Z-value')
plt.title('Filled Contour Plot')
plt.xlabel('X-axis')
plt.ylabel('Y-axis')
plt.show()
```

# 1.11 Advanced Chart - 3D Surfaces with mplot3d

- The mplot3d toolkit in Matplotlib allows for the creation of 3D plots.
- Import Axes3D from mpl\_toolkits.mplot3d to enable 3D plotting capabilities.
- Common types of 3D plots include:
  - 3D Surface Plots: Uses plot\_surface() to represent 3D surfaces based on meshgrid data.
  - Randomly Elevated Surface Plots: Applies random variations to surface heights.
  - **3D Scatter Plots:** Plots individual points in 3D space.
- Use view\_init() to adjust the elevation and azimuthal angle of the 3D view.

#### Example 1: Basic 3D Surface Plot.

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
# Generate grid data
x = np.linspace(-5, 5, 100)
y = np.linspace(-5, 5, 100)
X, Y = np.meshgrid(x, y)
Z = np.sin(np.sqrt(X**2 + Y**2))
# Create a 3D surface plot
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.plot_surface(X, Y, Z, cmap='viridis', edgecolor='none')
ax.set_title('3D Surface Plot')
ax.set_xlabel('X-axis')
ax.set_ylabel('Y-axis')
ax.set_zlabel('Z-axis')
plt.show()
```

#### Example 2: Randomly Elevated 3D Surface Plot.

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
# Generate grid data with random elevations
x = np.linspace(-5, 5, 100)
y = np.linspace(-5, 5, 100)
X, Y = np.meshgrid(x, y)
Z = np.sin(np.sqrt(X**2 + Y**2)) + 0.5 *
   np.random.randn(*X.shape)
# Create a 3D surface plot with random elevations
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.plot_surface(X, Y, Z, cmap='plasma', edgecolor='none',
ax.set_title('Randomly Elevated 3D Surface Plot')
ax.set_xlabel('X-axis')
ax.set_ylabel('Y-axis')
ax.set_zlabel('Z-axis')
plt.show()
```

#### Example 3: 3D Scatter Plot.

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
# Generate random data points for scatter plot
N = 100
\# Random values between -5 and 5
x = np.random.rand(N) * 10 - 5
y = np.random.rand(N) * 10 - 5
z = np.random.rand(N) * 10 - 5
colors = np.random.rand(N)
sizes = 100 * np.random.rand(N)
# Create a 3D scatter plot
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
scatter = ax.scatter(x, y, z, c=colors, s=sizes,
                  cmap='cool', alpha=0.7)
ax.set_title('3D Scatter Plot')
ax.set_xlabel('X-axis')
ax.set_ylabel('Y-axis')
ax.set_zlabel('Z-axis')
plt.colorbar(scatter, label='Color Scale')
plt.show()
```

# 1.12 Subplots Within Other Subplots

- A subplot within another subplot can be created using the add\_axes() function.
- This approach allows you to define a smaller plotting area inside an existing plot by specifying the position and size as a list of relative coordinates.
- Example: Create a main line chart with an inner line chart in the top-right corner.
  - The add\_axes() argument takes a list of four values:
    - \* The x-coordinate of the bottom-left corner of the inner subplot.
    - \* The y-coordinate of the bottom-left corner of the inner subplot.
    - \* The width of the inner subplot (relative to the main figure).
    - \* The height of the inner subplot (relative to the main figure).

Example: Line Chart with an Inner Line Chart in the Top Right.

```
import matplotlib.pyplot as plt
import numpy as np
# Main plot data
x_main = np.linspace(0, 10, 100)
y_main = np.sin(x_main)
# Inner plot data
x_inner = np.linspace(0, 2 * np.pi, 50)
y_inner = np.cos(x_inner)
# Create the main plot
fig, ax_main = plt.subplots()
ax_main.plot(x_main, y_main, label='Main Plot', color='blue')
ax_main.set_title('Main Plot with Inner Subplot')
ax_main.set_xlabel('X-axis')
ax_main.set_ylabel('Y-axis')
ax_main.legend()
# Add the inner subplot
ax_inner = fig.add_axes([0.65, 0.65, 0.25, 0.25]) # [x, y, width,
   height]
ax_inner.plot(x_inner, y_inner, label='Inner Plot', color='red')
ax_inner.set_title('Inner Plot', fontsize=10)
ax_inner.set_xticks([])
ax_inner.set_yticks([])
plt.show()
```