

Python libraries

NumPy

NumPy (Numerical Python) is a powerful library in Python used for numerical computing. It provides support for large, multi-dimensional arrays and matrices, along with a collection of mathematical functions to operate on these data structures efficiently.

Key Features of NumPy:

1. **Ndarray (N-dimensional Array)** – Provides fast and efficient array operations.
2. **Mathematical Functions** – Includes functions for linear algebra, statistics, and Fourier transforms.
3. **Broadcasting** – Enables operations on arrays of different shapes.
4. **Indexing and Slicing** – Similar to Python lists but more powerful.
5. **Integration with Other Libraries** – Used in Pandas, Matplotlib, Scikit-learn, and TensorFlow.

Why Use NumPy Instead of Python Lists?

Feature	NumPy Arrays	Python Lists
Speed	Faster	Slower
Memory Usage	Less	More
Functionality	Supports mathematical operations	Limited operations
Broadcasting	Yes	No

Installing NumPy

If you don't have NumPy installed, you can install it using:

```
//In bash
pip install numpy
```

Basic Operations in NumPy

1. Importing NumPy

```
//In python
import numpy as np
```

2. Creating Arrays

```
//In python
```

```
# Creating a 1D array
arr1 = np.array([1, 2, 3, 4, 5])
print(arr1)
```

```
# Creating a 2D array
arr2 = np.array([[1, 2, 3], [4, 5, 6]])
print(arr2)
```

3. Checking Array Shape and Type

//In python

```
print(arr1.shape) # Output: (5,)
print(arr2.shape) # Output: (2,3)
print(arr1.dtype) # Output: int32 or int64 (depends on system)
```

4. Creating Special Arrays

//In python

```
# Array of zeros
zeros = np.zeros((3, 3))
print(zeros)
```

```
# Array of ones
ones = np.ones((2, 2))
print(ones)
```

```
# Identity matrix
identity = np.eye(3)
print(identity)
```

```
# Random numbers
random_array = np.random.rand(3, 3)
print(random_array)
```

5. Array Operations

//In python

```
a = np.array([1, 2, 3])
b = np.array([4, 5, 6])
```

```
# Element-wise addition
print(a + b) # Output: [5 7 9]
```

```
# Element-wise multiplication
print(a * b) # Output: [4 10 18]
```

```
# Dot Product
print(np.dot(a, b)) # Output: 32 (1*4 + 2*5 + 3*6)
```

6. Indexing & Slicing

//In python

```
arr = np.array([10, 20, 30, 40, 50])
print(arr[2]) # Output: 30
print(arr[1:4]) # Output: [20 30 40]
print(arr[::-1]) # Reverse array
```

7. Reshaping Arrays

//In python

```
arr = np.array([1, 2, 3, 4, 5, 6])
reshaped_arr = arr.reshape(2, 3)
print(reshaped_arr)
```

```
# Output:  
# [[1 2 3]  
#   [4 5 6]]
```

8. Statistical Functions

```
//In python
```

```
arr = np.array([10, 20, 30, 40, 50])  
  
print(np.mean(arr)) # Mean: 30.0  
print(np.median(arr)) # Median: 30.0  
print(np.std(arr)) # Standard Deviation  
print(np.sum(arr)) # Sum of elements: 150  
print(np.min(arr)) # Minimum value: 10  
print(np.max(arr)) # Maximum value: 50
```

9. Create a Random Array

```
//In python
```

```
rand_arr = np.random.rand(4, 4) # 4x4 matrix with random values  
print("Random Array:\n", rand_arr)
```

10. Reshape an Array

```
//In python
```

```
arr = np.arange(1, 10) # Array from 1 to 9  
reshaped = arr.reshape(3, 3) # Reshape into 3x3  
print("Reshaped Array:\n", reshaped)
```

8. Statistical Functions

```
//In python
```

```
arr = np.array([10, 20, 30, 40, 50])  
  
print(np.mean(arr)) # Mean: 30.0  
print(np.median(arr)) # Median: 30.0  
print(np.std(arr)) # Standard Deviation  
print(np.sum(arr)) # Sum of elements: 150  
print(np.min(arr)) # Minimum value: 10  
print(np.max(arr)) # Maximum value: 50
```

11. Matrix multiplication

```
import numpy as np  
A = np.array([[1, 2], [3, 4]])  
B = np.array([[5, 6], [7, 8]])  
result = np.dot(A, B)  
print("Matrix Multiplication Result:\n", result)
```

12. Generating Random Data for Testing

```
//In python
import numpy as np

# Generate random data (100 rows, 5 columns)
data = np.random.rand(100, 5)

print(data[:5]) # Display first 5 rows
```

13. Convert a coloured image to numpy array

```
import numpy as np
from PIL import Image

# Load an image and convert to a numpy array
img = Image.open('sample.jpg')
img_array = np.array(img)

# Convert to grayscale
gray = 0.2989 * img_array[:, :, 0] + 0.5870 * img_array[:, :, 1] + 0.1140 * img_array[:, :, 2]

//• img_array[:, :, 0] – Extracts the Red channel values

    • img_array[:, :, 1] – Extracts the Green channel values
    • img_array[:, :, 2] – Extracts the Blue channel values

# Convert back to an image and display
gray_img = Image.fromarray(gray.astype('uint8'))
gray_img.show()
```

PANDAS

The **Pandas** library in Python is used for data manipulation and analysis. It provides powerful data structures like **Series** and **DataFrame** to handle and analyze structured data efficiently.

◆ Installation

Install Pandas using pip:

```
//in bash

pip install pandas
```

◆ Main Data Structures

1. **Series** – One-dimensional array-like object.
2. **DataFrame** – Two-dimensional table with rows and columns (like a spreadsheet).

Creating a Series

```
//In python

import pandas as pd

data = [10, 20, 30, 40]
series = pd.Series(data)
print(series)
```

Output:

```
0    10
1    20
2    30
3    40
dtype: int64
```

Creating a DataFrame

In **Pandas**, a **DataFrame** is a two-dimensional, tabular data structure with labeled rows and columns (like a spreadsheet or SQL table).

```
//In python
data = {
    'Name': ['Alice', 'Bob', 'Charlie'],
    'Age': [25, 30, 35],
    'City': ['New York', 'Los Angeles', 'Chicago']
}
df = pd.DataFrame(data)
print(df)
```

Output:

	Name	Age	City
0	Alice	25	New York
1	Bob	30	Los Angeles
2	Charlie	35	Chicago

Reading/Writing Data

- **Read from CSV:**

```
//In python
df = pd.read_csv('data.csv')
```

- **Write to CSV:**

```
//In python
df.to_csv('output.csv', index=False)
```

1. Write to CSV

You can create a DataFrame and write it to a CSV file using `to_csv()`.

Example: Write to CSV

```
//In python

import pandas as pd

# Create a DataFrame
data = {
    'Name': ['Alice', 'Bob', 'Charlie'],
    'Age': [25, 30, 35],
    'City': ['New York', 'Los Angeles', 'Chicago']
}

df = pd.DataFrame(data)

# Write to CSV
df.to_csv('data.csv', index=False)

print("Data written to data.csv")
```

🔗 **Output:** A CSV file `data.csv` will be created with the following content:

```
Name, Age, City
Alice, 25, New York
Bob, 30, Los Angeles
Charlie, 35, Chicago
```

2. Read from CSV

You can read a CSV file using `read_csv()`.

Example: Read from CSV

```
//In python
import pandas as pd

# Read CSV file
df = pd.read_csv('data.csv')

# Display the data
print(df)
```

🔗 **Output:**

	Name	Age	City
0	Alice	25	New York
1	Bob	30	Los Angeles

2 Charlie 35 Chicago

3. Append New Data to CSV

You can add new data to an existing CSV using `mode='a'` (append mode).

Example: Append Data

```
//In python
new_data = {'Name': ['David'], 'Age': [40], 'City': ['Houston']}
new_df = pd.DataFrame(new_data)

# Append to existing CSV (without header)
new_df.to_csv('data.csv', mode='a', index=False, header=False)

print("Data appended to data.csv")
```

🔗 **Output:** data.csv now contains

```
Name, Age, City
Alice, 25, New York
Bob, 30, Los Angeles
Charlie, 35, Chicago
David, 40, Houston
```

4. Write CSV Without Index

You can remove the index while writing to CSV using `index=False`.

Example: Write Without Index

```
//In python
df.to_csv('data.csv', index=False)
```

5. Read CSV with Specific Columns

You can read only specific columns from a CSV file.

Example: Read Specific Columns

```
//In python
df = pd.read_csv('data.csv', usecols=['Name', 'Age'])
print(df)
```

🔗 **Output:**

	Name	Age
0	Alice	25
1	Bob	30
2	Charlie	35
3	David	40

6. Read CSV Without Headers

You can read CSV files without headers using `header=None`.

Example: Read Without Headers

```
//In python
df = pd.read_csv('data.csv', header=None)
print(df)
```

🔗 Output:

	0	1	2
0	Name	Age	City
1	Alice	25	New York
2	Bob	30	Los Angeles
3	Charlie	35	Chicago
4	David	40	Houston

7. Save CSV with a Custom Separator

You can change the separator (like `;` or `|`) using the `sep` parameter.

Example: Write with Custom Separator

```
//In python
df.to_csv('data.csv', sep='|', index=False)
```

🔗 Output:

Name	Age	City
Alice	25	New York
Bob	30	Los Angeles
Charlie	35	Chicago
David	40	Houston

Selecting Data

- Select column:

```
//In python
print(df['Name'])
```

- Select row by index:

```
//In python
print(df.loc[0])
```

- Select row by index position:

```
// In python
print(df.iloc[0])
```

Filtering Data

```
//In python
filtered = df[df['Age'] > 25]
print(filtered)
```

Updating Data

```
//In python
df.loc[0, 'Age'] = 28
```

Adding New Column

```
//In python
df['Country'] = 'USA'
```

Deleting a Column

```
//In python
df.drop('City', axis=1, inplace=True)
```

Descriptive Statistics

```
//In python
print(df.describe())
```

Handling Missing Data

- Drop missing rows:

```
//In python
df.dropna(inplace=True)
```

Fill missing values:

```
//In python
df.fillna(0, inplace=True)
```

Drop rows with missing values using `dropna()`

This program creates a DataFrame, introduces some missing values (NaN), and removes rows with missing values using `dropna()`

```
import pandas as pd

# Sample data with missing values
data = {'Name': ['John', 'Anna', 'Peter', 'Linda', None],
        'Age': [28, 22, None, 32, 29],
        'City': ['New York', 'Paris', 'London', None, 'Tokyo']}

df = pd.DataFrame(data)

print("Original DataFrame:")
print(df)

# Remove rows with missing values
df.dropna(inplace=True)

print("\nDataFrame after dropping rows with missing values:")
print(df)
```

Output:

```
Original DataFrame:
   Name  Age  City
0  John  28.0 New York
1  Anna  22.0   Paris
2 Peter   NaN  London
3 Linda  32.0    None
4  None  29.0   Tokyo
```

```
DataFrame after dropping rows with missing values:
   Name  Age  City
0  John  28.0 New York
1  Anna  22.0   Paris
```

2. Fill missing values using `fillna()`

This program creates a DataFrame and fills missing values using `fillna()`.

```
import pandas as pd
```

```
# Sample data with missing values
data = {'Name': ['John', 'Anna', 'Peter', 'Linda', None],
        'Age': [28, 22, None, 32, 29],
        'City': ['New York', 'Paris', 'London', None, 'Tokyo']}

df = pd.DataFrame(data)

print("Original DataFrame:")
print(df)

# Fill missing values with default values
df.fillna({'Name': 'Unknown', 'Age': 0, 'City': 'Unknown'}, inplace=True)

print("\nDataFrame after filling missing values:")
print(df)
```

Output:

Original DataFrame

Functions

1. Creating Data

Function	Description	Example
<code>pd.Series()</code>	Creates a Pandas Series	<code>s = pd.Series([1, 2, 3])</code>
<code>pd.DataFrame()</code>	Creates a DataFrame	<code>df = pd.DataFrame({'A': [1, 2], 'B': [3, 4]})</code>
<code>pd.read_csv()</code>	Reads data from a CSV file	<code>df = pd.read_csv('data.csv')</code>
<code>pd.read_excel()</code>	Reads data from an Excel file	<code>df = pd.read_excel('data.xlsx')</code>
<code>pd.read_json()</code>	Reads data from a JSON file	<code>df = pd.read_json('data.json')</code>

2. Viewing Data

Function	Description	Example
<code>df.head()</code>	Displays the first 5 rows	<code>df.head()</code>
<code>df.tail()</code>	Displays the last 5 rows	<code>df.tail()</code>
<code>df.info()</code>	Displays DataFrame info	<code>df.info()</code>
<code>df.describe()</code>	Displays statistical summary	<code>df.describe()</code>
<code>df.shape</code>	Returns dimensions of DataFrame	<code>df.shape</code>
<code>df.columns</code>	Returns column names	<code>df.columns</code>

1. `df.head()` – Displays the first 5 rows

Returns the top 5 rows of the DataFrame. You can pass a number to display a specific number of rows.

Example:

```
import pandas as pd

# Sample DataFrame
data = {
    'Name': ['John', 'Anna', 'Mike', 'Tom', 'Sara', 'Alex'],
    'Age': [25, 30, 22, 28, 24, 27],
    'City': ['NY', 'LA', 'SF', 'TX', 'DC', 'CHI']
}

df = pd.DataFrame(data)

# Display first 5 rows
print(df.head())
```

Output:

	Name	Age	City
0	John	25	NY
1	Anna	30	LA
2	Mike	22	SF
3	Tom	28	TX
4	Sara	24	DC

☞ You can display the first **n** rows using `df.head(n)`.

2. `df.tail()` – Displays the last 5 rows

Returns the bottom 5 rows of the DataFrame. You can pass a number to display a specific number of rows.

Example:

```
# Display last 5 rows
print(df.tail())
```

Output:

	Name	Age	City
1	Anna	30	LA
2	Mike	22	SF
3	Tom	28	TX
4	Sara	24	DC
5	Alex	27	CHI

☞ You can display the last **n** rows using `df.tail(n)`.

3. `df.info()` – Displays DataFrame information

Returns a summary of the DataFrame, including:

✓ Number of rows and columns

- ✓ Column names and data types
- ✓ Non-null values in each column

Example:

```
# Display DataFrame information
df.info()
```

Output:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6 entries, 0 to 5
Data columns (total 3 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0   Name    6 non-null        object
 1   Age     6 non-null        int64
 2   City    6 non-null        object
dtypes: int64(1), object(2)
memory usage: 272.0 bytes
```

🔗 Useful for checking data types, missing values, and data structure.

🔗 3. Selecting Data

Function	Description	Example
<code>df['column']</code>	Select a column	<code>df['Age']</code>
<code>df[['col1', 'col2']]</code>	Select multiple columns	<code>df[['Age', 'Name']]</code>
<code>df.loc[]</code>	Select rows by label	<code>df.loc[0]</code>
<code>df.iloc[]</code>	Select rows by index	<code>df.iloc[0]</code>
<code>df.at[]</code>	Fast access to a single value by label	<code>df.at[0, 'Age']</code>
<code>df.iat[]</code>	Fast access to a single value by index	<code>df.iat[0, 1]</code>

🔗 4. Filtering Data

Function	Description	Example
<code>df[df['Age'] > 25]</code>	Filter rows based on a condition	<code>df[df['Age'] > 25]</code>
<code>df.query()</code>	Query data	<code>df.query('Age > 25')</code>
<code>df['column'].isin()</code>	Filter by values in a list	<code>df[df['City'].isin(['New York', 'Los Angeles'])]</code>

🔗 5. Adding and Removing Data

Function	Description	Example
<code>df['new_col'] = value</code>	Add a new column	<code>df['Country'] = 'USA'</code>
<code>df.drop()</code>	Remove columns or rows	<code>df.drop('Age', axis=1)</code>

Function	Description	Example
Program using insert function <pre>import pandas as pd # Create a sample DataFrame data = { 'Name': ['John', 'Alice', 'Bob'], 'Age': [25, 30, 22] } df = pd.DataFrame(data) print("Original DataFrame:") print(df) # Insert a new column at index 1 (after 'Name') df.insert(1, 'City', ['New York', 'Los Angeles', 'Chicago']) print("\nDataFrame after inserting 'City' column:") print(df)</pre>		
<code>df.insert()</code>	Insert column at specific position	
<code>df.append()</code>	Append rows	<code>df.append(new_row)</code>
<code>df.rename()</code>	Rename columns	<code>df.rename(columns={'A': 'Alpha'})</code>

🔗 6. Handling Missing Data

Function	Description	Example
<code>df.isnull()</code>	Check for missing values	<code>df.isnull()</code>
<code>df.dropna()</code>	Drop rows with missing values	<code>df.dropna()</code>
<code>df.fillna()</code>	Fill missing values	<code>df.fillna(0)</code>
<code>df.interpolate()</code>	Fill missing values using interpolation	<code>df.interpolate()</code>

```
import pandas as pd
import numpy as np

df = pd.DataFrame({
    'A': [1, np.nan, 3, np.nan, 5],
    'B': [np.nan, 2, np.nan, 4, 5]
})

df_interpolated = df.interpolate()
print(df_interpolated)
```

Function	Description	Example
<p><code>np.nan</code> is a special constant in NumPy that represents "Not a Number", used to denote missing or undefined values in numerical arrays or pandas DataFrames.</p>		

Key Facts:

- `np.nan` stands for **"Not a Number"**.
- It is of **float** type.
- Used frequently in **pandas** and **NumPy** to represent **missing data**.

7. Grouping and Aggregation

Function	Description	Example
<code>df.groupby()</code>	Group data	<code>df.groupby('City').mean()</code>
<code>df.agg()</code>	Aggregate data	<code>df.agg({'Age': 'mean'})</code>
<code>df.pivot_table()</code>	Create a pivot table	<code>df.pivot_table(index='City', values='Age', aggfunc='mean')</code>
<code>df.cumsum()</code>	Cumulative sum	<code>df['Age'].cumsum()</code>

```
import pandas as pd
data = {
    'Name': ['John', 'Alice', 'Bob'],
    'Age': [25, 30, 22],
    'City': ['New York', 'Los Angeles', 'New York']
}
df = pd.DataFrame(data)

result = df.pivot_table(index='City', values='Age', aggfunc='mean')
print(result)
```

8. Sorting and Ranking

Function	Description	Example
<code>df.sort_values()</code>	Sort by values	<code>df.sort_values(by='Age')</code>
<code>df.sort_index()</code>	Sort by index	<code>df.sort_index()</code>
<code>df.rank()</code>	Rank values	<code>df['Age'].rank()</code>

9. Merging and Joining

Function	Description	Example
<code>pd.concat()</code>	Concatenate DataFrames	<code>pd.concat([df1, df2])</code>

Function`df.merge()``df.join()`**Description**

Merge
DataFrames

Join DataFrames

Example

```
df1.merge(df2,  
on='ID')
```

```
df1.join(df2,  
on='ID')
```


1. `pd.concat()` – Concatenate DataFrames

Used to combine two or more DataFrames along rows (`axis=0`) or columns (`axis=1`).

Example:

```
import pandas as pd

# Create two sample DataFrames
data1 = {'ID': [1, 2, 3], 'Name': ['John', 'Anna', 'Mike']}
data2 = {'ID': [4, 5, 6], 'Name': ['Tom', 'Sara', 'Alex']}

df1 = pd.DataFrame(data1)
df2 = pd.DataFrame(data2)

# Concatenate along rows (default axis=0)
result = pd.concat([df1, df2])

print(result)
```

Output:

	ID	Name
0	1	John
1	2	Anna
2	3	Mike
0	4	Tom
1	5	Sara
2	6	Alex

Note: Indexes are not reset after concatenation.

2. `merge()` – Merge DataFrames based on a common column

Used to combine DataFrames like an SQL JOIN based on common keys.

Example:

```
data1 = {'ID': [1, 2, 3], 'Name': ['John', 'Anna', 'Mike']}
data2 = {'ID': [1, 2, 4], 'Age': [25, 30, 28]}

df1 = pd.DataFrame(data1)
df2 = pd.DataFrame(data2)

# Merge on 'ID' column (INNER JOIN)
```

Function	Description	Example
----------	-------------	---------

```
result = df1.merge(df2, on='ID')

print(result)
```

Output:

	ID	Name	Age
0	1	John	25
1	2	Anna	30

3. join() – Join DataFrames based on the index

Used to combine DataFrames on the index or a key column.

Example:

```
data1 = {'Name': ['John', 'Anna', 'Mike'], 'Age': [25, 30, 22]}
data2 = {'City': ['NY', 'LA', 'SF']}

df1 = pd.DataFrame(data1, index=[1, 2, 3])
df2 = pd.DataFrame(data2, index=[1, 2, 3])

# Join on index
result = df1.join(df2)

print(result)
```

Output:

	Name	Age	City
1	John	25	NY
2	Anna	30	LA
3	Mike	22	SF

Note: join() is similar to merge() but works based on index by default.

Use on='column' to join on a specific column instead of the index.

🔗 10. Exporting Data

Function	Description	Example
df.to_csv()	Save to CSV	df.to_csv('output.csv')
df.to_excel()	Save to Excel	df.to_excel('output.xlsx')
df.to_json()	Save to JSON	df.to_json('output.json')

Example workflow:

```
import pandas as pd

# Create a DataFrame
data = {'Name': ['Alice', 'Bob', 'Charlie'], 'Age': [25, 30, 35]}
df = pd.DataFrame(data)

# View the data
print("Initial DataFrame:")
print(df.head())

# Add a column
df['Country'] = 'USA'

# Filter data
filtered = df[df['Age'] > 28]
print("\nFiltered DataFrame (Age > 28):")
print(filtered)

# Group data
grouped = df.groupby('Country').mean(numeric_only=True)

# Export to CSV
df.to_csv('output5.csv', index=False)

print("\nGrouped Data (mean by Country):")
print(grouped)
```

MATPLOTLIB

Matplotlib is a popular Python library used for data visualization. It allows you to create a wide range of plots, including line plots, bar plots, scatter plots, histograms, and more.

Install `matplotlib` using `pip`:

```
pip install matplotlib
```

1.Example of a line plot using `matplotlib`

```
import matplotlib.pyplot as plt

# Data
x = [1, 2, 3, 4, 5]
y = [2, 4, 1, 3, 5]

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

```
# Add title and labels
plt.title("Simple Line Plot")
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
```

```
# Show plot
plt.show()
```

Common Plot Types with Examples

1. Line Plot

Used to visualize trends over time or continuous data.

```
//In python

plt.plot([1, 2, 3, 4], [10, 20, 25, 30], marker='o', linestyle='--',
color='b')
plt.title('Line Plot Example')
plt.show()
```

2. Bar Plot

Used to compare categorical data.

```
//In python

categories = ['A', 'B', 'C', 'D']
values = [4, 7, 1, 8]

plt.bar(categories, values, color='skyblue')
plt.title('Bar Plot Example')
plt.show()
```

3. Scatter Plot

Used to show the relationship between two variables

```
x = [5, 7, 8, 7, 2, 17, 2, 9]
y = [99, 86, 87, 88, 100, 86, 103, 87]

plt.scatter(x, y, color='red')
plt.title('Scatter Plot Example')
plt.show()
```

4. Histogram

Used to show the distribution of data.

```
//In python
data = [22, 87, 5, 43, 56, 73, 55, 54, 11, 20, 51, 5, 79, 31, 27]
plt.hist(data, bins=5, color='green', edgecolor='black')
plt.title('Histogram Example')
plt.show()
```

5. Pie Chart

Used to show proportions.

```
//In python
labels = ['Python', 'Java', 'C++', 'JavaScript']
sizes = [45, 25, 15, 15]
colors = ['blue', 'red', 'yellow', 'green']

plt.pie(sizes, labels=labels, colors=colors, autopct='%1.1f%%',
startangle=90)
plt.title('Pie Chart Example')
plt.show()
```

6. Multiple Plots in One Figure

Use `plt.subplot()` to create multiple plots in one figure.

```
//In python
# First plot
plt.subplot(1, 2, 1)
plt.plot([1, 2, 3], [4, 5, 6])
plt.title('Plot 1')

# Second plot
plt.subplot(1, 2, 2)
plt.bar(['A', 'B', 'C'], [3, 5, 7])
plt.title('Plot 2')

plt.tight_layout()
plt.show()
```

Customization Tips

- **Color:** 'red', 'blue', '#ff5733'
- **Line style:** '-', '--', '-.', ':'
- **Markers:** 'o', '^', 's', 'd'
- **Labels:** `plt.xlabel()`, `plt.ylabel()`, `plt.title()`
- **Legend:** `plt.legend()`
- **Grid:** `plt.grid()`
- **Figure size:** `plt.figure(figsize=(10, 5))`

Seaborn