

UNIT = 5

Introduction to NumPy

NumPy (Numerical Python) is a powerful Python library used for **numerical and scientific computing**. It provides support for **large, multi-dimensional arrays** and **mathematical functions** to operate on these arrays efficiently.

Why NumPy is important

Normal Python lists are slow and limited for heavy calculations.

NumPy solves this by providing:

- Fast execution (written in C internally)
- Easy handling of large datasets
- Built-in mathematical functions

That's why NumPy is widely used in:

- Data Science
- Machine Learning
- AI
- Scientific Computing
- Image Processing

Core feature: NumPy Array

The main object in NumPy is the **ndarray (N-dimensional array)**.

Example (Python list vs NumPy array)

Example (Python list vs NumPy array)

python

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```
# Python list
a = [1, 2, 3, 4]
b = [5, 6, 7, 8]

# This will NOT work as expected
print(a + b)  # [1,2,3,4,5,6,7,8]
```

python

 Copy code

```
import numpy as np

# NumPy array
a = np.array([1, 2, 3, 4])
b = np.array([5, 6, 7, 8])

print(a + b)  # [ 6  8 10 12 ]
```

 NumPy performs **element-wise operations** automatically.

Key features of NumPy

1. Fast performance

- Uses optimized C code
- Much faster than Python loops

2. Multidimensional arrays

You can create:

- 1D arrays (vectors)
- 2D arrays (matrices)
- 3D+ arrays (tensors)

```
arr = np.array([[1, 2], [3, 4]])
```

3. Mathematical operations

NumPy supports:

- Addition, subtraction, multiplication, division
- Trigonometry

- Statistics

```
python
arr = np.array([1, 2, 3, 4])

print(np.mean(arr))    # Average
print(np.sum(arr))    # Sum
print(np.sqrt(arr))   # Square root
```

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4. Broadcasting

You can perform operations between arrays of different shapes.

```
python
arr = np.array([1, 2, 3])
print(arr + 10)  # [11 12 13]
```

 Copy code

5. Used as a base for other libraries

Many popular libraries are built on NumPy:

- Pandas
- Matplotlib
- Scikit-learn
- TensorFlow
- PyTorch

Installing NumPy

```
bash
pip install numpy
```

 Copy code

Simple real-life example

Imagine you have marks of students and want to calculate average:

```
python
import numpy as np

marks = np.array([78, 85, 90, 88, 76])
average = np.mean(marks)

print("Average marks:", average)
```

Introduction to NumPy

NumPy (Numerical Python) is a powerful Python library used for **numerical and scientific computing**. It provides support for **large, multi-dimensional arrays** and **mathematical functions** to operate on these arrays efficiently.

Unlike Python lists, NumPy arrays:

- Are **faster**
- Consume **less memory**
- Support **vectorized operations**

NumPy is widely used in:

- Data Science
- Machine Learning
- Scientific calculations
- Image processing

The core object of NumPy is the **ndarray (N-dimensional array)**.

Creating Arrays in NumPy

NumPy arrays can be created in **two main ways**:

1. Using **existing arrays (lists/tuples)**
 2. Using **scalars and built-in functions**
-

1. Creating Arrays Using Existing Arrays

We can create NumPy arrays from:

- Python lists
- Tuples

1D Array (Vector)

```
python

import numpy as np

arr = np.array([10, 20, 30, 40])
print(arr)
```

Output:

```
csharp
```

```
[10 20 30 40]
```

2D Array (Matrix)

```
python
```

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

Output:

```
lua
```

```
[[1 2 3]
 [4 5 6]]
```

3D Array

```
python
```

```
arr = np.array([[[1, 2], [3, 4]],
                [[5, 6], [7, 8]]])
```

Key Points

- All elements must be of the **same data type**
- NumPy automatically converts data types if needed
- Arrays support **element-wise operations** 

Key Points

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- NumPy automatically converts data types if needed
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2. Creating Arrays Using Scalars

A **scalar** is a single value (number). NumPy provides functions to create arrays using scalars.

The screenshot shows a Jupyter Notebook interface with two sections: "Array of Zeros" and "Array of Ones".

Array of Zeros

python

```
arr = np.zeros(5)
print(arr)
```

Output:

csharp

```
[0. 0. 0. 0. 0.]
```

Array of Ones

python

```
arr = np.ones(4)
print(arr)
```

Output:

csharp

```
[1. 1. 1. 1.]
```

Constant Value Array

python

Copy code

```
arr = np.full(5, 7)
print(arr)
```

Output:

csharp

Copy code

```
[7 7 7 7 7]
```

Range of Values

python

Copy code

```
arr = np.arange(1, 10, 2)
print(arr)
```

Output:

csharp

Copy code

```
[1 3 5 7 9]
```

Equally Spaced Values

python

```
arr = np.linspace(1, 5, 4)
print(arr)
```

Output:

csharp

```
[1. 2.33 3.67 5.]
```

Identity Matrix

python

```
arr = np.eye(3)
print(arr)
```

Output:

lua

```
[[1. 0. 0.]
 [0. 1. 0.]
 [0. 0. 1.]]
```

Introduction to Pandas

Pandas is a Python library built on NumPy that is used for **data manipulation and analysis**.

While NumPy handles **numerical arrays**, Pandas handles **structured data** like:

- Tables
- CSV files
- Excel sheets
- Databases

Main Data Structures in Pandas

1. Series

- One-dimensional labeled array
- Similar to a column in a table

```
python

import pandas as pd

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

2. DataFrame

- Two-dimensional table (rows & columns)
- Similar to an Excel sheet

```
python

data = {
    "Name": ["Amit", "Rahul", "Neha"],
    "Marks": [85, 90, 88]
}

df = pd.DataFrame(data)
print(df)
```

Why Pandas is Used

- Easy data cleaning
- Fast data filtering
- Handling missing values
- Works smoothly with NumPy

Relationship Between NumPy and Pandas

NumPy

Works with arrays

Faster for math

No labels

Low-level operations

Pandas

Works with tables

Better for data analysis

Labeled data

High-level operations

What is Pandas?

Pandas is an **open-source Python library** used for **data manipulation and data analysis**.

It provides **easy-to-use data structures** that help us store, clean, modify, and analyze **structured data** efficiently.

Pandas is built **on top of NumPy**, which means it uses NumPy internally for fast numerical operations, but adds **labels (rows and columns)** to make data easier to understand and work with.

👉 In simple words:

Pandas helps us work with data in table form (like Excel) using Python.

Key Features of Pandas

- Works with **tabular data**
 - Handles **missing values**
 - Easy **filtering, sorting, and grouping**
 - Fast and efficient
 - Supports multiple data formats (CSV, Excel, SQL, JSON)
-

Main Data Structures in Pandas

1. Series

- One-dimensional labeled data
- Similar to a single column in a table

```
import pandas as pd

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

 Copy code

2. DataFrame

- Two-dimensional data (rows & columns)
- Similar to an Excel spreadsheet

```
python
```

 Copy code

```
data = {
    "Name": ["Amit", "Rahul", "Neha"],
    "Marks": [85, 90, 88]
}

df = pd.DataFrame(data)
print(df)
```

Where is Pandas Used?

1. Data Analysis

- Analyzing large datasets
- Finding patterns and trends
- Calculating statistics

Example:

```
python
df["Marks"].mean()
```

 Copy code



2. Data Cleaning

- Removing duplicate data
- Filling missing values
- Correcting wrong data

Example:

```
df.fillna(0)
```

3. Data Science & Machine Learning

- Preparing data before training models
- Feature selection and transformation

Used with:

- Scikit-learn
 - TensorFlow
 - PyTorch
-

4. Business & Financial Analysis

- Sales reports
 - Profit/loss analysis
 - Market trend analysis
-

5. Working with Files

Pandas can read and write:

- CSV files
- Excel files
- JSON data
- SQL databases

```
df = pd.read_csv("data.csv")
```

6. Academic & Research Work

- Scientific data analysis
 - Research statistics
 - Survey data processing
-

Difference Between NumPy and Pandas

NumPy

Works with arrays

Faster for math

No labels

Numerical data only

Pandas

Works with tables

Better for data analysis

Labeled rows & columns

Mixed data types

Series in Pandas

What is a Series?

A **Series** in Pandas is a **one-dimensional labeled array** that can store:

- Integers
- Floats
- Strings
- Boolean values

Each value in a Series has an **index (label)**.

👉 In simple words:

A Pandas Series is like a single column of a table with labels.

Creating a Series

1. Series from a list

```
python Copy code
import pandas as pd

s = pd.Series([10, 20, 30, 40])
print(s)

Output:
go Copy code

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

2. Series with custom index

```
python Copy code
s = pd.Series([85, 90, 88], index=["Amit", "Rahul", "Neha"])
print(s)

Output:
nginx Copy code

Amit    85
Rahul   90
Neha    88
```

3. Series from dictionary

```
python
```

 Copy code

```
data = {"Math": 80, "Science": 85, "English": 78}  
s = pd.Series(data)  
print(s)
```

Accessing Series Elements

```
python
```

 Copy code

```
print(s["Math"])  
print(s[0])
```

Operations on Series

```
python
```

 Copy code

```
s = pd.Series([10, 20, 30])  
  
print(s + 5)  
print(s * 2)  
print(s.mean())
```

Features of Pandas Series

- One-dimensional
- Indexed data
- Supports mathematical operations
- Handles missing values (NaN)
- Built on NumPy

Matplotlib

What is Matplotlib?

Matplotlib is a **Python plotting library** used for **data visualization**.

It helps us represent data using **graphs and charts**.

👉 In simple words:

Matplotlib is used to draw graphs in Python.

Where is Matplotlib Used?

- Data analysis
 - Data science
 - Machine learning
 - Academic projects
 - Business reports
-

Basic Plot Using Matplotlib

```
python
import matplotlib.pyplot as plt

x = [1, 2, 3, 4]
y = [10, 20, 25, 30]

plt.plot(x, y)
plt.xlabel("X Axis")
plt.ylabel("Y Axis")
plt.title("Simple Line Graph")
plt.show()
```

[Copy code](#)

Common Types of Plots

1. Line Plot

Used to show trends.

```
python
plt.plot([1,2,3],[10,20,30])
plt.show()
```

[Copy code](#)

2. Bar Chart

Used for comparison.

python

 Copy code

```
plt.bar(["A", "B", "C"], [20, 35, 30])
plt.show()
```

3. Histogram

Used for data distribution.

python

 Copy code

```
plt.hist([1,2,2,3,3,3,4])
plt.show()
```

4. Scatter Plot

Used to show relationship between data.

python

 Copy code

```
plt.scatter([1,2,3],[4,5,6])
plt.show()
```

Difference Between Pandas Series and Matplotlib

Pandas Series

Data structure

Stores data

Built on NumPy

Used for analysis

Matplotlib

Visualization tool

Displays data

Built on NumPy

Used for plotting

Python for Data Visualization

What is Data Visualization?

Data Visualization is the process of **representing data graphically** using charts, graphs, and plots so that patterns, trends, and insights are easy to understand.

👉 Instead of reading numbers, we **see the data visually**.

Why Python for Data Visualization?

Python is widely used for data visualization because:

- Easy to learn and use
 - Powerful visualization libraries
 - Works well with data analysis tools
 - Used in industry and academics
-

Popular Python Libraries for Data Visualization

1. Matplotlib

- Basic and most popular library
- Used for simple to complex plots

2. Seaborn

- Built on Matplotlib
- Used for statistical visualization
- Better design and themes

3. Pandas (Built-in Plotting)

- Quick plotting from Series & DataFrames
- Uses Matplotlib internally

4. Plotly

- Interactive graphs
 - Used in dashboards and web apps
-

1. Matplotlib for Data Visualization

Line Plot

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

```
x = [1, 2, 3, 4]
```

```
y = [10, 20, 25, 30]
```

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

```
plt.xlabel("X Axis")
```

```
plt.ylabel("Y Axis")
```

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

```
plt.show()
```

Used for: trends over time

Bar Chart

```
plt.bar(["A", "B", "C"], [20, 35, 30])
```

```
plt.title("Bar Chart")
```

```
plt.show()
```

Used for: comparison between categories

Histogram

```
plt.hist([10, 20, 20, 30, 30, 30])
```

```
plt.title("Histogram")
```

```
plt.show()
```

Used for: data distribution

Scatter Plot

```
plt.scatter([1, 2, 3], [4, 5, 6])
```

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

```
plt.show()
```

Used for: relationship between variables

2. Pandas for Visualization

Pandas makes visualization easier when data is in **Series or DataFrame** form.

Series Plot

```
import pandas as pd
```

```
s = pd.Series([10, 20, 30, 40])
```

```
s.plot()
```

```
plt.show()
```

DataFrame Plot

```
data = {
```

```
    "Math": [80, 85, 90],
```

```
    "Science": [75, 88, 92]
```

```
}
```

```
df = pd.DataFrame(data)
```

```
df.plot()
```

```
plt.show()
```

3. Seaborn for Visualization

Seaborn provides **attractive and statistical plots**.

```
import seaborn as sns
```

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

```
sns.barplot(x=["A","B","C"], y=[10,20,30])  
plt.show()
```

4. Plotly for Interactive Visualization

Plotly allows zooming, hovering, and interaction.

```
import plotly.express as px
```

```
df = px.data.iris()  
fig = px.scatter(df, x="sepal_width", y="sepal_length")  
fig.show()
```

Applications of Python Data Visualization

- Data Science
 - Machine Learning
 - Business analytics
 - Academic research
 - Dashboard creation
-

Advantages of Python Visualization

- Easy to learn
- High flexibility
- Supports large datasets
- Strong community support