

NumPy (Numerical Python) is a fundamental Python library for data science (DS) and machine learning. It provides fast, powerful, and flexible tools for handling large datasets efficiently.

#### **Array:**

- **Fixed Size** – The size of an array is usually defined at the time of creation.
- **Same Data Type** – All elements in an array must be of the same type.
- **Indexed Access** – Elements are accessed using an index, starting from 0 (zero-based index in most programming languages).
- **Efficient Retrieval** – Arrays allow fast access to elements using their index.

## **Why Use NumPy in Data Science?**

1. **Performance:** NumPy arrays are much faster than Python lists because they use fixed memory allocation and optimized C-based operations.
2. **Convenience:** It provides built-in mathematical functions for complex operations.
3. **Efficiency:** NumPy uses less memory than lists and enables vectorized operations (operations on entire arrays instead of element-by-element loops).

### **1. Installing and Importing NumPy**

```
pip install numpy
```

Then, import it in your Python script:

```
import numpy as np
```

### **2. Creating NumPy Arrays**

A NumPy array (ndarray) is similar to a list but more efficient.

### **From a Python list**

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

### **Multi-dimensional Array**

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

### **Array with all zeros**

```
zeros = np.zeros((3, 3)) # 3x3 matrix of zeros  
print(zeros)
```

### **Array with all ones**

```
ones = np.ones((2, 4)) # 2x4 matrix of ones  
print(ones)
```

### **Random Numbers**

```
random_arr = np.random.rand(3, 3) # 3x3 matrix with random values  
print(random_arr)
```

## **3. Array Attributes**

check various properties of a NumPy array:

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

```
print(arr.shape) # (2, 3) -> 2 rows, 3 columns
```

```
print(arr.size) # 6 -> total number of elements
```

```
print(arr.dtype) # int64 -> data type of elements
print(arr.ndim) # 2 -> number of dimensions
```

## 4. Indexing and Slicing

NumPy indexing is similar to Python lists.

### Access Elements

```
arr = np.array([10, 20, 30, 40, 50])
print(arr[0]) # 10
print(arr[1]) # 20
print(arr[-1]) # 50
```

### Access Multi-Dimensional Elements

```
matrix = np.array([[1, 2, 3], [4, 5, 6]])
print(matrix[0, 0]) # 1 (first row, first column)
print(matrix[1, 2]) # 6 (second row, third column)
```

### Slicing Arrays

```
arr = np.array([10, 20, 30, 40, 50])
print(arr[1:4]) # [20 30 40] (index 1 to 3)
print(arr[:3]) # [10 20 30] (first 3 elements)
print(arr[::2]) # [10 30 50] (every second element)
```

## 5. Mathematical Operations

NumPy makes mathematical operations simple.

### Basic Math

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

```
print(arr + 2) # [3 4 5 6]
print(arr * 3) # [3 6 9 12]
print(arr / 2) # [0.5 1. 1.5 2. ]
```

## **Element-wise Operations**

```
arr1 = np.array([1, 2, 3])
arr2 = np.array([4, 5, 6])
print(arr1 + arr2) # [5 7 9]
print(arr1 * arr2) # [4 10 18]
```

## **Matrix Multiplication**

```
A = np.array([[1, 2], [3, 4]])
B = np.array([[5, 6], [7, 8]])
print(np.dot(A, B))
```

```
# [[19 22]
#  [43 50]]
```

## **Aggregate Functions**

```
arr = np.array([1, 2, 3, 4, 5])
print(np.sum(arr)) # 15 (sum of all elements)
print(np.mean(arr)) # 3.0 (average)
print(np.max(arr)) # 5 (maximum value)
print(np.min(arr)) # 1 (minimum value)
print(np.std(arr)) # 1.41 (standard deviation)
```

## **Reshaping and Transposing**

### **Reshape an Array**

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