

## Task 2: NumPy

### 1. What is vectorization in NumPy?

Vectorization in NumPy refers to applying operations on entire arrays without using explicit loops. Vectorization is important because it:

- Improves Performance: Eliminates Python-level loops and leverages fast low-level implementations.
- Produces Cleaner Code: Fewer lines, easier to maintain.
- Scales Better: Can efficiently handle large scientific data and machine learning workloads.

Example 1: Add a number to each element

Performs element-wise addition across the entire array without using loops, making the operation fast and efficient.

```
import numpy as np
```

```
a1 = np.array([2, 4, 6, 8, 10])
num = 2
res = a1 + num
print(res)
```

#### Output

```
[ 4  6  8 10 12]
```

### 2. What is the difference between reshape() and resize()?

Both the `numpy.reshape()` and `numpy.resize()` methods are used to change the size of a NumPy array. The difference between them is that the `reshape()` does not change the original array but only returns the changed array, whereas the `resize()` method returns nothing and directly changes the original array.

### 3. What are NumPy dimensions and axes?

In NumPy, dimensions and axes describe the shape and direction of data stored in an array.

Dimensions (`ndim`)

- A dimension refers to the number of levels (or nested layers) of data in an array.

- It tells how many directions are needed to locate an element.
- Also called the rank of an array.
- Given by the attribute `ndim`.

Examples in words:

- 1D array → single line of elements
- 2D array → rows and columns (like a table)
- 3D array → multiple 2D tables stacked together

### Axes

- An axis represents a specific direction along which data is arranged or operated on.
- Axes are numbered starting from 0.
- Each axis corresponds to one dimension.

For a 2D array:

- Axis 0 → vertical direction (rows)
- Axis 1 → horizontal direction (columns)

For higher dimensions:

- Axis numbers increase as dimensions increase.
- Operations like sum, mean, or max can be applied along a specific axis.

## 4. What is slicing in NumPy arrays?

Slicing in NumPy is a way to extract a portion (subset) of an array by specifying a range of indices. It allows you to access:

Specific rows or columns

Subarrays from multidimensional arrays

Continuous blocks of data efficiently

### Important Characteristics

- **Zero-based indexing**

Indexing starts from 0.

- **End index is excluded**

The stop index is not included in the result.

- **Works across dimensions**

Each dimension can be sliced independently.

- **Returns a view, not a copy**

Changes to the sliced array affect the original array (important difference from Python lists).

## **5. How does NumPy help in mathematical computations for AI?**

NumPy is a core library for AI and machine learning because it provides fast, efficient, and expressive numerical computation.

### **1. Efficient Handling of Large Data**

AI models work with large datasets (vectors, matrices, tensors). NumPy stores data in contiguous memory blocks, making access and computation much faster than native Python structures.

### **2. Fast Mathematical Operations**

NumPy performs computations using optimized C and Fortran libraries under the hood. This makes operations like addition, multiplication, dot products, and matrix inversion significantly faster, which is critical for training AI models.

### **3. Linear Algebra Support**

AI heavily relies on linear algebra. NumPy provides built-in support for:

- Matrix multiplication
- Transpose
- Determinants
- Eigenvalues and eigenvectors
- Solving linear equations

These are essential for algorithms like regression, neural networks, and dimensionality reduction.

### **4. Foundation for AI Libraries**

Most AI and ML libraries are built on top of NumPy, including:

- Data preprocessing tools
- Scientific computing libraries
- Deep learning frameworks (internally)

Understanding NumPy makes it much easier to learn and use these advanced AI tools.