

# Numpy :-

⇒ Numpy provides efficient storage.

It provides an Array Data Structure, that is diff. from native Python List, in terms that in a Numpy Array, data of only same data type can be stored.

⇒ Access to data in Numpy Array is faster than Python List.

⇒ In Numpy, we can use as much memory as we need. Thus, it gives a better control over memory management.

⇒ Importing :-

import numpy as np

⇒ Creating array :-

my\_arr = np.array([33, 4, 14])

we here pass any sequence (ex:- list) that we want to get converted to array.

⇒ Setting datatype of array :-

⇒ Pass the datatype explicitly in the 2<sup>nd</sup> param of the np.array() method.

myarr = np.array([44, 3, 2], np.int8)

np.int8: Each element of array will be of integer datatype, occupying 8 bytes each.

[Options available:-] np.int32, np.int64

⇒ Accessing Array Elements :-

⇒ Elements are accessed using indexes in same way as in lists.

[my\_arr[2]] ⇒ 3<sup>rd</sup> element.

⇒ Array elements can be changed dynamically by assigning new value to any index:-

[my\_arr[2] = 45] ⇒ 2<sup>nd</sup> index value will be changed to 45.

## Creating & Accessing 2D Array :-

```
arr_2d = np.array([[1, 4, 3], [2, 4, 1]])
print(arr_2d[1][0])
```

o/p  
2

## Getting Dimension of Array:-

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

o/p  
(2, 4)

means 2 rows, 4 columns

shape is a member variable of the NP array object that contains a tuple, in which from right-to-left the number denotes dimension of array, and size of each level of array.

## > Checking Datatype of Array Elements :-

> dtype member variable of Numpy array object contains datatype of the array elements.

print(my\_arr.dtype) O/p  
int32

## > Getting No. of Elements of an Array:-

my\_arr.size O/p  
5

"size" member variable contains an integer equal to the no. of elements in the array.



When we create an Array from a Python SET or DICTIONARY, then array of 'object' datatype will be formed.

Entire SET/DICTIONARY will be at 0<sup>th</sup> index of the numpy array.

# Intrinsic Array Creation

## Objects:

↳ Creating array of 0's [zero]:-

[  
zero\_arr = np.zeros ((2,6))  
print (zero\_arr)

Output

[ [ 0,0,0,0,0,0 ],  
[ 0,0,0,0,0,0 ] ]

Here, we pass a tuple that contains the dimension of the array that we want.

⇒ zeros method of the Numpy array object is used.

Syntax

[ np.zeros (<shape>, <datatype>) ]

numpy module datatypes are supported.  
Optional parameter :- "float64" is the default datatype.

## 2) Creating Array from a Range of

Integers : `arange()` method :-

Syntax:-

[ `range_arr = np.arange(beg, end, steps)` ]

compulsory      optional  
"end" is excluded.

→ Takes the same parameter as the Python `range()` function, and creates a 1-D array out of the integers satisfying the parameters range.

## 3) `linspace()` method :- Getting

equally spaced floating numbers :-

Syntax:-

[ `arr = np.linspace(beg, end, no_of_elements)` ]

inclusive

Ex-5

```
arr = np.linspace(2, 4, 9)
print(arr)
```

Output

[ 2. , 2.25, 2.5, 2.75, 3.0, 3.25, 3.5, 3.75, 4.0 ]
---

→ Gives an array containing floats equal to the 3<sup>rd</sup> param, that are equally spaced with lower & upper bound as 1<sup>st</sup> and 2<sup>nd</sup> param, respec, both inclusive.

↳ empty() method :- Getting an empty array of fixed dimensions :-

```
em_arr = np.empty((2, 6))
```

Gives an empty array having random elements of the dimensions passed in the parameters.

## 5) Creating Identity Matrix:- `identity()`

[ `id_arr = np.identity(3)` ]

OR

$\begin{bmatrix} [1, 0, 0], \\ [0, 1, 0], \\ [0, 0, 1] \end{bmatrix}$

will give a  $3 \times 3$  identity matrix.

## Reshaping an existing

### Array : `reshape()` method:-

[ `arr = np.arange(10)` ]

[ `re_arr = arr.reshape((2, 5))` ]

Reshaped callee obj array into an array of dimension passed in the parameter.

⇒ Dimension passed in `reshape()` must fit all the elements in the original array exactly (not more or less), else will throw an error.

## ⇒ Converting Multi-D array to

### 1D array : `ravel()`

```
[ arr = np.array([[4,2,6],[1,8,7]])  
re_arr = arr.ravel()  
print(re_arr)
```

Output  
[ 4, 2, 6, 1, 8, 7 ]

32:10

## ⇒ Numpy Axis :-

⇒ Dimension of an array is called Axis.

Counting of Axis starts from 0.

## Sample Array

## AXES (plural)

① 1D  $\rightarrow [1, 2, 3, 4, 5] \rightarrow 1$  [Axis 0]

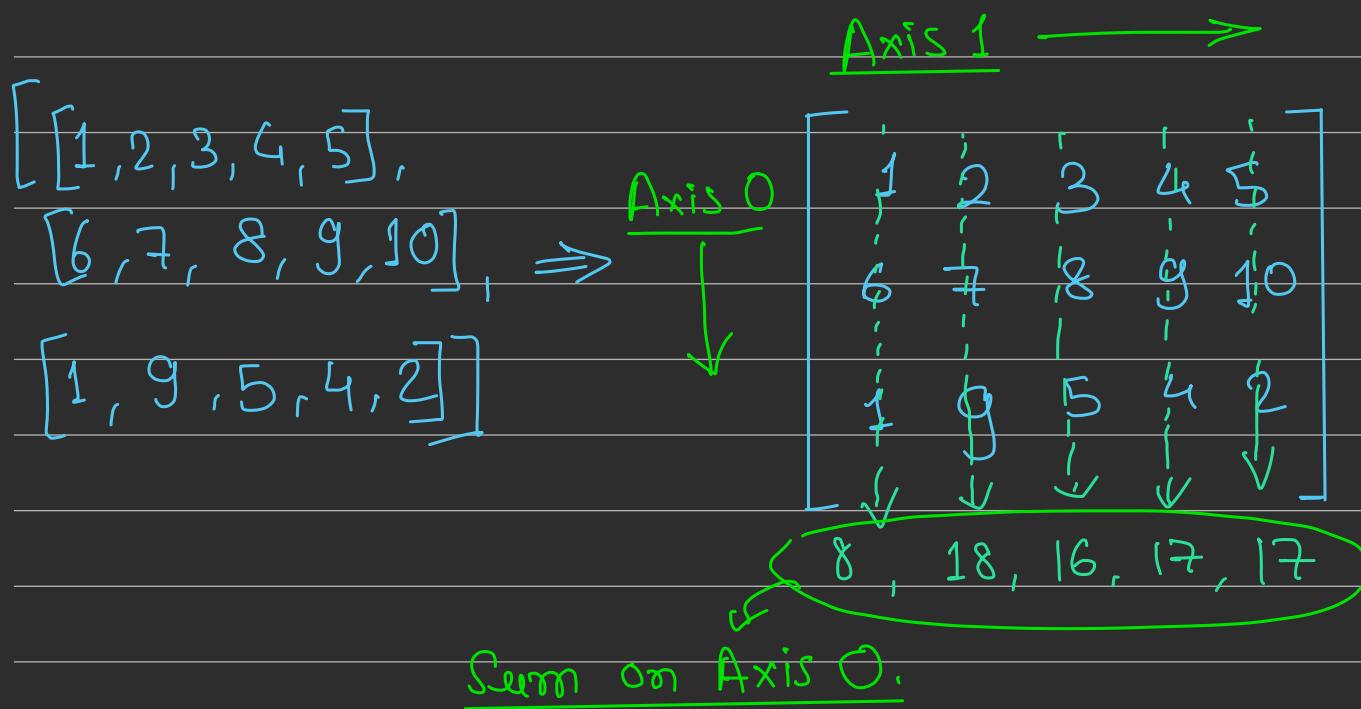
② 2D  $\rightarrow [[1, 2, 3, 4, 5], [6, 7, 8, 9, 10], [1, 9, 5, 4, 2]] \rightarrow 2$  [Axis 0, Axis 1]

⇒ The lowest level of array will be at the highest axis, i.e., at Axis( $n-1$ ) for an  $n$ -Dimensional Array.

↓  
For ex:- For a 2D array.  
Row  $\rightarrow$  Axis 0 (elements of this axis)  
Column  $\rightarrow$  Axis 1 (array contains array element)  
(lowest level, containing data of other datatype.)

Suppose arr is a 2D array :-





This means, elements intersecting Axis 0 will be summed up.

### Operation on Any Axis :-

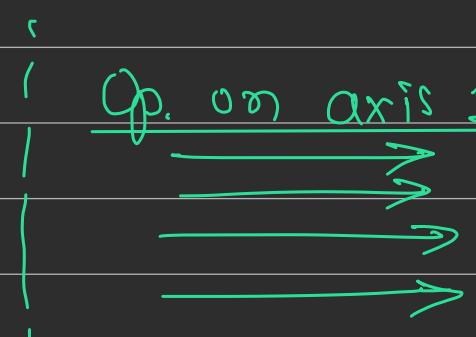
Those elements will be operated which intersects that particular axis. (being given)

### For a 2D Array :-

Operation on Axis 0 :-



Op. on axis 1 :-



## Operation on Axis :-

sum(), max(), min()

Ex:-  $n = [[1, 2, 3], [4, 5, 6], [7, 1, 0]]$

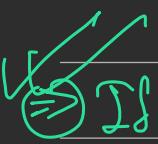
$ar = \text{np.array}(n)$

$$ar.sum(\text{axis} = 0) \rightarrow [12, 8, 9]$$

sum along axis 0

$$ar.sum(\text{axis} = 1) \rightarrow [6, 15, 8]$$

sum along axis 1.



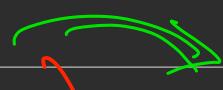
If no parameter is passed, performs the operation with all the elements of the array.

## \* Getting Transpose of Array :-

T attribute / member variable of the array object contains transpose matrix of the numpy array.

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

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

print (arr.T) 

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

\* Looping through all items in any Dimensional Array :-

④ flat attribute of Np array contains an iterator that gives all the elements of an array sequentially (even without handling indexes using variable)

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

for i in arr.flat:  
    print (i)

Output

1
2
3
4
5
6

\* `ndim` attribute - contains no. of

dimensions :-

[arr.ndim]



[OP  
2]

as it is a 2-D array.

\* `nbytes` attribute :-

⇒ Contains the size of array in terms of number of bytes it occupies.

[arr.nbytes]

\* `sqrt()` Numpy function :-

[np.sqrt(arr)]

Returns an array in which each element is a square root of the elements in the array passed in the parameter.

## \* Argmax() & Argmin() methods :-

- ④ Returns the index of that element which is maximum, minimum respectively.

```
[arr = np.array([8, 2, 634, 17, 24, 1, 19])
 point(arr.argmax())
 point(arr.argmin())]
```

The diagram illustrates the application of the argmax and argmin functions to the array [8, 2, 634, 17, 24, 1, 19]. The array elements are indexed from 0 to 6. The max value is 634 at index 2, and the min value is 1 at index 5. The word 'of' is written above the index 2, indicating it is the index of the maximum value.

- ⑤ When applied to 2D or multi-dim array (without any parameter), then it pavel()'s the array, and then gives the index of min/max element.

- ⑥ Applied on multi-dimensional array with axis parameter :-

- ⑦ When applied on multi-dimensional array by applying an axis parameter, then it returns an array of dimension n-1 (n being dimension of original array), with each element giving min/max elements' index along the axis specified.

Ex:  
`arr = np.array([`  
`[1, 2, 3],`  
`[4, 5, 6],`  
`[7, 1, 0],`  
`]) [2, 1, 1]` index of max element  
across axis 0  
`arr.argmax(axis=0)` op  
[2, 1, 1]

Here, vertically the numbers are considered as array, as axis is set to 0, and in that each, index of max element is given in the op array.

## \* `argsort()` method :-

⇒ Returns an array which contains the indexes of original array, in a sequence of ascending sort of the original array.

```
[ arr = mp.array([4, 1, 3, 7])  
print(arr.argsort()) ]
```

$\xrightarrow{\text{OP}}$  [ 1, 2, 0, 3 ]

indexes of elements in sorted order.

⇒ `argsort()` can also be used for multi-dim array by passing axis parameter, and along that axis, we will get index of elements in sorted order.

# MATRIX Operations:

⇒ In Numpy, we can take multiple 2D array, and can perform matrix operations on them.

```
arr1 = np.array([[],[],[]])  
arr2 = "
```

print (arr1 + arr2) → addition

print (arr1 - arr2) → subtraction

print (arr1 \* arr2) → gives element by element multiplication,  $\neq$  matrix multiplication.

# \* Numpy METHODS :-

1) np.sqrt()

2) np.where() any condition.

~~syntax:-~~ [ np.where (arr\_name > 5) ]

Returns a tuple of index, where the elements matching the particular condition occurs.

3) np.count\_nonzero(<arr>)

Returns the no. of non-zero elements in the array.

## => Converting Numpy array to List :-

[ np.arr.tolist() ]

converts numpy array to Python list  
and returns it.