

NUMPY

*import numpy as np

- Numpy stands for Numerical Python
- It is a fundamental python package for scientific computing
- Can be used for arrays, Linear algebra, Random Numbers, Broadcasting

`a= np.array([3,6,9,12])` #print the array.

`a/3` # divides all the elements in the array and prints float values

- Array: it is collection of elements of same data type
- Create arrays from lists or Tuples
- `Array()`

`np.array(object, dtype, copy, order, subok, ndmin)` #internal parameters

`np.array([1,2,3])` #1d array

`np.array([1,2,3.0])` # 1d array with different data types, also called upcasting

`np.array([[1,2],[3,4]])` #2d array, nested list

`np.array([1,2,3], ndimn=2)` # prints 2d array without nested list

`np.array([1,2,3], dtype=complex)` #changes the data type, prints complex values

- `Arange()`: creates an array of evenly spaced values.

`np.arange([start].stop.[step], dtype=None)` #includes start value but excludes stop value

`np.arange(1,10)` # prints array of 1-10, since we didn't gave step value it takes step as 1

`np.arange(3.0)` # we have given the stop value, so it starts from 0 by default and

`np.arange(1,10,2)` #starting value 1, stop value is 10 and step is 2

`np.arange(20,dtype= "complex")` #prints 0-20 complex numbers

- **zeros()**: Creates an array filled with zeros

`np.zeros(shape, dtype=float , order= 'C')`

`np.zeros(5)` # prints an Array of 5 zeros of float data type.

`np.zeros(5,dtype= "int")` # print an array of 5 zeros of integer data type

`np.zeros((2,3))` #we have given a tuple of int's,it prints the matrix of zeros with 2 rows 3 columns

`np.zeros([3,4])` #we have give an list of ints,it prints the matrix of zeros with 3 row, 4 columns

*order doesn't affect the o/p, because it is for storing the multi dimensional data

- **ones()**: Creates an array filled with ones

`np.ones(shape, dtype=None, order= "C")`

`np.ones(8)` #prints an array of 8 ones of float data type

`np.ones(7, dtype='int')` #prints an array of 7 ones of integer data type

`np.ones((3,4))` #we have given tuple of integers, it prints the matrix of 1's with 3 rows and 4 columns

- **empty()**: Creates an array without initializing the entries

`np.empty(6)` #prints an array of 6 arbitrary values of float data type

`np.empty([3,4], dtype= "int")` #prints a matrix of values with 3 rows and 4 columns of int data type.

- **linspace()**: Creates array of filled evenly spaced values

`np.linspace(strat, stop, num=50, endpoint=True, retstep=False, dtype=None, axis=0)`

- `arange()` and `linspace()` will do the same thing but parameters are different and `linspace` returns 'num' evenly space samples and the end point is optionally excluded.

`np.linspace(2.0, 3.0, num=5)` #prints 5 values in between 2 to 3 which are evenly spaced and prints along with ending point as a float data type

`np.linspace(2.0, 3.0, num=5, endpoint=False)` #prints 5 values in between 2 to 3 which are evenly spaced and prints along without ending point as float data type

`np.linspace(2, 3, num=5, retstep=True)` #prints 5 values in between 2 to 3 which are evenly spaced and prints along with ending point and prints step value as well as a float data type

- **eye()**: Returns array filled with zeros except in the k^{th} diagonal, whose values are equal to 1.

`np.eye(N, M=None, k=0, dtype=<class 'float'>, order = 'C')` # n= no.of rows, m= no.of columns

`np.eye(5)` #prints 1 in the main diagonal and remaining as zeros with float data type

`np.eye(2,3)` # 2 rows , 3 columns

`np.eye(4, k=-1)` #main diagonal will be zero and 1's will be printed in below the main diagonal

- **identity()**: Returns the identity array

`np.identity(shape, dtype = 'float')`

`np.identity(4)` #prints the square identity matrix of float numbers

- **random()**: Random number generation

- `np.rand()`: uniformly distributed value.
- `np.randn()`: Normally distributed value.
- `np.randf()`: Uniformly distributed floating point numbers.
- `np.randint()`: uniformly distributed integers in a given range.

1) rand()

`np.random.rand(d0,d1,d2,...)` # prints the random values in a given shape.

`np.random.rand(5)` # prints an array of 5 random values.

`np.random.rand(4,5)` # print the matrix with 4 rows and 5 columns of random values.

2) randn()

`np.random.randn(d0,d1,d2,...)` # returns a sample from the standard normal distribution.

`np.random.randn()` # prints single random value

`np.random.randn(5)` # prints 5 random values from the Standard normal distribution

3) randf()

`np.random.randf(size)` # returns random floats in half open interval

`np.random.randf(5)` # prints 5 random float values

`5*np.random.random_sample(3,3)`

4)randint()

`np.random.randint(low, high=None, size=None, dtype= 'i')` # returns a random integer from low to high in half open interval.

`np.random.randint(1, size=10)` #2 is assigned to low, generates 10 random values, we get 0,1 because high value is not mentioned

`np.random.randint(5, size=(2,4))` # prints random values 2d array with 2 rows and 4 columns

- Attributes of arrays
 - Dimension: N-dimensional array, if you want to know the dimension of an array
We can use 'ndim' attribute

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

`a.ndim` #prints the dimension of array.

- Shape: tuple of elements indicating the number of elements that are stored along each dimension of the array.

`a=np.zeros(5)`

`A.shape` #prints 5 number of elements in array

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

`b.shape` # prints the number of rows and number of columns (2,2)

- Size: This will tell the total number of elements in the array.

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

`b.size` #prints the number of elements in the 2 d array (4)

- Dtype: Describes the data types
- Itemsize: the size of each element in bytes.

INDEXING

- We can access the element of array through index, we can access single value of array through index
- Array follows Zero based indexing i.e, first element index is zero. We can also use negative indexing for arrays

`a=np.array([1,2,3,4,5])` #1D array with list

`a[3]` # prints 4

`a[-2]` #prints 4 again

`b=np.array([[1,2],[3,4]])` #2D array with list

`b[0][0]` # zero row , zero column , prints 1

`b[1]` #prints row ([3,4])

`c=np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]], [[13,14,15,16], [17,18,19,20], [21,22,23,24]])`

`c[0][0][1]` #zero i, zero j, 1 k, prints 2

`c[-2][-3][-3]` # prints -2

SLICING

- To retrieve a collection of arrays, we use slicing.

`a[start : stop: step]`

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

`a[1:4]` # prints ([2,3,4]) , step=1 by default

`a[1::2]` #prints ([2,4]), it prints every second element from starting point

`a[:]` #prints entire array

`b[start : end: step, start:end:step]`

```

b=np.array([[1,2],[2,3],[3,4]])
b[1:,1:] #prints ([[3],[4]])
c[start : end: step, start : end : step, start : end : step] #number of matrices, rows, columns
c=np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]], [[13,14,15,16], [17,18,19,20], [21,22,23,24]])
c[:, :, 1] #prints ([[1],[5],[9]],[[13],[17],[21]])

```

ARITHMETIC OPERATION

- We can do addition, subtraction, multiplication, division on arrays
- Arithmetic operations on arrays are applied element by element in arrays

```

a=np.array([1,2,3,4])
a+2 #add 2 to the each element of array ([3,4,5,6])
    • Similarly subtraction, multiplication, division
b=np.array([6,7,8,9])
a+b #prints sum of two array ([7,9,11,13])
    • Similarly subtraction, multiplication, division
    • Similarly for 2-D arrays of same shape as well
1) a+b = np.add(a,b)
2) a-b = np.subtract(a,b)
3) a*b = np.multiply(a,b)
4) a/b = np.divide(a,b)
5) a%b = np.mod(a,b)
6) a**b = np.power(a,b)

```

BROADCASTING

- Can we perform arithmetic operations on arrays of different shape and size.
- Broad Casting allows us to perform arithmetic operations on arrays of different size or shape.
- Broadcasting will stretch the value/array to the required shape, then performs arithmetic operation
- Broadcasting Rules
 - The size of each dimension should be same
 - The size of one of the dimension should be one
 - 1) If the two arrays differ in their number of dimensions, the shape of the one with fewer dimension is padded with ones on its leading side(left side)
 - 2) If the shape of the two arrays does not match in any dimensions, the array with shape equal to 1 in that dimension is stretched to match the other shape.
 - 3) If in any dimension the size disagrees and neither equal to 1, an error is raised.

```

a=np.array([10,20,30])
b=np.array([1,2,3,4])
a+b # raises error due to broadcasting error

```

```

a=np.array([[1,2],[3,4],[5,6]])
b=np.array([10,20])
a.shape #prints(3,2)

```

```
b.shape #prints(2,)  
a+b #prints suum of the arrays
```

Because,

- 1) **Rule 1:** According to Rule 1, if there are two different dimensions, we need to add 1 for the b array.
a=(3,2) b=(1,2)
- 2) **Rule 2:** Right values of both the arrays were matched(2=2), (3!=1)
- 3) **Rule 3:** According to rule 3, any one of the dimension should be 1, in b array it is (1,2) it satisfied
- 4) Hence, we will get output in higher dimensions.

ARRAY MANIPULATION

- **reshape():** Gives the new shape to the array without changing the array data .
np.reshape(array, shape, order='C')
a=np.arange(10) #prints array of 0-10
b=np.reshape(a,(5,2)) # changes 1D array to 2D array without changing data (rowwise)
b=np.reshape(a,(5,2), order='F') # changes 1D array to 2D array without changing data (column wise)
 - Size of the array is nothing but product of the shape
 - When we change the shape the size should remain the same, that shouldn't change.
- **resize():** This array will change the data of array
 - If the size of new array is larger than the original array, then it makes the repeated copies of datanp.resize(array, shape)
a = np.arange(5) #creates a 1d array from 0 to 5
np.resize(a,(2,3)) #changes the shape and size of the array without

FLATTEN

- **flatten():** Returns the copy of array which collapsed into 1 Dimension
- If we give 2d or 3d array it makes a copy of array collapsed into 1d
array.flatten(order= 'C')
a=np.array([[[1,2,3],[3,4,5]],[[6,7,8],[9,10,11]]])
a.flatten()
 - This gives the copy of array in 1D, it converts 3D to 1D in row wisea.flatten(order= "F")
 - This gives the copy of array in 1D, columns wise
 - **ravel():** Also used to flatten the array, but copy is made only if needed
np.ravel(array, order= 'C')
a.ravel() # This also flatten the given array.

TRANSPOSE AND SWAPAXES

- **transpose():** Rearrange the dimension of the array

```
np.transpose(array, axes=None)
```

```
a = np.arange(1,11).reshape(5,2) #prints the matrix with 5 rows and 2 columns
```

```
np.transpose(a) # transposes the matrix into 2 row and 5 columns
```

- It just reverses the shape of the array or matrix
- If we apply transpose on 1D array we get the same array

```
a.T # this also transposes the matrix
```

- **swapaxes():** This is used to interchange the any two axes of a given array.

```
np.swapaxes(array, axis1,axis2)
```

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

```
np.swapaxes(a,0,1) #swaps 1st column into 1st row, 2nd column into 2nd row
```

- This will be useful when we are dealing with 3D or 4D array, when we want to change the two columns