

NumPy

Introduction

- Stands for numerical python.
- Fundamental Package for numerical Computations.
- Supports N dimensional array objects that can be used for processing multi-dimensional data.
- Supports different data types.

Introduction

- Using Numpy we can perform
 - Mathematical and logical operations on arrays.
 - Fourier Transform
 - Linear Algebra Operations
 - Random Number generations

Crate an array

- Array is the ordered collection of basic data types of given length
- Syntax: `numpy.array(object)`

```
In [1]: import numpy as np
```

```
In [2]: x = np.array([2,3,4,8])
```

```
In [3]: type(x)
```

```
Out[3]: numpy.ndarray
```

```
In [4]: x
```

```
Out[4]: array([2, 3, 4, 8])
```

Array

- Numpy can handle different categorical entries

```
In [5]: y = np.array([2, 'LPU', 4, 8])
```

```
In [6]: y
```

```
Out[6]: array(['2', 'LPU', '4', '8'], dtype='<U11')
```

Generate Array using linspace

- Return equally spaced numbers within the given range.
- `numpy.linspace(start, stop, num, dtype, retstep, dtype)`
 - Start : start of interval range
 - Stop: The end value of the interval range
 - Num: number of samples to be generated
 - dtype: type of output array
 - restep: return the samples, step value

Generate Array using linspace

- Generate an array b with start value 1 and stop value 5.

```
In [9]: b = np.linspace(1, 5, num=10, endpoint = True, retstep=False)
```

```
In [10]: b
```

```
Out[10]: array([1.          , 1.44444444, 1.88888889, 2.33333333, 2.77777778,  
               3.22222222, 3.66666667, 4.11111111, 4.55555556, 5.          ])
```

```
In [17]: b = np.linspace(1, 5, num=10, endpoint = True, retstep=True)
```

```
In [18]: b
```

```
Out[18]: (array([1.          , 1.44444444, 1.88888889, 2.33333333, 2.77777778,  
               3.22222222, 3.66666667, 4.11111111, 4.55555556, 5.          ]),  
         0.4444444444444444)
```

```
In [19]: b = np.linspace(1, 5, num=10, endpoint = False, retstep=True)
```

```
In [20]: b
```

```
Out[20]: (array([1. , 1.4, 1.8, 2.2, 2.6, 3. , 3.4, 3.8, 4.2, 4.6]), 0.4)
```

Generate array using np.arange

- np.arange returns equally spaced numbers within the given range based on step size.
- Syntax: np.arange(start, stop, stepsize)
 - Start : start of interval range
 - Stop: The end value of the interval range
 - Stepsize: stepsize in the interval

Generate array using np.arange

- Generate an array with start value = 1, stop value = 10 and stepsize = 2

```
In [23]: d = np.arange(1,10,2)
```

```
In [24]: d
```

```
Out[24]: array([1, 3, 5, 7, 9])
```

Generate array using ones

- Numpy.ones returns an array of given shape and type filled with ones.
- Syntax: `numpy.ones(shape, dtype)`
- Shape: integer or sequence of integers
- Dtype: data type (default: float)

```
In [27]: e = np.ones((3,4))
```

```
In [28]: e
```

```
Out[28]: array([[1., 1., 1., 1.],  
               [1., 1., 1., 1.],  
               [1., 1., 1., 1.]])
```

Generate array using zeros

- Numpy.zeros returns an array of given shape and type filled with zeros.
- Syntax: `numpy.zeros(shape, dtype)`
- Shape: integer or sequence of integers
- Dtype: data type (default: float)

```
f = np.zeros((3,4))
```

```
f
```

```
array([[0., 0., 0., 0.],  
       [0., 0., 0., 0.],  
       [0., 0., 0., 0.]])
```

Generate array using random.rand()

- Numpy.random.rand() returns an array of given shape filled with random values.
- Syntax: numpy.random.rand(shape)

```
g = np.random.rand(5)
```

```
g
```

```
array([0.96591181, 0.17853905, 0.82199814, 0.08965516, 0.32435735])
```

```
In [43]: g = np.random.rand(5,4)
```

```
In [44]: g
```

```
Out[44]: array([[0.18836923, 0.25885422, 0.3509376 , 0.46863139],  
                [0.21263104, 0.18189903, 0.1027898 , 0.2155268 ],  
                [0.68860479, 0.49224318, 0.00386144, 0.41830482],  
                [0.73607213, 0.06082685, 0.86523608, 0.13471038],  
                [0.77180107, 0.43262638, 0.68347156, 0.1230697 ]])
```

Advantages of numpy

- Numpy supports vectorized operations.
- Array operations are carried out in c and hence universal functions in numpy are faster than operations carried out in Python Lists.

```
In [68]: x = range(1000)
```

```
In [69]: timeit(sum(x))
```

37.6 μ s \pm 4.86 μ s per loop (mean \pm std. dev. of 7 runs, 10000 loops each)

```
In [70]: y = np.array(x)
```

```
In [ ]: y
```

```
In [72]: timeit(np.sum(y))
```

9.93 μ s \pm 580 ns per loop (mean \pm std. dev. of 7 runs, 100000 loops each)

Advantages of numpy: Consumes less storage space

```
In [73]: import sys
```

```
In [87]: sys.getsizeof(1) * len(x)
```

```
Out[87]: 28000
```

```
In [90]: y.itemsize * y.size
```

```
Out[90]: 4000
```

Reshape an array

- Recast an array to new shape without changing its data

```
In [41]: g = np.arange(1,10).reshape(3,3)
```

```
In [42]: g
```

```
Out[42]: array([[1, 2, 3],  
               [4, 5, 6],  
               [7, 8, 9]])
```

Array Dimensions

- Shape Returns the dimensions of an array

```
In [43]: h = np.array([[1,4,8],[2,6,9],[3,5,7]])
```

```
In [44]: h
```

```
Out[44]: array([[1, 4, 8],  
               [2, 6, 9],  
               [3, 5, 7]])
```

```
In [46]: h.shape
```

```
Out[46]: (3, 3)
```


Numpy addition

- `Numpy.add()` returns the element wise addition between two arrays.

```
In [47]: g
```

```
Out[47]: array([[1, 2, 3],  
               [4, 5, 6],  
               [7, 8, 9]])
```

```
In [48]: h
```

```
Out[48]: array([[1, 4, 8],  
               [2, 6, 9],  
               [3, 5, 7]])
```

```
In [50]: np.add(g,h)
```

```
Out[50]: array([[ 2,  6, 11],  
               [ 6, 11, 15],  
               [10, 13, 16]])
```

Numpy Multiplication

- `Numpy.multiply()`: returns element wise multiplication between two arrays.

```
Out[47]: array([[1, 2, 3],  
               [4, 5, 6],  
               [7, 8, 9]])
```

```
In [48]: h
```

```
Out[48]: array([[1, 4, 8],  
               [2, 6, 9],  
               [3, 5, 7]])
```

```
In [51]: np.multiply(g,h)
```

```
Out[51]: array([[ 1,  8, 24],  
               [ 8, 30, 54],  
               [21, 40, 63]])
```

Other numpy functions

Function name	Description
<code>numpy.subtract</code>	performs element wise subtraction between two arrays
<code>numpy.divide</code>	returns an element wise division of inputs
<code>numpy.remainder</code>	Return element-wise remainder of division

Accessing Components of an array

- Components of an array can be accessed using index number.

	0	1	2
0	1	4	8
1	2	6	9
2	3	5	7

- Extract element 2 in array a : a(0,1)

```
In [55]: h
Out[55]: array([[1, 4, 8],
                [2, 6, 9],
                [3, 5, 7]])
```

```
In [57]: h[0,1]
```

```
Out[57]: 4
```

```
In [60]: h[1:3]
```

```
Out[60]: array([[2, 6, 9],
                [3, 5, 7]])
```

```
In [61]: h[:,0]
```

```
Out[61]: array([1, 2, 3])
```

```
In [62]: h
```

```
Out[62]: array([[1, 4, 8],
                [2, 6, 9],
                [3, 5, 7]])
```

```
In [64]: h[0,:]
```

```
Out[64]: array([1, 4, 8])
```

Modifying an array using transpose

- `Numpy.transpose(array)`: returns the transpose of an array.

```
In [73]: h
```

```
Out[73]: array([[ 1, 10,  8],  
               [ 2, 20,  9],  
               [ 3,  5,  7]])
```

```
In [74]: np.transpose(h)
```

```
Out[74]: array([[ 1,  2,  3],  
               [10, 20,  5],  
               [ 8,  9,  7]])
```

Modify array using append

- Add the values at the end of array.
- Syntax: `numpy.append(array,axis)`

```
In [99]: h = np.append(h, [[12,34,65]], axis = 0)
```

```
In [106]: h
```

```
Out[106]: array([[ 1, 10,  8],  
                 [ 2, 20,  9],  
                 [ 3,  5,  7],  
                 [12, 34, 65],  
                 [12, 34, 65]])
```

```
In [107]: col = np.array([[23,45,67,98,24]]).reshape(5,1)
```

```
In [108]: 1 h = np.append(h,col, axis = 1)
```

```
In [109]: h
```

```
Out[109]: array([[ 1, 10,  8, 23],  
                 [ 2, 20,  9, 45],  
                 [ 3,  5,  7, 67],  
                 [12, 34, 65, 98],  
                 [12, 34, 65, 24]])
```

```
Out[109]: array([[ 1, 10,  8, 23],
                 [ 2, 20,  9, 45],
                 [ 3,  5,  7, 67],
                 [12, 34, 65, 98],
                 [12, 34, 65, 24]])
```

```
In [111]: h = np.insert(h, 1, [[10,20,30, 40]], axis = 0)
```

```
In [112]: h
```

```
Out[112]: array([[ 1, 10,  8, 23],
                 [10, 20, 30, 40],
                 [ 2, 20,  9, 45],
                 [ 3,  5,  7, 67],
                 [12, 34, 65, 98],
                 [12, 34, 65, 24]])
```

```
In [114]: h = np.delete(h, 2, axis = 0)
```

```
In [115]: h
```

```
Out[115]: array([[ 1, 10,  8, 23],
                 [10, 20, 30, 40],
                 [ 3,  5,  7, 67],
                 [12, 34, 65, 98],
                 [12, 34, 65, 24]])
```
