```
In [2]: import numpy as np
 In [3]: #creating arrays using numpy
         arr = np.array([1, 2, 3, 4, 5, 6])
         print("Arr: ", arr)
         two_arr = np.array([[1, 2, 3, 4], [5, 6, 7, 8], [9, 10, 11, 12]])
         print("Two dimentional Arr: \n", two_arr)
        Arr: [1 2 3 4 5 6]
        Two dimentional Arr:
         [[1 2 3 4]
         [5 6 7 8]
         [ 9 10 11 12]]
 In [4]: print(two_arr[1])
        [5 6 7 8]
         np.zeros() creates array containing zeros of size given
 In [5]: arr0 = np.zeros(5)
         print(arr0)
        [0. 0. 0. 0. 0.]
         np.ones() creates array containing ones of size given
 In [6]: arr1 = np.ones(10)
         print(arr1)
        [1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
In [38]: mat = np.ones((5, 6))
         print(mat)
        [[1. 1. 1. 1. 1. 1.]
         [1. 1. 1. 1. 1. 1.]
         [1. 1. 1. 1. 1. 1.]
         [1. 1. 1. 1. 1. 1.]
         [1. 1. 1. 1. 1. 1.]]
In [55]: np.ones((4,4), dtype = bool)
Out[55]: array([[ True, True, True, True],
                 [ True, True, True],
                 [ True, True, True],
                 [ True, True, True, True]])
 In [7]: arr2 = np.arange(5)
         print(arr2)
         arr3 = np.arange(2, 11, 2)
         print(arr3)
        [0 1 2 3 4]
        [ 2 4 6 8 10]
```

np.sort() will return sorted array

```
In [8]: arr4 = np.array([4, 3, 6, 7, 2, 8, 1, 0, 9, 5])
          arr4 = np.sort(arr4)
          print(arr4)
        [0 1 2 3 4 5 6 7 8 9]
          concatenate function is used to attach one are to another
 In [9]: a = np.array([1, 2, 3, 4])
          b = np.array([5, 6, 7, 8])
          np.concatenate((a, b))
 Out[9]: array([1, 2, 3, 4, 5, 6, 7, 8])
In [10]: type(arr)
          type(two_arr)
Out[10]: numpy.ndarray
          ndarray.size will return the total number of elements of the array
In [11]:
          print(arr.size)
          print(two_arr.size)
        12
          Shape function will return dimentions of array i.e. (rows,cols)
In [12]:
          print(arr.shape)
          print(arr1.shape)
          print(two_arr.shape)
        (6,)
        (10,)
        (3, 4)
          reshape((rows, cols)) to reshape your array to x * y
In [39]: arr5 = np.arange(10)
          print(arr5)
        [0 1 2 3 4 5 6 7 8 9]
In [40]: b = arr5.reshape(5, 2)
          print(b)
        [[0 1]
         [2 3]
         [4 5]
         [6 7]
         [8 9]]
In [41]: c = arr5.reshape(2, 5)
          print(c)
```

```
[[0 1 2 3 4]
[5 6 7 8 9]]
```

dtype returns data type of the elements in array

```
In [42]: print(arr.dtype)
          print(c.dtype)
         int32
         int32
          transpose() will return transpose of the array
In [43]:
         c.transpose()
Out[43]: array([[0, 5],
                  [1, 6],
                  [2, 7],
                  [3, 8],
                  [4, 9]])
          np.empty() will generate a random array of size row * col containing of dtype values
          syntax : np.empty((row,col), dtype = data_type)
In [47]: np.empty((3,4), dtype = int)
Out[47]: array([[ 1670598114, -2147483303, -2125552816, -998675401],
                  [ 1752669196,
                                          345, 1752669220,
                                                                       345],
                  Γ
                             11,
                                             0,
                                                           7,
                                                                         0]])
          Reshaping and flattening multidimensional arrays:
In [59]: arr7 = np.empty((4,4), dtype = int)
          print(arr7)
         [[0 0 0 0]]
          [0 0 0 0]
          [0 0 0 0]
          [0 0 0 0]]
          There are two popular ways to flatten an array: .flatten() and .ravel(). The primary
          difference between the two is that flatten() does not make changes in same array instead
          it creates new copy of the array while ravel() does operations on original array itself.
In [60]: arr7.flatten()
```

```
In [61]: arr7.reshape((4,4))
Out[61]: array([[0, 0, 0, 0],
           [0, 0, 0, 0],
           [0, 0, 0, 0],
           [0, 0, 0, 0]])
In [62]: | arr7.ravel()
```

```
Indexing and slicing
In [85]: data = np.arange(1,26)
         data = data.reshape((5,5))
         print(data)
        [[ 1 2 3 4 5]
         [678910]
        [11 12 13 14 15]
         [16 17 18 19 20]
         [21 22 23 24 25]]
         arr[row] will return row at intex [row]
In [86]: data[0]
Out[86]: array([1, 2, 3, 4, 5])
         arr[i][j] will return element at location row [i] and col [j]
In [87]: data[3][4]
Out[87]: 20
         arr[r1:r2,c1:c2] to slice array at locations row1 to row2 and col1 to col2
In [95]: data[0:4,2:5]
Out[95]: array([[ 3, 4, 5],
                [8, 9, 10],
                [13, 14, 15],
                [18, 19, 20]])
         Useful array operations
           1. max() gives maximum in array
           2. min() gives minimum in array
           3. sum() gives addition of all array elements
           4. mean() returns mean of array elements
In [96]: data.max()
Out[96]: 25
In [97]: data.min()
Out[97]: 1
In [98]: data.sum()
Out[98]: 325
In [99]: data.mean()
```

```
Out[99]: 13.0

In [100... data.sum(axis = 1)

Out[100... array([ 15, 40, 65, 90, 115])
```