Modules

 A module comprises of various functions which can be created and imported whenever required.

Creating Module

We can create modules in Jupyter Notebook with steps below:

- 1. Create a new Jupyter notebook
- 2. define required functions in it
- 3. Go to File > Download as > Python(.py)
- 4. Save it in the same working directory

Working with Module

We can import module by using below options.

- 1. import module name (We can call function as "module name.function())
- 2. import module_name as md (We can call function as md.function())
- 3. from module_name import function (We can call the function as function())

Downloading a Module

Modules can be downloaded with pip option. We can check the source on google and use pip as below:

· !pip install pandas

Note- To see all functions in module, write "module name." and press tab key

```
In [1]: import calculator
   import calculator as cal
   from calculator import div
```

```
In [2]: calculator.add(5,2)
     cal.add(5,2)
     div(5,2)
```

Out[2]: 2.5

pwd (print working directory) command returns the current directory

```
In [3]: pwd
```

Out[3]: 'C:\\Users\\srbhk\\Downloads\\Python'

Modules to learn

- 1. Numpy (Numerical Python) Mathematical Operations
- 2. Pandas Data Manipulation
- 3. Matplotlib Data Visualization
- 4. Scikitlearn Machine Learning algorithms
- 5. Scipy (Scientic Operations) Advanced Mathematical/Scientific Operations
- 6. Seaborn Advanced Visualization

Numpy

- · It is an open source module
- · It is programmed in C so it is fast
- Adds support for large, multi-dimensional arrays and matrices, along with a large collection
 of high-level mathematical functions to operate on these arrays like One-Dimensional
 Numpy, Two-Dimensional Numpy
- Helpful in linear algebra, fourier transformations, integrating databases/C/C++

```
In [4]: import numpy as np
In [5]: list1 = [1,2,3] #each element in lists take 14 bytes of memory
arr1 = np.array([1,2,3]) #each element in array takes 4 byte of memory so faste
```

- numpy.full(shape, fill_value, dtype = None, order = 'C') Return a new array with the same shape and type as a given array filled with a fill_value, dtype is float(by Default)
- np.zeros(shape) or np.ones(shape) Returns arrays of 0s/1s with given shape. We can also give dtype as additional argument

- np.random.function(args) returns array of random values as per the function and arguments
- random(size = num) or rand(num) returns array of random float values. returns 1 value if no argument given
- randint(highest, size = num) returns random int values between 0 and highest of given size.

• **choice(array, size = num)** method takes an array as a parameter and randomly returns one of the values if no size given. If size is given, it returns array of random elements from the input array

- np.linspace(start, stop, num = 50, endpoint = True, retstep = False, dtype = None) Start is 0 by default, restep : If True, return (samples, step/gap). By default restep = False
- np.arange(start, stop, step, dtype): Returns an array with evenly spaced elements as per the interval. Stop is not included. step: step size of interval. By default step size = 1
- · arange can be combined with reshape
- np.arange(start,stop,step,dtype).reshape(size) gives array as per the reshape size given

Note - np.linspace allows you to define how many values you get including the specified min and max value. It infers the stepsize **whereas** np.arange allows you to define the stepsize and infers the number of steps(the number of values you get).

- np.concatenate(array1, array2) flatens the arrays and merges them along required axis. The axis along which the arrays will be joined. If axis is None, arrays are flattened before use. Default is 0.
- np.vstack(tup) is used to stack the sequence of input arrays vertically to make a single array.
- **np.hstack(tup)** is used to stack the sequence of input arrays horizontally (i.e. column wise) to make a single array.
- np.column stack(tup)

```
In [10]: a = np.array([[1,2,3],[4,5,6]])
         b = np.array([[7,8,9],[1,5,8]])
         print("Concat:\n",np.concatenate((a,b)))
         \#print("Concat1:\n",np.concatenate((a,b), axis =0)) is same as above statement
         print("Concat2:\n",np.concatenate((a,b), axis =1))
         print("vstack:\n",np.vstack((a,b)))
         print("hstack:\n",np.hstack((a,b)))
         print("column_stack:\n",np.column_stack((a,b)))
         Concat:
          [[1 2 3]
          [4 5 6]
          [7 8 9]
          [1 5 8]]
         Concat2:
          [[1 2 3 7 8 9]
          [4 5 6 1 5 8]]
         vstack:
          [[1 2 3]
          [4 5 6]
          [7 8 9]
          [1 5 8]]
         hstack:
          [[1 2 3 7 8 9]
          [4 5 6 1 5 8]]
         column stack:
          [[1 2 3 7 8 9]
          [4 5 6 1 5 8]]
```

- array.shape returns a tuple with each index having the number of corresponding elements.
- array.ndim returns the number of dimensions of array
- array.size returns the total number of elements in array.
- numpy.size(arr, axis) gives the size of array across particluar axis. 0=x axis, 1 = y axis.
- array.dtype returns datatype of the array
- array.itemsize returns the length of one array element in bytes.

```
In [11]: a = np.array([[1,2,3],[4,5,6]])
    print("Shape: ", a.shape)
    print("dimension: ", a.ndim)
    print("Size: ", a.size)
    print("datatype :", a.dtype)
    print("a.itemsize: ", a.itemsize)

Shape: (2, 3)
    dimension: 2
    Size: 6
    datatype : int32
    a.itemsize: 4
```

Note: We can change the shape of array keeping in mind that elements are equally accommodated.

- np.equal(arr1, arr2) used to compare and return bool values elementwise
- np.sum(arr, axis, dtype, out) axis =0 means along the column, 1 means sum along the row. Out if we want array to be assigned to other array
- np.subtract(arr1, arr2) gives arr1-arr2 elementwise
- Similarly we have divide, multiply, exponetial (e^n), sqrt etc.

```
b = np.array([[1,1,1],[2,2,2]])
         print("Equal:\n", np.equal(a,b))
         print(np.sum([[1,2,3],[4,5,6]]))
         print(np.sum([[1,2,3],[4,5,6]], axis = 0))
         print(np.sum([[1,2,3],[4,5,6]], axis = 1))
         print("Subtract:\n", np.subtract(a,b))
         print("Divide:\n", np.divide(a,b))
         print("Multiply:\n",np.multiply(a,b))
         print("exponential:\n", np.exp(a))
         print("root:\n", np.sqrt(a))
         Equal:
          [[ True False False]
          [False False False]]
         21
         [5 7 9]
         [ 6 15]
         Subtract:
          [[0 1 2]
          [2 3 4]]
         Divide:
          [[1. 2. 3.]
          [2. 2.5 3. ]]
         Multiply:
          [[ 1 2 3]
          [ 8 10 12]]
         exponential:
          [[ 2.71828183
                            7.3890561
                                        20.08553692]
          [ 54.59815003 148.4131591 403.42879349]]
         root:
          [[1.
                       1.41421356 1.73205081]
          [2.
                       2.23606798 2.44948974]]
         We also have functions to find min, max, mean, median
In [14]: a = np.array([[1,2,3],[4,5,6]])
         print(np.min(a))
         print(np.max(a))
         print(np.mean(a))
         print(np.median(a))
         print(np.min(a))
         print(np.std(a)) #Standard Deviation
         print(np.corrcoef(a)) #Correlation Coifficient
         1
         6
         3.5
         3.5
         1
         1.707825127659933
         [[1. 1.]
          [1. 1.]]
```

In [13]: a = np.array([[1,2,3],[4,5,6]])

Indexing and Slicing in Numpy

```
In [15]: a = np.array([1,2,3,4])
         b = np.array([[5,6,7],[8,9,10]])
         print(a[0])
         print(a[-1]) # or print(a[3])
         print(a[0:2]) #prints 1 less than index
         print(a[-2:-1])
         1
         4
         [1 2]
         [3]
In [16]: print("\n2D arrays:\n", b[0])
         print(b[0,2]) # or b[0][2] gives same result)
         print(b[-1,-1])
         print(b[0:2,1:3])
         2D arrays:
          [5 6 7]
         7
         10
         [[ 6 7]
          [ 9 10]]
```

Create the following NumPy arrays:

- a) A 1-D array called arr_zeros having 10 elements and all the elements are set to zero.
- b) A 1-D array called arr_vowels having the elements 'a', 'e', 'i', 'o' and 'u'.
- c) A 2-D array called arr_ones having 2 rows and 5 columns and all the elements are set to 1 and dtype as int.
- d) Use nested Python lists to create a 2-D array called arr_myarray1 having 3 rows and 3 columns and store the following data: 2.7, -2, -19 0, 3.4, 99.9 10.6, 0, 13
- e) A 2-D array called arr_myarray2 using arange() having 3 rows and 5 columns with start value = 4, step size 4 and dtype as float.

```
In [17]: #a
         arr_zeros = np.full(10, 0) #or np.zeros(10)
         #print(arr_zeros)
         #b
         arr_vowels = np.array(["a","e","i","o","u"])
         #print(arr_vowels)
         arr_ones = np.full((2, 5), 1, dtype=np.int32) # or <math>np.ones((2,5), dtype = np.int.
         #print(arr_ones)
         #d
         list1 = [[2.7, -2, -19], [0, 3.4, 99.9], [10.6, 0, 13]]
         arr myarray1 = np.array(list1)
         #print(arr myarray1)
         #e
         temp_myarray2 = np.arange(4,64,4,float)
         arr_myarray2 = temp_myarray2.reshape(3,5)
         print(arr_myarray2)
         [[ 4. 8. 12. 16. 20.]
          [24. 28. 32. 36. 40.]
          [44. 48. 52. 56. 60.]]
```

Using the arrays created in question 1 above, write NumPy commands for the following:

- a) Find the dimensions, shape, size, data type of the items and itemsize of arrays arr_zeros, arr_vowels, arr_ones, arr_myarray1 and arr_myarray2.
- b) Reshape the array arr_ones to have all the 10 elements in a single row.
- c) Display the 2nd and 3rd element of the array arr_vowels.
- d) Display all elements in the 2nd and 3rd row of the array arr_myarray1.
- e) Display the elements in the 1st and 2nd column of the array arr_myarray1.
- f) Display the elements in the 1st column of the 2nd and 3rd row of the array arr myarray1.
- g) Reverse the array of arr_vowels.

```
In [18]: #a
         np.ndim(arr_zeros)
         np.shape(arr_zeros)
         np.size(arr zeros)
         arr_zeros.dtype
         arr_zeros.itemsize
         #b
         arr_ones.reshape(10,)
         #c
         arr_vowels[1:3]
         #d
         arr_myarray1[1:3,]
         #e
         arr_myarray1[:,0:2]
         #f
         arr myarray1[1:3,0]
         #a
         arr_vowels[: : -1] # Check or arr_vowels[range(4,-1,-1)]
```

```
Out[18]: array(['u', 'o', 'i', 'e', 'a'], dtype='<U1')</pre>
```

Using the arrays created in question1 above, write NumPy commands for the following:

- a) Divide all elements of array arr ones by 3.
- b) Add the arrays arr_myarray1 and arr_myarray2.
- c) Subtract arr_myarray1 from arr_myarray2 and store the result in a new array.
- d) Multiply arr myarray1 and arr myarray2 elementwise.
- e) Do the matrix multiplication of arr_myarray1 and arr_myarray2 and store the result in a new array arr_myarray3.
- f) Divide arr myarray1 by arr myarray2.
- g) Find the cube of all elements of arr myarray1 and divide the resulting array by 2.
- h) Find the square root of all elements of arr_myarray2 and divide the resulting array by 2. The result should be rounded to two places of decimals.

Using the arrays created in Question 1 above, write NumPy commands for the following:

- a) Find the transpose of arr_ones and arr_myarray2.
- b) Sort the array arr_vowels in reverse.
- c) Sort the array arr_myarray1 such that it brings the lowest value of the column in the first row and so on.

```
In [20]: #a
         arr ones.T
         arr_myarray2.T
         #b
         arr_vowels[::-1]
         print("old array is:\n",arr_myarray1)
         print("\nSorted array is:")
         np.sort(arr_myarray1, axis =0)
         old array is:
          [[ 2.7 -2. -19.]
          [ 0.
                  3.4 99.9]
          [ 10.6
                0. 13. ]]
         Sorted array is:
Out[20]: array([[ 0. , -2. , -19. ],
               [ 2.7, 0., 13.],
               [ 10.6, 3.4, 99.9]])
```