

# **Library Usage Python Concepts to Solve Exercise Problems (P\_VTP4)**

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# Numpy Arrays

- Numpy is a **library** for the Python, adding support for large, **multi-dimensional arrays** and **matrices**, along with a large collection of high-level **mathematical functions** to operate on these arrays.
- **Importing** a numpy can be seen as follow:

|                     |                               |
|---------------------|-------------------------------|
| import numpy        | a = numpy.array([4, 6, 2, 9]) |
| import numpy as np  | a = np.array([4, 6, 2, 9])    |
| from numpy import * | a = array([4, 6, 2, 9])       |

# Creating Numpy Arrays

**Example-1: To create an array of `int` datatype**

```
a = array([10, 20, 30, 40, 50], int)
print(a) >> [10 20 30 40 50]
```

**Example-2: To create an array of `float` datatype**

```
a = array([10.1, 20.2, 30.3, 40.4, 50.5], float)
print(a) >> [10.1 20.2 30.3 40.4 50.5]
```

**Example-3: To create an array of `char` datatype**

```
a = array(['a', 'b', 'c', 'd'])
print(a) >> ['a' 'b' 'c' 'd']
```

**Note: No need to specify explicitly the char datatype**



**Example-4: To create an array of `str` datatype**

```
a = array(['abc', 'bcd', 'cde', 'def'], dtype=str)
print(a) >> ['abc' 'bcd' 'cde' 'def']
```

**Example-5: To create an array from tuple**

```
a = array((1, 3, 2))
print(a) >> [1 3 2]
```

**Example-6: To create an 2D array with 2 rows and 3 cols**

```
a = array([[1, 2, 3],
           [4, 5, 6]])
print(a) >> [[1 2 3]
              [4 5 6]]
```

# Creating an array with numpy-arange()

|                       |  |   |
|-----------------------|--|---|
| <b>Syntax</b>         | <code>arange(start, stop, stepsize)</code> |   |
| <b>Example<br/>-1</b> | <code>arange(10)</code>                    | Produces items from 0 – 9 >>[0 1 2 3 4 5 6 7 8 9] |
| <b>Example<br/>-2</b> | <code>arange(5, 10)</code>                 | Produces items from 5 - 9) >> [5 6 7 8 9]         |
| <b>Example<br/>-3</b> | <code>arange(1, 10, 3)</code>              | Produces items from 1, 4, 7 >> [1 4 7]            |
| <b>Example<br/>-4</b> | <code>arange(10, 1, -1)</code>             | Produces items >> [10 9 8 7 6 5 4 3 2]            |
| <b>Example<br/>-5</b> | <code>arange(0, 10, 1.5)</code>            | Produces [0. 1.5 3. 4.5 6. 7.5 9.]                |

# Creating an array with numpy-zeros()

|                       |                                 |  |
|-----------------------|---------------------------------|--|
| <b>Syntax</b>         | <code>zeros(n, datatype)</code> |  |
| <b>Example<br/>-1</b> | <code>zeros(5)</code>           | Produces items [0. 0. 0. 0. 0.]<br>Default datatype is float.  |
| <b>Example<br/>-2</b> | <code>zeros(5, int)</code>      | Produces items [0 0 0 0 0]                                     |
| <b>Example<br/>-3</b> | <code>zeros(1, 2)</code>        | Creating a 1*2 array with all zeros and produces items [0. 0.] |

# Attributes of an Array

- The '**ndim**' attribute represents the number of dimensions or axes of an array.
- The number of dimensions are also called as '**rank**'.

## Example

```
a = array ([1, 2, 3]) # one dimensional array
```

```
print (a.ndim) >> 1
```

```
b = numpy.array([[1, 2, 3],  
                 [4, 5, 6]]) # two dimensional array
```

```
print(b.ndim) >> 2
```

- The **‘size’** attribute gives the total number of items in an array.

## Example

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

```
print (a.size) >> 3
```

```
b = numpy.array([[1, 2, 3],  
                 [4, 5, 6]])
```

```
print(b.size) >> 6
```

- The **‘shape’** attribute gives the shape of the array with corresponding rows and columns.

### Example

```
a = array ([1, 2, 3])  
print (a.shape) >> (1, 3)  
b = numpy.array([[1, 2, 3],  
                 [4, 5, 6]])  
print(b.shape) >> (2, 3)
```

- The **‘dtype’** attribute gives the data type of the elements in the array.

### Example

```
a = array ([1, 2, 3])  
print (a.dtype) >> int32  
b = array ([1.3, 2.1, 3.9])  
print(b.dtype) >> float64
```

# Methods of an Array

- The **'reshape'** method is useful to **change the shape** of an array.

|  |   |
|--|---|
| <b>Example-1:</b>  |   |
| <pre>a = arange(10) #Change the shape as 2 Rows, 5 Cols a = a.reshape(2, 5) print(a)</pre> | <b>Outputs:</b><br>[[0 1 2 3 4]<br>[5 6 7 8 9]] |

|   |  |
|---|--|
| <b>Example-2:</b>   |  |
| <pre>#Change the shape to 5 rows, 2 cols a = a.reshape(5, 2) print(a)</pre> | <b>Outputs:</b><br>[[0 1]<br>[2 3]<br>[4 5]<br>[6 7]<br>[8 9]] |

- The ‘**flatten**’ method is useful to **return copy** of an array collapsed into **one dimension**.

|  |                               |
|--|-------------------------------|
| Example-1:   |                               |
| <pre>#flatten() method a = array([[1, 2], [3, 4]]) print(a) #Change to 1D array a = a.flatten() print(a)</pre> | <pre>Outputs: [1 2 3 4]</pre> |



- The **append()** method appends values along the mentioned axis at the end of the array

| Example-1: Working on 1D   |  |
|--|--|
| <pre>import numpy as np  arr1 = np.arange(3) print("1D arr1 : ", arr1)  arr2 = np.arange(3, 6) print("\n1D arr2 : ", arr2)  # appending the arrays print("\nAppended arr3 : ", <b>np.append</b>(arr1, arr2))</pre> | <p>Outputs:</p> <p>1D arr1 : [0 1 2]</p> <p>1D arr2 : [3 4 5]</p> <p>Appended arr3 :<br/>[0 1 2 3 4 5]</p> |

| Example-2: Working on 2D   |  |
|--|--|
| <pre>import numpy as np  arr1 = np.arange(4).reshape(2, 2) print("2D arr1 : \n", arr1)  arr2 = np.arange(8, 12).reshape(2, 2) print("\n2D arr2 : \n", arr2)  # appending the arrays arr3 = <b>np.append</b>(arr1, arr2) print("\nAppended arr3 by flattened : ", arr3)</pre> | <p>Outputs:</p> <p>2D arr1 :</p> <pre>[[0 1]  [2 3]]</pre> <p>2D arr2 :</p> <pre>[[ 8  9]  [10 11]]</pre> <p>Appended arr3 by flattened : [ 0  1  2 3 8 9 10 11]</p> |

- The **vstack()** function is used to stack the sequence of input arrays vertically to make a single array.

**Example:**

```
import numpy as np

in_arr1 = np.array([ 8, 1, 3] )
print ("1st Input array : \n", in_arr1)

in_arr2 = np.array([ 2, 5, 4] )
print ("2nd Input array : \n", in_arr2)

# Stacking the two arrays vertically
out_arr = np.vstack((in_arr1, in_arr2))
print ("Output vertically stacked
array:\n ", out_arr)
```

Outputs:  
1st Input array :  
[8 1 3]  
2nd Input array :  
[2 5 4]  
Output vertically  
stacked array:  
[[8 1 3]  
[2 5 4]]

# Indexing of an Array

## Example

```
from numpy import *
#Create an 2D array with 3 rows, 3 cols
a = [[1, 2, 3], [4, 5, 6], [7, 8, 9]]
#Display only rows
for i in range(len(a)):
    print(a[i])
#display item by item
for i in range(len(a)):
    for j in range(len(a[i])):
        print(a[i][j], end=' ')
x = array([1, 2, 3, 4, 5])
arr = x[array([1, 3, -1])]
print("\n Elements are : \n",arr)
```

## Output :

```
[1, 2, 3]
[4, 5, 6]
[7, 8, 9]
1 2 3 4 5 6 7 8 9
Elements are :
[2 4 5]
```

- In **Boolean Array Indexing**, elements are returned which **satisfy Boolean expression**.
- It is used **for filtering** the desired element values.

|  |  |
|--|--|
| <b>Example-1:</b>  |  |
| <pre> a = array([10, 40, 80, 50, 100]) # Select numbers greater than 40 print(a[a&gt;40]) a = array([10, 40, 80, 50, 100]) # print the squaring to the multiples of 50 print(a[a%50==0]**2) </pre> | <p><b>Outputs:</b></p> <pre> [50 80 100] [2500 10000] </pre> |

# Basic Slicing

## Example

```
import numpy as np
```

```
# Arrange elements from 0 to 10
```

```
a = np.arange(10)
```

```
print(a) >> [0 1 2 3 4 5 6 7 8 9]
```

```
# a[start:stop:step]
```

```
print(a[-4:8:1]) >> [6 7]
```

```
# The : operator means all elements till the end
```

```
print(a[7:]) >> [7 8 9]
```

# A 3 dimensional array.

```
b = np.array([[[1, 2, 3],[7, 1, 6]],  
              [[4, 3, 1],[1, 2, 2]])
```

# Equivalent to `b[:, :, 1]` and it will print values from every row and column in index 1 using basic slicing with ellipsis

```
print(b[...,1]) >> [[2 1]  
                    [3 2]]
```

# Basic (Vectorized) Operations

## Importance of vectorized operations

1. Operations are faster
  - Adding two arrays in the form  $a + b$  is faster than taking corresponding items of both arrays and then adding them.
2. Syntactically clearer
  - Writing  $a + b$  is clearer than using the loops
3. Provides compact code



|                       |  |
|-----------------------|--|
| <p>Example-<br/>1</p> | <pre> a = array([1, 2, 3, 4]) #Adds 5 to each item of an array print(a + 5) &gt;&gt; [6, 7, 8, 9] #Subtracts 1 from each item of an array print(a - 1) &gt;&gt; [0, 1, 2, 3] #Multiply 2 to each item of an array print(a * 2) &gt;&gt; [2, 4, 6, 8] #Divide 1 to each item of an array print(a / 1) &gt;&gt; [1, 2, 3, 4] #Sum of array elements performing unary operation print(a.sum()) &gt;&gt; 10 #Squaring each element print(a ** 2) &gt;&gt; [1, 4, 9, 16] </pre> |
| <p>Example-<br/>2</p> | <pre> a1 = array([0, 2, 3, 1]) a2 = array([1, 2, 3, 4]) #Adds each item of a1 and a2 Print(a1 + a2) &gt;&gt; [1, 4, 6, 5] </pre>   |

# Bitwise Operations

- `bitwise_and()` function is used to compute the bit-wise AND of two array element-wise.
- `bitwise_or()` function is used to compute the bit-wise OR of two array element-wise.
- `bitwise_xor()` function is used to compute the bit-wise XOR of two array element-wise.

## Example

`a = 11`

`b = 10`

`print(bitwise_and(a, b)) >> 10`

`print(bitwise_or(a, b)) >> 11`

`print(bitwise_xor(a, b)) >> 1`

# Arithmetic Functions

|                         |  |
|-------------------------|--|
| <code>add()</code>      | Add arguments element-wise.  |
| <code>subtract()</code> | Subtract arguments element-wise.   |
| <code>multiply()</code> | Multiply arguments element-wise.   |
| <code>divide()</code>   | Array element from first array is divided by elements from second element. |
| <code>negative()</code> | Numerical negative, element-wise.  |
| <code>power()</code>    | First array elements raised to powers from second array, element-wise.     |

- **Example**

```
import numpy as np
arr1 = [2, 4, 6, 2]
arr2 = [2, 2, 3, 1]
print (np.divide(arr1, arr2)) >> [ 1.  2.  2.  2.]
print (np.multiply(arr1, arr2)) >> [ 4  8 18  2]
in_num1 = 3
in_num2 = 1
print (np.negative(in_num1)) >> -3
print (np.add(in_num1, in_num2)) >> 4
print (np.subtract(in_num1, in_num2)) >> 2
arr1 = [2, 2, 2, 2, 2]
arr2 = [2, 3, 4, 5, 6]
print (np.power(arr1, arr2) ) >> [ 4  8 16 32 64]
```

# Mathematical Functions

- **min(a)** returns the **min value** in the array a.
- **max(a)** returns the **max value** in the array a.
- **around()** helps user to **evenly round** array elements to the given number of decimals.
- **dot()** returns the dot **product value** of elements.
- **isreal()** tests element-wise whether it is **a real number or not** and return the result as a **boolean array**.
- **conj()** helps the user to conjugate any complex number. The conjugate of a complex number is obtained by changing **the sign of its imaginary part**.

- **Example**

(for max() and min())

```
a = [1, 4, 5]
```

```
b = [7, 3, 1]
```

```
print(np.maximum(a,b)) >> [ 7 4 5]
```

```
print(np.minimum(a,b)) >> [ 1 3 1]
```

(for around())

```
in_array = [.4, 2.2, 1.1, 8.6]
```

```
print (around(in_array) ) >> [ 0.  2.  1.  9.]
```

```
in_array = [.43, 3.53, .11]
```

```
print (around(in_array)) >> [ 0.  4.  0.]
```

```
in_array = [.3338, 1.55454, .73415]
```

```
print (around(in_array, decimals = 3)) >> [ 0.334  1.555  0.734]
```

- **Example**

(for dot())

```
import numpy as np
```

```
print("Dot Product of scalar values :", np.dot(3, 2))
```

```
>> Dot Product of scalar values : 6
```

```
vector_a = 3 + 4j
```

```
vector_b = 2 + 5j
```

```
print("Dot Product of vector values :", np.dot(vector_a, vector_b))
```

```
>> Dot Product of vector values : (-14+23j)
```

```
arr_a = np.array([[1, 1], [5, 3]])
```

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

```
print("Dot Product in 2D array : \n", np.dot(arr_a, arr_b))
```

```
>> Dot Product in 2D array : [[ 5  3]
```

```
    [19 11]]
```

- **Example**

(for `isreal()`, `conj()`)

```
import numpy as np
```

```
print(np.isreal([2+1j, 0j]), "\n") >> [False True]
```

```
print(np.isreal([3, 0]), "\n") >> [ True True]
```

```
in_complex1 = 1+3j
```

```
print (np.conj(in_complex1)) >> (1-3j)
```

```
in_complex2 = 8-5j
```

```
print (np.conj(in_complex2)) >> (8+5j)
```



# String Operations

- **lower( )** returns the **lowercase** string from the given string.
- **upper( )** returns the **uppercase** string from the given string.
- **split()** returns a **list of strings** after breaking the given string by the **specified separator**.
- **join()** returns a string in which the elements of sequence have been **joined by str separator**.
- **count()** returns the number of **occurrences of a substring** in the given string.

- **rfind()** returns the **highest index** of the substring if found in given string. If not found then it returns -1.
- **isnumeric()** returns **True** if all characters in the string are numeric characters, Otherwise, it returns **False**.

- **Example**

# converting to lowercase

```
print(np.char.lower('WELCOME')) >> welcome
```

# converting to uppercase

```
print(np.char.upper('hi John')) >> HI JOHN
```

# splitting a string

```
print(np.char.split('Today is holiday')) >> ['Today', 'is', 'holiday']
```

# splitting a string with another format

```
print(np.char.split('Today, is, holiday', sep = ','))
```

```
>>['Today', ' is', ' holiday']
```

# Joining a string with str separator

```
print(np.char.join('-', 'welcome')) >> w-e-l-c-o-m-e
```

# Joining a string with another format by str separator

```
print(np.char.join(['-', ':'], ['geeks', 'for'])) >> ['g-e-e-k-s' 'f:o:r']
```

```
a=np.array(['Welcome', 'from', 'this place'])  
# counting a substring and the output will be printed like [0 0 1]  
print(np.char.count(a,'this')) >> [0 0 1]
```

```
# counting a substring and the output will be printed like [0 0 1]  
print(np.char.count(a, 'om')) >> [1 1 0]
```

```
# Finding a substring and the output will be printed like [0 -1 0]  
print(np.char.rfind(a,'from')) >> [-1  0 -1]
```

# Checking numeric or not and the output will be printed as True or False.

```
print(np.char.isnumeric('Welcome')) >> False
```

```
print(np.char.isnumeric('12')) >> True
```

# String Comparision

- **equal()** checks for string1 == string2 element wise.
- **not\_equal()** checks whether two string is **unequal or not**.
- **greater()** checks whether string1 is **greater** than string2 or not.
- **greater\_equal()** checks whether string1 >= string2 or not.
- **less\_equal()** checks whether string1 is <= string2 or not.
- **less()** checks whether string1 is **lesser than** string2 or not.

## Example

```
import numpy as np
print(np.char.equal('Welcome','hi')) >> False
print(np.char.not_equal('welcome','hi')) >> True
print(np.char.greater('welcome','hi')) >> True
print(np.greater_equal([2., 3.], [1., 2.]) ) >> [ True  True]
a = np.array([1,2])
b = np.array([4,2])
print(a >= b) >> [False  True]
print(a < b) >> [ True False]
print(np.less_equal([4., 2.], [3., 3.]) ) >> [False  True]
print(np.less([4., 2.], [3., 3.])) >> [False  True]
```

# Shuffle Usage

- **random.shuffle()** is used to shuffle the list in place. i.e., it randomizes the order of items in a list.

- **Example**

```
import random
```

```
number_list = [7, 4, 1, 8]
```

```
# Assume the output result after shuffle be [1, 8, 7, 4]
```

```
print(random.shuffle(number_list)) >> [1, 8, 7, 4]
```

```
# To Shuffle two List at once with the same order
```

```
list1_names = ['Jack', 'Emma', 'Smitt']
```

```
list2_id = [70, 50, 90]
```



```
mapIndexPosition = list(zip(list1_names, list2_id))  
random.shuffle(mapIndexPosition)  
list1_names, list2_id = zip(*mapIndexPosition)
```

```
print(" \nLists after Shuffling")
```

```
>> Lists after Shuffling
```

```
print("Employee Names: ", list1_names)
```

```
>> Employee Names: ('Emma', 'Smitt', 'Jack')
```

```
print("Employee ID: ", list2_id)
```

```
>> Employee ID: (50, 90, 70)
```

- **random.shuffle()** does **not work with string** and so, shuffling string can be done by following step by step.
  - **Convert** String to list
  - **Shuffle** the list randomly
  - **Convert** the shuffled list into String

- **Example**

```
import random
sampleStr = "Welcome"
char_list = list(sampleStr) # convert string into list
random.shuffle(char_list) # shuffle list
finalStr = "".join(char_list) # convert list to string
# Assume the resulted shuffled string is wemeocl
print(finalStr) >> wemeoc
```

# Iterating Over Array

- NumPy package contains an iterator object **numpy.nditer**.
- It is an efficient **multidimensional iterator object** using which it is possible to iterate over an array.
- **Example**

```
import numpy as np
```

```
a = np.arange(8) # creating an array using arrange method
```

```
a = a.reshape(2,4) # shape array with 2 rows and 4 columns
```

```
print(a) >> [[0 1 2 3]
```

```
             [4 5 6 7]]
```

```
print('Iterating an array is:')
```

```
for x in np.nditer(a):
```

```
    print(x, end = ",")    >> 0,1,2,3,4,5,6,7,
```

```
print()
```

```
# Creating second array using array method
```

```
b = np.array([5, 6, 7, 8], dtype = int)
```

```
print(b) >> [5 6 7 8]
```

# If two arrays are broadcastable, a combined nditer object is able to iterate upon them concurrently.

```
print('Modified array is:')
```

```
for x,y in np.nditer([a,b]):
```

```
    print("%d:%d" % (x,y), end = ",")
```

```
output >> 0:5,1:6,2:7,3:8,4:5,5:6,6:7,7:8,
```

- Array values can also **be modified** by using **op\_flags** using the iterator `nditer`.
- Its **default** value is **read-only**, but can be set to read-write or write-only mode.
- **Example**

```
import numpy as np
a = np.arange(4)
a = a.reshape(2,2) # shape array with 2 rows and 2 columns
print(a) >> [[0 1]
              [2 3]]

# modifying array values
for x in np.nditer(a, op_flags = ['readwrite']):
    x[...] = 3*x
print('Modified array is:', a) >> [[0 3]
                                   [6 9]]
```

# Statistical functions

- **mean(arr, axis = None)** computes the **arithmetic mean** (average) of the given data (array elements) along the specified axis along which we want to calculate the arithmetic mean.
- **var(arr, axis = None)** computes the **variance** of the given data (array elements) along the specified axis.
- **std(arr, axis = None)** computes the **standard deviation** of the given data (array elements) along the specified axis.
- **axis = 0** means along **the column** and **axis = 1** means working along **the row**.
- To **calculate mean**, if  $\text{arr} = [2, 3, 4, 5]$ , then  $(2+3+4+5)/4 = 3.5$  and the output will be printed like 3.5.

- To **calculate variance**,  $x = 1\ 1\ 1\ 1\ 1$  Standard Deviation = 0 . Variance = 0,  $y = 9, 2, 5, 4, 12, 7, 8, 11, 9, 3, 7, 4, 12, 5, 4, 10, 9, 6, 9, 4$

Step 1 : Mean of distribution  $4 = 7$

Step 2 : Summation of  $(x - x.\text{mean}())**2 = 178$

Step 3 : Finding Mean =  $178 / 20 = 8.9$

This Result is Variance.

- To **calculate standard Deviation**,  $x = 1\ 1\ 1\ 1\ 1$  Standard Deviation = 0 . Variance = 0,  $y = 9, 2, 5, 4, 12, 7, 8, 11, 9, 3, 7, 4, 12, 5, 4, 10, 9, 6, 9, 4$

Step 1 : Mean of distribution  $4 = 7$

Step 2 : Summation of  $(x - x.\text{mean}())**2 = 178$

Step 3 : Finding Mean =  $178 / 20 = 8.9$

This Result is Variance.

Step 4 : Standard Deviation =  $\text{sqrt}(\text{Variance}) = \text{sqrt}(8.9) = 2.983$ .

- **Example**

```
arr = [2, 3, 4, 5] # 1D array
```

```
print("mean of arr : ", np.mean(arr)) >> 3.5
```

```
print("Variance of arr : ", np.var(arr)) >> 1.25
```

```
print("std of arr : ", np.std(arr)) >> 1.11803398875
```

```
# 2D array
```

```
arr = [[4, 1, 0],  
       [5, 6, 2],  
       [3, 2, 4]]
```

```
# mean of the flattened array, calculate the sum of all values and then  
divided by 9
```

```
print("\nmean of arr, axis = None : ", np.mean(arr)) >> 3.0
```

```
# var of the flattened array
```

```
print("\nvar of arr, axis = None : ", np.var(arr)) >> 3.333333333333
```

```
print("\nstd of arr, axis = None : ", np.std(arr)) >> 1.82574185835
```



# mean along the axis = 0 that calculates mean value along each column

```
print("\nmean of arr, axis = 0 : ", np.mean(arr, axis = 0))  
>>[ 4.  3.  2.]
```

# var along the axis = 0

```
print("\nvar of arr, axis = 0 : ", np.var(arr, axis = 0))  
>>[ 0.66666667  4.66666667  2.66666667]
```

# std along the axis = 0

```
print("\nstd of arr, axis = 0 : ", np.std(arr, axis = 0))  
[ 0.81649658  2.1602469  1.63299316]
```

# mean along the axis = 1 that calculates mean value along each row

```
print("\nmean of arr, axis = 1 : ", np.mean(arr, axis = 1))  
>> [ 1.66666667  4.33333333  3.      ]
```

# var along the axis = 1

```
print("\nvar of arr, axis = 1 : ", np.var(arr, axis = 1))  
>> [ 2.88888889  2.88888889  0.66666667]
```

# std along the axis = 1

```
print("\nstd of arr, axis = 1 : ", np.std(arr, axis = 1))  
>> [ 1.69967317  1.69967317  0.81649658]
```

# Sorting functions

- `numpy.sort()` : This function returns a **sorted copy** of an array.
- `numpy.argsort()`:This function returns **the indices** that would sort an array.
- `numpy.lexsort()`:This function returns **an indirect stable sort** using a sequence of keys.

- **Example** (for sort())

```
import numpy as np
```

```
# sort along the first axis
```

```
a = np.array([[1, 5], [7, 3]])
```

```
print("original array is:\n",a) >> [[1 5]  
                                     [7 3]]
```

```
# sorted values for each column with axis = 0
```

```
print ("Along first axis = 0 : \n", np.sort(a, axis = 0))
```

```
>>[[1 3]
```

```
    [7 5]]
```

```
print ("\nAlong none axis : \n", np.sort(a, axis = None) )
```

```
>> [1 3 5 7]
```

- **Example** (for argsort())

```
import numpy as np
```

```
a = np.array([6, 3, 1, 2, 3])
```

```
# unsorted array print
```

```
print('Original array:\n', a) >> [6 3 1 2 3]
```

```
# Sort array indices
```

```
b = np.argsort(a)
```

```
print('Sorted indices of original array->', b) >> [2 3 1 4 0]
```

```
# To get sorted array using sorted indices, c is temp array created of  
same len as of b
```

```
c = np.zeros(len(b), dtype = int)
```

```
for i in range(0, len(b)):
```

```
    c[i]= a[b[i]]
```

```
print('Sorted array->', c) >> [1 2 3 3 6]
```

### **Example (lexsort())**

```
import numpy as np
# Numpy array created First column
a = np.array([3, 1, 3, 6])

# Second column
b = np.array([7, 1, 3, 7])
print('column a, column b')

for (i, j) in zip(a, b):
    print(i, ' ', j)
# Sort by a then by b
ind = np.lexsort((b, a))
print('Sorted indices->', ind)
```

### **Output :**

column a, column b

3 7

1 1

3 3

6 7

Sorted indices-> [1 2 0 3]

## Correspondence Between Each Topic and Related VTPs

| Library Usage Python Concept Topic | Related Problem Number |
|------------------------------------|------------------------|
| Creating Numpy Arrays              | 1, 4                   |
| Attributes of an Array             | 3                      |
| Methods of an Array                | 5, 28, 29              |
| Indexing of an Array               | 6, 8                   |
| Basic Slicing                      | 7                      |
| Basic (Vectorized) Operations      | 2                      |
| Bitwise Operations                 | 12                     |

| Library Usage Python Concept Topic | Related Problem Number |
|------------------------------------|------------------------|
| Arithmetic Functions               | 14                     |
| Mathematical Functions             | 13, 15, 22, 23, 24     |
| String Operations                  | 16, 17, 18             |
| Shuffle Usage                      | 19, 20, 21             |
| Iterating Over Array               | 9, 10, 11              |
| Statistical Functions              | 25, 26, 27             |
| Sorting Functions                  | 30, 31, 32             |



# Conclusion

- This slide introduces numpy library usage concepts for Python Programming.