

✓ Numpy Programming Lab

1

```
!pip install numpy
```

Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (1.23



2

```
import numpy as np
```

3

```
a=[1,2,3,4,5]
arr=np.array(a)
print(arr)
```

```
➞ [1 2 3 4 5]
```

4

```
arr=np.array([[1,2,3],[4,5,6]])
print(arr.shape)
```

```
(2, 3)
```

5

```
arr2=np.array([[[1,2,3]]])
print(arr2.shape)
```

```
(1, 1, 3)
```

6

```
arr= np.array([1,2,3,4], ndmin=3) # 3D array created using ndmin
print(arr)
print(arr.shape)
```

```
arr= np.array([1,2,3,4], ndmin=5) # 5D array created using ndmin
print(arr)
print(arr.shape)
```

```
[[[1 2 3 4]]]
(1, 1, 4)
```

```
[[[[[1 2 3 4]]]]]  
(1, 1, 1, 1, 4)
```

7 Numpy array with random values

```
Data = np.random.random((3,4))  
print("The array is:\n", Data)  
print(type(Data))  
print(np.ndim(Data))  
print(Data.size)  
print(Data.shape)
```

```
The array is:  
[[0.93613865 0.42445364 0.52029089 0.46929546]  
 [0.33137612 0.9693958  0.85476669 0.81680548]  
 [0.5241322  0.14721576 0.4027964  0.18530804]]  
<class 'numpy.ndarray'>  
2  
12  
(3, 4)
```

8 Numpy array of zeros

```
Data = np.zeros((3, 4))  
print("The array is:\n", Data)  
print(type(Data))  
print(np.ndim(Data))  
print(Data.size)  
print(Data.shape)
```

```
The array is:  
[[0. 0. 0. 0.]  
 [0. 0. 0. 0.]  
 [0. 0. 0. 0.]]  
<class 'numpy.ndarray'>  
2  
12  
(3, 4)
```

9 Numpy array of ones

```
Data = np.ones((2, 3))  
print("The array is:\n", Data)  
print(type(Data))  
print(np.ndim(Data))  
print(Data.size)  
print(Data.shape)
```

```
The array is:  
[[1. 1. 1.]  
 [1. 1. 1.]]  
<class 'numpy.ndarray'>  
2
```

```
6
(2, 3)
```

```
# 10 Empty numpy array
```

```
Data = np.empty((3, 4))
print("The array is:\n", Data)
print(type(Data))
print(np.ndim(Data))
print(Data.size)
print(Data.shape)
```

```
The array is:
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]]
<class 'numpy.ndarray'>
2
12
(3, 4)
```

```
# 11 Array with range of elements
```

```
Data = np.arange(12).reshape(3, 4)
print("The array is:\n", Data)
print(type(Data))
print(np.ndim(Data))
print(Data.size)
print(Data.shape)
```

```
The array is:
[[ 0  1  2  3]
 [ 4  5  6  7]
 [ 8  9 10 11]]
<class 'numpy.ndarray'>
2
12
(3, 4)
```

```
# 12 numpy array using linspace
```

```
var_arr10=np.linspace(2,25,num=5) # starting from 2 to 25 it genrates 5 no's who are equa
print(var_arr10)
```

```
[ 2.    7.75 13.5  19.25 25. ]
```

```
# 13
```

```
arr=np.array([[1,2,3],[4,5,6]])
print(arr.shape)
```

```
(2, 3)
```

```
# 14 Number of elements in an array
```

```
print(arr.size) # arrayname,size is used to return number of elements in an array
```

```
6
```

```
# 15
```

```
print(np.ndim(arr)) #get the dimension of arr
```

```
2
```

```
# 16
```

```
arr= np.array([1,2,3],dtype=int)
print(arr)
```

```
arr_type=np.float64(arr)
print(arr_type)
```

```
[1 2 3]
[1. 2. 3.]
```

```
# 17
```

```
arr= np.array([1,2,3],dtype=float)
print(arr)
```

```
arr_type=arr.astype(int)
print(arr_type)
```

```
[1. 2. 3.]
[1 2 3]
```

```
# 18
```

```
arr=np.array([3,7,1,8,4,6,0,2,5])
min_value= np.min(arr,axis = 0)
print(min_value)
max_value= np.max(arr,axis = 0)
print(max_value)
```

```
0
8
```

```
# 19
```

```
arr=np.array([3,7,1,8,4,6,0,2,5])
Data = arr.reshape(3, 3) #Reshaping the arr array to 3D array
print(Data)
```

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

```
# 20
```

```
arr=[1,2,3,4,5,6,7,8]
new_arr=np.array_split(arr,3)
print(new_arr)
```

```
[array([1, 2, 3]), array([4, 5, 6]), array([7, 8])]
```

```
# 21
```

```
arr_v_split=np.array([[1, 2, 3],
                      [4, 5, 6],
                      [7, 8, 9],
                      [10, 11, 12]])

arr_h_split=np.array([[1, 2, 3, 4],
                      [5, 6, 7, 8],
                      [9, 10, 11, 12]])

print(np.vsplit(arr_v_split,2))
print(np.hsplit(arr_h_split,2))
```

```
[array([[1, 2, 3],
        [4, 5, 6]]), array([[ 7,  8,  9],
        [10, 11, 12]])]
[array([[ 1,  2],
        [ 5,  6],
        [ 9, 10]]), array([[ 3,  4],
        [ 7,  8],
        [11, 12]])]
```

```
# 22 adding new dimension
```

```
arr=np.array([1,2,3,4])
new_arr=arr[:,np.newaxis]
print(new_arr)
```

```
[[1]
 [2]
 [3]
 [4]]
```

```
# 23 flattening arrays
```

```
flat_array=new_arr.flatten()
print(flat_array)
```

```
[1 2 3 4]
```

```
# 24 Sort
```

```
arr=np.array([3,7,1,8,4,6,0,2,5])  
np.sort(arr)
```

```
array([0, 1, 2, 3, 4, 5, 6, 7, 8])
```

```
# 25 Concatenate
```

```
x=np.array([[1,2],[3,4]])  
y=np.array([[5,6],[7,8]])  
arr_concat=np.concatenate((x,y))  
print(arr_concat)
```

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

```
# 26
```

```
x=np.array([[1,2],[3,4]])  
y=np.array([[5,6],[7,8]])  
result_add=x+y  
print(result_add)
```

```
[[ 6  8]  
 [10 12]]
```

```
# 27
```

```
x=np.array([1,2,3,4])  
y=np.array([5,6,7,8])  
result_power=np.power(y,x) #y as a power of x  
print(result_power)
```

```
[ 5 36 343 4096]
```

28

```
x=np.array([1,2,3])
mean_value=np.mean(x)
median_value=np.median(x)
sum_value=np.sum(x)
print("Mean:",mean_value)
print("Median:",median_value)
print("Sum:",sum_value)
```

```
Mean: 2.0
Median: 2.0
Sum: 6
```

29

```
x=np.array([[1,2],[3,4]])
y=np.array([[5,6],[7,8]])
result_mat_mul=np.dot(x,y)
print("Matrix Mul: ")
print(result_mat_mul)
```

```
Matrix Mul:
[[19 22]
 [43 50]]
```

30

```
x=np.array([[1,2],[3,4]])
transpose_mat=np.transpose(x)
print(transpose_mat)
```

```
[[1 3]
 [2 4]]
```

✓ 31

np.dot function is used to perform matrix multiplication on arrays in numpy

32

```
x = np.array([[1,2],[3,4]])
y = np.linalg.inv(x) # inversion of x done in y
# print(x)
# print(y)
# print(np.dot(x,y))
```

```
[[1 2]
 [3 4]]
[[-2.  1. ]
 [ 1.5 -0.5]]
[[1.0000000e+00 0.0000000e+00]
 [8.8817842e-16 1.0000000e+00]]
```

33 linear algebra operations

```
arr1=np.array([[1,2,3],[4,5,6],[7,8,9]])
arr2=np.array([[11,12,13],[4,5,6],[17,18,19]])
new1=np.matmul(arr1,arr2)
print(new1)
```

```
[[ 70  76  82]
 [166 181 196]
 [262 286 310]]
```

✓ 34

Missing values or nan values can be done by using the function "numpy.isnan()" it will give us the indexes which are having nan values and when combined with other function which is "numpy.logical_not()" where the boolean values will be reversed.

34

```
a = np.array([[1,np.nan,3,2,5],[6,8,7,np.nan,10]])
print(np.isnan(a))
```

```
[[False  True False False False]
 [False False False  True False]]
```

35 explore np.where

```
arr_where=np.array([[1, 2, 3],
                    [4, 5, 6],
                    [7, 8, 9],
                    [10, 11, 12]])
print(np.where(arr_where==3))
```

```
(array([0]), array([2]))
```

36 Element wise multiplication

```
arr1=np.array([[1,2,3],[4,5,6],[7,8,9]])
arr2=np.array([[11,12,13],[4,5,6],[17,18,19]])
new=np.multiply(arr1,arr2)
print(new)
```



```
[[ 11  24  39]
 [ 16  25  36]
 [119 144 171]]
```

37 Calculate mean and median

```
arr1=np.array([[1,2,3],[4,5,6]])
j=1
for i in arr1:
    print(j,"row meadian is ",np.median(i)," mean is ",np.mean(i))
    j+=1

1 row meadian is  2.0  mean is  2.0
2 row meadian is  5.0  mean is  5.0
```

38 Standard deviation

```
arr1=np.array([3,7,1,8,4,6,0,2,5])
print(np.std(arr1))

2.581988897471611
```

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copy(view) has a base while the deep copy(copy) has no base since it is a copy of base array.Changes made in base array cant be viewed in a deep copy(copy) while the base can be viewed in a view(shallow copy).

39 copy vs views

```
ar=np.array([1,2,3])
print(ar)
x=ar.copy()      # copy created
print(x)

y=ar.view()      # view created
print(y)
y[1]=45

print(y)
print(ar)

[1 2 3]
[1 2 3]
[1 2 3]
[ 1 45  3]
[ 1 45  3]
```

```
# 40 Slicing for extracting sub array
```

```
arr=np.array([[1,2,3],[4,5,6],[7,8,9]])
sub= arr[0:2,:]
print(sub)
```

```
[[1 2 3]
 [4 5 6]]
```

```
#41 Indexing
```

```
y=np.array([10,20,30,40,50])
print(y[4]) # value at index 4 of y
```

```
50
```

```
#42 How to slice elements
```

```
er=np.array([1,2,3,4,5])
sl=er[:4] # slicing from start to 4th element (4th index is not inclusive)
print(sl)
```

```
[1 2 3 4]
```

```
#43 select every nth element
```

```
nth=np.array([1,2,3,4,5,6,7,8,9,10])
sl=nth[::2]
print(sl)
```

```
[1 3 5 7 9]
```

```
#44 extract specific column
```

```
arr = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
col = arr[:, 1] #extracting second column
print(col)
```

```
[2 5 8]
```

```
#45 Reversing of rows
```

```
import numpy as np
arr = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
arr_reversed = np.flip(arr, axis=0) # reversing using flip function
print(arr_reversed)
```

```
[[7 8 9]
 [4 5 6]]
```

```
[1 2 3]]
```

#46 Extracting diagonal elements

```
arr = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
print(np.diag(arr))
```

```
[1 5 9]
```

✓ 47

In NumPy, boolean arrays are NumPy arrays with array components that are either “True” or “False” 1. Slicing in NumPy is performed in the same way as it is performed in Python lists 23. To slice a NumPy array using a boolean array, we can use the boolean array as a mask to filter the elements of the original array. For example, if we have an array a and a boolean array b, we can slice a using b as follows: a[b] This will return a new array that contains only the elements of a that correspond to True values in b 4.

#47 Slicing with Boolean arrays

```
ar=np.array([True,False,False,True,True])
print(ar)
print(ar[1:])
```

```
[ True False False  True  True]
[False False  True  True]
```

#48

#To select elements based on a condition in Python,
#we can use list comprehension to filter the list and create a new list with only the des

```
arr = np.array([1, 2, 3, 4, 5, 6, 7, 8, 9, 10])
result = arr[arr > 5]
print(result)
```

```
[ 6  7  8  9 10]
```

#49 Slice a 3D Numpy array

```
arr_3d = np.array([[[1, 2, 3], [4, 5, 6]], [[7, 8, 9], [10, 11, 12]]])
sliced_arr = arr_3d[0:2, 0:2, 0:2]
print(sliced_arr)
```

```
[[[ 1  2]
  [ 4  5]]
```

```
[[ 7  8]
 [10 11]]
```

#50 Extract specific rows based on condition

```
arrrr = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
row = arrrr[0]
print(row)
```

```
[1 2 3]
```

✓ 51

"The term broadcasting describes how NumPy treats arrays with different shapes during arithmetic operations. Subject to certain constraints, the smaller array is "broadcast" across the larger array so that they have compatible shapes"

#52 Saving numpy array to a file and loading it back

```
with open('test.npy', 'wb') as f:
    np.save(f, np.array([1, 2]))
    np.save(f, np.array([1, 3]))
with open('test.npy', 'rb') as f:
    a = np.load(f)
    b = np.load(f)
print(a, b)
```

```
[1 2] [1 3]
```

53

```
# You can check if two NumPy arrays are equal element-wise using the == operator.
# When you use the == operator on two NumPy arrays, it performs element-wise
# comparison and returns a boolean array of the same shape, where each element
# indicates whether the corresponding elements in the two arrays are equal or n
```

```
# Import numpy
import numpy as np
```

```
arr = np.array([2,4,5,7,9])
print("First array:", arr)
arr1 = np.array([2,4,6,7,10])
print("Second array:", arr1)
```

```
# Check array equal element wise
# Using == operator
result = arr == arr1
print("Check the element-wise equality:", result)
```

```

First array: [2 4 5 7 9]
Second array: [ 2  4  6  7 10]
Check the element-wise equality: [ True  True False  True False]

```

54. USE OF AND, OR, NOT

```

arr_and = np.arange(5)
print(np.logical_and(arr_and>1, arr_and<4))
print(np.logical_or(arr_and>1, arr_and<4))
print(np.logical_not([True, False, 0, 1]))

```

```

[False False  True  True False]
[ True  True  True  True  True]
[False  True  True False]

```

#55

The numpy.meshgrid function is used to create a rectangular grid out of two given
one-dimensional arrays representing the Cartesian indexing or Matrix indexing.

```

ar1=np.array([1,2,3,4])
ar2=np.array([1,2,3,4])
print(np.meshgrid(ar1,ar2))

[array([[1, 2, 3, 4],
        [1, 2, 3, 4],
        [1, 2, 3, 4],
        [1, 2, 3, 4]]), array([[1, 1, 1, 1],
        [2, 2, 2, 2],
        [3, 3, 3, 3],
        [4, 4, 4, 4]])]

```

56 Element wise division

```

ar1=np.array([1,2,3,4])
ar2=np.array([1,2,2,2])
print(np.divide(ar1,ar2))

```

```

[1.  1.  1.5 2. ]

```

#57 np.unique function

With the help of np.unique() method, we can get the unique values from an array
given as parameter in np.unique() method.

```

a=np.unique([1, 1, 2, 2, 3, 3])
print(a)

```

```

[1 2 3]

```

#58 correlation coefficient of two numpy arrays

```
ar1=np.array([1,2,3,4])
ar2=np.array([1,2,2,2])
print(np.corrcoef(ar1, ar2) )

[[1.          0.77459667]
 [0.77459667 1.          ]]
```

#59 Use of np.pad function

numpy.pad() function is used to pad the Numpy arrays. Sometimes there is a need to perform padding in Numpy arrays, then numPy.pad() function is used. The function returns the padded array of rank equal to the given array and the shape will increase according to pad_width.

```
arr = np.array([1, 3, 2, 5, 4] )
```

```
pad_arr = np.pad(arr, (3, 2), 'linear_ramp',
                  end_values=(-4, 5))
```

```
print(pad_arr)
```

```
arr = np.array([1, 3, 2, 5, 4] )
```

```
for i in arr:
    print(i)
```

```
ar = np.array([[1, 2, 3], [5, 6, 7], [9, 10, 11]])
```

```
for i in ar:
    for x in i:
        print(x)
```

```
[-4 -3 -1  1  3  2  5  4  4  5]
1
3
2
5
4
1
2
3
5
6
7
9
10
11
```

```
# 60 Iterate over each element in a numpy array
```

```
a=np.array([1,2,3,4])
for x in a:                #for 1D array
    print(x)
```

```
b=np.array([[1,2,3],[4,5,6]])
for x in b:                #for 2D array
    for y in x:
        print(y)
```

```
1
2
3
4
1
2
3
4
5
6
```

```
# 61. Generate an array of 20 elements linearly spaced between 5 and 15.
```

```
var_arr = np.linspace(5, 15, 20)
print(var_arr)
```

```
[ 5.          5.52631579  6.05263158  6.57894737  7.10526316  7.63157895
  8.15789474  8.68421053  9.21052632  9.73684211 10.26315789 10.78947368
 11.31578947 11.84210526 12.36842105 12.89473684 13.42105263 13.94736842
 14.47368421 15.          ]
```

```
# 62. Write a function to compute the Manhattan distance between two 1D arrays,
# array1 and array2, of equal length.
```

```
def manhattan_distance(array1, array2):
    return np.sum(np.abs(array1 - array2))
```

```
a=np.array([1,2,3,4])
b=np.array([6,7,8,9])
```

```
x= manhattan_distance(a,b)
print(x)
```

```
20
```

```
# 63. Flatten a given 3x3 2D array into a 1D array.
```

```
array_2d = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
array_1d = array_2d.flatten()
print(array_1d)
```

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

64. Write a script to create an array and then check if it contains any NaN values.

```
array = np.array([1, 2, np.nan, 4, 5])
new_nan = np.isnan(array).any()
print(new_nan)
```

True

65. Create a 5x5 identity matrix using NumPy.

```
identity_matrix = np.eye(5)
print(identity_matrix)
```

```
[[1. 0. 0. 0. 0.]
 [0. 1. 0. 0. 0.]
 [0. 0. 1. 0. 0.]
 [0. 0. 0. 1. 0.]
 [0. 0. 0. 0. 1.]]
```

66. Determine the rank of a given 4x4 matrix.

```
matrix = np.array([[1,4,-1,1],[2,4,-2,2],[3,6,-6,3],[4,8,-4,4]])
rank = np.linalg.matrix_rank(matrix)
print(matrix)
print("Rank:",rank)
```

```
[[ 1  4 -1  1]
 [ 2  4 -2  2]
 [ 3  6 -6  3]
 [ 4  8 -4  4]]
Rank: 3
```

67. Sort the elements of a given 1D array in descending order.

```
array = np.array([6,3,7,9,3,5,2,1,8])
array_sorted = np.sort(array)
```

```
print(array_sorted)
```

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

68. Slice a 3x3x3 array to extract a 2x2x2 sub-array from its corner.

```
array_3d = np.array([[[3,4,5],[1,2,3],[2,3,4]],[[2,3,4],[6,7,8],[4,5,6]],[[5,6,7],[9,8,7]]])
sub_array = array_3d[:2, :2, :2]
```

```
print(sub_array)
```

```
[[[3 4]
 [1 2]]
 [[2 3]
 [6 7]]]
```



```
# 69. Calculate the median of a given 1D array without using NumPy's median function.
```

```
array = np.array([7,3,4,7,9,1,2,6,9])
```

```
array_sorted = np.sort(array)
```

```
n = len(array_sorted)
```

```
if n % 2 == 0:
```

```
    median = (array_sorted[n//2-1] + array_sorted[n//2]) / 2
```

```
else:
```

```
    median = array_sorted[n//2]
```

```
print("Median is:",median)
```

```
    Median is: 6
```

```
# 70. Concatenate two given 2D arrays of the same shape along the second axis.
```