Numpy Array

- 1. NumPy is a Python library used for working with arrays.
- 2. It also has functions for working in domain of linear algebra, fourier transform, and matrices.
- 3. NumPy was created in 2005 by Travis Oliphant. It is an open source project and you can use it freely.
- 4. NumPy is short for "Numerical Python".

Why We Use

- 1. We can use lists in place of array but they are slow and takes more time to process
- 2. Numpy array are faster than lists
- 3. The array object in NumPy is called ndarray, it provides a lot of supporting functions that make working with ndarray very easy.
- 4. Arrays are very frequently used in data science, where speed and resources are very important.

Numpy Array Written In

NumPy is a Python library and is written partially in Python, but most of the parts that require fast computation are written in C or C++.

The source code for NumPy is located at this github repository https://github.com/numpy/numpy

Working with Numpy

Install

pip install numpy

Import

· import numpy

Create Alias

- · import numpy as np

```
%timeit [i**1000 for i in range(10)]
⊋ 21.6 μs ± 701 ns per loop (mean ± std. dev. of 7 runs, 10000 loops each)
#Numpy Array
import numpy as np
%timeit np.arange(10)**1000
\rightarrow 2.08 \mus \pm 87.6 ns per loop (mean \pm std. dev. of 7 runs, 100000 loops each)
def p1(n):
  1=[]
  for i in range(10):
    1.append(i**n)
%timeit p1(1000)
\rightarrow 21.9 µs ± 1.15 µs per loop (mean ± std. dev. of 7 runs, 10000 loops each)
def p2(n):
  1=[2,1,3,4]
  return 1
%timeit p2(1000)
```

```
102 ns ± 0.975 ns per loop (mean ± std. dev. of 7 runs, 10000000 loops each)
def p3(n):
  l=[2,'c','python',3.4,8]
  return l
%timeit p3(1000)
\rightarrow 107 ns ± 2.03 ns per loop (mean ± std. dev. of 7 runs, 10000000 loops each)
def p4(n):
  l=['c++','c','python','java']
  return l
%timeit p4(1000)
\longrightarrow 218 ns \pm 10 ns per loop (mean \pm std. dev. of 7 runs, 1000000 loops each)
%timeit [2,1,3,4]*1000
\rightarrow 7.75 \mus \pm 1.37 \mus per loop (mean \pm std. dev. of 7 runs, 100000 loops each)
import numpy as np
%timeit np.array([2,1,3,4])*1000
\rightarrow 2.04 \mus \pm 146 ns per loop (mean \pm std. dev. of 7 runs, 100000 loops each)

    Create Our First Numpy Array

import numpy
arr = numpy.array([2,1,4,5])
print(arr)
→ [2 1 4 5]

    Create List

1=[2,1,4,5]
print(1)
→ [2, 1, 4, 5]

    Create Alias and Use

import numpy as np
arr = np.array([2,1,4,5])
print(arr)
→ [2 1 4 5]
np.__version__
Check Type of Numpy Array
type(arr)
→ numpy.ndarray

    Create Different Dimension Arrays
```

```
#Zero Dimension Array
import numpy as np
a = np.array(2)
print(a)
→ 2
#One Dimension Array
import numpy as np
b = np.array([2,1,4])
print(b)
⋺▼ [2 1 4]
#Two Dimension Array
import numpy as np
c = np.array([[1,2],[5,6]])
print(c)
→ [[1 2]
      [5 6]]
#Three Dimension Array
import numpy as np
d = np.array([[[1,2,3],[4,5,6]],[[4,2,1],[6,5,9]]])
print(d)
→ [[[1 2 3]
       [4 5 6]]
      [[4 2 1]
       [6 5 9]]]
#Four Dimension Array
import numpy as np
e = np.array([[[[1,2,3],[4,5,6]],[[4,2,1],[6,5,9]]],[[[1,2,3],[4,5,6]],[[4,2,1],[6,5,9]]]))
→ [[[[1 2 3]
       [4 5 6]]
       [[4 2 1]
       [6 5 9]]]
      [[[1 2 3]
       [4 5 6]]
       [[4 2 1]
       [6 5 9]]]]

→ Check Dimensions

a.ndim
→ 0
b.ndim
→ 1
c.ndim
→ 2
d.ndim
→ 3
e.ndim
```

_____ 4

```
import numpy as np
a = np.array(42)
b = np.array([1, 2, 3, 4, 5])
c = np.array([[1, 2, 3], [4, 5, 6]])
d = np.array([[[1, 2, 3], [4, 5, 6]], [[1, 2, 3], [4, 5, 6]]])
print(a.ndim)
print(b.ndim)
print(c.ndim)
print(d.ndim)
→ 0
     1
     2
#Make Six Dimension Array
import numpy as np
arr = np.array([45,56], ndmin=6)
print(arr)
print('number of dimensions :', arr.ndim)
→ [[[[[45 56]]]]]
     number of dimensions : 6
#Create a list of five elements
1 = []
for i in range(5):
  x = int(input('enter element : '))
  1.append(x)
print(1)
→ enter element : 23
     enter element : 56
     enter element : 78
     enter element : 12
     enter element : 89
     [23, 56, 78, 12, 89]
#Create an array of list
import numpy as np
print(np.array(1))
→ [23 56 78 12 89]

→ Different Type of Arrays

   1. 1 D Array - For Zeros and Ones: np.zeros(columns)
   2. 2 D Array - For Zeros and Ones: np.zeros((rows,columns))
   3. 3 D Array - For Zeros and Ones: np.zeros((blocks,rows,columns))
Note: Use the same pattern for N Dimension Array

✓ 1D Array

np.zeros(5)
→ array([0., 0., 0., 0., 0.])

✓ 2D Array

np.zeros((2,5))
⇒ array([[0., 0., 0., 0., 0.],
            [0., 0., 0., 0., 0.]])
```

```
np.zeros((3,2))
 → array([[0., 0.],
                  [0., 0.],
[0., 0.]])

→ 3D Array

np.zeros((2,4,3))
 \rightarrow array([[[0., 0., 0.],
                   [0., 0., 0.],
                   [0., 0., 0.],
[0., 0., 0.]],
                  [[0., 0., 0.],
[0., 0., 0.],
                   [0., 0., 0.],
[0., 0., 0.]]])
np.ones(5)
 \Rightarrow array([1., 1., 1., 1., 1.])
np.ones((2,5))
 ⇒ array([[1., 1., 1., 1., 1.], [1., 1., 1., 1.]])
np.ones((3,2))
 \rightarrow array([[1., 1.],
                  [1., 1.],
                  [1., 1.]])
np.ones((2,4,3))
 → array([[[1., 1., 1.],
                   [1., 1., 1.],
[1., 1., 1.],
[1., 1., 1.]],
                  [[1., 1., 1.],
                   [1., 1., 1.],
                   [1., 1., 1.],
[1., 1., 1.]])

✓ nD Array

np.zeros((2,4,3,5))
 → array([[[[0., 0., 0., 0., 0.],
                     [0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]],
                   [[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]],
                   [[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]],
                   [[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]]],
                  [[[0., 0., 0., 0., 0.],
                     [0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]],
                    [[0., 0., 0., 0., 0.],
                     [0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]],
```

```
[[0., 0., 0., 0., 0.],
                   [0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]],
                 [[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.],
[0., 0., 0., 0., 0.]]]])
np.ones((2,4,3,5))
 → array([[[[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.]],
                  [[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.]]],
                 [[[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.]],
                  [[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.]]])

→ Range Array

import numpy as np
np.arange(10)
 \rightarrow array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
Empty Array
arr = np.empty(4)
arr
 ⇒ array([4.64730676e-316, 0.00000000e+000, 6.42301491e+246, 1.07220128e+200])
arr = np.empty((4,3))
arr
array([[0.3, 1. , 1. ], [1. , 1. ], [1. , 1. ], [1. , 1. ],
                [1., 0.5, 1.]])
arr = np.empty((4,3,2))
 → array([[[-0.25, -0.5],
                  [ 0. , -0.25],
[ 0.25, -0.5 ]],
                 [[ 0.5 , -0.25],
                 [ 0.25, 0. ],
[ 0.5 , 0.25]],
```

```
[[ 0.25, 0.5 ],
 [ 0. , 0.25],
 [-0.25, 0.5 ]],
 [[-0.5 , 0.25],
 [-0.25, 0. ],
 [-0.5 , -0.25]]])
```

→ Diagonal Array

✓ Unit Array

→ Diagonal Array with diagonal value 1 for any array

 ✓ Scaler Array

→ Diagonal Array having any value on diagonal

→ Requirement of elements in a range

```
np.linspace(5,20,num=5)
 \rightarrow array([ 5. , 8.75, 12.5 , 16.25, 20. ])
np.linspace(1,2,num=6)
 \rightarrow array([1., 1.2, 1.4, 1.6, 1.8, 2.])
   Array of Random Numbers

✓ rand():

function defined in the random module and generate random numbers between 0 and 1
np.random.rand(5)
 array([0.78353799, 0.35884583, 0.7781049 , 0.53631117, 0.0480623 ])
np.random.rand(2)
 array([0.72876549, 0.0213757])
np.random.rand(2,5)
 array([[0.25509119, 0.02836408, 0.66851913, 0.43954464, 0.61377896],
             [0.80932241, 0.65691926, 0.02561393, 0.75664088, 0.43993807]])
np.random.rand(4,3)
 → array([[0.45290732, 0.01048686, 0.72177979],
             [0.45867937, 0.11581793, 0.42573422],
             [0.03891808, 0.75602106, 0.71881213],
            [0.58199685, 0.67852448, 0.04859443]])
np.random.rand(2,3,4)
 ⇒ array([[[0.97870652, 0.63394042, 0.20357622, 0.40777274],
              [0.93031236, 0.48700301, 0.17537171, 0.91525033],
             [0.78691878, 0.15725978, 0.13209009, 0.25560429]],
            [[0.46974439, 0.55123979, 0.4414393 , 0.21367058],
             [0.16929465, 0.40143534, 0.27216358, 0.59436984],
             [0.95838275, 0.61196414, 0.02827833, 0.97535392]]])
randn():
function defined in the random module and generate both positive negative values close to zero
np.random.randn(5)
 ⇒ array([-1.43807162, 0.03075063, -1.87372668, 0.84186615, -0.60564534])
np.random.randn(5,2)
 array([[ 0.4604701 , -0.09062114], [ 1.15499609, -0.02187291],
             [-0.33198836, -1.05932534],
            [-2.57421669, 0.18440957],
[0.77772461, 2.48276682]])

✓ ranf()

function used to generate float values from 0 to 1 but 1 is excluded. This function is already defined in random module
np.random.ranf(5)
 ⇒ array([0.83455828, 0.30537477, 0.80257385, 0.61140916, 0.62422214])
```

```
np.random.ranf((5,3))
 → array([[0.53061725, 0.63117972, 0.7988413 ],
             [0.51479241, 0.22930213, 0.55897052],
             [0.41294443, 0.56159002, 0.48164991],
             [0.63415499, 0.50495113, 0.43211107],
             [0.27391068, 0.6133115 , 0.23488454]])
np.random.ranf((5,3,2))
 → array([[[0.80630164, 0.73211639],
              [0.22788459, 0.25205377],
              [0.97699895, 0.6047011 ]],
             [[0.11525073, 0.19391645],
[0.0893254 , 0.56887413],
[0.80538081, 0.21248227]],
             [[0.95817743, 0.89354202], [0.78114798, 0.03682275],
              [0.23164338, 0.44349187]],
             [[0.04295169, 0.45659137],
              [0.15534465, 0.69145448],
              [0.27471862, 0.13690532]],
             [[0.97604934, 0.57625531],
              [0.0345954 , 0.61233601],
[0.6625823 , 0.95718834]]])

→ randint(): generate numbers between range

np.random.randint(1,20,5)
 → array([ 7, 15, 15, 14, 19])
Check Data Type
x = np.array([3,2,5])
x.dtype
 dtype('int64')
x = np.array([3.0,2.6,5.5])
x.dtype
 dtype('float64')
x = np.array(['p','y'])
x.dtype
 → dtype('<U1')
x = np.array(['cpp','python'])
x.dtype
 → dtype('<U6')</pre>

    Change Data Type

x = np.array([2,4.7], dtype='int8')
x.dtype
 dtype('int8')
x = np.array([2,4.7], dtype='i')
x.dtype
 → dtype('int32')
```

```
x = np.array([2,4.7], dtype='float16')
x.dtype
dtype('float16')
x = np.array([2,4.7], dtype='f')
x.dtype
→ dtype('float32')
x = np.array([2,4.7])
np.int32(x).dtype
dtype('int32')
x = np.array([2,4.7])
np.int64(x).dtype
→ dtype('int64')
x = np.array([2,4.7])
np.float64(x).dtype
→ dtype('float64')
x = np.array([2,4,7])
y = x.astype(float)
→ array([2., 4., 7.])

    Arithematic Operations

   1. add(a1,a2)
   2. subtract(a1,a2)
   3. multiply(a1,a2)
   4. divide(a1,a2)
   5. mod(a1,a2)
   6. power(a1,a2)
   7. reciprocal(a1)
#Add with every element of array
x = np.array([2,4,6])
\rightarrow array([12, 14, 16])
#Subtract with every element of array
x = np.array([2,4,6])
x-2
\rightarrow array([0, 2, 4])
#Multiply with every element of array
x = np.array([2,4,6])
→ array([ 4, 8, 12])
#Divide with every element of array
x = np.array([2,4,6])
x/2
→ array([1., 2., 3.])
#Floor Divide with every element of array
x = np.array([2,4,6])
```

```
x//2
```

```
\rightarrow array([1, 2, 3])
#Mod with every element of array
x = np.array([2,4,6])
\rightarrow array([0, 0, 0])
#Add arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
х+у
\rightarrow array([ 3, 7, 11])
x = np.array([2,4,6])
y = np.array([1,3,5])
np.add(x,y)
→ array([ 3, 7, 11])
x = np.array([2,4,6])
y = np.array([1,3,5])
z = np.array([3,1,3])
x+y+z
→ array([ 6, 8, 14])
#with add function, we can add two arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
z = np.array([3,1,3])
np.add(x,y,z)
→ array([ 3, 7, 11])
#with nesting, we can add
x = np.array([2,4,6])
y = np.array([1,3,5])
z = np.array([3,1,3])
np.add(np.add(x,y),z)
→ array([ 6, 8, 14])
#Add arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
х-у
\rightarrow array([1, 1, 1])
#Add arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
np.subtract(x,y)
\rightarrow array([1, 1, 1])
x = np.array([2,4,6])
y = np.array([1,3,5])
z = np.array([3,1,3])
x-y-z
→ array([-2, 0, -2])
#only work with two arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
```

```
z = np.array([3,1,3])
np.subtract(x,y,z)
\rightarrow array([1, 1, 1])
x = np.array([2,4,6])
y = np.array([1,3,5])
z = np.array([3,1,3])
np.subtract(np.subtract(x,y),z) \\
→ array([-2, 0, -2])
#Multiply arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
х*у
→ array([ 2, 12, 30])
x = np.array([2,4,6])
y = np.array([1,3,5])
np.multiply(x,y)
→ array([ 2, 12, 30])
#Divide arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
x/y
                , 1.33333333, 1.2
→ array([2.
                                              ])
x = np.array([2,4,6])
y = np.array([1,3,5])
np.divide(x,y)
→ array([2.
                , 1.33333333, 1.2
x = np.array([2,4,6])
y = np.array([1,3,5])
x//y
\rightarrow array([2, 1, 1])
#mod of arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
np.mod(x,y)
\rightarrow array([0, 1, 1])
x = np.array([2,4,6])
y = np.array([1,3,5])
х%у
\rightarrow array([0, 1, 1])
#power of arrays
x = np.array([2,4,6])
y = np.array([1,3,5])
np.power(x,y)
→ array([ 2, 64, 7776])
x = np.array([2,4,6])
y = np.array([1,3,5])
x**y
→ array([ 2, 64, 7776])
```

```
#reciprocal of array
x = np.array([2,4,6])
np.reciprocal(x)
\rightarrow array([0, 0, 0])
x = np.array([2,4,6])
→ array([0, 0, 0])
x = np.array([[2,4,6],[1,6,5]])
print(np.add(x,y))
print(np.subtract(x,y))
print(np.multiply(x,y))
print(np.divide(x,y))
print(np.mod(x,y))
print(np.power(x,y))
print(np.reciprocal(x))
[[ 3 7 11]
[ 2 9 10]]
     [[1 1 1]
      [0 3 0]]
     [[ 2 12 30]
      [ 1 18 25]]
     [[2.
                  1.33333333 1.2
                                        ĺ)
      [1.
                             1.
     [[0 1 1]
      [0 0 0]]
     [[ 2 64 7776]
[ 1 216 3125]]
     [[0 0 0]]
      [1 0 0]]

→ Other Arithematic Functions:

   1. min()
   2. max()
   3. sqrt()
   4. sin()
   5. cos()
   6. cumsum()
import numpy as np
x = np.array([12,4,26])
print(np.min(x))
print(np.max(x))
print(np.argmin(x)) #tell us the min value index
print(np.argmax(x)) #tell us the max value index
print(np.sqrt(x))
print(np.sin(x))
print(np.cos(x))
print(np.cumsum(x))
₹
    4
     26
                            5.09901951]
     [3.46410162 2.
     [-0.53657292 -0.7568025  0.76255845]
     [ 0.84385396 -0.65364362 0.64691932]
     [12 16 42]
x = np.array([[12,4,26],[3,6,5]])
print(np.min(x))
print(np.max(x))
print(np.argmin(x)) #tell us the min value index
print(np.argmax(x)) #tell us the max value index
print(np.sqrt(x))
print(np.sin(x))
print(np.cos(x))
print(np.cumsum(x))
```

```
→ 3
     3
     [[3.46410162 2.
                            5.09901951]
      [1.73205081 2.44948974 2.23606798]]
     [ 0.14112001 -0.2794155 -0.95892427]]
     [[ 0.84385396 -0.65364362  0.64691932]
      [-0.9899925 0.96017029 0.28366219]]
     [12 16 42 45 51 56]
#column wise
x = np.array([[12,4,26],[3,6,5]])
print(np.min(x,axis=0))
print(np.max(x,axis=0))
print(np.argmin(x,axis=0)) #tell us the min value index
print(np.argmax(x,axis=0)) #tell us the max value index
print(np.sqrt(x))
print(np.cumsum(x,axis=0))
→ [3 4 5]
     [12 6 26]
     [1 0 1]
     [0 1 0]
     [[3.46410162 2.
                            5.09901951]
      [1.73205081 2.44948974 2.23606798]]
     [[12 4 26]
     [15 10 31]]
#row wise
x = np.array([[12,4,26],[3,6,5]])
print(np.min(x,axis=1))
print(np.max(x,axis=1))
print(np.argmin(x,axis=1)) #tell us the min value index
\label{eq:print(np.argmax(x,axis=1))} \text{ #tell us the max value index}
print(np.sqrt(x))
print(np.cumsum(x,axis=1))
→ [4 3]
     [26 6]
    [1 0]
     [2 1]
     [[3.46410162 2.
                            5.09901951]
     [1.73205081 2.44948974 2.23606798]]
     [[12 16 42]
     [ 3 9 14]]

✓ Shape of Array

#1D Array
x = np.array([12,4,26])
x.shape
→ (3,)
#2D Array
x = np.array([[12,4,26],[3,6,5]])
x.shape
\rightarrow (2, 3)
#3D Array
x = np.array([[[12,4,26],[3,6,5]],[[12,4,26],[3,6,5]]])
x.shape
→ (2, 2, 3)
```

```
#make 6 dimension array
x = np.array([12,4,26], ndmin=6)
print(x)
print(x.ndim)
print(x.shape)

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```

```
Change the shape of Array
#one dimension to two dimension
x = np.array([1,2,3,4,5,6,7,8])
print(x)
print(x.shape)
print(x.ndim)
rx = x.reshape(4,2)
print(rx)
print(rx.shape)
print(rx.ndim)
→ [1 2 3 4 5 6 7 8]
     (8,)
     [[1 2]
     [3 4]
      [5 6]
      [7 8]]
     (4, 2)
\#one dimension to three dimension
x = np.array([1,2,3,4,5,6,7,8])
print(x.shape)
print(x.ndim)
rx = x.reshape(2,2,2)
print(rx)
print(rx.shape)
print(rx.ndim)
→ [1 2 3 4 5 6 7 8]
     (8,)
     [[[1 2]
      [3 4]]
      [[5 6]
       [7 8]]]
     (2, 2, 2)
x = np.array([[1,2,3,4],[5,6,7,8]])
print(x)
print(x.shape)
print(x.ndim)
rx = x.reshape(-1)
print(rx)
print(rx.shape)
print(rx.ndim)
→ [[1 2 3 4]
      [5 6 7 8]]
     (2, 4)
     [1 2 3 4 5 6 7 8]
     (8,)
x = np.array([[1,2,3,4],[5,6,7,8]])
print(x)
print(x.shape)
print(x.ndim)
rx = x.reshape(4,2)
print(rx)
```

```
print(rx.shape)
print(rx.ndim)
→ [[1 2 3 4]
     [5 6 7 8]]
(2, 4)
     [[1 2]
      [3 4]
      [5 6]
      [7 8]]
     (4, 2)
x = np.array([[1,2,3,4],[5,6,7,8]])
print(x)
print(x.shape)
print(x.ndim)
rx = x.reshape(2,2,2)
print(rx)
print(rx.shape)
print(rx.ndim)
→ [[1 2 3 4]
      [5 6 7 8]]
     (2, 4)
     [[[1 2]
      [3 4]]
      [[5 6]
      [7 8]]]
     (2, 2, 2)
```

✓ Indexing and Slicing of Array

✓ 1D Array

```
x = np.array([3,2,1,7,9,8])
print(x[2])
print(x[-2])
print(x[2:])
print(x[:2])
print(x[2:4])
print(x[-2:])
print(x[:-2])
print(x[-4:-2])
print(x[:])
print(x[::])
print(x[::1])
print(x[::2])
print(x[::3])
print(x[::4])
print(x[1::2])
print(x[::-1])
print(x[-6:-2:1])
 → 1
     [1 7 9 8]
     [3 2]
     [1 7]
     [9 8]
     [3 2 1 7]
[1 7]
     [3 2 1 7 9 8]
     [3 2 1 7 9 8]
     [3 2 1 7 9 8]
     [3 1 9]
     [3 9]
     [2 7 8]
     [8 9 7 1 2 3]
```

[3 2 1 7]

```
print(x[-6:-2:2])
print(x[-6:-1:3])
print(x[-2:-6:-1])
print(x[-2:-6:-2])
→ [3 1]
     [3 7]
     [9 7 1 2]
     [9 1]

✓ 2D Array

x = np.array([[2,1,4,3],[5,1,4,7]])
print(x[1,2])
print(x[1,2:])
print(x[1,:2])
print(x[1,2:4])
print(x[1, 2])
print(x[-1, -2])
print(x[1, 2:])
print(x[-1, -2:])
print(x[1, :2])
print(x[-1, :-2])
print(x[1, 2:4])
print(x[-1, -2:])
→ 4
     [4 7]
[5 1]
     [4 7]
     4
     [4 7]
     [4 7]
     [5 1]
     [5 1]
     [4 7]
[4 7]

→ 3D Array

x = np.array([[[2,1,4,3],[5,1,4,7]],[[2,1,4,3],[5,1,4,7]]])
print(x[1,1,2])
print(x[1,1,2:])
print(x[-1, -1, -2])
print(x[-1, -1, -2:])
₹
     [4 7]
4
     [4 7]

✓ Iteration on Array

#1D array
x = np.array([2,1,5,4])
for i in x:
  print(i)
→ 2
     1
     5
```

```
#2D array
x = np.array([[2,1,7,8],[3,2,9,8]])
for i in x:
 for j in i:
    print(j)
₹
     8
     2
     9
#3D array
x = np.array([[[2,1,4,3],[5,1,4,7]],[[2,1,4,3],[5,1,4,7]]])
for i in x:
 for j in i:
    for k in j:
      print(k)
<del>_</del>_
    2
     1
     4
     3
     5
     7
     2
     4
     3
     5
     1
     4
#Iterate using nditer() function over 3D Array
x = np.array([[[2,1,4,3],[5,1,4,7]],[[2,1,4,3],[5,1,4,7]]])
for i in np.nditer(x):
 print(i)
₹
     4
     3
     1
     4
     7
     2
     1
     4
     3
     5
     1
     4
#Iterate using ndenumerate() function over 3D Array
x = np.array([[[2,1,4,3],[5,1,4,7]],[[2,1,4,3],[5,1,4,7]]])
for i in np.ndenumerate(x):
 print(i)
→ ((0, 0, 0), np.int64(2))
     ((0, 0, 1), np.int64(1))
     ((0, 0, 2), np.int64(4))
     ((0, 0, 3), np.int64(3))
     ((0, 1, 0), np.int64(5))
     ((0, 1, 1), np.int64(1))
     ((0, 1, 2), np.int64(4))
((0, 1, 3), np.int64(7))
     ((1, 0, 0), np.int64(2))
     ((1, 0, 1), np.int64(1))
((1, 0, 2), np.int64(4))
     ((1, 0, 3), np.int64(3))
     ((1, 1, 0), np.int64(5))
((1, 1, 1), np.int64(1))
```

```
((1, 1, 2), np.int64(4))
     ((1, 1, 3), np.int64(7))
x = np.array([[[2,1,4,3],[5,1,4,7]],[[2,1,4,3],[5,1,4,7]]])
for i,j in np.ndenumerate(x):
  print(i,j)
→ (0, 0, 0) 2
     (0, 0, 1) 1
     (0, 0, 2) 4
     (0, 0, 3) 3
     (0, 1, 0) 5
     (0, 1, 1) 1
     (0, 1, 2) 4
(0, 1, 3) 7
     (1, 0, 0) 2
     (1, 0, 1) 1
     (1, 0, 2) 4
     (1, 0, 3) 3
     (1, 1, 0) 5
     (1, 1, 1) 1
     (1, 1, 2) 4
     (1, 1, 3) 7
```

Copy and View in array

- 1. copy does not affect the original data
- 2. view affect the original data

```
x = np.array([2,1,4,3])
y = x.copy()
y[0] = 10
x[-1]=50
print(x)
print(y)

→ [ 2 1 4 50]
     [10 1 4 3]
x = np.array([2,1,4,3])
y = x.view()
x[-1]=50
y[0] = 10
print(x)
print(y)
[10 1 4 50]
[10 1 4 50]

✓ Join Array

x = np.array([2,1,4,3])
y = np.array([5,1,4,7])
z = np.concatenate((x,y))
print(z)
→ [2 1 4 3 5 1 4 7]
x = np.array([[2,1],[4,3]])
y = np.array([[5,1],[4,7]])
z = np.concatenate((x,y))
print(z)
→ [[2 1]
      [4 3]
      [5 1]
      [4 7]]
#concatenate along column wise
x = np.array([[2,1],[4,3]])
y = np.array([[5,1],[4,7]])
```

```
z = np.concatenate((x,y),axis=0)
print(z)
→ [[2 1]
     [4 3]
      [5 1]
      [4 7]]
#concatenate along row wise
x = np.array([[2,1],[4,3]])
y = np.array([[5,1],[4,7]])
z = np.concatenate((x,y),axis=1)
print(z)
→ [[2 1 5 1]
     [4 3 4 7]]
x = np.array([[2,1],[4,3]])
y = np.array([[5,1],[4,7]])
z = np.stack((x,y))
print(z)
→ [[[2 1]
      [4 3]]
      [[5 1]
      [4 7]]]
x = np.array([[2,1],[4,3]])
y = np.array([[5,1],[4,7]])
z = np.hstack((x,y))
print(z)
→ [[2 1 5 1]
      [4 3 4 7]]
x = np.array([[2,1],[4,3]])
y = np.array([[5,1],[4,7]])
z = np.vstack((x,y))
print(z)
→ [[2 1]
      [4 3]
      [5 1]
     [4 7]]
```

End of Code

Represent Data in the Form of Classical Bits

```
bits = ''.join(format(ord(char), '08b') for char in 'a')
print("Classical bits:", bits)

Classical bits: 01100001

bits = format(ord('a'), '08b')
print("Classical bit:", bits)

Classical bit: 01100001

bits = format(ord('b'), '08b')
print("Classical bit:", bits)

Classical bit: 01100010

bits = ''.join(format(ord(i), '08b') for i in 'python')
print("Classical bit:", bits)

Classical bit: 01110000011110010111010001101000011011101101110
```

```
bits = ''.join(format(ord(i), '08b') for i in 'python programming')
print("Classical bit:", bits)
 bits = ''.join(format(ord(i), '08b') for i in 'mycode')
print("Classical bit:", bits)
 bits = ''.join(format(ord(i), '08b') if type(i) == str else format(i, '08b') for i in [2,1,6,'p'])
print("Classical bits:", bits)
 Transical bits: 0000001000000010000011001110000
import struct
data = [2, 1, 6.7, 'p']
bits = ''
for i in data:
      if isinstance(i, str):
            bits += format(ord(i), '08b')
      elif isinstance(i, int):
            bits += format(i, '08b')
      elif isinstance(i, float):
            float_bytes = struct.pack('>d', i) # IEEE 754 double precision
            bits += ''.join(format(b, '08b') for b in float_bytes)
      else:
            raise TypeError(f"Unsupported type: {type(i)}")
print(bits)
 import pickle
data = [123, 4.56, 'hello', True, None, (1, 2), [3, 4], {'a': 5}, complex(1, 2)]
# Serialize the entire list to bytes
binary_data = pickle.dumps(data)
# Convert bytes to classical bits (8-bit per byte)
bits = ''.join(format(byte, '08b') for byte in binary_data)
print(bits)
 pip install qiskit

→ Collecting qiskit
           Downloading qiskit-2.1.0-cp39-abi3-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (12 kB)
        Collecting rustworkx>=0.15.0 (from qiskit)
           \label{lower_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_power_pow
        Requirement already satisfied: numpy<3,>=1.17 in /usr/local/lib/python3.11/dist-packages (from qiskit) (2.0.2)
        Requirement already satisfied: scipy>=1.5 in /usr/local/lib/python3.11/dist-packages (from qiskit) (1.15.3)
        Requirement already satisfied: dill>=0.3 in /usr/local/lib/python3.11/dist-packages (from qiskit) (0.3.7)
        Collecting stevedore>=3.0.0 (from qiskit)
           Downloading stevedore-5.4.1-py3-none-any.whl.metadata (2.3 kB)
        Requirement already satisfied: typing-extensions in /usr/local/lib/python3.11/dist-packages (from qiskit) (4.14.0)
        Collecting pbr>=2.0.0 (from stevedore>=3.0.0->qiskit)
           Downloading pbr-6.1.1-py2.py3-none-any.whl.metadata (3.4 kB)
        Requirement already satisfied: setuptools in /usr/local/lib/python3.11/dist-packages (from pbr>=2.0.0->stevedore>=3.0.0->qiskit) (75.2.0
        Downloading\ qiskit-2.1.0-cp39-abi3-manylinux\_2\_17\_x86\_64.manylinux2014\_x86\_64.whl\ (7.5\ MB)
                                                                          - 7.5/7.5 MB 68.2 MB/s eta 0:00:00
        Downloading rustworkx-0.16.0-cp39-abi3-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (2.1 MB)
                                                                          - 2.1/2.1 MB 79.8 MB/s eta 0:00:00
        Downloading stevedore-5.4.1-py3-none-any.whl (49 kB)
```

```
- 49.5/49.5 kB <mark>3.4 MB/s</mark> eta 0:00:00
     Downloading pbr-6.1.1-py2.py3-none-any.whl (108 kB)
                                                - 109.0/109.0 kB 7.9 MB/s eta 0:00:00
     Installing collected packages: rustworkx, pbr, stevedore, qiskit
     Successfully installed pbr-6.1.1 qiskit-2.1.0 rustworkx-0.16.0 stevedore-5.4.1
from qiskit import QuantumCircuit
binary = '101010'
qc = QuantumCircuit(len(binary))
# Initialize qubits according to bits
for i, bit in enumerate(reversed(binary)): # Qiskit uses little-endian
   if bit == '1':
       qc.x(i) # Apply X-gate to flip |0> to |1>
print(qc.draw('text'))
<del>_</del>__
!pip install qiskit-aer
→ Collecting qiskit-aer
       Downloading qiskit aer-0.17.1-cp311-cp311-manylinux 2 17 x86 64.manylinux2014 x86 64.whl.metadata (8.3 kB)
     Requirement already satisfied: qiskit>=1.1.0 in /usr/local/lib/python3.11/dist-packages (from qiskit-aer) (2.1.0)
     Requirement already satisfied: numpy>=1.16.3 in /usr/local/lib/python3.11/dist-packages (from qiskit-aer) (2.0.2)
     Requirement already satisfied: scipy>=1.0 in /usr/local/lib/python3.11/dist-packages (from qiskit-aer) (1.15.3)
     Requirement already satisfied: psutil>=5 in /usr/local/lib/python3.11/dist-packages (from qiskit-aer) (5.9.5)
     Requirement already satisfied: python-dateutil>=2.8.0 in /usr/local/lib/python3.11/dist-packages (from qiskit-aer) (2.9.0.post0)
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.0->qiskit-aer) (1.17.0)
     Requirement already satisfied: rustworkx>=0.15.0 in /usr/local/lib/python3.11/dist-packages (from qiskit>=1.1.0->qiskit-aer) (0.16.0)
     Requirement already satisfied: dill>=0.3 in /usr/local/lib/python3.11/dist-packages (from qiskit>=1.1.0->qiskit-aer) (0.3.7)
     Requirement already satisfied: stevedore>=3.0.0 in /usr/local/lib/python3.11/dist-packages (from qiskit>=1.1.0->qiskit-aer) (5.4.1)
     Requirement already satisfied: typing-extensions in /usr/local/lib/python3.11/dist-packages (from qiskit>=1.1.0->qiskit-aer) (4.14.0)
     Requirement already satisfied: pbr>=2.0.0 in /usr/local/lib/python3.11/dist-packages (from stevedore>=3.0.0->qiskit>=1.1.0->qiskit-aer)
     Requirement already satisfied: setuptools in /usr/local/lib/python3.11/dist-packages (from pbr>=2.0.0->stevedore>=3.0.0->qiskit>=1.1.0->
     Downloading qiskit_aer-0.17.1-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (12.4 MB)
                                                - 12.4/12.4 MB 64.8 MB/s eta 0:00:00
     Installing collected packages: qiskit-aer
     Successfully installed qiskit-aer-0.17.1
!pip install qiskit qiskit-aer --quiet
from qiskit import QuantumCircuit
from qiskit_aer.primitives import Sampler
# Step 1: Encode 42 → binary '101010'
binary = format(42, '06b')
# Step 2: Build the circuit with 6 qubits and 6 classical bits
qc = QuantumCircuit(6, 6) # 6 qubits, 6 classical bits
# Initialize qubits according to the binary string
for i, bit in enumerate(reversed(binary)): # Little-endian
   if bit == '1':
        qc.x(i)
# Add measurement to all qubits
qc.measure(range(6), range(6))
# Step 3: Simulate using Sampler
sampler = Sampler()
```

```
job = sampler.run([qc]) # Wrap the circuit in a list
result = job.result()
# Output
print("Quantum Circuit:\n", qc.draw('text'))
print("\nQubit Output (quasi-probabilities):")
print(result.quasi_dists[0])
→ Quantum Circuit:
     q_2:
     q_3:
     q_4:
     q_5:
                    2 4 1 3
     Qubit Output (quasi-probabilities):
     {42: 1.0}
from \ qiskit.quantum\_info \ import \ Statevector
# Create a superposition state |\psi\rangle = \alpha|\theta\rangle + \beta|1\rangle
sv = Statevector([1/2**0.5, 1/2**0.9])
print("\alpha =", sv.data[0])
print("β =", sv.data[1])
\Rightarrow \alpha = (0.7071067811865475+0j)
     \beta = (0.5358867312681466+0j)
```

from qiskit.quantum_info import Statevector

```
# Create a superposition state |\psi\rangle = \alpha|\theta\rangle + \beta|1\rangle sv = Statevector([1/2**0.5j+5, 1/2**0.9j+9]) print("\alpha =", sv.data[0]) print("\beta =", sv.data[1])
```

 α = (5.940542104683244-0.3396771251026685j) β = (9.811645689625399-0.5841500445198216j)

Start coding or generate with AI.