

Part-2

Python Libraries

Numpy Fundamental :-

What is numpy?

- Numpy is fundamental package for scientific computing.
- It is Library provide multidim array object (such as matrix) and assortment for fast operation on arrays including mathematical.

Numpy array -

At the core of Numpy package is the ndarray object. The encapsulate n-dim array of homogenous data type.

Numpy Arrays v/s Python Sequence :-

- Numpy

- fixed array.
- if change size create new array delete previous one.
- Convenience in Mathematical concept

• Matplotlib, Scikit learn originate from it

Python List

- dynamic array.
- Manipulate old array

Numpy \leftrightarrow scientific computing

for

in \Rightarrow stage w/ main

library

\Rightarrow Numpy array \Rightarrow array in Number \Rightarrow

store in \Rightarrow 3D in computation file \Rightarrow

Numpy is a God father.

Importing library -

import Numpy as np.

(4,5)

RANKA

DATE / /

PAGE

Creating Numpy Array

1) np.array

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

```
print(a)
```

2) 2D + 3D

Vector - 1D

Matrix - 2D

Tensor - 3D

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

3) dtype

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

complex

4) np.arange - like a range - in list

```
np.arange(1,11,2)
```

5) np with reshape

```
np.arange(1,13).reshape(5,2)
```

```
np.arange(16).reshape(2,2,2,2)
```

6) np.ones((3,4)) / np.zeros((3,1))

Note: default value
in most of case is
float

```
array([[1., 1., 1., 1.],  
       [1., 1., 1., 1.],  
       [1., 1., 1., 1.]])
```

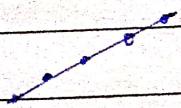
7) np.random.random((3,4))

random number array

8) np.linspace

Linearly separable points generate

```
np.linspace(1,10,10)
```



9) np.identity

```
np.identity(3)
```

→ to draw Identity matrix.

Array Attributes $a_1 = np.array(10)$ $a_2 = np.array(12, dtype=float) \circ reshape(3,4)$ $a_3 = np.array(8) \circ reshape(2,2,2)$ $\rightarrow ndim$

No. of dimension

1-D

 $a_1.ndim = 1$

2-D

 $a_2.ndim = 2$

3-D

 $a_3.ndim = 3$

\rightarrow Shape \rightarrow dimension of first rows & column exist \rightarrow $(2,2,2)$

`print(a3.shape) \rightarrow (2,2,2)` array (E)

 \rightarrow size \rightarrow No. of items \rightarrow \sum $a_1.size = 10$ $a_2.size = 12$ $a_3.size = 8$ \rightarrow item size \rightarrow Item size occupy \rightarrow Σ \rightarrow dtype`print(a2.dtype) \rightarrow float 64``(a2.dtype) \rightarrow int 32`

Changing Data type→ `a.astype`

→ If array → datatype change array
`a3.astype(np.int32)`

Array OperationsEx:`a1 = np.arange(12).reshape(3,4)``a2 = np.arange(12,24).reshape(3,4)`Scalar Operation

1) Arithmetic Operation → Multiply each entry by scalar
`a1 ** 2`

2) Relational → `a2 == 15` → Compare each element

Vector Operations

1) Arithmetic →

`a1 ** a2`→ Array function`a1 = np.random.random(3,3)``np.mean(a1)`

sum

↑

1) `mean, min, sum, prod``np. 0 → col 1 → row``np.mean(a1, axis=1)`2) `mean / median / std / var``np.mean(a1)``np.median(a1, axis=0)`

Trigonometric function

`np.sin(a)`

dot product

Criteria \rightarrow $m \times n$ $n \times m$

$a_2 = np.arange(12).reshape(3, 4)$

$a_3 = np.arange(12, 24) \rightarrow reshape(4, 3)$

`np.dot(a_2, a_3)`

$$\begin{bmatrix} 2 & 4 & 5 \\ 5 & 6 & 8 \end{bmatrix} \cdot \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix} = \begin{bmatrix} 2 \times 1 + 4 \times 2 + 5 \times 3 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

\rightarrow Logarithm & Exponent -

Logarithm

`np.log(a)`

Exponent

`np.exp`

\rightarrow round / floor / ceil

`np.ceil()`

a_2 [row, column]

RANKA

DATE: / /

PAGE: / /

Indexing and slicing

$a_1 = \text{np.array}(10)$
 a

Indexing → a_2 [row, column]
 $a_2[2, 3]$

Slicing → $a_1 = [1, 2, 3, 4, 5, 6, 7]$
 $a_1[1:3]$

$$a_1 = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

1st row

$a_1[0, :]$

2nd row

3rd column

$a_1[:, 2]$

(i) $[[0, 1, 2, 3]]$

$a_2 = [4, 5, 6, 7]$
 $[8, 9, 10, 11]]$

(i) $\begin{bmatrix} 0 & 3 \\ 8 & 11 \end{bmatrix}$, $a_2[:, ::2, ::3]$

ii) $\begin{bmatrix} 1 & 3 \\ 9 & 11 \end{bmatrix}$, $a_2[:, ::2, ::::2]$

iii) $[4, 9]$, $a_2[1, 0::3]$

Iterating:

1) $a_1 = [1, 2, 3, 4]$
 for i in a_1 :
 print(i)

2) $a_2 = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$
 for i in a_2 :
 print(i)

Every elements

for i in np.nditer(a):
 print(i)

Reshape &

1) reshape

2) Transpose

np.transpose(a2)

$a_2.T$

3) ravel - convert any n dim array in 1D

a2.ravel()

Stacking:

horizontal stacking

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

Horizontal

$$\begin{bmatrix} 1 & 2 & 5 & 6 \\ 3 & 4 & 7 & 8 \end{bmatrix}$$

Vertical

np.hstack((a1, a2))

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \\ 7 & 8 \end{bmatrix}$$

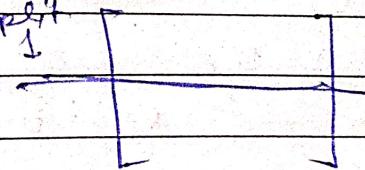
np.vstack(a2)

splitting: opposite of straiting

$$\begin{bmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ a_7 & a_8 & a_9 \end{bmatrix}$$

→ n split

VSP⁴



List can store non homogeneous data

RANKA

DATE / /

PAGE

Numpy array vs Python list

- 1) Speed
- 2) Memory
- 3) Convenience.

Speed

1) List

```
a = [i for i in range(10**7)]
```

```
b = [i for i in range(
```

```
c = a + [ ]
```

```
for i in range(len(a)):
```

```
c.append(a[i] + b[i])
```

~~REMEMBER~~

2) arrays

```
import numpy as np
```

```
a1 = np.arange(10**7)
```

```
a2 = np.arange(10**7, 20**7)
```

```
c = a1 + a2
```

Memory

int - 8 (8B)

```
a = [i for i in range(10000000)]
```

int - 16 (320000)

```
import sys
```

int - 32 "

```
sys.getsizeof(a)
```

Advanced Indexing

1) Fancy Indexing.

$$a_1 = \begin{bmatrix} 0 & 1 & 2 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \\ 9 & 10 & 11 \end{bmatrix}$$

$a_1[0:2, 3]$

$a_1[:, [0, 2, 3]]$

2) Boolean Indexing.

$$a = \begin{bmatrix} 50 & 65 \\ 72 & 24 \end{bmatrix}$$

i) Find all number greater than 50

ii) Find all number which are even

i) $a[a > 50]$

ii) $a[a \% 2 == 0]$

iii) $a[(a > 50) \& (a \% 2 == 0)]$

Broadcasting

The term broadcasting describes how NumPy treats arrays with different shapes during arithmetic operations.

The smaller array is "broadcast" across the larger array so that they have compatible shapes.

Same Shape

$a = np.array(6) \cdot reshape(2, 3)$

$b = np.array(6, 12) \cdot reshape(2, 3)$

$print(a)$

$print(b)$

$print(a+b)$

different shape

$a = np.array(6) \cdot reshape(2, 3)$

$b = np.array(6, 9) \cdot reshape(1, 3)$

$print(a)$

$print(b)$

$print(a+b)$

$$\begin{bmatrix} 0 & 1 & 2 \end{bmatrix} + \begin{bmatrix} 5 & 5 & 5 \end{bmatrix} = \begin{bmatrix} 5 & 6 & 7 \end{bmatrix}$$

$$\begin{array}{|c|c|c|} \hline 1 & 1 & 1 \\ \hline 1 & 1 & 1 \\ \hline 1 & 1 & 1 \\ \hline \end{array} + \begin{array}{|c|c|c|} \hline 0 & 1 & 2 \\ \hline 0 & 1 & 2 \\ \hline 0 & 1 & 2 \\ \hline \end{array} = \begin{array}{|c|c|c|} \hline 1 & 2 & 3 \\ \hline 1 & 2 & 3 \\ \hline 1 & 2 & 3 \\ \hline \end{array}$$

$$(3, 2) \quad (3,) \rightarrow (1, 3)$$

2D 1D

$$(3, 3, 3) \quad (3,) \rightarrow (1, 1, 3)$$

Working with Mathematical formulas :-

1) Sigmoid

$$\frac{1}{1+e^{-x}}$$

2) Mean Square Error 3) ^{Gregorian cross} entropy,

2) @ Sigmoid

def sigmoid (array):

return 1 / (1 + np.exp(-array))

a = np.arange(10)

Sigmoid (a)

2) Mean Square Error

$$= \frac{1}{n} \sum_{i=1}^n (y_i - \bar{y})^2$$

meshgridLec Numpy Trick1) np.sort $a = np.random.randint(1, 100, 25)$ a $np.sort(a)$ $np.sort(:: -1)$ 2d- $b = np.random.randint(1, 100, 24) \rightarrow \text{reshape}(6, 4)$ b $np.sort(b) \leftarrow \text{row wise}$ $np.sort(b, axis=1) \leftarrow \text{column wise}$ 2) np.append. \approx $np.append(a, 200)$ $np.append(b, np.ones(b.shape[0], 1)), axis=1)$

3) np.concatenate

 $c = \text{np.array}(6).reshape(2, 3)$
 $d = \text{np.array}(6, 12).reshape(2, 3)$

print(c)

print(d)

np.concatenate((c, d), axis=1)

4) np.unique — provide all the unique vals

 $e = \text{np.array}(1, 2, 3, 4, 5, 2, 4, 1, 1, 2, 3, 4)$

np.unique(e)

5) np.expand_dims — convert 1d - 2d - 3d

np.expand_dims(a, axis=0), shape

np.expand_dims()

6) np.where → This function returns the indices of elements in an input array where the given condition is satisfied

* Find all value greater than 50

np.where(a > 50)

* replace all values > 50 with 0

np.where(condition, true, false)

np.where(a > 50, 0, a)

7) np.argmax → It return the maximum element of the array in a particular axis

np.argmax(a)

np.argmin

↳ Index position

axis → 0 Columbus
1 row

RANKA

DATE / /

PAGE

np.cumsum & (cumulative sum)
Virat Kohli & 3147 career & first over 212

a

np.cumsum(a)

b

np.cumsum(b)

/ np.cumsum(b, axis=1)

np.comprod &

np.comprod(a)

percentile & used to compute n^{th} percentile of given data along specified axis
np.percentile.

np.percentile(9, 100)

np.percentile(a, 50)

$$P = \frac{n}{N} \times 100$$

$n \rightarrow$ No. of values below 'x'

$N \rightarrow$ Total no. of population

np.histogram &

5 15 9 17 25 23 29 → histogram.

a

np.histogram(a, bins = [0, 10, 20, 30, 40, 50, 60, 70, 80, 90, 100])

np.correlcoef

Return Pearson product-moment correlation coefficient.

Salary = np.array([20000, 40000, 25000, 35000, 60000])

Experience = np.array([1, 3, 2, 4, 2])

np.correlcoef(Salary, Experience)

np.isin

We can see that one array having values are checked in different

→ Array → Multiple Item to search for which exist? & if it
Item given → $\{20, 40\}$ in $\{10, 20, 30, 40, 50, 60, 70, 80, 90, 100\}$

a.

items = [10, 20, 30, 40, 50, 60, 70, 80, 90, 100]

a[np.isin(a, items)]

np.flip

Array → Flip → $\{20, 40\}$

np.flip(a)

np.flip(b, axis=1)

np.put

change the

a

np.put(a, [0, 1], [110, 530])

np.delete → delete item on given index

np.delete(a, 10, 1)

Set functions

- 1) np.union1d 2) np.intersect1d 3) np.setdiff1d
 4) np.setxor1d 5) np.in1d

$m = \text{np.array}([1, 2, 3, 4, 5])$

$n = \text{np.array}([3, 4, 5, 6, 7])$

$\text{np.union1d}(m, n)$

$\text{np.intersect1d}(m, n)$

$\text{np.setdiff1d}(m, n)$

$\text{np.setxor1d}(m, n)$

np.clip :- used to clip (limits) the values in an array.

a

$\text{np.clip}(a, \text{min.})$

$\text{np.clip}(a, a-\text{min}=25, a-\text{max}=75)$

75, 75, 28, 50, 38