
Numpy

(Numerical Python)

Why Numpy ?



NumPy

- Designed for efficiency on large arrays of data.
- stores data in a contiguous block of memory
- uses much less memory than Python lists

NumPy ndarray vs list

- One of the key features of NumPy is its N-dimensional array object, or ndarray, which is a fast, flexible container for large datasets in Python.
- Whenever you see “array,” “NumPy array,” or “ndarray” in the text, with few exceptions they all refer to the same thing: the ndarray object.

Creating ndarrays

Using list of lists

```
import numpy as np
```

```
data2 = [[1, 2, 3, 4],  
         [5, 6, 7, 8] ]
```

```
arr = np.array(data2)
```

```
arr.ndim -----> # 2
```

```
arr.shape -----> # (2,4)
```

```
arr.size -----> 2*4 = 8
```

NdArrays

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

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

np.ndim ???

np.shape ???

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Creating ndarrays

```
array = np.zeros((2,3))
```

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

```
array = np.arange(start , end, step_size)
```

```
array = np.arange(0, 10, 2)
```

```
[0, 2, 4, 6, 8]
```

```
array = np.ones((2,3))
```

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

```
array = np.random.randint(0, 10, (3,3))
```

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

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Element-Wise Operations

- Can we perform this in python ?
- `arr1 = [1,2,3]`
- `arr2 = [2,3,4]`

`arr1 * arr2`

Let's see

Arithmetic with NumPy Arrays

- Any arithmetic operations between equal-sized arrays applies the operation element-wise:

```
arr = np.array(  
[[1., 2., 3.],  
 [4., 5., 6.]  
)  
print(arr * arr)  
[[ 1.  4.  9.]  
 [16. 25. 36.]]  
print(arr - arr)  
[[0. 0. 0.]  
 [0. 0. 0.]]
```

Arithmetic with NumPy Arrays

- Comparisons between arrays of the same size yield boolean arrays:

```
arr2 = np.array([[0., 4., 1.], [7., 2., 12.]])
```

```
print(arr2)
```

```
[[ 0.  4.  1.]  
 [ 7.  2. 12.]]
```

```
print(arr2 > arr)
```

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

Slicing

```
arr = np.arange(10)
print(arr)      # [0 1 2 3 4 5 6 7 8 9]
```

arr [start , end , step_size] (step_size by default 1)

```
print(arr[5:8]) #[5 6 7]
```

```
arr[5:8] = 12
print(arr)     #[ 0 1 2 3 4 12 12 12 8 9]
```

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Slicing

- As we know that the **end** is exclusive, so if we have:

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

```
arr [9:0:-1]
```

It will give us : [9, 8, 7, 6, 5, 4, 3, 2, 1] # 0 is not included

I want to get the 0th element as well i.e.

[9, 8, 7, 6, 5, 4, 3, 2, 1, **0**].

What will be the result if instead of

```
[9,0,-1]
```

I write

```
[9,-1,-1]
```

Will I get the 0th element ?

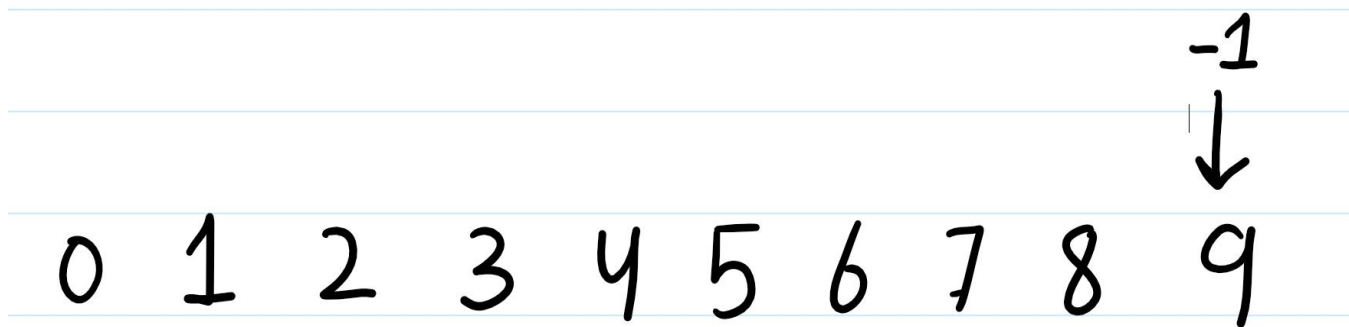
We will get :

[] (an empty numpy array)

Why ?

We have `[9:-1:-1]`

Due to negative indexing , it is considering that index -1 to be :



(We provided the same start and end points)

Slicing in multi-dimensional arrays

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16
mat [<input type="text"/> , <input type="text"/>]			
↙ Rows ?		↙ Columns ?	

.

			Columns	
Rows	1	2	3	4
	5	6	7	8
	9	10	11	12
	13	14	15	16

mat [: , 2:4]

↓
All Rows

↓
Columns from
2 to 4 (Exclusive)

1	2	3	4	
5	6	7	8	
9	10	11	12	
13	14	15	16	
mat [1:3 , 1:3]				
Row #		Col #		
1, 2		1, 2		

1	2	3	4	
5	6	7	8	
9	10	11	12	
13	14	15	16	
mat [3 , 2:4]				

Activity

How can we slice it
to display the last 2 elements
of middle array ?

1	2	3
4	5	6
7	8	9

Activity

How can we slice it
to display the last 2 elements
of middle array?

1	2	3
4	5	6
7	8	9

`mat [1 , 1:3]`

Activity

index	0	1	2	3	
0	1	2	3	4	
1	5	6	7	8	
2	9	10	11	12	
3	13	14	15	16	

Activity

index	0	1	2	3		
0	1	2	3	4		
1	5	6	7	8		
2	9	10	11	12		
3	13	14	15	16		
		arr [0:3:2 , 1:4:2]				

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Broadcasting

- We did the following earlier in numpy (Element-wise operation) :
- $[1,2,3] * [2,3,4]$ (np arrays)

What if we have this scenario :

$[1,2,3] * 2$

Let's first see what happens with python lists.

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Broadcasting

$$[1, 2, 3] * 2 = [2, 4, 6]$$

$$[1, 2, 3] * [2 \text{ } 2 \text{ } 2]$$

2 is actually
copied (Broadcasted)



(3,3)

1	1	1
1	1	1
1	1	1

+

(1,3)

0	1	2
0	1	2
0	1	2

=

(3,3)

1	2	3
1	2	3
1	2	3

Examples

Rules

- 1) Dimensions of the array A and B are compared from **right to left**.
- 2) If there is a mismatch , either of them **should be 1**.
- 3) If there is **no dimension** , it is supposed to be 1

A (2d array): 5 x 4
B (1d array): 1
Result (2d array): 5 x 4

A (2d array): 5 x 4
B (1d array): 4
Result (2d array): 5 x 4

A (3d array): 15 x 3 x 5
B (3d array): 15 x 1 x 5
Result (3d array): 15 x 3 x 5

A (3d array): 15 x 3 x 5
B (2d array): 3 x 5
Result (3d array): 15 x 3 x 5

A (3d array): 15 x 3 x 5
B (2d array): 3 x 1
Result (3d array): 15 x 3 x 5

Examples (Failure)

Example 1:

A (2d array): 2 x 1
B (3d array): 8 x 4 x 3 # second from last dimensions mismatched

Example 2:

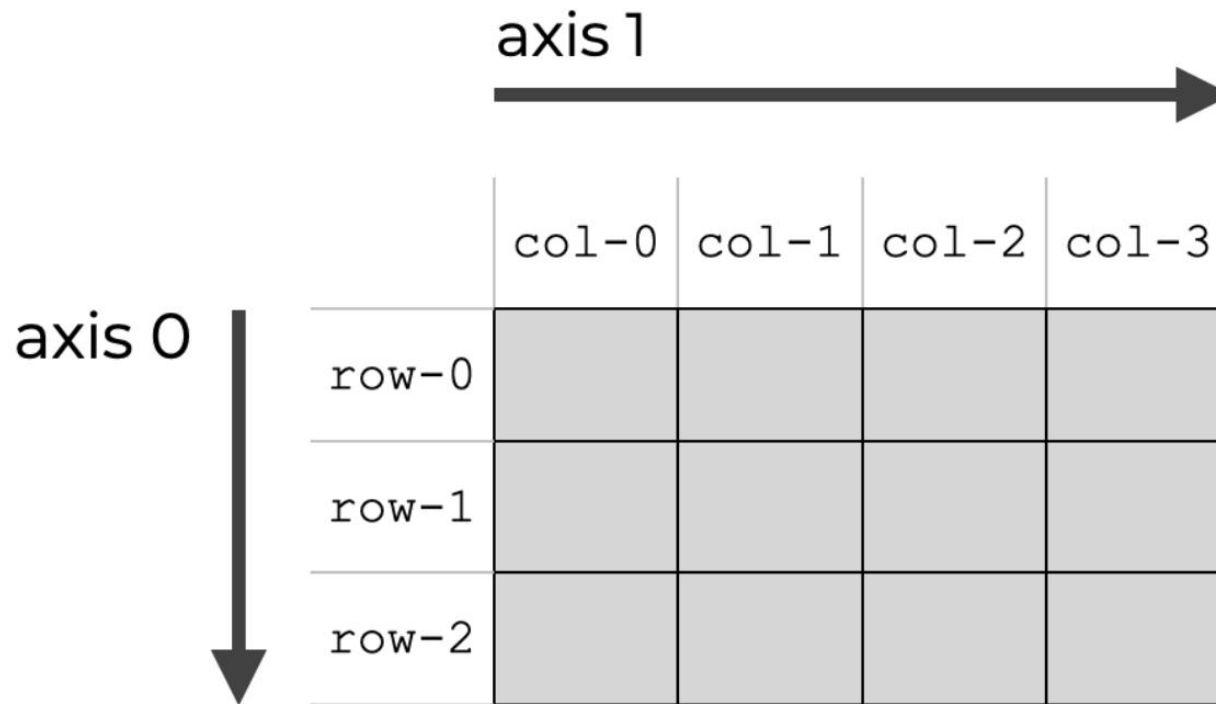
$$\begin{array}{c} (3,3) \\ \left[\begin{array}{ccc} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{array} \right] \end{array} + \begin{array}{c} (1,2) \\ [3,4] \end{array}$$

A: (3, 3)
B: (1, 2) Mismatch

Mathematical Functions

- `np.sum()`
- `np.mean()`
- `np.log()`
- `np.exp()` `# raise to power`
- `np.min()`
- `np.max()`

Axes in Numpy

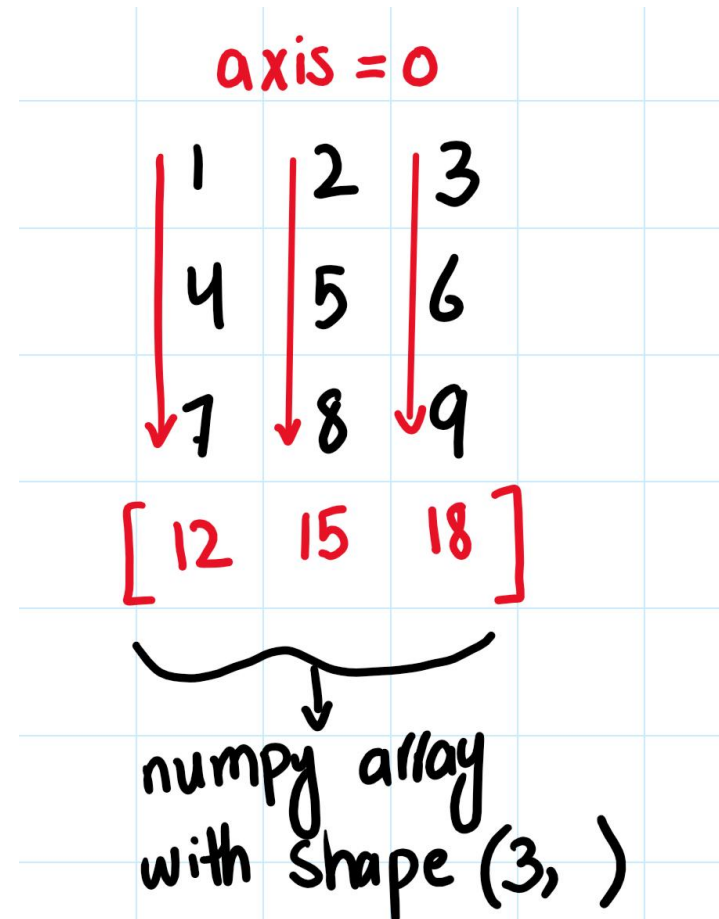


np.sum() with axes

np.sum(my_matrix)

1	2	3
4	5	6
7	8	9
sum 45		

`np.sum(mat , axis = 0)`



`np.sum(mat , axis = 1)`

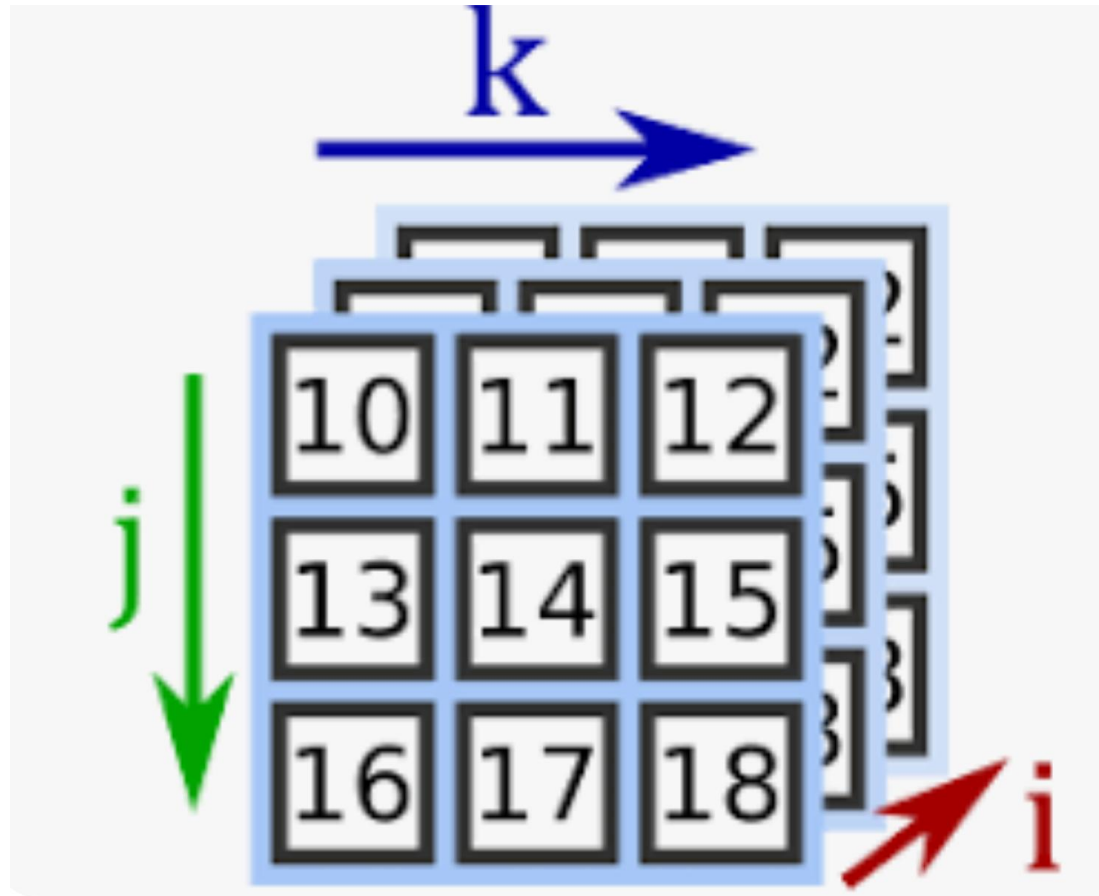
Similary , you can use :

`np.mean(mat , axis=?)`
`np.min(mat , axis=?)`
etc....

axis = 1				
1	2	3	→	6
4	5	6	→	15
7	8	9	→	24

shape
(3,)

Axis in 3D



$$\left[\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \right]$$

What would be the result of following :

`np.sum(cube , axis=0)`

`np.sum(cube , axis=1)`

`np.sum(cube , axis=2)`

Quiz !!!

Take out a sheet

Question 1 : How can we do slicing to get these elements ?

index	0	1	2	3	
0	1	2	3	4	
1	5	6	7	8	
2	9	10	11	12	
3	13	14	15	16	

Question 2

index	0	1	2	3
0	1	2	3	4
1	5	6	7	8
2	9	10	11	12
3	13	14	15	16

Question 3: BroadCasting Result ? .

Array 1 Shape	2, 3, 1	4, 5	3, 2, 1
Array 2 Shape	2, 1, 4	1, 2, 5	2, 2
Resultant Shape	?	?	?

Question 4 : Sum across an axis ?

```
arr = np.arange(8).reshape((2,2,2))  
print(arr)
```

$$\begin{bmatrix} \begin{bmatrix} 0 & 1 \end{bmatrix} & \begin{bmatrix} 4 & 5 \end{bmatrix} \\ \begin{bmatrix} 2 & 3 \end{bmatrix} & \begin{bmatrix} 6 & 7 \end{bmatrix} \end{bmatrix}$$

$\text{np.sum(arr, axis=0)} = ?$

$\text{np.sum(arr, axis=1)} = ?$

$\text{np.sum(arr, axis=2)} = ?$