

Note:-
To know the present working directory (pwd)

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To change directory cd (path)

VB:-
python for datascience

Numpy:-

It stands for Numerical python.

used for mathematical ops.

It can create upto 0d, 1d, 2d, 3d, ... nd arrays.

Ex:- a = 94 # 0D array.

```
temp = np.array(a)
print(type(temp)) # class <array>
print(temp.ndim) # 0
print(temp.shape) # ()
```

Ex:- a = [1, 2, 3, 4]

```
temp = np.array(a)
print(type(temp)) # class <array>
print(temp.ndim) # 1
print(temp.shape) # (4,) } no. of items in tuple is 1
                     ↓
                   no. of items
```

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Note:-

shape

- ① 0 dim array \rightarrow ()
- ② 1 dim array \rightarrow (no. of items,)
- ③ 2 dim array \rightarrow (no. of rows, no. of cols)

* functions for creating 1D array:-

① arange:-

\rightarrow arange() is a array-valued version of the built in python range function.

\rightarrow Syntax:- ~~np.array~~

np.arange(start, end, step)

Ex:-

a = np.arange(1, 10, 2)

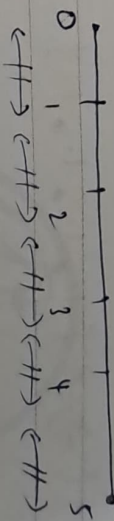
print(a) # [1 3 5 7 9]

② linspace:-

\rightarrow Returns evenly spaced numbers over a specified interval.

\rightarrow Syntax:- np.linspace(start, end, no-of-samples)

Ex:- np.linspace(0, 5, 6) # [0. 1. 2. 3. 4. 5.]



Note:- start, end both are inclusive.

③ zeros:-

\rightarrow Generates arrays of zeros.

\rightarrow Syntax:- np.zeros(shape)

Ex:- a = np.zeros((4, 10)) # (4, 10) \rightarrow 1D array with 14 items
print(a) # [0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0]

Note:- by default datatype = float, so if we want int \rightarrow

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$a = \text{np.zeros}(4, \text{dtype}=\text{int})$
 $\text{print}(a) \# [0 \ 0 \ 0 \ 0]$

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④ Ones:-

→ Here it all values filled with ones.

→ Syntax:- $\text{np.ones}(\text{shape})$

Ex:- $a = \text{np.ones}(4, 1)$
 $\text{print}(a) \# [1. \ 1. \ 1. \ 1.]$

⑤ random.randint:-

→ Return random integers from start (inclusive) and end (exclusive)

→ Syntax:-
 $\text{np.random.randint}(\text{start}, \text{end}, \text{no of samples})$

Ex:- $a = \text{np.random.randint}(1, 1000, 5)$
 $\text{print}(a) \# [294 \ 803 \ 670 \ 272 \ 901]$
↓
type # class < array >

* Accessing values from array:-

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Ex:- $l = [30, 40, 50, 60, 70]$
 $a = \text{np.array}(l)$
 $\text{print}(a) \# [30 \ 40 \ 50 \ 60 \ 70]$
 $\text{print}(a[0]) \# 30$
 $\text{print}(a[0:4:2]) \# [30 \ 50]$

Note:-

→ indexing and slicing operations are same as list and as well as same in '1D' array.

→ indexing and slicing can't be applied to '0D' array.

→ Accessing a multiple values based on condition (Mask indexing or Fancy indexing):-

Syntax:- $\text{array}[\text{condition}]$

→ first it will check the item which condition, the item which satisfies the condition that item will get selected.

Ex:- $a = \text{np.array}([45, 46, 47])$
 $\text{print}(a[a > 45]) \# [46 \ 47]$

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- insertion: `l = np.append(l, 900)`
- deletion: `l = np.delete(l, index)`
- updation.
- Arithmetic ops.

`del l,`
↓
To del from memory.

→ `np.array_equal(arr1, arr2)`

checks if both numpy arrays contains same el. or not.

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* 2D array: (list of 1D arrays)

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

so, 'arr' is a (2x3) array.

→ To fill a value as 'NaN' then -

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

So,
 $\begin{bmatrix} 1 & 2 & \text{nan} \\ 4 & \text{nan} & 5 \end{bmatrix}$

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→ To create a array of (3x3) with value '4' in every item then -

`arr = np.zeros(3, 3, dtype=int) + 4`
(or)

→ `np.ones(3, 3, dtype=int) + 3`

(diagonals will be '1')

→ To create identity matrix (generally it is a square matrix)

i.e. $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ (2x2), $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ (3x3)

So, to create above identity matrix -

`arr1 = np.eye(2, 2)`

`arr2 = np.eye(3, 3)`

→ Now to have our own values in diagonal then -

`arr = np.diag([1, 2, 4])` ⇒ $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix}$

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* indexing in 2D array:-

→ To extract a single value from 2D.

Syntax:-

arr_name[row_index, col_index]

Ex:- $\begin{bmatrix} 0 & 1 \\ 1 & 2 \\ 2 & 3 \\ 3 & 4 \end{bmatrix}$

to get '4' so position is (3, 1)

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

`print(arr[1, 1])` # 4

→ To extract a single row-

`print(arr[1])` # [3, 4]

→ To extract multiple rows-

`print(arr[0:3])` # row 0, 1, 2 returned.

→ To replace a value -

`a[2, 2] = 500`

→ To replace a entire row -

`a[1] = [4, 6]`

* flattening:- Converting ND array to 1D array

we use '.ravel()' method

i.e

Consider a 2D array

$\begin{bmatrix} 3 & 4 & 5 \\ 9 & 6 & 8 \\ 10 & 9 & 6 \end{bmatrix}$ $\xrightarrow[\text{into 1D}]{\text{to convert}}$ $[3, 4, 5, 9, 6, 8, 10, 9, 6]$

So, `arr1 = np.array([[3, 4, 5], [9, 6, 8], [10, 9, 6]])`

`arr2 = arr1 arr1 .ravel()`

`arr2 = arr1.ravel()`

`print(arr2)` # [3, 4, 5, 9, 6, 8, 10, 9, 6]

* To convert rows \Leftrightarrow cols & vice versa then it is called Transpose matrix.

i.e

$a \Rightarrow a.T$

$\begin{bmatrix} 3 & 4 & 5 \\ 9 & 6 & 8 \\ 10 & 9 & 6 \end{bmatrix} \xrightarrow{a_2} \begin{bmatrix} 3 & 9 & 10 \\ 4 & 6 & 9 \\ 5 & 8 & 6 \end{bmatrix}$

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* In 2D arrays, we have two ways of sorting-

- ① Row wise Sorting
- ② Column wise sorting.

Ex:- Consider a matrix 'A'

$$\begin{bmatrix} 4 & 3 & 1 \\ 9 & 6 & 4 \\ 10 & 12 & 1 \end{bmatrix} \xrightarrow{\text{Row wise sort}} \begin{bmatrix} 1 & 3 & 4 \\ 4 & 6 & 9 \\ 1 & 10 & 12 \end{bmatrix}$$

$$\begin{bmatrix} 4 & 3 & 1 \\ 9 & 6 & 4 \\ 10 & 12 & 1 \end{bmatrix} \xrightarrow{\text{col wise sort}} \begin{bmatrix} 4 & 3 & 1 \\ 9 & 6 & 1 \\ 10 & 12 & 4 \end{bmatrix}$$

So,

`arr1 = np.array([[4,3,1],[9,6,4],[10,12,1]])`

`arr1 = np.sort(arr1, axis=1) # Row wise sort`

`arr1 = np.sort(arr1, axis=0) # col wise sort.`

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* Arithmetic ops:-

→ Addition:-

`arr1 + arr2`

$$\begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} + \begin{bmatrix} 4 & 5 \\ 6 & 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 5 & 8 \\ 10 & 9 \end{bmatrix}$$

→ multiplication:- Two ways

- ① dot product
- ② matrix multiplication

$$\text{①:- } \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} \times \begin{bmatrix} 4 & 5 \\ 6 & 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 4 & 15 \\ 24 & 26 \end{bmatrix}$$

②:- `print(a * b)`

$$\text{②:- } \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} \times \begin{bmatrix} 4 & 5 \\ 6 & 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 \times 4 + 3 \times 6 & 1 \times 5 + 3 \times 4 \\ 4 \times 4 + 5 \times 6 & 4 \times 5 + 5 \times 4 \end{bmatrix}$$

`c = np.matmul(a, b)`

* Concatenation:-

We can concatenate two arrays in two ways-

- ① vertically
- ② horizontally.

① Vertically:-

$$\begin{array}{cc}
 A & B \\
 \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} & \begin{bmatrix} 6 & 7 \\ 9 & 8 \end{bmatrix} \\
 \Rightarrow & C \\
 & \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 6 & 7 \\ 9 & 8 \end{bmatrix}
 \end{array}$$

Ex: $\text{arr1} \Rightarrow \text{np.array}([[1,2], [3,4]])$

$\text{arr2} \Rightarrow \text{np.array}([[6,7], [9,8]])$

$\text{arr3} \Rightarrow \text{np.vstack}(\text{arr1}, \text{arr2})$

Vice versa

to do horizontal $[A][B] \Rightarrow [AB]$

do,

$\text{arr4} \Rightarrow \text{np.hstack}(\text{arr1}, \text{arr2})$