

DATE :

Q6 :- 22/07/15

deleting a variable:-

Using Keyword 'del'

a=55

del a

print()#error.

a(55)

↓

x→deleted.

del a,b,c #to del multiples variables.

Q7:-

id() → gives the address

type() → gives the type of object.

To create str → " (or) " (or) " "

To multiline str → " " (or) " " " " "

Ex: sc = " " pavan

addtly, " " "

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name="pavan"

age = 22.

print("my name is:",name,"age:",age);

print(f"my name is :{name} age is :{age}")

print("my name is:{} age is {}".format(name,age));

all three output is:-

my name is pavan age is: 22 ,

Note: int("12.67") X (.) is considered as char

float("12.67") ✓

Q8:-

Indexing and slicing.

Q9:-



DATE:  
 ex:-  
 del replace  
 ↗ overdriven just &  
 ↙ keep forward it does!

→ .insert(index, ele)

→ .clear() → will delete all items in a object. but object itself will not be removed.

→ del l → now same as clear but object will also get deleted.

→ .count(item) =) returns how many times the item got repeated.

→ .pop() → last item will be deleted.

.pop(i) → index 'i' item will be removed

→ .reverse()

ex:- l=[1, 1, [3, 4, 5]]

l.reverse() # [3, 4, 5], 2, 1]

but i want only nested element to be reversed

i.e. [1, 2, [5, 4, 3]]

So) l[-1].reverse();

(or)

l[2].reverse();

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ex: count('5') in list?

l=[1, 1, 1, 2, [2, 3, 4, 5, 5], 6, 6, 7, 8, 9, 6, 6, 3]

l.count()

l[4].count(5);

throws error if not found

→ .remove(item)

removes the specified item. only first occurring ele from left

→ .sort() # homogeneous type should only be allowed.

→ .sort(reverse=True) # desc order.

→ diff b/w .sort() & sorted(l)

.sort()

→ It is a method

→ Automatically stores in called variable

ex:- l=[1, 2, 4, 3]

l.sort()

print(l)

[1, 2, 3, 4]

sorted(l)

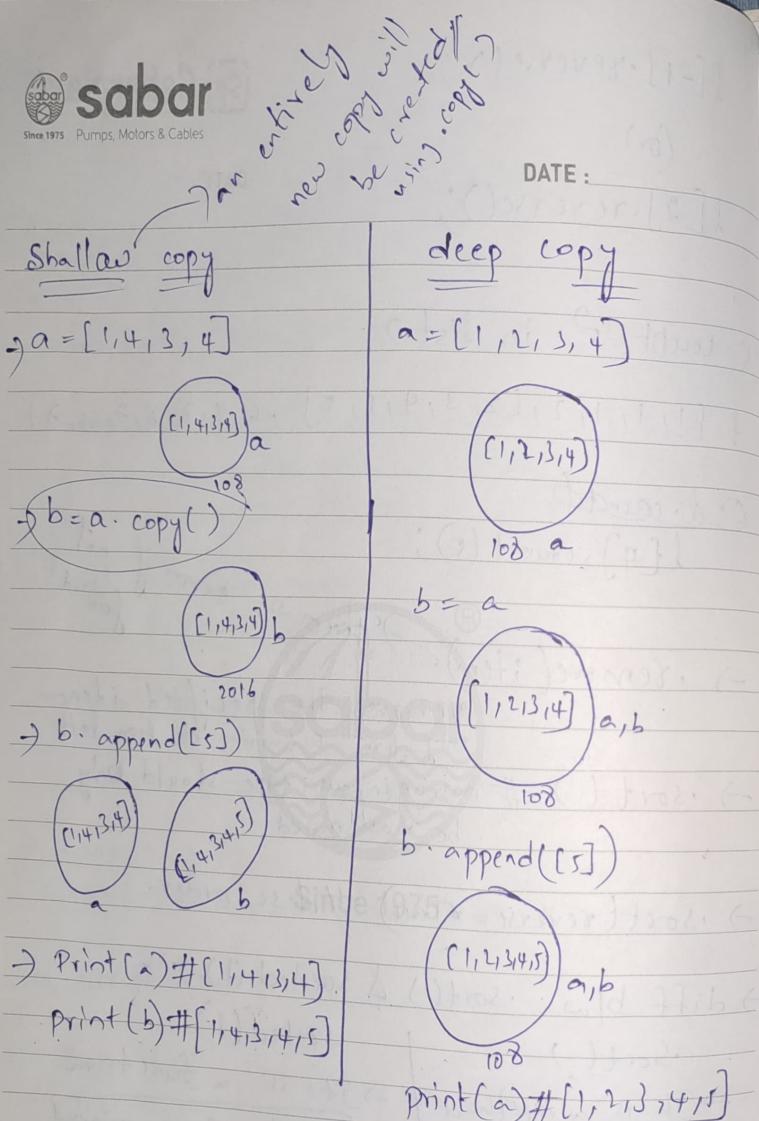
→ It is a function

→ Not stores in original val but we need to store in other val.

Sorted(l) → print(l) # [1, 2, 4]

So) temp=sorted(l)

↳ [1, 2, 3, 4].



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(V1)

Tuple:-

Same as list, but it is immutable.  
→ Using ()

Ex:-  $t = (1, 2, 3)$

Nested tuples:-

$t = (1, 2, 3, [4, 5], (6, 7), 8)$

$\text{print}(t[3]) \# [4, 5]$

So, it is since list type all the methods of list will be able to applied on it  
i.e.

$t[3].append(60) \# [4, 5, 60]$

$t[3].insert(0, 99) \# [0, 4, 5, 60]$

$\text{print}(t) \# (1, 2, 3, [0, 4, 5, 60], (6, 7), 8)$

but, it is applied on tuples (or) nested tuple

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concatenation:-

$$t_1 = (1, 2, 3)$$

$$t_2 = (4, 5, 6)$$

$$t_3 = t_1 + t_2$$

Note:-

In List →

i) `.extend()` will change the items in existing obj.

i.e

$$l_1 = (1, 2, 3, 4) \# id=234$$

$$l_1.extend([4, 5])$$

$$id(l_1) \# id=234$$

but

if `l_1 + [4, 5, 6]` will create entirely new obj of 'l\_1'

i.e

$$id(l_1) \# 234$$

$$l_1 = l_1 + [4, 5, 6]$$

$$id(l_1) \# 536$$

To convert a list to tuple → Use `tuple()` function

Ex:-  $l_1 = [1, 2, 3]$   
 $=$   
 $l_1 = \text{tuple}(l_1)$

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Vice versa from tuple to list

$$t_1 = (1, 2, 3)$$

$$l_1 = \text{list}(t_1)$$

$$\text{print(type}(l_1)) \# \text{list}$$

\* Interview Question:-

Q) Select a tuple of your choice, where 1 item should be a list, Now inside that list can i add 1 value of my choice?

A) Yes.

$$\text{Ex:- } \text{tup} = (1, 2, 3, [4, 5], "pavan")$$

$$= \text{tup}[3].insert(0, "adithya")$$

$$\text{print(tup)}$$

$$[1, 2, 3, ["adithya", 4, 5], "pavan"]$$

↑  
added

Q) Can we replace a value in a list item inside a tuple?

A) Yes.

$$\text{Ex:- } \text{tup}[3][0] = 400$$

$$\text{tup} = (1, 2, 3, [400, 5], "pavan")$$

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N12 -

Sets:-

Sets can be used to perform mathematical set operations like union, intersection, symmetric difference etc,

empty list  $\Rightarrow l = []$  (or)  $l = list()$   
tuple  $\Rightarrow t = ()$  (or)  $t = tuple()$   
set  $\Rightarrow s = set()$

$\rightarrow$  set is a unordered collection of item.  
 $\rightarrow$  duplicates are not allowed  
 $\rightarrow$  defined by {}.

$\rightarrow$  set can only store immutable items  
i.e it can store only - int, float,  
bool, str,  
tuple.

$\rightarrow$  Nested set is not allowed since  
set is ~~mutable~~.

$\rightarrow .add(item)$

adds single item to set if not present  
if already present it will not add.

$\rightarrow$  To add multiple item we use `.update()` method, but we need to add items inside that list:

i.e ex:-  $s_1 = \{1, 2, 3\}$

$s_1.update([4, 5, "true"])$

$print(s_1)$  # all items are seperately added  
 $\{1, 2, 3, 4, 5, "true"\}$

Simple:-

To add one item  $\Rightarrow$  use `.add()`  $\rightarrow$  as one item  
To multiple items  $\Rightarrow$  use `.update`  $\rightarrow$  as separate item.

i.e  $s_1 = \{1, 2, 3\}$

$s_1.add((40, 50))$  |  $s_1.update((40, 50))$

#  $\{1, 2, 3, (40, 50)\}$  | #  $\{1, 2, 3, \underline{40, 50}\}$

Note:-

$\Leftarrow$  Use tuple brackets only for you  
to avoid confusion.

DATE :

→ .discard(item)

removes specified item if available  
if not available nothing happens.

→ .remove(item)

if available well and good it will  
remove.

if not available it will throw  
error.

→ .clear() }  
→ del s } same concept as list

→ .copy() ⇒ to create shallow copy

Note:- In set 'replace' is not possible  
bcuz we don't know index  
right 😊.

Set operations:-

$$S_A = \{1, 2, 3\}$$

$$S_B = \{2, 3, 4\}$$

$$S_C = S_A \cup S_B$$

print(S\_C)

⇒ S\_A & S\_B remains same.

→ .intersection()  
S\_A.intersection(S\_B)

→ .isdisjoint()

S\_A.isdisjoint(S\_B) ↗ True (if no common ele)  
↘ False (if common ele)

→ S\_A - S\_B

S\_A do nunchi S\_B ele del chey.

i.e print(S\_A - S\_B) ⇒ {1}

print(S\_B - S\_A) ⇒ {4}

→ .symmetric\_difference() ↗ (A ∪ B - A ∩ B)

S\_A △ (S\_B) ⇒ {1, 4}

→ .issubset(.) → True  
i.e.  $S_A \cdot \text{issubset}(S_B)$

( $S_B$  contains all ele. of  $S_A$ )

then,

→ ~~subset~~ A → subset  
B → superset.

\* dictionaries:-

empty dict → dict() or {}

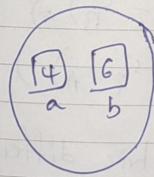
→ key value pair

Key : value

N<sub>13</sub> :-

$d_1 = \{ "a": 4, "b": 6 \}$

print( $d_1["a"]$ ) # 4



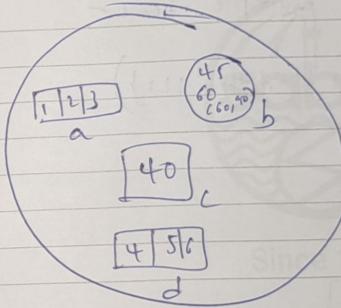
→ Keys can be of fundamental data types generally immutable  
→ Values can be of any data type.

Ex:-  $d = \{ "a": [1, 2, 3] \}$

$"b": \{ 45, 60, [60, 90] \}$

$"c": 40$

$"d": (4, 5, 6) \}$



to access (4, 5, 6)  
we do,  
 $\text{print}(d["d"])$

→ len(d) = 4.

→  $\text{print}(d["d"][1])$  # 5

like these also possible.

To get items (key-value pair)

→ .items() → returns list of tuple  
[( , )]

② To get keys → .keys() → returns list. [ ]  
To get values → .values() → returns list. [ ]

→ To add single (or) multiple items -

update({'pavan': 22, 'adithy': 22})

→ To delete a item.

~~del [key]~~

del d[keyname]

Ex:

= del d["pavan"]

(or)

d.pop("pavan")

→ To replace a value of a key.

Ex: ~~d = {"Hindi": 65, "pavan": 72}~~

Now to replace ~~72~~ with 22 →

i) ~~d["pavan"] = 22~~

(but it will give key error  
if 'key' not present)

ii) ~~d.get("pavan") = 22.~~

→ To retrieve a value of respective key.

Ex: ~~d = {"maths": 90, "Hindi": 65}~~

Now to retrieve "maths" value -

i) print(d["maths"])

(but it may give  
error if 'key' is not  
present)

ii) ~~print(d.get("maths"))~~

('None' will be returned  
if 'key' is not  
present)

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→ To replace a key in dict.

```
d = {"mathses": 90,
      "Hindi": 65}
```

to replace "mathses" to "maths"

```
d["maths"] = d.pop("mathses");
print(d) # {"maths": 90, "Hindi": 65}
```

Ques:-

String methods

Note: String is immutable

→ .upper()

will make all alphabets to uppercase and does nothing to special char and num.

```
Ex: S = "True...12"
S = S.upper()
print(S) # TRUE...12
```

→ s.lower()

→ s.swapcase() # up → low  
low → up

→ s.capitalize() # (entire string first word letter will be capital)

→ s.title() # each word first word will be cap

→ s.strip()

→ s.replace(from, to) → replaces all, no need of loop

→ l = s.split() # split and returns list.  
based on space as delimiter.

→ list to str

i)  $s_1 + " " + s_2 + " " + \dots$

ii) " ".join(listname)

→ .index("item") # returns starting index number.

if not found → error.

→ .find("item") # same as index but if not found error will not be raised but -1 is returned.

DATE:

- .count("item")
- .startswith("str") → True  
→ False
- .endswith("str") → True  
→ False
- .isalnum() → True  
→ False.
- .isdigit()
- .isalpha()
- .islower()
- .isupper()
- .isspace() # only spaces?
- .istitle()
- ~~.lstrip()~~

5 + 18 - 0.9  
29.0

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Interview Q)  $s = \text{"Hello"}$   
 $s[0] = "g"$

- whether it will be replaced?  
 → ~~Yes, it can be replaced but entirely new obj.~~  
 NO, since strings are immutable.

~~(V15)~~

→ will always give float

// → (trunc) part is returned of quotient.

BODMAS  
 $\downarrow$   
 $\downarrow$   
 exponent

( )	high priority ex: $5 + 5 * 5 - 5 / 5 \Rightarrow 29.0$
**	$= 5 + 5 * 5 - 0.0$
//	$= 5 + 25 - 0.0$
*, /, %	$= 30 - 0.0$
+,-	$\Rightarrow 30.0$

$$\begin{aligned} & \text{Ex: } 5 + 2^2 * 9 - 4 // 6 * 2 / 9 \\ & = \\ & \Rightarrow 23.0 \quad (\text{not } 23X) \end{aligned}$$

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Note:-

DATE:

→ To know the present working directory (pwd)

→ To change directory cd (path)

V23 :-

\*python for data science

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 = ~~a~~ np.array(a)

print(type(temp)) # class arrays

print(temp.ndim) # 0

print(temp.shape) # ()

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

= temp = np.array(a)

print(type(atemp)) # (1ss array)

print(temp.ndim) # 1

print(temp.shape) # (4,) } no. of items in  
↓ tuple is  
no. of items

Note:-

shape

DATE:

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)

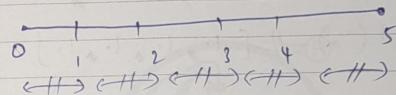
Ex:-  
 $a = np.arange(1, 10, 2)$   
print(a) # [1, 3, 5, 7, 9]

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) # [0.0, 1.0, 2.0, 3.0, 4.0]



Note:- start, end both are inclusive.

③ zeros()-

$\rightarrow$  Generates arrays of zeroes.

$\rightarrow$  Syntax:- np.zeros(shape)

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

Note:- by default datatype = float, so

if we want int ->

```
a = np.zeros((4,), dtype=int)
print(a) # [0 0 0 0]
```

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#### ④ Ones:-

- Here it all values filled with Ones.
- Syntax:- `np.ones(shape)`

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

#### ⑤ random.randint:-

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

Syntax:-  
 $= np.random.randint(start, end, nof samples)$

Ex:-  
 $a = np.random.randint(1, 1000, 5)$   
 $= print(a) # [294 803 670 292 901]$   
 ↓  
 type# class<array>

#### \*Accessing Values from array:-

Ex:-  
 $I = [30, 40, 50, 60, 70]$   
 $a = np.array(I)$

$print(a) # [30 40 50 60 70]$   
 $print(a[0]) # 30$   
 $print(a[0:4:2]) # [30 50]$

#### 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:- `array[condition]`

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

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

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- insertion:  $\text{arr} = \text{np.append}(\text{arr}, 900)$  → np.where( $\text{arr}$ )
- deletion:  $\text{arr} = \text{np.delete}(\text{arr}, \text{index})$  → del  $\text{arr}$ ,  
↓  
To del from  
Memory.
- updation:
- arithmetic op.

→  $\text{np.array_equal}(\text{arr1}, \text{arr2})$

checks if both numpy arrays contains same ele or not

Q24  
→ 2D array: (list of 1D arrays)  $\Rightarrow$

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

so, arr is a  $(2 \times 3)$  array.

→ To fill a value as 'NaN' then -

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

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

→ To create a array of  $(3 \times 3)$  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.  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

do, 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])  $\Rightarrow$   $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix}$

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\* indexing in 2D array -

→ To extract a single value from 2D

Syntax:-

= arr\_name[row\_index, col\_index]

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

to get '4' so position is (1, 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 -

$$arr[2, 2] = 500$$

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→ To replace a entire row -

$$arr[1] = [4, 6]$$

\* flattening: Converting ND array to 1D array

We use 'ravel()' method

i.e.

Consider a 2D array

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

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

arr1 ~~2D array~~ → 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

$$a \Rightarrow a.T$$

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

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

① Row wise sorting

② column wise sorting.

Ex:- Consider a matrix 'A'

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

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

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) # colwise sort

\* Arithmetic ops

\* Addition:-

= arr1 + arr2

$$\begin{bmatrix} 1 & 3 \\ 4 & 6 \\ 10 & 12 \end{bmatrix} + \begin{bmatrix} 4 & 5 \\ 6 & 4 \\ 12 & 1 \end{bmatrix} \Rightarrow \begin{bmatrix} 5 & 8 \\ 10 & 9 \end{bmatrix}$$

\* multiplication:- Two ways

① dot product

② matrix multiplication

$$\text{①: } \begin{bmatrix} 1 & 3 \\ 4 & 6 \end{bmatrix} \times \begin{bmatrix} 4 & 5 \\ 6 & 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 4 & 15 \\ 24 & 20 \end{bmatrix}$$

②: print(a \* b)

$$\text{②: } \begin{bmatrix} 1 & 3 \\ 4 & 6 \end{bmatrix} \times \begin{bmatrix} 4 & 5 \\ 6 & 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 \times 4 + 3 \times 6 & 1 \times 5 + 3 \times 4 \\ 4 \times 4 + 6 \times 6 & 4 \times 5 + 6 \times 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 22 & 17 \\ 46 & 40 \end{bmatrix}$$

c = np.matmul(a, b)

## \* Concatenation:-

We can concatenate two arrays in two ways.

① vertically.

② horizontally.

① Vertically:-

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

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

Ex:- arr1 = np.array([[1, 2], [3, 4]])

arr2 = np.array([[6, 7], [9, 8]])

arr3 = np.vstack((arr1, arr2))

Vice versa

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

do:- arr4 = np.hstack((arr1, arr2))

Pandas

Series

dataframe.

Pandas Series:-

→ It is 1D array holding data of any type.

→ A pandas series is like a column in a table.

Ex:- import numpy as np  
import pandas as pd

l = [10, 20, 30]

arr = np.array(l) # [10 20 30]

arr1 = pd.Series(l) # [0 10  
1 20  
2 30]

→ A series can be created Using '3' ways:-

- ① list
- ② 1D array
- ③ Dict

Row name  
(by)  
DATE:  
index names.

### ① list:-

$l = [10, 20, 30]$   
 $\text{arr} \Rightarrow \text{pd.Series}(l) \# \begin{bmatrix} 0 & 10 \\ 1 & 20 \\ 2 & 30 \end{bmatrix}$

index names:- by default it will be index values.

Ex:-  $\text{arr} = \text{pd.Series}(l, \text{index}=[\text{"a"}, \text{"b"}, \text{"c"}])$

$$\begin{bmatrix} \text{a} & 10 \\ \text{b} & 20 \\ \text{c} & 30 \end{bmatrix}$$

### ② Using 1D array:-

Same as above.

### ③ Using dict:-

$d \Rightarrow \{'a': 10, 'b': 20, 'c': 30, 40\}$

$\text{arr} \Rightarrow \text{pd.Series}(d)$

$$\begin{bmatrix} \text{a} & 10 \\ \text{b} & 20 \\ \text{c} & 30 \\ 40 & 40 \end{bmatrix}$$

Note:- Key values of dictionary  
will be the index names  
when we use dictionary.  
and index name should be immutable.

DATE:  
and index name should be immutable.

xDataframe:-

It is a combination of multiple series.

### ④ difference b/w Series, DataFrame

Series

$l = [10, 20, 30]$

$\rightarrow \text{pd.Series}(l)$

$$\begin{bmatrix} 0 & 10 \\ 1 & 20 \\ 2 & 30 \end{bmatrix}$$

Dataframe

$l = \{'a': 10, 'b': 20, 'c': 30\}$

$\rightarrow \text{pd.DataFrame}(l)$

$$\begin{bmatrix} 0 & a & b & c \\ 1 & 10 & 20 & 30 \end{bmatrix}$$

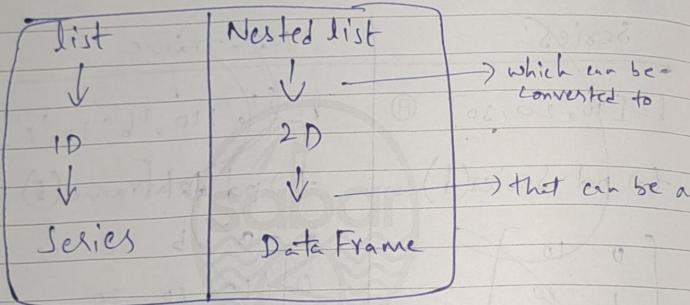
Here

Ex:-  $d = \{'a': [10, 20], 'b': [30, 40], 'c': [10, 30, 40]\}$   
 $\rightarrow \text{pd.DataFrame}(d)$

Row	a	b	c
0	10	30	10
1	20	40	30
2	-	-	40

Note:-

- 1) A normal list can't be a DataFrame.
- 2) A normal list can be a series.
- 3) A ~~skewed~~ list which can be converted to 2D array those list can be a DataFrame.



→ DataFrame can be created using two ways

① Using 2D arrays

② Using dict type.

DATE:

Q1  
Ex: DF

= l = np.array([ [10, 20, 30],

[40, 50, 60],

[60, 30, 40] ])

Q2  
pri

d = pd.DataFrame(l, index=['a', 'b', 'c'], columns=['i', 'j', 'k'])

print(d)

	i	j	k
a	10	20	30
b	40	50	60
c	60	30	40

Row names

Q3  
Using dict :- (Row names will become column names only in dict)

di = {'a': [10, 20, 30], 'b': [40, 50, 60], 'c': [60, 30, 40]}

df = pd.DataFrame(di)

print(df)

	a	b	c
0	10	40	60
1	20	50	30
2	30	60	40

Observe

DATE:

No = 1). \* How to add new column in a data frame

Ex:-  $d \Rightarrow \{ "Age": [10, 20, 30, 40, 50], "Gender": ["M", "F", "M", "F", "M"] \}$

Now i want to add 'S.no' column.  
So,

$\text{df} \Rightarrow \text{pd.DataFrame}(d)$

$\text{df}[["S.No"]] = [a, b, c, d, e]$

print(df)

S.No	Age	Gender
a	10	M
b	20	F
c	30	M
d	40	F
e	50	M

To set index as "S.No"

$\text{df} = \text{df.set_index("S.No")}$

To retrieve the data we have two ways

① loc

② iloc

↳ Based on rownames, colnames

Syntax:  $\text{df}[\text{rownames}, \text{colnames}]$  # To retrieve single row, single column

Ex:-  $\text{df}[a, "Age"] \# 10$

To retrieve multiple rows/columns

$\text{df}[["a", "c", "e"], ["Gender"]]$

S.No Gender

a M

c M

e M

→ To retrieve all rows of a single column

$\text{df.loc[:, "Age"]}$

S.No	Age
a	10
b	20
c	30
d	40
e	50

② iloc - Based on index numbers.

	0	1	2	col index
	Age	Sal	Gender	number
-5	0	22	17000	M
-4	1	23	18000	F
-3	2	29	19000	M
-2	3	30	40000	M
-1	4	35	60000	F

row index

numbers

index names

Syntax:- df.iloc[row index number,  
column index num]

→ To retrieve all the rows of 'Age' col,

df.iloc[:, 0]

(or)

df.iloc[:, -3]

→ To retrieve 1st and last row of 'Age',  
gender column -

df.iloc[0, 4], 0, 2]

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→ Reset index:-

df.reset\_index()

→ How to drop a row! -

① Single row(or)multiple

Syntax:- df.drop(index=[1, 2, 3])

→ How to drop column! -

df.drop(columns=[1, 2, 3])

→ Sorting!

Sorting is done by considering any particular column.

`df.sort_values(by = "Age") # A ESC`

`df.sort_values(by = "Age") //`

`df.sort_values(by = "Age", ascending = False) # DFG`

→ Combining multiple dataframes-

df1	city	temp
0	mumbai	32
1	delhi	45
2	benglore	40
3	hyd	36

df2	city	humidity
0	delhi	68
1	mumbai	65
2	benglore	75
3	Chennai	70

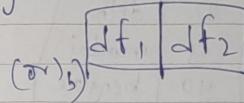
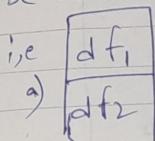
① there are two different ways to combine dataframes-

(1) Concat

(2) merge.

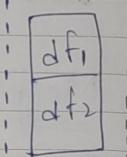
① Concat:

Two dataframes will be concatenated no matter the duplicates, it will be added again.



a) `pd.concat([df1, df2], axis=0)`

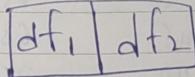
	city	temp	humidity
0	Mumbai	32	NAN
1	delhi	45	NAN
2	benglore	40	NAN
3	hyd	36	NAN
0	delhi	68	68
1	mumbai	NAN	65
2	benglore	NAN	75
3	Chennai	NAN	70



To make (0, 1, 2, 3, 4, 5, 6, 7, 8, ...)

`pd.concat([df1, df2], axis=0, ignore_index=True)`

and, To make



pd.concat([df1, df2], axis=1)				
	city	temp	city	humidity
0	mumbai	32	delhi	68
1	bengaluru	45	mumbai	65
2	bengalore	40	bengalore	75
3	hyd	36	chennai	70

## ② Merge:-

To merge two dataframes

As we can see if we consider city column we can observe 'mumbai', 'bengalore' repeated two times since they are same places.  
So, to avoid it we use merge.

pd.merge(df1, df2, on="city", how="left")

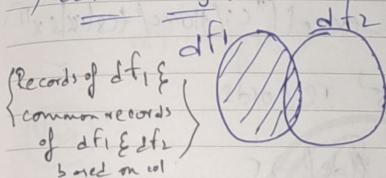
	city	temp	humidity
0	mumbai	32	65
1	delhi	45	68
2	bengalore	40	75
3	hyd	36	NAN

how → will take four Arg  
they are how = { inner (default)  
outer  
left  
right }

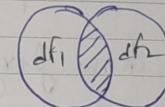
DATE:

Note:- SQL Joins = merge (pandas).

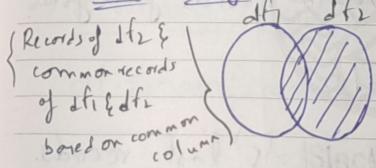
i) Left merge:-



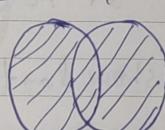
iii) inner:- {Common Records in both df1 & df2 based on col}



ii) Right merge:-



iv) outer:- {All the records}



Use → if no column is common then no merge is possible  
merge => when you want to combine two df based on common column.

concat => when you want to attach two df no matter the common columns.

## \* Importing datasets

① `pd.read_excel("path")`

path should be H (or) /

Ex: `df = pd.read_excel("C:/Home/test.xlsx")`

→ To show first 5 records.

`df.head()` // df[5]

`df.head(n=10)` # 10 records from first

→ To show last records.

`df.tail()` // df[-5:] last records.

`df.tail(n=10)` # last 10 records.

→ To know the dimension of df →

`df.shape`

Ex: (4,3)

→ To know total no. of values.

`df.size`

Ex:  $4 \times 3 \Rightarrow 12$  values.

DATE:

→ To know the column names -

`df.columns`

(or)

`df.keys()`

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→ To rename column names:-

`df.rename(columns={ 'AGE': "AGE",  
'sum': "SUM" })`

→ data types of each column:-

`df.dtypes`

object (if one value is str  
in column it is)  
float (if one float is present)  
int (if integer)

```

    if(column has at least one string)
    {
        output = object;
    }
    else if(column has at least one float)
    {
        output = float;
    }
    else
    {
        output = int;
    }

priority maximum float
priority minimum int

```

df.info()

shape + dtype + columns

df.isnull().sum()

DATE:

int

How to convert pandas series to 1D array

df

df['column'].values

(or)

df[" "].to\_numpy()

\* if it is a series then it will be converted to '1D' array.

\* if it is dataframe then it will be converted to '2D' array.

and to convert 2D → 1D we use .ravel().

ie  
df[" "].to\_numpy().ravel()

## \*filtering:-

→ Selecting/extracting the data based on condition.

Ex:- Consider a dataframe:

	X	Y
0	9	5
1	3	6
2	4	6
3	5	2
4	7	4
5	2	2

→ To get only records which are  $X \geq 3$

so,  $df[df['X'] \geq 3]$

→ To get only record where  $X \geq 3$  &  $Y \leq 4$  and

so,  $df[(df['X'] \geq 3) \& (df['Y'] \leq 4)]$

→ To get records where  $X == 3$

so,  $df[df['X'] == 3]$

— X —

→ To get what categories of data present in a column.

$df['Gender'].unique()$

#(M, F)

	Gender
0	M
1	M
2	F
3	M
4	F

→ To get how many count that each category has

$df['Gender'].value_counts()$

→ To get total how records were

DATE: \_\_\_\_\_

## \*Discretization:-

→ converting numerical data into non numerical data.

	Marks
Ex: 0	70
1	60
2	50
3	90
4	95
5	72

Now marks  $< 70$  are 'fail' to be marked  
 marks  $> 90$  are 'excellent'  
 $70 < 90$  are 'moderate'

So,  
 $\text{df}["status"] = \text{pd.cut}(\text{df}["marks"],$

Output:-

	Marks	Status
0	70	fail
1	60	fail
2	50	fail
3	90	moderate
4	95	excellent
5	72	moderate

$\text{bins} = [0, 70, 90, 100]$   
 $\text{labels} = ["fail", "moderate", "excellent"]$

\*crosstab:-

→ applied only on '2' discrete categorical variables.

Ex:-  $\text{pd.crosstab}(\text{df}['time'], \text{df}['gender'])$

time \ gender	female	male
time	female	male
Dinner	52	124
lunch	35	33

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## \*Statistics\*

\*Analytics:- Complete analysis of data.

Types of Analytics:-

- ① Descriptive Analytics:- Describe the data. (Statistics)
- ② Diagnostic Analytics:- Check the reason behind that.
- ③ Predictive Analytics:- Predict (ML & AI)
- ④ Prescriptive Analytics:- What we have to do.