

Section A: Theory

1. What is a NumPy array, and how is it different from a Python list?

Q.1] What is NumPy Array & how is it different from Python list?

NumPy - It stands for numerical python. It is a powerful python library which is used to perform scientific computing & perform numerical computing task.

- It provides efficient multidimensional array object.

- It can store elements of same data type.

Diff	NumPy Array	List
Data Type	① Homogeneous (all elements must be of the same type)	② Heterogeneous (all elements can be of different type)
mem efficiency	③ box of fixed size & type memory efficiency is more	③ It has less memory efficiency box of dynamic size
Performance	③ It is faster than list due to optimized C implementation	④ Slower, especially for large data
Functionality	④ It supports advanced mathematical functions	④ Does not have built-in functions for elementary operations.

3. What are ufuncs in NumPy? Provide an example.

What are ufunc in Numpy?
Provide example.

- Stands for universal function.
- It perform element-wise opn on arrays
- It allows you to apply opn on entire arrays without using explicit loops

Types of ufunction

- ① Unary function → operate on single array (e.g. square, square root) e.g. `np.sqrt(arr)`
- ② Binary function → operate on two array element-wise (e.g. addition, sub, mul)
- ③ Other ufuncs → more complex fun that might involve conditionals or other advance logic (like user defined their own ufunc with help of `np.frompyfunc`)

Note: you can check type of fun which is ufunc or not
e.g. `type(np.add)` → `ufunc`

```
x=[1,2,34,4]
y=[10,10,20,10]
print(np.add(x,y))
print(type(np.add))
print(np.sqrt(x))
print(type(np.sqrt))
print(np.less(x,y))
```

```
def each_add_10(x,y):
    return x+y+10
arr=np.frompyfunc(each_add_10,2,1)
print(arr(x,y))
```

eg:-

```
x = [1, 2, 34, 4]
y = [10, 10, 20, 10]
print(np.add(x, y))
print(type(np.add))
o/p. [11, 12, 54, 14]
<class numpy.ufunc>
```

eg:- mathematical function

- ① `np.abs()`
- ② `np.log()`
- ③ `np.add()`
- ④ `np.subtract()`
- ⑤ `np.divide()`
- ⑥ `np.sin()` etc

Statistical functions

- ① `np.mean()`
- ② `np.median()`
- ③ `np.percentile()` etc

Comparison fun

- ① `np.equal()`
- ② `np.less()`

You can define your own ufunc by using `np.frompyfunc()`
It take no. of i/p arguments & no. of output arguments as parameter.

eg:-

```
def each_add_10(x, y):
    return x+y+10
arr = np.frompyfunc(each_add_10, 2, 1)
```

2nd name no. of arrays in i/p no. of arrays output

2. Explain the concept of broadcasting in NumPy with an example.

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.

Broadcasting Rules :-

- ① Make the two arrays have the same number of dimensions - If the no. of dimensions of the two arrays are different, add new dimensions with size 1 to the head of the array within the smaller ~~arr~~ dimension

eg - 1

arr 1 arr 2 →

(3, 2) (3,) (↓, 3)

20 10 head

added sets small


diminut

eq - 2

$$(3, 3, 3) - (9) = (1, 1, 3)$$

② Make each dimension of the two arrays the same size

If sizes of each dimension of the two array do not match, dimensions with size 1 are stretched to the size of the other array.

$(3, 3)$ (3) $(1, 3)$

 not same

 $(3, 3)$ $(2, 3)$

If there is a dimension whose size is not 1 or n either of the two arrays, it cannot be broadcasted, and an error is raised.

eq 2 - arr1 = $\begin{bmatrix} 0 & 1 & 2 \end{bmatrix}$
 $\& \begin{bmatrix} 3 & 4 & 5 \end{bmatrix}$
 $\begin{bmatrix} 9 & 10 & 11 \end{bmatrix}$
 arr2 = $\begin{bmatrix} 0 & 1 & 2 \end{bmatrix}$

$[4, 3]$

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

$\rightarrow (4, 3)$

$(4, 3)$ $(4, 3)$
 opp $[[0 \ 2 \ 4]$
 $[3 \ 5 \ 7]$
 $[6 \ 8 \ 10]$
 $[9 \ 11 \ 13]$

eg 82 arr 1 shape (3, 4)
 arr 2 shape (3,)

Step 1 $(3, 4) \rightarrow (3, 3)$
Step 2 Stretch 1 $\rightarrow (3, 3)$

$$(3, 4) \times (3, 3)$$

It shows Broadcasting is not possible
Broadcasting error

eg 2 $\begin{pmatrix} 0 & -a_1 \\ a_2 & 0 \end{pmatrix}$ $(1, 3)$ $a_2 (3, 1)$

Step 1 $(1, 3)$ $(3, 1)$
Step 2 stretch 1 by same dimension
 $(3, 3)$ $(3, 3)$

e438- a1 $\begin{bmatrix} 0, 1, 2 \\ (1, 3) \end{bmatrix}$ a2 $\begin{bmatrix} 10 \\ 23 \\ 37 \end{bmatrix} = (3, 1)$

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

e4 3 - (1, 3) (4, 1)
 (4, 3) (4, 3)

ex: $(2, 2) \quad (1) \rightarrow (1, 1)$
Stretch $(2, 2)$

4. What is the purpose of the numpy.random module? Name three functions it provides.

4] what is purpose of the numpy.random module?
 - It provides funns for generate random numbers.
 - also it can provide random generated array from range.
 - It also enables of data shuffly.

1] `np.random.rand()` → Generate random number between 0 — 1
 eg: `np.random.rand()` → 0.8756
 eg:-
`np.random.rand(2, 2)` → `[[0.081, 0.57], [0.61, 0.03]]`
 dimensions

2] `np.random.randint()` — Generate random integers from specified range (inclusive of low, exclusive of high) by default start from 0
 from numpy, import random as rd.
 eg 1 → `rd.randint(1, 10)` → 5
`rd.randint(1, 10, (2, 2))` → `[[3, 2], [5, 8]]`
 dimension

5] **Shuffle** — it shuffles the original array
 return, None
`arr = [10, 20, 30, 40]`
`rd.shuffle(arr)`
`print(arr)` → `[30, 20, 10, 40]`

11] **Seed function** → we know that, when we generate random number or array each time when we executing random number get changed. But when if we want to fix the generation of this random number. in that case we use seed, this generates fix set of random numbers.

eg:-
`rd.seed(12)` → it can be any number
`rd.rand()`
 o/p: 0.15 — 1st execution
 0.15 — 2nd execution.

3] `np.random.randn()` — Generates random numbers from a standard normal distribution (mean = 0, std = 1) — by default
 eg
`np.random.randn(2, 2)`
 o/p `[[0.4, 0.8], [-0.66, -1.63]]`
 dimension

4] `np.random.choice(a, size = None, replace = True)`
 - Generates a random sample from a given 1-D array. You can specify sample size and whether sample is with or without duplicate Or replacement.
 replace — by default is True means duplicate element.
 false — unique.
 if you given size is bigger than choice array and replace is false you will get an error.
 note — If your array having repeated elements and you give replace = false then also you can get duplicate elements of that array.
 eg: `rd.choice([1, 2, 3, 4, 5])`
 o/p 4

6] `random.permutation(x)` — it return random shuffled / permutation of x rather than changing entire original array (x)

7] 8] 9] `rd.randn()` — (loc = 0.0, scale = 0.0, size) generates random numbers from normal distribution with given mean (loc) and standard deviation (scale)
`rd.binomial(n, p, size = None)`
 - Draws samples from binomial distribution, where n is the number of trials & p is the probability of success

`rd.poisson(lam = 1.0, size = None)`
 Generates random samples from a poisson distribution with rate lam

10] `Uniform(low = 0.0, high = 1.0, size = None)`
 - Generates random numbers from uniform distribution over [low, high]. — each element in range has equally chances for being selected. It returns random number in float

```
from numpy import random as rd
```

```
# rand
```

```
# print(rd.rand(2,2))
```

```
# [[0.08217302 0.5734172 ]
```

```
# [0.61766962 0.03358355]]
```

```
#randint
```

```
# print(rd.randint(1,10)) ->5
```

```
# print(rd.randint(1,10,(2,2)))
```

```
# [[3 2]
```

```
# [5 8]]
```

```
# print(rd.randint(1,10,size=5))
```

```
# [3 4 1 2 1]
```

```
# randn
```

```
# print(rd.randn(2,2))
```

```
# [[ 2.24035776  1.23595522]
```

```
# [-0.39494653 -0.6499075 ]]
```

```
# choice
```

```
# print(rd.choice([1,2,3,4]))#2
```

```
print(rd.choice([1,1,4,5,6],size=6))#[6 1 6 5 1 1]
```

```
print(rd.choice([1,2,3,4],size=3,replace=False))#[2 1 4]
```

```
print(rd.choice([1,2,1,3,4],size=4,replace=False))#[2 4 1 1]
```

```
# print(rd.choice([1,2,3,4,1],size=6,replace=False))#Error
```

```
# Shuffle
```

```
arr=[10,20,30,40]
```

```
rd.shuffle(arr)
```

```
print(arr)
```

```
arr2=np.array([[10,20,30],[11,22,33],[1,2,3]])
```

```
rd.shuffle(arr2)
```

```
print(arr2)
```

```
# Permutation
```

```
print(rd.permutation([1,2,3,7,8,9,0]))
```

```

#uniform
print("Uniform",rd.uniform(1,100))

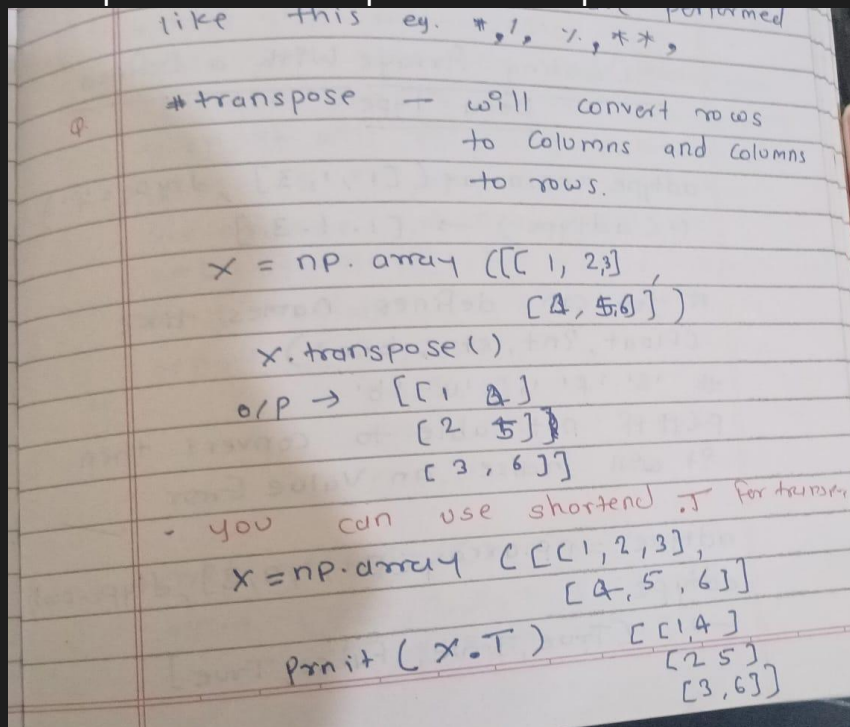
#normal
print(rd.normal(0,1,(2,2)))

#binomial
print(rd.binomial(10,0.5,3))

#seed
rd.seed(12)#it can be any number
print(rd.rand())

```

5. Explain how to compute the transpose of a matrix in NumPy.



```

arr=np.array([[1,2,3],[4,5,6]])
print(arr)
trans=np.transpose(arr)
print(trans)
witht=trans.T
print(witht)
# output:
# [[1 2 3]

```


[4 5 6]]
 # [[1 4]
 # [2 5]
 ## [3 6]]
 # [[1 2 3]
 # [4 5 6]]

6. Define the inverse of a matrix. When does a matrix not have an inverse?

6) Inverse of Matrix \rightarrow Inverse of matrix denoted by A^{-1} (A) matrix should be square matrix

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$A^{-1} = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

change of sign
inter change

Inverse of
d is $A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

Rules for inverse

- ① Only square matrix can have an inverse (a matrix has same no. of rows & columns).
- ② The determinant is non-zero

$$\det(A) = ad - bc$$

$$\det(A) \neq 0$$

eg-① $A = \begin{bmatrix} 4 & 7 \\ 2 & 6 \end{bmatrix}$ - is square matrix.

step ① calculate the determinant
 $(6 \times 4) - (7 \times 2) = 10$

step ② Inverse the compute

$$A^{-1} = \frac{1}{10} \begin{bmatrix} 6 & -7 \\ -2 & 4 \end{bmatrix}$$

$$A^{-1} = \begin{bmatrix} 0.6 & -0.7 \\ -0.2 & 0.4 \end{bmatrix}$$

code

```
arr = [[4,7], [2,6]]
if np.linalg.det(arr):
    P(np.linalg.det(arr))
    P(np.linalg.inv(arr))
```

O/P 10.000
 $\begin{bmatrix} 0.6 & -0.7 \\ -0.2 & 0.4 \end{bmatrix}$

7. What are the primary features of a pandas Series?

Q. 7] what is the primary features of a pandas Series?

Series - is a 1D Array that can hold homogeneous type elements, like int, float, string.

- A Series is essentially a labeled
- All elements in Series are of the same type you can have mixed-type - but it is not recommended.
- It support vectorization (element-wise opⁿ),
- you can access series with label-based.
- elements in series are associated with index that labels the data.

8. How does a pandas Series differ from a NumPy array?

Q. 8] How does pandas Series Different from numpy array.

Series

Array numpy

- | | |
|---|--|
| ① In Series is a label based indexing. | ① In Array Integer-based indexing. |
| ② It can be store mixed type data. | ② It is homogeneous data. |
| ③ In Series you can handle missing data like nan with fun ⁿ like <code>isna()</code> <code>fillna()</code> | ③ Array don't have built-in support for missing data |
| ④ In Series has metadata such as <code>Custom "name"</code>
e.g. <code>Series([1,2], name="Num")</code> | ④ Numpy array doesn't have built-in metadata |
| ⑤ Slower than Array | ⑤ faster than array |
| ⑥ use case is for Data Analysis & handling labeled data | ⑥ use case is Numerical & scientific computation |

9. What is the role of the dtype parameter in a pandas Series?

Q 9] what is the role of the dtype parameter in pandas Series?

- It defines how the data is stored, processed & interpreted which affects both mem usage & performance.

The dtype allows you to explicitly set the data type of Series.

- if you don't specify the dtype pandas will automatically infer the data type based on the input data

- if the data is mixed or ambiguous, you might get an undesirable type eg (object)

- It improves performance & memory efficiency eg. using int32 or float32 instead of int64 or float 64 if your data doesn't require such large ranges.

eg → `pd.Series([1,2,3])` → int64

`pd.Series([1,2,"Hello"])` → object

`pd.Series([1,1,2], dtype='float32')` → float32

you can change after creation using

`astype()`

`s = pd.Series([1,2,3,"Hello"])`

`print(s)` → type object

`s1 = s.astype('string')`

`print(s1)`

→ string type.

`print(s1[0])` type → dtype.string

```
print("Q-9-----")
```

```
s=pd.Series([1,2,3,"hello"])
```

```
print(s)
```

```
s1=s.astype('string')
```

```
print(s1)
```

```
print(type(s1[0]))
```

```
s2=pd.Series([1,1,2],dtype='float32')  
print(s2)
```

10. Discuss the advantages of using pandas over NumPy for data analysis.

Q. 10] Discuss the advantages of using pandas over numpy for data analysis.

→ **pandas** — is specifically designed for data manipulation & analysis especially with labeled heterogeneous & real-world data.

NumPy — is better suitable for efficient numerical computation, homogeneous data & scientific computing tasks.

pandas — supports label axis to make work with real world data where labels are important to understand the context of the data.

numpy — numpy stores multidimensional data without labels which makes it less understandable for analysis.

pandas — it handles heterogeneous data.
numpy — it handles homogeneous data.

pandas — provides built-in method for handling missing data.

NumPy — There is no built-in function for handling missing data.

pandas — works seamlessly with libraries like matplotlib & Seaborn. enables plotting & visualizing of data. You can directly plot ^{from} DFs.

NumPy — is used for data manipulation before plotting.

Section B: Correct the Error

```
# 1. arr = np.array[1, 2, 3]
```

```
#ANS
```

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

```
# 2. result = np.random.randint(10, 20, size=5, replace=False)
```

```
# ANS:
```

```
# result = np.random.randint(10, 20, size=5) #replace is present choice function
```

```
# 3. np.add([1, 2, 3], [4, 5])
```

```
# ANS
```

```
# print(np.add([1, 2], [4, 5]) )
```

```
# 4. matrix = np.array([[1, 2], [3, 4]]) print(matrix.T())
```

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

```
# ANS: print(matrix.T)
```

```
# 5. inv_matrix = np.linalg.inv([[1, 2], [3, 4]])
```

```
# inv_matrix = np.linalg.inv([[1, 2], [3, 4]])
```

```
# ANS
```

```
# print(inv_matrix)
```

```
# [[-2.  1.]
```

```
# [ 1.5 -0.5]]
```

```
# 6. series = pd.Series[1, 2, 3]
```

```
# ANS
```

```
series = pd.Series([1, 2, 3])
```

```
print(series)
```

```
# 7. data = {'a': 1, 'b': 2, 'c': 3} pd.Series(data, index=data.keys)
```

```
ANS
```

```
data = {'a': 1, 'b': 2, 'c': 3}
```

```
print(pd.Series(data))
```

```
# 8. random_arr = np.random.random(2.5)
```

```
ANS
```

```
random_arr = np.random.rand(2,5)
```



```
print(random_arr)
print(np.random.random(2))#random takes only 1 positional argument
```

```
# 9. matrix = np.array([1, 2], [3, 4])
```

ANS

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

```
# 10. print(pd.Series([1, 2, 3, 4], dtype=int64))
```

ANS

```
print(pd.Series([1, 2, 3, 4], dtype='int64'))#you have to give it in string form if you giving with bytes other wise int is ok
```

Section C: Write a Program

```
from numpy import random as rd
```

```
# 1. Create a 3x3 NumPy array filled with random integers between 1 and 10.
```

```
# arr=rd.randint(1,10,(3,3))
```

```
# print(arr)
```

```
# output:
```

```
# [[3 2 1]
```

```
# [9 1 5]
```

```
# [4 8 4]]
```

```
# 2. Write a program to find the transpose of a 4x4 matrix.
```

```
# arr=np.arange(1,17).reshape(4,4)
```

```
# print(arr.T)
```

```
# output:
```

```
# [[ 1  5  9 13]
```

```
# [ 2  6 10 14]
```

```
# [ 3  7 11 15]
```

```
# [ 4  8 12 16]]
```

```
# 3. Generate a NumPy array of 10 random floats between 0 and 1.
```

```
# print(rd.rand(10))
```

```
# output:
```

```
# [0.57443462 0.19231697 0.25399767 0.39233393 0.26131864 0.12850515
```

```
# 0.24464098 0.6318173 0.26389712 0.31208749]
```

```

# 4. Write a program to compute the inverse of a 2x2 matrix.
# arr=rd.randint(1,10,(2,2))
# print(np.linalg.inv(arr))
# output:
# [[ 1.16666667 -0.33333333]
# [-0.66666667  0.33333333]]

# 5. Create a NumPy array and apply the np.sqrt ufunc on all its elements.
# arr=np.array([1,2,3,4,5,8,16])
# print(np.sqrt(arr))
# output:
# [1.      1.41421356 1.73205081 2.      2.23606798 2.82842712
#  4.      ]

# 6. Write a program to create a pandas Series from a Python dictionary.
d={'Nitis':78,'Lavanya':90,'Samarth':98,'Kirana':34}
# print(pd.Series(d))
# output:
# Nitis    78
# Lavanya  90
# Samarth  98
# Kirana   34
# dtype: int64

# 7. Create a pandas Series with custom indices and retrieve values using indices.
# s=pd.Series([78,89,90,56],index=['nitis','samarth','lavanya','kiran'])
# print(s['lavanya'])
# output:90

# 8. Generate a 5x5 identity matrix using NumPy.
# print(np.eye(5,5,dtype=int))
# output:
# [[1 0 0 0 0]
# [0 1 0 0 0]
# [0 0 1 0 0]
# [0 0 0 1 0]
# [0 0 0 0 1]]

# 9. Write a program to create a NumPy array of even numbers between 10 and 30.
# arr=np.arange(10,30,2)
# print(arr)

```

```
# output:
```

```
# [10 12 14 16 18 20 22 24 26 28]
```

```
# 10. Create a pandas Series from a NumPy array and display its dtype.
```

```
# nprarray=np.array([1,2,3,4,5])
```

```
# arr=pd.Series(nprarray)
```

```
# print(arr.dtype)
```

```
# print(nprarray.dtype)
```

```
# output:
```

```
# 0    1
```

```
# 1    2
```

```
# 2    3
```

```
# 3    4
```

```
# 4    5
```

```
# dtype: int64
```

```
# int64
```

```
# int64
```