```
In [ ]: Lab Assignment 6 - Python
         UTKARSH
In [ ]:
In [ ]: """ 1. Create a 4X2 integer array and Prints its attributes """
In [9]: import numpy as np
         p= np.array([[1, 2], [3, 4], [5, 6], [7, 8]])
         print(p.ndim)
         print(p.shape)
         print(p.size)
         print(p.T)
         print(p.dtype)
        2
        (4, 2)
        [[1 3 5 7]
        [2 4 6 8]]
        int32
In [ ]: """ 2. Create a 5X2 integer array from a range between 100 to 200 such
         that the difference between each element is 10 . """
In [29]: p = np.arange(100, 200, 10).reshape(5, 2)
         print("Array:")
         print(p)
        Array:
        [[100 110]
         [120 130]
         [140 150]
         [160 170]
         [180 190]]
In [ ]: """ 3. Create a numPy array. Return array of items by taking the third
         column from all rows """
In [35]: p= np.array([[1, 2, 3], [3, 4, 9], [5, 6, 11], [7, 8, 13]])
         p_drop=p[:,2]
         print(p_drop)
        [ 3 9 11 13]
In [ ]: """ 4. Create a numpy array with 3 elements. Do all the arithmetic
         operations with scalar element. """
In [45]: p1= np.array([1, 2, 3])
         p2=np.array([4, 5, 7])
         add1=np.add(p1,p2)
         print(add1)
         sub1=np.subtract(p1,p2)
```

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print(sub1)
         prod1=np.multiply(p1,p2)
         print(prod1)
         divd=np.divide(p1,p2)
         print(divd)
       [ 5 7 10]
       [-3 -3 -4]
       [ 4 10 21]
       [0.25
                   0.4
                             0.42857143]
In [ ]: """ 5. Create two 2-D numpy array. Do array concatenation, stack, vstack
         and hstack on that. """
In [70]: a1 = np.array([[1, 2, 3], [4, 5, 6]])
         a2 = np.array([[7, 8, 9], [10, 11, 12]])
         concatenated = np.concatenate((a1, a2))
         stacked = np.stack((a1, a2))
         vstacked = np.vstack((a1, a2))
         hstacked = np.hstack((a1, a2))
         print("Concatenated:")
         print(concatenated)
         print("\nStacked:")
         print(stacked)
         print("\nVertical Stack:")
         print(vstacked)
         print("\nHorizontal Stack:")
         print(hstacked)
       Concatenated:
       [[ 1 2 3]
        [456]
        [7 8 9]
        [10 11 12]]
       Stacked:
       [[[ 1 2 3]
         [ 4 5 6]]
        [[ 7 8 9]
         [10 11 12]]]
       Vertical Stack:
       [[ 1 2 3]
        [456]
        [789]
        [10 11 12]]
       Horizontal Stack:
       [[123789]
        [ 4 5 6 10 11 12]]
```

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In [ ]:
In [ ]: 6. Create an Nd numpy array. Using array filter retrieve all the even
         elements in that array.
In [75]: import numpy as np
         b1 = np.array([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
         even_elements = b1[b1 \% 2 == 0]
         print("Even Elements:")
         print(even_elements)
        Even Elements:
        [2 4 6 8]
In [ ]:
In [ ]: 7. Write a python program to read a dataset from titanic file and do
         the following:
         • Read values to dataframe and print the first 6 rows.
         • Print the concise summary of the dataframe.
         • calculate the mean of each numeric column in dataframe.
         • find the count of null values in dataframe.
         • Take the average people survived.
         • Take the average male and female survived.
         • Group the entire dataframe based on Gender and Pclass.
In [84]: import pandas as pd
         dfw=pd.read_csv('titanic.csv')
         dfw.head(6)
```

Out[84]:		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.925(
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500
	5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583

## In [86]: print(dfw.info())

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	PassengerId	891 non-null	int64
1	Survived	891 non-null	int64
2	Pclass	891 non-null	int64
3	Name	891 non-null	object
4	Sex	891 non-null	object
5	Age	714 non-null	float64
6	SibSp	891 non-null	int64
7	Parch	891 non-null	int64
8	Ticket	891 non-null	object
9	Fare	891 non-null	float64
10	Cabin	204 non-null	object
11	Embarked	889 non-null	object

dtypes: float64(2), int64(5), object(5)

memory usage: 83.7+ KB

None

```
Out[90]: PassengerId 446.000000
          Survived
                         0.383838
          Pclass
                          2.308642
                         29.699118
          Age
          SibSp
                          0.523008
                          0.381594
          Parch
          Fare
                          32.204208
          dtype: float64
 In [91]: dfw.isnull().sum()
Out[91]: PassengerId
          Survived
                           0
          Pclass
                           0
          Name
                           0
          Sex
                           0
          Age
                         177
          SibSp
                           0
          Parch
                           0
          Ticket
                           0
          Fare
                           0
          Cabin
                         687
          Embarked
                           2
          dtype: int64
In [92]: dfw['Survived'].mean()
Out[92]: 0.3838383838383838
          dfw.groupby("Sex")["Survived"].mean()
In [100...
Out[100...
          Sex
          female
                    0.742038
                    0.188908
          male
          Name: Survived, dtype: float64
In [108...
          dfw.groupby(["Pclass","Sex"]).describe()
```

Out[108... Passengerld Si

		count	mean	std	min	25%	50%	75%	max	count	
Pclass	Sex										
1	female	94.0	469.212766	247.476723	2.0	293.50	447.0	698.25	888.0	94.0	0.
	male	122.0	455.729508	247.026449	7.0	255.50	480.5	660.75	890.0	122.0	0.
2	female	76.0	443.105263	243.627288	10.0	269.75	439.5	616.75	881.0	76.0	0.
	male	108.0	447.962963	256.922546	18.0	225.75	416.5	677.50	887.0	108.0	0.
3	female	144.0	399.729167	267.232416	3.0	165.25	376.0	636.00	889.0	144.0	0.
	male	347.0	455.515850	261.921251	1.0	209.50	466.0	687.50	891.0	347.0	0.

6 rows × 48 columns

- #• Read values to dataframe and print the first 6 rows.
- #• Print the concise summary of the dataframe.
- #• calculate the mean of each numeric column in dataframe.
- #• find the count of null values in dataframe.
- #• Take the maximum temperature from the weather data.
- #• Retrieve the average windspeed from the weather data.
- #• Make a new dataframe which contains minmum values of all columns based on event.
- #• Make a new dataframe which contains minmum temperature based on event.

In [3]: import pandas as pd
wd=pd.read\_csv('weather.csv')
wd.head(6)

Out[3]:		Unnamed: 0	day	temperature	windspeed	event
	0	0	01-01-2022	32.0	6.0	Rain
	1	1	01-04-2022	NaN	9.0	Sunny
	2	2	01-05-2022	28.0	NaN	Snow
	3	3	01-06-2022	NaN	7.0	NaN
	4	4	01-07-2022	32.0	NaN	Rain
	5	5	01-08-2022	NaN	NaN	Sunny

In [4]: wd.info()

```
Data columns (total 5 columns):
           Column
                       Non-Null Count Dtype
       --- -----
                       -----
          Unnamed: 0 11 non-null
       0
                                        int64
       1
           day 11 non-null
                                        object
           temperature 7 non-null
                                        float64
        3
           windspeed 7 non-null
                                        float64
       4
           event
                        9 non-null
                                        object
      dtypes: float64(2), int64(1), object(2)
      memory usage: 572.0+ bytes
In [5]: wd.describe()
Out[5]:
               Unnamed: 0 temperature windspeed
                 11.000000
                              7.000000
                                         7.000000
        count
        mean
                  5.000000
                             34.285714
                                         8.857143
          std
                  3.316625
                              4.386125
                                         2.340126
          min
                  0.000000
                             28.000000
                                         6.000000
         25%
                  2.500000
                             32.000000
                                         7.500000
         50%
                  5.000000
                             34.000000
                                         8.000000
         75%
                  7.500000
                             37.000000
                                        10.500000
                 10.000000
                             40.000000
         max
                                        12.000000
In [6]:
       wd.mean(numeric_only=True)
Out[6]: Unnamed: 0
                        5.000000
        temperature
                       34.285714
        windspeed
                        8.857143
        dtype: float64
In [7]: wd.isnull().sum()
Out[7]: Unnamed: 0
        day
                       0
        temperature
                       4
        windspeed
                       4
        event
                       2
        dtype: int64
In [8]: wd['temperature'].max()
Out[8]: 40.0
In [9]: wd['windspeed'].mean()
Out[9]: 8.857142857142858
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 11 entries, 0 to 10

```
In [10]: wd.groupby('event').min()
Out[10]:
                  Unnamed: 0
                                     day temperature windspeed
           event
          Cloudy
                            7 01-10-2022
                                                  34.0
                                                               0.8
            Rain
                            0 01-01-2022
                                                  32.0
                                                               6.0
           Snow
                            2 01-05-2022
                                                  28.0
                                                             NaN
           Sunny
                            1 01-04-2022
                                                  40.0
                                                               9.0
In [11]: wd.groupby('event')['temperature'].min()
Out[11]: event
          Cloudy
                    34.0
                    32.0
          Rain
          Snow
                    28.0
                    40.0
          Sunny
          Name: temperature, dtype: float64
 In [ ]:
 In [ ]: #9. Write a python program to read a dataset from diamonds file and
          #do the following:
          #• Read values to dataframe and print the first 6 rows.
          #• Print the concise summary of the dataframe.
          #• calculate the mean of each numeric column in dataframe.
          #• find the count of null values in dataframe.
          #• Calculate min, max price for each cut of diamonds.
          #• Take the count of all duplicate rows in the dataframe.
In [15]: import pandas as pd
          di=pd.read_csv('diamonds.csv')
          di.head(6)
Out[15]:
             Unnamed: 0 carat
                                      cut color clarity depth table price
                                                                               X
                                                                                    у
                                                                                          Z
          0
                       1
                          0.23
                                     Ideal
                                              Ε
                                                    SI2
                                                                 55.0
                                                                        326 3.95 3.98 2.43
                                                          61.5
                       2
                          0.21
                                 Premium
                                              Ε
                                                    SI1
                                                          59.8
                                                                 61.0
                                                                            3.89
                                                                                 3.84 2.31
          1
                                                                        326
          2
                       3
                          0.23
                                    Good
                                              Ε
                                                          56.9
                                                                 65.0
                                                                        327 4.05 4.07 2.31
                                                   VS1
          3
                          0.29
                                 Premium
                                                   VS2
                                                          62.4
                                                                 58.0
                                                                        334 4.20 4.23 2.63
          4
                       5
                          0.31
                                    Good
                                              J
                                                    SI2
                                                          63.3
                                                                 58.0
                                                                        335 4.34 4.35 2.75
          5
                          0.24 Very Good
                                                  VVS2
                                                          62.8
                                                                 57.0
                                                                        336 3.94 3.96 2.48
In [16]: di.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 53940 entries, 0 to 53939
Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	Unnamed: 0	53940 non-null	int64
1	carat	53940 non-null	float64
2	cut	53940 non-null	object
3	color	53940 non-null	object
4	clarity	53940 non-null	object
5	depth	53940 non-null	float64
6	table	53940 non-null	float64
7	price	53940 non-null	int64
8	X	53940 non-null	float64
9	у	53940 non-null	float64
10	Z	53940 non-null	float64

dtypes: float64(6), int64(2), object(3)

memory usage: 4.5+ MB

## In [17]: di.describe()

Out[17]:

	Unnamed: 0	carat	depth	table	price	х
count	53940.000000	53940.000000	53940.000000	53940.000000	53940.000000	53940.000000
mean	26970.500000	0.797940	61.749405	57.457184	3932.799722	5.731157
std	15571.281097	0.474011	1.432621	2.234491	3989.439738	1.121761
min	1.000000	0.200000	43.000000	43.000000	326.000000	0.000000
25%	13485.750000	0.400000	61.000000	56.000000	950.000000	4.710000
50%	26970.500000	0.700000	61.800000	57.000000	2401.000000	5.700000
75%	40455.250000	1.040000	62.500000	59.000000	5324.250000	6.540000
max	53940.000000	5.010000	79.000000	95.000000	18823.000000	10.740000

```
In [18]: di.mean(numeric_only=True)
```

```
Out[18]: Unnamed: 0
                       26970.500000
         carat
                          0.797940
         depth
                          61.749405
         table
                          57.457184
                        3932.799722
         price
                           5.731157
         Х
                           5.734526
         У
                           3.538734
         dtype: float64
```

```
In [19]: di.groupby('cut')['price'].agg(['min', 'max'])
```

```
Out[19]: min max

cut

Fair 337 18574

Good 327 18788

Ideal 326 18806

Premium 326 18823

Very Good 336 18818

In [20]: di.duplicated().sum()

Out[20]: 0

In []: In []:
```