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COURSE : B.VOC SOFTWARE DEVELOPMENT

SUBJECT: DATA ANALYSIS AND VISUALIZATION USING PYTHON

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```
#1. Write programs in Python using NumPy library to do the following:
#a. Compute the mean, standard deviation, and variance of a two dimensional random integer array along
#the second axis
import numpy as np
x=np.random.randint(10,50,5)
y=np.random.randint(5,80,5)
arr=np.array((x))
arr1=np.array((y))
z=np.array([[arr],[arr1]])
print(z)
print(z)
print(z.shape)
print("Mean\n",z.mean(axis=1))
print("standard deviation\n",z.std(axis=1))
print("variance\n",z.var(axis=1))
```

```
[[[18 16 13 12 23]]

[[21 41 9 60 69]]]
(2, 1, 5)

Mean

[[18. 16. 13. 12. 23.]

[21. 41. 9. 60. 69.]]

standard deviation

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

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

variance

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

```
#b. Create a 2-dimensional array of size m x n integer elements, also print the shape, type and data type of #the array and then reshape it into an n x m array, where n and m are user inputs given at the run time. import numpy as np m=eval(input("enter a row number")) n=eval(input("enter a column number")) x=np.random.randint(4,50,m) y=p.random.randint(5,80,n) arr=np.array((x)) arr1=np.array((y)) z=np.array([[arr],[arr1]]) print(z) print(z.shape) print(z.shape)
```

```
enter a row number5
enter a column number5
[[[35 5 31 20 25]]

[[58 18 24 35 60]]]
(2, 1, 5)
int64
```

```
#c. Test whether the elements of a given 1D array are zero, non-zero and NaN. Record the indices of these
    #elements in three separate arrays.
    import numpy as np
    arr = np.array([5,69,8,np.nan,89,0,5,0,56,0,np.nan,np.nan,89,569])
   11 = []
   12 = []
   13 = []
    for i in range(len(arr)):
     if (np.isnan(arr[i])):
       11.append(i)
      elif (i == 0):
      12.append(i)
     else:
       13.append(i)
    print("null value index",11)
    print("zero value index",12)
    print("not null value index",13)
```

```
→ null value index [3, 10, 11]
zero value index [0]
not null value index [1, 2, 4, 5, 6, 7, 8, 9, 12, 13]
```

```
#d. Create three random arrays of the same size: Array1, Array2 and Array3. Subtract Array 2 from Array3
    #and store in Array4. Create another array Array5 having two times the values in Array1.
    # Find Covariance and Correlation of Array1 with Array4 and Array5 resp
    import numpy as np
    x=np.random.randint(0,100,5)
    y=np.random.randint(0,100,5)
    z=np.random.randint(0,100,5)
    array_1=np.array((x))
    array_2=np.array((y))
    array_3=np.array((z))
    array_4=np.array((array_2 - array_3))
    print("array_4",array_4)
    array_5=np.array((2*array_1))
    print ("array_5",array_5)
    print("correlation",np.corrcoef([array_1],[array_4]))
    print("correlation",np.corrcoef([array_1],[array_5]))
    print("covariance",np.cov(array_1,array_4))
    print("covariance",np.cov(array_1,array_5))
```

```
#e. Create two random arrays of the same size 10: Array1, and Array2. Find the sum of the first half of both
#the arrays and product of the second half of both the arrays.
import numpy as np
x=np.random.randint(0,50,10)
y=np.random.randint(0,100,10)
arr=np.array((x))
arr1=np.array((y))
z=arr[:len(x)//2]
#print(arr[:z])
z1=arr1[:len(y)//2]
#print("sum of z and z1",z+z1)
m = arr[len(x)//2:]
m1 = arr1[len(x)//2:]
print("multiple of m and m1",m*m1)
```

```
⇒ sum of z and z1 [121 117 77 87 66]
multiple of m and m1 [3040 1599 1092 1887 611]
```

```
#2. Do the following using PANDAS Series:
import pandas as pd
series = pd.Series((12,23,45,67,876,87,6,5445,43,45,5,66,7,7,6,6,6,56,0,78,8,89,6664,865,78,65))

#a. Create a series with 5 elements. Display the series sorted on index and also sorted = series.sort_index()
sorted_values = series.sort_values()
print(sorted)
print("sorted_values", sorted_values)
```

# **D**

0	12
1	23
2	45
3	67
4	876
5	87
6	6
7	5445
8	43
9	45
10	5
11	66
12	7
13	7
14	6
15	6
16	6
17	56
18	0
19	78
20	8
21	89
22	6664
23	865
24	78
25	65
dtype:	int64

```
U --,,--- -----
   sorted_values 18
₹ 10
          5
   16
           6
   15
           6
   14
           6
   6
           6
   12
           7
   13
          7
   20
          8
   0
          12
   1
          23
   8
          43
   2
          45
   9
          45
   17
          56
   25
          65
   11
          66
   3
          67
   24
          78
          78
   19
   5
         87
   21
         89
   23 865
   4
        876
   7
        5445
   22
       6664
   dtype: int64
```

```
#b. Create a series with N elements with some duplicate values. Find the minimum and maximum ranks
#assigned to the values using 'first' and 'max' methods
print("According to first method\n", series.rank(method = "first"))
print("\n According to max method\n", series.rank(method = "max"))
```

```
According to max method
According to first method
                                                   10.0
             10.0
     0
                                            1
                                                  11.0
           11.0
    2
           13.0
                                            2
                                                  14.0
    3
           18.0
                                            3
                                                  18.0
                                            4
                                                  24.0
    4
           24.0
    5
                                            5
           21.0
                                                  21.0
            3.0
                                            6
                                                  6.0
    6
    7
           25.0
                                            7
                                                  25.0
    8
           12.0
                                            8
                                                  12.0
                                            9
                                                  14.0
    9
           14.0
            2.0
    10
                                            10
                                                  2.0
                                                  17.0
    11
           17.0
                                            11
    12
            7.0
                                            12
                                                  8.0
    13
            8.0
                                            13
                                                   8.0
    14
            4.0
                                            14
                                                   6.0
    15
            5.0
                                            15
                                                   6.0
    16
            6.0
                                            16
                                                  6.0
    17
           15.0
                                                  15.0
                                            17
            1.0
    18
                                            18
                                                  1.0
    19
           19.0
                                                  20.0
                                            19
    20
            9.0
                                            20
                                                  9.0
    21
           22.0
                                            21
                                                  22.0
    22
           26.0
                                                  26.0
                                            22
    23
           23.0
                                            23
                                                  23.0
    24
           20.0
                                                  20.0
                                            24
    25
           16.0
                                            25
                                                  16.0
    dtype: float64
                                             dtype: float64
```

```
#c. Display the index value of the minimum and maximum element of a Series
x = series.max()
y = series.min()
print("Max value", series.loc[series.idxmax()])
print("Max values index", series.idxmax())
print("Min value", series.loc[series.idxmin()])
print("Min values index", series.idxmin())
```

```
Max value 6664
Max values index 22
Min value 0
Min values index 18
```

```
#3. Create a data frame having at least 3 columns and 50 rows to store numeric data generated using a random
    #function. Replace 10% of the values by null values whose index positions are generated using random function.
    #Do the following:
    import pandas as pd
    import numpy as np
    series = np.random.rand(10,3)
    data = pd.DataFrame(series,columns = ['column1' ,'column2','column3'])
    #print(data)
    null\_values = np.random.choice(data.size , size = int(0.1 * data.size))
    data.values.ravel()[null_values] = np.nan
    #print(data)
    #a. Identify and count missing values in a data frame.
    missing = data.isnull().sum()
    print(missing)
    #b. Drop the column having more than 5 null values.
    #data = data.dropna(axis = 1 ,thresh = len(data) - 5)
    #c. Identify the row label having maximum of the sum of all values in a row and drop that row.
    #max row = data.sum(axis = 1).idxmax()
    #data= data.drop(index = max_row)
```

```
#d. Sort the data frame on the basis of the first column.
#data = data.sort_values("column1")

#e. Remove all duplicates from the first column.
#data = data.drop_duplicates("column1")

#f. Find the correlation between first and second column and covariance between second and third
#column.
correlation = data["column1"].corr(data["column2"])
covariance = data["column2"].cov(data["column3"])
print(correlation)
print(covariance)

#g. Discretize the second column and create 5 bins.
data["column2_bins"] = pd.cut(data["column2"] , bins = 5)
print(data)
```

```
→ column1

       column2
                         a
       column3
       dtype: int64
       -0.236751857037486
       -0.04964966980899943
      column1 column2 column3 column2_bins
0 0.003349 0.714170 0.467341 (0.642, 0.817]
1 0.931825 0.423493 0.823253 (0.292, 0.467]
2 0.690441 0.992039 0.058476 (0.817, 0.992]
      3 NaN 0.615732 0.612446
4 0.876113 0.117184 0.969109
                                                            (0.467, 0.642]
                                                            (0.116, 0.292]
           0.570485 0.916646 0.759010 (0.817, 0.992]
0.492786 0.404287 0.622129 (0.292, 0.467]
       5
           0.063221 0.453420 0.674204 (0.292, 0.467]
       7
                   2525 0.185338 NAN (0.116, 0.292]
NAN 0.674480 0.726861 (0.642, 0.817]
       8 0.612525
```

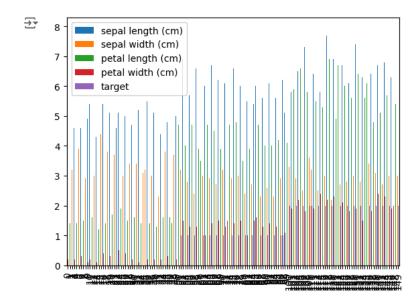
```
#4. Consider two excel files having attendance of two workshops. Each file has three fields 'Name', 'Date, duration
       #(in minutes) where names are unique within a file. Note that duration may take one of three values (30, 40, 50)
       #only. Import the data into two data frames and do the following:
       import pandas as pd
       df1 = pd.read_excel('workshop1.xlsx')
       df2 = pd.read_excel('workshop2.xlsx')
       # a. Perform merging of the two dataframes to find the names of students who attended both workshops
       attended_both = pd.merge(df1, df2, on='Name', how='inner')
       print("names of students who attended both workshops\n",attended both)
       # b. Find names of all students who attended a single workshop only
       attended_one = pd.concat([df1, df2]).drop_duplicates(keep=False)['Name']
       print("names of all students who attended a single workshop\n",attended_one)
       \ensuremath{\text{\#}} c. Merge two dataframes row-wise and find the total number of records
       rowwise = pd.concat([df1, df2])
       print(" the total number of records\n",len(rowwise))
       # d. Merge two dataframes row-wise and use two columns (names and dates) as multi-row indexes
       rowwise_indexes = pd.concat([df1.set_index(['Name', 'Date']), df2.set_index(['Name', 'Date'])])
       # e. Generate descriptive statistics for this hierarchical dataframe
       statistics = rowwise_indexes.describe()
       print(statistics)
```

```
names of students who attended both workshops
```

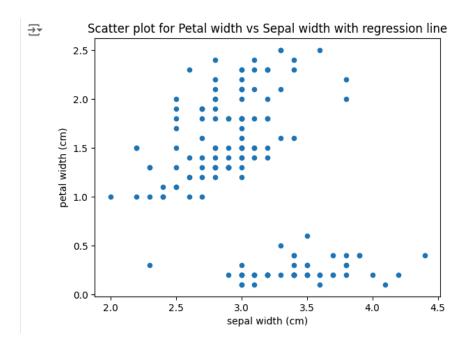
```
Date_y Duration_y
             Date_x Duration_x
     RAM 2024-02-03
                             45 2024-05-03
0
1 NAREN 2024-02-04
                             45 2024-03-04
                                                     45
                             45 2024-03-06
2 DHRUV 2024-02-06
                                                     45
   AMAN 2024-02-08
                             45 2024-03-08
                                                     45
4 MOHAN 2024-02-10
                             45 2024-03-10
                                                     45
names of all students who attended a single workshop
       SHIVAM
         RAM
1
2
       NAREN
3
    ASHWANI
4
      DHRUV
5
      VIKAS
6
        AMAN
7
      DEEPAK
8
     MOHAN
0
     RUPESH
1
         RAM
2
     NAREN
3
         DEV
4
      DHRUV
5
      ASHISH
6
        AMAN
7
       SHLOK
       MOHAN
```

Name: Name, dtype: object

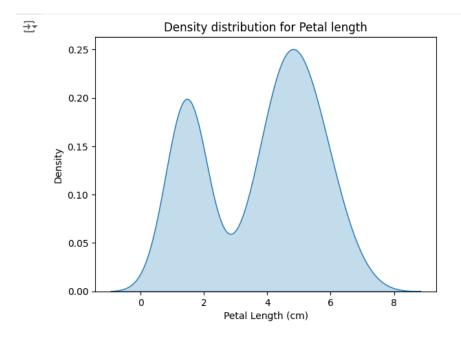
```
the total number of records
18
       Duration
           18.0
count
           45.0
mean
            0.0
std
min
           45.0
25%
           45.0
50%
           45.0
75%
           45.0
           45.0
max
```



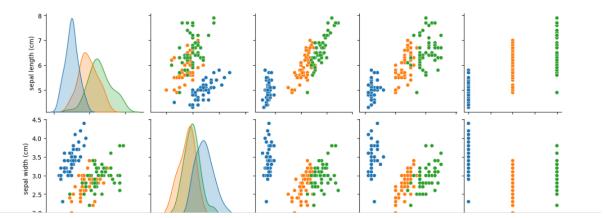
```
#b. Draw a scatter plot for Petal width vs sepal width and fit a regression line
iris_df.plot.scatter(x='sepal width (cm)', y='petal width (cm)')
plt.title('Scatter plot for Petal width vs Sepal width with regression line')
plt.show()
```



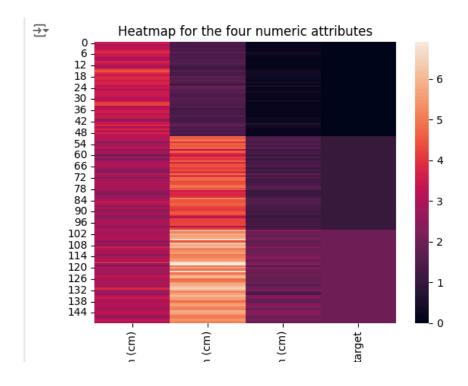
```
#c. Plot density distribution for feature petal length.
sns.kdeplot(data=iris_df['petal length (cm)'], shade=True)
plt.title('Density distribution for Petal length')
plt.xlabel('Petal Length (cm)')
plt.ylabel('Density')
plt.show()
```



```
#d. Use a pair plot to show pairwise bivariate distribution in the Iris Dataset.
sns.pairplot(iris_df, hue='species')
plt.suptitle('Pairwise bivariate distribution in the Iris Dataset', y=1.02)
plt.show()
```



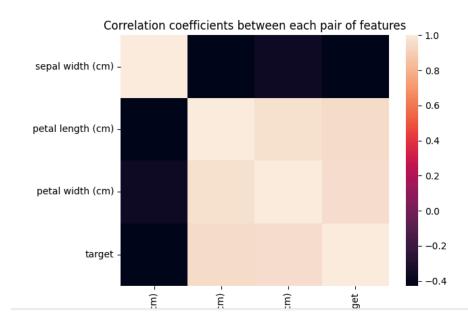
```
#e. Draw heatmap for the four numeric attributes
sns.heatmap(iris_df[iris_df.columns[1:5]])
plt.title('Heatmap for the four numeric attributes')
plt.show()
```



```
#f. Compute mean, mode, median, standard deviation, confidence interval and standard error for each
#feature
print(iris_df.describe())
print(iris_df[iris_df.columns[1:-2]].mode(axis=1))
```

	Sec	peti	pet		
count mean std	sepal length ( 150.000 5.843 0.828	000 333	width (cm) 150.000000 3.057333 0.435866	petal length (cm) 150.000000 3.758000 1.765298	\
min 25% 50% 75% max	4.300 5.100 5.800 6.400 7.900	000 000 000	2.000000 2.800000 3.000000 3.300000 4.400000	1.000000 1.600000 4.350000 5.100000 6.900000	
count mean std min 25% 50% 75% max	petal width (c 150.0000 1.1993 0.7622 0.1000 0.3000 1.3000 1.8000 2.5000	00 150.000 33 1.000 38 0.819 00 0.000 00 0.000 00 1.000 00 2.000	9000 9232 9000 9000 9000		

```
#g. Compute correlation coefficients between each pair of features and plot heatmap
correlation_matrix = (iris_df[iris_df.columns[1:5]].corr())
sns.heatmap(correlation_matrix)
plt.title('Correlation coefficients between each pair of features')
plt.show()
```



```
#6. Consider the following data frame containing a family name, gender of the family member and her/his monthly 
#income in each record.
import pandas as pd
import numpy as np
data = {
    "name":['shah','vats','vats','kumar','vats','kumar','shah','shah','shah','vats'],
    "Gender":['male','male','female','female','male','male','male','male'],
    "MonthlyIncome":[114000.0,65000.0,43150.0,69500.0,155000.0,103000,55000,112400,81030,71900]
}
data1 = pd.DataFrame(data)
#print(data1)

#a. Calculate and display familywise gross monthly income.
familywise = data1.groupby("name")["MonthlyIncome"].sum()
print(familywise)
```

```
name
kumar 253530.0
shah 281400.0
vats 335050.0
Name: MonthlyIncome, dtype: float64
```

```
#b. Calculate and display the member with the highest monthly income.
member = data1.loc[data1["MonthlyIncome"].idxmax()]
print(member)
```

```
name vats
Gender female
MonthlyIncome 155000.0
Name: 4, dtype: object
```

```
#c. Calculate and display monthly income of all members with income greater than
monthly = data1[data1["MonthlyIncome"]>60000]
print(monthly)
```

```
мате: 4, атуре: орјест
   name Gender MonthlyIncome
Θ
   shah male
                   114000.0
          male
1
   vats
                    65000.0
3 kumar female
                    69500.0
4
  vats female
                    155000.0
        male
5
  kumar
                    103000.0
7
  shah female
                   112400.0
8 kumar female
                    81030.0
        male
                    71900.0
  vats
```

```
#d. Calculate and display the average monthly income of the female members
avg_income = data1[data1["Gender"]=="female"]["MonthlyIncome"].mean()
print("avg_income",avg_income")
```

avg income 92216.0

```
#7. Using Titanic dataset, to do the following:
 import pandas as pd
 import numpy as np
 data = pd.read_csv("titanic.csv")
 #print(data)
 #a. Find total number of passengers with age less than 30
 age = data[data["age"]<30]
 print("total number of passengers with age less than 30: ",len(age))
 #b. Find total fare paid by passengers of first class
 first = data[data["pclass"]== 1]["fare"].sum()
 print(" total fare paid by passengers of first class",first)
 #c. Compare number of survivors of each passenger class
 survivors = data.groupby('pclass')['survived'].sum()
 print("number of survivors of each passenger class", survivors)
 #d. Compute descriptive statistics for any numeric attribute genderwise
 statistics = data.groupby('sex')['age'].describe()
 print(statistics)
```

```
→ total number of passengers with age less than 30: 384
    total fare paid by passengers of first class 18177.4125
    number of survivors of each passenger class pclass
    1
        136
    2
         87
    Name: survived, dtype: int64
                                  std min 25% 50%
                                                        75%
           count
                      mean
                                                              max
    sex
    female 261.0 27.915709 14.110146 0.75 18.0 27.0 37.0 63.0
    male
           453.0 30.726645 14.678201 0.42 21.0 29.0 39.0 80.0
```