RAMANUJAN COLLEGE UNIVERSITY OF DELHI



DSE – DATA ANALYSIS AND VISUALIZATION (PRACTICAL FILE)

SUBMITTED TO: Mrs. Sheetal Singh Ma'am

SUBMITTED BY: Sukaina Inam Naqvi

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Examination Roll No.- 20020570033

B.Sc. (H) Computer Science | V Semester

PRACTICAL - QUESTIONS

Program No.

Question

1.	Given below is a dictionary having two keys 'Boys' and 'Girls' and having two lists of heights of five Boys and Five Girls respectively as values associated with these keys
2.	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
3.	Create a dataframe 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:
4.	Consider two excel files having attendance of a workshop's participants for two days. Each file has three fields 'Name', 'Time of joining', 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 dataframes and do the following:
5.	Taking Iris data, plot the following with proper legend and axis labels: (Download IRIS data from: https://archive.ics.uci.edu/ml/datasets/iris or import it from sklearn.datasets)
6.	Consider any sales training/ weather forecasting dataset
7.	Consider a data frame containing data about students i.e. name, gender and passing division:
8.	Consider the following data frame containing a family name, gender of the family member and her/his monthly income in each record.

1. Given below is a dictionary having two keys 'Boys' and 'Girls' and having two lists of heights of five Boys and Five Girls respectively as values associated with these keys Original dictionary of lists:

```
{'Boys': [72, 68, 70, 69, 74], 'Girls': [63, 65, 69, 62, 61]}
From the given dictionary of lists create the following list of dictionaries:
[{'Boys': 72, 'Girls': 63}, {'Boys': 68, 'Girls': 65}, {'Boys': 70, 'Girls': 69}, {'Boys': 69, 'Girls': 62}, {'Boys': 74, 'Girls': 61]
```

Solution:

```
test_dict = {'Boys':[72,68,70,69,74],'Girls':[63,65,69,62,61]}
print("Original Dictionary is : " + str(test_dict))
res = [{key:value[i] for key,value in test_dict.items()} for i in range(5)]
print("The converted list of dictionaries is : " + str(res))
```

Output:

```
Original Dictionary is : {'Boys': [72, 68, 70, 69, 74], 'Girls': [63, 65, 69, 62, 61]}

The converted list of dictionaries is :
[{'Boys': 72, 'Girls': 63}, {'Boys': 68, 'Girls': 65}, {'Boys': 70, 'Girls': 69}, {'Boys': 69, 'Girls': 62}, {'Boys': 74, 'Girls': 61}]
```

- 2. 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.

Solution:

```
import numpy as np
from numpy import random
arr = np.random.randint(6, size=(3,4))
print("2D random integer array is \n", arr)
print("mean of array is : ", arr.mean(axis= 1) )
print("standard deviation of array is : ", arr.std(axis= 1) )
print("variance of array is : ", arr.var(axis= 1) )
```

```
2D random integer array is
[[5 4 2 4]
[2 4 0 0]
[0 5 2 0]]
mean of array is: [3.75 1.5 1.75]
standard deviation of array is: [1.08972474 1.6583124 2.04633819]
variance of array is: [1.1875 2.75 4.1875]
```

b. Get the indices of the sorted elements of a given array. a. B = [56, 48, 22, 41, 78, 91, 24, 46, 8, 33]

Solution:

```
import numpy as np
B = [56, 48, 22, 41, 78, 91, 24, 46, 8, 33]
print("given array : ", B)
indices = np.argsort(B)
print("indices of the sorted element of a given array are : ", indices)
```

Output:

```
given array: [56, 48, 22, 41, 78, 91, 24, 46, 8, 33] indices of the sorted element of a given array are: [8 2 6 9 3 7 1 0 4 5]
```

c. Create a 2-dimensional array of size $m \times n$ integer elements, also print the shape, type and data type of the array and then reshape it into $n \times m$ array, n and m are user inputs given at the run time.

Solution:

```
import numpy as np

from numpy import random

arr = np.random.randint(100,size=(3,4))

print("2D array of dimension 3x4 : \n",arr)

print("Shape of array is : ", np.shape(arr))

print("data type of array is : ", arr.dtype)

print("type of array is :", type(arr))

print("after reshaping the array it will be 4x3:\n", arr.reshape(4,3))
```

Output:

```
2D array of dimension 3x4 :

[[26 36 89 95]

[ 0 22 39 64]

[ 2 3 68 94]]

Shape of array is : (3, 4)

data type of array is : int32

type of array is : <class 'numpy.ndarray'>
after reshaping the array it will be 4x3 :

[[26 36 89]

[95 0 22]

[39 64 2]

[ 3 68 94]]
```

d. Test whether the elements of a given array are zero, non-zero and NaN. Record the indices of these elements in three separate arrays.

```
import numpy as np
arr = np.array([1, 0, 2, 0, 3, np.nan, 0, 5, np.nan])
```

```
print("original array is : ", arr)

res = np.where(arr== 0)[0]

array1 = np.array(res)

print("indices of zero elements in given array : ", array1)

res1 = np.where(arr !=0)[0]

array2 = np.array(res1)

print("indices of non zero elements in given array : ",array2)

res2 = np.where(np.isnan(arr))

array3 = np.array(res2)

print("indices of NaN in given array : ",array3)
```

```
original array is: [ 1. 0. 2. 0. 3. nan 0. 5. nan]
indices of zero elements in given array: [1 3 6]
indices of non zero elements in given array: [0 2 4 5 7 8]
indices of NaN in given array: [[5 8]]
```

3. Create a dataframe 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:

Solution

```
before replacing 10% values by null the dataframe becomes :
   A B C
0 11 28 61
1 84 80 18
2 94 82 25
     85 70
3 77
4 99 59
         8
after replacing 10% values by null the dataframe becomes :
0 11.0 28.0 61.0
1 84.0 80.0
             NaN
  94.0 82.0 25.0
3
  77.0
       NaN 70.0
4 99.0 59.0 NaN
```

a) Identify and count missing values in a dataframe.

Solution:

print(" \nTotal null values in each column :\n",df.isnull().sum())

Output:

```
Total null values in each column:
A 4
B 6
C 5
dtype: int64
```

b) Drop the column having more than 5 null values.

Solution:

```
df1 = df.dropna(thresh=45,axis=1 )
print(df1)
```

Output:

c.) Identify the row label having maximum of the sum of all values in a row and drop that row.

```
sum=df.sum(axis=1)
print("Sum of rows :\n",sum.head())
print("\nMaximum Sum is :",sum.max())
max sum row = df.sum(axis=1).idxmax()
```

```
print("\nRow index having maximum sum is :" ,max_sum_row)

df = df.drop(max_sum_row ,axis =0)
print("\nData frame after removing the row having maximum sum value: \n",df.head())
```

```
Sum of rows
     100.0
    164.0
    201.0
    147.0
    158.0
dtype: float64
Maximum Sum is : 216.0
Row index having maximum sum is : 15
Data frame after removing the row having maximum sum value:
 11.0 28.0 61.0
  84.0 80.0
              NaN
  94.0 82.0 25.0
  77.0
         NaN 70.0
4 99.0 59.0
              NaN
```

d.) Sort the dataframe on the basis of the first column.

Solution:

```
sorted_df = df.sort_values(by='A')
print(sorted_df.head())
```

Output:

e.) Remove all duplicates from the first column.

Solution:

```
df.drop_duplicates(subset="A")
print(df.head())
```

	Α	В	C
0	11.0	28.0	61.0
1	84.0	80.0	NaN
2	94.0	82.0	25.0
3	77.0	NaN	70.0
4	99.0	59.0	NaN

f.) Find the correlation between first and second column and covariance between second and third column.

Solution:

```
correlation = df['A'].corr(df['B'])
print("CORRELATION between column A and B: ", correlation)
covariance = df['B'].cov(df['C'])
print("COVARIANCE between column B and C:",covariance)
```

Output:

```
CORRELATION between column A and B: 0.08926569675122807 COVARIANCE between column B and C: -280.7904761904761
```

g.) Detect the outliers and remove the rows having outliers.

Solution:

```
from scipy import stats
import numpy as np

z = np.abs(stats.zscore(df['A']))
print(z)
threshold = 1
outlier_position = np.where(z > 1)
df.drop(outlier_position[0])
print(df)
```

Output:

```
0
   NaN
1
   NaN
2
   NaN
3
   NaN
4
   NaN
Name: A, dtype: float64
           В
                C
     Α
0 11.0 28.0 61.0
1 84.0 80.0 NaN
2 94.0 82.0 25.0
3 77.0
       NaN 70.0
4 99.0 59.0 NaN
```

h.) Discretize second column and create 5 bins

Solution:

```
df1 = pd.cut(df['B'],bins=5).head()
df1
```

```
Out[151]: 0 (20.8, 39.6]
1 (77.2, 96.0]
2 (77.2, 96.0]
3 NaN
4 (58.4, 77.2]
Name: B, dtype: category
Categories (5, interval[float64, right]): [(1.906, 20.8] < (20.8, 39.6] < (39.6, 58.4] < (58.4, 77.2] < (77.2, 96.0]]
```

4. Consider two excel files having attendance of a workshop's participants for two days. Each file has three fields 'Name', 'Time of joining', 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 dataframes and do the following:

Solution:

```
import numpy as np
import pandas as pd
d1_df = pd.read_excel(r'C:\Users\SUKAINA\Downloads\day1.xlsx')
d2_df = pd.read_excel(r'C:\Users\SUKAINA\Downloads\day2.xlsx')
print("Day1 data : \n",d1_df)
print("\nDay2 data : \n",d2_df)
```

Output:

```
Day1 data:
        Name Time of joining Duration
    Akansha
                  10:00:00
                                   30
     Sakshi
                                  50
1
                   10:05:00
2
       Esha
                   10:08:01
                                  40
3 Himanshi
                   10:05:20
                                  50
4
     Vinay
                                  50
                  10:06:10
5
     Sumeg
                   10:07:00
                                  40
6
      Naina
                                  30
                  10:12:00
7
     Adarsh
                  10:22:00
                                  30
Day2 data:
      Name Time of joining Duration
Ø Akansha
                10:00:00
1
                                40
   Harsh
                10:09:12
2
   Aparna
                10:09:15
                                30
3 Sakshi
                10:11:10
                                40
                                50
4
     Esha
                10:11:20
5
  Yogita
                10:15:00
                                40
```

10:21:00

a.) Perform merging of the two dataframes to find the names of students who had attended the workshop on both days.

30

Solution:

```
pd.merge(d1 df,d2 df,how='inner',on='Name')
```

6 sukaina

Out[15]:

	Name	Time of joining	Duration_x	Time of joining	Duration_y
0	Akansha	10:00:00	30	10:00:00	30
1	Sakshi	10:05:00	50	10:11:10	40
2	Esha	10:08:01	40	10:11:20	50

b.) Find names of all students who have attended workshop on either of the days.

Solution:

```
either_day = pd.merge(d1_df,d2_df,how='outer',on='Name')
either_day
```

Output:

Out[16]:

	Name	Time of joining	Duration_x	Time of joining	Duration_y
0	Akansha	10:00:00	30.0	10:00:00	30.0
1	Sakshi	10:05:00	50.0	10:11:10	40.0
2	Esha	10:08:01	40.0	10:11:20	50.0
3	Himanshi	10:05:20	50.0	NaN	NaN
4	Vinay	10:06:10	50.0	NaN	NaN
5	Sumeg	10:07:00	40.0	NaN	NaN
6	Naina	10:12:00	30.0	NaN	NaN
7	Adarsh	10:22:00	30.0	NaN	NaN
8	Harsh	NaN	NaN	10:09:12	40.0
9	Aparna	NaN	NaN	10:09:15	30.0
10	Yogita	NaN	NaN	10:15:00	40.0
11	sukaina	NaN	NaN	10:21:00	30.0

c.) Merge two data frames row-wise and find the total number of records in the data frame.

Solution:

```
either_day['Name'].count()
```

Output:

```
In [17]: either_day['Name'].count()
Out[17]: 12
```

d.) Merge two data frames and use two columns names and duration as multi-row indexes. Generate descriptive statistics for this multi-index.

```
either_day.set_index(['Name','Duration_x'],inplace = True,append = True,drop = False)
```

```
print(either_day.index)
print(either day.describe(include='all')) #descriptive statistic of multi index multi index
```

```
(0, 'Akansha', 30.0, 'Akansha', 30.0, 'Akansha', 30.0, ...), (1, 'Sakshi', 50.0, 'Sakshi', 50.0, 'Sakshi', 50.0, ...), (2, 'Esha', 40.0, 'Esha', 40.0, 'Esha', 40.0, ...), (3, 'Himanshi ', 50.0, 'Himanshi ', 50.0, 'Himanshi ', 50.0, ...), (4, 'Vinay ', 50.0, 'Vinay ', 50.0, 'Vinay ', 50.0, ...), (5, 'Sumeg', 40.0, 'Sumeg', 40.0, 'Sumeg', 40.0, ...),
MultiIndex([( 0,
                                    'Vinay ', 50.0, 'Sumeg ', 40.0,
                                   'Naina', 30.0,
'Adarsh', 30.0,
'Harsh', nan,
'Aparna', nan,
                                                                      'Naina', 30.0,
'Adarsh', 30.0,
'Harsh', nan,
'Aparna', nan,
                                                                                                         'Naina', 30.0, ...),
'Adarsh', 30.0, ...),
'Harsh', nan, ...),
'Aparna', nan, ...),
                      (8,
                      (9,
                                    'Yogita', nan,
                                                                      'Yogita', nan,
                                                                                                         'Yogita', nan, ...),
                      (10,
                   (10, Yogita, nan, Yogita, nan, Yogita, nan, (11, 'sukaina', nan, 'sukaina', nan, 'sukaina', nan, names=[None, 'Name', 'Duration_x', 'Name', 'Duration_x', 'Name Time of joining Duration_y Duration_y
                                                                                                      'sukaina', nan, ...)],
puration_x', 'Name', 'Duration_x', 'Name', 'Duration_x'])
                                                 8 8.000000
count
unique
                       12
                                                                          NaN
              Akansha
                                        10:00:00
                                                                          NaN
                                                                                          10:00:00
                                                                                                                             NaN
top
                                                                          NaN
freq
                        1
                                                       1
                                                                                                          1
                                                                                                                             NaN
                                                    NaN 40.000000
                                                                                                       NaN 37,142857
mean
                     NaN
                                                                9.258201
std
                     NaN
                                                    NaN
                                                                                                       NaN
                                                                                                                   7,559289
min
                     NaN
                                                    NaN
                                                               30.000000
                                                                                                       NaN
                                                                                                                 30,000000
25%
                     NaN
                                                    NaN
                                                               30.000000
                                                                                                       NaN
                                                                                                                 30.000000
50%
                                                    NaN
                                                               40.000000
                                                                                                       NaN
                                                                                                                  40.000000
75%
                                                    NaN
                                                               50.000000
                                                                                                       NaN
                                                                                                                  40.000000
                     NaN
                                                    NaN
                                                               50.000000
                                                                                                       NaN
                                                                                                                  50.000000
max
```

5. Taking Iris data, plot the following with proper legend and axis labels: (Download IRIS data from: https://archive.ics.uci.edu/ml/datasets/iris or import it from sklearn.datasets)

Solution:

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
iris = sns.load_dataset('iris')
iris

Oι

Output:

ut[40]:						
		sepal_length	sepal_width	petal_length	petal_width	species
	0	5.1	3.5	1.4	0.2	setosa
	1	4.9	3.0	1.4	0.2	setosa
	2	4.7	3.2	1.3	0.2	setosa
	3	4.6	3.1	1.5	0.2	setosa
	4	5.0	3.6	1.4	0.2	setosa
	145	6.7	3.0	5.2	2.3	virginica
	146	6.3	2.5	5.0	1.9	virginica
	147	6.5	3.0	5.2	2.0	virginica
	148	6.2	3.4	5.4	2.3	virginica
	149	5.9	3.0	5.1	1.8	virginica

150 rows × 5 columns

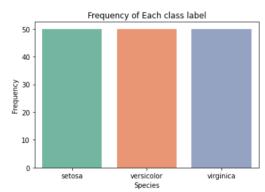
a.) Plot bar chart to show the frequency of each class label in the data.

Solution:

```
sns.countplot(x='species',data=iris,palette='Set2')
plt.xlabel('Species')
plt.ylabel('Frequency')
plt.title('Frequency of Each class label')
```

Output:





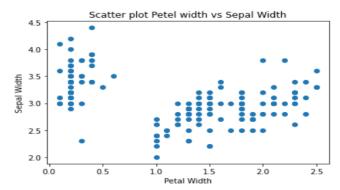
b.) Draw a scatter plot for Petal width vs sepal width.

Solution:

```
plt.scatter(x='petal_width',y='sepal_width',data=iris)
plt.xlabel('Petal Width')
plt.ylabel('Sepal Width')
plt.title("Scatter plot Petel width vs Sepal Width")
```

Output:

Out[42]: Text(0.5, 1.0, 'Scatter plot Petel width vs Sepal Width')

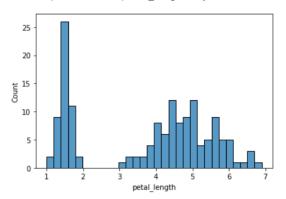


c.) Plot density distribution for feature petal length.

Solution:

sns.histplot(iris['petal_length'],kde=False,bins=30)

Out[43]: <AxesSubplot:xlabel='petal_length', ylabel='Count'>



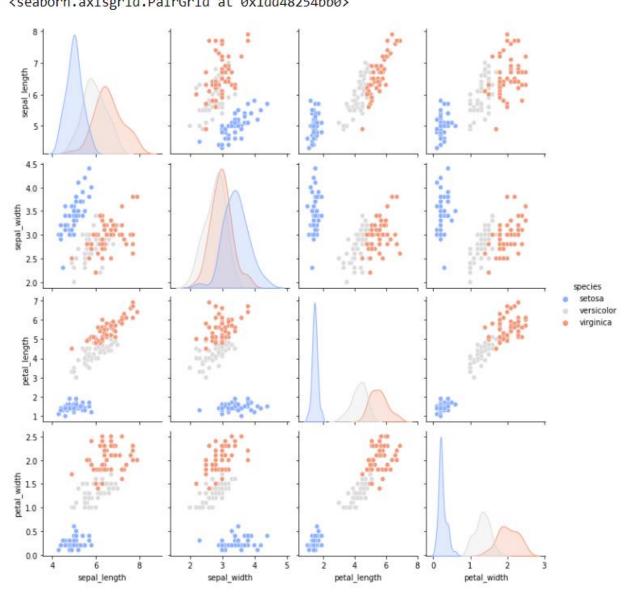
d.) Use a pair plot to show pairwise bivariate distribution in the Iris Dataset.

Solution:

sns.pairplot(iris,hue='species',palette='coolwarm')

Output:

Out[44]: <seaborn.axisgrid.PairGrid at 0x1dd48254bb0>



- 6. Consider any sales training/ weather forecasting dataset
- > I take retail sales forecasting dataset from Kaggle Retail Sales Forecasting | Kaggle ...

Solution:

```
import pandas as pd
import numpy as np
data = pd.read_csv(r"C:\Users\SUKAINA\Downloads\archive (3)\mock_kaggle.csv")
df = pd.DataFrame(data)
df
```

Output:

Out[93]:

	date	venda	estoque	preco
0	01-01-2014	0	4972	1.29
1	02-01-2014	70	4902	1.29
2	03-01-2014	59	4843	1.29
3	04-01-2014	93	4750	1.29
4	05-01-2014	96	4654	1.29
932	27-07-2016	98	3179	2.39
933	28-07-2016	108	3071	2.39
934	29-07-2016	128	4095	2.39
935	30-07-2016	270	3825	2.39
936	31-07-2016	183	3642	2.39

937 rows × 4 columns

a.) Compute mean of a series grouped by another series

Solution:

```
mean_series = df.groupby(['date']).mean()
mean_series.head()
```

Output:

Out[154]:

	venda	estoque	preco
date			
01-01-2014	0.0	4972.0	1.29
01-01-2015	0.0	542.0	1.29
01-01-2016	0.0	595.0	1.39
01-02-2014	369.0	2145.0	0.99
01-02-2015	88.0	197.0	1.29

b.) Fill an intermittent time series to replace all missing dates with values of previous non-missing date.

```
df = df.set_index('date')
```

```
# to_datetime() method converts string
# format to a DateTime object

df.index = pd.to_datetime(df.index)
# dates which are not in the sequence
# are returned
print(pd.date_range (start="2014-1-1", end="2015-12-31").difference(df.index))

df.sort_values(['date','venda','estoque','preco']).groupby('date').ffill()
```

Out[155]:

	venda	estoque	preco
date			
2014-01-01	0	4972	1.29
2014-01-02	369	2145	0.99
2014-01-03	94	6237	1.09
2014-01-04	62	3164	1.09
2014-01-05	129	1263	1.29
2016-12-03	67	1712	2.19
2016-12-04	62	1955	2.59
2016-12-05	134	213	1.89
2016-12-06	317	1870	1.66
2016-12-07	164	1967	1.89

937 rows × 3 columns

c.) Perform appropriate year-month string to dates conversion.

Solution:

```
from datetime import datetime

df['date'] = pd.to_datetime(df['date'])

change_format = df['date'].dt.strftime('%y/%m/%d')

print("After Performing appropriate year-month string to dates conversion: \n",change_format)
```

```
After Performing appropriate year-month string to dates conversion:
0
       14/01/01
1
       14/02/01
      14/03/01
2
3
      14/04/01
4
      14/05/01
      16/07/27
932
      16/07/28
933
934
      16/07/29
935
      16/07/30
936
       16/07/31
Name: date, Length: 937, dtype: object
```

d.) Split a dataset to group by two columns and then sort the aggregated results within the groups.

Solution:

```
df_agg = df.groupby(['date','venda']).agg({'estoque':sum})
result = df_agg['estoque'].groupby(level=0, group_keys=False)
print("\nGroup on 'date', 'venda' and then sort sum of estoque within the groups:\n")
print(result.nlargest())
```

Output:

```
Group on 'date', 'venda' and then sort sum of estoque within the groups:
date
           venda
2014-01-01 0
                    4972
2014-01-02 369
                    2145
2014-01-03 94
                    6237
2014-01-04 62
                    3164
2014-01-05 129
                    1263
2016-12-03 67
                    1712
2016-12-04 62
                    1955
2016-12-05 134
                     213
2016-12-06 317
                    1870
2016-12-07 164
                    1967
Name: estoque, Length: 937, dtype: int64
```

e.) Split a given dataframe into groups with bin counts.

Solution:

```
groups = df.groupby(['date', pd.cut(df.venda, 3)])
result = groups.size().unstack()
print(result)
```

Output:

venda date	(-0.542, 180.667]	(180.667, 361.333]	(361.333, 542.0]
2014-01-01	1	0	0
2014-01-02	0	0	1
2014-01-03	1	0	0
2014-01-04	1	0	0
2014-01-05	1	0	0
2016-12-03	1	0	0
2016-12-04	1	0	0
2016-12-05	1	0	0
2016-12-06	0	1	0
2016-12-07	1	0	0

[937 rows x 3 columns]

7. Consider a data frame containing data about students i.e. name, gender and passing division:

	Name	Birth_Month	Gender	Pass_Division
0	Mudit Chauhan	December	M	III
1	Seema Chopra	January	F	Ш
2	Rani Gupta	March	F	I
3	Aditya Narayan	October	М	I
4	Sanjeev Sahni	February	М	Ш
5	Prakash Kumar	December	М	III
6	Ritu Agarwal	September	F	I
7	Akshay Goel	August	M	I
8	Meeta Kulkarni	July	F	II
9	Preeti Ahuja	November	F	Ш
10	Sunil Das Gupta	April	М	III
11	Sonali Sapre	January	F	I
12	Rashmi Talwar	June	F	III
13	Ashish Dubey	May	М	Ш
14	Kiran Sharma	February	F	Ш
15	Sameer Bansal	October	М	ı

a.) Perform one hot encoding of the last two columns of categorical data using the get_dummies() function.

Solution:

```
import pandas as pd
#categorical data
categorical_cols = ['Gender','pass_division']
df = pd.get_dummies(data, columns = categorical_cols)
```

b.) Sort this data frame on the "Birth Month" column (i.e. January to December). Hint: Convert Month to Categorical.

8. Consider the following data frame containing a family name, gender of the family member and her/his monthly income in each record.

Solution:

Name	Gender	MonthlyIncome (Rs.)
Shah	Male	114000.00
Vats	Male	65000.00
Vats	Female	43150.00
Kumar	Female	69500.00
Vats	Female	155000.00
Kumar	Male	103000.00
Shah	Male	55000.00
Shah	Female	112400.00
Kumar	Female	81030.00
Vats	Male	71900.00

Write a program in Python using Pandas to perform the following:

a.) Calculate and display familywise gross monthly income

Solution:

```
import pandas as pd

df = pd.DataFrame({

'Name': ['Shah','Vats','Vats','Kumar','Vats','Kumar','Shah','Shah','Kumar','Shah'],

'Gender': ['Male','Male','Female','Female','Female','Male','Male','Female','Fem ale','Male'],

'Monthly_Income ': [114000,65000,43150,69500,155000,103000,55000,112400,81030,71900]})

df

gross_salary = df.groupby(by=['Name'], as_index=False)['Monthly_Income (Rs)'].sum()

print (gross_salary)
```

Output:

```
Family wise monthly gross income
Name Monthly_Income (Rs)

Kumar 253530
Shah 353300
Vats 263150
```

b.) Calculate and display the member with the highest monthly income in a family.

```
max\_salary = data.groupby(by=['Name','Gender']).apply(lambda x : x[x['MonthlyIncome'] == x['MonthlyIncome'].max()])
max\_salary
```

```
s = max(max_salary['MonthlyIncome'])
res = max_salary[max_salary['MonthlyIncome'] == s ]
print("the member with the highest monthly income in a family :\n ",res)
```

```
the member with the highest monthly income in a family:

Name Gender MonthlyIncome

Name Gender

Vats Female 4 Vats Female 155000.0
```

c.) Calculate and display monthly income of all members with income greater than Rs. 60000.00.

Solution:

```
greater_income = data[data['MonthlyIncome'] > 60000.00]

print(" monthly income of all members with income greater than Rs. 60000.00: \n",greater_income)
```

Output:

monthly income of all members with income greater than Rs. 60000.00:

Name	Gender	MonthlyIncome
shah	Male	114000.0
Vats	Male	65000.0
Kumar	Female	69500.0
Vats	Female	155000.0
Kumar	Male	103000.0
shah	Female	112400.0
Kumar	Female	81030.0
Vats	Male	71900.0
	shah Vats Kumar Vats Kumar shah Kumar	shah Male Vats Male Kumar Female Vats Female Kumar Male shah Female Kumar Female

d.) Calculate and display the average monthly income of the female members in the Shah family.

Solution:

```
average = data[(data['Name']== 'shah') & (data['Gender']=='Female')].mean()
print("average monthly income of the female members in the Shah family: \n ",average)
```

```
average monthly income of the female members in the Shah family:
MonthlyIncome 112400.0
dtype: float64
```