DATA ANALYSIS AND VISUALIZATION PRACTICAL FILE

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SCIENCE
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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 x n integer elements, also print the shape, type and data type of the array and then reshape it into nx 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))
```

```
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.

Solution:

```
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

Output:

```
before replacing 10% values by null the dataframe becomes :
      B C
0 11 28 61
1 84 80 18
2 94 82 25
     85 70
3 77
4 99
after replacing 10% values by null the dataframe becomes :
           В
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
```

a) Identify and count missing values in a dataframe.

Solution:

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

```
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:

```
A C
0 11.0 61.0
1 84.0 NaN
2 94.0 25.0
3 77.0 70.0
4 99.0 NaN
```

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

Solution:

```
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:
            В
  11.0 28.0 61.0
  84.0 80.0
              NaN
2 94.0 82.0
              25.0
        NaN 70.0
  77.0
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:

```
A B C

35 0.0 81.0 3.0

9 0.0 41.0 27.0

40 8.0 65.0 73.0

31 8.0 74.0 40.0

19 10.0 22.0 69.0
```

e.) Remove all duplicates from the first column.

Solution:

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

Output:

```
A B 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)
```

```
0
   NaN
1
   NaN
2
   NaN
3
   NaN
   NaN
Name: A, dtype: float64
     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
```

h.) Discretize second column and create 5 bins

Solution:

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

Output:

```
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:

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

| Da | y1 data : | | |
|----|-----------|-------------------|----------|
| | Name | Time of joining | Duration |
| 0 | Akansha | 10:00:00 | 30 |
| 1 | Sakshi | 10:05:00 | 50 |
| 2 | Esha | 10:08:01 | 40 |
| 3 | Himanshi | 10:05:20 | 50 |
| 4 | Vinay | 10:06:10 | 50 |
| 5 | Sumeg | 10:07:00 | 40 |
| 6 | Naina | 10:12:00 | 30 |
| 7 | Adarsh | 10:22:00 | 30 |
| _ | - 11 | | |
| Da | y2 data : | | |
| | Name T | ime of joining Du | ration |
| 0 | Akansha | 10:00:00 | 30 |
| 1 | Harsh | 10:09:12 | 40 |
| 2 | Aparna | 10:09:15 | 30 |
| 3 | Sakshi | 10:11:10 | 40 |
| 4 | Esha | 10:11:20 | 50 |
| 5 | Yogita | 10:15:00 | 40 |
| 6 | sukaina | 10:21:00 | 30 |

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

Solution:

pd.merge(d1_df,d2_df,how='inner',on='Name')

Output:

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.

Solution:

```
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
```

```
'Akansha', 30.0,
'Sakshi', 50.0,
                                                                              'Akansha', 30.0, ...),
'Sakshi', 50.0, ...),
MultiIndex([( 0,
                                                    'Akansha', 30.0,
                (1,
                                                      'Sakshi', 50.0,
                       'Esha', 40.0,
'Himanshi ', 50.0,
                                                 'Esha', 40.0, 'Esha', 40.0, ...),
'Himanshi ', 50.0, 'Himanshi ', 50.0, ...),
                (2,
                (3,
                                    , 50.0,
                                                     'Vinay', 50.0,
'Sumeg', 40.0,
'Naina', 30.0,
'Adarsh', nan,
'Anarna', nan
                                                                                'Vinay ',
                            'Vinay '
                                                                                            50.0, ...),
                (4,
                            'Sumeg
                                      , 40.0,
                                                                                'Sumeg
                (5,
                                                                                          ', 40.0, ...),
                             'Naina',
                                                                                 'Naina', 30.0,
                (6,
                                         30.0,
                           'Adarsh', 30.0,
'Harsh', nan,
                                                                                'Adarsh', 30.0, ...),
'Harsh', nan, ...),
                (7,
                (8,
                            'Aparna',
                                                      'Aparna', nan,
                                                                                'Aparna',
                (9,
                                          nan,
                                                                                              nan, ...),
                                                     'Yogita', nan,
                           'Yogita', nan,
                                                                               'Yogita', nan, ...),
                (10,
              (10, Yogita, Man, Yogita, Man, 'Sukaina', nan, 'Sukaina', nan, names=[None, 'Name', 'Duration_x', 'Name', 'Duration_x', 'Name Time of joining Duration_x Time of joining Duration_y
                                                                              'sukaina', nan, ...)],
ration_x', 'Name', 'Duration_x', 'Name', 'Duration_x'])
count
                 12
                                                 8.000000
                                                                                       7.000000
unique
                 12
                                          8
                                                        NaN
                                                                                              NaN
top
          Akansha
                                10:00:00
                                                        NaN
                                                                       10:00:00
                                                                                              NaN
                                                        NaN
freq
                NaN
                                       NaN
                                                40.000000
                                                                              NaN
                                                                                      37.142857
mean
                                                 9.258201
                                                                                       7,559289
std
                NaN
                                       NaN
                                                                             NaN
                NaN
                                       NaN
                                                30,000000
                                                                              NaN
                                                                                      30,000000
min
                                                                                      30.000000
25%
                                                30.000000
                NaN
                                       NaN
                                                                             NaN
50%
                NaN
                                       NaN
                                                40.000000
                                                                              NaN
                                                                                      40.000000
75%
                NaN
                                       NaN
                                                50.000000
                                                                              NaN
                                                                                      40.000000
                                       NaN
                                                50.000000
                                                                                      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

Output:

| 0 | $\Gamma \Lambda C$ | ٠. ٦٠ |
|---|------------------------|-------|
| U | 148 | , , , |

| | 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

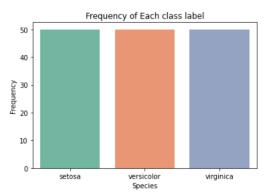
a.) Plot bar chart to show the frequency of each class label in the data.

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')

Scatter plot Petel width vs Sepal Width

4.5

3.5

2.0

2.5

2.0

0.0

0.5

1.0

1.5

2.0

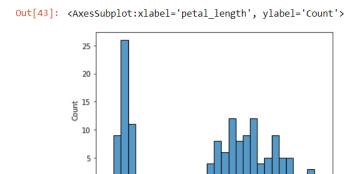
2.5
```

c.) Plot density distribution for feature petal length.

Solution:

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

Output:

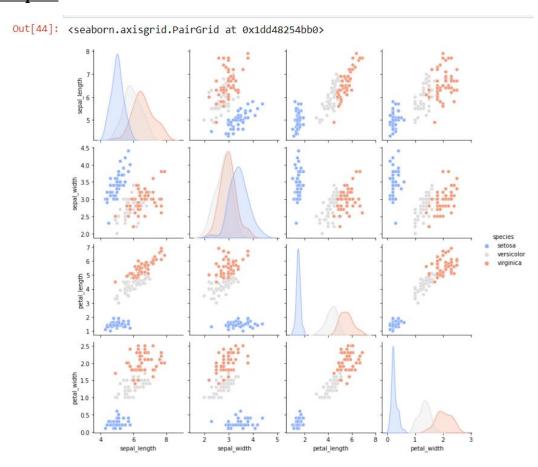


d.) Use a pair plot to show pairwise bivariate distribution in the Iris Dataset.

Solution:

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

Output:



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\MOHD SHOHEL\Downloads\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.

Solution:

```
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()
```

Output:

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.

```
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)

Output:

```
After Performing appropriate year-month string to dates conversion:
       14/01/01
1
       14/02/01
2
       14/03/01
3
       14/04/01
      14/05/01
        . . .
932
      16/07/27
933
      16/07/28
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.

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

| venda | (-0.542, 180.667] | (180.667, 361.333] | (361.333, 542.0] |
|------------|-------------------|--------------------|------------------|
| date | | | |
| 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 | М | III |
| 1 | Seema Chopra | January | F | II |
| 2 | Rani Gupta | March | F | I |
| 3 | Aditya Narayan | October | M | I |
| 4 | Sanjeev Sahni | February | M | II |
| 5 | Prakash Kumar | December | M | III |
| 6 | Ritu Agarwal | September | F | I |
| 7 | Akshay Goel | August | M | I |
| 8 | Meeta Kulkarni | July | F | Ш |
| 9 | Preeti Ahuja | November | F | Ш |
| 10 | Sunil Das Gupta | April | M | III |
| 11 | Sonali Sapre | January | F | I |
| 12 | Rashmi Talwar | June | F | III |
| 13 | Ashish Dubey | May | M | Ш |
| 14 | Kiran Sharma | February | F | Ш |
| 15 | Sameer Bansal | October | М | I |

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

```
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.

Solution:

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

```
import pandas as pd

df = pd.DataFrame({
    'Name': ['Shah','Vats','Kumar','Vats','Kumar','Shah','Kumar','Shah'],
```

```
'Gender': ['Male','Male','Female','Female','Female','Male','Male','Female','Female','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)
```

```
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.

Solution:

```
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)
```

Output:

```
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)
```

```
monthly income of all members with income greater than Rs. 60000.00:
    Name Gender MonthlyIncome
   shah
          Male
                   114000.0
  Vats Male
                    65000.0
3 Kumar Female
                    69500.0
4 Vats Female
5 Kumar Male
                   155000.0
                   103000.0
  shah Female
                   112400.0
8 Kumar Female
                    81030.0
9 Vats Male
                    71900.0
```

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
```