```
Lab Assignment 3

AIM:-Descriptive Statistics - Measures of Central Tendency and
variability
Perform the following operations on any open source dataset (e.g., data.
csv)
1. Provide summary statistics (mean, median, minimum, maximum,
standard deviation) for
a dataset (age, income etc.) with numeric variables grouped by one of
the qualitative
(categorical) variable. For example, if your categorical variable is
age groups and
quantitative variable is income, then provide summary statistics of
income grouped by the
age groups. Create a list that contains a numeric value for each
response to the categorical
variable.
2. Write a Python program to display some basic statistical details
like percentile, mean,
standard deviation etc. of the species of 'Iris-setosa', 'Iris-
versicolor' and 'Iris-versicolor'
of iris.csv dataset.
Provide the codes with outputs and explain everything that you do in
this step.

import pandas as pd
file_path=r"C:\Users\shrey\OneDrive\Desktop\MALL_CUSTOMER.csv"
df=pd.read_csv(file_path)
df.head()
```

	CustomerID	Age	Annual Income($)	Spending Score	Gender
0	1	33.0	186.0	56.0	male
1	2	18.0	127.0	26.0	male
2	3	25.0	132.0	37.0	male
3	4	25.0	100.0	63.0	male
4	5	29.0	104.0	42.0	male

```
df
```

	CustomerID	Age	Annual Income($)	Spending Score	Gender
0	1	33.0	186.0	56.0	male
1	2	18.0	127.0	26.0	male
2	3	25.0	132.0	37.0	male
3	4	25.0	100.0	63.0	male
4	5	29.0	104.0	42.0	male
..	...	...	...	...	...
195	196	25.0	161.0	93.0	male
196	197	25.0	189.0	40.0	male
197	198	33.0	125.0	5.0	male
198	199	19.0	108.0	14.0	male

```
199          200  34.0                112.0                36.0    male

[200 rows x 5 columns]

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
 #   Column             Non-Null Count  Dtype
---  ------             --------------  -----
 0   CustomerID         200 non-null    int64
 1   Age                184 non-null    float64
 2   Annual Income($)   184 non-null    float64
 3   Spending Score     185 non-null    float64
 4   Gender             200 non-null    object
dtypes: float64(3), int64(1), object(1)
memory usage: 7.9+ KB

df.head

<bound method NDFrame.head of       CustomerID   Age  Annual Income($)
Spending Score Gender
0                1  33.0              186.0                56.0    male
1                2  18.0              127.0                26.0    male
2                3  25.0              132.0                37.0    male
3                4  25.0              100.0                63.0    male
4                5  29.0              104.0                42.0    male
..             ...   ...                ...                 ...     ...
195            196  25.0              161.0                93.0    male
196            197  25.0              189.0                40.0    male
197            198  33.0              125.0                 5.0    male
198            199  19.0              108.0                14.0    male
199            200  34.0              112.0                36.0    male

[200 rows x 5 columns]>

df.tail

<bound method NDFrame.tail of       CustomerID   Age  Annual Income($)
Spending Score Gender
0                1  33.0              186.0                56.0    male
1                2  18.0              127.0                26.0    male
2                3  25.0              132.0                37.0    male
3                4  25.0              100.0                63.0    male
4                5  29.0              104.0                42.0    male
..             ...   ...                ...                 ...     ...
195            196  25.0              161.0       93.0             male
196            197  25.0              189.0                40.0    male
197            198  33.0              125.0                 5.0    male
198     199         19.0              108.0                 14.0  male
```

```
199          200  34.0                 112.0                 36.0   male

[200 rows x 5 columns]>
```

```
df.describe()
```

```
        CustomerID         Age  Annual Income($)  Spending Score
count   200.000000  184.000000        184.000000      185.000000
mean    100.500000   26.342391        148.244565       49.470270
std      57.879185    5.133959         29.339728       28.099985
min       1.000000   18.000000        100.000000        1.000000
25%      50.750000   22.000000        122.000000       26.000000
50%     100.500000   26.000000        150.000000       47.000000
75%     150.250000   30.000000        170.250000       72.000000
max     200.000000   35.000000        200.000000      100.000000
```

```
df.Age.mean()
```

```
26.342391304347824
```

```
df.Age.mode()
```

```
0    30.0
Name: Age, dtype: float64
```

```
df.Age.median()
```

```
26.0
```

```
df.groupby(['Age']).count()
```

```
      CustomerID  Annual  Income($)  Spending  Score  Gender
Age
18.0          15                  14                13      15
19.0          12                  11                11      12
20.0           3                   3                 3       3
21.0           8                   8                 7       8
22.0          13                  12                12      13
23.0           9                   7                 9       9
24.0           5                   5                 5       5
25.0          16                  15                16      16
26.0          14                  14                12      14
27.0          12                   9                12      12
28.0           6                   5                 6       6
29.0          10                  10                 9      10
30.0          18                  17                16      18
31.0          10                   7                10      10
32.0           8                   7                 7       8
33.0           5                   5                 4       5
34.0           9                   9                 7       9
35.0          11                  10                10      11
```

```
df.groupby(['Gender']).count()
```

```
        CustomerID  Age  Annual Income($)  Spending Score
Gender
female          20   20                 8              20
male           180  164               176             165
```

```
df.Age.std()
```

```
5.133959234335101
```

```
df[['Age' , 'Annual Income($)', 'Spending Score']].mean()
```

```
Age                    26.3423
Annual                      91
Income($)              148.244
dtype: float64
```

```
df[['Age' , 'Annual Income($)', 'Spending Score']].mode()
```

```
    Age  Annual Income($)  Spending Score
0  30.0             170.0            26.0
```

```
df[['Age' , 'Annual Income($)', 'Spending Score']].median()
```

```
Age                    26.
Annual                  0
Income($)              15

dtype: float64
```

```
df[['Age' , 'Annual Income($)', 'Spending Score']].max()
```

```
Age                    35.
Annual                  0
Income($)              20

dtype: float64
```

```
df[['Age' , 'Annual Income($)', 'Spending Score']].std()
```

```
Age                    5.1339
Annual                     59
Income($)              29.339

dtype: float64
```

```
df2 = df.groupby('Gender')
```

```
df
```

```
     CustomerID   Age  Annual Income($)  Spending Score Gender
0             1  33.0             186.0            56.0   male
1             2  18.0             127.0            26.0   male
2             3  25.0             132.0            37.0   male
```

```
3                4   25.0                  100.0                  63.0    male
4                5   29.0                  104.0                  42.0    male
..             ...    ...                    ...                   ...     ...
195            196   25.0                  161.0                  93.0    male
196            197   25.0                  189.0                  40.0    male
197            198   33.0                  125.0                   5.0    male
198            199   19.0                  108.0                  14.0    male
199            200   34.0                  112.0                  36.0    male

[200 rows x 5 columns]
```

```
for Gender, Gender_f in df2:
    print(Gender)
    print(Gender_f)
```

```
female
     CustomerID   Age  Annual Income($)  Spending Score  Gender
5               6  35.0               174.0              68.0  female
6               7  32.0               114.0              71.0  female
7               8  32.0               127.0              49.0  female
8               9  28.0                 NaN              19.0  female
9              10  30.0                 NaN              58.0  female
10             11  35.0                 NaN              34.0  female
11             12  32.0                 NaN              17.0  female
12             13  27.0                 NaN              18.0  female
13             14  27.0                 NaN              26.0  female
14             15  31.0                 NaN              65.0  female
15             16  22.0                 NaN              39.0  female
16             17  25.0                 NaN              65.0  female
17             18  19.0                 NaN              89.0  female
18             19  31.0                 NaN              76.0  female
22             23  23.0                 NaN              93.0  female
28             29  29.0               198.0               4.0  female
33             34  31.0               176.0              30.0  female
56             57  24.0               107.0              74.0  female
94             95  28.0               106.0               9.0  female
172           173  25.0               152.0              93.0  female
male
     CustomerID   Age  Annual Income($)  Spending Score Gender
0               1  33.0               186.0              56.0   male
1               2  18.0               127.0              26.0   male
2               3  25.0               132.0              37.0   male
3               4  25.0               100.0              63.0   male
4               5  29.0               104.0              42.0   male
..            ...   ...                 ...               ...    ...
195           196  25.0               161.0              93.0   male
196           197  25.0               189.0              40.0   male
197           198  33.0               125.0               5.0   male
198           199  19.0               108.0              14.0   male
```

199	200	34.0	112.0	36.0	male

```
[180 rows x 5 columns]
```

```
df2.get_group('male')
```

```
     CustomerID   Age  Annual  Income($)  Spending  Score  Gender
0             1  33.0                186.0              56.0    male
1             2  18.0                127.0              26.0    male
2             3  25.0                132.0              37.0    male
3             4  25.0                100.0              63.0    male
4             5  29.0                104.0              42.0    male
..          ...   ...                  ...               ...     ...
195         196  25.0                161.0              93.0    male
196         197  25.0                189.0              40.0    male
197         198  33.0                125.0               5.0    male
198         199  19.0                108.0              14.0    male
199         200  34.0                112.0              36.0    male
```

```
[180 rows x 5 columns]
```

```
df2.get_group('female')
```

```
     CustomerID   Age  Annual  Income($)  Spending  Score  Gender
5             6  35.0                174.0              68.0  female
6             7  32.0                114.0              71.0  female
7             8  32.0                127.0              49.0  female
8             9  28.0                  NaN              19.0  female
9            10  30.0                  NaN              58.0  female
10           11  35.0                  NaN              34.0  female
11           12  32.0                  NaN              17.0  female
12           13  27.0                  NaN              18.0  female
13           14  27.0                  NaN              26.0  female
14           15  31.0                  NaN              65.0  female
15           16  22.0                  NaN              39.0  female
16           17  25.0                  NaN              65.0  female
17           18  19.0                  NaN              89.0  female
18           19  31.0                  NaN              76.0  female
22           23  23.0                  NaN              93.0  female
28           29  29.0                198.0               4.0  female
33           34  31.0                176.0              30.0  female
56           57  24.0                107.0              74.0  female
94           95  28.0                106.0               9.0  female
172         173  25.0                152.0              93.0  female
```

```
df2[['Age' , 'Annual Income($)', 'Spending Score']].median()
```

```
          Age  Annual Income($)  Spending Score
Gender
female   28.5             139.5            53.5
male     26.0             150.0            47.0
```

```
df2[['Age' , 'Annual Income($)', 'Spending Score']].mean()
```

```
                Age  Annual Income($)  Spending Score
Gender
female  28.300000         144.250000        49.850000
male    26.103659         148.426136        49.424242
```

```
df2[['Age' , 'Annual Income($)', 'Spending Score']].max()
```

```
           Age  Annual Income($)  Spending Score
Gender
female  35.0                198.0              93.0
male    35.0                200.0             100.0
```

```
df2[['Age' , 'Annual Income($)', 'Spending Score']].min()
```

```
           Age  Annual Income($)  Spending Score
Gender
female  19.0                106.0               4.0
male    18.0                100.0               1.0
```

```
df2[['Age' , 'Annual Income($)', 'Spending Score']].std()
```

```
                Age  Annual Income($)  Spending Score
Gender
female  4.317650         35.668113        28.995962
male    5.185656         29.129371        28.079841
```

```
url =
"https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.d
ata"

df3 = pd.read_csv(url)

df3
```

```
     5.1  3.5  1.4  0.2      Iris-setosa
0    4.9  3.0  1.4  0.2      Iris-setosa
1    4.7  3.2  1.3  0.2      Iris-setosa
2    4.6  3.1  1.5  0.2      Iris-setosa
3    5.0  3.6  1.4  0.2      Iris-setosa
4    5.4  3.9  1.7  0.4      Iris-setosa
..   ...  ...  ...  ...               ...
144  6.7  3.0  5.2  2.3   Iris-virginica
145  6.3  2.5  5.0  1.9   Iris-virginica
146  6.5  3.0  5.2  2.0   Iris-virginica
147  6.2  3.4  5.4  2.3   Iris-virginica
148  5.9  3.0  5.1  1.8   Iris-virginica

[149 rows x 5 columns]
```

```
df3
```

```
        A    B   C    D                 E
0     4.9  3.0  1.4  0.2     Iris-setosa
1     4.7  3.2  1.3  0.2     Iris-setosa
2     4.6  3.1  1.5  0.2     Iris-setosa
3     5.0  3.6  1.4  0.2     Iris-setosa
4     5.4  3.9  1.7  0.4     Iris-setosa
..    ...  ...  ...  ...              ...
144   6.7  3.0  5.2  2.3  Iris-virginica
145   6.3  2.5  5.0  1.9  Iris-virginica
146   6.5  3.0  5.2  2.0  Iris-virginica
147   6.2  3.4  5.4  2.3  Iris-virginica
148   5.9  3.0  5.1  1.8  Iris-virginica

[149 rows x 5 columns]
```

```
df4.get_group("Iris-setosa")
```

```
      A    B   C    D            E
0   4.9  3.0  1.4  0.2  Iris-setosa
1   4.7  3.2  1.3  0.2  Iris-setosa
2   4.6  3.1  1.5  0.2  Iris-setosa
3   5.0  3.6  1.4  0.2  Iris-setosa
4   5.4  3.9  1.7  0.4  Iris-setosa
5   4.6  3.4  1.4  0.3  Iris-setosa
6   5.0  3.4  1.5  0.2  Iris-setosa
7   4.4  2.9  1.4  0.2  Iris-setosa
8   4.9  3.1  1.5  0.1  Iris-setosa
9   5.4  3.7  1.5  0.2  Iris-setosa
10  4.8  3.4  1.6  0.2  Iris-setosa
11  4.8  3.0  1.4  0.1  Iris-setosa
12  4.3  3.0  1.1  0.1  Iris-setosa
13  5.8  4.0  1.2  0.2  Iris-setosa
14  5.7  4.4  1.5  0.4  Iris-setosa
15  5.4  3.9  1.3  0.4  Iris-setosa
16  5.1  3.5  1.4  0.3  Iris-setosa
17  5.7  3.8  1.7  0.3  Iris-setosa
18  5.1  3.8  1.5  0.3  Iris-setosa
19  5.4  3.4  1.7  0.2  Iris-setosa
20  5.1  3.7  1.5  0.4  Iris-setosa
21  4.6  3.6  1.0  0.2  Iris-setosa
22  5.1  3.3  1.7  0.5  Iris-setosa
23  4.8  3.4  1.9  0.2  Iris-setosa
24  5.0  3.0  1.6  0.2  Iris-setosa
25  5.0  3.4  1.6  0.4  Iris-setosa
26  5.2  3.5  1.5  0.2  Iris-setosa
27  5.2  3.4  1.4  0.2  Iris-setosa
28  4.7  3.2  1.6  0.2  Iris-setosa
29  4.8  3.1  1.6  0.2  Iris-setosa
30  5.4  3.4  1.5  0.4  Iris-setosa
31  5.2  4.1  1.5  0.1  Iris-setosa
```

```
32   5.5  4.2  1.4  0.2  Iris-setosa
33   4.9  3.1  1.5  0.1  Iris-setosa
34   5.0  3.2  1.2  0.2  Iris-setosa
35   5.5  3.5  1.3  0.2  Iris-setosa
36   4.9  3.1  1.5  0.1  Iris-setosa
37   4.4  3.0  1.3  0.2  Iris-setosa
38   5.1  3.4  1.5  0.2  Iris-setosa
39   5.0  3.5  1.3  0.3  Iris-setosa
40   4.5  2.3  1.3  0.3  Iris-setosa
41   4.4  3.2  1.3  0.2  Iris-setosa
42   5.0  3.5  1.6  0.6  Iris-setosa
43   5.1  3.8  1.9  0.4  Iris-setosa
44   4.8  3.0  1.4  0.3  Iris-setosa
45   5.1  3.8  1.6  0.2  Iris-setosa
46   4.6  3.2  1.4  0.2  Iris-setosa
47   5.3  3.7  1.5  0.2  Iris-setosa
48   5.0  3.3  1.4  0.2  Iris-setosa
```

```
 df4.get_group("Iris-virginica")
```

```
        A    B   C     D                 E
99    6.3  3.3  6.0  2.5  Iris-virginica
100   5.8  2.7  5.1  1.9  Iris-virginica
101   7.1  3.0  5.9  2.1  Iris-virginica
102   6.3  2.9  5.6  1.8  Iris-virginica
103   6.5  3.0  5.8  2.2  Iris-virginica
104   7.6  3.0  6.6  2.1  Iris-virginica
105   4.9  2.5  4.5  1.7  Iris-virginica
106   7.3  2.9  6.3  1.8  Iris-virginica
107   6.7  2.5  5.8  1.8  Iris-virginica
108   7.2  3.6  6.1  2.5  Iris-virginica
109   6.5  3.2  5.1  2.0  Iris-virginica
110   6.4  2.7  5.3  1.9  Iris-virginica
111   6.8  3.0  5.5  2.1  Iris-virginica
112   5.7  2.5  5.0  2.0  Iris-virginica
113   5.8  2.8  5.1  2.4  Iris-virginica
114   6.4  3.2  5.3  2.3  Iris-virginica
115   6.5  3.0  5.5  1.8  Iris-virginica
116   7.7  3.8  6.7  2.2  Iris-virginica
117   7.7  2.6  6.9  2.3  Iris-virginica
118   6.0  2.2  5.0  1.5  Iris-virginica
119   6.9  3.2  5.7  2.3  Iris-virginica
120   5.6  2.8  4.9  2.0  Iris-virginica
121   7.7  2.8  6.7  2.0  Iris-virginica
122   6.3  2.7  4.9  1.8  Iris-virginica
123   6.7  3.3  5.7  2.1  Iris-virginica
124   7.2  3.2  6.0  1.8  Iris-virginica
125   6.2  2.8  4.8  1.8  Iris-virginica
126   6.1  3.0  4.9  1.8  Iris-virginica
127   6.4  2.8  5.6  2.1  Iris-virginica
```

```
128  7.2  3.0  5.8  1.6  Iris-virginica
129  7.4  2.8  6.1  1.9  Iris-virginica
130  7.9  3.8  6.4  2.0  Iris-virginica
131  6.4  2.8  5.6  2.2  Iris-virginica
132  6.3  2.8  5.1  1.5  Iris-virginica
133  6.1  2.6  5.6  1.4  Iris-virginica
134  7.7  3.0  6.1  2.3  Iris-virginica
135  6.3  3.4  5.6  2.4  Iris-virginica
136  6.4  3.1  5.5  1.8  Iris-virginica
137  6.0  3.0  4.8  1.8  Iris-virginica
138  6.9  3.1  5.4  2.1  Iris-virginica
139  6.7  3.1  5.6  2.4  Iris-virginica
140  6.9  3.1  5.1  2.3  Iris-virginica
141  5.8  2.7  5.1  1.9  Iris-virginica
142  6.8  3.2  5.9  2.3  Iris-virginica
143  6.7  3.3  5.7  2.5  Iris-virginica
144  6.7  3.0  5.2  2.3  Iris-virginica
145  6.3  2.5  5.0  1.9  Iris-virginica
146  6.5  3.0  5.2  2.0  Iris-virginica
147  6.2  3.4  5.4  2.3  Iris-virginica
148  5.9  3.0  5.1  1.8  Iris-virginica
```

```
df4.mean()
```

```
                    A         B         C           D
E
Iris-setosa       5.004082  3.416327  1.465306   0.244898
Iris-versicolor   5.936000  2.770000  4.260000   1.326000
Iris-virginica    6.588000  2.974000  5.552000   2.026000
```

```
df4.std()
```

```
                    A         B         C           D
E
Iris-setosa       0.355879  0.384787  0.175061   0.108130
Iris-versicolor   0.516171  0.313798  0.469911   0.197753
Iris-virginica    0.635880  0.322497  0.551895   0.274650
```

```
 Sushant Jawale -13209
```