

Assignment No 3A

```
import pandas as pd
import numpy as np
import io
import statistics
import seaborn as sns
```

```
from google.colab import files
uploaded=files.upload()
```

<IPython.core.display.HTML object>

Saving Mall_Customers.csv to Mall_Customers.csv

```
df=pd.read_csv(io.BytesIO(uploaded['Mall_Customers.csv']))
```

```
print(df)
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40
..
195	196	Female	35	120	79
196	197	Female	45	126	28
197	198	Male	32	126	74
198	199	Male	32	137	18
199	200	Male	30	137	83

[200 rows x 5 columns]

```
df.describe()
```

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
count	200.000000	200.000000	200.000000	200.000000
mean	100.500000	38.850000	60.560000	50.200000
std	57.879185	13.969007	26.264721	25.823522
min	1.000000	18.000000	15.000000	1.000000
25%	50.750000	28.750000	41.500000	34.750000
50%	100.500000	36.000000	61.500000	50.000000
75%	150.250000	49.000000	78.000000	73.000000
max	200.000000	70.000000	137.000000	99.000000

```
statistics.mode(df['Age'])
```

```
32
```

```
statistics.mean(df['Annual Income (k$)'])
```

```
60.56
```

```
statistics.median(df['Age'])
```

```
36.0
```

```
df[['Age', 'Annual Income (k$)']].groupby('Age').describe()
```

```
Annual Income (k$)
count      mean      std   min   25%   50%   75%
Age
18         4.0  51.250000  14.056434  33.0  44.25  53.5  60.50
19         8.0  57.000000  20.632845  15.0  47.50  63.5  67.25
20         5.0  41.600000  24.815318  16.0  21.00  37.0  61.00
21         5.0  38.800000  19.018412  15.0  30.00  33.0  54.00
22         3.0  31.333333  22.278540  17.0  18.50  20.0  38.50
23         6.0  41.500000  23.441416  16.0  20.75  41.5  60.00
24         4.0  39.250000  16.357975  20.0  33.50  38.5  44.25
25         3.0  57.666667  29.263174  24.0  48.00  72.0  74.50
26         2.0  58.000000   5.656854  54.0  56.00  58.0  60.00
27         6.0  63.166667  18.400181  40.0  49.50  63.5  75.25
28         4.0  85.250000  11.615363  76.0  76.75  82.0  90.50
29         5.0  63.600000  28.866936  28.0  40.00  73.0  79.00
30         7.0  76.142857  39.612408  19.0  56.00  78.0  93.50
31         8.0  48.375000  23.323425  17.0  35.50  41.5  70.50
32        11.0  87.181818  26.809768  48.0  74.00  77.0  100.00
33         3.0  80.333333  35.837597  42.0  64.00  86.0  99.50
34         5.0  79.000000  15.968719  58.0  78.00  78.0  78.00
35         9.0  46.666667  38.632888  18.0  21.00  24.0  74.00
36         6.0  81.000000  22.556595  37.0  85.50  87.0  87.00
37         3.0  65.000000  40.112342  20.0  49.00  78.0  87.50
38         6.0  74.500000  20.462160  54.0  64.75  69.0  76.25
39         3.0  72.666667   4.725816  69.0  70.00  71.0  74.50
40         6.0  61.666667  19.551641  29.0  55.50  64.5  70.50
41         2.0 101.000000   2.828427  99.0 100.00 101.0 102.00
42         2.0  60.000000  36.769553  34.0  47.00  60.0  73.00
43         3.0  65.666667  15.695010  48.0  59.50  71.0  74.50
44         2.0  75.500000   3.535534  73.0  74.25  75.5  76.75
45         3.0  69.333333  50.767444  28.0  41.00  54.0  90.00
46         3.0  59.000000  36.755952  25.0  39.50  54.0  76.00
47         6.0  70.166667  27.694163  43.0  51.75  65.5  76.25
48         5.0  58.200000  13.700365  39.0  54.00  60.0  61.00
49         7.0  51.000000  12.884099  33.0  40.50  54.0  62.00
50         5.0  58.600000  18.420098  40.0  43.00  58.0  67.00
```

51	2.0	55.500000	16.263456	44.0	49.75	55.5	61.25
52	2.0	55.500000	45.961941	23.0	39.25	55.5	71.75
53	2.0	39.500000	9.192388	33.0	36.25	39.5	42.75
54	4.0	59.750000	30.998656	28.0	42.25	55.0	72.50
55	1.0	57.000000	NaN	57.0	57.00	57.0	57.00
56	1.0	79.000000	NaN	79.0	79.00	79.0	79.00
57	2.0	64.500000	14.849242	54.0	59.25	64.5	69.75
58	2.0	54.000000	48.083261	20.0	37.00	54.0	71.00
59	4.0	65.250000	21.792583	43.0	51.25	62.5	76.50
60	3.0	43.333333	11.547005	30.0	40.00	50.0	50.00
63	2.0	56.500000	12.020815	48.0	52.25	56.5	60.75
64	1.0	19.000000	NaN	19.0	19.00	19.0	19.00
65	2.0	50.500000	17.677670	38.0	44.25	50.5	56.75
66	2.0	63.000000	0.000000	63.0	63.00	63.0	63.00
67	4.0	45.500000	18.699376	19.0	40.00	50.5	56.00
68	3.0	56.666667	7.767453	48.0	53.50	59.0	61.00
69	1.0	44.000000	NaN	44.0	44.00	44.0	44.00
70	2.0	47.500000	2.121320	46.0	46.75	47.5	48.25

	max
Age	
18	65.0
19	81.0
20	73.0
21	62.0
22	57.0
23	70.0
24	60.0
25	77.0
26	62.0
27	88.0
28	101.0
29	98.0
30	137.0
31	81.0
32	137.0
33	113.0
34	103.0
35	120.0
36	103.0
37	97.0
38	113.0
39	78.0
40	87.0
41	103.0
42	86.0
43	78.0
44	78.0
45	126.0

```
46    98.0
47   120.0
48    77.0
49    65.0
50    85.0
51    67.0
52    88.0
53    46.0
54   101.0
55    57.0
56    79.0
57    75.0
58    88.0
59    93.0
60    50.0
63    65.0
64    19.0
65    63.0
66    63.0
67    62.0
68    63.0
69    44.0
70    49.0
```

```
my_var=df.corr().round(2)
```

```
sns.heatmap(data=my_var,annot=True)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f4842b09250>
```



```
df.groupby('Age').agg({'Annual Income (k$)': ['mean', 'min', 'max']})
```

```

Annual Income (k$)
      mean min  max
Age
18      51.250000  33  65
19      57.000000  15  81
20      41.600000  16  73
21      38.800000  15  62
22      31.333333  17  57
23      41.500000  16  70
24      39.250000  20  60
25      57.666667  24  77
26      58.000000  54  62
27      63.166667  40  88
28      85.250000  76 101
29      63.600000  28  98
30      76.142857  19 137
31      48.375000  17  81
32      87.181818  48 137
33      80.333333  42 113
34      79.000000  58 103
35      46.666667  18 120
36      81.000000  37 103

```

37	65.000000	20	97
38	74.500000	54	113
39	72.666667	69	78
40	61.666667	29	87
41	101.000000	99	103
42	60.000000	34	86
43	65.666667	48	78
44	75.500000	73	78
45	69.333333	28	126
46	59.000000	25	98
47	70.166667	43	120
48	58.200000	39	77
49	51.000000	33	65
50	58.600000	40	85
51	55.500000	44	67
52	55.500000	23	88
53	39.500000	33	46
54	59.750000	28	101
55	57.000000	57	57
56	79.000000	79	79
57	64.500000	54	75
58	54.000000	20	88
59	65.250000	43	93
60	43.333333	30	50
63	56.500000	48	65
64	19.000000	19	19
65	50.500000	38	63
66	63.000000	63	63
67	45.500000	19	62
68	56.666667	48	63
69	44.000000	44	44
70	47.500000	46	49