

import necessary packages

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [2]: # Load the file
```

```
In [4]: f = pd.read_csv(r"D:\Data Science with AI\4th, 5th dec 2023 Inferential stats\Descriptive stats _PRA
```

A horizontal scrollbar is visible below the code input area, indicating that the text is truncated. The scrollbar has a dark grey track and a lighter grey slider.

```
In [5]: income_df
```

Out[5]:

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_HH_Income	Highest_Qualified_M
0	5000	8000	3	2000	64200	Under-Gi
1	6000	7000	2	3000	79920	I
2	10000	4500	2	0	112800	Under-Gi
3	10000	2000	1	0	97200	I
4	12500	12000	2	3000	147000	Gi
5	14000	8000	2	0	196560	Gi
6	15000	16000	3	35000	167400	Post-Gi
7	18000	20000	5	8000	216000	Gi
8	19000	9000	2	0	218880	Under-Gi
9	20000	9000	4	0	220800	Under-Gi
10	20000	18000	4	8000	278400	Under-Gi
11	22000	25000	6	12000	279840	I
12	23400	5000	3	0	292032	I
13	24000	10500	6	0	316800	Gi
14	24000	10000	4	0	244800	Gi
15	25000	12300	3	0	246000	Gi
16	25000	20000	3	3500	261000	Gi
17	25000	10000	6	0	258000	Under-Gi
18	29000	6600	2	2000	348000	Gi
19	30000	13000	4	0	385200	Gi
20	30500	25000	5	5000	351360	Under-Gi
21	32000	15000	4	0	445440	Profe
22	34000	19000	6	0	330480	Profe
23	34000	25000	3	4000	469200	Profe
24	35000	12000	3	0	466200	Gi
25	35000	25000	4	0	449400	Profe
26	39000	8000	4	0	556920	Under-Gi
27	40000	10000	4	0	412800	Under-Gi
28	42000	15000	4	0	488880	Gi
29	43000	12000	4	0	619200	Gi
30	45000	25000	6	0	523800	Gi
31	45000	40000	6	3500	507600	Profe
32	45000	10000	2	1000	437400	Post-Gi
33	45000	22000	4	2500	610200	Post-Gi
34	46000	25000	5	3500	596160	Gi
35	47000	15000	7	0	456840	Profe
36	50000	20000	4	0	570000	Profe
37	50500	20000	3	0	581760	Profe
38	55000	45000	6	12000	600600	Gi
39	60000	10000	3	0	590400	Post-Gi
40	60000	50000	6	10000	590400	Gi
41	65000	20000	4	5000	647400	I

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_HH_Income	Highest_Qualified_M
42	70000	9000	2	0	756000	Gi
43	80000	20000	4	0	1075200	Gi
44	85000	25000	5	0	1142400	Under-Gi
45	90000	48000	7	0	885600	Post-Gi
46	98000	25000	5	0	1152480	Profe
47	100000	30000	6	0	1404000	Gi
48	100000	50000	4	20000	1032000	Profe
49	100000	40000	6	10000	1320000	Post-Gi

In [6]: income_df.head()

Out[6]:

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_HH_Income	Highest_Qualified_Me
0	5000	8000	3	2000	64200	Under-Græ
1	6000	7000	2	3000	79920	III
2	10000	4500	2	0	112800	Under-Græ
3	10000	2000	1	0	97200	III
4	12500	12000	2	3000	147000	Græ

In [7]: income_df.tail()

Out[7]:

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_HH_Income	Highest_Qualified_M
45	90000	48000	7	0	885600	Post-Gi
46	98000	25000	5	0	1152480	Profe
47	100000	30000	6	0	1404000	Gi
48	100000	50000	4	20000	1032000	Profe
49	100000	40000	6	10000	1320000	Post-Gi

In [8]: # analyze the data

In [9]: income_df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50 entries, 0 to 49
Data columns (total 7 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Mthly_HH_Income                       50 non-null    int64
1   Mthly_HH_Expense                       50 non-null    int64
2   No_of_Fly_Members                     50 non-null    int64
3   Emi_or_Rent_Amt                       50 non-null    int64
4   Annual_HH_Income                       50 non-null    int64
5   Highest_Qualified_Member               50 non-null    object
6   No_of_Earning_Members                  50 non-null    int64
dtypes: int64(6), object(1)
memory usage: 2.9+ KB
```

```
In [11]: income_df.shape
```

```
Out[11]: (50, 7)
```

```
In [12]: income_df.describe().T
```

```
Out[12]:
```

	count	mean	std	min	25%	50%	75%	max
Mthly_HH_Income	50.0	41558.00	26097.908979	5000.0	23550.0	35000.0	50375.0	100000.0
Mthly_HH_Expense	50.0	18818.00	12090.216824	2000.0	10000.0	15500.0	25000.0	50000.0
No_of_Fly_Members	50.0	4.06	1.517382	1.0	3.0	4.0	5.0	7.0
Emi_or_Rent_Amt	50.0	3060.00	6241.434948	0.0	0.0	0.0	3500.0	35000.0
Annual_HH_Income	50.0	490019.04	320135.792123	64200.0	258750.0	447420.0	594720.0	1404000.0
No_of_Earning_Members	50.0	1.46	0.734291	1.0	1.0	1.0	2.0	4.0

```
In [15]: income_df.isna().head()
```

#isna()-check whether the value is na or not
#whether is true means value is na and false means value is not na

```
Out[15]:
```

Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_HH_Income	Highest_Qualified_Member	No_of_Earning_Members
False	False	False	False	False	False
False	False	False	False	False	False
False	False	False	False	False	False
False	False	False	False	False	False
False	False	False	False	False	False

```
In [17]: income_df.isna().any()
```

```
Out[17]: Mthly_HH_Income      False
Mthly_HH_Expense      False
No_of_Fly_Members      False
Emi_or_Rent_Amt        False
Annual_HH_Income       False
Highest_Qualified_Member False
No_of_Earning_Members  False
dtype: bool
```

no null value in the dataset

what is the mean expense of a household ?

```
In [18]: income_df["Mthly_HH_Expense"].mean()
```

```
Out[18]: 18818.0
```

what is the median household expense?

```
In [19]: income_df["Mthly_HH_Expense"].median()
```

```
Out[19]: 15500.0
```

what is the mothly expenses for most of the household ?

```
In [20]: mth_exp_tmp = pd.crosstab(index=income_df["Mthly_HH_Expense"], columns="count")
mth_exp_tmp.reset_index(inplace=True)
mth_exp_tmp[mth_exp_tmp['count'] == income_df.Mthly_HH_Expense.value_counts().max()]
```

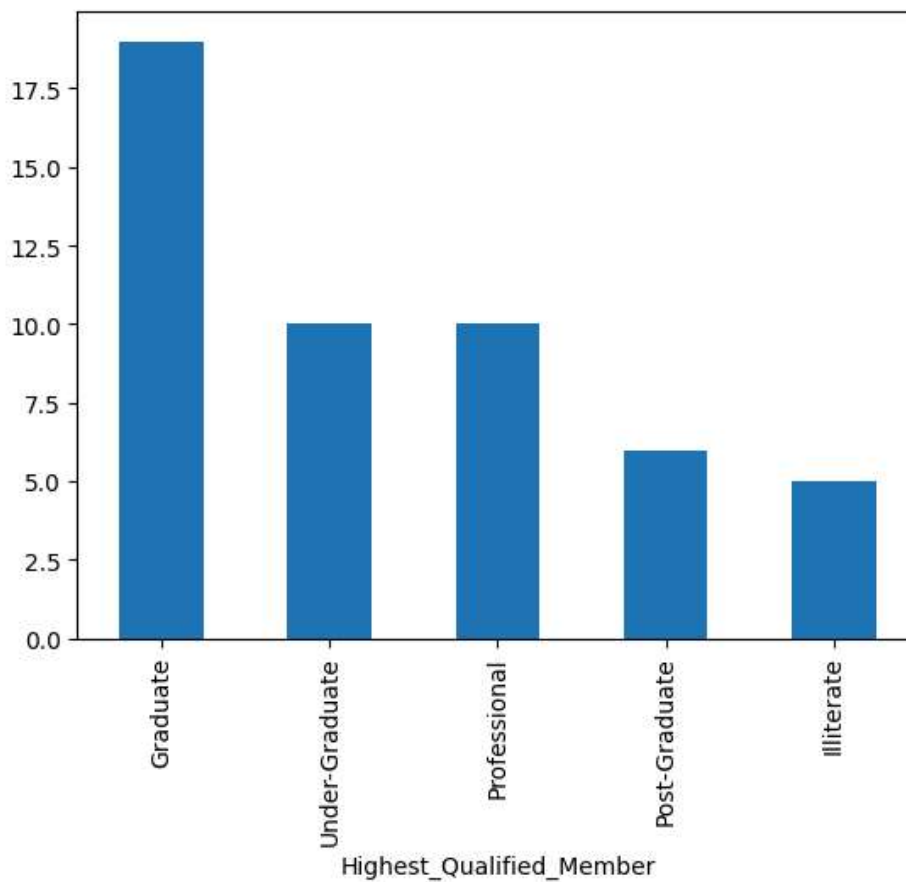
```
Out[20]:
```

col_0	Mthly_HH_Expense	count
18	25000	8

Plot the Histogram to count the Highest qualified member

```
In [22]: income_df["Highest_Qualified_Member"].value_counts().plot(kind="bar")
```

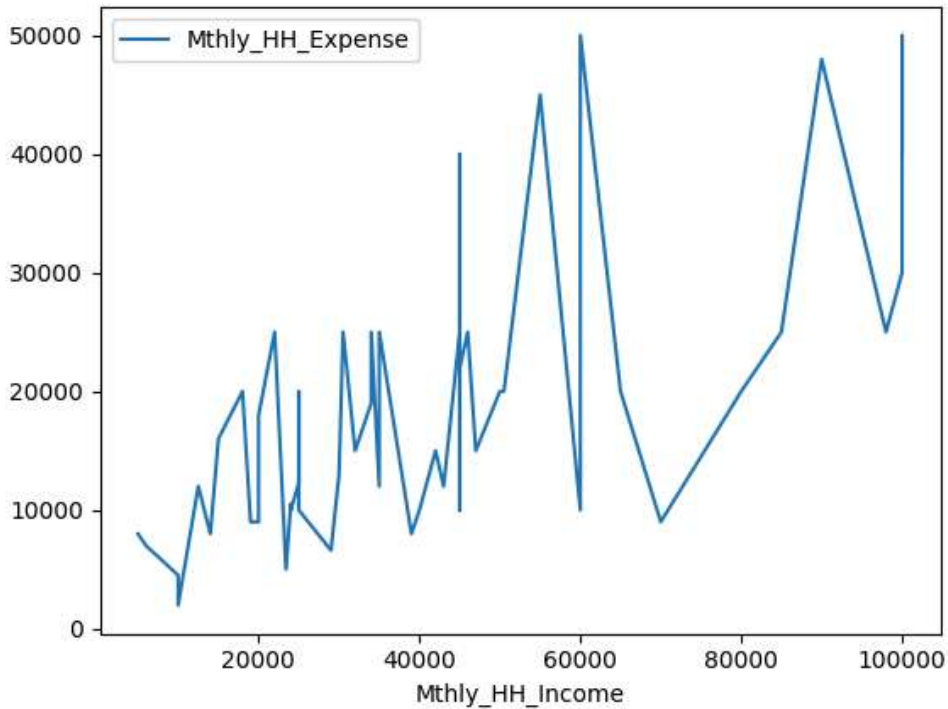
```
Out[22]: <Axes: xlabel='Highest_Qualified_Member'>
```



Calculate IQR(difference between 75% and 25% quartile)

```
In [23]: income_df.plot(x="Mthly_HH_Income", y="Mthly_HH_Expense")
IQR=income_df["Mthly_HH_Expense"].quantile(0.75)-income_df["Mthly_HH_Expense"].quantile(0.25)
IQR
```

Out[23]: 15000.0



```
In [26]: #Calculate Variance for first 3 columns.
pd.DataFrame(income_df.iloc[:,0:4].var().to_frame()).T
```

Out[26]:

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt
0	6.811009e+08	1.461733e+08	2.302449	3.895551e+07

```
In [28]: #11.Calculate the count of Highest qualified member.
income_df["Highest_Qualified_Member"].value_counts().to_frame().T
```

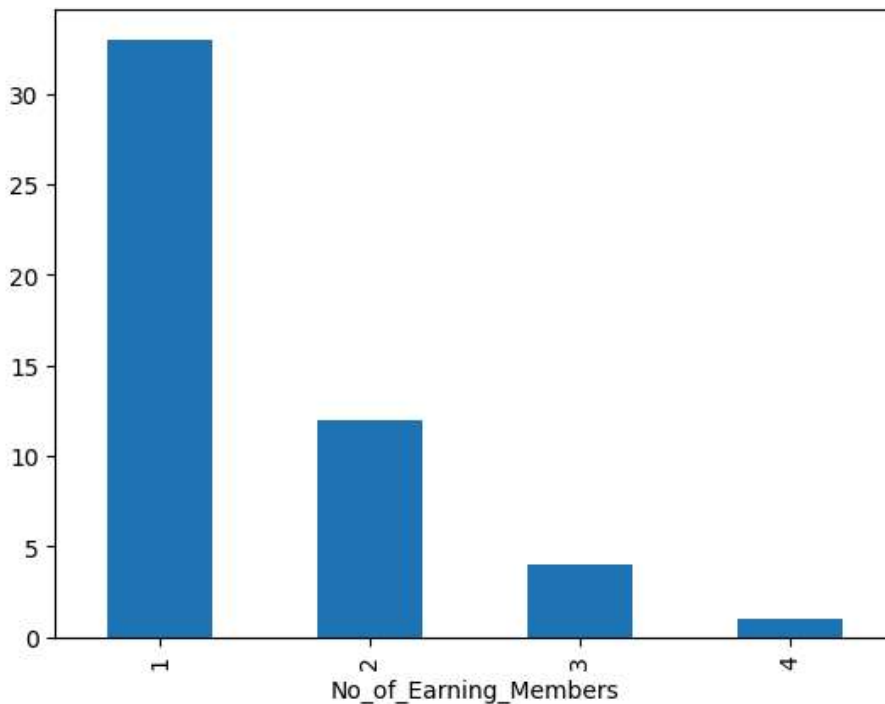
Out[28]:

	Highest_Qualified_Member	Graduate	Under-Graduate	Professional	Post-Graduate	Illiterate
count		19	10	10	6	5

plot the histogram to count the no_of_earning_members

```
In [30]: income_df["No_of_Earning_Members"].value_counts().plot(kind="bar")
```

```
Out[30]: <Axes: xlabel='No_of_Earning_Members'>
```



Suppose you have option to invest in Stock A or Stock B. The stocks • have different expected returns and standard deviations. The expected return of Stock A is 15% and Stock B is 10%. Standard Deviation of the returns of these stocks is 10% and 5% respectively.¶

```
In [31]: #here we need to calculate the coeff of variation
```

```
Coeff_of_var_StockA=10/15  
print(Coeff_of_var_StockA)  
Coeff_of_var_StockB=5/10  
print(Coeff_of_var_StockB)
```

```
0.6666666666666666  
0.5
```

```
In [ ]:
```