

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

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

2.Load the file

```
In [3]: income_df=pd.read_csv(r"D:\Data Science with AI\Data Science With AI\6th, 7th -a
```

```
In [4]: income_df.head()
```

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_HH_Income
0	5000	8000	3	2000	10000
1	6000	7000	2	3000	12000
2	10000	4500	2	0	20000
3	10000	2000	1	0	10000
4	12500	12000	2	3000	25000

3.Analyze the data

```
In [5]: 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 [6]: income_df['Highest_Qualified_Member']=income_df['Highest_Qualified_Member'].astype(str)
```

```
In [7]: 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  category 
 6   No_of_Earning_Members 50 non-null    int64  
dtypes: category(1), int64(6)
memory usage: 2.7 KB
```

In [8]: `income_df.shape`

Out[8]: (50, 7)

In [9]: `income_df.describe().T`

	count	mean	std	min	25%	50%
Mthly_HH_Income	50.0	41558.00	26097.908979	5000.0	23550.0	35000.0
Mthly_HH_Expense	50.0	18818.00	12090.216824	2000.0	10000.0	15500.0
No_of_Fly_Members	50.0	4.06	1.517382	1.0	3.0	4.0
Emi_or_Rent_Amt	50.0	3060.00	6241.434948	0.0	0.0	0.0
Annual_HH_Income	50.0	490019.04	320135.792123	64200.0	258750.0	447420.0
No_of_Earning_Members	50.0	1.46	0.734291	1.0	1.0	1.0

In [10]: `income_df.isna.any()`

```
-----
AttributeError                               Traceback (most recent call last)
Cell In[10], line 1
----> 1 income_df.isna.any()

AttributeError: 'function' object has no attribute 'any'
```

In [11]: `income_df.isna().any()`

```
Out[11]: 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
```

4.What is the Mean Expense of a Household

```
In [12]: income_df["Mthly_HH_Expense"].mean()
```

```
Out[12]: np.float64(18818.0)
```

```
In [14]: income_df['Mthly_HH_Expense'].median()
```

```
Out[14]: 15500.0
```

```
In [15]: 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()]
```

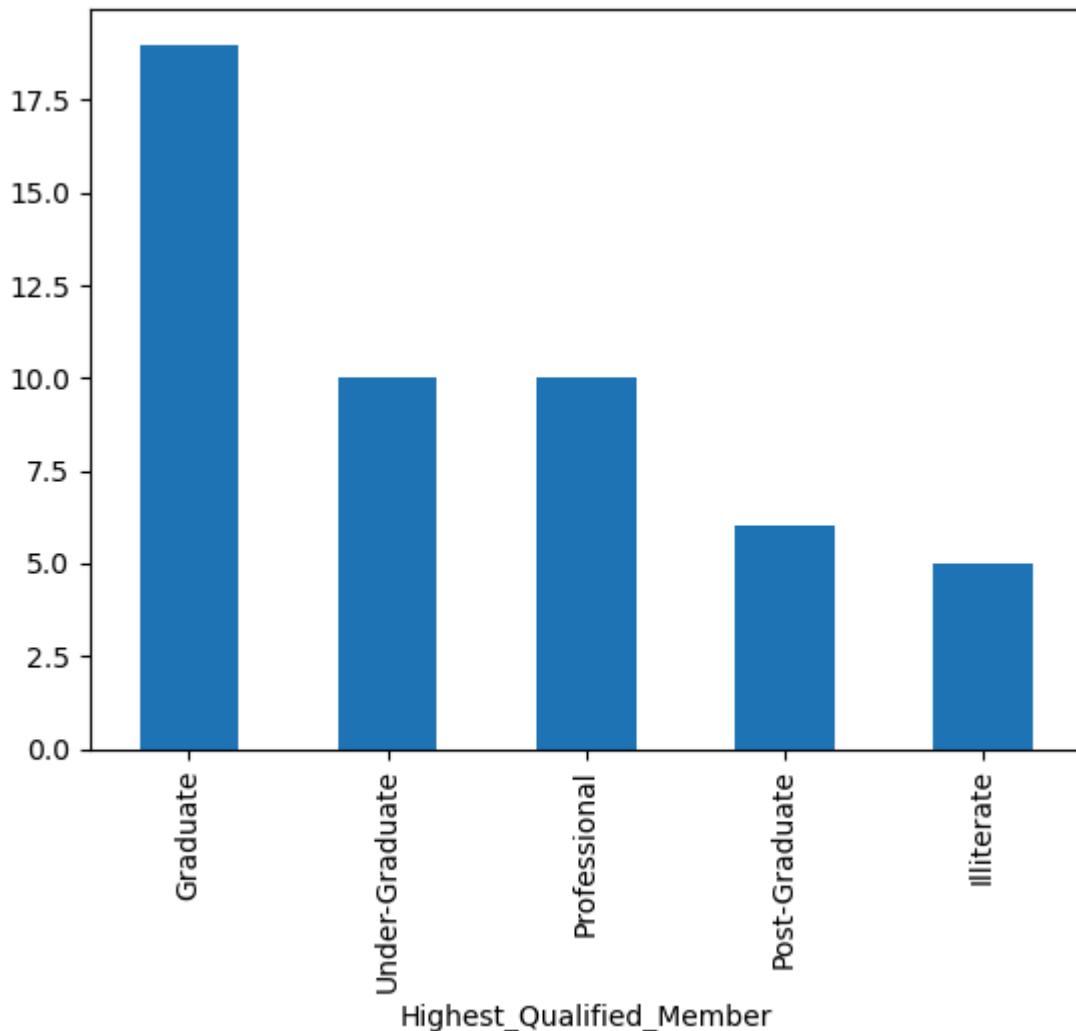
```
Out[15]: col_0 Mthly_HH_Expense count
```

col_0	Mthly_HH_Expense	count
18	25000	8

7.Plot the Histogram to count the Highest qualified member

```
In [17]: income_df['Highest_Qualified_Member'].value_counts().plot(kind='bar')
```

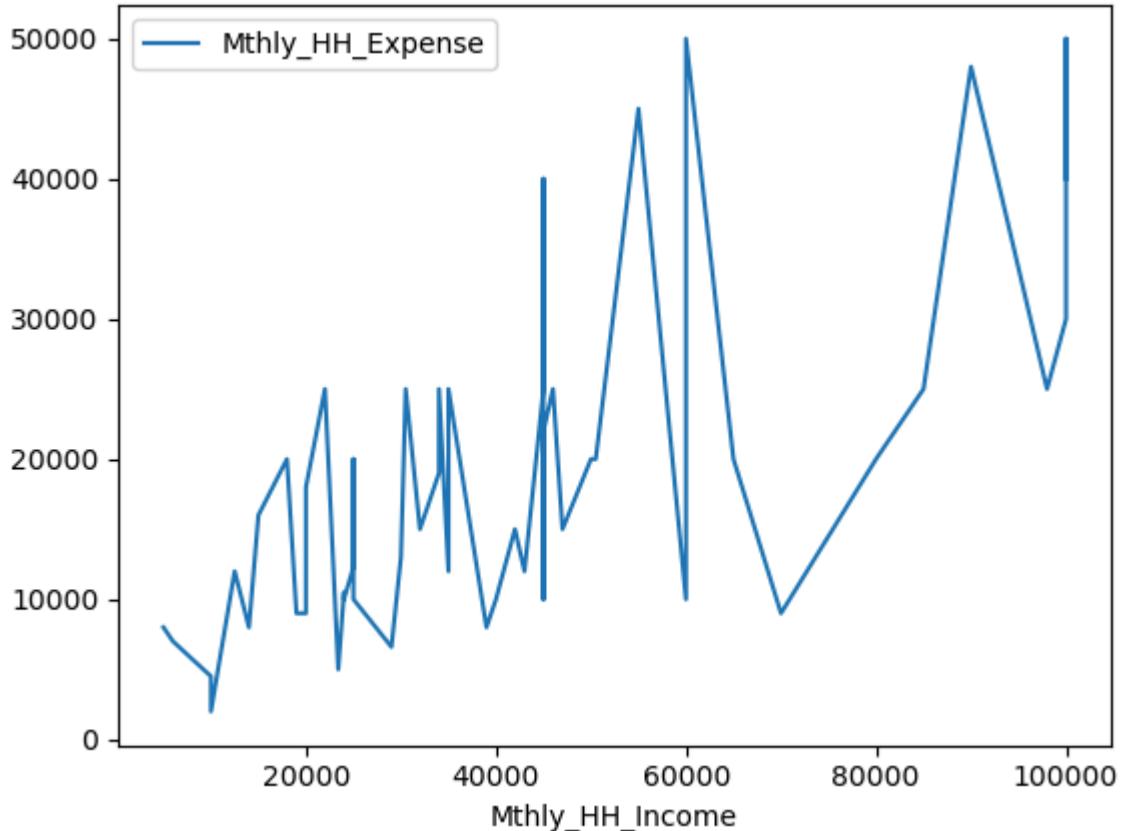
```
Out[17]: <Axes: xlabel='Highest_Qualified_Member'>
```



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

```
In [18]: 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'].q
IQR
```

Out[18]: np.float64(15000.0)



9.Calculate Standard Deviation for first 4 columns

```
In [19]: pd.DataFrame(income_df.iloc[:,0:5].std().to_frame()).T
```

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annual_Goal
0	26097.908979	12090.216824	1.517382	6241.434948	31000.0

10.Calculate Variance for first 3 columns

```
In [20]: pd.DataFrame(income_df.iloc[:,0:4].var().to_frame()).T
```

Out[20]:

	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 [21]:

```
pd.DataFrame(income_df.iloc[:,0:4].var().to_frame())
```

Out[21]:

	0
Mthly_HH_Income	6.811009e+08
Mthly_HH_Expense	1.461733e+08
No_of_Fly_Members	2.302449e+00
Emi_or_Rent_Amt	3.895551e+07

11. Calculate the count of Highest qualified member.

In [22]:

```
income_df['Highest_Qualified_Member'].value_counts().to_frame().T
```

Out[22]:

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

In [23]:

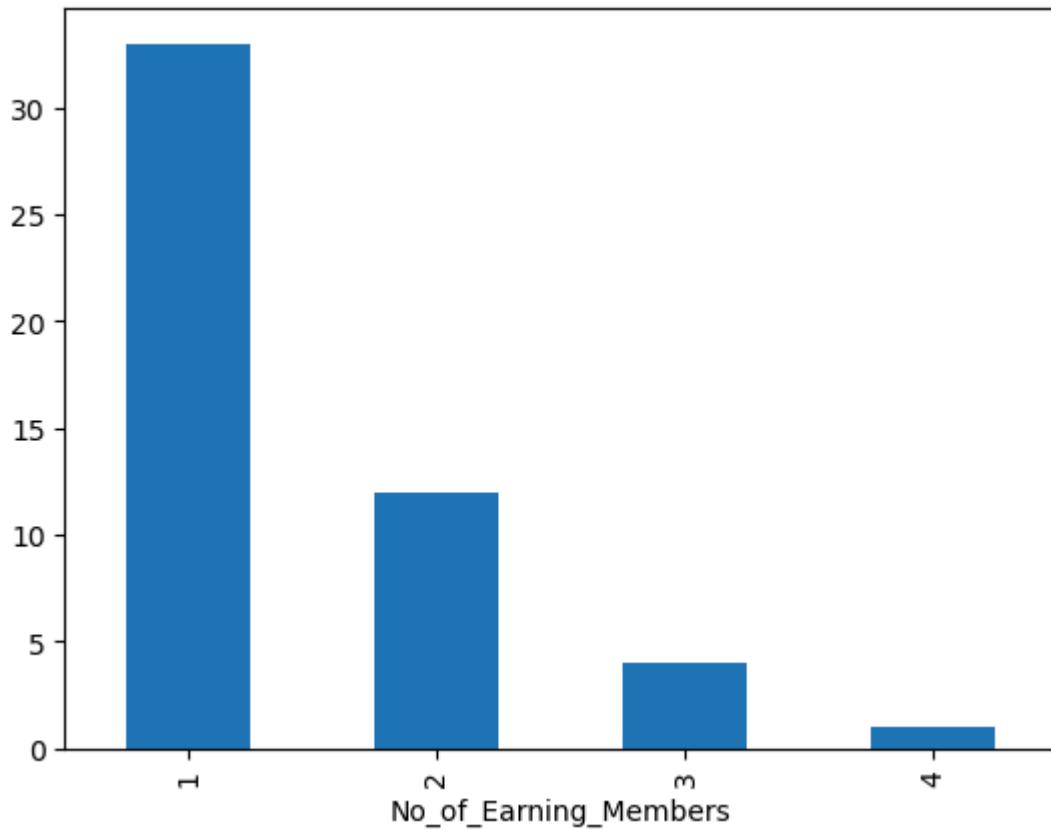
```
income_df['Highest_Qualified_Member']
```

```
Out[23]: 0      Under-Graduate
1          Illiterate
2      Under-Graduate
3          Illiterate
4          Graduate
5          Graduate
6    Post-Graduate
7          Graduate
8      Under-Graduate
9      Under-Graduate
10     Under-Graduate
11     Illiterate
12     Illiterate
13     Graduate
14     Graduate
15     Graduate
16     Graduate
17     Under-Graduate
18     Graduate
19     Graduate
20     Under-Graduate
21     Professional
22     Professional
23     Professional
24     Graduate
25     Professional
26     Under-Graduate
27     Under-Graduate
28     Graduate
29     Graduate
30     Graduate
31     Professional
32     Post-Graduate
33     Post-Graduate
34     Graduate
35     Professional
36     Professional
37     Professional
38     Graduate
39     Post-Graduate
40     Graduate
41     Illiterate
42     Graduate
43     Graduate
44     Under-Graduate
45     Post-Graduate
46     Professional
47     Graduate
48     Professional
49     Post-Graduate
Name: Highest_Qualified_Member, dtype: category
Categories (5, object): ['Graduate', 'Illiterate', 'Post-Graduate', 'Professional', 'Under-Graduate']
```

12.Plot the Histogram to count the No_of_Earning_Members

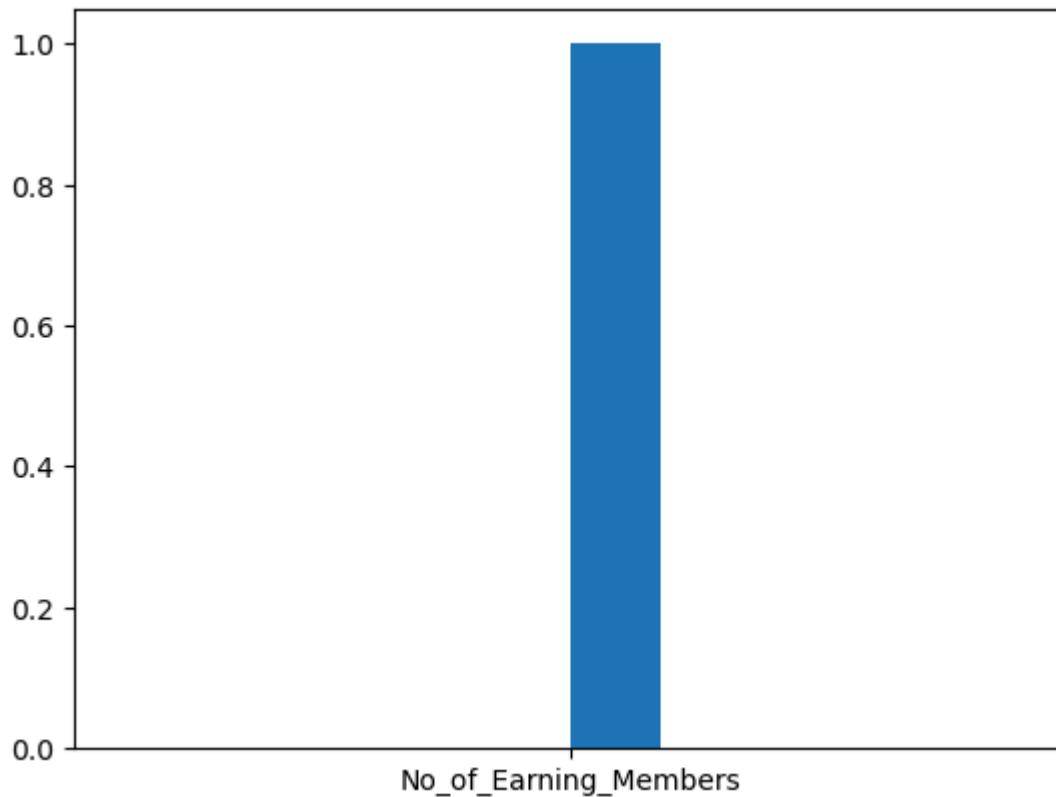
```
In [24]: income_df['No_of_Earning_Members'].value_counts().plot(kind='bar')
```

```
Out[24]: <Axes: xlabel='No_of_Earning_Members'>
```



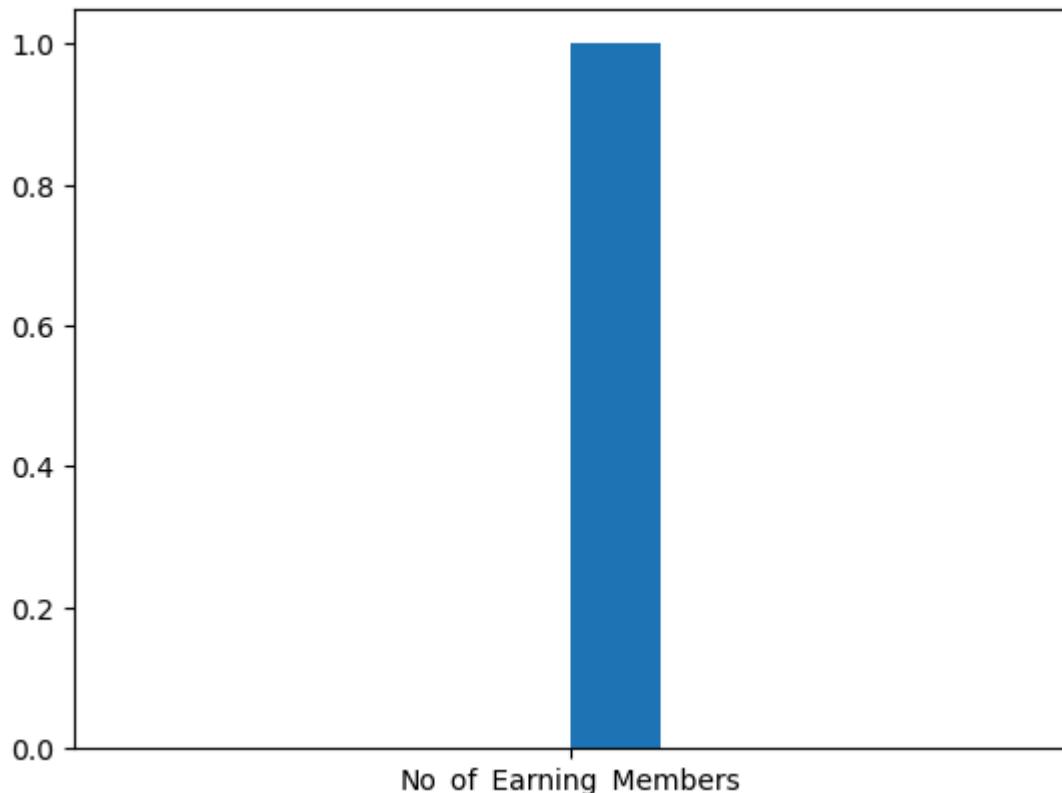
```
In [26]: plt.hist('No_of_Earning_Members')
```

```
Out[26]: (array([0., 0., 0., 0., 0., 1., 0., 0., 0., 0.]),  
 array([-0.5, -0.4, -0.3, -0.2, -0.1, 0., 0.1, 0.2, 0.3, 0.4, 0.5]),  
 <BarContainer object of 10 artists>)
```



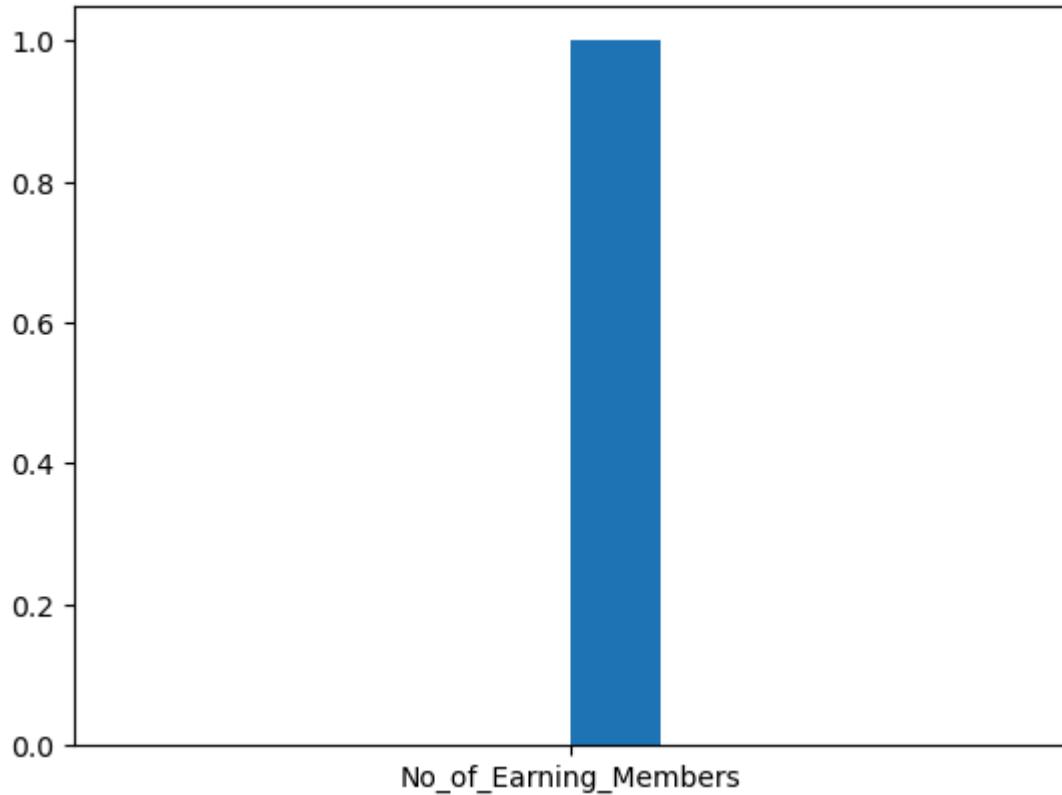
```
In [27]: plt.hist('No_of_Earning_Members').value_counts()
```

```
-----  
AttributeError                               Traceback (most recent call last)  
Cell In[27], line 1  
----> 1 plt.hist()                           .value_counts()  
  
AttributeError: 'tuple' object has no attribute 'value_counts'
```



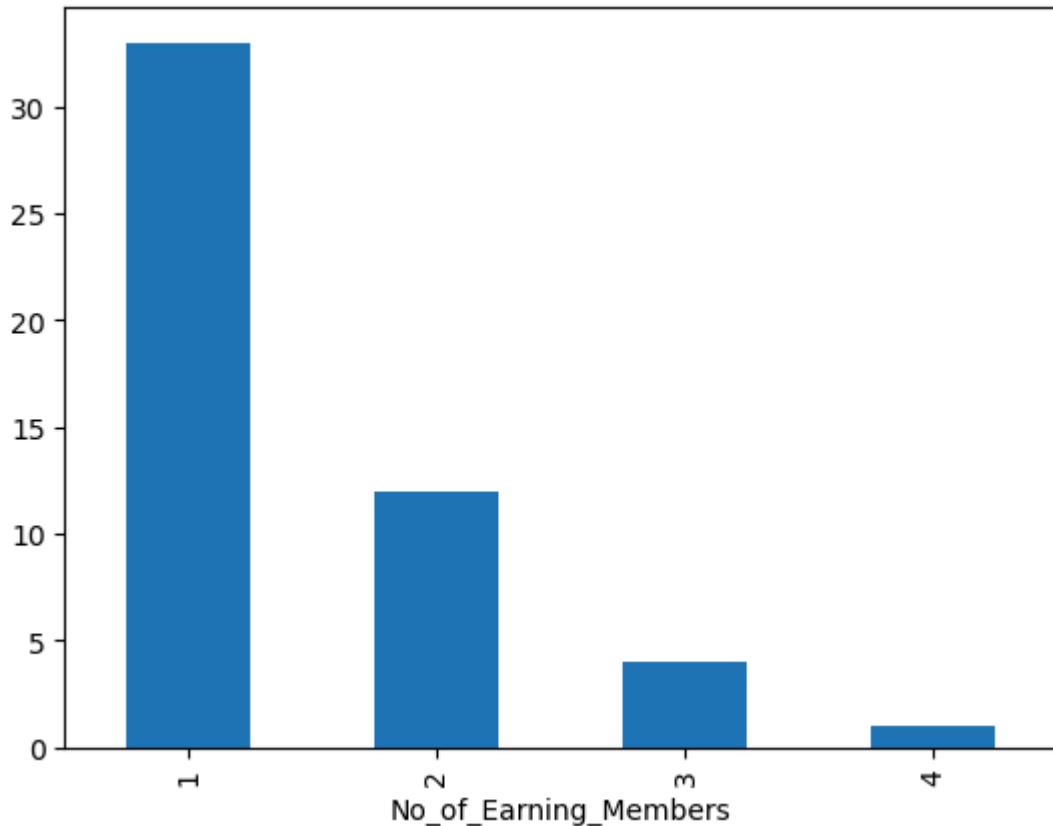
```
In [28]: plt.hist('No_of_Earning_Members').value_counts().plot(kind='bar')
```

```
-----  
AttributeError                               Traceback (most recent call last)  
Cell In[28], line 1  
----> 1 plt.hist()                           .value_counts().plot(kind='bar')  
  
AttributeError: 'tuple' object has no attribute 'value_counts'
```



```
In [29]: income_df['No_of_Earning_Members'].value_counts().plot(kind='bar')
```

```
Out[29]: <Axes: xlabel='No_of_Earning_Members'>
```



```
In [31]: income_df['No_of_Earning_Members'].value_counts()
```

```
Out[31]: No_of_Earning_Members
1    33
2    12
3     4
4     1
Name: count, dtype: int64
```

13. 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 os 10%. Standard Deviation of the returns of these stocks is 10% and 5 % respectively

```
In [33]: 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 [ ]:
```