

```
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()
```

```
Out[4]:
```

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

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'].asty
```

```
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`

Out[9]:

	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(
```

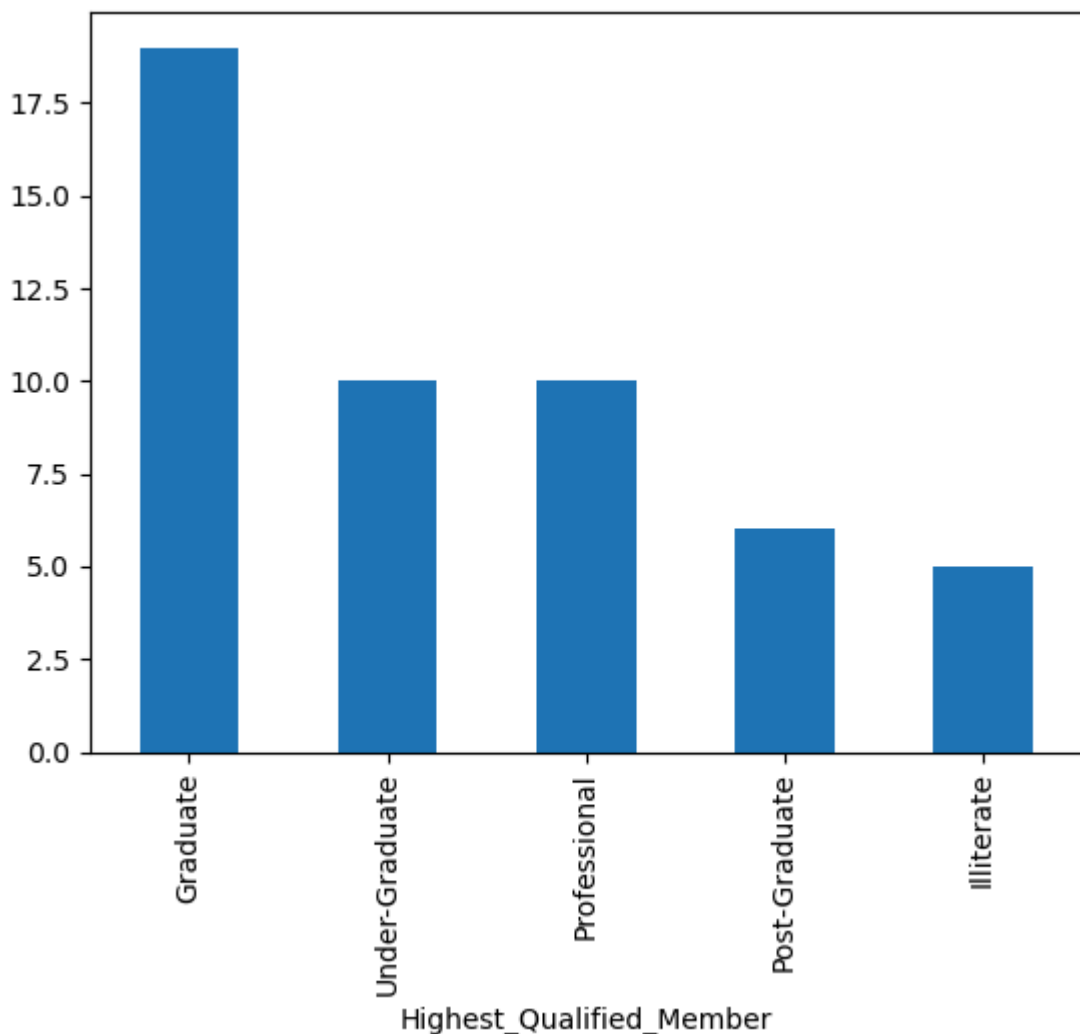
```
Out[15]:
```

col_0	Mthly_HH_Expense	count
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')
```

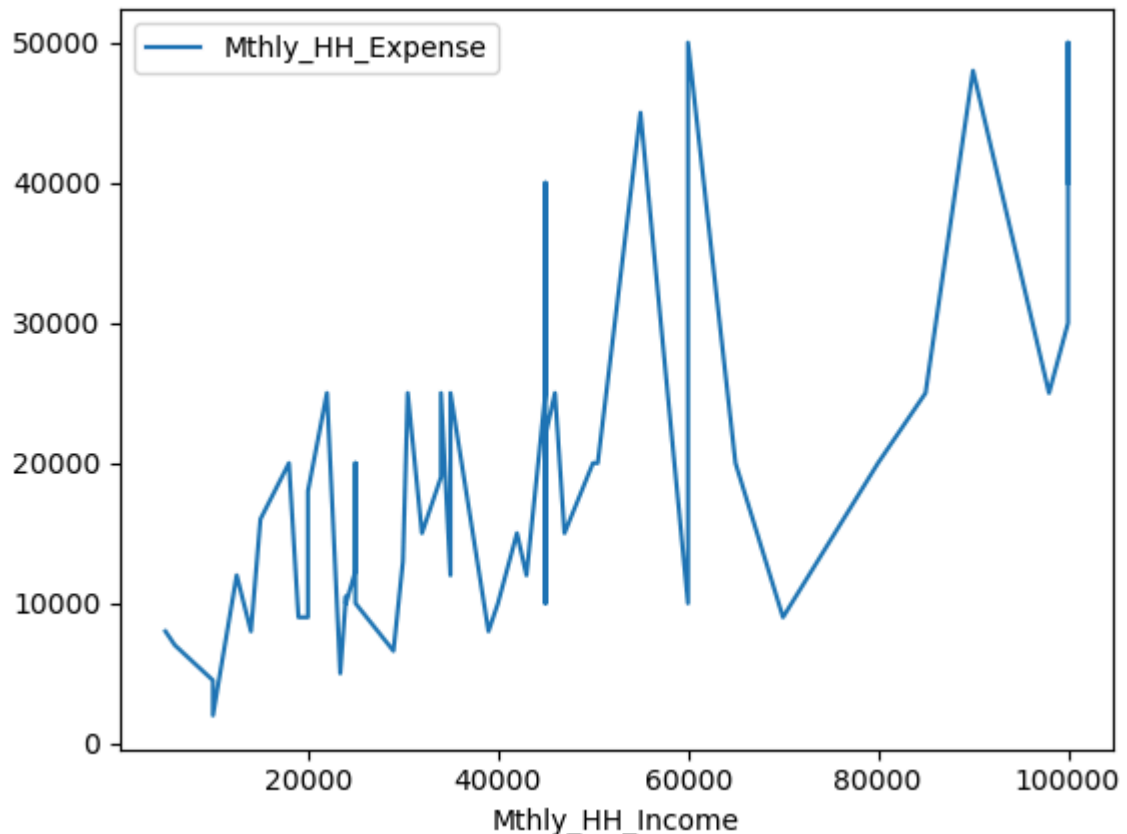
```
Out[17]: <Axes: xlabel='Highest_Qualified_Member'>
```



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
```

```
Out[19]:
```

	Mthly_HH_Income	Mthly_HH_Expense	No_of_Fly_Members	Emi_or_Rent_Amt	Annua
0	26097.908979	12090.216824	1.517382	6241.434948	3

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']
```

```

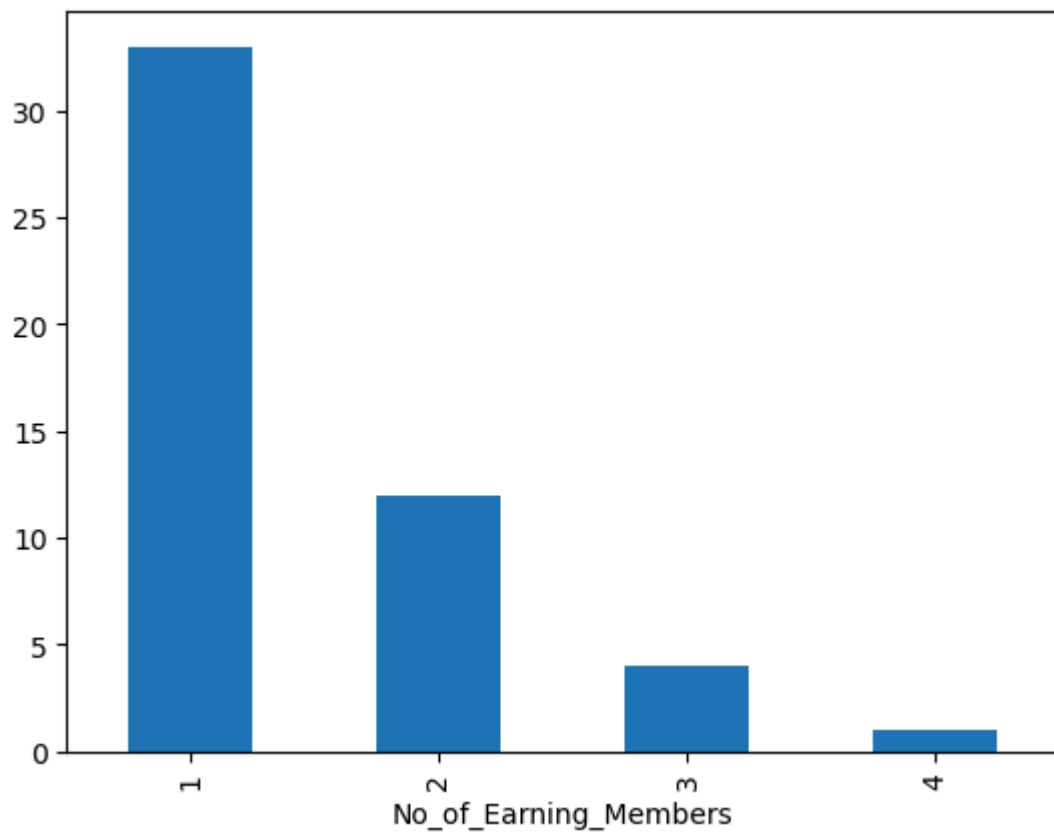
Out[23]: 0      Under-Graduate
        1      Illiterate
        2      Under-Graduate
        3      Illiterate
        4      Graduate
        5      Graduate
        6      Post-Graduate
        7      Graduate
        8      Under-Graduate
        9      Under-Graduate
       10      Under-Graduate
       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', 'Profession
al', 'Under-Graduate']

```

12. Plot the Histogram to count the No_of_Earning_Members

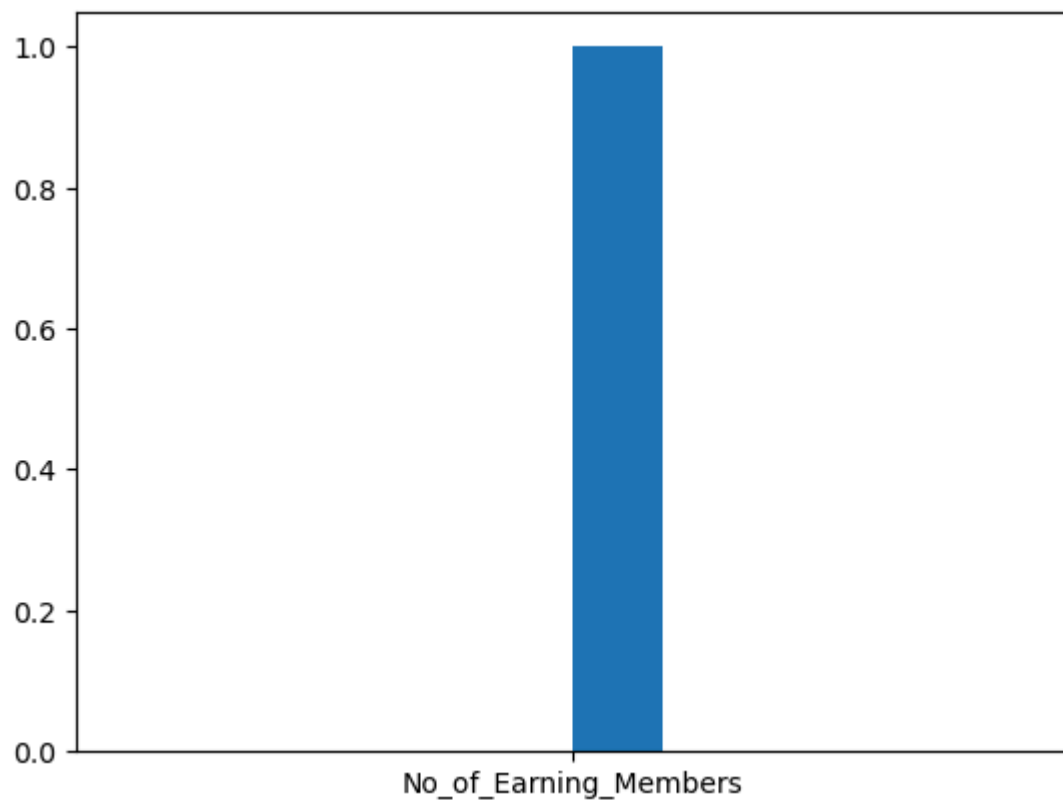
```
In [24]: income_df['No_of_Earning_Members'].value_counts().plot(kind='bar')
```

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



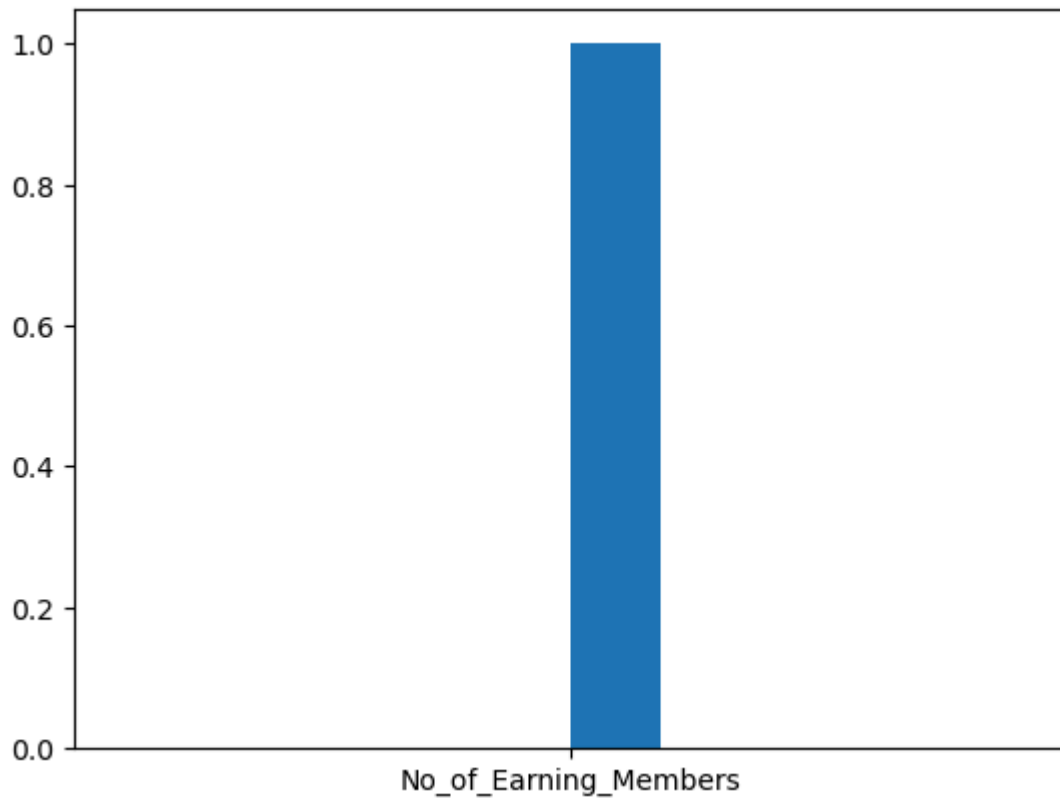
```
In [26]: plt.hist('No_of_Earning_Members')
```

```
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>)
```



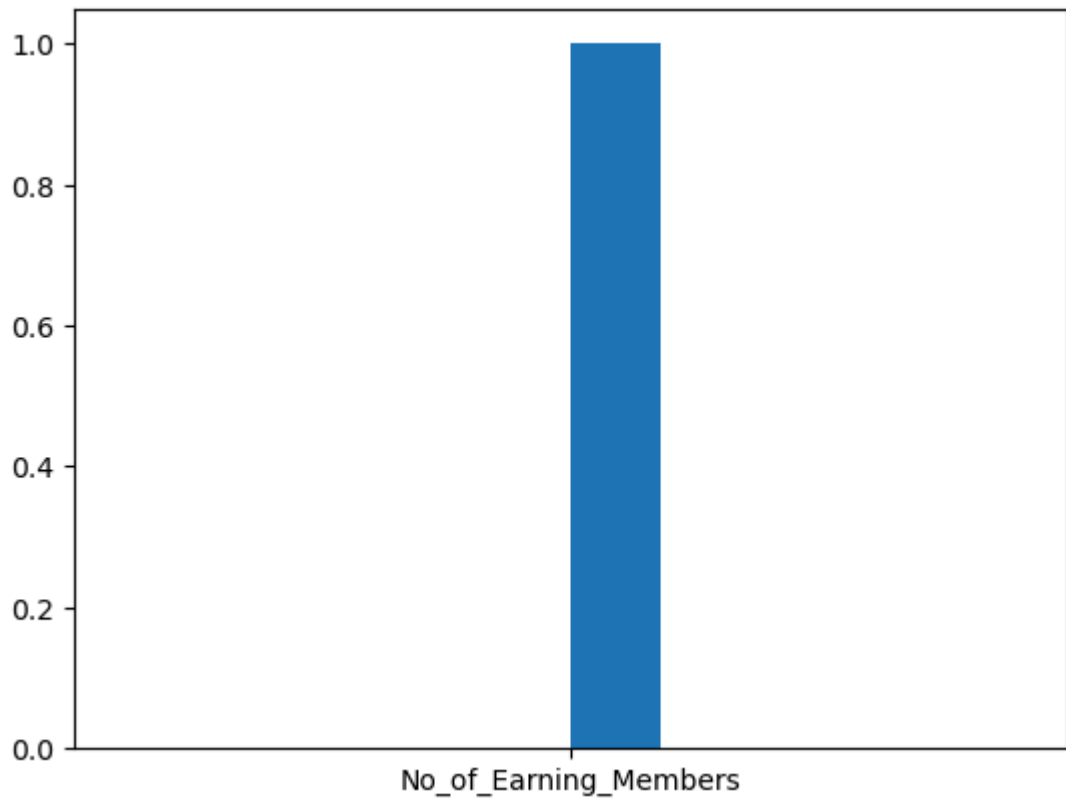
```
In [27]: plt.hist('No_of_Earning_Members').value_counts()
```

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



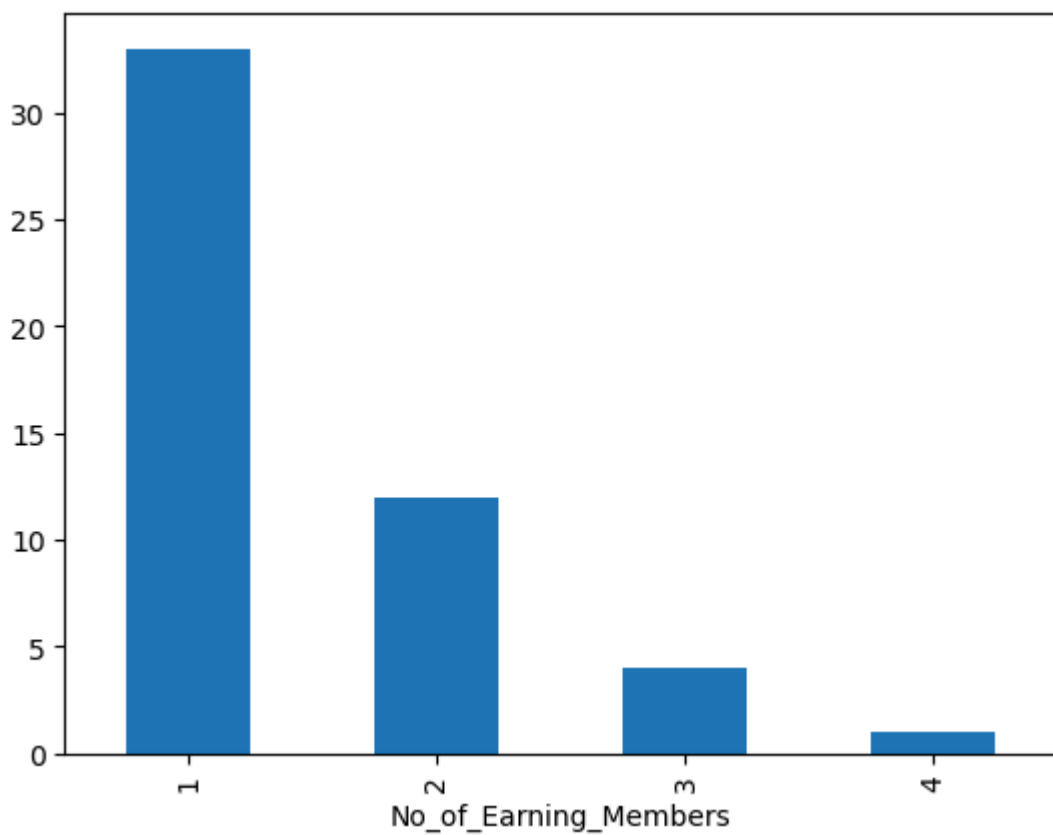
```
In [28]: plt.hist('No_of_Earning_Members').value_counts().plot(kind='bar')
```

```
-----  
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 is 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 [ ]:
```