

```
In [1]: from mpl_toolkits.mplot3d import Axes3D
from sklearn.preprocessing import StandardScaler
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
sns.set_style('darkgrid')
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
plt.rcParams["figure.figsize"] = (12,7)
```

```
In [2]: data = pd.read_csv('income_evaluation.csv')
```

```
In [3]: data.head(1)
```

Out[3]:

	age	workclass	fnlwgt	education	education-num	marital-status	occupation	relationship	race	sex
0	39	State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male

```
In [4]: data.shape
```

Out[4]: (32561, 15)

```
In [5]: data.columns
```

```
Out[5]: Index(['age', ' workclass', ' fnlwgt', ' education', ' education-num',
               ' marital-status', ' occupation', ' relationship', ' race', ' sex',
               ' capital-gain', ' capital-loss', ' hours-per-week', ' native-countr
y',
               ' income'],
              dtype='object')
```

In [6]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32561 entries, 0 to 32560
Data columns (total 15 columns):
age                32561 non-null int64
workclass          32561 non-null object
fnlwgt             32561 non-null int64
education          32561 non-null object
education-num      32561 non-null int64
marital-status     32561 non-null object
occupation         32561 non-null object
relationship       32561 non-null object
race               32561 non-null object
sex               32561 non-null object
capital-gain       32561 non-null int64
capital-loss       32561 non-null int64
hours-per-week     32561 non-null int64
native-country     32561 non-null object
income             32561 non-null object
dtypes: int64(6), object(9)
memory usage: 3.7+ MB
```

In [7]: `del data['fnlwgt']`
`del data['education-num']`
`del data['relationship']`

In [8]: data.columns

Out[8]: Index(['age', 'workclass', 'education', 'marital-status', 'occupation',
 'race', 'sex', 'capital-gain', 'capital-loss', 'hours-per-week',
 'native-country', 'income'],
 dtype='object')

In [9]: data.columns=['age', 'workclass', 'education', 'married', 'occupation',\
 'race', 'sex', 'gain', 'loss', 'hours', 'country', 'income']

In [10]: data.columns

Out[10]: Index(['age', 'workclass', 'education', 'married', 'occupation', 'race', 'se
x',
 'gain', 'loss', 'hours', 'country', 'income'],
 dtype='object')

In [11]: data = data[data.country == 'United-States']
`del data['country']`

In [12]: data = data[data.gain == 0]
`del data['gain']`

In [13]: data = data[data.loss == 0]
`del data['loss']`

In [14]: `data.count()`

Out[14]:

age	25320
workclass	25320
education	25320
married	25320
occupation	25320
race	25320
sex	25320
hours	25320
income	25320
dtype:	int64

In [15]: `data.head(1)`

Out[15]:

	age	workclass	education	married	occupation	race	sex	hours	income
1	50	Self-emp-not-inc	Bachelors	Married-civ-spouse	Exec-managerial	White	Male	13	<=50K

In [16]: `data.shape`

Out[16]: (25320, 9)

In [17]: `data = data[data.workclass != ' ?']`
`data = data[data.occupation != ' ?']`

In [18]: `data.shape`

Out[18]: (23816, 9)

In [19]: `data.head(3)`

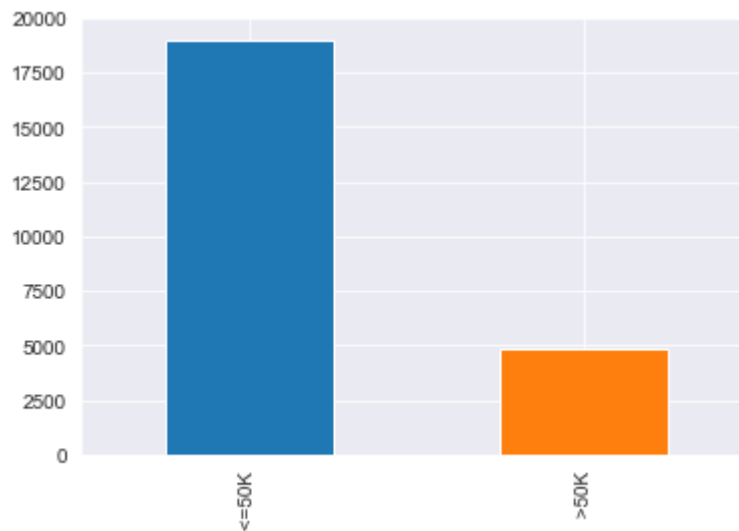
Out[19]:

	age	workclass	education	married	occupation	race	sex	hours	income
1	50	Self-emp-not-inc	Bachelors	Married-civ-spouse	Exec-managerial	White	Male	13	<=50K
2	38	Private	HS-grad	Divorced	Handlers-cleaners	White	Male	40	<=50K
3	53	Private	11th	Married-civ-spouse	Handlers-cleaners	Black	Male	40	<=50K

Under 5,000 people make 50K from the dataset

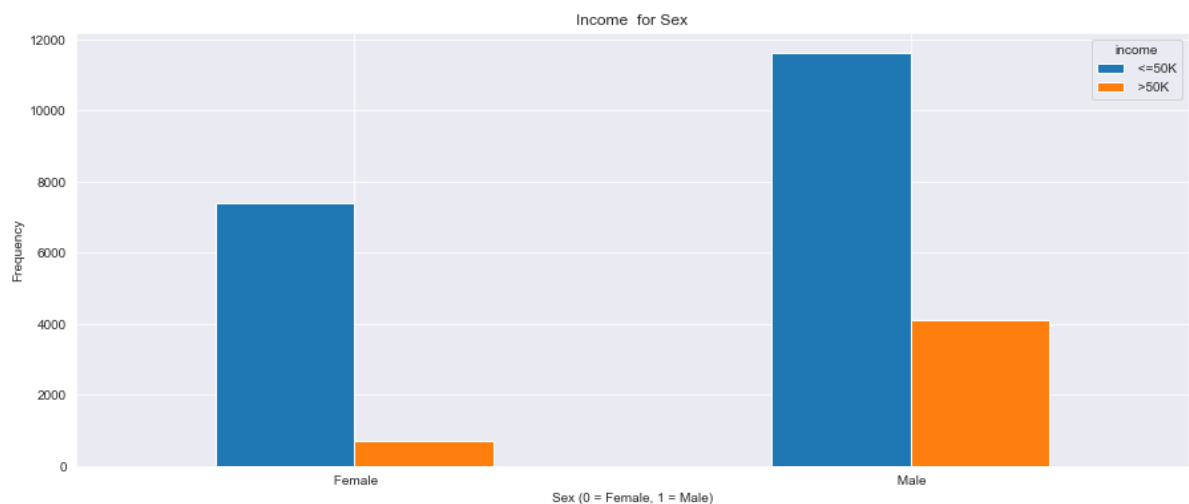
```
In [20]: data.income.value_counts().plot.bar()
```

```
Out[20]: <matplotlib.axes._subplots.AxesSubplot at 0x12862e35fd0>
```



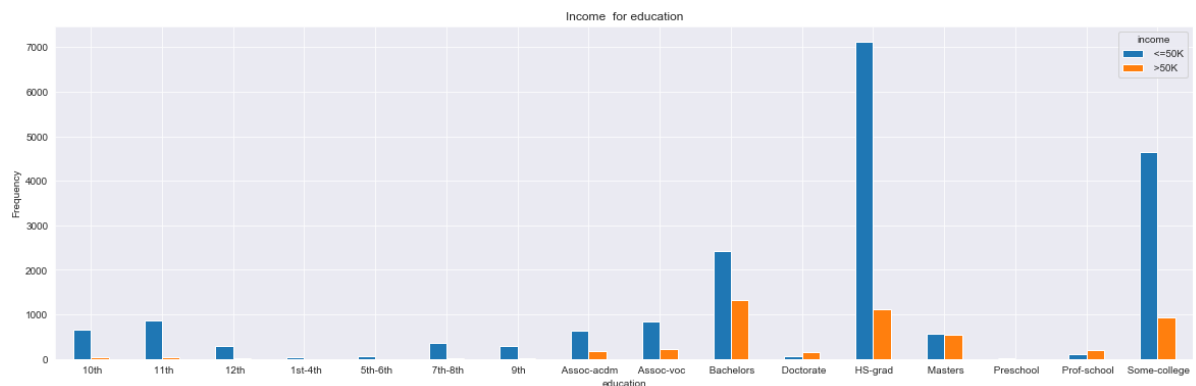
A greater ratio of men make above 50K

```
In [21]: pd.crosstab(data.sex,data.income).plot(kind="bar",figsize=(15,6))  
plt.title('Income for Sex')  
plt.xlabel('Sex (0 = Female, 1 = Male)')  
plt.xticks(rotation=0)  
plt.ylabel('Frequency')  
plt.show()
```



A Masters degree gives almost a 50% chance of making more than 50K

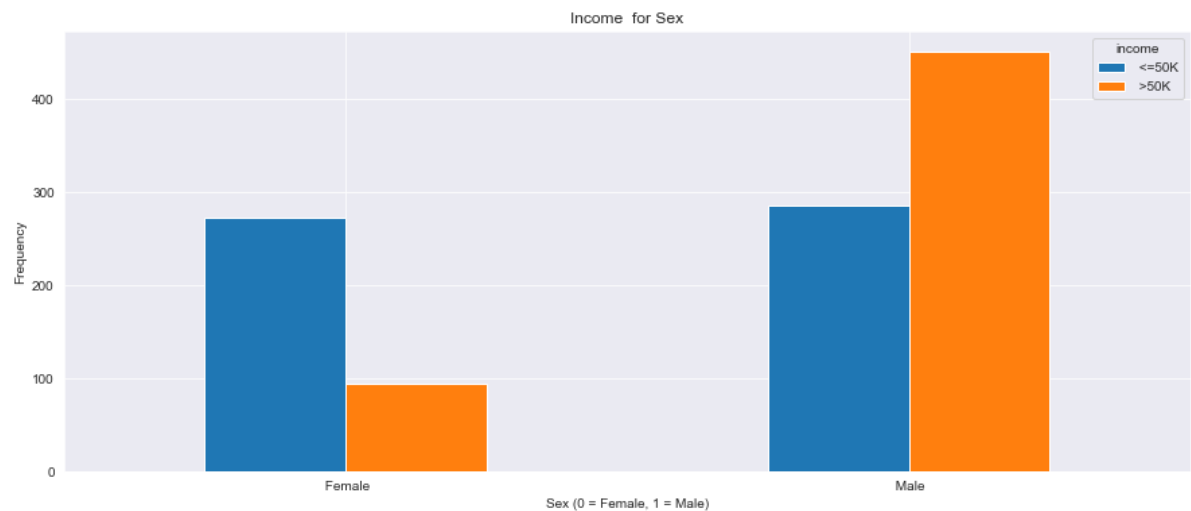
```
In [22]: pd.crosstab(data.education,data.income).plot(kind="bar",figsize=(20,6))
plt.title('Income for education')
plt.xticks(rotation=0)
plt.ylabel('Frequency')
plt.show()
```



Even with a Masters degree a male is still 4x more likely to make over 50K

```
In [23]: m = data[data.education == 'Masters']
```

```
In [24]: pd.crosstab(m.sex,data.income).plot(kind="bar",figsize=(15,6))
plt.title('Income for Sex')
plt.xlabel('Sex (0 = Female, 1 = Male)')
plt.xticks(rotation=0)
plt.ylabel('Frequency')
plt.show()
```



```
In [25]: m[m.sex == 'Female'].income.value_counts()
```

```
Out[25]: <=50K    273
         >50K     94
         Name: income, dtype: int64
```

```
In [26]: m[m.sex == ' Male'].income.value_counts()
```

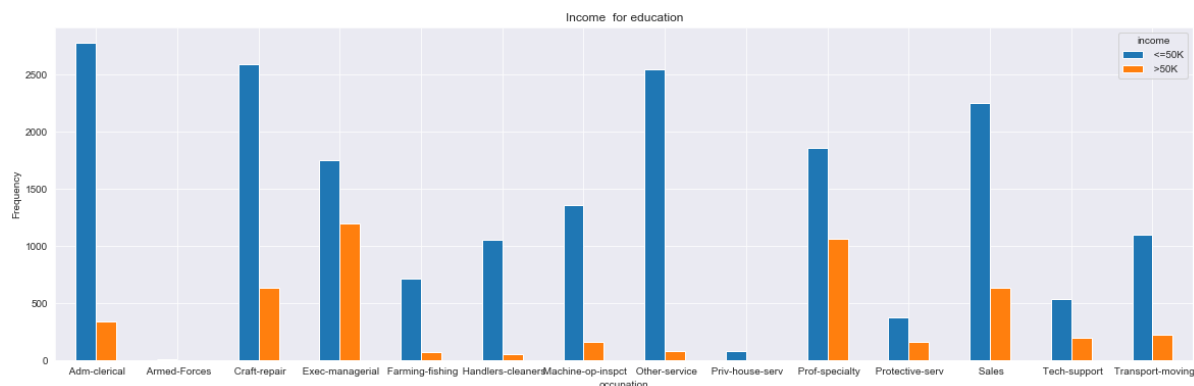
```
Out[26]: >50K      451
         <=50K    286
         Name: income, dtype: int64
```

```
In [27]: data[data.education != ' Masters'].income.value_counts()
```

```
Out[27]: <=50K    18450
         >50K     4262
         Name: income, dtype: int64
```

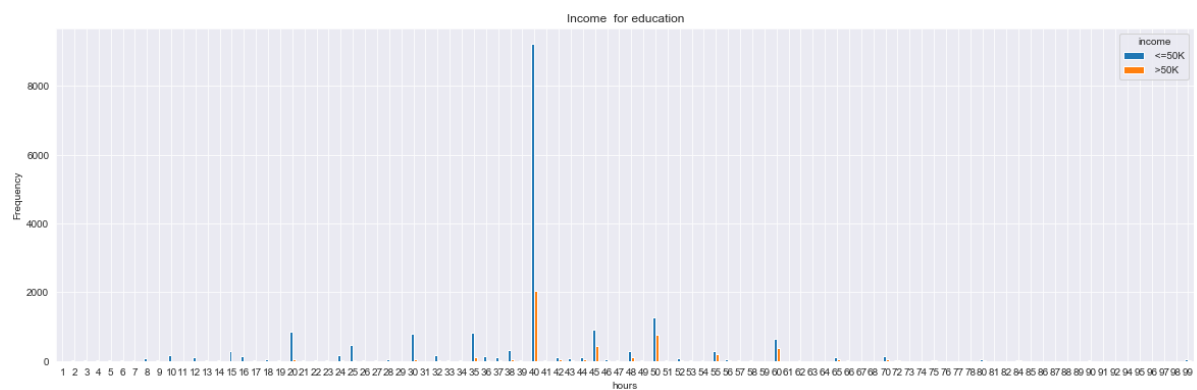
If you are in management or a Profesional, you have a much better chance of making more than 50K

```
In [28]: pd.crosstab(data.occupation,data.income).plot(kind="bar",figsize=(20,6))
         plt.title('Income for education')
         plt.xticks(rotation=0)
         plt.ylabel('Frequency')
         plt.show()
```



Working more than 60 hours a week greatly improves your chances.

```
In [29]: pd.crosstab(data.hours,data.income).plot(kind="bar",figsize=(20,6))  
plt.title('Income for education')  
plt.xticks(rotation=0)  
plt.ylabel('Frequency')  
plt.show()
```



```
In [ ]:
```