

```
In [1]:  
import matplotlib.pyplot as plt  
import pandas as pd  
import numpy as np  
  
df=pd.read_csv('canada_per_capita_income.csv')  
df.head()
```

	year	per capita income (US\$)
0	1970	3399.299037
1	1971	3768.297935
2	1972	4251.175484
3	1973	4804.463248
4	1974	5576.514583

by sklearn

```
In [5]:  
df.info()  
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 47 entries, 0 to 46  
Data columns (total 2 columns):  
 #   Column            Non-Null Count  Dtype     
---  --  
 0   year              47 non-null      int64    
 1   per capita income (US$) 47 non-null      float64  
dtypes: float64(1), int64(1)  
memory usage: 884.0 bytes
```

```
In [7]:  
plt.scatter(df['year'],df['per capita income (US$)'])  
plt.xlabel('YEAR')  
plt.ylabel('INCOME(in $)')  
Text(0, 0.5, 'INCOME(in $)')
```

In [1]:

In [3]:

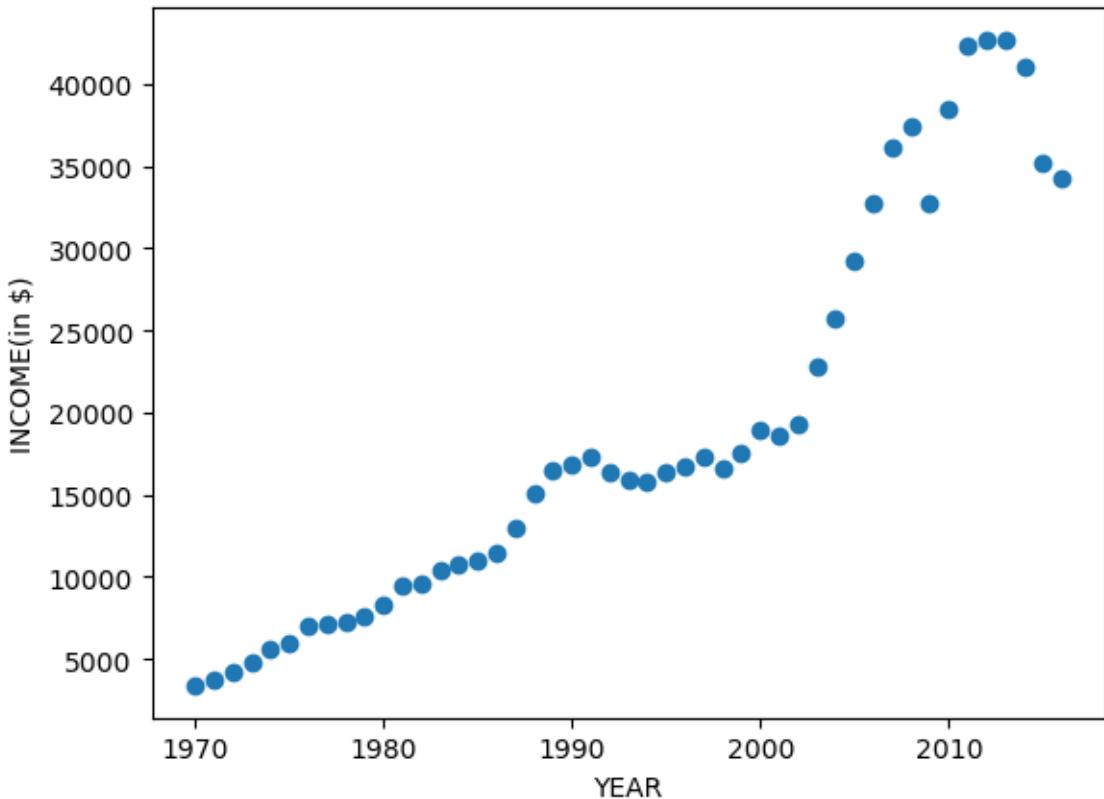
In [4]:

Out[4]:

In [5]:

In [7]:

Out[7]:



```
In [8]:
x = df.iloc[:,0:1]
y = df.iloc[:, -1]

y.head()
```

```
0    3399.299037
1    3768.297935
2    4251.175484
3    4804.463248
4    5576.514583
Name: per capita income (US$), dtype: float64
```

```
x.head()
```

	year
0	1970
1	1971
2	1972
3	1973
4	1974

```
In [12]:
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test =
train_test_split(x,y,test_size=0.2,random_state=2)
```

```
from sklearn.linear_model import LinearRegression
```

In [9]:

Out[9]:

In [11]:

Out[11]:

In [12]:

In [13]:

```
lr = LinearRegression()  
lr.fit(X_train,y_train)  
LinearRegression  
LinearRegression()
```

```
x_test.head()
```

	year
23	1993
12	1982
42	2012
16	1986
35	2005

```
y_test.head()
```

```
23    15875.586730  
12    9619.438377  
42    42665.255970  
16    11482.891530  
35    29198.055690  
Name: per capita income (US$), dtype: float64
```

```
lr.predict(X_test.iloc[0].values.reshape(1,1))  
D:\ProgramData\anaconda3\Lib\site-packages\sklearn\base.py:464:  
UserWarning: X does not have valid feature names, but LinearRegression was  
fitted with feature names  
warnings.warn(
```

```
array([18602.90731472])
```

```
plt.scatter(df['year'],df['per capita income (US$)'])  
plt.plot(X_train,lr.predict(X_train),color='red')  
plt.xlabel('YEAR')  
plt.ylabel('INCOME(in $)')
```

```
Text(0, 0.5, 'INCOME(in $)')
```

In [14]:

In [15]:

Out[15]:

In [16]:

Out[16]:

In [17]:

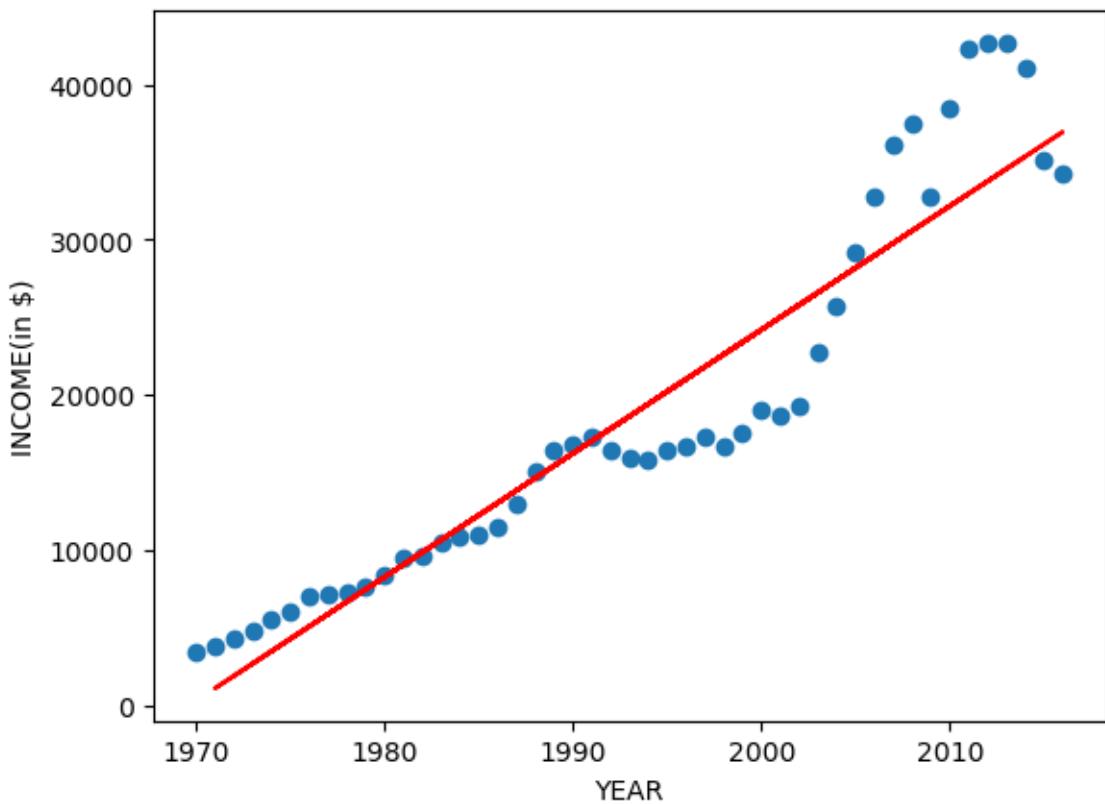
Out[17]:

In [18]:

Out[18]:

In [20]:

Out[20]:



```
m = lr.coef_
```

In [21]:

```
m
```

In [23]:

```
array([796.93681988])
```

Out[23]:

```
b = lr.intercept_
```

In [24]:

```
b
```

In [25]:

```
-1569692.1747126186
```

Out[25]:

by formula

```
dm=pd.read_csv('canada_per_capita_income.csv')
```

In [27]:

```
dm.head()
```

In [28]:

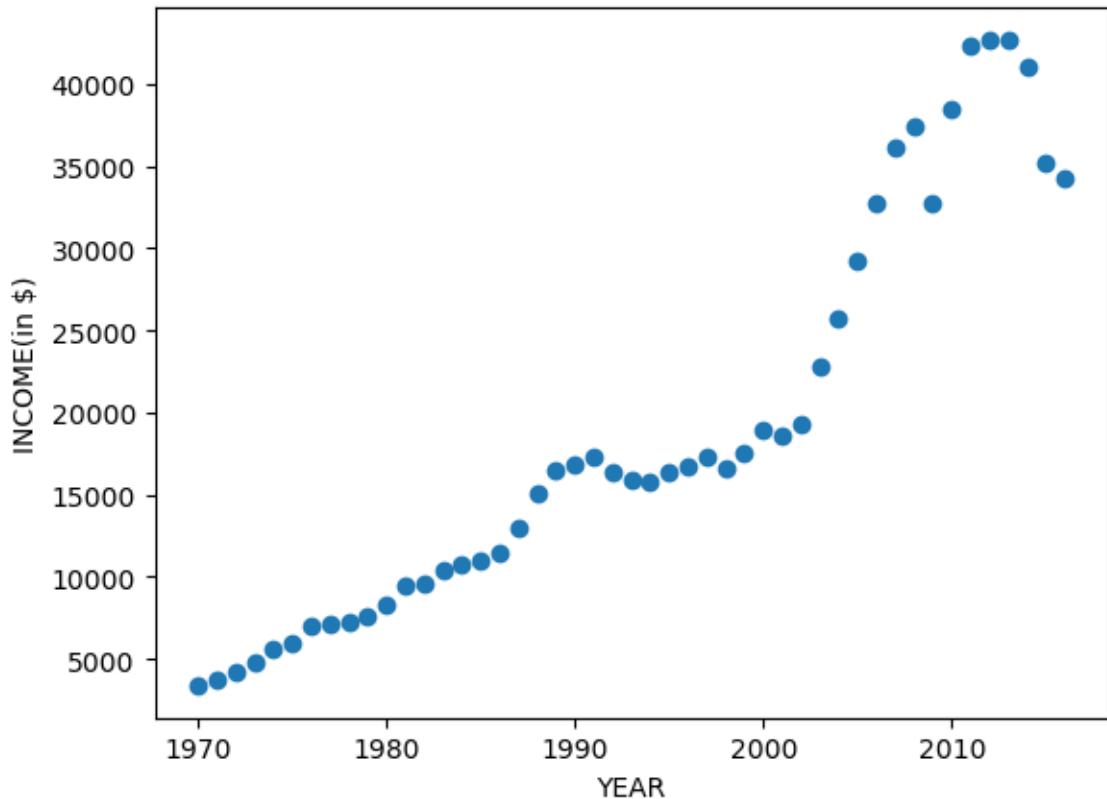
	year	per capita income (US\$)
0	1970	3399.299037
1	1971	3768.297935
2	1972	4251.175484
3	1973	4804.463248
4	1974	5576.514583

Out[28]:

In [29]:

```
plt.scatter(dm['year'],dm['per capita income (US$)'])  
plt.xlabel('YEAR')  
plt.ylabel('INCOME(in $)')  
Text(0, 0.5, 'INCOME(in $)')
```

Out[29]:



```
X = dm.iloc[:,0].values  
y = dm.iloc[:,1].values
```

In [35]:

X

```
array([[1970],  
       [1971],  
       [1972],  
       [1973],  
       [1974],  
       [1975],  
       [1976],  
       [1977],  
       [1978],  
       [1979],  
       [1980],  
       [1981],  
       [1982],  
       [1983],  
       [1984],  
       [1985],  
       [1986],
```

In [38]:

Out[38]:

```
[1987],  
[1988],  
[1989],  
[1990],  
[1991],  
[1992],  
[1993],  
[1994],  
[1995],  
[1996],  
[1997],  
[1998],  
[1999],  
[2000],  
[2001],  
[2002],  
[2003],  
[2004],  
[2005],  
[2006],  
[2007],  
[2008],  
[2009],  
[2010],  
[2011],  
[2012],  
[2013],  
[2014],  
[2015],  
[2016]], dtype=int64)
```

In [39]:

Y

```
array([ 3399.299037,  3768.297935,  4251.175484,  4804.463248,  
      5576.514583,  5998.144346,  7062.131392,  7100.12617 ,  
      7247.967035,  7602.912681,  8355.96812 ,  9434.390652,  
      9619.438377, 10416.53659 , 10790.32872 , 11018.95585 ,  
     11482.89153 , 12974.80662 , 15080.28345 , 16426.72548 ,  
     16838.6732 , 17266.09769 , 16412.08309 , 15875.58673 ,  
     15755.82027 , 16369.31725 , 16699.82668 , 17310.75775 ,  
     16622.67187 , 17581.02414 , 18987.38241 , 18601.39724 ,  
     19232.17556 , 22739.42628 , 25719.14715 , 29198.05569 ,  
     32738.2629 , 36144.48122 , 37446.48609 , 32755.17682 ,  
     38420.52289 , 42334.71121 , 42665.25597 , 42676.46837 ,  
     41039.8936 , 35175.18898 , 34229.19363 ])
```

Out[39]:

class MeraLR:

```
def __init__(self):
```

In [40]:

```

        self.m = None
        self.b = None

    def fit(self,X_train,y_train):

        num = 0
        den = 0

        for i in range(X_train.shape[0]):

            num = num + ((X_train[i] - X_train.mean())*(y_train[i] -
y_train.mean()))
            den = den + ((X_train[i] - X_train.mean())*(X_train[i] -
X_train.mean()))

        self.m = num/den
        self.b = y_train.mean() - (self.m * X_train.mean())
        print(self.m)
        print(self.b)

    def predict(self,X_test):

        print(X_test)

        return self.m * X_test + self.b

```

```

from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test =
train_test_split(X,y,test_size=0.2,random_state=2)

```

```
X_train.shape
```

```
(37, 1)
```

```
lr = MeraLR()
```

```

lr.fit(X_train,y_train)
[796.93681988]
[-1569692.17471262]

```

```
X_train.shape[0]
```

```
37
```

```
X_train[0]
```

```
array([1972], dtype=int64)
```

```
X_train.mean()
```

In [41]:

In [42]:

Out[42]:

In [43]:

In [44]:

In [45]:

Out[45]:

In [46]:

Out[46]:

In [47]:

Out[47]:

```
1992.6486486486488
```

```
x_test[0]
```

```
array([1993], dtype=int64)
```

```
print(lr.predict(x_test[0]))
```

```
[1993]
```

```
[18602.90731472]
```

In [48]:

Out[48]:

In [49]:

In []: