In [42]: import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt
 import seaborn as seabornInstance
 from sklearn.model_selection import train_test_split
 from sklearn.linear_model import LinearRegression
 from sklearn import metrics
 %matplotlib inline

In [43]: #read file
 df=pd.read_csv("USA_Housing.csv")
 df.head()

Out[43]:

	Avg_Area_Income	Avg_Area_House_Age	Avg_Area_Number_of_Rooms	Avg_Area_Number_of_Bec
0	79545.45857	5.682861	7.009188	
1	79248.64245	6.002900	6.730821	
2	61287.06718	5.865890	8.512727	
3	63345.24005	7.188236	5.586729	
4	59982.19723	5.040555	7.839388	
4				>

In [44]: #To see the statistical details of the dataset
df.describe()

Out[44]:

	Avg_Area_Income	Avg_Area_House_Age	Avg_Area_Number_of_Rooms	Avg_Area_Number_of
count	5000.000000	5000.000000	5000.000000	Ę
mean	68583.108984	5.977222	6.987792	
std	10657.991214	0.991456	1.005833	
min	17796.631190	2.644304	3.236194	
25%	61480.562390	5.322283	6.299250	
50%	68804.286405	5.970429	7.002902	
75%	75783.338665	6.650808	7.665871	
max	107701.748400	9.519088	10.759588	
4				>

```
In [45]: #plot our data points on a 2-D graph to eyeball our dataset and see if we can mar
df.plot(x='Avg_Area_House_Age', y='Price', style='o')
plt.title('Avg_Area_House_Age vs Price')
plt.xlabel('Avg_Area_House_Age')
plt.ylabel('Price')
plt.show()
```



```
In [46]: # the data into "attributes" and "labels".
#Attributes are the independent variables while labels are dependent variables wh
X = df['Avg_Area_House_Age'].values.reshape(-1,1)
y = df['Price'].values.reshape(-1,1)
```

In [47]: #we split 20% of the data to the training set while 80% of the data to test set u
#The test_size variable is where we actually specify the proportion of the test s
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.8, random_s

```
In [48]: #to train our algorithm. For that, we need to import LinearRegression class
    regressor = LinearRegression()
    regressor.fit(X_train, y_train) #training the algorithm
```

Out[48]: LinearRegression()

```
In [49]: #To retrieve the intercept:
    print(regressor.intercept_)
    #For retrieving the slope:
    print(regressor.coef_)
```

[309418.81589309] [[155235.07217097]]

```
In [50]: #we will use our test data and see how accurately our algorithm predicts
y_pred = regressor.predict(X_test)
```

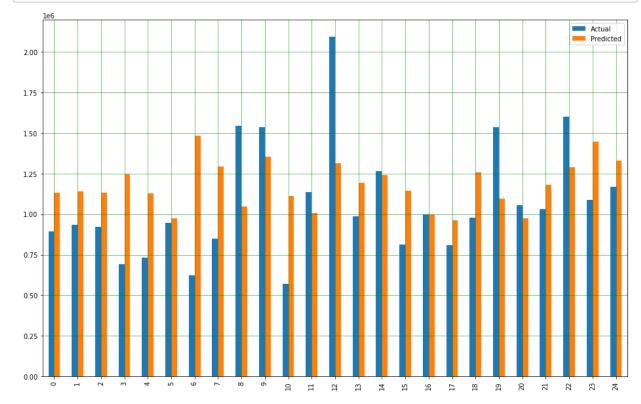
```
In [51]: #compare the actual output values for X_test with the predicted values
df = pd.DataFrame({'Actual': y_test.flatten(), 'Predicted': y_pred.flatten()})
df
```

Out[51]:

	Actual	Predicted
0	8.942511e+05	1.132117e+06
1	9.329794e+05	1.139098e+06
2	9.207479e+05	1.132215e+06
3	6.918549e+05	1.245093e+06
4	7.327332e+05	1.128760e+06
3995	9.792828e+05	1.227820e+06
3996	1.182888e+06	1.023557e+06
3997	1.423025e+06	1.277973e+06
3998	9.907257e+05	1.036107e+06
3999	1.086829e+06	1.167474e+06

4000 rows × 2 columns

```
In [52]: #visualize comparison result as a bar graph
     df1 = df.head(25)
     df1.plot(kind='bar',figsize=(16,10))
     plt.grid(which='major', linestyle='-', linewidth='0.5', color='green')
     plt.grid(which='minor', linestyle=':', linewidth='0.5', color='black')
     plt.show()
```



```
In [ ]:
```

In []: