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

from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn import metrics

In [2]: link = 'https://raw.githubusercontent.com/AdiPersonalWorks/Random/master/student_scores%20-%20student_scores.csv'
df = pd.read_csv(link)
print("successful import")
```

In [3]: df.head(5)

Out[3]:

Hours Scores

0 2.5 21

1 5.1 47

2 3.2 27 **3** 8.5 75

successful import

4 3.5 30

In [4]: df.head()

Out[4]:

	Hours	Scores
0	2.5	21
1	5.1	47
2	3.2	27
3	8.5	75
4	3.5	30

In [5]: df.tail()

Out[5]:

	Hours	Scores
20	2.7	30
21	4.8	54
22	3.8	35
23	6.9	76
24	7.8	86

```
In [6]: df.columns
 Out[6]: Index(['Hours', 'Scores'], dtype='object')
 In [7]: df.shape
 Out[7]: (25, 2)
 In [8]: df.info()
              <class 'pandas.core.frame.DataFrame'>
RangeIndex: 25 entries, 0 to 24
              Hangeindex: 25 entries, 0 to 24
Data columns (total 2 columns):

# Column Non-Null Count Dtype

O Hours 25 non-null float6
1 Scores 25 non-null int64
                                                           float64
              dtypes: float64(1), int64(1) memory usage: 528.0 bytes
 In [9]: df.describe()
 Out[9]:
                             Hours
               count 25.000000 25.000000
                mean 5.012000 51.480000
                std 2.525094 25.286887
                  min 1.100000 17.000000
                 25% 2.700000 30.000000
                 50% 4.800000 47.000000
               75% 7.400000 75.000000
                 max 9.200000 95.000000
In [10]: df.corr()
Out[10]:
                             Hours Scores
               Hours 1.000000 0.976191
                Scores 0.976191 1.000000
In [11]: # Plotting the distribution of scores
    df.plot(x='Hours', y='Scores', style='o')
    plt.title('Hours vs Percentage')
    plt.xlabel('Hours Studied')
    plt.ylabel('Percentage Score')
    plt.show()
                                             Hours vs Percentage
                          Scores
                   90
                   80
                Dercentage Score
                   30
```

Hours Studied

```
In [12]: sns.jointplot(y= 'Hours' ,x='Scores', data= df)

Out[12]: <seaborn.axisgrid.JointGrid at 0x16a77b6a100>

Data Preprocessing

In [13]: | x = df.iloc[:, :-1].values | y = df.iloc[:, 1].values |
```

Model Training

```
In [14]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=0)
    model = LinearRegression()
    model.fit(x_train.reshape(-1,1), y_train)
    print('done')
    done
```

Predictions

Out[16]:

	Reg	Predicted
0	20	16.884145
1	27	33.732261
2	69	75.357018
3	30	26.794801
4	62	60.491033

```
In [17]: #Estimating training and test score
print("Training Score:",model.score(x_train,y_train)*100,'%')
                Training Score: 95.15510725211553 %
In [18]: print("Test Score:",model.score(x_test,y_test) * 100,'%')
                Test Score: 94.54906892105356 %
In [19]: # Plotting the Bar graph to depict the difference between the actual and predicted value
               newDF.plot(kind='bar',figsize=(5,5))
plt.grid(which='major', linewidth='0.5', color='black')
plt.grid(which='minor', linewidth='0.5', color='black')
plt.show()
                                                                  Reg
Predicted
                  70
                  60
                  50
                  40
                  20
                  10
 In [20]: # testing the given sample
hours = 9.25
                 newTest=np.array([hours])
newTest = newTest.reshape(-1,1)
newPred = model.predict(newTest)
 In [21]: print("No of Hours = {}".format(hours))
print("Predicted Score = {}".format(newPred[0]))
                 No of Hours = 9.25
Predicted Score = 93.69173248737538
 In [22]: print('Mean Absolute Error:',metrics.mean_absolute_error(y_test, y_pred))
    print('Mean Squared Error:', metrics.mean_squared_error(y_test, y_pred))
    print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test, y_pred)))
    print('R-2:', metrics.r2_score(y_test, y_pred))
```

Mean Absolute Error: 4.183859899002975 Mean Squared Error: 21.5987693072174 Root Mean Squared Error: 4.6474476121003665 R-2: 0.9454906892105356