17-studentsmarksprediction

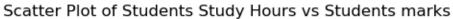
September 21, 2025

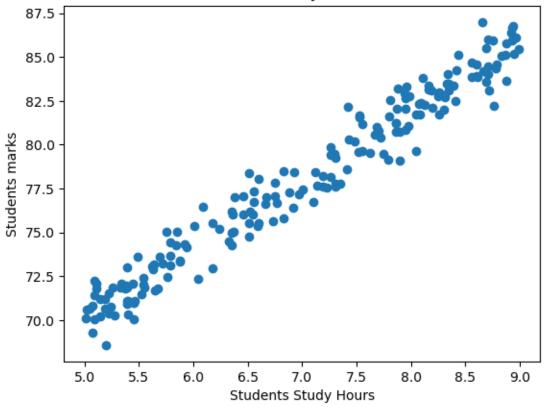
1 STUDENTS MARKS PREDICTION

```
[1]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     %matplotlib inline
[2]: df = pd.read_csv(r"D:\Naresh It Classes\4. September\17- Production, Into to_
      →AI\Students Marks Prediction\student_info.csv")
[3]: df.head()
[3]:
        study_hours student_marks
     0
               6.83
                             78.50
     1
               6.56
                             76.74
     2
                NaN
                             78.68
     3
               5.67
                             71.82
               8.67
                             84.19
[4]: df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200 entries, 0 to 199
    Data columns (total 2 columns):
        Column
                        Non-Null Count
                                        Dtype
     0
         study_hours
                        195 non-null
                                         float64
         student_marks 200 non-null
                                         float64
    dtypes: float64(2)
    memory usage: 3.3 KB
[6]: df.shape
[6]: (200, 2)
[7]: df.describe()
```

```
[7]:
            study_hours
                          student_marks
             195.000000
                              200.00000
     count
               6.995949
                               77.93375
     mean
     std
               1.253060
                                4.92570
               5.010000
     min
                               68.57000
     25%
               5.775000
                               73.38500
     50%
               7.120000
                               77.71000
     75%
               8.085000
                               82.32000
     max
               8.990000
                               86.99000
```

```
[8]: plt.scatter(x=df.study_hours, y=df.student_marks)
   plt.xlabel("Students Study Hours")
   plt.ylabel("Students marks")
   plt.title("Scatter Plot of Students Study Hours vs Students marks")
   plt.show()
```





1.1 Data Cleaning

```
[9]: df.isna()
 [9]:
           study_hours
                        student_marks
                 False
                                 False
      1
                 False
                                 False
      2
                  True
                                 False
      3
                 False
                                 False
      4
                 False
                                 False
      195
                 False
                                 False
                 False
                                 False
      196
      197
                 False
                                 False
      198
                 False
                                 False
      199
                 False
                                 False
      [200 rows x 2 columns]
[10]: df.isna().sum()
[10]: study_hours
                       5
      student_marks
                       0
      dtype: int64
[11]: df2=df.fillna(df.mean())
[12]: df2.isna().sum()
[12]: study_hours
                       0
      student_marks
                       0
      dtype: int64
          Split data
[13]: X = df2.drop('student_marks', axis='columns')
      y = df.drop('study_hours', axis='columns')
      print("shape of X = ", X.shape)
      print("shape of y = ", y.shape)
     shape of X = (200, 1)
     shape of y = (200, 1)
[14]: 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=0)
      print("shape of X_train = ", X_train.shape)
```

```
print("shape of y_train = ", y_train.shape)
      print("shape of X_test = ", X_test.shape)
      print("shape of y_test = ", y_test.shape)
     shape of X_{train} = (160, 1)
     shape of y_{train} = (160, 1)
     shape of X_{test} = (40, 1)
     shape of y_{test} = (40, 1)
     1.3 Choosing Linear Regression model and training it
[17]: from sklearn.linear_model import LinearRegression
      lr = LinearRegression()
[18]: lr.fit(X_train,y_train)
[18]: LinearRegression()
[24]: m = lr.coef_
      m
[24]: array([[3.93037294]])
[25]: c = lr.intercept_
      С
[25]: array([50.45063632])
[27]: y = m*4 + c #for 4 hours of study
      у
[27]: array([[66.17212807]])
[28]: lr.predict([[4]])[0][0].round(2)
     D:\Program Files\Lib\site-packages\sklearn\base.py:493: UserWarning: X does not
     have valid feature names, but LinearRegression was fitted with feature names
       warnings.warn(
[28]: 66.17
[29]: y_pred = lr.predict(X_test)
      y_pred
[29]: array([[83.50507271],
             [70.84927186],
             [72.93236952],
```

```
[73.20749562],
             [84.48766595],
             [80.12495199],
             [81.85431608],
             [80.91102657],
             [82.20804964],
             [78.98514384],
             [84.84139951],
             [77.84533568],
             [77.68812077],
             [83.22994661],
             [85.78468901],
             [84.9593107],
             [72.61793968],
             [78.71001773],
             [79.18166248],
             [84.2911473],
             [85.6274741],
             [74.74034107],
             [81.3433676],
             [72.02838374],
             [80.40007809],
             [78.98514384],
             [82.09013845],
             [77.94732382],
             [82.24735337],
             [75.44780819],
             [84.60557713],
             [71.63534645],
             [75.48711192],
             [70.29901965],
             [78.98514384],
             [75.32989701],
             [84.52696967],
             [74.07217767],
             [71.4388278]])
[30]: pd.DataFrame(np.c_[X_test, y_test, y_pred], columns = ["study_hours",__

¬"student_marks_original", "student_marks_predicted"])
[30]:
          study_hours
                        student_marks_original
                                                 student_marks_predicted
      0
             8.410000
                                          82.50
                                                                83.505073
      1
             5.190000
                                          71.18
                                                                70.849272
      2
             5.720000
                                          73.25
                                                                72.932370
      3
             8.880000
                                          83.64
                                                                85.352348
      4
             5.790000
                                          73.64
                                                                73.207496
```

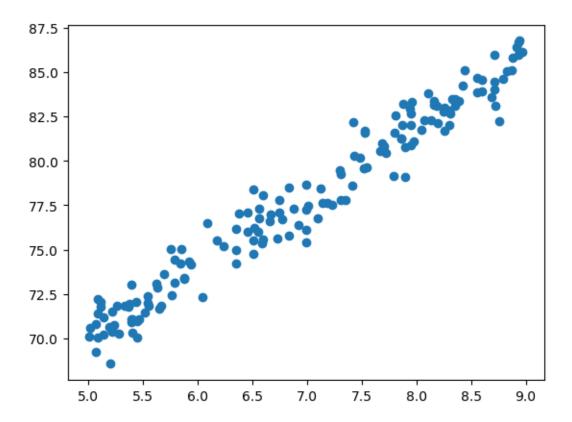
[85.35234799],

```
5
       8.660000
                                     86.99
                                                           84.487666
6
                                     81.18
       7.550000
                                                           80.124952
7
       7.990000
                                     82.75
                                                           81.854316
8
       7.750000
                                     79.50
                                                           80.911027
9
       8.080000
                                     81.70
                                                           82.208050
10
       7.260000
                                     79.41
                                                           78.985144
11
       8.750000
                                     85.95
                                                           84.841400
12
       6.970000
                                    77.19
                                                           77.845336
13
                                                           77.688121
       6.930000
                                     78.45
14
       8.340000
                                     84.00
                                                           83.229947
15
                                     85.46
                                                           85.784689
       8.990000
16
       8.780000
                                     84.35
                                                           84.959311
17
       5.640000
                                     73.19
                                                           72.617940
18
       7.190000
                                     78.21
                                                           78.710018
19
                                     77.59
       7.310000
                                                           79.181662
20
       8.610000
                                     83.87
                                                           84.291147
21
       8.950000
                                     85.15
                                                           85.627474
22
       6.180000
                                     72.96
                                                           74.740341
23
       7.860000
                                     80.72
                                                           81.343368
24
       5.490000
                                     73.61
                                                           72.028384
25
       7.620000
                                     79.53
                                                           80.400078
26
       7.260000
                                     78.17
                                                           78.985144
27
       8.050000
                                     79.63
                                                           82.090138
       6.995949
28
                                     76.83
                                                           77.947324
29
       8.090000
                                     82.38
                                                           82.247353
30
       6.360000
                                     76.04
                                                           75.447808
31
       8.690000
                                     85.48
                                                           84.605577
32
       5.390000
                                     71.87
                                                           71.635346
33
       6.370000
                                     75.04
                                                           75.487112
34
       5.050000
                                     70.67
                                                           70.299020
35
       7.260000
                                     79.87
                                                           78.985144
36
       6.330000
                                     74.49
                                                           75.329897
37
       8.670000
                                                           84.526970
                                     84.19
38
                                                           74.072178
       6.010000
                                     75.36
39
       5.340000
                                     72.10
                                                           71.438828
```

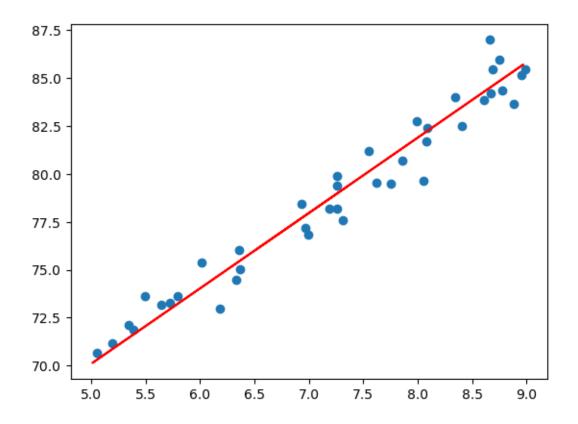
[31]: lr.score(X_test,y_test) #R2

[31]: 0.9521841793508595

[35]: plt.scatter(X_train,y_train)
plt.show()



```
[34]: plt.scatter(X_test, y_test)
plt.plot(X_train, lr.predict(X_train), color = "r")
plt.show()
```



1.4 Save the model as a Pickle File

```
[36]: import joblib
    joblib.dump(lr, "student_mark_predictor.pkl")

[36]: ['student_mark_predictor.pkl']

[37]: model = joblib.load("student_mark_predictor.pkl")

[38]: model.predict([[5]])[0][0]

D:\Program Files\Lib\site-packages\sklearn\base.py:493: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names warnings.warn(

[38]: 70.10250100162847

[]:
```