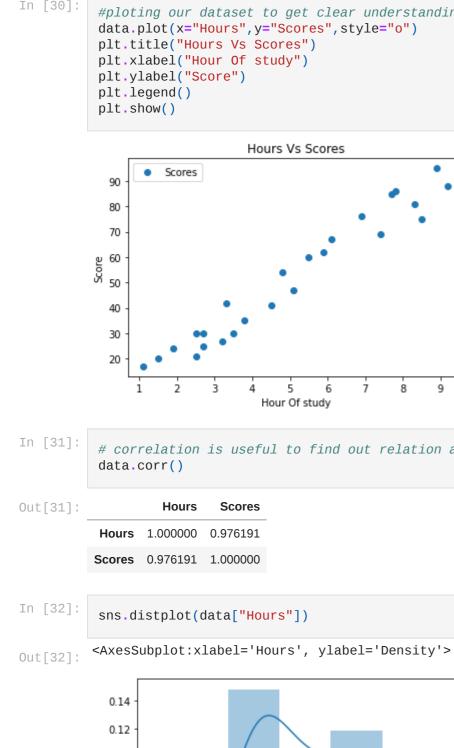
GRADUATE ROTATIONAL INTERNSHIP PROGRAM (GRIP) THE SPARK FOUNDATION NAME: AKASH GAWAS TASK1:PREDICTION USING SUPERVISED ML Predict the percentage of student on basis of how many hour in a day they study. In [24]: #Import libraries import warnings warnings.filterwarnings("ignore") import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt import sklearn %matplotlib inline In [25]: #import dataset data=pd.read_csv("My_data.csv") data.head() **Hours Scores** Out[25]: 2.5 21 5.1 47 3.2 27 3.5 30 In [26]: #check the information about our dataset data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 25 entries, 0 to 24 Data columns (total 2 columns): # Column Non-Null Count Dtype 0 Hours 25 non-null float64 1 Scores 25 non-null dtypes: float64(1), int64(1)memory usage: 464.0 bytes In [27]: data.describe() Out[27]: Hours Scores count 25.000000 25.000000 mean 5.012000 51.480000 std 2.525094 25.286887 1.100000 17.000000 **25%** 2.700000 30.000000 4.800000 47.000000 7.400000 75.000000 9.200000 95.000000 In [28]: #checking for missing or null value are present or not data.isna().sum() 0 Hours Out[28]: Scores 0 dtype: int64 In [29]: data.shape (25, 2)Out[29]: In [30]: #ploting our dataset to get clear understanding about our dataset data.plot(x="Hours", y="Scores", style="o") plt.title("Hours Vs Scores") plt.xlabel("Hour Of study") plt.ylabel("Score") plt.legend() plt.show() Hours Vs Scores Scores 80 70 9 60 50 50 40 30 20 Hour Of study In [31]: # correlation is useful to find out relation among them. data.corr() Out[31]: Hours Scores **Hours** 1.000000 0.976191



From the above graph we conclude that hour of study and score are strongly corelation with each other

100

x_train, x_test, y_train, y_test=train_test_split(x, y, test_size=0.2, random_state=0)

0.10

0.08

0.06

0.02

0.0175

0.0150

0.0125

0.0100

0.0075

0.0050

0.0025

0.0000

y_test

19

16

11

20 27

69

30 62

Name: Scores, dtype: int64

model=LinearRegression() model.fit(x_train,y_train)

y_pred=model.predict(x_test)

#visualize the training test result

plt.scatter(x_train, y_train, color="blue")

plt.scatter(x_test, y_test, color="blue") plt.plot(x_test, y_pred, color="red")

plt.title("Hours vs Scores (Testing set)")

Hours vs Scores (Testing set)

Hours

df=pd.DataFrame({"Actual":y_test, "Predict":y_pred})

plt.title("Hours vs Scores (Training set)")

plt.plot(x_train, model.predict(x_train), color="red")

Hours vs Scores (Training set)

training our model

LinearRegression()

plt.xlabel("Hours") plt.ylabel("Score")

plt.xlabel("Hours") plt.ylabel("Score")

#pred=algo.predict(x_test)

Predict

20 16.884145 27 33.732261 69 75.357018

30 26.794801 62 60.491033

prediction=model.predict([[9.5]])

plt.show()

70

60

40

30

20

Actual

prediction

0.0200 0.0175 0.0150

0.0125 خِي 0.0100 0.0075 0.0050 0.0025 0.0000

h=9.25

array([96.16939661])

sns.distplot(y_pred)

<AxesSubplot:ylabel='Density'>

from sklearn import metrics

s=model.predict([[h]])

Mean Absolute Error: 4.18385989900298

print('Mean Absolute Error:', metrics.mean_absolute_error(y_test, y_pred))

print(f"If students studies for {h} hour per/day then he/she will score {s}% marks")

If students studies for 9.25 hour per/day then he/she will score [93.69173249]% marks

16

In [41]:

In [42]:

Out[42]:

In [43]:

Out[43]:

In [44]:

Out[44]:

In [45]:

In [46]:

plt.show()

80

60

20

In [40]:

-20

x=data.iloc[:,:-1] y=data.iloc[:,-1]

Now we building linear regression model

#splitting the dataset into train and test set from sklearn.model_selection import train_test_split

 $\textbf{from} \ \text{sklearn.linear_model} \ \textbf{import} \ \text{LinearRegression}$

sns.distplot(data["Scores"]);

In [33]:

In [34]:

In [35]:

In [36]:

In [37]:

Out[37]:

In [38]:

In [39]: