Task-1

## Analysis of percentage of an student based on the no. of study hours

Import the necessary libraries import pandas as pd In [29]: import numpy as np import matplotlib.pyplot as plt import seaborn as sns %matplotlib inline storing the data in a csv file

In [2]: data=pd.read\_csv("http://bit.ly/w-data") In [3]: data

**Hours Scores** Out[3]: 2.5 21 5.1 47 3.2 27 2 8.5 75

3.5 30 1.5 20 6 9.2 88 5.5 60 8 8.3 81 2.7 25 10 7.7 85 62 11 5.9 41 12 4.5 13 3.3 14 1.1 17 15 8.9 95 2.5 30 16 17 1.9 24 18 6.1 67 19 7.4 2.7 30 20 4.8

27 75

1.100000 17.000000

2.700000 30.000000

4.800000 47.000000

data['Scores'].unique()

min

**25**% **50%** 

In [7]:

Out[7]:

In [8]

In [22]:

In [26]:

y\_pred

35

76

86

Description of the dataset

22

24

3.8

6.9

7.8

3.2

8.5

data.head() **Hours Scores** Out[4]: 2.5 5.1 47

30 In [5]: data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 25 entries, 0 to 24 Data columns (total 2 columns): Column Non-Null Count Dtype Hours 25 non-null float64 0

Scores 25 non-null int64 dtypes: float64(1), int64(1)memory usage: 528.0 bytes data.describe() Hours Out[6]: **Scores** count 25.000000 25.000000 5.012000 51.480000 mean 2.525094 25.286887 std

> 7.400000 75.000000 9.200000 95.000000 max => The dataset contains 25 rows and 2 columns => The first column denote the number of hour spend by the students.It contains quantitative values. = >The second column denote the marks obtained by the student.It contains quantiative values.

= >There is zero null values in both columns. Plotting the data

data['Hours'].unique()

array([2.5, 5.1, 3.2, 8.5, 3.5, 1.5, 9.2, 5.5, 8.3, 2.7, 7.7, 5.9, 4.5,

3.3, 1.1, 8.9, 1.9, 6.1, 7.4, 4.8, 3.8, 6.9, 7.8])

Out[8]: array([21, 47, 27, 75, 30, 20, 88, 60, 81, 25, 85, 62, 41, 42, 17, 95, 24, 67, 69, 54, 35, 76, 86], dtype=int64)

In [9]: sns.lineplot(data=data, x="Hours", y="Scores") Out[9]: <AxesSubplot:xlabel='Hours', ylabel='Scores'>

90 80 70 £ 60 й<sub>50</sub> 40 30 We can clearly notice a linear relation between hours and scores

X\_data=data.iloc[:, :-1].values Y\_data=data.iloc[:, 1].values

splitting the dataset into train and test dataset

from sklearn.model\_selection import train\_test\_split In [23]:

Applying the Linear Regression Model

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X\_data, Y\_data, test\_size=0.2, random\_state=0)

## In [24]: from sklearn.linear\_model import LinearRegression regressor = LinearRegression() regressor.fit(X\_train, y\_train)

Out[24]: LinearRegression()

y\_pred=regressor.predict(X\_test) In [25]:

array([16.88414476, 33.73226078, 75.357018 , 26.79480124, 60.49103328])

y\_test In [27]:

array([20, 27, 69, 30, 62], dtype=int64)

plt.scatter(X\_test,y\_pred) In [36]: plt.scatter(X\_test,y\_test) plt.legend(['y\_pred','y\_test']) Out[36]: <matplotlib.legend.Legend at 0x1e2640a4b20>

y\_pred y\_test 60 50 40 30 20

Evaluating the model In [33]: from sklearn import metrics 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)))

Root Mean Squared Error: 4.647447612100373

Mean Absolute Error: 4.183859899002982 Mean Squared Error: 21.598769307217456