The Sparks Foundation - # Aug21

import numpy as np import pandas as pd

# Reading the data

data.head(10)

**Hours Scores** 

5.1

3.2

8.5

3.5

1.5

9.2

5.5 8.3

2.7

data.shape

In [7]:

Out[7]:

In [8]:

In [9]:

Out[9]:

In [10]:

In [11]:

In [12]:

In [13]:

In [14]:

Out[8]: (25, 2)

import seaborn as sns

url='http://bit.ly/w-data' data=pd.read\_csv(url)

21

47

27

75

30

20

88 60

81

25

# shape of the dataset

# Describing the data

count 25.000000 25.000000 mean 5.012000 51.480000

std 2.525094 25.286887

**75**% 7.400000 75.000000

max 9.200000 95.000000

Scores

1.100000 17.000000

2.700000 30.000000 4.800000 47.000000

Scores

# Plotting the distribution of scores

plt.title('Hours vs Percentage') plt.xlabel('Hours Studied') plt.ylabel('Percentage Score')

data.plot(x='Hours', y='Scores', style='\*')

Hours vs Percentage

Hours Studied

# x is independent variable and y is dependent

# iloc[] indexing of dataset X = data.iloc[: , :-1].valuesy = data.iloc[: , 1].values

Extracting Independent and Dependent Variable

[21 47 27 75 30 20 88 60 81 25 85 62 41 42 17 95 30 24 67 69 30 54 35 76

test\_size=0.25, random\_state=42)

Spliting the data into traning and test set

# train\_test\_split randomly shuffles and divides all the rows

Training and implementing the model

from sklearn.model\_selection import train\_test\_split X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y,

# Importing the linear regression model

regressor = LinearRegression() regressor.fit(X\_train, y\_train)

# Plotting the regression line

# Plotting for the test data

print("Training complete.")

Training complete.

plt.scatter(X, y) plt.plot(X, line);

plt.show()

print(X\_test)

print(y\_pred)

34.53687133]

Comparing

Actual Predicted

81 83.107332 30 26.765598

21 26.765598 76 69.507603 62 59.793511 25 28.708416 42 34.536871

Model evaluation

# prediction of values

y\_pred = regressor.predict(X\_test)

# Comparing Actual vs Predicted values

[83.10733229 26.76559757 26.76559757 69.50760322 59.79351103 28.70841601

from sklearn.metrics import mean\_absolute\_error, mean\_squared\_error,r2\_score

print("Mean Absolute Error:" , mean\_absolute\_error(y\_test,y\_pred))
print("mean squared Error" , mean\_squared\_error(y\_test,y\_pred))

print("R^2 score:", r2\_score(y\_test,y\_pred))

print("predicted percentage: %.2f"%(own\_pred))

#print("Predicted Score = {}".format(own\_pred[0]))

metrics.mean\_absolute\_error(y\_test, y\_pred))

Mean Absolute Error: 4.425394675156183 mean squared Error 23.516396034595225

hours = float(input("enter the hour")) own\_pred = regressor.predict([[hours]])

#print("No of Hours = {}".format(hours))

R^2 score: 0.9553509219739938

predicted percentage: 94.76

from sklearn import metrics print('Mean Absolute Error:',

Mean Absolute Error: 4.425394675156183

1) A student will score 80.19 % if he/she study for 8 hours

2) The r^2 value for this Linear Regression model is 0.955

enter the hour9.5

Conclusion:

df = pd.DataFrame({'Actual': y\_test, 'Predicted': y\_pred})

[[8.3] [2.5] [2.5] [6.9] [5.9] [2.7] [3.3]]

60

40

In [15]:

In [16]:

In [17]:

Out[17]:

In [18]:

In [21]:

In [20]:

from sklearn.linear\_model import LinearRegression

line = regressor.coef\_\*X+regressor.intercept\_

data.describe()

min 25%

plt.show()

90 80

> 30 20

print(X) print() print(y)

[[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] [2.5] [1.9] [6.1][7.4][2.7] [4.8] [3.8] [6.9][7.8]]

# printing the first 10 elements.

we have succesfully imported the data

from matplotlib import pyplot as plt

regression task as it involves just two variables.

# Importing the libraries required for the task

print("we have succesfully imported the data")

Simple Linear Regression In this regression task we will predict the percentage of marks that a student is expected to score based upon the number of hours they studied. This is a simple linear

Task 1: Prediction using supervised ML

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