**GRIP**: The Sparks Foundation Data Science and Business Analytics Internship Task 1: Prediction Using Supervised Machine Learning Predict the percentage of an student based on the number of study hours Python 3 Jupyter Notebook Linear Regression Chetana Thorat Step 1: Import the Dataset Step 2: Visualize and Analyze the Dataset Step 3: Preparation of Data Step 4: Design and Train the Machine Learning Model Step 5: Visualize the Model Step 6: Make Predictions Step 7: Evaluate the model In [45]: import pandas as pd import numpy as np import matplotlib.pyplot as plt In [46]: # Reading data from remote link using the url url = "http://bit.ly/w-data" student\_data = pd.read\_csv(url) print("Data imported successfully") student\_data Data imported successfully Out[46]: **Hours Scores** 2.5 0 21 47 5.1 3.2 27 75 3.5 30 1.5 20 9.2 88 60 8 8.3 81 25 7.7 85 10 11 62 4.5 41 12 13 3.3 42 14 1.1 17 15 8.9 95 2.5 30 16 17 1.9 24 18 6.1 67 19 7.4 69 20 2.7 30 21 4.8 54 22 3.8 35 6.9 76 24 7.8 86 student\_data.shape #Here we can see that there are 25 rows and 2 colums in the dataset Out[47]: (25, 2) student\_data.describe() Out[48]: Scores 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 4.800000 47.000000 7.400000 75.000000 9.200000 95.000000 In [53]: student\_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 int64 dtypes: float64(1), int64(1)memory usage: 528.0 bytes In [54]: student\_data.isnull().sum() #Here,we can see that there are no NULL VALUES in the dataset that can affect the training of our algorithm Out[54]: Hours Scores dtype: int64 student\_data.corr(method = 'pearson') Out[79]: Hours Scores **Hours** 1.000000 0.976191 Scores 0.976191 1.000000 In [80]: student\_data.corr(method='spearman') Out[80]: Hours Scores Hours 1.000000 0.971891 **Scores** 0.971891 1.000000 # Plotting the distribution of scores and number of hours studied on a 2D graph student\_data.plot(x='Hours', y='Scores', style='ro') plt.title('Hours vs Percentage') plt.xlabel('Hours Studied') plt.ylabel('Percentage Score') plt.show() Hours vs Percentage Scores 80 rcentage Score 2 40 30 20 In [56]: #we are ectracting values of Hours Data into variable X and the values of Score Data into variable Y x = student\_data.iloc[:, :-1].values y = student\_data.iloc[:, 1].values In [57]: #number of hours studied Out[57]: 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], [2.5], [1.9], [6.1], [7.4], [2.7], [4.8], [3.8], [6.9], [7.8]]) In [58]: #Score obtained Out[58]: array([21, 47, 27, 75, 30, 20, 88, 60, 81, 25, 85, 62, 41, 42, 17, 95, 30, 24, 67, 69, 30, 54, 35, 76, 86], dtype=int64) In [59]: #We now split the data into train and test dataset using Scikit-Learn's --built-in train\_test\_split() 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) In [60]: #no of rows in training dataset (x-colums) x\_train.shape Out[60]: (20, 1) In [61]: #no of rows in training dataset (y-colums) y\_train.shape Out[61]: (20,) In [62]: #no of rows in testing dataset (x-colums) x\_test.shape Out[62]: **(5, 1)** In [63]: #no of rows in testing dataset (y-colums) y\_test.shape Out[63]: (5,) In [64]: x\_train [6.9], [1.1], [5.1], [7.7], [3.3], [8.3], [9.2], [6.1], [3.5], [2.7], [5.5], [2.7], [8.5], [2.5], [4.8], [8.9], [4.5]]) In [65]: y\_train Out[65]: array([35, 24, 86, 76, 17, 47, 85, 42, 81, 88, 67, 30, 25, 60, 30, 75, 21, 54, 95, 41], dtype=int64) In [66]: x\_test Out[66]: array([[1.5], [3.2], [7.4], [2.5], [5.9]]) In [67]: y\_test Out[67]: array([20, 27, 69, 30, 62], dtype=int64) In [68]: from sklearn.linear\_model import LinearRegression regressor = LinearRegression() regressor.fit(x\_train, y\_train) print("Training complete") Training complete In [69]: # Plotting the regression line (y=mx+c) line = regressor.coef\_\*x+regressor.intercept\_ # Plotting for the test data plt.scatter(x, y) plt.plot(x, line,color='red'); plt.show() 80 60 20 · print(x\_test) # Testing data - In Hours [[1.5] [3.2] [7.4] [2.5] [5.9]] In [71]: print(y\_test) # Testing data - In percentage [20 27 69 30 62] # Predicting the scores y\_pred = regressor.predict(x\_test) print(y\_pred) [16.88414476 33.73226078 75.357018 26.79480124 60.49103328] In [73]: # Comparing Actual vs Predicted compdata = pd.DataFrame({'Actual': y\_test, 'Predicted': y\_pred}) Out[73]: **Actual Predicted** 20 16.884145 27 33.732261 69 75.357018 30 26.794801 62 60.491033 #Testing with the custom of 9 hrs/day hours = 9.25 own\_pred = regressor.predict([[hours]]) print(f"No of Hours = {hours}") print(f"Predicted Score = {own\_pred[0]}") No of Hours = 9.25Predicted Score = 93.69173248737538 from sklearn import metrics print('Mean Absolute Error:', metrics.mean\_absolute\_error(y\_test, y\_pred)) Mean Absolute Error: 4.183859899002975 print('Max Error:',metrics.max\_error(y\_test, y\_pred)) Max Error: 6.732260779489842

print('Mean Squared Error:', metrics.mean\_squared\_error(y\_test, y\_pred))

Mean Squared Error: 21.5987693072174

In [ ]: