The Sparks Foundation Data Science and Business Analytics (Grip 2021) Task 1: Prediction using Supervised ML Problem: predict the percentage of students based on no. of study hours. Author: Venepally Tharun 1. Importing all Libraries required #importing libraries import pandas as pd import numpy as np import matplotlib.pyplot as plt %matplotlib inline import seaborn as sns #for data visualisation. 1. Reading Dataset In [2]: #Reading data from link url = r"http://bit.ly/w-data" df = pd.read\_csv(url) print("Data imported successfully") Data imported successfully #printing first 10 rows of the data set. In [3]: df.head(10) **Hours Scores** Out[3]: 2.5 21 5.1 1 47 2 3.2 27 3 8.5 75 4 3.5 30 5 1.5 20 6 9.2 88 5.5 60 8.3 81 2.7 25 #checking the number of rows and columns. In [4]: df.shape Out[4]: (25, 2) df.describe() In [5] Out[5]: Hours **Scores** count 25.000000 25.000000 5.012000 51.480000 mean 2.525094 25.286887 1.100000 17.000000 min 2.700000 30.000000 25% **50**% 4.800000 47.000000 7.400000 75.000000 **75**% 9.200000 95.000000 df.info() In [6]: <class 'pandas.core.frame.DataFrame'> RangeIndex: 25 entries, 0 to 24 Data columns (total 2 columns): Column 25 non-null float64 Hours Scores 25 non-null int64 dtypes: float64(1), int64(1)memory usage: 528.0 bytes #check if there is any null value in the dataset. In [7]: df.isnull().sum() Out[7]: Hours Scores 0 dtype: int64 3. DATA VISUALISATION #Plotting the graph to see the relation and distribution of datapoints. df.plot(x = 'Hours', y = 'Scores', figsize = (9,5), style = '\*')plt.title('Hours vs Percentage') plt.xlabel('Hours studied') plt.ylabel('Percentage scores') plt.show() Hours vs Percentage Scores 90 80 70 Percentage scores 60 50 40 30 20 Hours studied from the above graph we can see that there is positive linear relation between the no of hours studied and percentage scores. sns.regplot( x = 'Hours', y = 'Scores', data = df) Out[9]: <AxesSubplot:xlabel='Hours', ylabel='Scores'> 100 80 60 #Distribution of hours data In [10]: plt.hist(x = 'Hours', data = df)plt.show() 4.0 3.5 2.5 2.0 1.5 1.0 0.5 In [11]: #Distribution of scores data plt.hist(x = 'Scores', data = df)plt.show() 4 3 2 1 0 20 30 40 60 70 80 90 4. DATA PREPARATION #dividing the datasets int attributes(inputs) and labels(columns) In [12]: x = df.iloc[:,:-1].valuesy = df.iloc[:,1].valuesslpit this data into train and testdata using the train\_test\_split() method from the skicit learn library. In [13]: from sklearn.model\_selection import train\_test\_split  $x_{train}, x_{test}, y_{train}, y_{test} = train_{test}, y_{test} = 0.2, random_{state} = 0)$ 5 . TRAINING THE MODEL we have to train our data using Linear regression algorithm and check the values for test data from sklearn.linear\_model import LinearRegression In [14]: lr = LinearRegression() lr.fit(x\_train,y\_train) LinearRegression(copy\_X=True, fit\_intercept=True, n\_jobs=None, normalize=False) Out[14]: LinearRegression() plot the regression line in the scatter In [15]: r\_line = lr.coef\_\*x+lr.intercept\_ #visualising the training dataset In [16]: plt.scatter(x\_train, y\_train) plt.title('Training set') plt.plot(x\_train, lr.predict(x\_train)) plt.xlabel("Hours") plt.ylabel("Scores") Out[16]: Text(0, 0.5, 'Scores') Training set 80 Ŋ 40 20 6. MAKING PREDICTIONS we have done tarining and we have to make some predictions. y\_pred = lr.predict(x\_test) In [17]: y\_pred Out[17]: array([16.88414476, 33.73226078, 75.357018 , 26.79480124, 60.49103328]) compare actual values with predicted values df = pd.DataFrame({"Actual":y\_test, "Predicted":y\_test}) df Out[18]: **Actual Predicted** 20 20 27 1 27 2 69 69 30 30 62 62 In [19]: #score for test data accuracy = lr.score(x\_test, y\_test) print("Accuracy:", accuracy \* 100) Accuracy: 94.54906892105356 what is the predicted score if student studies for 9.25 hours per day? # we can test for any input In [20]: # here we are calculating the score for 9.25 hours hours = [[9.25]]pred = lr.predict(hours) pred Out[20]: array([93.69173249]) 1. MODEL EVALUATION In this step we have to evaluate the performance of algorithm. we can evaluate this by calculating mean squared error or mean absolute error. MEAN ABSOLUTE ERROR In [21]: from sklearn import metrics MAE = metrics.mean\_absolute\_error(y\_test,y\_pred) print("Mean Absolute Error: ", MAE) Mean Absolute Error: 4.183859899002975 MEAN SQUARED ERROR In [22]: from sklearn import metrics MSE = metrics.mean\_squared\_error(y\_test,y\_pred) print("Mean Squared Error: ", MSE) Mean Squared Error: 21.5987693072174 Small value of Mean Absolute Error states that the model is quite good. THANK YOU.