Ritu Upadhyay Data Science & Business Analytics Intern (Batch - May 2021) Task 1: Prediction using Supervised ML ---Linear Regression with Python Scikit Learn---In this section we will see how the Python Scikit-Learn library for machine learning can be used to implement regression functions. We will start with simple linear regression involving two variables. ---Problem statement---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 regression task as it involves just two variables. --To Predict--What will be predicted score if a student studies for 9.25 hrs/ day? **Importing Libraries** import numpy as np import pandas as pd %matplotlib inline import matplotlib.pyplot as plt **Loading Dataset** url= "http://bit.ly/w-data " data=pd.read_csv(url) print("***** Data imported successfully! *****") data #Displaying the data **** Data imported successfully! **** Out[2]: **Hours Scores** 2.5 21 5.1 47 3.2 27 8.5 75 3.5 30 1.5 20 9.2 88 5.5 60 8.3 81 2.7 25 7.7 10 85 11 5.9 62 12 4.5 41 42 13 3.3 17 14 1.1 15 8.9 95 16 2.5 30 17 1.9 24 18 6.1 67 19 7.4 69 2.7 30 21 4.8 54 22 3.8 35 23 6.9 76 24 7.8 86 df.head() Out[3]: **Hours Scores** 2.5 21 5.1 47 3.2 27 8.5 75 3.5 30 df.describe() Out[4]: Hours Scores **count** 25.000000 25.000000 5.012000 51.480000 std 2.525094 25.286887 min 1.100000 17.000000 2.700000 30.000000 4.800000 47.000000 7.400000 75.000000 9.200000 95.000000 **Checking Null Values** df.isnull().sum() Hours Out[5]: Scores dtype: int64 No Null values found, so no need to clean this data Plotting the distribution of scores plt.scatter(df['Hours'], df['Scores'], color = 'red') plt.title('Hours vs Percentage(%)') plt.xlabel('Hours Studied') plt.ylabel("Percentage Score(%)") plt.show() Hours vs Percentage(%) 90 80 (%) 70 60 50 40 30 Linear Regression model - Preparing the data and splitting it in testing # Dividing the data into attributes and labels. x = df.iloc[:, :-1].valuesy = df.iloc[:, 1].valuesSplitting the data into Training and Testing Sets # Using in-built method of sci-kit learn of train_test_split() from sklearn.model_selection import train_test_split x_{train} , x_{test} , y_{train} , y_{test} = $train_{test}$, y_{test} , y_{test} = 0.25, $random_{test}$ Training the model from sklearn.linear_model import LinearRegression model = LinearRegression() In [10]: model.fit(x_train, y_train) print("Model Trained!") Model Trained! Plotting the Regression Line # Plotting the regression line # formula for line is y=m*x + cline = model.coef_*x + model.intercept_ plt.scatter(x, y, color = 'red') plt.plot(x, line) plt.show() 80 60 **Making Predictions** #Predicting scores for model print(x_test) y_pred = model.predict(x_test) [3.2] [7.4] [2.5] [3.8] [1.9]] Comparing Actual vs Predicted df1 = pd.DataFrame({'Actual' : y_test, 'Predicted' : y_pred}) df1 Out[13]: **Actual Predicted** 20 16.844722 27 33.745575 69 75.500624 30 26.786400 62 60.588106 35 39.710582 24 20.821393 #Checking the accuracy of training and test scores print('Test Score') print(model.score(x_test, y_test)) print('Training Score') print(model.score(x_train, y_train)) Test Score 0.9367661043365055 Training Score 0.9484509249326872 Testing with custom data In [15]: hrs = [[9.25]] predict = model.predict(hrs) print("No. of Hours = {}".format(hrs)) print("Predicted Score = {}".format(predict[0]))

No. of Hours = [[9.25]]

In [1]: from sklearn import metrics

NameError

Predicted Score = 93.89272889341655

Evaluating the Model

NameError: name 'y_test' is not defined

print('Mean Absolute Error(MAE) :', metrics.mean_absolute_error(y_test, y_pred))
print('Mean Squared Error(MSE) :', metrics.mean_squared_error(y_test, y_pred))

print('Root Mean Squared Error(RMSE) :', np.sqrt(metrics.mean_squared_error(y_test, y_pred)))

---> 2 print('Mean Absolute Error(MAE) :', metrics.mean_absolute_error(y_test, y_pred))
3 print('Mean Squared Error(MSE) :', metrics.mean_squared_error(y_test, y_pred))

Traceback (most recent call last)

4 print('Root Mean Squared Error(RMSE) :', np.sqrt(metrics.mean_squared_error(y_test, y_pred)))