AUTHOR NAME: THEJAS ELDHOS KURIAN DATASET: http://bit.ly/w-data Predict the score of student who works 9.25 hours/day Step 1: Importing neccessary libraries/packages In [1]: import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression from sklearn.metrics import mean_squared_error,r2_score,mean_absolute_error Step 2: Importing the dataset In [2]: URL = "http://bit.ly/w-data" data = pd.read_csv(URL) data.head(10) **Hours Scores** Out[2]: 2.5 21 47 5.1 3.2 27 8.5 75 3.5 30 1.5 20 88 5.5 60 8.3 81 2.7 25 Step 3: Choosing the appropriate variables # x denotes independent variable x = data["Hours"] # y denotes ependent variable y = data["Scores"] Step 4: Plotting the graph of given data In [4]: plt.scatter(x,y,marker = 'o') plt.title("Scatter Plot of Data") plt.xlabel("No. of study hours") plt.ylabel("Percentage of marks obtained") plt.show() Scatter Plot of Data marks obtained 09 04 08 ъ 50 40 30 20 No. of study hours Step 5: Splitting the data into testing and training data $x_{train}, x_{test}, y_{train}, y_{test} = train_{test}, y_{train}, z_{test}, y_{train}, z_{test}, z_{te$ Step 6: Building the model with training data In [6]: $model = LinearRegression().fit(np.array(x_train).reshape(-1,1),np.array(y_train).reshape(-1,1))$ print("Model Coefficient") print("Intercept : ",float(model.intercept_)) print("Coefficient : " ,float(model.coef_)) Model Coefficient Intercept : 1.2162641848495284 Coefficient: 9.810924136707005 Step 7 : Plotting the Regression line in the scatter plot line = float(model.intercept_) + x*float(model.coef_) In [9]: plt.scatter(x,y) plt.plot(x,line,color = "red") plt.show() 80 60 40 20 Step 8: Making prediction on the test data In [10]: y_predicted = float(model.intercept_) + x_test*float(model.coef_) y_predicted 25.743575 Out[10]: 16 27.705759 33.592314 76.760380 10 17 19.857020 77.741472 38.497776 22 73.817103 19 Name: Hours, dtype: float64 Another method for predicting values from Model created In [11]: y_predict_1 = model.predict(np.array(x_test).reshape(-1,1)) y_predict = pd.DataFrame(y_predict_1 , columns=["y_predicted"]) y_predict Out[11]: y_predicted 25.743575 **1** 27.705759 33.592314 76.760380 19.857020 77.741472 38.497776 73.817103 Step 9: Comparing the predicted and test data set for evalutaing the model In [12]: dataframe = pd.DataFrame({"Actual":y_test, "Predicted":y_predicted}) dataframe.reset_index().drop(columns = ["index"]) **Actual Predicted** Out[12]: 30 25.743575 30 27.705759 42 33.592314 85 76.760380 24 19.857020 86 77.741472 35 38.497776 69 73.817103 In [13]: $x_axis = np.arange(8)$ plt.bar(x_axis-0.2, dataframe["Actual"], 0.4, label = "Actual") plt.bar(x_axis+0.2, dataframe["Predicted"], 0.4, label = "Predicted")

plt.legend(loc = "upper left")

plt.ylabel("Percentage of Mark")

Step 10 : Prediction using given data

print("Number of Hours =",float(x_new))

print("Mean Absolute Error = ", m_a_s)

print("Mean Squared Error = ", m_s_e)

Mean Absolute Error = 5.489294806376118 Mean Squared Error = 35.345962190770884 Root Mean Squared Error = 5.945247025210213

r2 = r2_score(y_test,y_predicted)

Predicted percentage of score = 91.96731244938931

m_a_s = mean_absolute_error(y_test,y_predicted)

m_s_e = mean_squared_error(y_test,y_predicted)

print("Coefficient of Determination = ",r2)

print("Root Mean Squared Error = ",np.sqrt(m_s_e))

Coefficient of Determination = 0.9391488478760097

Since coefficient of determination is 0.94 approximately. The model which we build best fits the data

I was successfully able to carry out the Prediction using Supervised Machine Learning Model. And evaluate model's performance using various parameters.

Step 11 : Importance statistics relating to Model

y_predict_new = float(model.intercept_) + x_new*float(model.coef_)

print("Predicted percentage of score =",float(y_predict_new))

 $x_{new} = np.array([9.25])$

Number of Hours = 9.25

Conclusion

THANK YOU

plt.xlabel("Indicies")

Actual
Predicted

plt.show()

Percentage of Mark & 8

In [14]:

In [15]:

plt.title("Comparison Between Predicted and Actual")

Comparison Between Predicted and Actual

Task 1: Use Simple Linear Regression for fitting the best model for the given data set