import libraries import pandas as pd import numpy as np import dataset data_df = pd.read_csv("H:\Level 4 Information Systems\Plastikat\Plastikat Data\companiesPlastic-data-ML.csv") data_df.head() status number_of_delegates number_of_offers number_of_users company_rateing years_of_experince plastic_quantity Out[18]: 0 IN_APPROVAL 18 0 90 0 0 0 11 17 55 1 SUSPENDED 113 25 5 5 10 **APPROVED** 1 86 SUSPENDED 275 17 45 7 **APPROVED** 99 Define x and y x = data_df.drop(['status', 'plastic_quantity'], axis=1).values y = data_df['plastic_quantity'].values split the data into teaining and testing data from sklearn.model_selection import train_test_split x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=0) train the model on the training data In [5]: from sklearn.linear_model import LinearRegression ml=LinearRegression() ml.fit(x_train,y_train) LinearRegression() Out[5]: In [6]: from sklearn.tree import DecisionTreeRegressor from sklearn.metrics import mean_squared_error as MSE dt = DecisionTreeRegressor(max_depth=4, min_samples_leaf=0.1, random_state=3) dt.fit(x_train,y_train) DecisionTreeRegressor(max_depth=4, min_samples_leaf=0.1, random_state=3) Out[6] predict the test result y_pred = ml.predict(x_test) ml.predict([[11,17,55,1,6]]) y_pred_2 = dt.predict(x_test) mse_dt = MSE(y_test,y_pred) $rmse_dt = mse_dt^{**}(1/2)$ [110.03333333 2.48888889 258.66666667 110.03333333 157.89795918 69.32653061 7.7 258.66666667 69.32653061 157.89795918 258.66666667 157.89795918 7.7 7.7 69.32653061 110.03333333 3.8125 3.8125 2.48888889 2.48888889 7.7 7.7 97.05714286 7.7 3.8125 69.32653061 2.48888889 2.48888889 2.48888889 2.4888889 110.03333333 110.03333333 97.05714286 7.7 157.89795918 2.48888889 157.89795918 258.66666667 258.66666667 110.03333333 7.7 2.48888889 97.05714286 157.89795918 110.03333333 157.89795918 3.8125 2.48888889 3.8125 2.48888889 3.8125 157.89795918 3.8125 97.05714286 69.32653061 69.32653061 258.66666667 2.48888889 157.89795918 3.8125 3.8125 97.05714286 7.7 97.05714286 7.7 110.03333333 157.89795918 157.89795918 2.48888889 157.89795918 7.7 157.89795918 110.03333333 110.03333333 110.03333333 157.89795918 3.8125 3.8125 157.89795918 110.03333333 2.48888889 69.32653061 2.48888889 69.32653061 157.89795918 69.32653061 110.03333333 97.05714286 157.89795918 157.89795918 69.32653061 157.89795918 69.32653061 7.7 97.05714286 69.32653061 3.8125 3.8125 69.32653061] Evaluate the model In [8]: from sklearn.metrics import r2_score r2_score(y_test,y_pred) 0.9533303352626061 Out[8] In [9]: r2_score(y_test,y_pred_2) 0.8967251807385168 Out[9]: plot the results In [12]: import matplotlib.pyplot as plt plt.figure(figsize=(15,10)) plt.scatter(y_test,y_pred, label="Actual test data") plt.xlabel('Actual') plt.ylabel('Predicted') plt.title('Actual vs Predicted') Text(0.5, 1.0, 'Actual vs Predicted') Actual vs Predicted 250 200 150 100 50 50 200 150 Actual plt.figure(figsize=(15,10)) plt.scatter(y_test,y_pred_2, label="Actual test data") plt.xlabel('Actual') plt.ylabel('Predicted') plt.title('Actual vs Predicted') Text(0.5, 1.0, 'Actual vs Predicted') Actual vs Predicted 250 200 150 100 50 150 Actual predicted values In [14]: pred_y_df = pd.DataFrame({'Actual Value' : y_test, 'predicted value' : y_pred, "Difference" : y_test-y_pred}) pred_y_df[0:20] Out[14]: predicted value Difference 110.574126 -16.574126 0.678293 -0.678293 2 225.253363 -1.253363 224 118.139431 -20.139431 8.604514 4 150 141.395486 85 82.008536 2.991464 6 0 -0.873214 0.873214 222 218.838004 3.161996 8 60 60.160399 -0.160399 -1.842906 1.842906 0 10 -0.485337 0.485337 11 173.403533 -2.403533 12 199 205.303425 -6.303425 13 104 135.008312 -31.008312 0 -0.291399 0.291399 14 59 15 67.853770 -8.853770 16 145 121.157346 23.842654 17 -1.648967 1.648967 18 0 -1.455029 1.455029 1.260108 -1.260108 In [16]: pred_y_df_2 = pd.DataFrame({'Actual Value' : y_test, 'predicted value' : y_pred_2, "Difference" : y_test-y_pred_2}) pred_y_df_2[0:20] Actual Value predicted value Difference Out[16]: 110.033333 -16.033333 0 2.488889 -2.488889 2 224 258.666667 -34.666667 98 110.033333 -12.033333 150 157.897959 -7.897959 5 85 69.326531 15.673469 6 0 7.700000 -7.700000 222 258.666667 -36.666667 8 60 69.326531 -9.326531 0 3.812500 -3.812500 10 0 7.700000 -7.700000 171 157.897959 13.102041 11 12 199 258.666667 -59.666667 13 104 157.897959 -53.897959 14 0 7.700000 -7.700000 59 69.326531 -10.326531 15 16 145 110.033333 34.966667 17 0 3.812500 -3.812500 18 0 3.812500 -3.812500 0 2.488889 -2.488889 19