In [75]: **%matplotlib** inline import matplotlib.pyplot as plt import numpy as np import pandas as pd from sklearn.linear_model import LinearRegression from sklearn.model_selection import train_test_split from sklearn.metrics import mean_squared_error In [105... df=pd.read_csv('diabetes.csv') df.describe() In [106.. **Pregnancies** Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Outcome Out[106]: Age 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 768.000000 count 3.845052 120.894531 69.105469 79.799479 31.992578 0.471876 33.240885 0.348958 mean 20.536458 31.972618 15.952218 115.244002 std 3.369578 19.355807 7.884160 0.331329 11.760232 0.476951 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.078000 21.000000 0.000000 min 27.300000 24.000000 25% 1.000000 99.000000 62.000000 0.000000 0.000000 0.243750 0.000000 50% 3.000000 117.000000 72.000000 23.000000 30.500000 32.000000 0.372500 29.000000 0.000000 **75**% 6.000000 140.250000 80.000000 32.000000 127.250000 36.600000 0.626250 41.000000 1.000000 122.000000 17.000000 199.000000 99.000000 846.000000 67.100000 2.420000 81.000000 1.000000 max df.head() In [107... Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome Out[107]: 0 6 148 72 35 0 33.6 50 0.627 1 1 66 29 0 1 85 0 26.6 0.351 31 2 8 183 64 0 32 0 23.3 0.672 1 3 23 0 1 89 66 94 28.1 0.167 21 4 40 35 0 137 168 43.1 2.288 33 1 df.isnull().any() False Pregnancies Out[108]: Glucose False BloodPressure False SkinThickness False Insulin False BMI False DiabetesPedigreeFunction False False Age Outcome False dtype: bool df.shape (768, 9)Out[80]: df.columns In [81]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'], dtype='object') df["SkinThickness"]=df["SkinThickness"].replace(0,df["SkinThickness"].mean()) In [82]: df Out[82]: Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome 0 6 148 72 35.000000 0 33.6 0.627 50 1 85 66 29.000000 31 0 0 26.6 0.351 2 0.672 32 8 183 64 20.536458 0 23.3 1 3 23.000000 0.167 0 89 66 94 28.1 4 0 137 40 35.000000 168 43.1 2.288 33 1 48.000000 763 10 101 76 180 32.9 0.171 63 0 2 70 27 764 122 27.000000 0 36.8 0.340 0 5 72 765 121 23.000000 112 26.2 0.245 30 0 766 126 60 20.536458 0 30.1 0.349 767 1 93 70 31.000000 0 30.4 0.315 23 0 768 rows × 9 columns df["Insulin"]=df["Insulin"].replace(0,df["Insulin"].mean()) In [83]: df Out[83]: Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome 0 6 79.799479 33.6 50 148 72 35.000000 0.627 1 1 1 85 66 29.000000 79.799479 26.6 0.351 31 0 2 8 183 64 20.536458 79.799479 23.3 0.672 32 1 3 1 89 66 23.000000 94.000000 28.1 0.167 21 0 4 0 137 40 35.000000 168.000000 43.1 33 2.288 1 48.000000 180.000000 32.9 763 10 101 76 0.171 63 0 764 2 122 70 27.000000 79.799479 36.8 0.340 27 5 765 121 72 23.000000 112.000000 26.2 0.245 30 0 766 1 126 60 20.536458 79.799479 30.1 0.349 47 767 1 70 0 93 31.000000 79.799479 30.4 0.315 23 768 rows × 9 columns plt.figure(figsize=(12,7)) plt.scatter("Pregnancies", "Insulin", data=df) plt.title("Pregnancies vs Insulin") plt.xlabel("Pregnancies") plt.ylabel("Insulin") plt.show() Pregnancies vs Insulin 800 600 ulnsulin 400 200 0.0 5.0 7.5 10.0 12.5 15.0 17.5 2.5 Pregnancies plt.figure(figsize=(12,7)) plt.scatter("SkinThickness", "Insulin", data=df) plt.title("SkinThickness vs Insulin") plt.xlabel("SkinThickness") plt.ylabel("Insulin") plt.show() SkinThickness vs Insulin 800 600 Insulin 400 200 0 20 40 60 80 100 SkinThickness In [86]: #To remove the Outliers def remove_outlier (dataFrame): for column_name in dataFrame.columns: Q1=df[column_name].quantile(0.25) Q3=df[column_name].quantile(0.75) IQR=Q3-Q1 lower_limit=Q1-1.5*IQR upper_limit=Q3+1.5*IQR print(f"{column_name}: \n Lower Limit: {lower_limit} \n Upper Limit: {upper_limit}") dataFrame = dataFrame[(dataFrame[column_name] > lower_limit)| (dataFrame[column_name] < upper_limit)]</pre> return dataFrame In [87]: df=remove_outlier(df) Pregnancies: Lower Limit: -6.5 Upper Limit: 13.5 Glucose: Lower Limit: 37.125 Upper Limit: 202.125 BloodPressure: Lower Limit: 35.0 Upper Limit: 107.0 SkinThickness: Lower Limit: 3.341145833333332 Upper Limit: 49.1953125 Insulin: Lower Limit: 8.623697916666671 Upper Limit: 198.42578125 BMI: Lower Limit: 13.35 Upper Limit: 50.550000000000004 DiabetesPedigreeFunction: Lower Limit: -0.329999999999999 Upper Limit: 1.2 Age: Lower Limit: -1.5 Upper Limit: 66.5 Outcome: Lower Limit: -1.5 Upper Limit: 2.5 In [88]: df Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome Out[88]: 0 6 148 72 35.000000 79.799479 33.6 0.627 50 1 1 1 85 66 79.799479 26.6 0.351 31 0 29.000000 2 8 183 64 79.799479 23.3 0.672 32 20.536458 1 94.000000 28.1 3 1 89 66 21 23.000000 0.167 0 35.000000 168.000000 43.1 4 137 40 2.288 33 1 ••• 10 101 76 48.000000 180.000000 32.9 0.171 63 0 763 764 2 122 70 27.000000 79.799479 36.8 0.340 27 765 5 121 72 23.000000 112.000000 26.2 0.245 30 0 79.799479 30.1 766 1 126 0.349 47 20.536458 767 1 93 70 31.000000 79.799479 30.4 0.315 23 0 768 rows × 9 columns In [89]: #Splitting the datas from sklearn.model_selection import train_test_split x=df.drop(["Outcome"], axis=1) y=df["Outcome"] In [90]: x_train, x_test, y_train, y_test=train_test_split(x, y, train_size=0.8, random_state=42) Out[90]: Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age 60 2 84 20.536458 79.799479 0 0.0 0.304 21 618 9 112 82 79.799479 28.2 50 24.000000 1.282 346 1 139 46 19.000000 83.000000 28.7 0.654 22 294 161 20.536458 79.799479 21.9 0.254 65 231 6 134 37.000000 370.000000 46.2 0.238 46 5 71 139 35.000000 140.000000 28.6 64 0.411 26 106 122 20.536458 79.799479 22.4 0.207 27 10 101 86 270 37.000000 79.799479 45.6 1.136 38 79.799479 42.4 435 0 141 29 20.536458 0.205 102 0 125 20.536458 79.799479 22.5 0.262 21 614 rows × 8 columns In [91]: y_train 60 0 Out[91]: 618 1 346 0 294 0 231 1 71 0 106 0 270 1 435 1 102 0 Name: Outcome, Length: 614, dtype: int64 x_train.shape In [92]: (614, 8)Out[92] x_test.shape (154, 8)Out[93]: In [94]: #Logistic Regression from sklearn.linear_model import LogisticRegression logReg=LogisticRegression() In [95]: logReg.fit(x_train,y_train) In [96]: C:\Users\S.Ramya\anaconda3\lib\site-packages\sklearn\linear_model_logistic.py:814: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT. Increase the number of iterations (max_iter) or scale the data as shown in: https://scikit-learn.org/stable/modules/preprocessing.html Please also refer to the documentation for alternative solver options: https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression n_iter_i = _check_optimize_result(LogisticRegression() Out[96]: logReg.score(x_test,y_test) 0.7532467532467533Out[97]: #LinearRegression In [98]: linearReg=LinearRegression() linearReg.fit(x_train,y_train) In [99]: LinearRegression() Out[99]: linearReg.score(x_test,y_test) 0.2586760188416003 Out[100]: In [101... #DecisionTreeRegressor from sklearn.tree import DecisionTreeRegressor dt=DecisionTreeRegressor(max_depth=4, min_samples_split=5, max_leaf_nodes=10) In [102... dt.fit(x_train,y_train) DecisionTreeRegressor(max_depth=4, max_leaf_nodes=10, min_samples_split=5) Out[102]: In [103... dt.score(x_train,y_train) 0.40328186716013226 Out[103]: dt.score(x_test,y_test) In [104... 0.2359826081685742 Out[104]: