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In [1]:
         # This Python 3 environment comes with many helpful analytics libraries inst
         # It is defined by the kaggle/python Docker image: https://github.com/kaggle
         # For example, here's several helpful packages to load
         import numpy as np # linear algebra
         import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
         # Input data files are available in the read-only "../input/" directory
         # For example, running this (by clicking run or pressing Shift+Enter) will l
         import os
         for dirname, _, filenames in os.walk('/kaggle/input'):
             for filename in filenames:
                  print(os.path.join(dirname, filename))
         # You can write up to 20GB to the current directory (/kaggle/working/) that
         # You can also write temporary files to /kaggle/temp/, but they won't be sav
         /kaggle/input/heart-disease-uci/heart.csv
In [2]:
         df = pd.read csv('/kaggle/input/heart-disease-uci/heart.csv')
In [3]:
         df.head()
Out[3]:
                    cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal ta
                sex
         0
                     3
                                 233
                                               0
                                                     150
                                                                               0
                                                                                    1
            63
                  1
                            145
                                                                    2.3
                                                                            0
         1
            37
                     2
                                 250
                                       0
                                                     187
                                                                                    2
                  1
                            130
                                               1
                                                                    3.5
                                                                            0
                                                                               0
         2
            41
                     1
                            130
                                 204
                                               0
                                                     172
                                                                    1.4
                                                                                    2
                  0
                                                             0
                                                                            2
                                                                               0
         3
                                                                                    2
            56
                            120
                                 236
                                               1
                                                     178
                                                             0
                                                                    8.0
                                                                            2
                  1
                     1
                                                                               0
                                 354
                                                                            2
                                                                                    2
            57
                  0
                     0
                            120
                                               1
                                                     163
                                                             1
                                                                    0.6
                                                                               0
In [4]:
         from sklearn.model selection import train test split
         X_train,X_test,y_train,y_test = train_test_split(df.iloc[:,0:-1],df.iloc[:,-
In [5]:
         from sklearn.linear model import LogisticRegression
         from sklearn.tree import DecisionTreeClassifier
In [6]:
         clf1 = LogisticRegression()
         clf2 = DecisionTreeClassifier()
In [7]:
         clf1.fit(X train,y train)
         clf2.fit(X_train,y_train)
         /opt/conda/lib/python3.7/site-packages/sklearn/linear_model/_logistic.py:765:
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ConvergenceWarning: Ibtgs tailed 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-regres
         sion
           extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG)
 Out[7]: DecisionTreeClassifier()
 In [8]:
          y_pred1 = clf1.predict(X_test)
          y pred2 = clf2.predict(X test)
 In [9]:
          from sklearn.metrics import accuracy score,confusion matrix
          print("Accuracy of Logistic Regression",accuracy_score(y_test,y_pred1))
          print("Accuracy of Decision Trees",accuracy_score(y_test,y_pred2))
         Accuracy of Logistic Regression 0.9016393442622951
         Accuracy of Decision Trees 0.8360655737704918
In [10]:
          confusion matrix(y test,y pred1)
Out[10]: array([[26, 6],
                [ 0, 29]])
In [11]:
          print("Logistic Regression Confusion Matrix\n")
          pd.DataFrame(confusion_matrix(y_test,y_pred1),columns=list(range(0,2)))
         Logistic Regression Confusion Matrix
Out[11]:
         0 26
            0 29
In [12]:
          print("Decision Tree Confusion Matrix\n")
          pd.DataFrame(confusion_matrix(y_test,y_pred2),columns=list(range(0,2)))
         Decision Tree Confusion Matrix
Out[12]:
         0 24 8
            2 27
In [13]:
          result = pd.DataFrame()
          result['Actual Label'] = y test
          result['Logistic Regression Prediction'] = y pred1
          result['Decision Tree Prediction'] = y pred2
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In [14]:
          result.sample(10)
Out[14]:
              Actual Label Logistic Regression Prediction Decision Tree Prediction
          257
                       0
                                                  0
                                                                       0
           65
                       1
                                                  1
                                                                       1
          164
                       1
                                                  1
           29
                       1
          296
          184
          161
          292
           53
          147
                       1
                                                  1
In [15]:
          from sklearn.metrics import recall_score,precision_score,f1_score
In [16]:
          print("For Logistic regression Model")
          print("-"*50)
          cdf = pd.DataFrame(confusion matrix(y test,y pred1),columns=list(range(0,2))
          print(cdf)
          print("-"*50)
          print("Precision - ",precision_score(y_test,y_pred1))
          print("Recall - ",recall_score(y_test,y_pred1))
          print("F1 score - ",f1_score(y_test,y_pred1))
          For Logistic regression Model
             0 1
           26 6
             0 29
         Precision - 0.8285714285714286
          Recall - 1.0
          F1 score - 0.90625
In [17]:
          print("For DT Model")
          print("-"*50)
          cdf = pd.DataFrame(confusion matrix(y test,y pred2),columns=list(range(0,2))
          print(cdf)
          print("-"*50)
          print("Precision - ",precision_score(y_test,y_pred2))
          print("Recall - ",recall score(y test,y pred2))
          print("F1 score - ",f1_score(y_test,y_pred2))
          For DT Model
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2 27
         Precision - 0.7714285714285715
         Recall - 0.9310344827586207
         F1 score - 0.8437500000000001
In [18]:
          precision_score(y_test,y_pred1,average=None)
Out[18]: array([1.
                         , 0.82857143])
In [19]:
          precision_score(y_test,y_pred2,average=None)
Out[19]: array([0.92307692, 0.77142857])
In [21]:
          recall_score(y_test,y_pred2,average=None)
Out[21]: array([0.75 , 0.93103448])
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