In [41]: # import pandas Library import pandas as pd from sklearn.model_selection import train_test_split import matplotlib.pyplot as plt import seaborn as sns from sklearn.preprocessing import LabelEncoder from sklearn.metrics import accuracy_score,confusion_matrix from sklearn.linear_model import LogisticRegression

In [2]: # Reading csv file
 data = pd.read_csv("Heart.csv")
 # creating a dataframe of it
 data = pd.DataFrame(data)
 data.head()

Out[2]: Unnamed: ChestPain RestBP Chol Fbs RestECG MaxHR ExAng Oldpeak Slope Ca Age Sex 0 1 63 1 typical 145 233 1 2 150 0 2.3 3 0.0 1 286 2 2 67 asymptomatic 160 0 108 1 1.5 2 3.0 2 2 3 67 1 asymptomatic 120 229 0 129 2.6 2 2.0 4 0 3 37 1 nonanginal 130 250 0 187 0 3.5 3 0.0 2 5 0 130 204 0 41 nontypical 0 172 1.4 1 0.0

In [3]: # Datatype of each column
data.dtypes

•

Out[3]: Unnamed: 0 int64 Age int64 int64 Sex object ChestPain RestBP int64 Chol int64 int64 Fbs RestECG int64 MaxHR int64 ExAng int64 01dpeak float64 Slope int64 Ca float64 Thal object AHD object dtype: object

In [4]: # Information about each column data data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 15 columns):
Column Non-Null Count Dtype

#	Column	Non-Null Count	Dtype
0	Unnamed: 0	303 non-null	int64
1	Age	303 non-null	int64
2	Sex	303 non-null	int64
3	ChestPain	303 non-null	object
4	RestBP	303 non-null	int64
5	Chol	303 non-null	int64
6	Fbs	303 non-null	int64
7	RestECG	303 non-null	int64
8	MaxHR	303 non-null	int64
9	ExAng	303 non-null	int64
10	Oldpeak	303 non-null	float64
11	Slope	303 non-null	int64
12	Ca	299 non-null	float64
13	Thal	301 non-null	object
14	AHD	303 non-null	object
dtyp	es: float64(2), int64(10),	object(3)

memory usage: 35.6+ KB

In [5]: # Count ,min,max ,etc of each column
data.describe()

Out[5]:

	Unnamed: 0	Age	Sex	RestBP	Chol	Fbs	RestECG	MaxHR	
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303
mean	152.000000	54.438944	0.679868	131.689769	246.693069	0.148515	0.990099	149.607261	0
std	87.612784	9.038662	0.467299	17.599748	51.776918	0.356198	0.994971	22.875003	0
min	1.000000	29.000000	0.000000	94.000000	126.000000	0.000000	0.000000	71.000000	0
25%	76.500000	48.000000	0.000000	120.000000	211.000000	0.000000	0.000000	133.500000	0
50%	152.000000	56.000000	1.000000	130.000000	241.000000	0.000000	1.000000	153.000000	0
75%	227.500000	61.000000	1.000000	140.000000	275.000000	0.000000	2.000000	166.000000	1
max	303.000000	77.000000	1.000000	200.000000	564.000000	1.000000	2.000000	202.000000	1

```
In [6]:
         #Drop unnnamed column
         data.drop("Unnamed: 0",axis=1,inplace=True)
         data.head()
Out[6]:
             Age Sex
                         ChestPain RestBP Chol Fbs RestECG MaxHR ExAng Oldpeak Slope Ca
                                                                                                      Thal A
                            typical
                                                            2
          0
              63
                    1
                                       145
                                            233
                                                                  150
                                                                           0
                                                                                   2.3
                                                                                           3 0.0
                                                                                                      fixed
          1
              67
                      asymptomatic
                                       160
                                            286
                                                   0
                                                            2
                                                                  108
                                                                            1
                                                                                   1.5
                                                                                           2 3.0
                                                                                                     normal
                                                            2
                       asymptomatic
                                            229
                                                                                   2.6
          2
              67
                                       120
                                                   0
                                                                  129
                                                                            1
                                                                                           2 2.0 reversable
          3
              37
                    1
                         nonanginal
                                       130
                                            250
                                                   0
                                                            0
                                                                  187
                                                                           0
                                                                                   3.5
                                                                                           3 0.0
                                                                                                     normal
          4
                    0
                         nontypical
                                       130
                                            204
                                                   0
                                                            2
                                                                  172
                                                                           0
                                                                                           1 0.0
              41
                                                                                   1.4
                                                                                                     normal
In [7]: #Finding null values in each column
         data.isna().sum()
Out[7]: Age
                        0
         Sex
                        0
         ChestPain
                        0
         RestBP
                        0
         Chol
                        0
         Fbs
                        0
                        0
         RestECG
         MaxHR
                        0
                        0
         ExAng
         01dpeak
                        0
         Slope
                        0
         Ca
                        4
         Thal
                        2
         AHD
         dtype: int64
In [8]: ca_mean = data['Ca'].mean()
         data['Ca'] = data['Ca'].fillna(ca_mean)
         data.isna().sum()
Out[8]: Age
                        0
                        0
         Sex
         ChestPain
                        0
                        0
         RestBP
         Chol
                        0
         Fbs
                        0
         RestECG
                        0
         MaxHR
                        0
                        0
         ExAng
                        0
         01dpeak
         Slope
                        0
                        0
         Ca
         Thal
                        2
         AHD
         dtype: int64
```

```
data['Thal'] = data['Thal'].fillna(ca_mode)
          data.isna().sum()
 Out[9]: Age
                        0
          Sex
                        0
          ChestPain
                        0
          RestBP
                        0
          Chol
                        0
          Fbs
                        0
                        0
          RestECG
          MaxHR
                        0
          ExAng
                        0
          01dpeak
          Slope
                        0
          Ca
                        0
          Thal
                        0
          AHD
          dtype: int64
In [10]: #Finding number of zeros in each column
          for i in data:
            count = (data[i] == 0).sum()
            print('Zeros in column',i,'->', count)
          Zeros in column Age -> 0
          Zeros in column Sex -> 97
          Zeros in column ChestPain -> 0
          Zeros in column RestBP -> 0
          Zeros in column Chol -> 0
          Zeros in column Fbs -> 258
          Zeros in column RestECG -> 151
          Zeros in column MaxHR -> 0
          Zeros in column ExAng -> 204
          Zeros in column Oldpeak -> 99
          Zeros in column Slope -> 0
          Zeros in column Ca -> 176
          Zeros in column Thal -> 0
          Zeros in column AHD -> 0
In [11]: #Label Encoding the 'AHD' column
          le1=LabelEncoder()
          data['AHD']=le1.fit_transform(data['AHD'])
          data.head()
Out[11]:
                         ChestPain RestBP Chol Fbs RestECG MaxHR ExAng Oldpeak Slope Ca
             Age Sex
                                                                                                     Thal I
           0
               63
                    1
                             typical
                                       145
                                            233
                                                   1
                                                            2
                                                                 150
                                                                          0
                                                                                  2.3
                                                                                         3
                                                                                            0.0
                                                                                                    fixed
           1
               67
                       asymptomatic
                                       160
                                            286
                                                   0
                                                            2
                                                                 108
                                                                          1
                                                                                  1.5
                                                                                         2 3.0
                                                                                                   normal
           2
               67
                       asymptomatic
                                       120
                                            229
                                                   0
                                                            2
                                                                 129
                                                                           1
                                                                                  2.6
                                                                                         2 2.0
                                                                                               reversable
           3
               37
                    1
                         nonanginal
                                       130
                                            250
                                                            0
                                                                 187
                                                                                  3.5
                                                                                         3 0.0
                                                                                                   normal
                    0
                                            204
                                                            2
                                                                 172
                                                                          0
                                                                                  1.4
               41
                          nontypical
                                       130
                                                   0
                                                                                         1 0.0
                                                                                                   normal
```

In [9]: | ca_mode = data['Thal'].mode()[0]

```
In [12]:
          le1=LabelEncoder()
          data['Thal']=le1.fit_transform(data['Thal'])
          data.head()
Out[12]:
                         ChestPain RestBP Chol Fbs RestECG MaxHR ExAng
             Age Sex
                                                                            Oldpeak Slope Ca Thal AHD
           0
              63
                    1
                                            233
                                                                 150
                                                                          0
                                                                                 2.3
                                                                                         3 0.0
                                                                                                  0
                             typical
                                      145
              67
                                            286
                                                           2
                                                                 108
                                                                                 1.5
           1
                                                  0
                                                                                        2 3.0
                                      160
                                                                          1
                                                                                                  1
                                                                                                       1
                       asymptomatic
                                                           2
                                                                                        2 2.0
                                                                                                  2
           2
              67
                                      120
                                            229
                                                  0
                                                                 129
                                                                          1
                                                                                 2.6
                                                                                                       1
                       asymptomatic
           3
              37
                    1
                                      130
                                            250
                                                  0
                                                           0
                                                                 187
                                                                          0
                                                                                 3.5
                                                                                         3 0.0
                                                                                                  1
                                                                                                       0
                         nonanginal
           4
              41
                    0
                                      130
                                            204
                                                  0
                                                           2
                                                                 172
                                                                          0
                                                                                 1.4
                                                                                         1 0.0
                                                                                                  1
                                                                                                       0
                          nontypical
In [13]:
          le1=LabelEncoder()
          data['ChestPain']=le1.fit_transform(data['ChestPain'])
          data.head()
Out[13]:
             Age Sex ChestPain RestBP Chol Fbs RestECG MaxHR ExAng Oldpeak Slope Ca Thal AHD
                                                         2
           0
              63
                    1
                              3
                                    145
                                         233
                                                               150
                                                                        0
                                                                              2.3
                                                                                      3 0.0
                                                                                               0
                                                                                                     0
                                                1
           1
              67
                    1
                              0
                                    160
                                         286
                                                         2
                                                               108
                                                                              1.5
                                                                                         3.0
                                                                                               1
                                                                                                     1
           2
              67
                              0
                                    120
                                         229
                                                         2
                                                               129
                                                                              2.6
                                                                                      2 2.0
                                                                                               2
                    1
                                                0
                                                                        1
                                                                                                     1
              37
                              1
                                    130
                                         250
                                                         0
           3
                    1
                                                0
                                                               187
                                                                              3.5
                                                                                      3 0.0
                                                                                               1
                                                                                                     0
              41
                              2
                                    130
                                         204
                                                         2
                                                               172
                                                                                               1
                                                                                                     0
                                                                              1.4
                                                                                      1 0.0
In [14]: | # splitting data using train test split
          x = data.drop(['AHD'], axis=1)
          y = data.AHD
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
          x_train.shape,x_test.shape,y_train.shape,y_test.shape
Out[14]: ((242, 13), (61, 13), (242,), (61,))
In [47]: #Classification model using Decision Tree
          from sklearn.tree import DecisionTreeClassifier
          tc=DecisionTreeClassifier(criterion='entropy')
          tc.fit(x_train,y_train)
          y_train_pred=tc.predict(x_train)
          print("Training Accuracy Score :",accuracy_score(y_train_pred,y_train))
          print("Training Confusion Matrix :",confusion_matrix(y_train_pred,y_train))
          Accuracy Score : 1.0
          Confusion Matrix : [[129]
                                        0]
           [ 0 113]]
In [48]: y_test_pred=tc.predict(x_test)
          print("Accuracy Score :",accuracy_score(y_test_pred,y_test))
          print("Confusion Matrix :",confusion_matrix(y_test_pred,y_test))
          Accuracy Score: 0.6721311475409836
          Confusion Matrix : [[25 10]
```

[10 16]]

```
In [50]: # Classification model using Logistic Regression
         lg = LogisticRegression(random_state=0)
         lg.fit(x_train, y_train)
         y_train_pred = lg.predict(x_train)
         print("Training Accuracy Score :",accuracy_score(y_train_pred,y_train))
         print("Training Confusion Matrix :",confusion_matrix(y_train_pred,y_train))
         Training Accuracy Score : 0.871900826446281
         Training Confusion Matrix : [[119 21]
          [ 10 92]]
         /home/student/.local/lib/python3.8/site-packages/sklearn/linear_model/_logistic.py:44
         4: 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 (https://scikit-learn.o
         rg/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 (htt
         ps://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)
           n_iter_i = _check_optimize_result(
In [51]: y_test_pred = lg.predict(x_test)
         print("Testing Accuracy Score :",accuracy_score(y_test_pred,y_test))
         print("Testing Confusion Matrix :",confusion_matrix(y_test_pred,y_test))
```

Testing Accuracy Score: 0.7868852459016393
Testing Confusion Matrix: [[31 9]
[4 17]]