# "EFFECTUAL PROPHECY OF HEART DISEASE PROGNOSIS USING MACHINE LEARNING TECHNIQUES"

## **IMPORTING LIBRARIES**

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt#FOR PLOTING GRAPHS
from sklearn.model_selection import train_test_split # Import_rain_test_split function
from sklearn import svm #Import svm model
from sklearn import metrics #Import scikit-learn metrics module for accuracy calculation
from sklearn.metrics import confusion_matrix
%matplotlib inline
import os
print(os.listdir())
import warnings
warnings.filterwarnings('ignore')
from sklearn.metrics import accuracy_score

□: ['.config', 'ptest.csv', 'heart2.csv', 'heart1.csv', 'sample_data']
```

#### **READ THE CSV HEART DATA BASE**

```
data = pd.read_csv("/content/heart1.csv")
print(data)
```

|      | age | sex | ср | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | \ |
|------|-----|-----|----|----------|------|-----|---------|---------|-------|---------|---|
| 0    | 52  | 1   | 0  | 125      | 212  | 0   | 1       | 168     | 0     | 1.0     |   |
| 1    | 53  | 1   | 0  | 140      | 203  | 1   | 0       | 155     | 1     | 3.1     |   |
| 2    | 70  | 1   | 0  | 145      | 174  | 0   | 1       | 125     | 1     | 2.6     |   |
| 3    | 61  | 1   | 0  | 148      | 203  | 0   | 1       | 161     | 0     | 0.0     |   |
| 4    | 62  | 0   | 0  | 138      | 294  | 1   | 1       | 106     | 0     | 1.9     |   |
|      |     |     |    | • • •    |      |     | • • •   | • • •   |       |         |   |
| 1020 | 59  | 1   | 1  | 140      | 221  | 0   | 1       | 164     | 1     | 0.0     |   |
| 1021 | 60  | 1   | 0  | 125      | 258  | 0   | 0       | 141     | 1     | 2.8     |   |
| 1022 | 47  | 1   | 0  | 110      | 275  | 0   | 0       | 118     | 1     | 1.0     |   |
| 1023 | 50  | 0   | 0  | 110      | 254  | 0   | 0       | 159     | 0     | 0.0     |   |
| 1024 | 54  | 1   | 0  | 120      | 188  | 0   | 1       | 113     | 0     | 1.4     |   |
|      |     |     |    |          |      |     |         |         |       |         |   |

|      | slope | ca | thal | target |
|------|-------|----|------|--------|
| 0    | 2     | 2  | 3    | 0      |
| 1    | 0     | 0  | 3    | 0      |
| 2    | 0     | 0  | 3    | 0      |
| 3    | 2     | 1  | 3    | 0      |
| 4    | 1     | 3  | 2    | 0      |
|      |       |    |      |        |
| 1020 | 2     | 0  | 2    | 1      |
| 1021 | 1     | 1  | 3    | 0      |
| 1022 | 1     | 1  | 2    | 0      |

 1023
 2
 0
 2
 1

 1024
 1
 1
 3
 0

[1025 rows x 14 columns]

# CHEACKING POSSIVE AND NEGATIVE DATA POINTS IN DATABASE

y = data["target"]

sns.countplot(y)

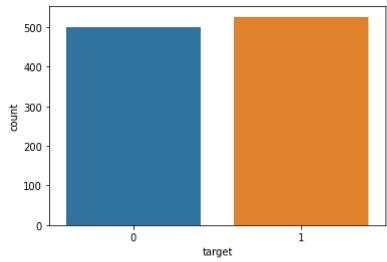
Loading...

target\_temp = data.target.value\_counts()

print(target\_temp)

526
 499

Name: target, dtype: int64



# **PRINTING THE FIRST 5 DATA POINTS**

data.head()

|   | age | sex | ср | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | slope | ca | tł |
|---|-----|-----|----|----------|------|-----|---------|---------|-------|---------|-------|----|----|
| 0 | 52  | 1   | 0  | 125      | 212  | 0   | 1       | 168     | 0     | 1.0     | 2     | 2  |    |
| 1 | 53  | 1   | 0  | 140      | 203  | 1   | 0       | 155     | 1     | 3.1     | 0     | 0  |    |
| 2 | 70  | 1   | 0  | 145      | 174  | 0   | 1       | 125     | 1     | 2.6     | 0     | 0  |    |
| 3 | 61  | 1   | 0  | 148      | 203  | 0   | 1       | 161     | 0     | 0.0     | 2     | 1  |    |
| 4 | 62  | 0   | 0  | 138      | 294  | 1   | 1       | 106     | 0     | 1.9     | 1     | 3  |    |
| 4 |     |     |    |          |      |     |         |         |       |         |       |    | •  |

# **DESCRIBE THE DATABASE**

data.describe()

|       | age         | sex         | ср          | trestbps                | chol       | fbs         |     |
|-------|-------------|-------------|-------------|-------------------------|------------|-------------|-----|
| count | 1025.000000 | 1025.000000 | 1025.000000 | 1025.000000             | 1025.00000 | 1025.000000 | 10: |
| mean  | 54.434146   | 0.695610    | 0.942439    | 131.611707              | 246.00000  | 0.149268    |     |
| std   | 9.072290    | 0.460373    | 1.029641    | 17.516718               | 51.59251   | 0.356527    |     |
| min   | 29.000000   | 0.000000    | 0.000000    | 94.000000               | 126.00000  | 0.000000    |     |
| 25%   | 48.000000   | 0.000000    | 0.000000    | 120.000000              | 211.00000  | 0.000000    |     |
| 50%   | 56.000000   | 1.000000    | 1.000000    | 130.00 <b>.000000</b> 0 | g240.00000 | 0.000000    |     |
| 75%   | 61.000000   | 1.000000    | 2.000000    | 140.000000              | 275.00000  | 0.000000    |     |
| max   | 77.000000   | 1.000000    | 3.000000    | 200.000000              | 564.00000  | 1.000000    |     |



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# **READIONG THE INFORMATION**

data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1025 entries, 0 to 1024
Data columns (total 14 columns):

| #  | Column   | Non-Null Count | Dtype   |
|----|----------|----------------|---------|
|    |          |                |         |
| 0  | age      | 1025 non-null  | int64   |
| 1  | sex      | 1025 non-null  | int64   |
| 2  | ср       | 1025 non-null  | int64   |
| 3  | trestbps | 1025 non-null  | int64   |
| 4  | chol     | 1025 non-null  | int64   |
| 5  | fbs      | 1025 non-null  | int64   |
| 6  | restecg  | 1025 non-null  | int64   |
| 7  | thalach  | 1025 non-null  | int64   |
| 8  | exang    | 1025 non-null  | int64   |
| 9  | oldpeak  | 1025 non-null  | float64 |
| 10 | slope    | 1025 non-null  | int64   |
| 11 | ca       | 1025 non-null  | int64   |
| 12 | thal     | 1025 non-null  | int64   |
| 13 | target   | 1025 non-null  | int64   |
|    |          |                |         |

dtypes: float64(1), int64(13)

memory usage: 112.2 KB

# FINDING THE SIZE OF DATA

data.shape

(1025, 14)

## SPLITTING OUTPUT AND INPUT

```
x = data.drop('target',axis = 1)
y = data.target
print(x)
print("\n\n\n")
print(y)
             age
                   sex
                         ср
                              trestbps
                                          chol
                                                 fbs
                                                       restecg
                                                                  thalach exang
                                                                                     oldpeak
      0
              52
                     1
                          0
                                    125
                                           212
                                                   0
                                                               1
                                                                       168
                                                                                  0
                                                                                          1.0
      1
              53
                          0
                                    140
                                                   1
                                                               0
                                                                       155
                                                                                  1
                                                                                          3.1
                     1
                                           203
                                                                    Loading...
      2
              70
                          0
                                    145
                                           174
                                                              1
                                                                                  1
                     1
                                                   0
                                                                                          2.6
      3
                          0
                                                               1
                                                                       161
                                                                                  0
              61
                     1
                                    148
                                           203
                                                   0
                                                                                          0.0
      4
              62
                     0
                          0
                                    138
                                           294
                                                    1
                                                              1
                                                                       106
                                                                                  0
                                                                                          1.9
                                    . . .
                                                                       . . .
                                                                                           . . .
                                           . . .
              59
                                    140
                                                                                  1
                                                                                          0.0
      1020
                     1
                          1
                                           221
                                                   0
                                                              1
                                                                       164
      1021
              60
                     1
                          0
                                    125
                                           258
                                                   0
                                                              0
                                                                       141
                                                                                  1
                                                                                          2.8
      1022
              47
                     1
                          0
                                    110
                                           275
                                                   0
                                                              0
                                                                       118
                                                                                  1
                                                                                          1.0
                          0
                                                               0
                                                                                  0
      1023
              50
                     0
                                    110
                                           254
                                                   0
                                                                       159
                                                                                          0.0
                     1
                          0
                                    120
                                                    0
                                                               1
                                                                                  0
                                                                                          1.4
      1024
              54
                                           188
                                                                       113
                          thal
             slope
                     ca
      0
                  2
                      2
                              3
      1
                  0
                      0
                              3
      2
                  0
                              3
                      0
      3
                  2
                       1
                              3
                              2
      4
                  1
                       3
      . . .
                            . . .
                . . .
                      . .
      1020
                  2
                      0
                              2
      1021
                  1
                      1
                              3
                              2
      1022
                  1
                       1
      1023
                  2
                       0
                              2
      1024
                  1
                       1
                              3
```

[1025 rows x 13 columns]

```
0
         0
1
         0
2
         0
3
         0
4
         0
1020
         1
1021
         0
         0
1022
1023
         1
1024
Name: target, Length: 1025, dtype: int64
```

# 70% training and 30% test

```
print(x_train)
print('\n')
print(y_train)
print('\n')
print(x_test)
print('\n')
print(y_test)
      [/1/ rows x 13 columns]
      301
               1
      891
               1
      739
               0
                                                                   Loading...
      19
               1
      841
               1
      801
               1
      587
               0
      399
               1
      1008
               1
      141
      Name: target, Length: 717, dtype: int64
                            trestbps
                                        chol
                                               fbs
                                                     restecg
                                                                thalach
                                                                           exang
                                                                                   oldpeak
            age
                  sex
                        ср
      297
             58
                         0
                                  150
                                         270
                                                  0
                                                             0
                                                                     111
                                                                                1
                                                                                        0.8
                    1
      215
             49
                    1
                         1
                                  130
                                          266
                                                  0
                                                             1
                                                                     171
                                                                                0
                                                                                        0.6
                                                             0
      224
             51
                         0
                                  140
                                                  0
                                                                     186
                                                                                        0.0
                    1
                                         261
                                                                                1
      912
             35
                    1
                         0
                                  120
                                         198
                                                  0
                                                             1
                                                                     130
                                                                                1
                                                                                        1.6
      361
             62
                         2
                                  130
                                         231
                                                             1
                                                                     146
                                                                                        1.8
                        . .
                                  . . .
                                          . . .
                                                           . . .
                                                                     . . .
                                                                                        . . .
      . .
      513
             54
                         0
                                  110
                                         206
                                                  0
                                                             0
                                                                     108
                                                                                1
                                                                                        0.0
                    1
      494
                         2
                                  125
                                         245
                                                             0
                                                                     166
                                                                                        2.4
             51
                                                  1
                                                                                0
                    1
      168
             43
                    1
                         2
                                  130
                                         315
                                                             1
                                                                     162
                                                                                0
                                                                                        1.9
             58
                    0
                         1
                                  136
                                          319
                                                  1
                                                             0
                                                                                        0.0
      911
                                                                     152
                                                                                0
      279
             41
                    0
                         1
                                  105
                                         198
                                                  0
                                                             1
                                                                     168
                                                                                0
                                                                                        0.0
            slope
                         thal
                    ca
      297
                2
                     0
                            3
                            2
      215
                2
                     0
      224
                2
                     0
                            2
      912
                1
                     0
                            3
      361
                1
                     3
                            3
              . . .
                          . . .
      513
                1
                     1
                            2
                            2
      494
                1
                2
                     1
                            2
      168
                2
                            2
      911
                     2
                2
                            2
      279
                     1
      [308 rows x 13 columns]
      297
              0
      215
              1
      224
              1
      912
              0
      361
              1
```

## SUPPORT VECTOR MACHINE

#### **LINEAR SVM**

```
Loading...
#Create a svm Classifier
ml = svm.SVC(kernel='linear') # Linear Kernel
#Train the model using the training sets
ml.fit(x train, y train)
#Predict the response for test dataset
y_pred = ml.predict(x_test)
y_pred
     array([0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1,
            1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1,
            0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1,
            0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1,
            0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0,
            1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1,
            1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1,
            1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0,
            0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1,
            0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1,
            0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1,
            0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1,
            1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1,
            1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1])
```

## **ACCURACY OF LINEAR SVM**

#### **POLYNOMIAL SVM**

```
#Create a svm Classifier
ml = svm.SVC(kernel='poly',C=1000,degree=3) # polynomial
#Train the model using the training sets
ml.fit(x_train, y_train)

#Predict the response for test dataset
y_pred = ml.predict(x_test)
```

## **ACCURACY OF POLYNOMIAL SVM**

## **RADIAL BASIS FUNCTION SVM**

```
#Create a svm Classifier
ml = svm.SVC(kernel='rbf',gamma=10,C=1000) # RBF
#Train the model using the training sets
ml.fit(x_train, y_train)
#Predict the response for test dataset
y_pred = ml.predict(x_test)
```

## **ACCURACY OF RADIAL BASIS FUNCTION SVM**

## **DECISION TREE**

```
from sklearn.tree import DecisionTreeClassifier
max_accuracy = 0
```

```
for x in range(200):
    dt = DecisionTreeClassifier(random_state=x)
    dt.fit(x_train,y_train)
    Y_pred_dt = dt.predict(x_test)
    current_accuracy = round(accuracy_score(Y_pred_dt,y_test)*100,2)
    if(current_accuracy>max_accuracy):
        max_accuracy = current_accuracy
        best_x = x

#print(max_accuracy)
#print(best_x)

Loading...

dt = DecisionTreeClassifier(random_state=best_x)
dt.fit(x_train,y_train)
Y_pred_dt = dt.predict(x_test)
```

#### **ACCURACY OF DECISION TREE**

```
score_dt = round(accuracy_score(Y_pred_dt,y_test)*100,2)
print("The accuracy score achieved using Decision Tree is: "+str(score_dt)+" %")
The accuracy score achieved using Decision Tree is: 99.03 %
```

## **RANDOM FOREST**

```
from sklearn.ensemble import RandomForestClassifier
max_accuracy = 0
for x in range(200):
    rf= RandomForestClassifier(random state=x)
    rf.fit(x_train,y_train)
   Y pred rf = rf.predict(x test)
    current accuracy = round(accuracy score(Y pred rf,y test)*100,2)
    if(current_accuracy>max_accuracy):
        max_accuracy = current_accuracy
        best_x = x
print(max_accuracy)
print(best_x)
rf = RandomForestClassifier(random state=best x)
rf.fit(x_train,y_train)
Y pred rf = rf.predict(x test)
     99.03
```

0

## **ACCURACY OF RANDOM FOREST**

```
score_rf = round(accuracy_score(Y_pred_rf,y_test)*100,2)
print("The accuracy score achieved using Decision Tree is: "+str(score_rf)+" %")
The accuracy score achieved using Decision Tree is: 99.03 %
```

#### **K NEAREST NEIGHBOR**

```
Loading...
from sklearn.neighbors import KNeighborsClassifier
max_accuracy=0
bk=0
for k in range(300):
    if k==0:
      continue
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(x_train,y_train)
    Y_pred_dt = knn.predict(x_test)
    current_accuracy = round(accuracy_score(Y_pred_dt,y_test)*100,2)
    if(current_accuracy>max_accuracy):
        max_accuracy = current_accuracy
        bk= k
knn = KNeighborsClassifier(n_neighbors=bk)
knn.fit(x_train,y_train)
```

#### **ACCURACY OF K NEAREST NEIGHBOR**

Y\_pred\_knn=knn.predict(x\_test)

```
score_knn = round(accuracy_score(Y_pred_knn,y_test)*100,2)
print("The accuracy score achieved using KNN is: "+str(score_knn)+" %")
The accuracy score achieved using KNN is: 97.08 %
```

## FINDING TEST ACCURASY FOR NEW UNSEEN 303 DATA IN A DIFFRENT DATASET

```
testdata=pd.read_csv("/content/heart2.csv")
x_newdata=testdata.drop('target',axis=1)
y_accual=testdata.target

y_pred=ml.predict(x_newdata)
y accual
```

## **SVM TEST ACCURACY WITH NEW DATABASE 97.7%**

Loading...

#### THE TEST ACCURACY OF RF FOR NEW UNSEEN DATA IS 99.6%

```
0
       1
1
       1
2
       1
3
       1
4
       1
298
       0
299
       0
300
       0
301
302
Name: target, Length: 303, dtype: int64
array([[138,
              0],
       [ 1, 164]])
```

## SO, THE TEST ACCURACY FOR DECISION TREE FOR NEW DATA SET IS 99.66%

```
y_predknn=knn.predict(x_newdata)
confusion_matrix(y_accual,y_predknn)
```

```
array([[137, 1], [ 2, 163]])
```

#### THE TEST ACCURACY OF KNN FOR UNSEEN DATASET IS 99%

#### **TEST SINGLE PERSON DATA**

```
print("enter the patient data (13 attributes )")
pdata=pd.read_csv("/content/ptest.csv")
y_prednt=dt.predict(pdata)
                                                         Loading...
print(y prednt)
print('\n')
if y_prednt[0]==0:
  print("The patient has no heart disease")
  print("The patient have heart disease")
     enter the patient data (13 attributes )
     [0]
     The patient has no heart disease
print("Enter the patient data (13 attributes )")
dist={}
dist["age"]=int(input("enter your age: "))
dist["sex"]=int(input("enter your sex (0-female 1-for male: "))
dist["cp"]=int(input("enter cp value: "))
dist["trestbps"]=int(input("enter trestbps: "))
dist["chol"]=int(input("enter chol: "))
dist["fbs"]=int(input("enter fbs value: "))
dist["restecg"]=int(input("enter restecg: "))
dist["thalach"]=int(input("enter thalach: "))
dist["exang"]=int(input("enter exang: "))
dist["old peak"]=float(input("enter oldpeak: "))
dist["slop"]=int(input("enter slop: "))
dist["ca"]=int(input("enter ca: "))
dist["thal"]=int(input("enter thal: "))
df=pd.DataFrame(dist,index=[0])
print(df)
print('\n')
y_prednt=dt.predict(df)
print(y_prednt)
print('\n')
if y_prednt[0]==0:
  print("The patient has no heart disease")
else:
  print("The patient have heart disease")
```

```
Enter the patient data (13 attributes )
enter your age: 52
enter your sex (0-female 1-for male: 1
enter cp value: 0
enter trestbps: 125
enter chol: 212
enter fbs value: 0
enter restecg: 1
enter thalach: 168
enter exang: 0
enter oldpeak: 1
enter slop: 2
enter ca: 2
enter thal: 3
  age sex cp trestbps chol fbs restecg thalacadimexang old peak slop \
  52
             0
                     125
                           212
                                  0
                                           1
                                                  168
                                                           0
                                                                   1.0
         1
                                                                           2
  ca thal
  2
```

[0]

The patient has no heart disease