In [2]:

```
# Step-1: Importing the libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

from sklearn import svm
from sklearn.svm import SVC
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix, accuracy_score


# Step-2: Load Data set
dataset= pd.read_csv("E:\\mylab\\dataset\\processed.cleveland.data.csv", names=['age','sex'
dataset_mean= dataset

#Step-3: Data Preprocessing

# Filling missing values Statistics measures
print("*****Before Fill Missing values Row 166,192,287,302********")
print(dataset_mean.loc[287])
dataset1=dataset_mean
df1=pd.DataFrame(dataset1)
#print(df1)

print("-------- Mean of Column 11 'ca' --------")
print(df1['ca'].mean())
df1.fillna(df1.mean(), inplace=True)
print("*****After Fill Missing values Row 166,192,287,302********")
print(df1.loc[[166,192,287,302]])

print("-------- Mean of Column 12 'thal' --------")
print(df1['thal'].mean())
df1.fillna(df1.mean(), inplace=True)
print("*****After Fill Missing values Row 87,266********")
print(df1.loc[[87,266]])


# Extract feature columns
feature_cols = list(dataset.columns[0:13])


# Show the list of columns
print("Feature columns:\n{}".format(feature_cols))

# Separate the data into feature data and target data (X_all and y_all, respectively)
X= dataset[feature_cols]
y= dataset['output'].values

# Show the feature information by printing the first five rows
print("\nFeature values:")
X.head()

# Step-3: Split the Dataset into Traning and Testing data

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state=5)
print(X_train)

# Normalization Step
```

```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()

scaler.fit(X_train)
X_train = scaler.transform(X_train)
print("----After Z-score Normalization on X_train-------")
print(X_train)

scaler.fit(X_test)
X_test = scaler.transform(X_test)
print("----After Z-score Normalization on X_test-------")
print(X_test)
```

```
*****Before Fill Missing values Row 166,192,287,302********
age          58.0
sex           1.0
cp            2.0
trestbps    125.0
chol        220.0
fbs           0.0
restecg       0.0
thalach     144.0
exang         0.0
oldpeak       0.4
slope         2.0
ca            NaN
thal          7.0
output        0.0
Name: 287, dtype: float64
-------- Mean of Column 11 'ca' --------
0.6722408026755853
*****After Fill Missing values Row 166,192,287,302********
     age  sex  cp  trestbps  chol  fbs  restecg  thalach  exang  oldpeak  \
166   52    1   3       138   223    0        0      169      0      0.0   
192   43    1   4       132   247    1        2      143      1      0.1   
287   58    1   2       125   220    0        0      144      0      0.4   
302   38    1   3       138   175    0        0      173      0      0.0   

     slope        ca  thal  output  
166      1  0.672241   3.0       0  
192      2  0.672241   7.0       1  
287      2  0.672241   7.0       0  
302      1  0.672241   3.0       0  
-------- Mean of Column 12 'thal' --------
4.73421926910299
*****After Fill Missing values Row 87,266********
     age  sex  cp  trestbps  chol  fbs  restecg  thalach  exang  oldpeak  \
87    53    0   3       128   216    0        2      115      0      0.0   
266   52    1   4       128   204    1        0      156      1      1.0   

     slope   ca      thal  output  
87       1  0.0  4.734219       0  
266      2  0.0  4.734219       2  
Feature columns:
['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thalach', 'exan
g', 'oldpeak', 'slope', 'ca', 'thal']

Feature values:
     age  sex  cp  trestbps  chol  fbs  restecg  thalach  exang  oldpeak  \
3     37    1   3       130   250    0        0      187      0      3.5
55    54    1   4       124   266    0        2      109      1      2.2
225   34    0   2       118   210    0        0      192      0      0.7
224   63    0   4       108   269    0        0      169      1      1.8
75    65    0   3       160   360    0        2      151      0      0.8
..   ...  ...  ..       ...   ...  ...      ...      ...    ...      ...
8     63    1   4       130   254    0        2      147      0      1.4
73    65    1   4       110   248    0        2      158      0      0.6
118   63    1   4       130   330    1        2      132      1      1.8
189   69    1   3       140   254    0        2      146      0      2.0
206   58    1   4       128   259    0        2      130      1      3.0

     slope   ca  thal
3        3  0.0   3.0
55       2  1.0   7.0
225      1  0.0   3.0
224      2  2.0   3.0
75       1  0.0   3.0
..     ...  ...   ...
8        2  1.0   7.0
73       1  2.0   6.0
118      1  3.0   7.0
189      2  3.0   7.0
206      2  2.0   7.0

[212 rows x 13 columns]
----After Z-score Normalization on X_train-------
[[-1.91736161  0.67975655 -0.16656264 ...  2.36151212 -0.68283167
  -0.93461042]
 [-0.06178394  0.67975655  0.8720044  ...  0.68151021  0.3635441
   1.13614677]
 [-2.24481649 -1.47111492 -1.20512967 ... -0.9984917  -0.68283167
  -0.93461042]
 ...
 [ 0.92058071  0.67975655  0.8720044  ... -0.9984917   2.45629564
   1.13614677]
 [ 1.57549048  0.67975655 -0.16656264 ...  0.68151021  2.45629564
   1.13614677]
 [ 0.37482257  0.67975655  0.8720044  ...  0.68151021  1.40991987
   1.13614677]]
----After Z-score Normalization on X_test-------
[[-1.85828815  0.70128687 -0.16222142 ... -0.9335927  -0.05302469
  -0.81856114]
 [ 0.78936134  0.70128687 -2.2710999  ...  0.58349544  1.48316063
  -0.81856114]
 [-1.62805776  0.70128687  0.89221782 ... -0.9335927  -0.83079106
   1.26892886]
 ...
 [ 1.48005251  0.70128687  0.89221782 ...  0.58349544  1.48316063
   1.26892886]
 [ 0.78936134  0.70128687  0.89221782 ...  0.58349544  0.32618478
  -0.81856114]
 [ 0.44401575  0.70128687 -0.16222142 ... -0.9335927   1.48316063
   1.26892886]]
```

In [5]:

```
#Build SVM Classifier Model
print("Linear SVM")
svm_model_linear = SVC(kernel = 'linear')
#lin_clf = svm.LinearSVC()
#lin_clf.fit(X_train, y_train)
svm_model_linear.fit(X_train, y_train)

y_predictions = svm_model_linear.predict(X_test)
cm1 = confusion_matrix(y_test,y_predictions)

print("Accuracy=",accuracy_score(y_test, y_predictions))
```

```
Linear SVM
Accuracy= 0.6593406593406593
```

In [4]:

```
# training and prediction through a Naive Bayes classifier
from sklearn.naive_bayes import GaussianNB
gnb = GaussianNB().fit(X_train, y_train)
y_predictions = gnb.predict(X_test)
cm1 = confusion_matrix(y_test,y_predictions)

print("Accuracy=",accuracy_score(y_test, y_predictions))
```

```
Accuracy= 0.6043956043956044
```

In [ ]:

```
```