Importing the Dependencies

import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn import metrics
from sklearn.metrics import accuracy_score

import matplotlib.pyplot as plt

import seaborn as sns

from google.colab import files
uploaded = files.upload()

Choose Files heart.csv

• **heart.csv**(application/vnd.ms-excel) - 11328 bytes, last modified: 2/21/2022 - 100% done Saving heart.csv to heart.csv

df=pd.read_csv('heart.csv')

#To print first five rows of the data

df.head(5)

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	th
0	63	1	3	145	233	1	0	150	0	2.3	0	0	
1	37	1	2	130	250	0	1	187	0	3.5	0	0	
2	41	0	1	130	204	0	0	172	0	1.4	2	0	
3	56	1	1	120	236	0	1	178	0	0.8	2	0	
4	57	0	0	120	354	0	1	163	1	0.6	2	0	

#To print last five rows of the data

df.tail()

		age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca
	298	57	0	0	140	241	0	1	123	1	0.2	1	0
	222	4.5	4	^	440	004	^	4	100	^	4.0	4	^
# number of rows and columns in the dataset													
	301	57	1	0	130	131	0	1	115	1	1.2	1	1
df.sh	ape												
	(303,	14)											

Details about the dataset

df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 303 entries, 0 to 302 Data columns (total 14 columns): Column Non-Null Count Dtype _____ ---0 age 303 non-null int64 1 sex 303 non-null int64 2 303 non-null int64 ср 3 trestbps 303 non-null int64 4 chol 303 non-null int64 5 fbs 303 non-null int64 6 restecg 303 non-null int64 7 thalach 303 non-null int64 8 exang 303 non-null int64 9 303 non-null float64 oldpeak 10 slope 303 non-null int64 11 ca 303 non-null int64 12 thal 303 non-null int64 13 target 303 non-null int64 dtypes: float64(1), int64(13) memory usage: 33.3 KB

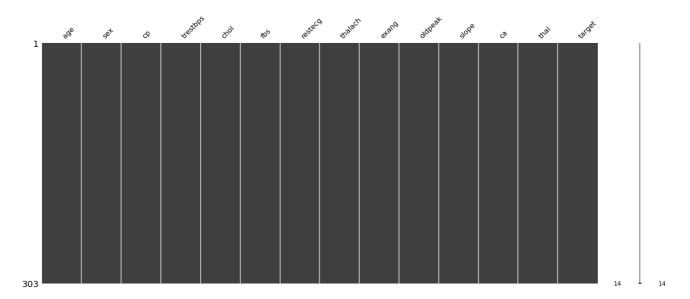
Cross checking the missing values

df.isnull().sum()

age 0 sex 0 ср trestbps 0 chol 0 fbs 0 restecg 0 thalach 0 exang 0 oldpeak 0 slope 0 0 ca

thal 0 target 0 dtype: int64

import missingno as msno
msno.matrix(df)
plt.show()



statistical measures about the data

df.describe()

```
trestbps
                                                                       chol
                                                                                      fbs
                     age
                                  sex
                                                ср
      count 303.000000 303.000000
                                       303.000000
                                                    303.000000 303.000000 303.000000
               E1 266227
                             N 602160
       maan
                                          0.066007
                                                    121 622762 246 264026
                                                                                ∩ 1/Q515
#Distribution of Target Variable
df['target'].value_counts()
     1
           165
     0
           138
     Name: target, dtype: int64
#Splitting the data into Features and Labels
X = df.drop(columns='target', axis=1)
Y = df['target']
print(X)
                          trestbps
                                     chol
                                                         oldpeak
                                                                   slope
                                                                                thal
           age
                sex
                      ср
                                            . . .
                                                  exang
                                                                            ca
     0
            63
                   1
                       3
                                145
                                       233
                                                      0
                                                              2.3
                                                                        0
                                                                             0
                                                                                    1
     1
            37
                       2
                                       250
                                                              3.5
                                                                                    2
                   1
                                130
                                                      0
                                                                        0
                                                                             0
     2
            41
                   0
                       1
                                130
                                       204
                                                      0
                                                              1.4
                                                                        2
                                                                             0
                                                                                    2
                                            . . .
     3
                                                                        2
                                                                                    2
            56
                   1
                       1
                                120
                                       236
                                                      0
                                                              0.8
                                                                             0
     4
            57
                       0
                                       354
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                                                                                    2
                   0
                                120
                                                      1
                                                              0.6
                                                                             0
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     298
            57
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                       0
                                140
                                       241
                                                      1
                                                              0.2
                                                                        1
                                                                             0
                                                                                    3
                                            . . .
     299
            45
                       3
                                                              1.2
                                                                                    3
                   1
                                110
                                       264
                                                      0
                                                                        1
                                                                             0
     300
            68
                   1
                       0
                                144
                                       193
                                                      0
                                                              3.4
                                                                        1
                                                                             2
                                                                                    3
     301
            57
                   1
                       0
                                130
                                       131
                                                      1
                                                              1.2
                                                                        1
                                                                             1
                                                                                    3
                                            . . .
     302
                                                              0.0
                                                                                    2
            57
                   0
                       1
                                130
                                       236
                                                      0
                                                                        1
                                                                             1
     [303 rows x 13 columns]
print(Y)
     0
             1
     1
             1
     2
             1
     3
             1
     4
             1
     298
             0
     299
             0
     300
             0
     301
             0
     302
     Name: target, Length: 303, dtype: int64
```

#Separating the Data into Training data & Test Data

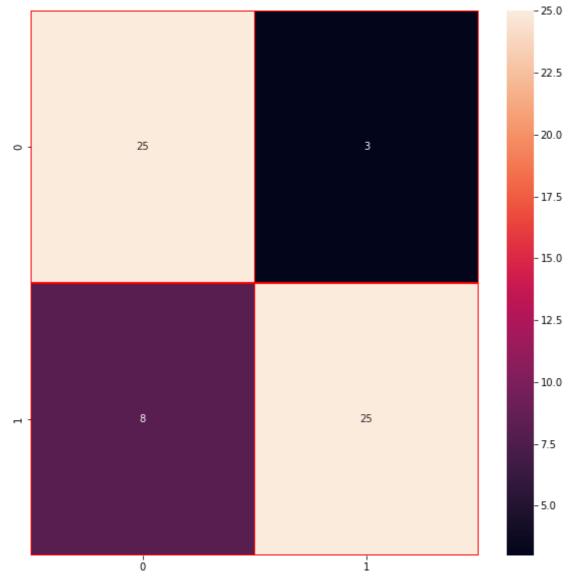
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```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, stratify=Y, rando
print(X.shape, X_train.shape, X_test.shape)
     (303, 13) (242, 13) (61, 13)
# Model Training
#Naive Bayes Classifier
from sklearn.naive_bayes import GaussianNB
from sklearn import metrics
classifier = GaussianNB()
classifier.fit(X_train, Y_train)
     GaussianNB()
#Model Evaluation
#Accuracy Score
# accuracy on training data
X train prediction = classifier.predict(X train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)
print('Accuracy on Training data : ', training_data_accuracy)
     Accuracy on Training data : 0.8471074380165289
# accuracy on training data
X_test_prediction = classifier.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
print('Accuracy on Test data : ', test_data_accuracy)
     Accuracy on Test data: 0.819672131147541
```

```
#Building a Predictive System
input data = (56,1,2,130,256,1,0,142,1,0.6,1,1,1)
input_data = (43,1,0,115,303,0,1,181,0,1.2,1,0,2)
# change the input data to a numpy array
input_data_as_numpy_array= np.asarray(input_data)
# reshape the numpy array as we are predicting for only on instance
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
prediction = classifier.predict(input_data_reshaped)
print(prediction)
     [0]
     /usr/local/lib/python3.7/dist-packages/sklearn/base.py:451: UserWarning: X does not h
       "X does not have valid feature names, but"
if (prediction[0]== 0):
  print('The Person does not have a Heart Disease')
else:
  print('The Person has Heart Disease')
     The Person does not have a Heart Disease
from sklearn.metrics import confusion_matrix
cm = confusion matrix(Y test, Y pred)
f,ax = plt.subplots(figsize=(10, 10))
sns.heatmap(cm, annot=True, linewidths=0.5,linecolor="red", fmt= '.0f',ax=ax)
plt.show()
plt.savefig('ConfusionMatrix.png')
```



<Figure size 432x288 with 0 Axes>

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