

```
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np

df=pd.read_csv('/content/drive/MyDrive/Heart.csv')

df.head()
```

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

```
#Find Shape of Data
df.shape
```

```
(303, 15)
```

```
#Find data type of each column
df.dtypes
```

```

Unnamed: 0      int64
Age             int64
Sex             int64
ChestPain      object
RestBP         int64
Chol           int64
Fbs            int64
RestECG        int64
MaxHR          int64
ExAng          int64
Oldpeak        float64
Slope          int64
Ca             float64
Thal           object
AHD            object
dtype: object
```

```
#Find Mean age of patients
mean1 = df['Age'].mean()
print ('Mean Age: ' + str(mean1))
```

```
Mean Age: 54.43894389438944
```

```
#Finding out Zero's
(df==0).sum()
```

```

Unnamed: 0      0
Age             0
Sex            97
```

```

ChestPain      0
RestBP         0
Chol           0
Fbs            258
RestECG        151
MaxHR          0
ExAng          204
Oldpeak        99
Slope          0
Ca             176
Thal           0
AHD            0
dtype: int64

```

```

#Find Missing Values
df.isna().sum()

```

```

Unnamed: 0      0
Age             0
Sex             0
ChestPain       0
RestBP          0
Chol            0
Fbs             0
RestECG         0
MaxHR           0
ExAng           0
Oldpeak         0
Slope           0
Ca              4
Thal            2
AHD             0
dtype: int64

```

```

from sklearn.model_selection import train_test_split

```

```

train,test=train_test_split(df[["Age","Sex","ChestPain","RestBP","Chol"]], test_size=0.2)
print("Training Data:",train)
print("Testing Data:",test)

```

```

Training Data:      Age  Sex      ChestPain  RestBP  Chol
83    68    1    nonanginal      180    274
108   61    1  asymptomatic      120    260
186   42    1    nonanginal      120    240
273   71    0  asymptomatic      112    149
141   59    1      typical      170    288
..    ...    ...          ...      ...
272   46    1  asymptomatic      140    311
239   42    1    nontypical      120    295
68    59    1  asymptomatic      170    326
283   35    1    nontypical      122    192
152   67    0    nonanginal      115    564

```

```

[242 rows x 5 columns]
Testing Data:      Age  Sex      ChestPain  RestBP  Chol
116   58    1    nonanginal      140    211
125   45    0    nontypical      130    234
190   50    1    nonanginal      129    196
291   55    0    nontypical      132    342
211   38    1      typical      120    231
..    ...    ...          ...      ...
212   41    1    nonanginal      130    214

```

131	51	1	nonanginal	94	227
37	57	1	asymptomatic	150	276
134	43	0	nonanginal	122	213
285	58	1	asymptomatic	114	318

[61 rows x 5 columns]

```
df[(df['AHD'] == 'Yes')].min()
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: FutureWarning: Dropping of nuisance
    """Entry point for launching an IPython kernel.
Unnamed: 0          2
Age                35
Sex                0
ChestPain          asymptomatic
RestBP             100
Chol               131
Fbs                0
RestECG            0
MaxHR              71
ExAng              0
Oldpeak            0.0
Slope              1
Ca                 0.0
AHD                Yes
dtype: object
```

```
df[(df['AHD'] == 'Yes')].max()
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: FutureWarning: Dropping of nuisance
    """Entry point for launching an IPython kernel.
Unnamed: 0          302
Age                77
Sex                1
ChestPain          typical
RestBP             200
Chol               409
Fbs                1
RestECG            2
MaxHR              195
ExAng              1
Oldpeak            6.2
Slope              3
Ca                 3.0
AHD                Yes
dtype: object
```

```
from sklearn import metrics
from sklearn.metrics import confusion_matrix
```

```
y_true = [2, 0, 2, 2, 0, 1]
y_pred = [0, 0, 2, 2, 0, 2]
confusion_matrix(y_true, y_pred)
print(metrics.classification_report(y_true, y_pred, labels=[0,1,2]))
```

	precision	recall	f1-score	support
0	0.67	1.00	0.80	2
1	0.00	0.00	0.00	1
2	0.67	0.67	0.67	3
accuracy			0.67	6
macro avg	0.44	0.56	0.49	6
weighted avg	0.56	0.67	0.60	6

```

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision is based on zero labeled samples.
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall is based on zero labeled samples.
  _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: F1-score is based on zero labeled samples.
  _warn_prf(average, modifier, msg_start, len(result))

```

```
from sklearn.metrics import *
```

```

predicted_positive=[0]*400+[1]*100;
actual_pos=[1]*50+[0]*450;
print("Confusion Matrix\n",confusion_matrix(actual_pos, predicted_positive))
print("Accuracy Score \n",accuracy_score(actual_pos,predicted_positive))
print("Precision Score \n",precision_score(actual_pos,predicted_positive))
print("Recall Score \n",recall_score(actual_pos, predicted_positive,average=None))
print("f1 Score \n",f1_score(actual_pos, predicted_positive, average=None))

```

```

Confusion Matrix
[[350 100]
 [ 50   0]]
Accuracy Score
0.7
Precision Score
0.0
Recall Score
[0.77777778 0.         ]
f1 Score
[0.82352941 0.         ]

```