

## \*\* PROJECT DAY - 17 \*\*

### [ Heart Disease Prediction Using Machine Learning ]

#### Importing the Dependencies

```
In [136... import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
```

#### Data collection and Processing

```
In [139... # Loading the csv data to a pandas DataFrame
data=pd.read_csv('heart_disease_data.csv')
```

```
In [140... # print first 5 rows of the dataset
data.head()
```

```
Out[140...   age  sex  cp  trestbps  chol  fbs  restecg  thalach  exang  oldpeak  slope  ca  thal  target
0    63     1    3      145    233     1      0     150     0     2.3     0     0     1     1     1
1    37     1    2      130    250     0      1     187     0     3.5     0     0     2     1
2    41     0    1      130    204     0      0     172     0     1.4     2     0     2     1
3    56     1    1      120    236     0      1     178     0     0.8     2     0     2     1
4    57     0    0      120    354     0      1     163     1     0.6     2     0     2     1
```

```
In [141... # print last 5 rows of the dataset
data.tail()
```

```
Out[141...   age  sex  cp  trestbps  chol  fbs  restecg  thalach  exang  oldpeak  slope  ca  thal  target
298   57     0    0      140    241     0      1     123     1     0.2     1     0     3     0
299   45     1    3      110    264     0      1     132     0     1.2     1     0     3     0
300   68     1    0      144    193     1      1     141     0     3.4     1     2     3     0
301   57     1    0      130    131     0      1     115     1     1.2     1     1     3     0
302   57     0    1      130    236     0      0     174     0     0.0     1     1     2     0
```

```
In [142... # number of rows and columns in the dataset
data.shape
```

```
Out[142... (303, 14)
```

```
In [143... # getting some info about data
data.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   cp        303 non-null   int64  
 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   oldpeak   303 non-null   float64 
 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
```

```
In [144... # checking for missing values
data.isnull().sum()
```

```
Out[144... age      0
sex      0
cp       0
trestbps 0
chol     0
fbs      0
restecg  0
thalach   0
exang    0
oldpeak   0
slope    0
ca       0
thal     0
target    0
dtype: int64
```

```
In [145... # statistical measures about the data
data.describe()
```

```
Out[145...   age      sex      cp      trestbps      chol      fbs      restecg      thalach      exang      oldpeak      slope      ca      thal      target
count  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000  303.000000
mean   54.366337  0.683168  0.966997  131.623762  246.264026  0.148515  0.528053  149.646865  0.326733  1.039604  1.399340  0.729373  2.313531  0.544554
std    9.082101  0.466011  1.032052  17.539143  51.830751  0.356198  0.525860  22.905161  0.469794  1.161075  0.616226  1.022606  0.612277  0.498835
min    29.000000  0.000000  0.000000  94.000000  126.000000  0.000000  0.000000  71.000000  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
25%   47.500000  0.000000  0.000000  120.000000  211.000000  0.000000  0.000000  133.500000  0.000000  1.000000  0.100000  0.000000  2.000000  0.000000
50%   55.000000  1.000000  1.000000  130.000000  240.000000  0.000000  1.000000  153.000000  0.000000  0.800000  1.000000  0.000000  2.000000  1.000000
75%   61.000000  1.000000  2.000000  140.000000  274.500000  0.000000  1.000000  166.000000  1.000000  1.600000  2.000000  1.000000  3.000000  1.000000
max    77.000000  1.000000  3.000000  200.000000  564.000000  1.000000  2.000000  202.000000  1.000000  6.200000  2.000000  4.000000  3.000000  1.000000
```

```
In [146... # checking the distribution of Target variable
data['target'].value_counts()
```

```
Out[146... target
1    165
0    138
Name: count, dtype: int64
```

1-> Defective Heart

2-> Healthy Heart

#### Splitting the Features and Target

```
In [149... X=data.drop(columns='target', axis=1)
Y=data['target']
```

```
In [150... X
```

```
Out[150...   age  sex  cp  trestbps  chol  fbs  restecg  thalach  exang  oldpeak  slope  ca  thal
0    63     1    3      145    233     1      0     150     0     2.3     0     0     1
1    37     1    2      130    250     0      1     187     0     3.5     0     0     2
2    41     0    1      130    204     0      0     172     0     1.4     2     0     2
3    56     1    1      120    236     0      1     178     0     0.8     2     0     2
4    57     0    0      120    354     0      1     163     1     0.6     2     0     2
...
298   57     0    0      140    241     0      1     123     1     0.2     1     0     3
299   45     1    3      110    264     0      1     132     0     1.2     1     0     3
300   68     1    0      144    193     1      1     141     0     3.4     1     2     3
301   57     1    0      130    131     0      1     115     1     1.2     1     1     3
302   57     0    1      130    236     0      0     174     0     0.0     1     1     2
```

303 rows × 13 columns

```
In [151... Y
```

```
Out[151... 0      1
1      1
2      1
3      1
4      1
...
298   0
299   0
300   0
301   0
302   0
Name: target, Length: 303, dtype: int64
```

#### Splitting the Data into Training data & Test data

```
In [153... X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.2,stratify=Y,random_state=2)
```

```
In [154... print(X.shape,X_train.shape,X_test.shape)
```

(303, 13) (242, 13) (61, 13)

#### Model Training

##### Logistic Regression

```
In [157... from sklearn.preprocessing import StandardScaler
```

```
scaler = StandardScaler()
```

```
X_train_scaled = scaler.fit_transform(X_train)
```

```
X_test_scaled = scaler.transform(X_test)
```

```
In [158... model = LogisticRegression(max_iter=2000)
```

```
In [159... # training the LogisticRegression model with training Data
model.fit(X_train, Y_train)
```

```
Out[159... LogisticRegression(max_iter=2000)
```

#### Model Evaluation

##### Accuracy Score

```
In [162... # accuracy on training data
X_train_prediction=model.predict(X_train)
```

```
training_data_accuracy=accuracy_score(X_train_prediction,Y_train)
```

```
In [163... print("Accuracy on Training data:",training_data_accuracy)
```

Accuracy on Training data: 0.8553719008264463

```
In [164... # accuracy on test data
X_test_prediction=model.predict(X_test)
```

```
test_data_accuracy=accuracy_score(X_test_prediction,Y_test)
```

```
In [164... print("Accuracy on Testing data:",test_data_accuracy)
```

Accuracy on Testing data: 0.8032786885245902

#### Buiding a Predictive System

```
In [187... input_data = (45,1,3,110,264,0,1,132,0,1,2,1,0,3)
```

```
# (37,1,2,130,250,0,1,187,0,3,5,0,0,2)
```

```
# convert input data to a DataFrame with feature names
```

```
input_df = pd.DataFrame([input_data],
```

```
           columns=X_train.columns # SAME columns used during training
```

```
)
```

```
# prediction
```

```
prediction = model.predict(input_df)
```

```
print(prediction)
```

```
If (prediction[0]==0):
```

```
    print("The person does not have Heart Disease")
```

```
else:
```

```
    print("The person has Heart Disease")
```

```
[1]
```

The person has Heart Disease

**\*\* END - PROJECT \*\***

In [1]: