## Importing the Dependencies

import numpy as np
import pandas as pd
from sklearn.model\_selection import train\_test\_split
from sklearn.linear\_model import LogisticRegression
from sklearn.metrics import accuracy\_score

## **Data Collection and Processing**

# loading the csv data to a Pandas DataFrame
heart\_data = pd.read\_csv('/content/data.csv')

# print first 5 rows of the dataset
heart\_data.head()

₹		age	sex	ср	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	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

# print last 5 rows of the dataset
heart\_data.tail()

<b>→</b>		age	sex	ср	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

# number of rows and columns in the dataset
heart\_data.shape

**→** (303, 14)

# getting some info about the data
heart\_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	ср	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						
dtvb	ltypes: float64(1), int64(13)									

dtypes: float64(1), int64(13)
memory usage: 33.3 KB

# checking for missing values
heart\_data.isnull().sum()

	^
age	U
sex	0
ср	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	4
	cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal target

# statistical measures about the data heart\_data.describe()

7	age	sex	ср	trestbps	chol	fbs	restecg	thalach	
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.
mean	54.366337	0.683168	0.966997	131.623762	246.264026	0.148515	0.528053	149.646865	0.
std	9.082101	0.466011	1.032052	17.538143	51.830751	0.356198	0.525860	22.905161	0.
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	71.000000	0.
25%	47.500000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	133.500000	0.
50%	55.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	153.000000	0.
75%	61.000000	1.000000	2.000000	140.000000	274.500000	0.000000	1.000000	166.000000	1.
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	202.000000	1.

# checking the distribution of Target Variable
heart\_data['target'].value\_counts()

```
₹
    1
           165
     0
           138
     Name: target, dtype: int64
1 --> Defective Heart
0 --> Healthy Heart
Splitting the Features and Target
X = heart_data.drop(columns='target', axis=1)
Y = heart_data['target']
print(X)
                                                          oldpeak slope
\overline{2}
                          trestbps
                                      chol
                                                  exang
                                                                             ca
                                                                                 thal
           age
                sex
                      ср
                                             . . .
                                                               2.3
            63
                  1
                       3
                                145
                                       233
                                                       0
                                                                         0
                                                                              0
                                             . . .
     1
            37
                       2
                                130
                                       250
                                                       0
                                                               3.5
                                                                         0
                                                                              0
                                                                                     2
                  1
                                             . . .
     2
            41
                  0
                                130
                                       204
                                                       0
                                                               1.4
                                                                         2
                                                                              0
                                                                                     2
                       1
                                             . . .
                                                                         2
                                                                                     2
     3
            56
                  1
                                                       0
                                                                              0
                       1
                                120
                                       236
                                                               0.8
                                             . . .
                       0
                                                                         2
                                                                                     2
     4
            57
                  0
                                120
                                       354
                                                       1
                                                               0.6
                                                                              0
                                             . . .
                                             . . .
     298
            57
                       0
                                140
                                       241
                                                               0.2
                                                                              0
                                                                                     3
                  0
                                                       1
                                                                         1
                                             . . .
     299
            45
                                                               1.2
                  1
                       3
                                110
                                       264
                                                       0
                                                                         1
                                                                              0
                                                                                     3
     300
                                                                                     3
            68
                   1
                       0
                                144
                                       193
                                                       0
                                                               3.4
                                                                         1
                                                                              2
                                             . . .
                                                                                     3
     301
            57
                   1
                       0
                                130
                                       131
                                                       1
                                                               1.2
                                                                         1
                                                                              1
                                             . . .
     302
            57
                   0
                       1
                                130
                                       236
                                                       0
                                                               0.0
                                                                         1
                                                                                     2
                                             . . .
     [303 rows x 13 columns]
print(Y)
→
     0
             1
     1
             1
     2
             1
     3
             1
     298
             0
     299
             0
     300
             0
     301
             0
     302
     Name: target, Length: 303, dtype: int64
Splitting the Data into Training data & Test Data
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, stratify=Y, random_state=
print(X.shape, X_train.shape, X_test.shape)
→ (303, 13) (242, 13) (61, 13)
```

**Model Training** 

```
model = LogisticRegression()
# training the LogisticRegression model with Training data
model.fit(X train, Y train)
→ /usr/local/lib/python3.7/dist-packages/sklearn/linear_model/_logistic.py:940: ConvergenceWarni
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear model.html#logistic-regression
      extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG)
    LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                        intercept_scaling=1, l1_ratio=None, max_iter=100,
                       multi_class='auto', n_jobs=None, penalty='12',
                       random_state=None, solver='lbfgs', tol=0.0001, verbose=0,
                       warm_start=False)
Model Evaluation
Accuracy Score
# accuracy on training data
X train prediction = model.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 : 0.8512396694214877
# accuracy on test data
X_test_prediction = model.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 = (62,0,0,140,268,0,0,160,0,3.6,0,2,2)
# 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 = model.predict(input_data_reshaped)
print(prediction)
```

```
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
```