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/heart disease dataset.csv')

#print first five rows of the dataset
heart_data.head()

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

Saved successfully!

X

heart_data.tail()

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak
298	57	0	0	140	241	0	1	123	1	0.2
299	45	1	3	110	264	0	1	132	0	1.2
300	68	1	0	144	193	1	1	141	0	3.4
301	57	1	0	130	131	0	1	115	1	1.2
302	57	0	1	130	236	0	0	174	0	0.0

#number of rows and columns in the Dataset
heart_data.shape

(303, 14)

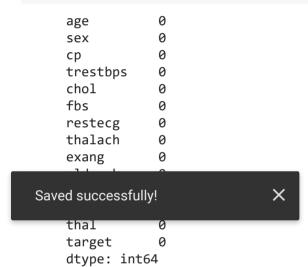
#getting some info about the data
heart_data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302

Data	columns (tota:	l 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		
dtype	dtypes: float64(1), int64(13)					
momony usage 22 2 VD						

memory usage: 33.3 KB

#checking for missing values heart_data.isnull().sum()



#statistical measures about the data heart_data.describe()

	age	sex	ср	trestbps	chol	fb
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.00000
mean	54.366337	0.683168	0.966997	131.623762	246.264026	0.14851
std	9.082101	0.466011	1.032052	17.538143	51.830751	0.35619
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.00000
25%	47.500000	0.000000	0.000000	120.000000	211.000000	0.00000
50%	55.000000	1.000000	1.000000	130.000000	240.000000	0.00000
75%	61.000000	1.000000	2.000000	140.000000	274.500000	0.00000
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.00000

```
#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 Target and Features column
X=heart_data.drop(columns='target',axis=1)
Y=heart_data['target']
print(X)
         age sex cp trestbps chol ... exang oldpeak slope ca
                                                                   thal
             1
    0
          63
                  3
                           145
                                233
                                     . . .
                                         0
                                                    2.3
    1
                                250 ...
                                                            0 0
          37 1 2
                           130
                                            0
                                                    3.5
                                                                      2
    2
          41 0 1
                           130
                                204 ...
                                            0
                                                    1.4
                                                           2 0
                                                                      2
          56 1 1
    3
                           120
                                236 ...
                                            0
                                                    0.8
                                                           2 0
                                                                      2
          57
              0 0
                                354 ...
                                            1
                                                           2 0
                                                                      2
                           120
                                                    0.6
                                241 ... 0
0
                  . .
                           . . .
                                                    . . .
                                                           . . . . . .
    298
          57
              0
                  0
                           140
                                             1
                                                    0.2
                                                           1 0
                                                                      3
                                                    1.2
                                                            1 0
    299
          45
                1
                   3
                           110
                                                                      3
                                            0
1
    300
          68
                   0
                           144
                                                   3.4
                                                           1 2
                                                                      3
          57
                           130
                                131 ...
                                                           1 1
                                                                      3
    301
                                                   1.2
                                          0
                                 236 ...
                                                    0.0 1 1
                                                                      2
 Saved successfully!
print(Y)
    0
           1
    1
    2
           1
    3
           1
    4
           1
    298
           0
    299
           0
    300
           0
    301
           0
    302
    Name: target, Length: 303, dtype: int64
Splitting the Data into Training data and Testing data
 X\_train, \ X\_test, Y\_train, \ Y\_test=train\_test\_split(X,Y, \ test\_size=0.2, \ stratify=Y, random\_state=2)
```

print(X_train.shape,X_test.shape)

(242, 13) (61, 13)

MODEL TRAINING

LOGISTIC REGRESSION

```
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)
    4
```

MODEL EVALUATION

```
#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

#accurracy on test data
X_test_prediction=model.predict(X_test)
test_data_accuracy=accuracy_score(X_test_prediction,Y_test)
```

Accuracy on Test data: 0.819672131147541

print('Accuracy on Test data:',test_data_accuracy)

BUILDING A PEDICTIVE SYSTEM

```
input_data = (66,0,3,150,226,0,1,114,0,2.6,0,0,2)
```

```
# 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 = 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')

[1]
    The Person has Heart Disease
```

Saved successfully!