

9_Predicting_Heart_Disease_Using_Logistic_Regression_and_ML_

July 6, 2024

0.1 Heart Disease Prediction Using Machine Learning: A Logistic Regression Approach - Vignesh Prabhu

This project focuses on predicting heart disease using machine learning, specifically employing logistic regression. Clinical data was analyzed to identify key risk factors and develop an accurate and reliable predictive model. The goal is to assist healthcare professionals in early diagnosis and effective treatment planning of heart disease.

Import Dependencies

```
[36]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
```

Data Collection and Preprocessing

```
[37]: #Load the data to Pandas dataframe
heart_data=pd.read_csv("/content/heart_disease_data.csv")
```

```
[38]: # To print first 5 data's in dataset
heart_data.head()
```

```
[38]:
```

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

	ca	thal	target
0	0	1	1
1	0	2	1
2	0	2	1
3	0	2	1

4 0 2 1

```
[39]: #To check Rows and columns
heart_data.shape
```

[39]: (303, 14)

```
[40]: #To check informations
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   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
```

```
[41]: #statistical datas
heart_data.describe()
```

```
[41]:
```

	age	sex	cp	trestbps	chol	fbs	\
count	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	
std	9.082101	0.466011	1.032052	17.538143	51.830751	0.356198	
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	
25%	47.500000	0.000000	0.000000	120.000000	211.000000	0.000000	
50%	55.000000	1.000000	1.000000	130.000000	240.000000	0.000000	
75%	61.000000	1.000000	2.000000	140.000000	274.500000	0.000000	
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	

	restecg	thalach	exang	oldpeak	slope	ca	\
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	

mean	0.528053	149.646865	0.326733	1.039604	1.399340	0.729373
std	0.525860	22.905161	0.469794	1.161075	0.616226	1.022606
min	0.000000	71.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	133.500000	0.000000	0.000000	1.000000	0.000000
50%	1.000000	153.000000	0.000000	0.800000	1.000000	0.000000
75%	1.000000	166.000000	1.000000	1.600000	2.000000	1.000000
max	2.000000	202.000000	1.000000	6.200000	2.000000	4.000000

	thal	target
count	303.000000	303.000000
mean	2.313531	0.544554
std	0.612277	0.498835
min	0.000000	0.000000
25%	2.000000	0.000000
50%	2.000000	1.000000
75%	3.000000	1.000000
max	3.000000	1.000000

```
[42]: #To check any Null values in Data
heart_data.isnull().sum()
```

```
[42]: 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
```

```
[43]: #checking Distribution of Target values
heart_data['target'].value_counts() # 1-Defective Heart 0- healthy heart
```

```
[43]: target
1      165
0      138
Name: count, dtype: int64
```

Spliting Feautres and Target

```
[44]: X=heart_data.drop(columns='target',axis=0)
      Y=heart_data['target']
```

```
[45]: print(X)
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	\
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	
..			
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	

	slope	ca	thal
0	0	0	1
1	0	0	2
2	2	0	2
3	2	0	2
4	2	0	2
..
298	1	0	3
299	1	0	3
300	1	2	3
301	1	1	3
302	1	1	2

[303 rows x 13 columns]

```
[46]: print(Y)
```

```
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 Training and Testing data

```
[47]: X_train, X_test, Y_train , Y_test=train_test_split(X,Y,test_size=0.  
↪2,stratify=Y,random_state=2) #Stratify used for split data even manner
```

```
[48]: print(X.shape,X_train.shape,X_test.shape)
```

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

Model Training

```
[49]: logistic_model=LogisticRegression()
```

```
[50]: logistic_model.fit(X_train,Y_train)
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:458:  
ConvergenceWarning: lbfgs failed to converge (status=1):  
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

```
n_iter_i = _check_optimize_result(
```

```
[50]: LogisticRegression()
```

Model Evaluation

```
[51]: #accuracy score  
X_train_prediction=logistic_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

```
[52]: #Accuracy For test data  
X_test_prediction=logistic_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

Bulding Predictive System

```
[57]: input_data=(44,1,2,130,233,0,1,179,1,0.4,2,0,2,)  
#change input data to numpy array  
input_data_as_numpy_array=np.asarray(input_data)  
#reshape the numpy array as we are predicting for only one instance
```

```

input_data_resaped=input_data_as_numpy_array.reshape(1,-1)
prediction=logistic_model.predict(input_data_resaped)
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

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but LogisticRegression was fitted with feature names

```
warnings.warn(
```

```

[58]: input_data=(62,0,0,140,268,0,0,160,0,3.6,0,2,2)
      #change input data to numpy array
      input_data_as_numpy_array=np.asarray(input_data)
      #reshape the numpy array as we are predicting for only one instance
      input_data_resaped=input_data_as_numpy_array.reshape(1,-1)
      prediction=logistic_model.predict(input_data_resaped)
      print(prediction)
      if(prediction[0]==0):
          print("The Person does not have a Heart Disease")
      else:
          print("The Person has Heart Disease")

```

[0]

The Person does not have a Heart Disease

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but LogisticRegression was fitted with feature names

```
warnings.warn(
```

This project developed a heart disease prediction model using logistic regression, which demonstrated strong predictive accuracy. By analyzing clinical data, we identified key risk factors and created a reliable tool for early diagnosis. Future work could enhance the model by integrating additional algorithms and more diverse datasets, further supporting healthcare professionals in improving patient outcomes.

0.2 Thank You!