logistic-regression-hd

August 25, 2023

```
[]: import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
```

##Name:

##Project title:To predict the heart attack disease for orgnization WHO(World Health Organization), using machine learning Algorithm rate of heart_Attack disease will increasing or decreasing manner

1 Problem Statement: A world health organization estimated 12 millions death records. One of them half of the daeath results is found in US. The Research Scholars point out the most relevant risk factor of Heart Attack AS a Data Science Engineer predict the overall risks using machine learning algorithm whichever them called as logistic Regression

##tasks

##import the libraries which is required for prediction ##import the dataset your using workspace ##use the approprite arguement of sklearn librry to train, test, split the dataset ## Chech ur model accuracy and precison using confusion matrix

```
[]: from google.colab import files files=files.upload()
```

<IPython.core.display.HTML object>

Saving framingham.csv to framingham.csv

```
[]: df=pd.read_csv('framingham.csv') df
```

```
[]:
                         education currentSmoker
                                                       cigsPerDay
                                                                     BPMeds
            \mathtt{male}
                   age
                1
                    39
                                4.0
                                                               0.0
                                                                         0.0
     0
                                                    0
     1
                0
                                                               0.0
                    46
                                2.0
                                                    0
                                                                         0.0
```

```
2
                   48
                             1.0
                                                         20.0
                                                                   0.0
              1
                                                1
     3
              0
                   61
                             3.0
                                                1
                                                         30.0
                                                                   0.0
     4
              0
                   46
                             3.0
                                                         23.0
                                                                   0.0
                                                1
     4233
              1
                   50
                             1.0
                                                1
                                                          1.0
                                                                   0.0
     4234
                   51
                             3.0
                                                1
                                                         43.0
                                                                   0.0
              1
     4235
                             2.0
                                                         20.0
                                                                   NaN
              0
                   48
                                                1
     4236
              0
                   44
                              1.0
                                                1
                                                         15.0
                                                                   0.0
     4237
                   52
                             2.0
                                                0
                                                          0.0
                                                                   0.0
              0
           prevalentStroke
                             prevalentHyp
                                            diabetes
                                                       totChol
                                                                 sysBP
                                                                         diaBP
                                                                                  BMI
     0
                          0
                                                    0
                                                         195.0
                                                                106.0
                                                                          70.0
                                                                                26.97
     1
                          0
                                         0
                                                         250.0 121.0
                                                                          81.0
                                                                                28.73
                                                    0
     2
                          0
                                         0
                                                    0
                                                         245.0
                                                                 127.5
                                                                          80.0
                                                                                25.34
     3
                          0
                                         1
                                                    0
                                                         225.0
                                                                 150.0
                                                                          95.0
                                                                                28.58
     4
                          0
                                         0
                                                    0
                                                         285.0
                                                                          84.0
                                                                                23.10
                                                                 130.0
     4233
                          0
                                                    0
                                                         313.0
                                                                179.0
                                                                          92.0
                                                                                25.97
                                         1
     4234
                          0
                                         0
                                                         207.0 126.5
                                                                          80.0 19.71
                                                    0
     4235
                                         0
                                                                131.0
                                                                                22.00
                          0
                                                    0
                                                         248.0
                                                                          72.0
     4236
                          0
                                         0
                                                    0
                                                         210.0
                                                                126.5
                                                                          87.0 19.16
     4237
                          0
                                         0
                                                         269.0 133.5
                                                    0
                                                                          83.0
                                                                                21.47
                       glucose
           heartRate
                                TenYearCHD
                 80.0
     0
                          77.0
                                          0
                 95.0
     1
                          76.0
                                          0
                 75.0
     2
                          70.0
                                          0
     3
                 65.0
                         103.0
                                          1
     4
                 85.0
                          85.0
                                          0
     4233
                 66.0
                          86.0
                                          1
     4234
                 65.0
                                          0
                          68.0
     4235
                 84.0
                          86.0
                                          0
     4236
                 86.0
                                          0
                           NaN
     4237
                 80.0
                         107.0
                                          0
     [4238 rows x 16 columns]
[]: x=df[['age']]
     y=df[['currentSmoker']]
[]:
[]: x_test,x_train,y_test,y_train=train_test_split(x,y,test_size=0.4,random_state=0)
     print(x train)
```

age

```
1669
       47
156
       58
87
       61
685
       45
666
       57
2790
       53
1855
       66
700
       60
2060
       38
2348
       48
```

[1696 rows x 1 columns]

[]: print(y_train)

currentSmoker
0
0
1
0
0
•••
0
0
0
0
1

[1696 rows x 1 columns]

[]: print(x_test)

```
age
       42
3218
590
       60
3880
       41
1548
       59
2601
       55
1033
       44
3264
       51
1653
       39
2607
       57
2732
       40
```

[2542 rows x 1 columns]

```
[]: print(y_test)
          currentSmoker
    3218
    590
                       1
    3880
                       0
                       0
    1548
    2601
                       1
                       0
    1033
    3264
                       1
    1653
                       1
    2607
                       0
    2732
                       1
    [2542 rows x 1 columns]
[]: from sklearn.preprocessing import StandardScaler
     sc =StandardScaler()
     x_train = sc.fit_transform(x_train)
     x_test = sc.transform(x_test)
[]: print(x_train)
    [[-0.2920309]
     [ 0.9841763 ]
     [ 1.33223281]
     [ 1.21621397]
     [-1.33620043]
     [-0.17601207]]
[]: print(x_test)
    [[-0.87212509]
     [ 1.21621397]
     [-0.98814392]
     [-1.2201816]
     [ 0.86815746]
     [-1.10416276]]
[]: from sklearn.linear_model import LogisticRegression
     classifier = LogisticRegression(random_state = 0)
     classifier.fit(x_train, y_train)
```

/usr/local/lib/python3.10/dist-packages/sklearn/utils/validation.py:1143: DataConversionWarning: A column-vector y was passed when a 1d array was

```
expected. Please change the shape of y to (n_samples, ), for example using
   ravel().
     y = column_or_1d(y, warn=True)
[]: LogisticRegression(random_state=0)
[]: y_pred = classifier.predict(x_test)
[]: y_pred
[]: array([1, 0, 1, ..., 1, 0, 1])
[]: from sklearn.metrics import confusion_matrix, accuracy_score
    cm = confusion_matrix(y_test, y_pred)
    print(cm)
    accuracy_score(y_test, y_pred)
    [[775 495]
    [517 755]]
[]: 0.6018882769472856
    ##CONCLUSIONS ##ACCORDING TO THE MODEL ANALYSIS THE LOGISTIC REGRE-
    SION ALGORITHM WORKS SUCCESSFULLY WITH 0.6 ACCURACY. ##THE ACCURACY
    SHOWS THAT BUILDING THE MODEL IS SUCCEEFUL
[]:
[]:
```