heart-attack3

August 25, 2023

NAME: SUMANTH

ROLL NO: 21X05A6704

BRANCH : CSE(DS)

NRCM

PROJECT TITLE

Predict the heart attack disease for organization whO using machine learning algorithm rate of heart attack disease will incresing manner or decressing manner

##PROBLEM STATMENT

A world heard organization estimated 12 millions Dead records. One of them half of the dead result is found in US. The research scsholer point out the most relevent risk factor of heartattack as a data science engineer predict the overall risk using machine learning algorithm is called logistic regression

##Task

- 1. Import the libraries which is the required for prediction.
- 2. Import the data set using work space.
- 3. Use a appropriate argument of sklearn librarie to train, test, and Split the data sets.
- 4. Fit your value with arrange function using future scalling.
- 5. Check your model accuracy and precission using confuser matrix.

```
[]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
[]: data=pd.read_csv("/content/framingham.csv")
data
```

```
[]:
            male
                        education
                                    currentSmoker
                                                      cigsPerDay
                                                                   BPMeds
                                                                            \
                   age
               1
                    39
                               4.0
                                                              0.0
                                                                       0.0
     0
     1
               0
                               2.0
                                                  0
                                                              0.0
                                                                       0.0
                    46
     2
               1
                    48
                               1.0
                                                  1
                                                             20.0
                                                                       0.0
     3
               0
                               3.0
                                                             30.0
                                                                       0.0
                    61
                                                  1
                    46
                               3.0
                                                             23.0
                                                                       0.0
```

```
4233
               1
                    50
                              1.0
                                                           1.0
                                                                    0.0
                                                1
      4234
                              3.0
                                                          43.0
                                                                    0.0
               1
                    51
                                                1
      4235
                              2.0
                                                          20.0
                    48
                                                1
                                                                    NaN
      4236
               0
                    44
                              1.0
                                                1
                                                          15.0
                                                                    0.0
      4237
               0
                    52
                              2.0
                                                0
                                                           0.0
                                                                    0.0
                                             diabetes
                                                        totChol
                                                                 sysBP
                                                                                   BMI
            prevalentStroke
                              prevalentHyp
                                                                         diaBP
                                                                                       \
      0
                                                                 106.0
                                                                                26.97
                                          0
                                                     0
                                                          195.0
                                                                          70.0
      1
                           0
                                          0
                                                     0
                                                          250.0
                                                                 121.0
                                                                          81.0
                                                                                28.73
      2
                           0
                                          0
                                                     0
                                                                 127.5
                                                                                25.34
                                                          245.0
                                                                          80.0
                                                                          95.0
      3
                           0
                                          1
                                                     0
                                                          225.0
                                                                 150.0
                                                                                28.58
      4
                           0
                                          0
                                                     0
                                                          285.0 130.0
                                                                          84.0
                                                                                23.10
      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
                           0
                                                     0
                                                          248.0 131.0
                                                                          72.0 22.00
      4236
                           0
                                          0
                                                     0
                                                          210.0
                                                                 126.5
                                                                          87.0 19.16
      4237
                           0
                                          0
                                                     0
                                                          269.0 133.5
                                                                          83.0
                                                                                21.47
            heartRate
                        glucose
                                 TenYearCHD
      0
                  80.0
                           77.0
                                           0
      1
                  95.0
                           76.0
                                           0
      2
                  75.0
                           70.0
                                           0
      3
                  65.0
                          103.0
                                           1
      4
                  85.0
                           85.0
                                           0
      4233
                  66.0
                           86.0
                                           1
      4234
                  65.0
                           68.0
                                           0
      4235
                  84.0
                                           0
                           86.0
      4236
                  86.0
                            NaN
                                           0
      4237
                  80.0
                                           0
                          107.0
      [4238 rows x 16 columns]
[35]: x=data[["age"]]
      y=data[["currentSmoker"]]
      from sklearn.model_selection import train_test_split
      x_train,x_test,y_train,y_test=train_test_split (x,y,test_size=0.
       →2,random_state=0)
[36]: print(x_train)
            age
     4203
             63
     3534
             52
     1736
             38
```

```
541
             49
     2526
             40
     1033
             44
     3264
             51
     1653
             39
     2607
             57
     2732
            40
     [3390 rows x 1 columns]
[37]: print(y_train)
            currentSmoker
     4203
     3534
                        1
     1736
                        1
     541
                        0
                        0
     2526
                        0
     1033
     3264
                        1
     1653
                        1
     2607
                        0
     2732
                        1
     [3390 rows x 1 columns]
[38]: print(x_test)
           age
     1669
            47
     156
            58
     87
             61
             45
     685
     666
             57
     245
             46
     4215
             63
     548
             50
     4225
             45
     2996
            50
      [848 rows x 1 columns]
[39]: print(y_test)
```

currentSmoker

```
156
                       0
     87
                       1
     685
                       0
     666
                       0
     245
                       1
     4215
     548
                       1
     4225
                       1
     2996
                       0
     [848 rows x 1 columns]
[40]: from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      x_train = sc.fit_transform(x_train)
      x_test = sc.transform(x_test)
 []: print(x_train)
 []: print(x_test)
[43]: 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)
[43]: LogisticRegression(random_state=0)
[46]: y_pred=classifier.predict(x_test)
[47]: from sklearn.metrics import confusion matrix, accuracy_score
      cm = confusion_matrix(y_test, y_pred)
      print(cm)
      accuracy_score(y_test, y_pred)
     [[268 185]
      [155 240]]
[47]: 0.5990566037735849
     ###Conclusion
```

1669

0

according to the model analysis the logistic regression algorithm work suuccesfully with 0.6 accuracy. the accuracy shows that building the model is succesful.

[]: