heart-attack-disease

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5 project title: predict the heart attack disease for organisation WHO. using machine learning algorithm rate of heart attack disease will increasing manner or decreasing manner

6 Problem Statement:

world heard organization estimated 12 millions Dead records. One of them half of the dead result is found in US. The research scholer 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

7 TASK:

1.import the libraries which is the required for prediction

2.import the dataset your using workspace.

3. use the appropriate argument of sklearn library to train, test and split the dataset

4.fit your values with arange function using future scaling

5.check your model accuracy and precision using confuser matrix

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

```
[3]: dataset=pd.read_csv("framingham.csv")
dataset
```

```
[3]:
            male
                        education currentSmoker
                                                    cigsPerDay
                                                                 BPMeds
                   age
                               4.0
                                                            0.0
                                                                     0.0
      0
                1
                    39
      1
                               2.0
                                                 0
                                                            0.0
                0
                    46
                                                                     0.0
      2
                1
                    48
                               1.0
                                                 1
                                                           20.0
                                                                     0.0
      3
                0
                    61
                               3.0
                                                 1
                                                           30.0
                                                                     0.0
      4
                0
                    46
                               3.0
                                                 1
                                                           23.0
                                                                     0.0
                                                 •••
                                                                     0.0
      4233
                1
                    50
                               1.0
                                                 1
                                                            1.0
      4234
                    51
                               3.0
                                                           43.0
                                                                     0.0
                1
                                                 1
      4235
                    48
                0
                               2.0
                                                 1
                                                           20.0
                                                                     NaN
      4236
                0
                    44
                               1.0
                                                 1
                                                           15.0
                                                                     0.0
      4237
                0
                    52
                               2.0
                                                 0
                                                            0.0
                                                                     0.0
            prevalentStroke
                               prevalentHyp
                                              diabetes
                                                         totChol
                                                                  sysBP
                                                                          diaBP
                                                                                    BMI \
      0
                                                           195.0
                                                                  106.0
                                                                           70.0
                                                                                  26.97
                            0
                                                     0
                            0
                                           0
                                                                                  28.73
      1
                                                     0
                                                           250.0 121.0
                                                                           81.0
      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
                                                                  130.0
                                                                           84.0
                                                                                  23.10
      4233
                                                           313.0
                                                                  179.0
                                                                           92.0
                                                                                 25.97
                            0
                                           1
                                                     0
      4234
                            0
                                           0
                                                     0
                                                           207.0 126.5
                                                                           80.0 19.71
                                                                           72.0 22.00
      4235
                            0
                                           0
                                                     0
                                                           248.0 131.0
      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
                  80.0
      0
                            77.0
                                            0
                  95.0
                                            0
      1
                            76.0
      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
                                            0
      4235
                  84.0
                            86.0
      4236
                                            0
                  86.0
                             NaN
      4237
                  80.0
                           107.0
                                            0
      [4238 rows x 16 columns]
[19]: x=dataset[["age"]]
      y=dataset[["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.
        →4,random_state=0)
```

```
[20]: print(x_train)
            age
     3218
             42
     590
             60
     3880
            41
     1548
             59
     2601
             55
     1033
            44
     3264
            51
     1653
            39
     2607
            57
     2732
            40
     [2542 rows x 1 columns]
[21]: print(y_train)
            currentSmoker
     3218
                        1
     590
                        1
     3880
                        0
     1548
                        0
     2601
                        1
     1033
                        0
     3264
                        1
     1653
                        1
                        0
     2607
     2732
     [2542 rows x 1 columns]
[22]: print(x_test)
           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]
[23]: print(y_test)
           currentSmoker
     1669
     156
                        0
     87
                        1
     685
                        0
     666
                        0
     2790
                        0
     1855
                        0
                        0
     700
     2060
                        0
     2348
                        1
     [1696 rows x 1 columns]
[24]: from sklearn.preprocessing import StandardScaler
      sc =StandardScaler()
      x_train = sc.fit_transform(x_train)
      x_test = sc.transform(x_test)
[25]: print(x_train)
     [[-0.89361628]
      [ 1.21446304]
      [-1.0107318]
      [-1.24496283]
      [ 0.86311649]
      [-1.12784731]]
[26]: print(x_test)
     [[-0.30803869]
      [ 0.980232 ]
      [ 1.33157856]
      [ 1.21446304]
      [-1.36207835]
      [-0.19092317]]
[27]: 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)

[27]: LogisticRegression(random_state=0)

[28]: y_pred = classifier.predict(x_test)

[30]: y_pred

[30]: array([1, 0, 0, ..., 0, 1, 1])

[31]: from sklearn.metrics import confusion_matrix, accuracy_score
        cm = confusion_matrix(y_test, y_pred)
        print(cm)
        accuracy_score(y_test, y_pred)

[[503 371]
        [303 519]]
```

[31]: 0.6025943396226415

###conclusion: According to the model analysis the logistic regression algorithm work successfully with 0.6 accuracy shows that building the model is successful