untitled0-1

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NRCM

1 PROJECRT TITLE: Predict the heart attack disase for organsization who are using meachine learning algorithm rate of heart attack disease will increasing manner or decreasing manner

##PROBLEM STATEMENT

A world heard organization estmated 12 million Dead records. One of them half of the dead result is found in US. The rescearch schlor point out the most revelent risk factor of heatattack as a data science engineer predict the overall risk using meachine learning algorithm is called logistic regression

$\#\#\mathrm{TASK}$

1.import the libraries which is the regired for preduction

2.import the data set using the work space

3. use a approprie argument or sklearn libraries to train, test, and split the data set.

4.fit your value with arrange function using future scalling.

5.check your model accuracyand precussion using confuser matrix

##CONCLUSION

According to the model analysis thre logsticregression algorithm work successfully with 0.6 accuracy done, The accuracy shows that building the model is successful.

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

```
[4]: data=pd.read_csv("/content/framingham (3).csv")
data
```

```
[4]:
            male
                        education currentSmoker
                                                   cigsPerDay
                                                                BPMeds
                   age
                              4.0
                                                           0.0
                                                                   0.0
      0
               1
                    39
                              2.0
                                                0
                                                           0.0
      1
               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
                                                          269.0
      4237
                           0
                                          0
                                                     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]
[13]: 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.
```

```
[14]: 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]
[15]: print(y_train)
            currentSmoker
     3218
                        1
     590
                        1
     3880
                        0
     1548
                        0
     2601
                        1
     1033
                        0
     3264
                        1
     1653
                        1
     2607
                        0
     2732
     [2542 rows x 1 columns]
[16]: 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]
[17]: 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]
[18]: from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      x_train = sc.fit_transform(x_train)
      x_test = sc.transform(x_test)
[19]: print(x_train)
     [[-0.89361628]
      [ 1.21446304]
      [-1.0107318]
      [-1.24496283]
      [ 0.86311649]
      [-1.12784731]]
[20]: print(x_test)
     [[-0.30803869]
      [ 0.980232 ]
      [ 1.33157856]
      [ 1.21446304]
      [-1.36207835]
      [-0.19092317]]
[22]: 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)

[22]: LogisticRegression(random_state=0)

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

[25]: 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]]

[25]: 0.6025943396226415

[]:
```