heart-attack

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#Project Title: ##To Predict the Heart attack disease for organization WHO (World Health Organization), using Machine Learning algorithm rate of heart attack disease will increasing manner or decresing manner

#Project Statement: ##A world health organization eastimated 12 millons death data. One of them half of the death result is found in US. ##The research scholars point out the most relevent risk factor of heart attack. As a Data Science Engineer predict the overall risk using Machine Learning algorithm which ever them called as Logistic Regression.

#Task: ##1.Import the libraries which is required for prediction ##2.import the dataset your using workspace ##3.Use the apporipriate argument of sklearn library to train ,test and split the dataset ##4.Fit your values with arange function using FeatureScaling ##5.Check your model accuracy and precision using confusion matrix

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

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

[40]:		m	ale	age	education	currentSmoker	cigsPerDay	BPMeds	\
	0		1	39	4.0	0	0.0	0.0	
	1		0	46	2.0	0	0.0	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	
	•••		•••		***		***		
	4233		1	50	1.0	1	1.0	0.0	
	4234		1	51	3.0	1	43.0	0.0	
	4235		0	48	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	

```
BMI \
            prevalentStroke
                            prevalentHyp
                                           diabetes
                                                     totChol
                                                               sysBP
                                                                      diaBP
      0
                                                   0
                                                        195.0
                                                               106.0
                                                                       70.0
                                                                             26.97
      1
                          0
                                        0
                                                   0
                                                              121.0
                                                                       81.0
                                                                             28.73
                                                        250.0
      2
                          0
                                        0
                                                        245.0 127.5
                                                                             25.34
                                                   0
                                                                       80.0
      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
                          0
                                                   0
                                                        313.0
                                                              179.0
                                                                       92.0
                                                                             25.97
                                         1
      4234
                          0
                                                        207.0 126.5
                                                                       80.0 19.71
                                        0
                                                   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
                                TenYearCHD
            heartRate glucose
      0
                 80.0
                          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
      4235
                 84.0
                          86.0
                                         0
      4236
                 86.0
                                         0
                           NaN
      4237
                 80.0
                         107.0
                                          0
      [4238 rows x 16 columns]
[38]:
     data.shape
[38]: (4238, 16)
[39]: X=data[["age"]]
      y=data[["currentSmoker"]]
      X_test,X_train,y_test,y_train = train_test_split(X,y,test_size=0.
       [25]: print(X_train)
           age
     1669
            47
     156
            58
     87
            61
            45
     685
     666
            57
     2790
            53
```

```
700
            60
     2060
            38
     2348
            48
     [1696 rows x 1 columns]
[26]: print(y_train)
           currentSmoker
     1669
     156
                        0
     87
                        1
     685
     666
                        0
     2790
                        0
     1855
                        0
     700
                        0
     2060
                        0
     2348
     [1696 rows x 1 columns]
[22]: print(X_test)
     [[42]
      [60]
      [41]
      [39]
      [57]
      [40]]
[27]: print(y_test)
           currentSmoker
     3218
                        1
     590
                        1
     3880
                        0
                        0
     1548
     2601
                        1
     1033
                        0
     3264
                        1
     1653
                        1
     2607
                        0
     2732
                        1
```

```
[2542 rows x 1 columns]
```

```
[28]: from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      X_train = sc.fit_transform(X_train)
      X_test = sc.transform(X_test)
[29]: print(X_train)
     [[-0.2920309]
      [ 0.9841763 ]
      [ 1.33223281]
      [ 1.21621397]
      [-1.33620043]
      [-0.17601207]]
[30]: print(X_test)
     [[-0.87212509]
      [ 1.21621397]
      [-0.98814392]
      [-1.2201816]
      [ 0.86815746]
      [-1.10416276]]
[31]: 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)
[31]: LogisticRegression(random_state=0)
[32]: y_pred = classifier.predict(X_test)
[33]: 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]]

[33]: 0.6018882769472856

###Conclusion: According to the model analysis the Logistic Regression algorithm
works successfully with 0.60 Accuracy. ###The accuracy shows that building the model is successful.