framing

August 25, 2023

[]:

####NAME:CH.Mahesh## ####21X05A6712## ####CSE-DS

##Project Title:- Predict the heart attack disease for organisation WHO(World Health Organization) . using machine learning algorithm rate of heart attack disease in increasing or decreasing manner.

###Problem Statement A WHO estimated 12 million death records. One of them half off the death result is found in US. The research intenders the researchers scholar pointout the most relevant risk factor of heartattack As a datascience engineer predict the overall risk using machine learning alogorithm which is called as logistic regression

###Task 1 ####>import the libraries ##Task2 ####>import the data set to your used workspace ###Task 3 ####>Use a appropriate argument of sklearn library to train, test and split the dataset ### Task 4 ##### hit your values with the arrange using feature scaling ###Task 5 ##### Check your model accuracy and precision using confusion matrix

```
[17]: #import the libraries
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
```

```
[18]: #Read the data data=pd.read_csv("framingham.csv") data
```

[18]:	n	nale	age	education	currentSmoker	cigsPerDay	BPMeds	\
0		1	39	4	0	0.0	0	
1		0	46	2	0	0.0	0	
2		1	48	1	1	20.0	0	
3		0	61	3	1	30.0	0	
4		0	46	3	1	23.0	0	
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42	33	1	50	1	1	1.0	0	

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            heartRate glucose TenYearCHD
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                            107
      [4238 rows x 16 columns]
[34]: from sklearn.model_selection import train_test_split
      X=data[["age"]]
      y=data["currentSmoker"]
      X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.
       →2,random_state=42)
[35]: print(X_test)
            age
             63
     3188
     764
             45
     3264
            51
     1967
             45
```

```
3303
             47
     4056
             44
     4210
             50
     3971
             64
     2540
             55
      [848 rows x 1 columns]
[36]: print(X_train)
            age
     3252
             40
     3946
             57
     1261
             47
     2536
             41
     4089
             64
     3444
             36
     466
             57
     3092
             60
     3772
             39
     860
             35
     [3390 rows x 1 columns]
[37]: print(y_test)
     3188
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     Name: currentSmoker, Length: 848, dtype: int64
[38]: print(y_train)
     3252
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3444
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     Name: currentSmoker, Length: 3390, dtype: int64
[39]: from sklearn.preprocessing import StandardScaler
      sc =StandardScaler()
      X_train = sc.fit_transform(X_train)
      X_test = sc.transform(X_test)
[40]: print(X_train)
      [[-1.11033368]
      [ 0.87196279]
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      [ 1.22177981]
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      [-1.69336206]]
     ####Conclusion: According to the modern analysis the LougisticRegression algorithm works
     successfully '0.6" accuracy. ####The accuracy shows that building the model is successful
[41]: print(X_test)
      [[ 1.57159684]
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[42]: from sklearn.linear_model import LogisticRegression
      classifier = LogisticRegression(random_state = 0)
      classifier.fit(X_train, y_train)
[42]: LogisticRegression(random_state=0)
[44]: y_pred = classifier.predict(X_test)
[45]: from sklearn.metrics import confusion_matrix, accuracy_score
      cm = confusion_matrix(y_test, y_pred)
      print(cm)
      accuracy_score(y_test, y_pred)
     [[263 183]
      [155 247]]
[45]: 0.6014150943396226
 []:
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