heartattack-data

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#Project Title To predict the heart attack disease for organization WHO, using machine learning algorithm rate of heart attack disease will increasing manner 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

#Conculsion According to the model analysis the logistic regression algorithm works successfully with 0.6 accuracy.

The accuracy shows that building the model is successfull.

```
[]: # import the libery's
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

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

[]:	male	age	education	currentSmoker	cigsPerDay	BPMeds	\
0	1	39	4	0	0	0	
1	0	46	2	0	0	0	
2	1	48	1	1	20	0	
3	0	61	3	1	30	0	
4	0	46	3	1	23	0	
			•••		•••		
4233	3 1	50	1	1	1	0	
4234	1	51	3	1	43	0	

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     4237
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                   52
                                                0
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           prevalentStroke
                             prevalentHyp
                                            diabetes
                                                       totChol
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                                                                         diaBP
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     4237
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     [4238 rows x 16 columns]
[]: 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)
[]: 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]
[]: print(y_train)
    3252
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    860
    Name: currentSmoker, Length: 3390, dtype: int64
[]: print(X_test)
           age
    3188
            63
    764
            45
    3264
            51
    1967
            45
    2185
            45
    3303
            47
    4056
            44
    4210
            50
    3971
            64
    2540
            55
    [848 rows x 1 columns]
[]: print(y_test)
    3188
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3303
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    Name: currentSmoker, Length: 848, dtype: int64
[]: from sklearn.preprocessing import StandardScaler
     sc =StandardScaler()
     X_train = sc.fit_transform(X_train)
     X_test = sc.transform(X_test)
[]: print(X_train)
    [[-1.11033368]
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[]: print(X_test)
    [[ 1.57159684]
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[]: from sklearn.linear_model import LogisticRegression
     classifier = LogisticRegression(random_state = 0)
     classifier.fit(X_train, y_train)
[]: LogisticRegression(random_state=0)
[]: y_pred = classifier.predict(X_test)
[]: 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]]
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[]:
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