heart-attack-disease-2

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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

##Porblem 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

##Task1: Import the libery which is required for prediction ##Task2: Import the dataset your using workspace ##Task3: Use a appropriate argument of sklearn libery to train, test and split the dataset ##Task4: Fit your values with a arrange function using future scaling ##Task5: Check your model accuracy and precision using confusion matrix

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

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

[16]:		m	ale	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		1	50	1	1	1	0	
	4234		1	51	3	1	43	0	
	4235		0	48	2	1	20	0	
	4236		0	44	1	1	15	0	
	4237		0	52	2	0	0	0	

prevalentStroke prevalentHyp diabetes totChol sysBP diaBP BMI \

```
0
                    0
                                  0
                                            0
                                                   195 106.0
                                                                70.0 26.97
1
                    0
                                  0
                                            0
                                                   250 121.0
                                                                81.0
                                                                      28.73
2
                                  0
                                                                      25.34
                    0
                                            0
                                                       127.5
                                                                80.0
                                                   245
3
                    0
                                  1
                                                       150.0
                                                                95.0 28.58
                                            0
                                                   225
4
                    0
                                  0
                                            0
                                                   285 130.0
                                                                84.0
                                                                      23.10
4233
                    0
                                            0
                                                   313 179.0
                                                                92.0 25.97
                                  1
                                                                80.0 19.71
4234
                    0
                                  0
                                            0
                                                   207 126.5
4235
                    0
                                  0
                                            0
                                                   248 131.0
                                                                72.0 22.00
4236
                    0
                                  0
                                            0
                                                   210 126.5
                                                                87.0 19.16
4237
                    0
                                  0
                                            0
                                                   269 133.5
                                                                83.0 21.47
```

	heartRate	glucose	${\tt TenYearCHD}$
0	80	77	0
1	95	76	0
2	75	70	0
3	65	103	1
4	85	85	0
•••	•••	•••	•••
4233	66	86	1
4234	65	68	0
4235	84	86	0
4236	86	0	0
4237	80	107	0

[4238 rows x 16 columns]

```
[37]: 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)
```

[38]: 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]

```
[39]: print(y_train)
     3218
              1
     590
              1
     3880
              0
     1548
              0
     2601
              1
     1033
              0
     3264
              1
     1653
              1
     2607
              0
     2732
              1
     Name: currentSmoker, Length: 2542, dtype: int64
[40]: print(X_test)
            age
     1669
             47
     156
             58
     87
             61
     685
             45
             57
     666
     2790
             53
     1855
             66
     700
             60
     2060
             38
     2348
             48
      [1696 rows x 1 columns]
[41]: print(y_test)
     1669
              0
     156
              0
     87
              1
     685
              0
     666
              0
             . .
     2790
              0
     1855
              0
     700
              0
     2060
              0
     2348
     Name: currentSmoker, Length: 1696, dtype: int64
```

```
[42]: from sklearn.preprocessing import StandardScaler
      sc =StandardScaler()
      X_train = sc.fit_transform(X_train)
      X_test = sc.transform(X_test)
[43]: print(X_train)
     [[-0.89361628]
      [ 1.21446304]
      [-1.0107318]
      [-1.24496283]
      [ 0.86311649]
      [-1.12784731]]
[44]: print(X_test)
     [[-0.30803869]
      [ 0.980232 ]
      [ 1.33157856]
      [ 1.21446304]
      [-1.36207835]
      [-0.19092317]]
[45]: from sklearn.linear_model import LogisticRegression
      classifier = LogisticRegression(random_state = 0)
      classifier.fit(X_train, y_train)
[45]: LogisticRegression(random_state=0)
[47]: y_pred = classifier.predict(X_test)
[48]: 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]]
[48]: 0.6025943396226415
```

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

The accuracy shows that building the model is successfull.