#NAME: K ABHIRAM

#PIN NO: 21X05A6724 COLLAGE: NARASIMHA REDDY ENGINEERING COLLAGE

PROJECT TITLE:To predict the heartattack diseases for organization (WHO:world Health Organization) using machine learning algorithm rate of heart attack disease will increasing manner or discreasing manner

Problem statement: A WHO estimated 12 millions death records. One of them half off the death result is found in US. Theresearch scholars point out the relevant risk factor of heart attack as a DataScience engineer predict the overall risk using ML algorithm(Logistic Regression)

Task:

- 1) Import the libraries required for prediction.
- 2) Import the Dataset using workspace.
- 3) Use the appropriate argument of sklearn Library to train, test and split the dataset
- 4) Fit your values with a range function using feature scalling
- 5) Check your model accuracy and precison using confusion matrix

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
data = pd.read csv("framingham.csv")
data
      male
             age
                  education
                               currentSmoker
                                               cigsPerDay
                                                            BPMeds \
                         4.0
                                                       0.0
0
              39
                                                                0.0
         1
                                            0
1
              46
                         2.0
                                            0
                                                       0.0
                                                                0.0
          0
2
                                                      20.0
          1
              48
                         1.0
                                            1
                                                                0.0
3
          0
              61
                         3.0
                                            1
                                                      30.0
                                                                0.0
                                                      23.0
4
                                            1
          0
              46
                         3.0
                                                                0.0
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                                                       1.0
4233
         1
              50
                         1.0
                                            1
                                                                0.0
4234
                         3.0
                                            1
                                                      43.0
                                                                0.0
              51
          1
4235
          0
              48
                         2.0
                                            1
                                                      20.0
                                                                NaN
4236
              44
                                            1
                                                      15.0
          0
                         1.0
                                                                0.0
              52
4237
                         2.0
                                            0
                                                       0.0
                                                                0.0
      prevalentStroke prevalentHyp
                                        diabetes totChol
                                                             sysBP
                                                                     diaBP
BMI
                      0
                                                                      70.0
                                                      195.0
                                                             106.0
26.97
                      0
                                     0
                                                      250.0
                                                             121.0
                                                                      81.0
28.73
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prevalentStroke prevalentHyp diabetes totChol sysBP diaBP

False

False

False

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False

False

False

4236 False False

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BMI \
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      heartRate glucose TenYearCHD
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4235
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4236
          False
                   True
                               False
          False
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4237
                               False
[4238 rows x 16 columns]
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=0.2, random state=0)
print(X train)
      age
4203
      63
3534
       52
```

```
1736
       38
541
       49
2526
       40
. . .
      . . .
1033
      44
       51
3264
1653
       39
2607
       57
2732 40
[3390 rows \times 1 columns]
print(y_train)
4203
       1
3534
        1
1736
        1
541
        0
2526
        0
1033
       0
3264
       1
1653
        1
2607
        0
2732
        1
Name: currentSmoker, Length: 3390, dtype: int64
print(X_test)
      age
1669
       47
156
       58
87
       61
       45
685
       57
666
245
      46
4215
       63
548
       50
4225
       45
2996
       50
[848 rows x 1 columns]
print(y_test)
1669
        0
156
        0
87
        1
685
        0
        0
666
```

```
245
        1
4215
        0
548
        1
4225
        1
2996
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)
[[ 1.57197395]
 [ 0.28779517]
 [-1.34661418]
 [-1.22987065]
 [ 0.8715128 ]
 [-1.11312713]]
print(X_test)
[[-0.29592245]
[ 0.98825632]
 [ 1.3384869 ]
 [-0.5294095]
 [ 0.8715128 ]
 [-0.5294095]
 [0.52128222]
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[-0.87964008]
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- [-1.22987065]
- [-1.58010123]
- [0.98825632]
- [-1.34661418]
- [0.63802575]
- [0.03002373]
- [1.22174337]
- [-0.5294095]
- [-0.5294095]
- [-1.11312713]
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- [-0.76289655]
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- [0.05430812]
- [-0.17917893]
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- [-0.64615303]
- [2.15569157]
- [0.63802575]
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[-0.17917893]

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[1.3384869]

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[-1.11312713]

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[-0.41266598]

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[2.038948051

[0.8715128]

[-0.17917893]

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[0.17105165]

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- [-1.9303318]

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[ 2.15569157]
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- 0.4045387 1
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- [-0.29592245]
- [0.52128222][-1.4633577]

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[-0.87964008]
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 [-1.34661418]
 [-1.11312713]
 [-0.41266598]
 [ 1.57197395]
 [ 0.05430812]
 [-0.5294095]
 [ 0.05430812]]
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)
y pred
array([1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1,
0,
```

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0,
       0, 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0])
from sklearn.metrics import confusion matrix, accuracy score
cm = confusion matrix(y test, y pred)
print(cm)
accuracy score(y test, y pred)
[[268 185]
[155 240]]
0.5990566037735849
```

Conclusion:

According to the model analysis the Logistic Regression Algorithm works successfully with 0.6 accuracy. The accuracy shows that building model is *successfull*