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#Project Title: ###To predict the Heart attach disease for organisation WHO(World Health Organisation), using Machine Learning algorithm rate of heart attack disease will increasing manner or decreasing manner.

##Problem Statement: ###The world health Organisation (WHO) estimated 12 million then records One of them half of the dead result is found in US. The research scholor point out the most relevent risk factor as a Datascience engineering predict the overall risk using Machine Learning algorith called as Logistic Regression.

##Task 1 ###Import the libraries which is required for prediction ##Task 2 ###Import the Dataset using your workspace. ##Task 3 ###Use a appropriate argument of sk learn library to train, test and split the datasets. ##Task 4 ###Fit your values with arrange function using feature scalling. ##Task 5 ###check your model accuracy and precision using confussion matrics.

##Import the Libraries

```
[3]: import numpy as np import pandas as pd import matplotlib.pyplot as plt
```

##Importing the dataset

```
[5]: dataset = pd.read_csv("framingham.csv")
dataset
```

[5]:		ma	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	

```
totChol sysBP
            prevalentStroke
                            prevalentHyp
                                            diabetes
                                                                       diaBP
                                                                                BMI
                                                               106.0
      0
                           0
                                                   0
                                                         195.0
                                                                        70.0
                                                                              26.97
                          0
                                         0
                                                   0
                                                         250.0 121.0
                                                                        81.0
                                                                              28.73
      1
      2
                           0
                                         0
                                                   0
                                                        245.0 127.5
                                                                        80.0
                                                                              25.34
      3
                                                        225.0 150.0
                                                                        95.0
                                                                              28.58
                          0
                                         1
                                                   0
      4
                           0
                                         0
                                                   0
                                                        285.0
                                                               130.0
                                                                        84.0 23.10
                                                                              25.97
      4233
                          0
                                                   0
                                                        313.0
                                                               179.0
                                                                        92.0
                                         1
      4234
                           0
                                         0
                                                   0
                                                        207.0 126.5
                                                                        80.0 19.71
      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
                                                         269.0 133.5
                                                                        83.0 21.47
            heartRate
                       glucose
                                TenYearCHD
                 80.0
                          77.0
      0
                 95.0
                          76.0
                                          0
      1
      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
                                          0
                          86.0
      4236
                 86.0
                           NaN
                                          0
      4237
                 80.0
                                          0
                         107.0
      [4238 rows x 16 columns]
     ##Splitting the dataset into the Training set and Test set
[12]: 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.
       [13]: print(y_train)
           currentSmoker
     3218
                        1
     590
                        1
                        0
     3880
     1548
                        0
     2601
                        1
     1033
                        0
```

3264

1

```
1653
                        1
     2607
                        0
     2732
                        1
     [2542 rows x 1 columns]
[14]: print(X_test)
            age
     1669
            47
     156
            58
     87
             61
             45
     685
     666
            57
     2790
            53
     1855
            66
     700
            60
     2060
            38
     2348
            48
     [1696 rows x 1 columns]
[15]: print(y_test)
            currentSmoker
     1669
                        0
     156
     87
                        1
     685
                        0
     666
                        0
     2790
                        0
     1855
                        0
                        0
     700
     2060
                        0
     2348
     [1696 rows x 1 columns]
     \#\#Feature Scaling
[16]: from sklearn.preprocessing import StandardScaler
      sc = StandardScaler()
      x_train = sc.fit_transform(x_train)
      X_test = sc.transform(X_test)
[17]: print(x_train)
```

```
[[-0.89361628]
      [ 1.21446304]
      [-1.0107318]
      [-1.24496283]
      [ 0.86311649]
      [-1.12784731]]
[18]: print(X_test)
     [[-0.30803869]
      [ 0.980232 ]
      [ 1.33157856]
      [ 1.21446304]
      [-1.36207835]
      [-0.19092317]]
     ##Training the Logistic Regression on the Training set
[19]: 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)
[19]: LogisticRegression(random_state=0)
     ##Predicting the Test set results
[21]: y_pred = classifier.predict(X_test)
     ##Making the Confusion Matrix
[22]: 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]]
[22]: 0.6025943396226415
```

##Conclusion: ### According to the model analysis the LogesticReggresion algorith works successfully with range [0.6] accuracy. ### The accuracy shows that building the model is successfull.

[]:[