who

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PROJECT TITLE: To predict the heart disease for oraganization "who", using machine learning alogorithm rate of heart attack disease will increasing or decreasing manner.

##PROBLEM STATEMENT : A "world health organization" estimated 12 millons death records. One of them half of the death result is found in US. The research scholar point out the most relevant rist factor of heart attack as a data science engineer predict the over all risk using machine learning alogorithm is called as logistic regreession

#TASK-1: Import the libraries which is required for prediction. #TASK-2: Import the dataset using your workspace. #TASK-3: Use a appropriate of sklearn library to train, test and split the datasets. #TASK-4: Fit your values with arrange function using feature scale. #TASK-5: Check your model accuracy and presidential presidential confusing matrix.

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

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

[5]:		${\tt male}$	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	

```
prevalentStroke
                            prevalentHyp
                                            diabetes
                                                      totChol sysBP
                                                                       diaBP
                                                                                 BMI \
     0
                                                   0
                                                         195.0
                                                                106.0
                                                                        70.0
                                                                               26.97
     1
                          0
                                         0
                                                   0
                                                               121.0
                                                                        81.0
                                                                               28.73
                                                         250.0
     2
                          0
                                         0
                                                         245.0 127.5
                                                                               25.34
                                                   0
                                                                        80.0
     3
                          0
                                         1
                                                   0
                                                         225.0 150.0
                                                                        95.0
                                                                               28.58
     4
                          0
                                         0
                                                   0
                                                         285.0
                                                               130.0
                                                                        84.0 23.10
     4233
                          0
                                                   0
                                                         313.0
                                                               179.0
                                                                        92.0
                                                                              25.97
                                         1
     4234
                          0
                                                                        80.0 19.71
                                         0
                                                   0
                                                        207.0 126.5
     4235
                          0
                                         0
                                                   0
                                                        248.0 131.0
                                                                        72.0 22.00
     4236
                          0
                                                        210.0 126.5
                                                                        87.0 19.16
                                         0
                                                   0
                                                         269.0 133.5
     4237
                          0
                                         0
                                                   0
                                                                        83.0
                                                                              21.47
                                TenYearCHD
           heartRate
                      glucose
     0
                80.0
                          77.0
                                          0
                95.0
     1
                          76.0
                                          0
     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
                          86.0
                                          0
     4236
                86.0
                                          0
                           NaN
     4237
                80.0
                         107.0
                                          0
     [4238 rows x 16 columns]
[6]: 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)
[7]: 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]

[8]: print(y_train)

```
{\tt currentSmoker}
3218
590
                   1
3880
                    0
1548
                    0
2601
                    1
                    0
1033
3264
                    1
1653
                    1
2607
                    0
2732
                    1
```

[2542 rows x 1 columns]

[9]: print(x_test)

```
age
1669
       47
156
       58
87
       61
685
       45
666
       57
2790
       53
1855
       66
700
       60
2060
       38
2348
       48
```

[1696 rows x 1 columns]

[10]: print(y_test)

	currentSmoker
1669	0
156	0
87	1
685	0
666	0
•••	•••
2790	0

```
1855
                       0
     700
                        0
     2060
                        0
     2348
                        1
     [1696 rows x 1 columns]
[11]: from sklearn.preprocessing import StandardScaler
      sc =StandardScaler()
      x_train = sc.fit_transform(x_train)
      x_test = sc.transform(x_test)
[12]: print(x_train)
     [[-0.89361628]
      [ 1.21446304]
      [-1.0107318]
      [-1.24496283]
      [ 0.86311649]
      [-1.12784731]]
[13]: print(x_test)
     [[-0.30803869]
      [ 0.980232 ]
      [ 1.33157856]
      [ 1.21446304]
      [-1.36207835]
      [-0.19092317]]
[14]: 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)
[14]: LogisticRegression(random_state=0)
[15]: y_pred = classifier.predict(x_test)
[16]: y_pred
```

```
[16]: array([1, 0, 0, ..., 0, 1, 1])

[17]: 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]]

[17]: 0.6025943396226415

###conclusion: According to the model analysis the logistic regression algorithm work successfully with 0.6 accuracy shows that building the model is successful