

```
In [1]: #Name : Ankita Gulde  
#Roll no : 44  
#section :3A
```

```
In [2]: #Aim : To perform operation on logistic regression algorithm
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```
In [3]: import pandas as pd  
import matplotlib.pyplot as plt  
import numpy as np  
import seaborn as sns  
from sklearn.model_selection import train_test_split  
import warnings  
warnings.filterwarnings('ignore')
```

```
In [4]: import os
```

```
In [5]: os.getcwd()
```

```
Out[5]: 'C:\\Users\\HP'
```

```
In [6]: os.chdir("C:\\Users\\HP\\Desktop")
```

```
In [7]: df=pd.read_csv("framingham.csv")
```

```
In [8]: df.head()
```

```
Out[8]:
```

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHypertension
0	1	39	4.0	0	0.0	0.0	0	0
1	0	46	2.0	0	0.0	0.0	0	0
2	1	48	1.0	1	20.0	0.0	0	0
3	0	61	3.0	1	30.0	0.0	0	1
4	0	46	3.0	1	23.0	0.0	0	0

```
In [9]: df.describe()
```

```
Out[9]:
```

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke
count	4238.000000	4238.000000	4133.000000	4238.000000	4209.000000	4185.000000	4238.0
mean	0.429212	49.584946	1.978950	0.494101	9.003089	0.029630	0.0
std	0.495022	8.572160	1.019791	0.500024	11.920094	0.169584	0.0
min	0.000000	32.000000	1.000000	0.000000	0.000000	0.000000	0.0
25%	0.000000	42.000000	1.000000	0.000000	0.000000	0.000000	0.0
50%	0.000000	49.000000	2.000000	0.000000	0.000000	0.000000	0.0
75%	1.000000	56.000000	3.000000	1.000000	20.000000	0.000000	0.0
max	1.000000	70.000000	4.000000	1.000000	70.000000	1.000000	1.0

```
In [10]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4238 entries, 0 to 4237
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  -
1.  male                  4238 non-null   int64
2.  age                   4238 non-null   int64
3.  education             4133 non-null   float64
4.  currentSmoker         4238 non-null   int64
5.  cigsPerDay            4209 non-null   float64
6.  BPMeds                4185 non-null   float64
7.  prevalentStroke       4238 non-null   int64
8.  prevalentHyp          4238 non-null   int64
9.  diabetes              4238 non-null   int64
10. totChol              4188 non-null   float64
11. sysBP                4238 non-null   float64
12. diaBP                4238 non-null   float64
13. BMI                  4219 non-null   float64
14. heartRate            4237 non-null   float64
15. glucose              3850 non-null   float64
16. TenYearCHD           4238 non-null   int64
dtypes: float64(9), int64(7)
memory usage: 529.9 KB
```

```
In [12]: df.isna().sum()
```

```
male                0
age                 0
education           105
currentSmoker       0
cigsPerDay          29
BPMeds              53
prevalentStroke     0
prevalentHyp        0
diabetes            0
totChol             50
sysBP               0
diaBP               0
BMI                 19
heartRate           1
glucose             388
TenYearCHD          0
dtype: int64
```

```
Out[12]:
```

In [13]:
Out[13]:

df									
	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHypertension	
0	1	39	4.0	0	0.0	0.0	0	0	
1	0	46	2.0	0	0.0	0.0	0	0	
2	1	48	1.0	1	20.0	0.0	0	0	
3	0	61	3.0	1	30.0	0.0	0	1	
4	0	46	3.0	1	23.0	0.0	0	0	
...
4233	1	50	1.0	1	1.0	0.0	0	1	
4234	1	51	3.0	1	43.0	0.0	0	0	
4235	0	48	2.0	1	20.0	NaN	0	0	
4236	0	44	1.0	1	15.0	0.0	0	0	
4237	0	52	2.0	0	0.0	0.0	0	0	

4238 rows × 16 columns

Missing Value Treatment

Since, 'glucose' and 'education' columns had a significant amount of all null values, so we replaced them with the mean of values for their respective columns

```
In [15]: df['glucose'].fillna(value = df['glucose'].mean(), inplace=True)

In [16]: df['education'].fillna(value = df['education'].mean(), inplace=True)

In [17]: df['heartRate'].fillna(value = df['heartRate'].mean(), inplace=True)

In [18]: df['BMI'].fillna(value = df['BMI'].mean(), inplace=True)

In [19]: df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(), inplace=True)
```

```
In [20]: df['totChol'].fillna(value = df['totChol'].mean(),inplace=True)
```

```
In [21]: df['BPMeds'].fillna(value = df['BPMeds'].mean(),inplace=True)
```

```
In [22]: df.isna().sum()
```

```
Out[22]: male                0
age                0
education          0
currentSmoker      0
cigsPerDay         0
BPMeds             0
prevalentStroke    0
prevalentHyp       0
diabetes           0
totChol            0
sysBP              0
diaBP              0
BMI                0
heartRate          0
glucose            0
TenYearCHD         0
dtype: int64
```

```
In [23]: #Splitting the dependent and independent variables.
x = df.drop("TenYearCHD",axis=1)
y = df['TenYearCHD']
```

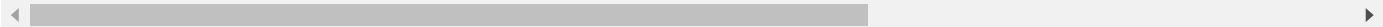
In [24]:

x #checking the features

Out[24]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHypertension
0	1	39	4.0	0	0.0	0.00000	0	0
1	0	46	2.0	0	0.0	0.00000	0	0
2	1	48	1.0	1	20.0	0.00000	0	0
3	0	61	3.0	1	30.0	0.00000	0	1
4	0	46	3.0	1	23.0	0.00000	0	0
...
4233	1	50	1.0	1	1.0	0.00000	0	1
4234	1	51	3.0	1	43.0	0.00000	0	0
4235	0	48	2.0	1	20.0	0.02963	0	0
4236	0	44	1.0	1	15.0	0.00000	0	0
4237	0	52	2.0	0	0.0	0.00000	0	0

4238 rows × 15 columns



Train Test Split

In [25]:

x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_sta

In [26]:

Out[26]:

In [29]:

from sklearn.linear_model import LogisticRegression
model = LogisticRegression().fit(x_train,y_train)
model.score(x_train, y_train)

Explore our de

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3252
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0
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3 0
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0
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1 0
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2
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2 0
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5
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4 0
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0
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3 0
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4
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4 0
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6
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3 0
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0
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3 0
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7
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8 0
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Logistic Regression Algorithm

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0.8495575221238938
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Out[29]:
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In [ ]:
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