LOGISTICAL REGRESSION

```
In [1]:
#Aim: To perform Logistical Regression
#Exp no:10
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#Sec:B
#Roll no:43
#Sub:ET-1
#Date: 11/10/2024
Importing Libraries
In [4]:
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 [5]:
import os
In [6]:
os.getcwd()
Out[6]:
'C:\\Users\\asus'
In [7]:
os.chdir("C:\\Users\\asus\\Desktop")
In [8]:
df=pd.read csv("framingham.csv")
In [9]:
#The "Framingham" heart disease dataset includes over 4,240 records, 15 attributes.
#The goal of the dataset is to predict whether the patient has 10-year risk of future (C
In [10]:
df.head()
```

Out[10]:

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

In [11]:

df.describe()

Out[11]:

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStrok |
|-------|-------------|-------------|-------------|---------------|-------------|-------------|----------------|
| count | 4238.000000 | 4238.000000 | 4133.000000 | 4238.000000 | 4209.000000 | 4185.000000 | 4238.00000 |
| mean | 0.429212 | 49.584946 | 1.978950 | 0.494101 | 9.003089 | 0.029630 | 0.00589 |
| std | 0.495022 | 8.572160 | 1.019791 | 0.500024 | 11.920094 | 0.169584 | 0.07658 |
| min | 0.000000 | 32.000000 | 1.000000 | 0.000000 | 0.000000 | 0.000000 | 0.00000 |
| 25% | 0.000000 | 42.000000 | 1.000000 | 0.000000 | 0.000000 | 0.000000 | 0.00000 |
| 50% | 0.000000 | 49.000000 | 2.000000 | 0.000000 | 0.000000 | 0.000000 | 0.00000 |
| 75% | 1.000000 | 56.000000 | 3.000000 | 1.000000 | 20.000000 | 0.000000 | 0.00000 |
| max | 1.000000 | 70.000000 | 4.000000 | 1.000000 | 70.000000 | 1.000000 | 1.00000 |

In [12]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4238 entries, 0 to 4237
Data columns (total 16 columns):

| # | Column | Non-Null Count | Dtype |
|----|-----------------|----------------|---------|
| | | | |
| 0 | male | 4238 non-null | int64 |
| 1 | age | 4238 non-null | int64 |
| 2 | education | 4133 non-null | float64 |
| 3 | currentSmoker | 4238 non-null | int64 |
| 4 | cigsPerDay | 4209 non-null | float64 |
| 5 | BPMeds | 4185 non-null | float64 |
| 6 | prevalentStroke | 4238 non-null | int64 |
| 7 | prevalentHyp | 4238 non-null | int64 |
| 8 | diabetes | 4238 non-null | int64 |
| 9 | totChol | 4188 non-null | float64 |
| 10 | sysBP | 4238 non-null | float64 |
| 11 | diaBP | 4238 non-null | float64 |
| 12 | BMI | 4219 non-null | float64 |
| 13 | heartRate | 4237 non-null | float64 |
| 14 | glucose | 3850 non-null | float64 |
| 15 | TenYearCHD | 4238 non-null | int64 |
| | £1+C4/O\ - | -+C1/7\ | |

dtypes: float64(9), int64(7)

memory usage: 529.9 KB

In [13]:

df.isna().sum()

Out[13]: 0 male age 0 105 education currentSmoker 0 29 cigsPerDay 53 **BPMeds** prevalentStroke 0 0 prevalentHyp diabetes 0 50 totChol sysBP 0 diaBP 0 BMI 19 heartRate 1 388 glucose TenYearCHD 0 dtype: int64

In [14]:

#Since, only a few rows have null values in them, we are only removing those rows from t
#df = df.dropna(subset=['heartRate','BMI','cigsPerDay','totChol','BPMeds'])

In [15]:

df

Out[15]:

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStroke | prevalentHyp | diabe |
|------|------|-----|-----------|---------------|------------|--------|-----------------|--------------|-------|
| 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 null values, so we replaced them with the mean of values for their respective columns

```
In [30]:
df['glucose'].fillna(value = df['glucose'].mean(),inplace=True)
In [31]:
df['education'].fillna(value = df['education'].mean(),inplace=True)
In [32]:
df['heartRate'].fillna(value = df['heartRate'].mean(),inplace=True)
In [361:
df['BMI'].fillna(value = df['BMI'].mean(),inplace=True)
In [38]:
df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(),inplace=True)
In [40]:
df['totChol'].fillna(value = df['totChol'].mean(),inplace=True)
In [42]:
df['BPMeds'].fillna(value = df['BPMeds'].mean(),inplace=True)
In [44]:
df.isna().sum()
Out[44]:
                   0
male
                   0
age
                   0
education
currentSmoker
                   0
                   0
cigsPerDay
BPMeds
                   0
                   0
prevalentStroke
prevalentHyp
                   0
diabetes
                   0
                   0
totChol
sysBP
                   0
diaBP
                   0
BMT
                   0
                   0
heartRate
                   0
glucose
TenYearCHD
                   0
dtype: int64
In [46]:
#Splitting the dependent and independent variables.
x = df.drop("TenYearCHD",axis=1)
y = df['TenYearCHD']
In [48]:
x #checking the features
Out[48]:
```

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStroke | prevalentHyp | diabe |
|------|------|-----|-----------|---------------|------------|---------|-----------------|--------------|-------|
| 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 [51]:
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=42)
In [53]:
y train
Out[53]:
3252
        0
3946
1261
        0
2536
        0
4089
        0
3444
        0
466
       0
3092
        0
3772
        0
Name: TenYearCHD, Length: 3390, dtype: int64
```

Logistic Regression Algorithm

```
In [56]:
    from sklearn.linear_model import LogisticRegression
    model = LogisticRegression().fit(x_train,y_train)
    model.score(x_train, y_train)
Out[56]:
```

0.848377581120944

In []: