```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
```

In [2]: from sklearn.linear_model import LogisticRegression

In [17]: df=pd.read_csv(r"E:\154\C4_framingham - C4_framingham.csv").dropna()
df

Out[17]:

		male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp
	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
42	31	1	58	3.0	0	0.0	0.0	0	1
42	32	1	68	1.0	0	0.0	0.0	0	1
42	33	1	50	1.0	1	1.0	0.0	0	1
42	34	1	51	3.0	1	43.0	0.0	0	0
42	37	0	52	2.0	0	0.0	0.0	0	0

3656 rows × 16 columns

In [18]: df.head()

Out[18]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	di
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	
4									•

```
In [19]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 3656 entries, 0 to 4237
Data columns (total 16 columns):

#	Column	Non-Null Count	Dtype			
0	male	3656 non-null	int64			
1	age	3656 non-null	int64			
2	education	3656 non-null	float64			
3	currentSmoker	3656 non-null	int64			
4	cigsPerDay	3656 non-null	float64			
5	BPMeds	3656 non-null	float64			
6	prevalentStroke	3656 non-null	int64			
7	prevalentHyp	3656 non-null	int64			
8	diabetes	3656 non-null	int64			
9	totChol	3656 non-null	float64			
10	sysBP	3656 non-null	float64			
11	diaBP	3656 non-null	float64			
12	BMI	3656 non-null	float64			
13	heartRate	3656 non-null	float64			
14	glucose	3656 non-null	float64			
15	TenYearCHD	3656 non-null	int64			
dt_{ypos} , $f_{1oo}+64(0)$ $in+64(7)$						

dtypes: float64(9), int64(7)
memory usage: 485.6 KB

In [20]: df.describe()

Out[20]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevale
count	3656.000000	3656.000000	3656.000000	3656.000000	3656.000000	3656.000000	3656
mean	0.443654	49.557440	1.979759	0.489059	9.022155	0.030361	(
std	0.496883	8.561133	1.022657	0.499949	11.918869	0.171602	(
min	0.000000	32.000000	1.000000	0.000000	0.000000	0.000000	(
25%	0.000000	42.000000	1.000000	0.000000	0.000000	0.000000	(
50%	0.000000	49.000000	2.000000	0.000000	0.000000	0.000000	(
75%	1.000000	56.000000	3.000000	1.000000	20.000000	0.000000	(
max	1.000000	70.000000	4.000000	1.000000	70.000000	1.000000	,

```
In [21]: df.columns
```

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In [22]: feature_matrix = df.iloc[:,0:15]
         target_vector = df.iloc[:,-1]
In [23]: fs=StandardScaler().fit_transform(feature_matrix)
         logr=LogisticRegression()
         logr.fit(fs,target_vector)
Out[23]: LogisticRegression()
In [24]: observation=[[1,2,3,4,5,6,7,8,9,10,11,12,13,14,15]]
In [25]: prediction=logr.predict(observation)
         print(prediction)
         [1]
In [26]:
         logr.classes
Out[26]: array([0, 1], dtype=int64)
In [27]: logr.predict_proba(observation)[0][0]
Out[27]: 0.0002214783507201723
In [28]: logr.predict_proba(observation)[0][1]
Out[28]: 0.9997785216492798
 In [ ]:
```