```
In [1]: import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
 In [2]: from sklearn.linear_model import LogisticRegression
 In [3]: df=pd.read_csv("C5 health.csv").dropna()
          df
            0
                              148
                                                                  0 33.6
                                                                                          0.627
            1
                               85
                                             66
                                                          29
                                                                  0 26.6
                                                                                          0.351
                                                                                                 31
                                                                                                           0
                        8
                              183
                                             64
                                                           0
                                                                  0 23.3
                                                                                          0.672
                                                                                                 32
            3
                               89
                                             66
                                                          23
                                                                 94 28.1
                                                                                          0.167
                                                                                                 21
                                                                                                           0
            4
                        0
                              137
                                             40
                                                          35
                                                                168 43.1
                                                                                          2.288
                                                                                                 33
          763
                       10
                              101
                                             76
                                                          48
                                                                180 32.9
                                                                                          0.171
                                                                                                 63
                                                                                                           0
          764
                        2
                              122
                                             70
                                                          27
                                                                 0 36.8
                                                                                          0.340
                                                                                                 27
                                                                                                           0
          765
                              121
                                             72
                                                          23
                                                                112 26.2
                                                                                          0.245
                                                                                                 30
                                                                                                           0
          766
                              126
                                             60
                                                           0
                                                                  0 30.1
                                                                                          0.349
                                                                                                 47
          767
                               93
                                             70
                                                          31
                                                                  0 30.4
                                                                                          0.315
                                                                                                 23
                                                                                                           0
          768 rows × 9 columns
 In [4]: | df.dropna(inplace=True)
 In [5]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          Int64Index: 768 entries, 0 to 767
          Data columns (total 9 columns):
           #
               Column
                                           Non-Null Count Dtype
           0
               Pregnancies
                                           768 non-null
                                                            int64
           1
               Glucose
                                           768 non-null
                                                           int64
               {\tt BloodPressure}
                                           768 non-null
                                                            int64
           3
               SkinThickness
                                           768 non-null
                                                            int64
                                           768 non-null
           4
               Insulin
                                                            int64
               BMI
                                           768 non-null
                                                            float64
               DiabetesPedigreeFunction
                                          768 non-null
                                                            float64
           6
                                           768 non-null
               Age
                                                           int64
               Outcome
                                           768 non-null
                                                            int64
          dtypes: float64(2), int64(7)
          memory usage: 60.0 KB
 In [7]: | feature_matrix = df[['Pregnancies','Glucose','BloodPressure','SkinThickness','Insulin','BMI','DiabetesPedigreeFunction','Age
          target vector = df['Outcome']
 In [8]: feature_matrix.shape
 Out[8]: (768, 8)
In [11]: target_vector.shape
Out[11]: (768,)
In [12]: from sklearn.preprocessing import StandardScaler
In [13]: | fs = StandardScaler().fit_transform(feature_matrix)
In [14]: logr = LogisticRegression()
          logr.fit(fs,target_vector)
Out[14]: LogisticRegression()
In [15]: feature_matrix.shape
Out[15]: (768, 8)
```

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In [16]: target_vector.shape
Out[16]: (768,)
In [17]: from sklearn.preprocessing import StandardScaler
In [18]: fs = StandardScaler().fit_transform(feature_matrix)
In [19]: logr = LogisticRegression()
   logr.fit(fs,target_vector)
Out[19]: LogisticRegression()
In [20]: | observation=df[['Pregnancies','Glucose','BloodPressure','SkinThickness','Insulin','BMI','DiabetesPedigreeFunction','Age']]
In [21]: prediction = logr.predict(observation)
   prediction
In [22]: logr.classes_
Out[22]: array([0, 1], dtype=int64)
In [23]: logr.predict_proba(observation)[0][1]
Out[23]: 1.0
In [ ]:
```