```
In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sea
 In [2]: | from sklearn.linear_model import LogisticRegression
 In [3]: | df = pd.read_csv(r"C:\Users\user\Downloads\C5_health care diabetes.csv")
          df
 Out[3]:
                Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunctio
                                                                       33.6
             0
                         6
                                148
                                              72
                                                            35
                                                                                              0.62
             1
                                85
                                              66
                                                            29
                                                                      26.6
                                                                                              0.35
             2
                         8
                                183
                                              64
                                                             0
                                                                      23.3
                                                                                              0.67
             3
                         1
                                89
                                                            23
                                                                   94
                                                                      28.1
                                                                                              0.16
                                              66
             4
                                137
                                              40
                                                            35
                                                                   168 43.1
                                                                                              2.28
                                                                    ...
           763
                        10
                                101
                                              76
                                                            48
                                                                   180
                                                                       32.9
                                                                                              0.17
           764
                         2
                               122
                                              70
                                                            27
                                                                    0 36.8
                                                                                              0.34
                         5
           765
                                121
                                              72
                                                            23
                                                                   112 26.2
                                                                                              0.24
           766
                         1
                                126
                                              60
                                                             0
                                                                       30.1
                                                                                              0.34
           767
                                93
                                              70
                                                            31
                                                                    0 30.4
                                                                                              0.31
          768 rows × 9 columns
 In [4]: | df.columns
 Out[4]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
                  'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
                 dtype='object')
 In [8]: | feature_matrix = df.iloc[:,0:8]
          target_vector = df.iloc[:,-1]
 In [9]: | feature_matrix.shape
 Out[9]: (768, 8)
          from sklearn.preprocessing import StandardScaler
In [10]:
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In [11]: fs = StandardScaler().fit_transform(feature_matrix)

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In [12]: logs = LogisticRegression()
logs.fit(fs,target_vector)
Out[12]: LogisticRegression()
In [13]: observation = [[1.4,1.5,1.6,2,2.1,1,1,3]]
    prediction = logs.predict(observation)
In [14]: print(prediction)
    [1]
In [15]: logs.classes_
Out[15]: array([0, 1], dtype=int64)
In [16]: logs.predict_proba(observation)[0][0]
Out[16]: 0.09599090836764246
In [ ]:
```