```
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("bmi.csv").dropna()
              Gender meight weight index
           0
                Male
                        174
                                96
                                       4
           1
                Male
                        189
                                87
                                       2
           2 Female
                        185
                               110
                                       4
           3 Female
                        195
                               104
                                       3
           4
                Male
                        149
                                61
                                       3
                         ...
         495 Female
                        150
                               153
                                       5
         496 Female
                        184
                               121
                                       4
         497 Female
                        141
                               136
                                       5
         498
                Male
                        150
                                       5
                                95
         499
                Male
                        173
                               131
                                       5
        500 rows × 4 columns
In [4]: | df.dropna(inplace=True)
In [5]: | df.info()
         <class 'pandas.core.frame.DataFrame'>
        Int64Index: 500 entries, 0 to 499
        Data columns (total 4 columns):
          #
              Column Non-Null Count Dtype
          0
              Gender 500 non-null
                                       object
          1
              Height 500 non-null
                                       int64
          2
              Weight 500 non-null
                                       int64
                      500 non-null
              Index
          3
                                       int64
        dtypes: int64(3), object(1)
        memory usage: 19.5+ KB
In [6]: feature_matrix = df[['Height','Weight','Index']]
        target_vector = df['Gender']
```

```
In [7]: | feature_matrix.shape
 Out[7]: (500, 3)
 In [8]: target_vector.shape
 Out[8]: (500,)
 In [9]: from sklearn.preprocessing import StandardScaler
In [10]: fs = StandardScaler().fit_transform(feature_matrix)
In [11]: logr = LogisticRegression()
         logr.fit(fs,target_vector)
Out[11]: LogisticRegression()
In [12]: feature_matrix.shape
Out[12]: (500, 3)
In [13]: target vector.shape
Out[13]: (500,)
In [14]: from sklearn.preprocessing import StandardScaler
In [15]: | fs = StandardScaler().fit_transform(feature_matrix)
In [16]: logr = LogisticRegression()
         logr.fit(fs,target_vector)
Out[16]: LogisticRegression()
In [18]: | observation=df[['Height','Weight','Index']]
```

```
In [19]:
                                                                                                                                                       prediction = logr.predict(observation)
                                                                                                                                                          prediction
                                                                                                                                                                                                                                                                                  'Female', 'Female', 'Female', 'Female', 'Female',
                                                                                                                                                                                                                                                                              'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Femal
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                                                                                                                                                                                                                                                                              'Female', 'Femal
                                                                                                                                                                                                                                                                              'Female', 'Femal
                                                                                                                                                                                                                                                                                'Female', 'Female', 'Female', 'Female', 'Female',
                                                                                                                                                                                                                                                                              'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Female', 'Femal
                                                                                                                                                                                                                                                                              'Female', 'Femal
                                                                                                                                                                                                                                                                              'Female', 'Femal
                                                                                                                                                                                                                                                                                  'Female', 'Female', 'Male', 'Female', 'Female', 'Female',
                                                                                                                                                                                                                                                                                  'Female', 'Female'], dtype=object)
In [20]: logr.classes
Out[20]: array(['Female', 'Male'], dtype=object)
In [21]: logr.predict proba(observation)[0][1]
Out[21]: 0.013560340101419254
```