```
In [1]: import numpy as np import pandas as pd
```

In [2]: df=pd.read_csv(r"C:\Users\user\Downloads\C6_bmi - C6_bmi.csv")
df

Out[2]:

	Gender	Height	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	95	5
499	Male	173	131	5

500 rows × 4 columns

```
In [3]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 4 columns):
    Column Non-Null Count Dtype
                             ____
 0
    Gender 500 non-null
                             object
    Height 500 non-null
                             int64
 1
 2
    Weight 500 non-null
                             int64
    Index
             500 non-null
                             int64
dtypes: int64(3), object(1)
memory usage: 15.8+ KB
```

In [4]: from sklearn.linear_model import LogisticRegression

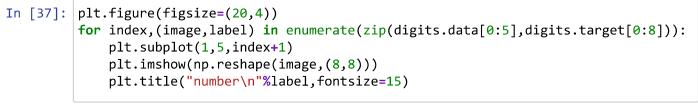
```
In [10]: df1=df[['Height','Weight','Index']]
```

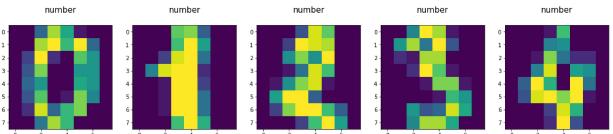
In [42]: feature_matrix = df1.iloc[:,0:3]
target_vector = df1.iloc[:,1]

```
In [43]: feature matrix.shape
Out[43]: (500, 3)
In [44]: target_vector.shape
Out[44]: (500,)
In [27]: | from sklearn.preprocessing import StandardScaler
In [28]: fs=StandardScaler().fit transform(feature matrix)
In [29]: |logr=LogisticRegression()
         logr.fit(fs,target vector)
Out[29]: LogisticRegression()
In [30]: | observation=[[5,7,9]]
         prediction=logr.predict(observation)
In [31]:
         print(prediction)
         [153]
In [32]: logr.classes
Out[32]: array([ 50,
                      51,
                           52,
                                 53,
                                      54,
                                           55,
                                                56,
                                                     57,
                                                          58,
                                                               59,
                                                                    60,
                                                                          61,
                                                                               62,
                                      67,
                                           68,
                                                69,
                                                     70,
                                                               72,
                                                                    73,
                                                                          74,
                                                                               75,
                  63,
                      64,
                           65,
                                 66,
                                                          71,
                  76, 77,
                           78,
                                 79,
                                      80,
                                           81,
                                                82,
                                                     83,
                                                          84,
                                                               85,
                                                                    86,
                                                                          87,
                 89,
                      90,
                           91,
                                 92,
                                      93,
                                          94,
                                               95,
                                                     96,
                                                          97,
                                                               98,
                                                                    99, 100, 101,
                102, 103, 104, 105, 106, 107, 108, 109, 110, 111, 112, 114, 115,
                116, 117, 118, 119, 120, 121, 122, 123, 124, 125, 126, 127, 128,
                129, 130, 131, 132, 133, 134, 135, 136, 137, 138, 139, 140, 141,
                142, 143, 144, 145, 146, 147, 148, 149, 150, 151, 152, 153, 154,
                155, 156, 157, 158, 159, 160], dtype=int64)
In [33]: logr.predict proba(observation)[0][0]
Out[33]: 4.012934825172367e-30
In [34]: logr.predict proba(observation)[0][0]
Out[34]: 4.012934825172367e-30
```

```
In [35]: import re
    from sklearn.datasets import load_digits
    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.model_selection import train_test_split
In [36]: digits = load digits()
```

```
In [36]: digits = load_digits()
          digits
             pixel_3_7'
             'pixel_4_0',
             'pixel 4 1',
             'pixel 4 2',
             'pixel_4_3',
             'pixel_4_4',
             'pixel 4 5',
             'pixel_4_6',
             'pixel 4 7',
             'pixel_5_0',
             'pixel 5 1',
             'pixel_5_2',
             'pixel 5 3',
             'pixel_5_4',
             'pixel_5_5',
             'pixel 5 6',
             'pixel_5_7',
             'pixel_6_0',
             'pixel_6_1',
```





```
x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_siz
In [38]:
         print(x train.shape)
         print(x_test.shape)
         print(y train.shape)
         print(y_test.shape)
         (1257, 64)
         (540, 64)
         (1257,)
         (540,)
In [39]: logre=LogisticRegression()
         logre.fit(x_train,y_train)
         C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:76
         3: ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-
         learn.org/stable/modules/preprocessing.html)
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
         on (https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
         on)
           n iter i = check optimize result(
Out[39]: LogisticRegression()
In [40]: |print(logre.predict(x test))
         [5 0 9 5 4 3 3 1 5 3 0 1 5 2 5 1 7 1 5 2 8 3 5 6 9 9 7 7 2 1 4 7 9 5 7 2 8
          4 6 2 1 2 9 8 3 2 5 4 5 4 0 4 0 8 3 8 2 9 4 9 5 1 8 1 4 8 6 6 2 0 8 4 7 7
          5 1 2 7 0 6 5 7 8 5 3 0 2 7 2 5 1 3 2 2 0 6 5 9 2 1 9 9 9 0 9 8 4 8 4 2 2
          7 5 7 9 5 7 1 5 2 3 7 1 0 1 9 2 9 4 3 5 9 0 7 3 1 3 0 1 2 8 4 6 6 7 7 2 0
          2 9 6 6 2 9 4 5 1 8 2 7 0 6 3 7 5 8 6 8 5 6 0 1 5 8 0 9 6 5 0 3 0 6 8 4 3
          9 5 9 4 9 6 0 0 6 9 1 6 6 8 3 2 3 4 4 0 1 6 5 1 0 9 6 9 1 2 9 2 0 3 4 4 3
          1 7 8 2 7 9 0 0 1 5 2 0 5 6 6 1 4 7 0 1 0 3 6 8 8 5 9 7 0 1 0 4 8 3 2 2 1
          5 0 0 9 6 3 4 0 7 6 6 4 9 5 4 7 0 7 2 6 1 1 1 1 7 8 6 8 6 7 7 7 1 2 9 7 9
          1 3 0 7 4 6 5 9 9 8 5 9 3 4 2 0 2 9 2 4 7 9 5 5 5 8 1 7 7 9 9 4 1 1 4 5 5
          2 6 0 1 4 1 4 7 3 8 1 2 8 0 6 2 0 7 0 0 4 5 3 2 0 1 1 9 1 3 0 3 6 9 7 5 3
          9 5 5 9 9 7 3 0 1 8 2 7 2 1 6 7 5 1 2 6 6 9 1 0 8 8 0 2 1 9 5 5 1 0 8 0 7
          7 2 4 7 9 2 6 2 6 2 7 4 8 1 5 3 9 4 9 2 3 5 7 6 8 0 4 5 9 7 8 7 3 2 4 5 6
          2 6 8 4 5 2 1 4 9 4 4 2 3 6 5 3 3 0 2 9 8 2 1 8 5 0 1 2 7 7 7 1 9 7 8 0 7
          9 0 4 0 3 4 1 7 1 7 5 7 8 5 2 1 1 7 5 8 8 0 0 0 5 0 5 3 9 5 5 2 0 1 3 5 9
          1802944930760427940894
In [41]:
         print(logre.score(x_test,y_test))
```

0.95555555555556