```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

In [2]:

In [3]: dftrain=pd.read\_csv(r"C:\USERS\user\Downloads\C5\_health care diabetes - C5\_hea

Out[3]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.62
1	1	85	66	29	0	26.6	0.38
2	8	183	64	0	0	23.3	0.67
3	1	89	66	23	94	28.1	0.10
4	0	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
765	5	121	72	23	112	26.2	0.24
766	1	126	60	0	0	30.1	0.34
767	1	93	70	31	0	30.4	0.3

768 rows × 9 columns

```
In [4]:
```

Out[6]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.62
1	1	85	66	29	0	26.6	0.3
2	8	183	64	0	0	23.3	0.6
3	1	89	66	23	94	28.1	0.16
4	0	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
765	5	121	72	23	112	26.2	0.24
766	1	126	60	0	0	30.1	0.34
767	1	93	70	31	0	30.4	0.3

768 rows × 9 columns

In [7]: b=dftrain.head(10)

## Out[7]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.627
1	1	85	66	29	0	26.6	0.351
2	8	183	64	0	0	23.3	0.672
3	1	89	66	23	94	28.1	0.167
4	0	137	40	35	168	43.1	2.288
5	5	116	74	0	0	25.6	0.201
6	3	78	50	32	88	31.0	0.248
7	10	115	0	0	0	35.3	0.134
8	2	197	70	45	543	30.5	0.158
9	8	125	96	0	0	0.0	0.232

```
In [8]: a=b[['Pregnancies','Glucose','BloodPressure','SkinThickness','Insulin','BMI','
 Out[8]:
              Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction
           0
                       6
                              148
                                             72
                                                           35
                                                                   0 33.6
                                                                                            0.627
           1
                       1
                              85
                                             66
                                                           29
                                                                   0 26.6
                                                                                            0.351
           2
                       8
                              183
                                             64
                                                           0
                                                                   0 23.3
                                                                                            0.672
           3
                       1
                              89
                                             66
                                                           23
                                                                  94 28.1
                                                                                            0.167
                       0
                              137
                                                           35
                                                                 168 43.1
                                                                                            2.288
                                             40
           5
                       5
                              116
                                             74
                                                           0
                                                                   0 25.6
                                                                                            0.201
                       3
                                                           32
                              78
                                             50
                                                                  88 31.0
                                                                                            0.248
           7
                      10
                              115
                                             0
                                                           0
                                                                   0 35.3
                                                                                            0.134
                       2
                                                           45
           8
                              197
                                             70
                                                                 543
                                                                     30.5
                                                                                            0.158
                       8
                              125
                                             96
                                                           0
                                                                   0
                                                                       0.0
                                                                                            0.232
 In [9]: c=a.iloc[:,0:9]
In [10]:
Out[10]: (10, 9)
In [11]:
Out[11]: (10,)
In [12]:
In [13]:
In [14]: logr=LogisticRegression()
Out[14]: LogisticRegression()
In [15]:
In [16]: prediction=logr.predict(observation)
Out[16]: array([1], dtype=int64)
In [17]:
Out[17]: array([0, 1], dtype=int64)
In [18]:
Out[18]: 2.220446049250313e-16
```

```
In [19]: 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
In [20]: digits=load_digits()
Out[20]: {'data': array([[ 0., 0., 5., ..., 0., 0.,
                 [0., 0., 0., \dots, 10., 0., 0.],
                 [0., 0., 0., ..., 16., 9., 0.],
                 [ 0., 0., 1., ..., 6.,
                                            0., 0.],
                 [0., 0., 2., ..., 12., 0., 0.],
                 [0., 0., 10., ..., 12., 1., 0.]]),
          'target': array([0, 1, 2, ..., 8, 9, 8]),
          'frame': None,
          'feature_names': ['pixel_0_0',
           'pixel_0_1',
           'pixel_0_2',
           'pixel_0_3',
           'pixel_0_4',
           'pixel_0_5',
           'pixel_0_6',
           'pixel_0_7',
            'pixel_1_0',
           'pixel_1_1',
In [21]:
         plt.figure(figsize=(20,4))
         for index,(image,label) in enumerate(zip(digits.data[0:5],digits.target[0:5]))
             plt.subplot(1,5,index+1)
             plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
                                                                             Number:4
In [22]:
In [23]: print(x_train.shape)
         print(x_test.shape)
         print(y_train.shape)
         (1257, 64)
         (540, 64)
         (1257,)
         (540,)
```

```
In [24]: logre=LogisticRegression(max_iter=10000)
Out[24]: LogisticRegression(max_iter=10000)
In [25]:
         [5 9 5 7 7 9 1 2 1 0 1 1 7 0 2 0 7 5 4 0 9 5 1 9 2 2 4 4 8 6 9 6 1 2 9 3 9
          6 8 2 3 9 1 8 6 5 5 0 8 0 3 3 8 4 1 2 8 0 0 1 2 6 6 0 8 0 3 6 1 3 3 6 2 7
          4 5 3 6 8 2 7 0 6 0 5 9 1 6 6 5 5 9 6 1 1 7 2 5 9 7 0 7 1 5 1 6 1 9 6 2 3
          3 2 9 4 8 3 6 1 4 2 0 8 8 1 1 9 9 1 6 2 2 5 0 0 2 2 4 4 3 3 1 9 7 8 0 8 1
          6\ 1\ 1\ 4\ 9\ 0\ 4\ 1\ 8\ 3\ 7\ 7\ 8\ 5\ 1\ 5\ 0\ 0\ 3\ 1\ 1\ 4\ 8\ 2\ 9\ 6\ 1\ 8\ 6\ 6\ 9\ 8\ 4\ 8\ 8\ 1\ 6
          1 4 5 3 5 9 0 5 4 4 5 9 1 2 9 9 8 4 3 6 9 3 9 1 4 2 0 5 5 1 5 5 0 6 0 0 8
          5 7 8 8 4 5 3 8 5 6 3 7 2 1 7 1 9 9 7 2 8 0 0 6 0 1 7 3 2 6 8 1 9 9 1 3 3
          9 4 8 7 0 8 0 4 4 9 7 3 4 6 7 8 0 3 2 4 9 7 5 9 1 5 4 5 0 5 5 5 1 0 1 6 0
          4 7 4 3 1 7 7 8 4 3 1 4 1 8 4 3 2 2 0 1 4 6 6 2 9 5 9 5 7 3 5 5 8 0 4 0 6
          3 6 2 4 6 4 2 8 8 6 7 3 7 5 4 5 3 4 0 6 3 2 7 0 4 7 6 8 4 1 2 7 0 9 5 0 4
          7 2 1 1 5 5 0 2 1 3 6 1 6 0 5 1 7 1 1 9 6 5 2 6 6 8 7 4 5 9 0 2 4 8 9 9 5
          3 9 2 6 4 3 1 1 3 9 9 9 6 4 2 5 2 9 2 0 5 4 0 8 4 6 4 5 6 7 4 8 7 6 6 7 3
          1 2 9 8 7 1 0 5 3 9 8 3 1 9 0 8 4 0 7 8 9 0 7 2 2 5 2 5 8 6 5 1 7 7 2 3 6
          9 2 6 2 6 8 7 7 6 0 1 7 0 5 4 0 6 8 1 5 2 3 9 2 8 0 3 2 2 5 2 1 3 8 9 2 1
          6 5 6 1 1 7 5 4 4 4 8 3 4 7 6 1 3 6 6 7 5 6
In [26]:
         0.9518518518518518
In [ ]:
```