In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

```
a=pd.read_csv(r"C:\Users\user\Downloads\C4_framingham.csv")
a
```

Out[2]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentH
0	1	39	4.0	0	0.0	0.0	0	
1	0	46	2.0	0	0.0	0.0	0	
2	1	48	1.0	1	20.0	0.0	0	
3	0	61	3.0	1	30.0	0.0	0	
4	0	46	3.0	1	23.0	0.0	0	
4233	1	50	1.0	1	1.0	0.0	0	
4234	1	51	3.0	1	43.0	0.0	0	
4235	0	48	2.0	1	20.0	NaN	0	
4236	0	44	1.0	1	15.0	0.0	0	
4237	0	52	2.0	0	0.0	0.0	0	

4238 rows × 16 columns

In [3]:

from sklearn.linear_model import LogisticRegression

```
In [4]:
a=a.head(10)
а
Out[4]:
   male age education currentSmoker cigsPerDay BPMeds prevalentStroke
                                                                              prevalentHyp
                                      0
                                                                                         0
0
           39
                     4.0
                                                0.0
                                                          0.0
                                                                           0
1
       0
           46
                     2.0
                                      0
                                                0.0
                                                          0.0
                                                                           0
                                                                                         0
2
                     1.0
                                      1
                                               20.0
                                                          0.0
                                                                           0
                                                                                         0
       1
           48
                                               30.0
                                                          0.0
3
       0
          61
                     3.0
                                      1
                                                                           0
                                                                                         1
4
       0
           46
                     3.0
                                      1
                                               23.0
                                                          0.0
                                                                           0
                                                                                         0
5
                     2.0
                                      0
                                                0.0
                                                          0.0
                                                                           0
       0
          43
                                                                                         1
6
       0
          63
                     1.0
                                      0
                                                0.0
                                                          0.0
                                                                           0
                                                                                         0
7
       0
          45
                     2.0
                                      1
                                               20.0
                                                          0.0
                                                                           0
                                                                                         0
8
                                      0
                                                0.0
                                                                           0
       1
           52
                     1.0
                                                          0.0
                                                                                         1
9
                     1.0
                                               30.0
                                                          0.0
                                                                           0
       1
           43
                                      1
                                                                                         1
In [14]:
c=a.iloc[:,0:16]
d=a.iloc[:,-1]
In [15]:
c.shape
Out[15]:
(10, 16)
In [16]:
d.shape
Out[16]:
(10,)
In [18]:
from sklearn.preprocessing import StandardScaler
In [19]:
```

fs=StandardScaler().fit_transform(c)

```
In [20]:
logr=LogisticRegression()
logr.fit(fs,d)
Out[20]:
LogisticRegression()
In [21]:
e=[[2,5,77,8,6,5,4,66,88,46,65,76,87,45,92,44]]
In [22]:
prediction=logr.predict(e)
prediction
Out[22]:
array([1], dtype=int64)
In [23]:
logr.classes_
Out[23]:
array([0, 1], dtype=int64)
In [24]:
logr.predict_proba(e)[0][0]
Out[24]:
0.0
In [25]:
logr.predict_proba(e)[0][1]
Out[25]:
1.0
In [5]:
import re
from sklearn.datasets import load_digits
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import sklearn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
```

In [6]:

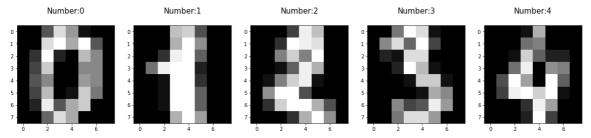
```
digits=load_digits()
digits
```

```
Out[6]:
```

```
{'data': array([[ 0., 0., 5., ..., 0., 0., 0.],
        [0., 0., 0., ..., 10., 0., 0.],
              0., 0., ..., 16., 9.,
        [ 0.,
        [0., 0., 1., ..., 6., 0.,
        [0., 0., 2., ..., 12., 0., 0.],
                                 1.,
        [ 0., 0., 10., ..., 12.,
                                       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',
  'nixel 1 0'.
```

In [7]:

```
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)
    plt.title('Number:%i\n'%label,fontsize=15)
```



In [8]:

```
x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_size=0.30)
```

In [9]:

```
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)

(1257, 64)
```

```
(1257, 64)
(540, 64)
(1257,)
(540,)
```

```
In [10]:
logre=LogisticRegression(max iter=10000)
logre.fit(x_train,y_train)
Out[10]:
LogisticRegression(max_iter=10000)
In [11]:
logre.predict(x_test)
Out[11]:
array([2, 3, 3, 3, 6, 1, 3, 3, 8, 2, 3, 0, 6, 6, 7, 6, 5, 0, 3, 1, 4, 3,
       6, 6, 3, 2, 8, 7, 0, 6, 6, 2, 9, 4, 7, 0, 0, 2, 5, 5, 4, 4, 8, 0,
       9, 4, 5, 1, 2, 4, 5, 5, 4, 6, 0, 4, 7, 6, 7, 0, 5, 1, 3, 7, 3, 2,
       9, 6, 8, 1, 6, 1, 6, 9, 3, 3, 1, 1, 5, 0, 4, 4, 7, 4, 3, 8, 5, 0,
       2, 0, 5, 2, 1, 6, 6, 8, 4, 7, 9, 8, 9, 4, 6, 1, 0, 5, 6, 3, 6, 7,
       6, 5, 9, 5, 8, 2, 5, 8, 8, 8, 2, 7, 0, 8, 7, 2, 7, 7, 0, 8, 3, 5,
       2, 9, 3, 2, 0, 7, 8, 0, 1, 2, 0, 4, 7, 0, 3, 6, 6, 8, 6, 1, 8, 8,
       1, 8, 8, 6, 4, 1, 4, 8, 8, 6, 9, 4, 8, 4, 9, 5, 3, 8, 9, 8, 1, 4,
       5, 3, 7, 3, 2, 3, 7, 4, 5, 5, 1, 6, 7, 9, 5, 8, 8, 3, 5, 6, 4, 3,
       5, 5, 6, 4, 0, 3, 8, 4, 6, 9, 0, 7, 0, 0, 1, 8, 7, 2, 6, 2, 6, 6,
       9, 5, 7, 2, 7, 6, 1, 9, 5, 5, 9, 4, 0, 4, 1, 1, 7, 2, 4, 0, 9, 8,
       6, 6, 4, 7, 1, 0, 3, 9, 2, 9, 2, 3, 4, 8, 0, 5, 8, 6, 4, 3, 1, 1,
       2, 2, 6, 8, 1, 1, 8, 8, 3, 4, 4, 7, 2, 5, 8, 0, 1, 0, 8, 5, 4, 6,
       3, 7, 3, 8, 3, 5, 4, 7, 6, 0, 5, 1, 8, 5, 9, 0, 3, 6, 9, 5, 5, 7,
       0, 8, 1, 2, 7, 5, 0, 0, 1, 8, 0, 0, 6, 8, 6, 2, 7, 2, 4, 2, 5, 9,
       3, 9, 6, 9, 2, 9, 3, 5, 3, 6, 7, 7, 4, 4, 4, 0, 3, 2, 5, 1, 6, 7,
       0, 9, 1, 7, 9, 2, 3, 0, 3, 3, 6, 7, 9, 8, 0, 5, 9, 4, 4, 3, 2, 3,
       7, 1, 4, 6, 0, 7, 5, 0, 1, 0, 4, 0, 6, 0, 4, 1, 9, 0, 7, 1, 4, 1,
       1, 6, 6, 9, 4, 3, 3, 1, 5, 6, 0, 7, 6, 7, 1, 4, 1, 3, 5, 2, 0, 7,
       3, 7, 2, 2, 7, 8, 5, 6, 8, 5, 0, 0, 7, 2, 5, 6, 0, 0, 0, 4, 3, 1,
       8, 2, 5, 7, 9, 2, 6, 3, 7, 4, 7, 8, 9, 3, 0, 1, 8, 9, 7, 2, 4, 4,
       5, 6, 8, 3, 1, 3, 6, 5, 0, 3, 7, 5, 7, 0, 0, 4, 5, 6, 9, 8, 0, 2,
       7, 3, 0, 1, 3, 6, 0, 6, 7, 2, 3, 2, 5, 1, 1, 5, 7, 9, 6, 7, 5, 3,
       9, 8, 4, 0, 5, 6, 1, 1, 6, 7, 2, 1, 7, 4, 3, 3, 2, 7, 5, 5, 8, 2,
       8, 6, 8, 2, 4, 8, 1, 7, 1, 1, 4, 6])
In [12]:
logre.score(x_test,y_test)
Out[12]:
0.9592592592592593
```

In []: