In [1]: 1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 import seaborn as sn

In [2]: 1 from sklearn.linear_model import LogisticRegression

In [3]: 1 df=pd.read_csv(r"framingham")

In [4]: 1 df

Out[4]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	ВМ
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	106.0	70.0	26.9 ⁻
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	121.0	81.0	28.7
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	127.5	80.0	25.3
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	150.0	95.0	28.5
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	130.0	84.0	23.10
													••
4233	1	50	1.0	1	1.0	0.0	0	1	0	313.0	179.0	92.0	25.9 ⁻
4234	1	51	3.0	1	43.0	0.0	0	0	0	207.0	126.5	80.0	19.7
4235	0	48	2.0	1	20.0	NaN	0	0	0	248.0	131.0	72.0	22.00
4236	0	44	1.0	1	15.0	0.0	0	0	0	210.0	126.5	87.0	19.10
4237	0	52	2.0	0	0.0	0.0	0	0	0	269.0	133.5	83.0	21.4 ⁻

4238 rows × 16 columns

```
In [5]:
          1 df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 4238 entries, 0 to 4237
        Data columns (total 16 columns):
             Column
                              Non-Null Count Dtype
         0
             male
                              4238 non-null
                                              int64
                              4238 non-null
         1
                                              int64
             age
             education
         2
                              4133 non-null
                                              float64
             currentSmoker
         3
                              4238 non-null
                                              int64
                              4209 non-null
             cigsPerDay
                                              float64
         5
             BPMeds
                              4185 non-null
                                              float64
             prevalentStroke 4238 non-null
                                              int64
             prevalentHyp
                              4238 non-null
                                              int64
             diabetes
                              4238 non-null
                                              int64
             totChol
         9
                              4188 non-null
                                              float64
         10 sysBP
                              4238 non-null
                                              float64
         11 diaBP
                              4238 non-null
                                              float64
         12 BMI
                              4219 non-null
                                             float64
         13 heartRate
                              4237 non-null
                                              float64
         14 glucose
                              3850 non-null float64
         15 TenYearCHD
                              4238 non-null
                                              int64
        dtypes: float64(9), int64(7)
        memory usage: 529.9 KB
In [6]:
          1 df.dropna(inplace=True)
          1 df.columns
In [7]:
Out[7]: Index(['male', 'age', 'education', 'currentSmoker', 'cigsPerDay', 'BPMeds',
               'prevalentStroke', 'prevalentHyp', 'diabetes', 'totChol', 'sysBP',
               'diaBP', 'BMI', 'heartRate', 'glucose', 'TenYearCHD'],
```

dtype='object')

```
In [8]:
            1 df
 Out[8]:
          rentSmoker cigsPerDay BPMeds prevalentStroke prevalentHyp diabetes totChol sysBP diaBP
                                                                                                       BMI heartRate glucose TenYearCHI
                   0
                                      0.0
                                                                                                70.0 26.97
                             0.0
                                                       0
                                                                    0
                                                                             0
                                                                                  195.0
                                                                                         106.0
                                                                                                                 80.0
                                                                                                                          77.0
                   0
                             0.0
                                      0.0
                                                       0
                                                                    0
                                                                                  250.0
                                                                                         121.0
                                                                                                81.0 28.73
                                                                                                                 95.0
                                                                                                                          76.0
                                                                             0
                            20.0
                                      0.0
                                                       0
                                                                    0
                                                                             0
                                                                                 245.0
                                                                                         127.5
                                                                                                80.0 25.34
                                                                                                                 75.0
                                                                                                                         70.0
                                                                                                95.0 28.58
                            30.0
                                      0.0
                                                       0
                                                                    1
                                                                             0
                                                                                 225.0
                                                                                         150.0
                                                                                                                 65.0
                                                                                                                         103.0
                            23.0
                                      0.0
                                                                    0
                                                                             0
                                                                                  285.0
                                                                                         130.0
                                                                                                84.0 23.10
                                                                                                                 85.0
                                                                                                                          85.0
                   1
                                                       0
                                       ...
                                                                    ...
                                                                            ...
                   0
                             0.0
                                      0.0
                                                       0
                                                                    1
                                                                             0
                                                                                  187.0
                                                                                         141.0
                                                                                                81.0 24.96
                                                                                                                 0.08
                                                                                                                          81.0
                   0
                             0.0
                                      0.0
                                                       0
                                                                    1
                                                                             0
                                                                                 176.0
                                                                                         168.0
                                                                                                97.0 23.14
                                                                                                                 60.0
                                                                                                                          79.0
                             1.0
                                      0.0
                                                       0
                                                                    1
                                                                             0
                                                                                 313.0
                                                                                         179.0
                                                                                                92.0 25.97
                                                                                                                 66.0
                                                                                                                          86.0
                            43.0
                                      0.0
                                                       0
                                                                    0
                                                                             0
                                                                                  207.0
                                                                                         126.5
                                                                                                80.0 19.71
                                                                                                                 65.0
                                                                                                                         68.0
                   0
                             0.0
                                      0.0
                                                                    0
                                                                             0
                                                                                  269.0
                                                                                         133.5
                                                                                                83.0 21.47
                                                                                                                 80.0
                                                                                                                        107.0
            1 from sklearn.linear model import LogisticRegression
 In [9]:
            1 logr =LogisticRegression()
In [10]:
In [12]:
               feature matrix=df[['age', 'education', 'cigsPerDay', 'BPMeds',
                        'prevalentStroke',
            2
                        'totChol', 'sysBP',
            3
                        'diaBP', 'BMI', 'heartRate', 'glucose']]
            5 target vector=df['currentSmoker']
In [13]:
            1 feature matrix.shape
Out[13]: (3656, 11)
```

```
1 target_vector.shape
In [14]:
Out[14]: (3656,)
In [15]:
           1 from sklearn.preprocessing import StandardScaler
In [16]:
           1 fs=StandardScaler().fit transform(feature matrix)
In [17]:
           1 logr=LogisticRegression()
           2 logr.fit(fs,target_vector)
Out[17]: LogisticRegression()
In [22]:
           1 observation=[[1,2,3,4,5,6,7,8,9,10,11]]
           1 prediction = logr.predict(observation)
In [23]:
           2 print(prediction)
         [1]
In [24]:
           1 logr.classes_
Out[24]: array([0, 1], dtype=int64)
In [25]:
           1 logr.predict proba(observation)[0][1]
Out[25]: 1.0
           1 logr.predict_proba(observation)[0][0]
In [26]:
Out[26]: 0.0
```

Linear regression 2

```
In [27]:
          1 import re
          2 from sklearn.datasets import load_digits
          3 import numpy as np
          4 import pandas as pd
          5 import matplotlib.pyplot as plt
          6 import seaborn as sns
          7 from sklearn.linear_model import LogisticRegression
          8 from sklearn.model_selection import train_test_split
In [28]:
          1 digits =load_digits()
          2 digits
Out[28]: {'data': array([[ 0., 0., 5., ..., 0., 0., 0.],
                [0., 0., 0., ..., 10., 0., 0.],
                [0., 0., 0., ..., 16., 9., 0.],
                [0., 0., 1., \ldots, 6., 0., 0.],
                [0., 0., 2., ..., 12., 0., 0.],
                [0., 0., 10., \ldots, 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 [29]:
           1 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)
           3
                  plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
           4
                  plt.title("Number:%i\n"%label,fontsize=15)
           5
                 Number:0
                                        Number:1
                                                              Number:2
                                                                                     Number:3
                                                                                                           Number:4
In [30]:
           1 x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_size=0.30)
In [31]:
           1 print(x train.shape)
           2 print(x test.shape)
           3 print(y train.shape)
             print(y test.shape)
         (1257, 64)
         (540, 64)
         (1257,)
         (540,)
           1 logre=LogisticRegression(max_iter=10000)
In [32]:
           2 logre.fit(x_train,y_train)
```

Out[32]: LogisticRegression(max_iter=10000)

```
1 print(logre.predict(x_test))
In [33]:
         [6 9 2 6 7 4 0 0 3 1 8 0 3 7 6 4 8 9 6 6 8 3 5 4 3 4 0 6 7 1 0 5 6 6 6 6 2
          0 0 6 5 8 2 3 5 7 1 7 8 8 2 0 3 9 9 4 5 1 1 5 8 4 1 3 8 5 8 3 1 2 3 4 0 6
          5 5 9 3 4 5 0 4 9 2 4 2 7 6 6 6 6 9 4 5 6 7 0 0 4 7 5 6 0 6 5 8 6 3 9 2 6
          3 6 4 9 0 4 7 5 9 3 0 6 5 3 2 6 1 9 8 1 1 8 7 8 3 3 4 2 2 5 8 7 8 4 2 8 9
          7 1 2 6 0 0 8 5 5 3 7 7 9 5 9 4 0 7 6 6 1 4 8 2 8 6 7 7 0 0 3 2 4 6 1 2 1
          3 8 2 9 7 9 5 0 5 8 3 3 7 9 3 5 0 7 2 2 3 1 1 3 6 2 2 5 3 5 7 1 7 2 4 5 8
          2 3 1 2 4 8 7 9 9 6 6 7 1 6 3 2 8 8 9 2 9 0 5 2 5 1 5 7 7 5 6 9 8 1 8 9 2
          9 2 7 3 2 5 7 0 4 9 5 4 9 7 7 5 2 0 0 5 2 0 4 1 0 6 5 0 8 5 6 3 8 2 4 1 9
          0 6 6 3 7 5 7 6 8 9 3 0 2 1 9 0 2 6 2 7 3 8 6 2 0 1 3 3 3 0 4 7 3 2 2 5 0
          3 6 1 5 0 2 8 4 4 0 7 8 6 6 4 0 5 7 9 3 6 5 5 1 0 1 1 0 0 1 8 0 6 5 9 3 5
          3 6 4 7 0 6 8 3 0 2 1 1 3 3 3 0 3 9 0 1 7 1 2 0 4 1 9 1 5 3 7 3 9 2 3 4 2
          4 5 9 4 8 3 6 4 6 2 0 3 4 3 5 1 7 3 3 1 6 4 0 5 5 3 8 9 7 6 1 8 0 1 0 7 1
          2 5 5 1 6 4 3 9 4 2 2 7 5 3 0 2 3 9 6 6 5 4 0 6 6 1 8 1 8 6 7 9 7 3 1 1 7
          2 2 8 9 3 4 3 8 5 7 2 5 3 0 6 6 6 8 5 1 6 7 8 4 8 7 1 8 5 4 1 3 5 6 3 2 3
          1 4 2 3 2 2 9 3 6 6 6 5 3 1 5 3 8 4 0 5 6 1
In [34]:
           1 print(logre.score(x test,y test))
```

0.95555555555556