In [1]:

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

In [3]:

a=pd.read_csv(r"C:\Users\user\Downloads\C2_train.gender_submission.csv")
a

Out[3]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Far
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.250
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.283
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.925
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.100
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.050
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.000
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.000
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.450
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.000
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.750

891 rows × 12 columns

In [4]:

from sklearn.linear_model import LogisticRegression

In [10]:

```
a=a.head(10)
a
```

Out[10]:

1		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
Mrs. John Bradley (Florence Briggs Th 1	0	1	0	3	Mr. Owen	male	22.0	1	0	A/5 21171	7.2500
2 3 1 3 Miss. Laina Futrelle, Mrs. Mrs. Jacques Heath (Lily May Peel) 4 5 0 3 Moran, Mr. James 6 7 0 1 Mr. Timothy J Palsson, Master. Timothy J Palsson, Mrs. Coscar W (Elisabeth Vilhelmina Berg) Nasser, Mrs. Macy Mrs. Pemale 26.0 0 0 3 TON/OZ. 3101282 7.9250 7	1	2	1	1	Mrs. John Bradley (Florence Briggs	female	38.0	1	0	PC 17599	71.2833
Mrs. Jacques Heath (Lily May Peel) Female 35.0 1 0 113803 53.1000 A	2	3	1	3	Miss.	female	26.0	0	0		7.9250
4 5 0 3 William Henry male 35.0 0 0 373450 8.0500 5 6 0 3 Moran, Mr. James male NaN 0 0 330877 8.4583 6 7 0 1 Mr. Mr. Timothy J male 54.0 0 0 17463 51.8625 7 8 0 3 Master. Gosta Leonard male 2.0 3 1 349909 21.0750 8 9 1 3 Oscar W (Elisabeth Vilhelmina Berg) female 27.0 0 2 347742 11.1333 9 10 1 2 Nasser, Mrs. Mrs. (Adele 14.0 1 0 237736 30.0708	3	4	1	1	Mrs. Jacques Heath (Lily May	female	35.0	1	0	113803	53.1000
6 7 0 1 McCarthy, Timothy J male 54.0 0 0 17463 51.8625 7 8 0 3 Palsson, Master. Gosta Leonard male 2.0 3 1 349909 21.0750 8 9 1 3 Oscar W (Elisabeth Vilhelmina Berg) female 27.0 0 2 347742 11.1333 9 10 1 2 Nicholas (Adele female 14.0 1 0 237736 30.0708	4	5	0	3	William	male	35.0	0	0	373450	8.0500
6 7 0 1 Mr. Timothy J male 54.0 0 0 17463 51.8625 7 8 0 3 Palsson, Master. Gosta Leonard male 2.0 3 1 349909 21.0750 8 9 1 3 Oscar W (Elisabeth Vilhelmina Berg) female 27.0 0 2 347742 11.1333 9 10 1 2 Nicholas (Adele female 14.0 1 0 237736 30.0708	5	6	0	3		male	NaN	0	0	330877	8.4583
7 8 0 3 Master. Gosta Leonard male 2.0 3 1 349909 21.0750 8 9 1 3 Oscar W (Elisabeth Vilhelmina Berg) female 27.0 0 2 347742 11.1333 9 10 1 2 Nicholas (Adele female 14.0 1 0 237736 30.0708	6	7	0	1	Mr.	male	54.0	0	0	17463	51.8625
8 9 1 3	7	8	0	3	Master. Gosta	male	2.0	3	1	349909	21.0750
Mrs. 9 10 1 2 Nicholas female 14.0 1 0 237736 30.0708 (Adele	8	9	1	3	Mrs. Oscar W (Elisabeth Vilhelmina	female	27.0	0	2	347742	11.1333
	9	10	1	2	Mrs. Nicholas (Adele	female	14.0	1	0	237736	30.0708

In [11]:

```
a.columns
```

Out[11]:

```
In [12]:
b=a[['PassengerId', 'Survived', 'Pclass', 'SibSp', 'Parch']]
Out[12]:
   Passengerld Survived Pclass SibSp Parch
 0
                     0
                             3
                                          0
 1
            2
                     1
                             1
                                   1
                                          0
 2
            3
                      1
                             3
                                   0
                                          0
 3
            4
                             1
                     1
                                   1
                                          0
 4
            5
                     0
                             3
                                   0
                                          0
            6
                     0
                             3
 5
                                   0
                                          0
            7
 6
                     0
                             1
                                   0
                                          0
 7
            8
                     0
                             3
                                   3
                                          1
 8
            9
                     1
                             3
                                          2
                                   0
 9
            10
                     1
                             2
                                          0
                                   1
In [13]:
c=b.iloc[:,0:5]
d=a.iloc[:,-1]
In [14]:
c.shape
Out[14]:
(10, 5)
In [15]:
d.shape
Out[15]:
(10,)
In [41]:
fs=StandardScaler().fit_transform(c)
In [17]:
logr=LogisticRegression()
```

logr.fit(fs,d)

LogisticRegression()

Out[17]:

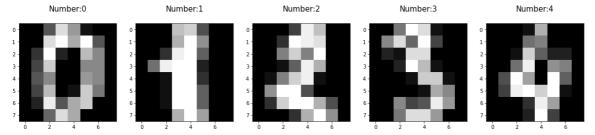
```
In [19]:
e=[[2,5,77,8,65]]
In [20]:
prediction=logr.predict(e)
prediction
Out[20]:
array(['S'], dtype=object)
In [21]:
logr.classes_
Out[21]:
array(['C', 'Q', 'S'], dtype=object)
In [22]:
logr.predict_proba(e)[0][0]
Out[22]:
1.3926038933604862e-33
In [23]:
logr.predict_proba(e)[0][1]
Out[23]:
8.694120921868689e-08
In [29]:
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 [30]:

```
digits=load_digits()
digits
  pıxeı_/_v ,
  'pixel_7_1',
  'pixel_7_2',
  'pixel_7_3',
  'pixel_7_4',
  'pixel_7_5',
  'pixel_7_6',
  'pixel 7 7'],
 'target_names': array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]),
 'images': array([[[ 0., 0., 5., ..., 1., 0., 0.],
        [0., 0., 13., ..., 15., 5., 0.],
        [ 0., 3., 15., ..., 11., 8.,
                                       0.],
         . . . ,
         [ 0.,
               4., 11., ..., 12., 7., 0.],
              2., 14., ..., 12.,
                                   0.,
        [ 0.,
               0., 6., ..., 0.,
                                   0.,
                                       0.]],
        [[ 0., 0., 0., ..., 5.,
                                 0.,
                                       0.],
        [ 0., 0., 0., ..., 9.,
                                   0., 0.],
        [0., 0., 3., ..., 6.,
                                   0., 0.],
```

In [34]:

```
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 [35]:

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

In [36]:

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

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

```
In [38]:
logre=LogisticRegression(max iter=10000)
logre.fit(x_train,y_train)
Out[38]:
LogisticRegression(max_iter=10000)
In [39]:
 1 logre.predict(x_test)
Out[39]:
array([6, 9, 4, 9, 6, 1, 1, 2, 9, 5, 8, 5, 3, 0, 3, 2, 4, 6, 3, 8, 2, 0,
       6, 7, 5, 1, 6, 2, 6, 2, 3, 6, 4, 0, 6, 9, 7, 8, 5, 5, 0, 9, 4, 4,
       7, 5, 1, 8, 2, 0, 6, 5, 6, 0, 1, 9, 8, 1, 3, 1, 6, 7, 9, 1, 2, 8,
       1, 2, 2, 5, 7, 6, 8, 4, 5, 4, 4, 3, 9, 0, 7, 8, 6, 4, 2, 2, 8, 8,
       5, 0, 7, 6, 6, 1, 1, 4, 1, 9, 6, 5, 7, 2, 6, 3, 0, 4, 2, 8, 1, 1,
       3, 2, 8, 7, 6, 5, 7, 7, 3, 0, 8, 5, 5, 2, 4, 3, 3, 7, 2, 0, 0, 0,
       9, 9, 6, 5, 6, 8, 8, 4, 6, 9, 6, 2, 6, 8, 9, 7, 1, 9, 8, 4, 3, 8,
       8, 1, 1, 8, 4, 9, 3, 1, 0, 2, 1, 4, 7, 2, 5, 4, 2, 0, 8, 5, 3, 8,
       8, 0, 0, 4, 3, 2, 6, 4, 6, 3, 8, 3, 1, 2, 3, 5, 9, 0, 0, 1, 7, 5,
       6, 5, 8, 1, 3, 0, 9, 2, 0, 7, 4, 8, 7, 2, 6, 7, 8, 5, 5, 9, 5, 4,
       4, 5, 7, 6, 9, 8, 3, 9, 3, 5, 3, 8, 4, 1, 6, 6, 7, 3, 8, 7, 1, 7,
       0, 7, 2, 8, 5, 7, 2, 1, 9, 4, 0, 1, 9, 4, 9, 1, 1, 6, 6, 9, 0, 1,
       6, 4, 4, 2, 8, 6, 5, 6, 4, 3, 8, 7, 2, 0, 8, 7, 3, 6, 2, 8, 2, 9,
       0, 5, 9, 9, 4, 8, 1, 3, 2, 2, 4, 5, 1, 3, 5, 0, 8, 4, 1, 1, 2, 6,
       4, 6, 0, 1, 2, 6, 2, 0, 4, 2, 5, 0, 2, 6, 9, 0, 6, 3, 9, 4, 4, 9,
       6, 6, 7, 8, 9, 8, 4, 9, 2, 8, 9, 0, 3, 7, 0, 1, 0, 2, 6, 0, 7, 2,
       9, 4, 1, 4, 2, 6, 4, 0, 9, 6, 0, 9, 5, 1, 6, 7, 4, 5, 7, 9, 8, 7,
       1, 0, 4, 3, 0, 7, 4, 1, 3, 6, 1, 3, 7, 4, 9, 9, 6, 3, 9, 6, 4, 7,
       9, 4, 0, 2, 9, 9, 0, 8, 2, 9, 9, 9, 0, 1, 4, 6, 5, 2, 1, 0, 8, 8,
       6, 7, 9, 3, 2, 7, 5, 5, 1, 7, 0, 8, 1, 0, 7, 1, 4, 7, 3, 6, 4, 8,
       4, 7, 7, 1, 4, 1, 1, 2, 7, 6, 8, 3, 2, 1, 8, 5, 7, 1, 5, 8, 7, 0,
       7, 1, 0, 1, 0, 4, 2, 5, 2, 6, 9, 9, 2, 7, 4, 6, 3, 7, 3, 1, 7, 9,
       6, 6, 4, 3, 4, 5, 1, 3, 9, 9, 2, 4, 9, 7, 1, 1, 8, 0, 5, 3, 4, 3,
       8, 5, 1, 2, 6, 8, 5, 8, 1, 5, 4, 1, 2, 5, 9, 8, 3, 4, 9, 5, 4, 1,
       2, 1, 0, 9, 2, 7, 6, 1, 5, 0, 8, 8])
In [40]:
   logre.score(x_test,y_test)
Out[40]:
0.9611111111111111
In [ ]:
```