

```
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
```

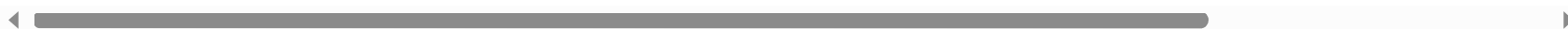
```
In [3]: 1 df=pd.read_csv(r"C8_loan-test")
```

```
In [4]: 1 df
```

Out[4]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amo
0	LP001015	Male	Yes	0	Graduate	No	5720	0	110.0	
1	LP001022	Male	Yes	1	Graduate	No	3076	1500	126.0	
2	LP001031	Male	Yes	2	Graduate	No	5000	1800	208.0	
3	LP001035	Male	Yes	2	Graduate	No	2340	2546	100.0	
4	LP001051	Male	No	0	Not Graduate	No	3276	0	78.0	
...	...	...	...	...	...	...	...	...	...	...
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	1777	113.0	
363	LP002975	Male	Yes	0	Graduate	No	4158	709	115.0	
364	LP002980	Male	No	0	Graduate	No	3250	1993	126.0	
365	LP002986	Male	Yes	0	Graduate	No	5000	2393	158.0	
366	LP002989	Male	No	0	Graduate	Yes	9200	0	98.0	

367 rows × 12 columns



In [5]: 1 df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 367 entries, 0 to 366
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Loan_ID                367 non-null   object
1   Gender                 356 non-null   object
2   Married                367 non-null   object
3   Dependents             357 non-null   object
4   Education              367 non-null   object
5   Self_Employed          344 non-null   object
6   ApplicantIncome        367 non-null   int64
7   CoapplicantIncome      367 non-null   int64
8   LoanAmount             362 non-null   float64
9   Loan_Amount_Term       361 non-null   float64
10  Credit_History          338 non-null   float64
11  Property_Area          367 non-null   object
dtypes: float64(3), int64(2), object(7)
memory usage: 34.5+ KB
```

In [6]: 1 df.dropna(inplace=True)

In [7]: 1 df.columns

Out[7]: Index(['Loan\_ID', 'Gender', 'Married', 'Dependents', 'Education',  
          'Self\_Employed', 'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',  
          'Loan\_Amount\_Term', 'Credit\_History', 'Property\_Area'],  
          dtype='object')

In [8]:

```
1 df
```

Out[8]:

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_An
0	LP001015	Male	Yes	0	Graduate	No	5720	0	110.0	
1	LP001022	Male	Yes	1	Graduate	No	3076	1500	126.0	
2	LP001031	Male	Yes	2	Graduate	No	5000	1800	208.0	
4	LP001051	Male	No	0	Not Graduate	No	3276	0	78.0	
5	LP001054	Male	Yes	0	Not Graduate	Yes	2165	3422	152.0	
...	...	...	...	...	...	...	...	...	...	...
361	LP002969	Male	Yes	1	Graduate	No	2269	2167	99.0	
362	LP002971	Male	Yes	3+	Not Graduate	Yes	4009	1777	113.0	
363	LP002975	Male	Yes	0	Graduate	No	4158	709	115.0	

In [9]:

```
1 from sklearn.linear_model import LogisticRegression
```

In [10]:

```
1 logr =LogisticRegression()
```

In [11]:

```
1 feature_matrix=df[['ApplicantIncome', 'CoapplicantIncome', 'LoanAmount',  
2 'Loan_Amount_Term', 'Credit_History']]  
3 target_vector=df['Married']
```

In [14]:

```
1 feature_matrix.shape
```

Out[14]: (289, 5)

In [15]:

```
1 target_vector.shape
```

Out[15]: (289,)

```
In [16]: 1 from sklearn.preprocessing import StandardScaler
```

```
In [17]: 1 fs=StandardScaler().fit_transform(feature_matrix)
```

```
In [18]: 1 logr=LogisticRegression()  
2 logr.fit(fs,target_vector)
```

```
Out[18]: LogisticRegression()
```

```
In [20]: 1 observation=[[1,2,3,4,5]]
```

```
In [21]: 1 prediction = logr.predict(observation)  
2 print(prediction)  
  
['Yes']
```

```
In [22]: 1 logr.classes_
```

```
Out[22]: array(['No', 'Yes'], dtype=object)
```

```
In [23]: 1 logr.predict_proba(observation)[0][1]
```

```
Out[23]: 0.9680271507223828
```

```
In [24]: 1 logr.predict_proba(observation)[0][0]
```

```
Out[24]: 0.03197284927761723
```

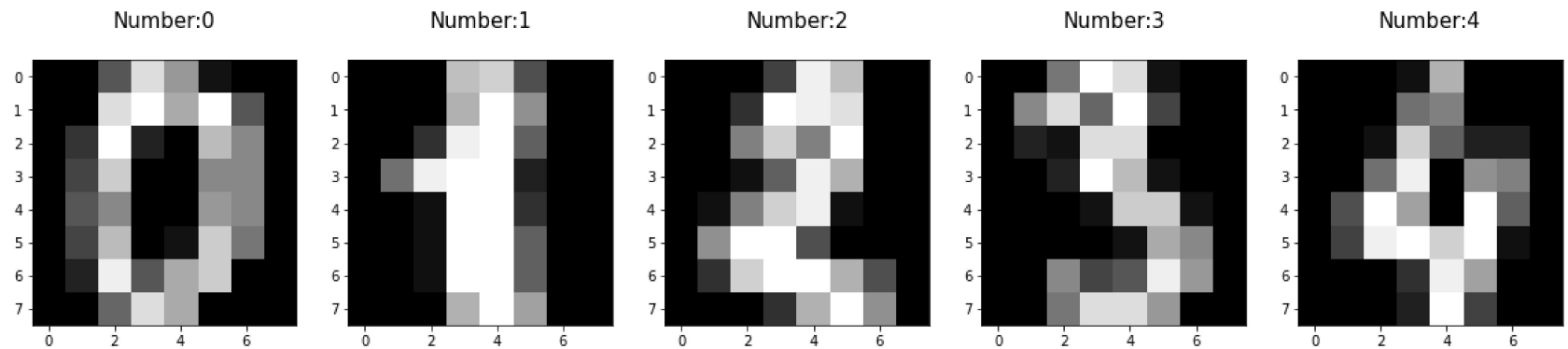
## Linear regression 2

```
In [25]: 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 [26]: 1 digits =load_digits()
          2 digits
```

```
Out[26]: {'data': array([[ 0.,  0.,  5., ...,  0.,  0.,  0.],
                          [ 0.,  0.,  0., ..., 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 [27]: 1 plt.figure(figsize=(20,4))
2 for index,(image,label) in enumerate(zip(digits.data[0:5],digits.target[0:5])):
3     plt.subplot(1,5,index+1)
4     plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
5     plt.title("Number:%i\n"%label,fontsize=15)
```



```
In [28]: 1 x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_size=0.30)
```

```
In [29]: 1 print(x_train.shape)
2 print(x_test.shape)
3 print(y_train.shape)
4 print(y_test.shape)
```

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

```
In [30]: 1 logre=LogisticRegression(max_iter=10000)
2 logre.fit(x_train,y_train)
```

```
Out[30]: LogisticRegression(max_iter=10000)
```

```
In [31]: 1 print(logre.predict(x_test))
```

```
[3 8 2 8 5 5 7 4 8 7 3 5 3 0 4 7 3 8 5 3 8 1 5 0 6 8 4 4 3 8 9 8 5 2 2 2 1
 4 1 9 2 4 8 0 1 9 4 6 4 5 1 4 8 4 0 8 7 9 4 5 4 0 8 0 5 6 6 2 1 0 1 9 5 3
 2 9 7 1 4 6 4 7 4 6 3 4 4 2 3 2 3 6 6 6 2 7 9 7 9 2 8 7 4 1 3 6 8 6 3 2 0
 0 1 2 4 4 3 3 7 8 6 2 1 8 1 2 6 4 8 0 7 0 4 6 7 7 4 3 0 1 6 2 7 0 2 6 5 9
 1 5 0 6 4 0 6 8 3 5 4 7 7 9 4 2 1 9 3 6 6 7 8 5 5 8 3 2 3 6 8 1 5 1 8 7 8
 4 1 6 9 2 7 3 1 8 2 9 4 7 0 7 0 9 1 6 1 9 6 4 2 3 2 3 0 4 6 2 1 8 4 0 7 1
 5 6 0 3 1 9 9 6 2 3 1 5 2 6 7 5 3 1 1 8 2 9 1 8 1 2 2 3 3 0 7 6 7 7 3 1 1
 3 5 8 7 0 1 1 6 4 5 8 0 1 6 0 0 2 7 9 7 6 3 2 1 9 4 2 6 0 9 4 2 2 4 6 0 3
 2 3 8 7 7 7 5 1 9 8 8 1 6 0 3 5 1 0 0 9 0 3 5 8 2 1 3 9 5 0 4 2 0 0 0 3 1
 4 3 3 9 5 3 2 7 9 5 5 0 3 7 5 1 7 2 1 5 8 3 3 8 9 6 0 9 0 9 1 0 8 2 5 6 0
 0 9 4 6 2 7 0 7 5 6 3 8 9 7 7 3 7 6 9 6 8 5 6 7 8 6 2 0 2 2 6 7 3 9 0 0 8
 9 7 7 1 2 1 2 5 8 2 2 0 5 0 8 9 5 4 2 4 0 7 9 1 6 4 6 1 8 5 4 5 9 3 9 0 9
 7 9 7 4 3 7 3 0 0 7 8 1 7 2 8 1 2 9 1 6 9 5 2 0 5 0 0 5 6 0 3 4 2 0 7 9 2
 4 6 7 6 9 2 0 8 6 9 3 5 5 7 4 3 8 6 9 8 2 2 5 9 1 9 4 1 8 6 1 4 2 6 6 7 2
 8 2 8 7 9 6 7 8 0 3 4 5 2 5 4 1 2 6 3 6 0 3]
```

```
In [32]: 1 print(logre.score(x_test,y_test))
```

```
0.9685185185185186
```