

mk 02-09-2023

```
In [544]: 1 import numpy as np
          2 import pandas as pd
          3 import matplotlib.pyplot as plt
          4 import seaborn as sns
```

```
In [545]: 1 from sklearn.linear_model import LogisticRegression
          2 a=pd.read_csv(r"C:\USERS\user\Downloads\C6_bmi.csv")
          3 a
```

	Gender	Height	Weight	Mass
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

500 rows × 4 columns

```
In [591]: 1 a=a.head(100)
          2 a
```

Out[591]:

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
5	Male	189	104	3
6	Male	147	92	5
7	Male	154	111	5
8	Male	174	90	3
9	Female	169	103	4
10	Male	195	81	2
11	Female	159	80	4
12	Female	192	101	3
13	Male	155	51	2
14	Male	191	79	2
15	Female	153	107	5
16	Female	157	110	5
17	Male	140	129	5
18	Male	144	145	5
19	Male	172	139	5
20	Male	157	110	5
21	Female	153	149	5
22	Female	169	97	4
23	Male	185	139	5
24	Female	172	67	2
25	Female	151	64	3
26	Male	190	95	3
27	Male	187	62	1
28	Female	163	159	5
29	Male	179	152	5
30	Male	153	121	5
31	Male	178	52	1
32	Female	195	65	1
33	Female	160	131	5
34	Female	157	153	5
35	Female	189	132	4
36	Female	197	114	3
37	Male	144	80	4
38	Female	171	152	5

	Gender	Height	Weight	Index
39	Female	185	81	2
40	Female	175	120	4
41	Female	149	108	5
42	Male	157	56	2
43	Male	161	118	5
44	Female	182	126	4
45	Male	185	76	2
46	Female	188	122	4
47	Male	181	111	4
48	Male	161	72	3
49	Male	140	152	5

```
In [592]: 1 from sklearn.linear_model import LogisticRegression
```

```
In [593]: 1 a.columns
```

```
Out[593]: Index(['Gender', 'Height', 'Weight', 'Index'], dtype='object')
```

```
In [594]: 1 b=a[[ 'Height', 'Weight', 'Index']]  
          2 b
```

Out[594]:

	Height	Weight	Index
0	174	96	4
1	189	87	2
2	185	110	4
3	195	104	3
4	149	61	3
5	189	104	3
6	147	92	5
7	154	111	5
8	174	90	3
9	169	103	4
10	195	81	2
11	159	80	4
12	192	101	3
13	155	51	2
14	191	79	2
15	153	107	5
16	157	110	5
17	140	129	5
18	144	145	5
19	172	139	5
20	157	110	5
21	153	149	5
22	169	97	4
23	185	139	5
24	172	67	2
25	151	64	3
26	190	95	3
27	187	62	1
28	163	159	5
29	179	152	5
30	153	121	5
31	178	52	1
32	195	65	1
33	160	131	5
34	157	153	5
35	189	132	4
36	197	114	3
37	144	80	4
38	171	152	5

	Height	Weight	Index
39	185	81	2
40	175	120	4
41	149	108	5
42	157	56	2
43	161	118	5
44	182	126	4
45	185	76	2
46	188	122	4
47	181	111	4
48	161	72	3
49	140	152	5

```
In [595]: 1 c=b.iloc[:,0:5]  
          2 d=b.iloc[:, -1]
```

```
In [596]: 1 c.shape
```

```
Out[596]: (50, 3)
```

```
In [597]: 1 d.shape
```

```
Out[597]: (50,)
```

```
In [598]: 1 from sklearn.preprocessing import StandardScaler
          2 fs=StandardScaler().fit_transform(c)
          3 fs
```

```
Out[598]: array([[ 0.23938063, -0.28077719,  0.23354968],
 [ 1.12379184, -0.58450252, -1.32344821],
 [ 0.88794885,  0.19168443,  0.23354968],
 [ 1.47755633, -0.01079912, -0.54494926],
 [-1.23463805, -1.46193123, -0.54494926],
 [ 1.12379184, -0.01079912, -0.54494926],
 [-1.35255954, -0.41576622,  1.01204863],
 [-0.93983431,  0.22543169,  1.01204863],
 [ 0.23938063, -0.48326074, -0.54494926],
 [-0.0554231 , -0.04454638,  0.23354968],
 [ 1.47755633, -0.78698607, -1.32344821],
 [-0.64503058, -0.82073332,  0.23354968],
 [ 1.30067409, -0.1120409 , -0.54494926],
 [-0.88087356, -1.79940382, -1.32344821],
 [ 1.24171334, -0.85448058, -1.32344821],
 [-0.99879506,  0.09044265,  1.01204863],
 [-0.76295207,  0.19168443,  1.01204863],
 [-1.76528477,  0.83288234,  1.01204863],
 [-1.52944179,  1.37283847,  1.01204863],
 [ 0.12145914,  1.17035492,  1.01204863],
 [-0.76295207,  0.19168443,  1.01204863],
 [-0.99879506,  1.50782751,  1.01204863],
 [-0.0554231 , -0.24702993,  0.23354968],
 [ 0.88794885,  1.17035492,  1.01204863],
 [ 0.12145914, -1.25944768, -1.32344821],
 [-1.11671655, -1.36068946, -0.54494926],
 [ 1.18275259, -0.31452445, -0.54494926],
 [ 1.00587035, -1.42818398, -2.10194715],
 [-0.40918759,  1.84530009,  1.01204863],
 [ 0.53418437,  1.60906928,  1.01204863],
 [-0.99879506,  0.56290427,  1.01204863],
 [ 0.47522362, -1.76565656, -2.10194715],
 [ 1.47755633, -1.3269422 , -2.10194715],
 [-0.58606983,  0.90037685,  1.01204863],
 [-0.76295207,  1.64281654,  1.01204863],
 [ 1.12379184,  0.93412411,  0.23354968],
 [ 1.59547782,  0.32667346, -0.54494926],
 [-1.52944179, -0.82073332,  0.23354968],
 [ 0.06249839,  1.60906928,  1.01204863],
 [ 0.88794885, -0.78698607, -1.32344821],
 [ 0.29834138,  0.52915701,  0.23354968],
 [-1.23463805,  0.12418991,  1.01204863],
 [-0.76295207, -1.63066753, -1.32344821],
 [-0.52710908,  0.46166249,  1.01204863],
 [ 0.71106661,  0.73164056,  0.23354968],
 [ 0.88794885, -0.95572236, -1.32344821],
 [ 1.0648311 ,  0.59665153,  0.23354968],
 [ 0.65210587,  0.22543169,  0.23354968],
 [-0.52710908, -1.09071139, -0.54494926],
 [-1.76528477,  1.60906928,  1.01204863]])
```

```
In [599]: 1 logr=LogisticRegression()
          2 logr.fit(fs,d)
```

```
Out[599]: LogisticRegression()
```



```
In [600]: 1 e=[[77,9,55]]
```

```
In [601]: 1 prediction=logr.predict(e)
          2 prediction
```

```
Out[601]: array([4], dtype=int64)
```

```
In [602]: 1 logr.classes_
```

```
Out[602]: array([1, 2, 3, 4, 5], dtype=int64)
```

```
In [603]: 1 logr.predict_proba(e)[0][0]
```

```
Out[603]: 3.415792265699517e-51
```

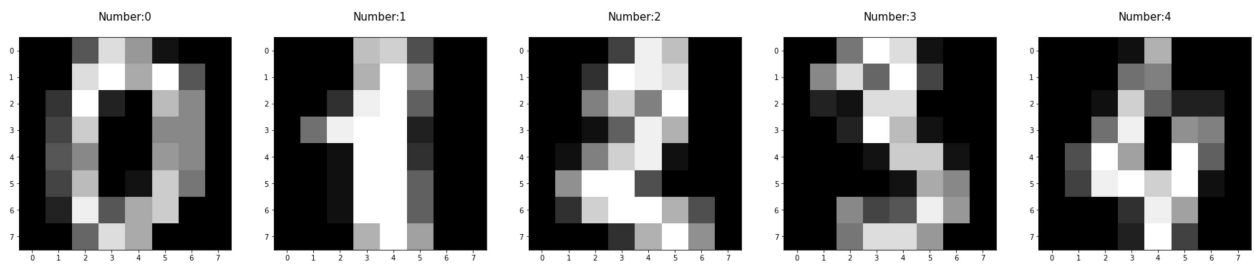
```
In [604]: 1 import re
          2 from sklearn.datasets import load_digits
          3 import numpy as np
          4 import pandas as pd
          5 import matplotlib.pyplot as plt
          6 import seaborn as sns
```

```
In [605]: 1 from sklearn.linear_model import LogisticRegression
          2 from sklearn.model_selection import train_test_split
```

```
In [606]: 1 digits=load_digits()
          2 digits
```

```
pixel_1_0',
'pixel_1_4',
'pixel_1_5',
'pixel_1_6',
'pixel_1_7',
'pixel_2_0',
'pixel_2_1',
'pixel_2_2',
'pixel_2_3',
'pixel_2_4',
'pixel_2_5',
'pixel_2_6',
'pixel_2_7',
'pixel_3_0',
'pixel_3_1',
'pixel_3_2',
'pixel_3_3',
'pixel_3_4',
'pixel_3_5',
'pixel_3_6',
'pixel_3_7'
```

```
In [607]: 1 plt.figure(figsize=(50,25))
2 for index,(image,label) in enumerate(zip(digits.data[0:8],digits.target[0:5])):
3     plt.subplot(1,8,index+1)
4     plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
5     plt.title('Number:%i\n'%label,fontsize=15)
```



```
In [608]: 1 x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_size=0.1)
```

```
In [609]: 1 print(x_train.shape)
2 print(x_test.shape)
3 print(y_train.shape)
4 print(y_test.shape)
```

```
(898, 64)
(899, 64)
(898,)
(899,)
```

```
In [610]: 1 logre=LogisticRegression(max_iter=10000)
2 logre.fit(x_train,y_train)
3
```

Out[610]: LogisticRegression(max_iter=10000)

```
In [611]: 1 print(logre.predict(x_test))
```

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

```
In [612]: 1 import numpy as np
          2 import pandas as pd
          3 import matplotlib.pyplot as plt
          4 import seaborn as sns
```

```
In [613]: 1 a=pd.read_csv(r"C:\USERS\user\Downloads\C6_bmi.csv")
```

```
In [614]: 1 a=a.head(100)
          2 a
```

Out[614]:

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...
95	Female	170	156	5
96	Male	142	69	4
97	Male	160	139	5
98	Male	195	69	1
99	Female	190	50	0

100 rows × 4 columns

```
In [615]: 1 b=a[['Gender', 'Height', 'Weight', 'Index']]
          2 b
```

Out[615]:

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...
95	Female	170	156	5
96	Male	142	69	4
97	Male	160	139	5
98	Male	195	69	1
99	Female	190	50	0

100 rows × 4 columns

```
In [616]: 1 b['Gender'].value_counts()
```

```
Out[616]: Male      51
          Female    49
          Name: Gender, dtype: int64
```

```
In [617]: 1 x=b.drop('Gender',axis=1)
          2 y=b['Gender']
          3 print(b)
```

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
..
95	Female	170	156	5
96	Male	142	69	4
97	Male	160	139	5
98	Male	195	69	1
99	Female	190	50	0

[100 rows x 4 columns]

```
In [618]: 1 g1={"Gender":{"g1":1}}
          2 a=a.replace(g1)
          3 print(a)
```

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
..
95	Female	170	156	5
96	Male	142	69	4
97	Male	160	139	5
98	Male	195	69	1
99	Female	190	50	0

[100 rows x 4 columns]

```
In [619]: 1 from sklearn.model_selection import train_test_split
          2 x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [620]: 1 from sklearn.ensemble import RandomForestClassifier
```

```
In [621]: 1 rfc=RandomForestClassifier()
          2 rfc.fit(x_train,y_train)
```

```
Out[621]: RandomForestClassifier()
```

```
In [622]: 1 parameters={'max_depth':[1,2,3,4,5],
          2              'min_samples_leaf':[5,10,15,20,25],
          3              'n_estimators':[10,20,30,40,50]}
```

```
In [623]: 1 from sklearn.model_selection import GridSearchCV
```

```
In [624]: 1 grid_search=GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="accuracy")  
2 grid_search.fit(x_train,y_train)
```

```
Out[624]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),  
                      param_grid={'max_depth': [1, 2, 3, 4, 5],  
                                'min_samples_leaf': [5, 10, 15, 20, 25],  
                                'n_estimators': [10, 20, 30, 40, 50]},  
                      scoring='accuracy')
```

```
In [625]: 1 grid_search.best_score_
```

```
Out[625]: 0.6142857142857143
```

```
In [626]: 1 rfc_best=grid_search.best_estimator_
```

```
In [627]: 1 from sklearn.tree import plot_tree
```

```
In [628]: 1 plt.figure(figsize=(20,10))  
2 plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['Yes','No'],fi  
3
```

```
Out[628]: [Text(558.0, 407.70000000000005, 'Index <= 4.5\nngini = 0.498\nnsamples = 47\nnvalue = [37,  
33]\nnclass = Yes'),  
          Text(279.0, 135.89999999999998, 'gini = 0.473\nnsamples = 26\nnvalue = [24, 15]\nnclass =  
Yes'),  
          Text(837.0, 135.89999999999998, 'gini = 0.487\nnsamples = 21\nnvalue = [13, 18]\nnclass =  
No')]
```

