

```
In [1]: import numpy as np  
import pandas as pd
```

```
In [2]: from sklearn.datasets import load_digits  
digits=load_digits()
```

```
In [3]: digits
```

```
Out[3]: {'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',
                           'pixel_1_2',
                           'pixel_1_3',
                           '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',
                           'pixel_4_0',
                           'pixel_4_1',
                           'pixel_4_2',
                           'pixel_4_3',
                           'pixel_4_4',
                           'pixel_4_5',
                           'pixel_4_6',
                           'pixel_4_7',
                           'pixel_5_0',
                           'pixel_5_1',
                           'pixel_5_2',
                           'pixel_5_3',
                           'pixel_5_4',
                           'pixel_5_5',
                           'pixel_5_6',
                           'pixel_5_7',
                           'pixel_6_0',
                           'pixel_6_1',
                           'pixel_6_2',
                           'pixel_6_3',
                           'pixel_6_4',
                           'pixel_6_5',
                           'pixel_6_6',
```

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'pixel_6_7',
'pixel_7_0',
'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.],
 [ 0.,  2., 14., ..., 12.,  0.,  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.],
 ...,
 [ 0.,  0.,  1., ...,  6.,  0.,  0.],
 [ 0.,  0.,  1., ...,  6.,  0.,  0.],
 [ 0.,  0.,  0., ..., 10.,  0.,  0.]],

 [[ 0.,  0.,  0., ..., 12.,  0.,  0.],
 [ 0.,  0.,  3., ..., 14.,  0.,  0.],
 [ 0.,  0.,  8., ..., 16.,  0.,  0.],
 ...,
 [ 0.,  9., 16., ...,  0.,  0.,  0.],
 [ 0.,  3., 13., ..., 11.,  5.,  0.],
 [ 0.,  0.,  0., ..., 16.,  9.,  0.]],

 ...,

 [[ 0.,  0.,  1., ...,  1.,  0.,  0.],
 [ 0.,  0., 13., ...,  2.,  1.,  0.],
 [ 0.,  0., 16., ..., 16.,  5.,  0.],
 ...,
 [ 0.,  0., 16., ..., 15.,  0.,  0.],
 [ 0.,  0., 15., ..., 16.,  0.,  0.],
 [ 0.,  0.,  2., ...,  6.,  0.,  0.]],

 [[ 0.,  0.,  2., ...,  0.,  0.,  0.],
 [ 0.,  0., 14., ..., 15.,  1.,  0.],
 [ 0.,  4., 16., ..., 16.,  7.,  0.],
 ...,
 [ 0.,  0.,  0., ..., 16.,  2.,  0.],
 [ 0.,  0.,  4., ..., 16.,  2.,  0.],
 [ 0.,  0.,  5., ..., 12.,  0.,  0.]],

 [[ 0.,  0., 10., ...,  1.,  0.,  0.],
 [ 0.,  2., 16., ...,  1.,  0.,  0.],
 [ 0.,  0., 15., ..., 15.,  0.,  0.],
 ...,
 [ 0.,  4., 16., ..., 16.,  6.,  0.],
 [ 0.,  8., 16., ..., 16.,  8.,  0.],
 [ 0.,  1.,  8., ..., 12.,  1.,  0.])),
'DESCR': ".. _digits_dataset:\n\nOptical recognition of handwritten digits datase
t\n-----\n\n**Data Set Characteristic
s:**\n\n      :Number of Instances: 1797\n      :Number of Attributes: 64\n      :Attrib
ute Information: 8x8 image of integer pixels in the range 0..16.\n      :Missing Att
ribute Values: None\n      :Creator: E. Alpaydin (alpaydin '@' boun.edu.tr)\n      :Da

```

te: July; 1998\n\nThis is a copy of the test set of the UCI ML hand-written digits datasets\nhttps://archive.ics.uci.edu/ml/datasets/Optical+Recognition+of+Handwritten+Digits\n\nThe data set contains images of hand-written digits: 10 classes where each class refers to a digit.\n\nPreprocessing programs made available by NIST were used to extract\nnormalized bitmaps of handwritten digits from a preprinted form. From a total of 43 people, 30 contributed to the training set and different 13 to the test set. 32x32 bitmaps are divided into nonoverlapping blocks of 4x4 and the number of on pixels are counted in each block. This generates an input matrix of 8x8 where each element is an integer in the range 0..16. This reduces dimensionality and gives invariance to small distortions.\n\nFor info on NIST preprocessing routines, see M. D. Garriss, J. L. Blue, G. T. Candela, D. L. Dimmick, J. Geist, P. J. Grother, S. A. Janet, and C. L. Wilson, NIST Form-Based Handprint Recognition System, NISTIR 5469, 1994.\n\n.. topic:: References\n\n - C. Kaynak (1995) Methods of Combining Multiple Classifiers and Their Applications to Handwritten Digit Recognition, MSc Thesis, Institute of Graduate Studies in Science and Engineering, Bogazici University.\n - E. Alpaydin, C. Kaynak (1998) Cascading Classifiers, Kybernetika.\n - Ken Tang and Ponnuthurai N. Suganthan and Xi Yao and A. Kai Qin.\n Linear dimensionality reduction using relevance weighted LDA. School of Electrical and Electronic Engineering Nanyang Technological University. 2005.\n - Claudio Gentile. A New Approximate Maximal Margin Classification Algorithm. NIPS. 2000.\n"}

In [6]: `df=pd.DataFrame(digits.data,columns=digits.feature_names)`

In [7]: `df.head()`

Out[7]:

	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
0	0.0	0.0	5.0	13.0	9.0	1.0	0.0	0.0	0.0
1	0.0	0.0	0.0	12.0	13.0	5.0	0.0	0.0	0.0
2	0.0	0.0	0.0	4.0	15.0	12.0	0.0	0.0	0.0
3	0.0	0.0	7.0	15.0	13.0	1.0	0.0	0.0	0.0
4	0.0	0.0	0.0	1.0	11.0	0.0	0.0	0.0	0.0

5 rows × 64 columns

In [9]: `df.shape`

Out[9]: (1797, 64)

In [10]: `digits.target`

Out[10]: array([0, 1, 2, ..., 8, 9, 8])

In [11]: `df['digits']=digits.target`

In [12]: `df.head(2)`

Out[12]:

	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	digits
0	0.0	0.0	5.0	13.0	9.0	1.0	0.0	0.0	0.0	0
1	0.0	0.0	0.0	12.0	13.0	5.0	0.0	0.0	0.0	1

2 rows × 65 columns

```
In [16]: df['digits'].value_counts()
```

```
Out[16]: 3    183
          1    182
          5    182
          4    181
          6    181
          9    180
          7    179
          0    178
          2    177
          8    174
          Name: digits, dtype: int64
```

```
In [17]: df.isnull().sum().sum()
```

```
Out[17]: 0
```

```
In [18]: x=df.drop(['digits'],axis=1)
         y=df['digits']
```

```
In [20]: from sklearn.model_selection import train_test_split
```

```
In [21]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=1)
```

```
In [24]: from sklearn.ensemble import RandomForestClassifier
```

```
In [28]: rfc1=RandomForestClassifier(n_estimators=200,max_depth=2)
```

```
In [29]: rfc1.fit(x_train,y_train)
```

```
Out[29]: RandomForestClassifier(max_depth=2, n_estimators=200)
```

```
In [30]: y_predict=rfc1.predict(x_test)
```

```
In [33]: from sklearn.metrics import confusion_matrix,accuracy_score
```

```
In [34]: confusion_matrix(y_test,y_predict)
accuracy_score(y_test,y_predict)
rfc1.estimators_[0]
```

```
Out[34]: DecisionTreeClassifier(max_depth=2, max_features='auto', random_state=786542976)
```

```
In [35]: from sklearn import tree
```

```
In [36]: import matplotlib.pyplot as plt
```

```
In [41]: plt.figure(figsize=(25,5))
         tree.plot_tree(rfc1.estimators_[0])
         plt.show()
```



