Introducing Scikit-Learn

There are several Python libraries that provide solid implementations of a range of machine learning algorithms. One of the best known is Scikit-Learn, a package that provides efficient versions of a large number of common algorithms. Scikit-Learn is characterized by a clean, uniform, and streamlined API, as well as by very useful and complete online documentation. A benefit of this uniformity is that once you understand the basic use and syntax of Scikit-Learn for one type of model, switching to a new model or algorithm is very straightforward.

This section provides an overview of the Scikit-Learn API; a solid understanding of these API elements will form the foundation for understanding the deeper practical discussion of machine learning algorithms and approaches in the following chapters.

We will start by covering data representation in Scikit-Learn, followed by covering the Estimator API, and finally go through a more interesting example of using these tools for exploring a set of images of handwritten digits.

Data Representation in Scikit-Learn

Machine learning is about creating models from data: for that reason, we'll start by discussing how data can be represented in order to be understood by the computer. The best way to think about data within Scikit-Learn is in terms of tables of data.

Data as table

A basic table is a two-dimensional grid of data, in which the rows represent individual elements of the dataset, and the columns represent quantities related to each of these elements. For example, consider the Iris dataset, famously analyzed by Ronald Fisher in 1936. We can download this dataset in the form of a Pandas DataFrame using the Seaborn library:

```
In[1]: import seaborn as sns
      iris = sns.load dataset('iris')
      iris.head()
         sepal length sepal width petal length petal width species
Out[1]:
       0
                 5.1
                             3.5
                                  1.4
                                                      0.2 setosa
                 4.9
                             3.0
                                          1.4
                                                      0.2 setosa
       1
                 4.7
                             3.2
                                          1.3
                                                      0.2 setosa
       3
                 4.6
                             3.1
                                          1.5
                                                      0.2 setosa
                                                      0.2 setosa
                  5.0
                             3.6
                                          1.4
```

Here each row of the data refers to a single observed flower, and the number of rows is the total number of flowers in the dataset. In general, we will refer to the rows of the matrix as *samples*, and the number of rows as n samples.