

## How can we determine whether a problem is linear/nonlinear?

Before running any classifier just like that, we need to understand what is a linear classification problem. Look at the 2D data below. In data A, blue and green points can be easily classified by drawing a simple line. The line shown is one of the many solutions that are possible and you will get maximum accuracy. A problem is that, if the separating line is drawn improperly, then even in a simple linear case we can mess up. Therefore, proper optimization of the parameters to find an optimum line/plane/hyper-plane is very important. In data B we cannot draw a line that can classify both green and blue points. Therefore, data A represents a linear classification problem and data B represents a non-linear classification problem.

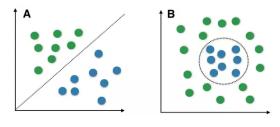


Figure 1

Let's get into details of linear and nonlinear classifiers.

## Linear Classifier

Let's say we have data from two classes (o and  $\chi$ ) distributed as shown in the figure below. To discriminate the two classes, one can draw an arbitrary line, s.t. all the 'o' are on one side of the line and  $\chi$ 's on the other side of the line. These two classes are called linearly-separable.

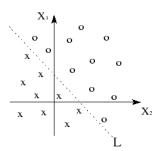


Figure 2



However, there are infinite number of lines that can be drawn to discriminate two separable classes, as shown below

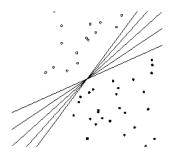


Figure 3

How you approximate the exact location of this discriminating line (or plane or hyperplane) depends on the type of a classifier called linear classifier. Some examples of linear classifier are: Linear Discriminant Classifier, Naive Bayes, Logistic Regression, Perceptron, SVM (with linear kernel).

## Non-linear Classifier

Let's consider the famous XOR problem. An XOR's truth table and the data distribution is shown below. In this situation, there is no way that a straight line can be drawn to discriminate the two classes (0 and 1).

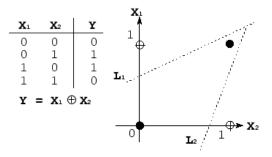


Figure 4

Another example is in the next page, where majority of points one one class can be discriminated from the other classes by drawing a straight line. But some points of that class are present in between the data points of second class (shown inside a circle). It is clear that drawing one straight line cannot identify all the points of one of the class correctly.

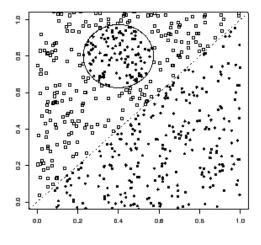


Figure 5

The above problems are called nonlinear classification problems and cannot be solved by drawing a linear classifier; therefore, other alternatives are required. We may need piece-wise linear (i.e. linear in parts), or non-linear classification boundaries to identify the two classes correctly. What would be that non-linear function (or approximation) and how that boundary may look like is defined by a non-linear classifier. Some examples of non-linear classifiers are: Quadratic Discriminant Classifier, Multi-layer Perceptron (most deep networks), Decision Trees, Random Forest, K-Nearest Neighbour.

## References:

For more details on Nonlinear Classification can refer below: http://www.moseslab.csb.utoronto.ca/alan/Lec6\_ML4bio\_2012.pdf