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# Chapter 1

## 1.1

### 1.1.1 Estimating class probabilities in multiclass classification via the softmax function

The softmax function is a soft form of the argmax function; instead of giving a single class index, it provides the probability of each class. Therefore, it allows us to compute meaningful class probabilities in multiclass settings (multinomial logistic regression).

In softmax, the probability of a particular sample with net input  $z$  belonging to the  $i$ th class can be computed with a normalization term in the denominator, that is, the sum of the exponentially weighted linear functions:

$$p(z) = \sigma(z) = \frac{e^{z_i}}{\sum_{j=1}^M e^{z_j}} \quad (1.1)$$

### 1.1.2 Broadening the output spectrum using a hyperbolic tangent

Another sigmoidal function that is often used in the hidden layers of artificial NNs is the hyperbolic tangent (commonly known as tanh), which can be interpreted as a rescaled version of the logistic function:

$$\begin{aligned} \sigma_{logistic}(z) &= \frac{1}{1 + e^{-z}} \\ \sigma_{tanh}(z) &= 2 \times \sigma_{logistic}(2z) - 1 = \frac{e^z - e^{-z}}{e^z + e^{-z}} \end{aligned} \quad (1.2)$$

The advantage of the hyperbolic tangent over the logistic function is that it has a broader output spectrum ranging in the open interval  $(-1, 1)$ , which can improve the convergence of the backpropagation algorithm.