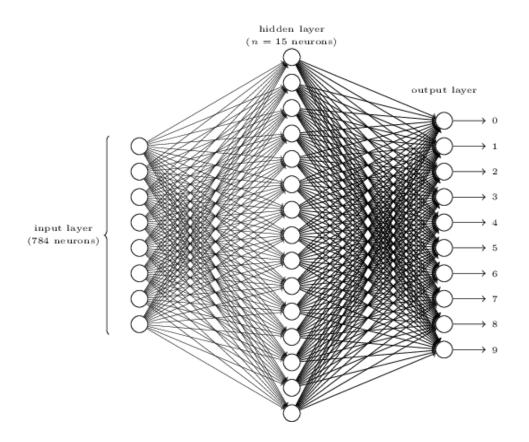
Machine Learning

Neural Networks



1. Basics

sigmoid function:

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

$$\sigma'(x) = \sigma(x) \cdot [1 - \sigma(x)]$$

hyperbolic function:

$$tanh(x) = \frac{e^{x} - e^{-x}}{e^{x} + e^{-x}}$$

$$\tanh'(x) = 1 - \tanh^2(x)$$

softmax function:

$$y = softmax(x)$$

$$y_i = \frac{e^{x_i}}{\sum_{j=1}^n e^{x_j}}$$

$$\frac{\partial y_i}{\partial x_j} = \begin{cases} -y_i \cdot y_j, & i \neq j \\ y_i \cdot (1 - y_i), & i = j \end{cases}$$

2. Model

input:

$$x \in \mathbb{R}^n$$

layer 1:

$$a^1 = x$$

layer l:

$$a^{l} = \sigma(w^{l}a^{l-1} + b^{l}) \quad (l = 2, ..., L)$$

layer L:

$$\hat{y} = a^L$$

output:

$$\hat{y} \in \mathbb{R}^m$$

3. Backpropagation

cost function:

$$C = C(\hat{y})$$

definition:

$$z^{l} = w^{l}a^{l-1} + b^{l} \quad (l = 2, ..., L)$$
$$\delta^{l} = \frac{\partial C}{\partial z^{l}} \quad (l = 2, ..., L)$$

output error δ^L :

$$\delta^{L} = \frac{\partial C}{\partial z^{L}} = \frac{\partial C}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial z^{L}}$$

$$= \frac{\partial C}{\partial a^{L}} \cdot \frac{\partial a^{L}}{\partial z^{L}}$$

$$= \frac{\partial C}{\partial a^{L}} \odot \sigma'(z^{L}) \quad (\text{need } a^{L}, z^{L})$$

backpropagate the error:

$$\delta^{l} = ((w^{l+1})^{T} \delta^{l+1}) \odot \sigma'(z^{l}) \quad (\text{need } z^{l}; l = L - 1, L - 2, \dots, 2)$$

output:

$$\frac{\partial C}{\partial b^{l}} = \delta^{l} \quad (l = L, L - 1, \dots, 2)$$

$$\frac{\partial C}{\partial w^{l}} = \delta^{l} \cdot (a^{l-1})^{T} \quad (\text{need } a^{l-1}; l = L, L - 1, \dots, 2)$$