

Mathematics and Artificial Neural Network

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1. Neural Block

Let's consider T function

$$T(x) = f(Wx + b), \quad (1.1)$$

where

$$W = \begin{pmatrix} w_{11} & w_{12} & \cdots & w_{1n} \\ w_{21} & w_{22} & \cdots & w_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ w_{m1} & w_{m2} & \cdots & w_{mn} \end{pmatrix}, \quad b = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}, \quad x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix}, \quad T = \begin{pmatrix} t_1 \\ t_2 \\ \vdots \\ t_m \end{pmatrix}.$$
$$t_i = f(y_i) = f\left(\sum_{k=1}^m w_{ik}x_k + b_i\right), \quad \frac{\partial t_i}{\partial w_{ik}} = x_k f'(y_i), \quad \frac{\partial t_i}{\partial b_i} = f'(y_i), \quad \frac{\partial y_i}{\partial x_k} = w_{ik}.$$

Let's define loss function L as

$$L(\hat{t}; t) = L(\hat{t}_1, \dots, \hat{t}_m; t_1, \dots, t_m) = \frac{1}{2} \sum_{i=1}^m (\hat{t}_i - t_i)^2. \quad (1.2)$$

where t is original vector and \hat{t} is computed using (1.1).

$$\delta w_{ij} = \frac{\partial L}{\partial w_{ij}} = \sum_{i=1}^m (\hat{t}_i - t_i) \frac{\partial \hat{t}_i}{\partial w_{ij}} = \sum_{i=1}^m (f(y_i) - t_i) x_j f'(y_i),$$
$$\delta b_i = \frac{\partial L}{\partial b_i} = \sum_{i=1}^m (\hat{t}_i - t_i) \frac{\partial \hat{t}_i}{\partial b_i} = \sum_{i=1}^m (f(y_i) - t_i) f'(y_i).$$

If we consider x_k to also change, thans

$$\delta x_k = \frac{\partial L}{\partial x_k} = \sum_{i=1}^m \frac{\partial y_i}{\partial x_k} \frac{\partial \hat{t}_i}{\partial y_i} \frac{\partial L}{\partial \hat{t}_i} = \sum_{i=1}^m w_{ik} f'(y_i) (f(y_i) - t_i)$$

Therefore, we obtain following

$$w_{ij} \leftarrow w_{ij} - \gamma \delta w_{ij}, \quad (1.3)$$

$$b_i \leftarrow b_i - \gamma \delta b_i, \quad (1.4)$$

$$x_i \leftarrow x_i - \gamma \delta x_i, \quad (1.5)$$

where γ is some small positive constant.

2. Neural Chain

Let's take T , T_1 and T_2 as

$$T_1(x) = f_1(W_1x + b_1), \quad T_2(x) = f_2(W_2x + b_2), \quad T = T_1 \circ T_2$$