Mathematics of a Deep Neural Network

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Note

This is intended to act as a reference to use while writing code; not as a comprehensive guide to how neural networks work. The goal is only to describe the mathematical operations required to implement stochastic gradient descent in CUDA.

Variable	Value
$\overline{S_n}$	Height of the <i>n</i> th layer
\overline{L}	Number of layers in the network. Does not include input vector
$\overline{S_L}$	Number of neurons in the last layer. Equal to number of outputs
$\overline{S_1}$	Number of neurons in the first layer
$\overline{S_0}$	Number of inputs to the network. Just a convention; there is no "0th layer",
	and the inputs are not part of the network shape
\overline{C}	Cost function. Mean squared error is used exclusively in this paper
σ_n	Activation function of the nth layer (any differentiable function - not just
	sigmoid)
w_{nij}	jth weight of the i th neuron in the n th layer. j corresponds to a neuron in
	the previous layer, or in the input vector if $i = 1$
b_{ni}	The bias of i th neuron in the n th layer
O_{ni}	Output of the i th neuron in the n th layer, prior to activation
a_{ni}	Output of the i th neuron in the n th layer, after activation
a_{0i}	The <i>i</i> th input value. Just a convention; like S_0
T_i	The <i>i</i> th target value. In a perfectly trained network, this is equal to a_{Li}

$$o_{ni} = \left[\sum_{j=1}^{S_{n-1}} w_{nij} \cdot a_{(n-1)j}\right] + b_{ni}$$
$$a_{ni} = \sigma_n(o_{ni})$$

$$C = \frac{1}{S_L} \sum_{i=1}^{S_L} (a_{Li} - T_i)^2$$

$$\frac{\partial C}{\partial a_{Li}} = 2 \cdot \frac{1}{S_L} \cdot (a_{Li} - T_i)$$

$$\frac{\partial C}{\partial o_{Li}} = \frac{\partial C}{\partial a_{Li}} \cdot \frac{\partial a_{Li}}{\partial o_{Li}} = \frac{\partial C}{\partial a_{Li}} \cdot \sigma'_L(o_{Li})$$

$$\frac{\partial C}{\partial a_{ni}} = \sum_{j=1}^{S_{n+1}} \frac{\partial C}{\partial o_{(n+1)j}} \cdot w_{(n+1)ji}$$
$$\frac{\partial C}{\partial o_{ni}} = \frac{\partial C}{\partial a_{ni}} \cdot \frac{\partial a_{ni}}{\partial o_{ni}} = \frac{\partial C}{\partial a_{ni}} \cdot \sigma'_{n}(o_{ni})$$

$$\frac{\partial C}{\partial w_{nij}} = \frac{\partial C}{\partial o_{ni}} \cdot \frac{\partial o_{ni}}{\partial w_{nij}} = \frac{\partial C}{\partial o_{ni}} \cdot a_{(n-1)j}$$
$$\frac{\partial C}{\partial b_{ni}} = \frac{\partial C}{\partial o_{ni}} \cdot \frac{\partial o_{ni}}{\partial b_{ni}} = \frac{\partial C}{\partial o_{ni}} \cdot 1 = \frac{\partial C}{\partial o_{ni}}$$