## **Backpropagation Implementation Helper Formulas**

## **Notations**

- ullet  $b_{j}^{l}$  for the bias of the  $j^{th}$  neuron in the  $l^{th}$  layer
- $w^l_{jk}$  for the weight for the connection from the  $k^{th}$  neuron in the  $(l-1)^{th}$  layer to the  $j^{th}$  neuron in the  $l^{th}$  layer
- $\bullet \hspace{0.2cm} z_j^l$  for the weighted input to the activation function for neuron j in layer l
- ullet  $a_j^l$  for the activation of the  $j^{th}$  neuron in the  $l^{th}$  layer
- ullet  $\sigma$  for the sigmoid function

## **Formulas**

1. the activation  $a^l_j$  of the  $j^{th}$  neuron in the  $l^{th}$  layer is related to the activations in the  $(l-1)^{th}$  layer

$$a_j^l = \sigma \left( \sum_k w_{jk}^l a_k^{l-1} + b_j^l 
ight)$$

which is equivalent to

$$a^l = \sigma(w^l a^{l-1} + b^l)$$

2. the cost function

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$$C = rac{1}{2}\|y - a^L\|^2 = rac{1}{2}\sum_j (y_j - a_j^L)^2$$

3. the gradient  $\delta^l_j$  of neuron j in layer l

$$\delta_j^l = rac{\partial C}{\partial z_j^l}$$

4. the gradient in the output layer,  $\delta^L$ 

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\begin{eqnarray}
\delta^L_j = \frac{\partial C}{\partial z^L_j} = \frac{\partial C}{\partial a^L_j} \sigma'(z^L_j)
= (a_j^L-y_j)\sigma'(z^L_j)
\end{eqnarray}
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$$\delta_j^L = rac{\partial C}{\partial z_i^L} = rac{\partial C}{\partial a_i^L} \sigma'(z_j^L) = (a_j^L - y_j) \sigma'(z_j^L)$$

which is equivalent to

$$\delta^L = (a^L - y) \odot \sigma'(z^L)$$

5. the gradient  $\delta^l$  in terms of the gradient in the next layer,  $\delta^{l+1}$  , where l < L

$$\delta^l = ((w^{l+1})^T \delta^{l+1}) \odot \sigma'(z^l)$$

Proof

$$\begin{split} \delta_j^l &= \frac{\partial C}{\partial z_j^l} \\ &= \sum_k \frac{\partial C}{\partial z_k^{l+1}} \frac{\partial z_k^{l+1}}{\partial z_j^l} \\ &= \sum_k \frac{\partial z_k^{l+1}}{\partial z_j^l} \delta_k^{l+1} \end{split}$$

where

$$z_k^{l+1} = \sum_j w_{kj}^{l+1} a_j^l + b_k^{l+1} = \sum_j w_{kj}^{l+1} \sigma(z_j^l) + b_k^{l+1}$$

SO

$$rac{\partial z_k^{l+1}}{\partial z_j^l} = w_{kj}^{l+1} \sigma'(z_j^l)$$

then we have

$$\delta_j^l = \sum_k w_{kj}^{l+1} \delta_k^{l+1} \sigma'(z_j^l)$$

6. the gradient of the bias  $b^l$  in layer l

$$\frac{\partial C}{\partial b_j^l} = \frac{\partial C}{\partial z_j^l} \frac{\partial z_j^l}{\partial b_j^l} = \frac{\partial C}{\partial z_j^l} \frac{\partial (w^l a^{l-1} + b_j^l)}{\partial b_j^l} = \delta_j^l$$

which is equivalent to

$$\frac{\partial C}{\partial b^l} = \delta^l$$

7. the gradient of the weight  $w^l$  in layer l

$$rac{\partial C}{\partial w_{jk}^l} = a_k^{l-1} \delta_j^l$$

which is equivalent to

$$\frac{\partial C}{\partial w^l} = a^{l-1} \delta^l$$