## CS224N PA4: Neural Networks for Named Entity Recognition

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## 1 Gradients for Backpropagation

Let  $w_{jk}^l$  denote the weight for connecting the  $k^{th}$  neuron in the  $(l-1)^{th}$  layer to the  $j^{th}$  neuron in the  $l^{th}$  layer;  $b_j^l$  denote the bias for the  $j^{th}$  neuron in the  $l^{th}$  layer;  $a_j^l$  denote the activation of the  $j^{th}$  neuron in the  $l^{th}$  layer;  $z_j^l$  denote the weighted input to the  $j^{th}$  neuron in the  $l^{th}$  layer;  $h_l(.)$  denote the activation function for the weighted input  $\mathbf{z}_l$ . Note that  $z_j^l = \sum_i w_{ji}^l a_i^{l-1} + b_j^l$  and  $a_j^l = h_l(z_j^l)$ . Let's define  $\delta_j^l = \frac{\partial J}{\partial z_j^l}$ , the error of neuron j in layer l. Then it can be easily derived that the following four equations are true for any backpropagation system:

$$\delta^L = \frac{\partial J}{\partial a^L} \odot h_L'(z^L) \tag{1a}$$

$$\delta^l = (W^{l+1})^T \delta^{l+1} \odot h'_l(z^l) \tag{1b}$$

$$\frac{\partial J}{\partial b_j^l} = \delta_j^l \tag{1c}$$

$$\frac{\partial J}{\partial w_{ik}^l} = a_k^{l-1} \delta_j^l \tag{1d}$$

For current system, we've three layers: input layer, hidden layer and output layer. The cost function is  $J = -[ylna^L + (1-y)ln(1-a^L)]$ . Using the above general system, we can obtain the following:

$$\delta^3 = p_\theta - y \tag{2a}$$

$$\delta^{2} = (W^{3})^{T} \delta^{3} \odot \tanh'(z^{2}) = U^{T}(p_{\theta} - y) \odot \tanh'(Wx + b^{(1)})$$
 (2b)

$$\delta^{1} = (W^{2})^{T} \delta^{2} \odot I'(x) = W^{T} \delta^{2} = W^{T} U^{T} \delta^{3} \odot \tanh'(Wx + b^{(1)})$$
 (2c)

And

$$\frac{\partial J}{\partial U} = a^2 \delta^3 = \tanh(Wx + b^{(1)})(p_\theta - y) \tag{3a}$$

$$\frac{\partial J}{\partial W} = a^1 \delta^2 = LU^T(p_\theta - y) \odot \tanh'(Wx + b^{(1)}) \tag{3b}$$

$$\frac{\partial J}{\partial b^{(2)}} = \delta^3 = p_\theta - y \tag{3c}$$

$$\frac{\partial J}{\partial b^{(1)}} = \delta^2 = U^T(p_\theta - y) \odot \tanh'(Wx + b^{(1)}) \tag{3d}$$

$$\frac{\partial J}{\partial L} = W^T U^T (p_\theta - y) \odot \tanh'(Wx + b^{(1)}) \tag{3e}$$