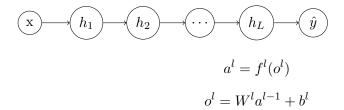
0.1 Notations

⊙ is the Hadamart productsss

1 Forward propagation



Therefore, if we omit the bias terms, we have that

$$y = f^{L}(W^{L}f^{L-1}(W^{L-1}(\ldots)))$$

2 Backward propagation

2.1 How to compute the gradient slice of w_i^L recursively

$$\begin{split} \nabla \frac{c}{w^l} &= (\sum_i \nabla \frac{a^l_i}{w^l}) [\nabla \frac{c}{a^l}]^T \odot \overrightarrow{\frac{\partial o^{L'}}{\partial w^L_i}} = \overrightarrow{\frac{\partial c}{\partial a^L}} \odot \overrightarrow{\frac{\partial a^{L'}}{\partial o^L}} \odot \overrightarrow{\frac{\partial o^{L'}}{\partial w^L_i}} \\ &\overrightarrow{\frac{\partial c}{\partial a^L}} = \overrightarrow{\frac{\partial o^{L+1}}{\partial a^L}} \sum_i \frac{\partial c}{\partial o^{L+1}_i} = \overrightarrow{\sum_j w^{L+1}_{i,j}} \sum_i \frac{\partial c}{\partial o^{L+1}_i} = \overrightarrow{\sum_k w^{L+1}_{\forall,k}} \frac{\partial c}{\partial o^{L+1}_k} \\ &\overrightarrow{\frac{\partial o^{L'}}{\partial w^L_i}} = a^{L-1} \end{split}$$

If we pose $\lambda=\frac{\partial c}{\partial o}$ and use the previous results we can obtain a compact recursive formulation of the gradient slice w_i^L

$$\frac{\overrightarrow{\partial c}}{\partial w_i^L} = \lambda^L a^{L-1},$$