

# Backpropagation

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11:04 AM

## Forward Propagation

$$\begin{aligned} \text{Input } X \in \mathbb{R}^D &\rightsquigarrow Z_1 = W_1 X + b_1 \rightsquigarrow A_1 = \sigma(Z_1) \\ &\quad \downarrow \\ Z_2 &:= W_2 A_1 + b_2 \\ &\quad \downarrow \\ A_2 &:= \text{Softmax}(Z_2) \end{aligned}$$

$W_1$  is  $h \times D$  matrix

$h$  is # of hidden nodes

$b_1, b_2 \in \mathbb{R}$

$W_2$  is  $O \times h$  matrix

$O$  is # of output nodes

$\sigma(x) = \frac{1}{1+e^{-x}}$  sigmoid function

$$\text{Softmax}\left(\begin{pmatrix} x_1 \\ \vdots \\ x_k \end{pmatrix}\right) = \frac{1}{\sum e^{x_i}} (e^{x_1}, \dots, e^{x_k})$$

We use CROSS ENTROPY LOSS

$$\mathcal{L}(y, \hat{y}) = -\sum y_i \log \hat{y}_i$$

## BACKPROPAGATION

$$\mathcal{L}(y, \hat{y}) = \mathcal{L}(y, A_2) = \mathcal{L}(y, \text{Softmax}(Z_2)) =$$

$$= -\sum y_i \log\left(\frac{e^{z_{2,i}}}{\sum_j e^{z_{2,j}}}\right) = -\sum y_i (z_{2,i} - \log(\sum_j e^{z_{2,j}})) =$$

$$= \log(\sum_j e^{z_{2,j}}) - \sum y_i z_{2,i} \quad (\text{we used } \sum y_i = 1).$$

$$\bullet \frac{\partial L}{\partial z_{2,i}} = \frac{e^{z_{2,i}}}{\sum_j e^{z_{2,j}}} - y_i = \hat{y}_i - y_i \quad \frac{\partial L}{\partial z_2} = \hat{y} - y$$

$$\bullet \frac{\partial L}{\partial A_1} = \frac{\partial L}{\partial z_2} \cdot \frac{\partial z_2}{\partial A_1} = \frac{\partial L}{\partial z_2} \cdot W_2$$

$$\bullet \frac{\partial L}{\partial b_2} = \frac{\partial L}{\partial z_2} \cdot \frac{\partial z_2}{\partial b_2} = \frac{\partial L}{\partial z_2} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \text{sum of components of } \frac{\partial L}{\partial z_2}$$

$$\bullet \frac{\partial L}{\partial z_1} = \frac{\partial L}{\partial A_1} \cdot \frac{\partial A_1}{\partial z_1} = \frac{\partial L}{\partial A_1} \cdot \left( \text{matrix } Z_1, \text{ but apply } \frac{d\sigma}{dx} \text{ to every element} \right)$$

$$\frac{d\sigma}{dx}(x) = \frac{e^{-x}}{(1+e^{-x})^2} \quad \leftarrow \text{sigmoid derivative}$$

$$\bullet \frac{\partial L}{\partial A_1} = \frac{\partial L}{\partial z_1} \cdot \frac{\partial z_1}{\partial A_1} = \frac{\partial L}{\partial z_1} \cdot W_1$$

$$\bullet \frac{\partial L}{\partial b_1} = \frac{\partial L}{\partial z_1} \cdot \frac{\partial z_1}{\partial b_1} = \frac{\partial L}{\partial z_1} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \text{sum of elements of } \frac{\partial L}{\partial z_1}$$