

## ReLU

$$f(x) = x^+ = \max(x, 0) \quad (1)$$

## Softmax

$$f(x_i) = \frac{\exp(x_i)}{\sum_j \exp(x_j)} \quad (2)$$

## Forward Pass

Let  $(x_1, y_1, \dots, (x_{n_b}, y_{n_b}))$  be the data in a mini-batch  $\mathcal{D}^{(t)}$ , where  $X \in R^{d \times n}$  and  $Y \in R^{o \times n}$ .

for  $i = 0$

$$H^0 = \text{ReLU}(W_i X_{batch} + b_i 1_{n_b}^T, 0) \quad (3)$$

for  $i = 1, \dots, k - 1$

$$H^i = \text{ReLU}(W_i H^{(i-1)} + b_i 1_{n_b}^T, 0) \quad (4)$$

Then,

$$P_{batch} = \text{Softmax}(W_k H^{(k-1)} + b_k 1_{n_b}^T) \quad (5)$$

## Backward Pass

$$G_{batch} = -(Y_{batch} - P_{batch}) \quad (6)$$

for  $l = k, k - 1, \dots, 2$

$$\frac{\partial L}{\partial W_l} = \frac{1}{n_b} G_{batch} H^{(l-1)T} \quad (7)$$

$$\frac{\partial L}{\partial b_l} = \frac{1}{n_b} G_{batch} 1_{n_b} \quad (8)$$

$$G_{batch} = W_l^T G_{batch} \quad (9)$$

$$G_{batch} = G_{batch} \odot \text{Ind}(X_{batch}^{l-1} > 0) \quad (10)$$

Then,

$$\frac{\partial L}{\partial W_1} = \frac{1}{n_b} G_{batch} X_{batch}^T \quad (11)$$

$$\frac{\partial L}{\partial b_1} = \frac{1}{n_b} G_{batch} 1_{n_b} \quad (12)$$