

1.

(a) Recall the softmax function  $y_k = \frac{\exp(a_k)}{\sum_j \exp(a_j)}$  Find  $\frac{\partial y_k}{\partial a_j}$   $\square$

If  $k = j$ ,

$$\begin{aligned} \frac{\partial y_k}{\partial a_k} &= \frac{\exp(a_k) \cdot \sum_j \exp(a_j) - \exp(a_k) \cdot \exp(a_k)}{\left( \sum_j \exp(a_j) \right)^2} \\ &= \frac{\exp(a_k)}{\sum_j \exp(a_j)} - \left[ \frac{\exp(a_k)}{\sum_j \exp(a_j)} \right]^2 = y_k - y_k^2 \end{aligned}$$

If  $k \neq j$ ,

$$\frac{\partial y_k}{\partial a_j} = \frac{-\exp(a_k) \cdot \exp(a_j)}{\left( \sum_j \exp(a_j) \right)^2} = - \left( \frac{\exp(a_k)}{\sum_j \exp(a_j)} \right) \left( \frac{\exp(a_j)}{\sum_j \exp(a_j)} \right) = -y_k y_j$$

Observe that

$$y_k (\delta_{kj} - y_j) = \begin{cases} y_k (1 - y_k) & \text{if } k = j \\ -y_k y_j & \text{if } k \neq j \end{cases}$$

Hence,  $\frac{\partial y_k}{\partial a_k} = y_k (\delta_{kj} - y_j)$

$\square$

(b). Recall the cross-entropy error function is

$$EW = - \sum_{i=1}^n \sum_{k=1}^c t_{ik} \ln y_{ik}$$

Since  $y_k(\vec{x}_i) = \frac{\exp(a_{ik})}{\sum_j \exp(a_{ij})}$ , from prob. (a),  $\frac{\partial y_{ik}}{\partial a_{ij}} = y_{ik} (\delta_{kj} - y_{ij})$

Differentiating  $EW$  w.r.t  $a_{ij}$  using chain rule,

$$\frac{\partial E}{\partial a_{ij}} = \sum_k \frac{\partial y_{ik}}{\partial a_{ij}} \frac{\partial E}{\partial y_{ik}} = - \sum_k y_{ik} (\delta_{kj} - y_{ij}) \times \frac{t_{ik}}{y_{ik}}$$

$$= - \sum_k \delta_{kj} t_{ik} + \sum_k y_{ij} t_{ik}$$

$$= - t_{ij} + y_{ij} \sum_k t_{ik} = y_{ij} - t_{ij} \quad \left( \because \vec{x}_i = (0, 0, \dots, 1, 0, 0, \dots, 0) \right)$$

$\Rightarrow \sum_k t_{ik} = 1$

Differentiating w.r.t.  $w_j$  using chain rule again,

$$\frac{\partial E}{\partial w_j} = \sum_i \frac{\partial a_{ij}}{\partial w_j} \frac{\partial E}{\partial a_{ij}} = \sum_{i=1}^n \vec{x}_i (y_{ij} - t_{ij}) \quad \left( \because a_{ij} = w_j^T \vec{x}_i \right)$$

$\frac{\partial a_{ij}}{\partial w_j} = \vec{x}_i$

So the batch learning rule is

$$\vec{w}_j \leftarrow \vec{w}_j + \eta \sum_{i=1}^n (y_{ij} - t_{ij}) \vec{x}_i$$

And the single-sample rule is

$$\vec{w}_j \leftarrow \vec{w}_j + \eta (y_{ij} - t_{ij}) \vec{x}_i$$

where  $\eta$  is small enough.