

2.

Let $a_i(x) = w_i^T x + b_i$, $\theta = \{w_i, b_i\}$,
 $i = 1, \dots, c$

$$\begin{aligned}
 \underset{\substack{\downarrow \\ \text{likelihood}}}{\mathcal{L}(\theta)} &= \prod_{i=1}^m p(y^{(i)} | x^{(i)}, \theta) \\
 &= \prod_{i=1}^m \text{softmax}_{y^{(i)}}(x^{(i)}) \\
 &= \prod_{i=1}^m \frac{e^{a_{y^{(i)}}(x^{(i)})}}{\sum_{j=1}^c e^{a_j(x^{(i)})}}
 \end{aligned}$$

$$\Rightarrow \log \mathcal{L}(\theta) = \sum_{i=1}^m \log \left[\frac{e^{a_{y^{(i)}}(x^{(i)})}}{\sum_{j=1}^c e^{a_j(x^{(i)})}} \right]$$

$$= \sum_{i=1}^m \left[a_{y^{(i)}}(x^{(i)}) - \log \left(\sum_{j=1}^c e^{a_j(x^{(i)})} \right) \right] \quad \#$$

$$\Rightarrow \underset{\theta}{\operatorname{argmax}} \log \mathcal{L}(\theta) = \underset{\theta}{\operatorname{argmin}} \underbrace{-\log \mathcal{L}(\theta)}$$

\downarrow negative log-likelihood
 we want in this question (L)

Let $J(\theta) = -\log L(\theta)$

$$\nabla_{w_i} J(\theta) = \sum_{k=1}^m \frac{e^{a_i(x^{(k)})}}{\sum_{j=1}^c e^{a_j(x^{(k)})}} x^{(k)} - \sum_{\{k: y^{(k)}=i, \forall k \in \{1, \dots, m\}\}} x^{(k)}$$

$$= \sum_{k=1}^m \text{softmax}_i(x^{(k)}) x^{(k)} - \sum_{\{k: y^{(k)}=i, \forall k \in \{1, \dots, m\}\}} x^{(k)}$$

$$\nabla_{b_i} J(\theta) = \sum_{k=1}^m \frac{e^{a_i(x^{(k)})}}{\sum_{j=1}^c e^{a_j(x^{(k)})}} - \sum_{\{k: y^{(k)}=i, \forall k \in \{1, \dots, m\}\}} 1$$

$$= \sum_{k=1}^m \text{softmax}_i(x^{(k)}) - \sum_{\{k: y^{(k)}=i, \forall k \in \{1, \dots, m\}\}} 1$$

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