

# CS 224N Assignment #1.

## 1. Softmax

$$(a) \quad \text{softmax}(x)_i = \frac{e^{x_i}}{\sum_j e^{x_j}}$$

$$\text{softmax}(x_i + c) = \frac{e^{x_i + c}}{\sum_j e^{x_j + c}} = \frac{e^c \cdot e^{x_i}}{e^c \cdot \sum_j e^{x_j}} = \frac{e^{x_i}}{\sum_j e^{x_j}} = \text{softmax}(x_i).$$

(b). CODE.

## 2. Neural Network Basics.

$$(a) \quad \sigma(x) = \frac{1}{1 + e^{-x}} \Rightarrow \frac{d}{dx} \sigma(x) = \frac{-(-e^{-x})}{(1 + e^{-x})^2} = (1 - \sigma(x)) \sigma(x).$$

(b)

$$\hat{y} = \text{softmax}(\theta)$$

$$\frac{\partial \mathcal{L}(y, \hat{y})}{\partial \theta} = \frac{\partial \mathcal{L}(y, \hat{y})}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial \theta}.$$

The first step is to calculate the derivative of softmax() function.

$$\begin{aligned} \hat{y}_i &= \frac{e^{\theta_i}}{\sum_k e^{\theta_k}} & \frac{\partial \hat{y}_i}{\partial \theta_i} &= \frac{\partial}{\partial \theta_i} \frac{1}{1 + A e^{-\theta_i}} \quad (\text{where } A = \sum_k e^{\theta_k} - e^{\theta_i}) \\ & & &= \frac{1}{(1 + A e^{-\theta_i})^2} \left( 1 - \frac{1}{1 + A e^{-\theta_i}} \right) = \hat{y}_i (1 - \hat{y}_i). \\ & & \frac{\partial \hat{y}_i}{\partial \theta_j} &= \frac{\partial}{\partial \theta_j} \frac{e^{\theta_i}}{A + e^{\theta_i}} \quad (\text{where } A = \sum_k e^{\theta_k} - e^{\theta_j}) \\ & & &= \frac{e^{\theta_i}}{A + e^{\theta_i}} \cdot \frac{-e^{\theta_j}}{A + e^{\theta_j}} = -\hat{y}_i \hat{y}_j \end{aligned}$$

$$\begin{aligned} \frac{\partial \mathcal{L}(y, \hat{y})}{\partial \theta_i} &= - \sum_k y_k \cdot \frac{\partial (\log \hat{y}_k)}{\partial \theta_i} = - \sum_k y_k \cdot \frac{1}{\hat{y}_k} \frac{\partial \hat{y}_k}{\partial \theta_i} \\ &= -(y_i \cdot \hat{y}_i \cdot \frac{1}{\hat{y}_i} (1 - \hat{y}_i) + \sum_{k \neq i} y_k \cdot \frac{1}{\hat{y}_k} (-\hat{y}_i \hat{y}_k)) \Rightarrow \frac{\partial \mathcal{L}(y, \hat{y})}{\partial \theta} = \hat{y} - y \\ &= -y_i + \sum_k \hat{y}_i y_k = (\hat{y}_i - y_i) \star \end{aligned}$$