

a). Because y is a hot vectors with 0 if $w \neq 0$ and 1 if $w=0$.

$$\text{Therefore } -\sum_{w \in \text{Vocab}} y_w \log(\hat{y}_w) = -(y_1 \log(\hat{y}_1) + y_2 \log(\hat{y}_2) + \dots + y_0 \log(\hat{y}_0) + \dots)$$

$$= -y_0 \log(\hat{y}_0)$$

$$\because y_0 = 1 \text{ when } w=0$$

$$\therefore = -\log(\hat{y}_0)$$

$$b). \text{Naive-softmax}(V_0, o, U) = -\log \frac{\exp(U_0^T V_c)}{\sum_{w \in \text{Vocab}} \exp(U_w^T V_c)}$$

$$\frac{\partial J}{\partial V_c} = -\frac{\partial}{\partial V_c} (\log(\exp(U_0^T V_c)) - \log(\sum_{w \in \text{Vocab}} \exp(U_w^T V_c)))$$

$$= -\frac{\partial}{\partial V_c} (U_0^T V_c) + \frac{1}{\sum_{w \in \text{Vocab}} \exp(U_w^T V_c)} \cdot \sum_{x \in \text{Vocab}} U_x \cdot \exp(U_x^T V_c)$$

$$= -U_0 + \sum_{x \in \text{Vocab}} \frac{\exp(U_x^T V_c)}{\sum_{w \in \text{Vocab}} \exp(U_w^T V_c)} \cdot U_x$$

$$= -U_0 + \sum_{x \in \text{Vocab}} P(U_x | V_c) \cdot U_x$$

$$= -U_0 + \sum_{x \in \text{Vocab}} \hat{y}_x \cdot U_x$$

c). case 1: $w=0$

$$\frac{\partial J}{\partial U_{w=0}} = -\frac{\partial}{\partial U_{w=0}} (\log(\exp(U_0^T V_c)) - \log(\sum_{w \in \text{Vocab}} (\exp(U_w^T V_c))))$$

$$= -\frac{\partial}{\partial U_{w=0}} (U_0^T V_c) + \frac{\partial}{\partial U_{w=0}} (\log(\sum_{w \in \text{Vocab}} (\exp(U_w^T V_c))))$$

$$= -V_c + \frac{\exp(U_0^T V_c)}{\sum_{w \in \text{Vocab}} (\exp(U_w^T V_c))} V_c$$

$$= -V_c + \hat{y}_0 V_c$$

$$= V_c \cdot (\hat{y}_0 - 1)$$

case 2: $w \neq 0$

$$\frac{\partial J}{\partial U_{w \neq 0}} = -\frac{\partial}{\partial U_{w \neq 0}} (\log(\exp(U_w^T V_c)) - \log(\sum_{w \in \text{Vocab}} (\exp(U_w^T V_c))))$$

$$= 0 - \frac{\exp(\mathbf{w}_{\#0}^T \mathbf{v}_c)}{\sum_{w \in \text{vocab}} (\exp(\mathbf{w}^T \mathbf{v}_c))} \cdot \mathbf{v}_c$$

$$= -\mathbf{w}_{\#0} \cdot \mathbf{v}_c$$

$$d). \quad \frac{\partial \sigma(x)}{\partial x} = \frac{1}{(1+e^x)^2} \cdot -e^{-x}$$

$$= -\frac{e^{-x}}{(1+e^x)^2}$$

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

$$= \sigma(x) \cdot (1 - \sigma(x))$$

$$e). \quad 1. \quad \frac{\partial J}{\partial \mathbf{v}_c} = -\frac{1}{\sigma(\mathbf{u}_0^T \mathbf{v}_c)} \cdot (\sigma(\mathbf{u}_0^T \mathbf{v}_c) - (1 - \sigma(\mathbf{u}_0^T \mathbf{v}_c))) \cdot \mathbf{u}_0$$

$$+ \sum_{k=1}^K \frac{1}{\sigma(\mathbf{u}_k^T \mathbf{v}_c)} \cdot (\sigma(\mathbf{u}_k^T \mathbf{v}_c) - (1 - \sigma(\mathbf{u}_k^T \mathbf{v}_c))) \cdot \mathbf{u}_k$$

$$= -(1 - \sigma(\mathbf{u}_0^T \mathbf{v}_c)) \cdot \mathbf{u}_0 + \sum_{k=1}^K (\sigma(\mathbf{u}_k^T \mathbf{v}_c) - (1 - \sigma(\mathbf{u}_k^T \mathbf{v}_c))) \cdot \mathbf{u}_k$$

$$2. \quad \frac{\partial J}{\partial \mathbf{u}_0} = -\mathbf{v}_c \frac{1}{\sigma(\mathbf{u}_0^T \mathbf{v}_c)} \cdot (\sigma(\mathbf{u}_0^T \mathbf{v}_c) - (1 - \sigma(\mathbf{u}_0^T \mathbf{v}_c))) - 0$$

$$= -\mathbf{v}_c \cdot (1 - \sigma(\mathbf{u}_0^T \mathbf{v}_c))$$

$$3. \quad \frac{\partial J}{\partial \mathbf{u}_k} = \sum_{k=1}^K \mathbf{v}_c \cdot (\sigma(\mathbf{u}_k^T \mathbf{v}_c) - (1 - \sigma(\mathbf{u}_k^T \mathbf{v}_c)))$$

Instead of calculating all the words in the Vocab, we now only need to calculate k samples.

$$f). \quad i) \quad \frac{\partial \text{skip}(\mathbf{u}_c, \mathbf{u}_k, \dots, \mathbf{u}_{\text{term}} | \mathbf{u})}{\partial \mathbf{u}} = \sum_{\substack{\text{term} \\ j \neq 0}} \frac{\partial \text{Jaccard}(\mathbf{u})}{\partial \mathbf{u}}$$

$$ii) = \sum_{\substack{\text{term} \\ j \neq 0}} \frac{\partial \text{Jaccard}(\mathbf{u})}{\partial \mathbf{v}_c}$$

$$iii) = 0$$