

Written :

a) As we're dealing with hot vectors, just $y_0 = 1$ and other elements are zeros so :

$$-\sum_{w \in V} y_w \log(\hat{y}_w) = -\log(\hat{y}_0)$$

b) $J_{\text{naive-softmax}}(v_c, u, U) = -\log \frac{\exp(u_0^T v_c)}{\sum_w \exp(u_w^T v_c)}$

$$\frac{d}{dv_c} -\log \frac{\exp(u_0^T v_c)}{\sum_w \exp(u_w^T v_c)}$$

$$\frac{d}{dv_c} -\log(\exp(u_0^T v_c)) + \frac{d}{dv_c} \log \sum_w \exp(u_w^T v_c)$$

$$\frac{d}{dv_c} -u_0^T v_c = -u_0$$

partial

$$\frac{d}{dv_c} \log \sum_w \exp(u_w^T v_c) = \frac{1}{\sum_w \exp(u_w^T v_c)} \cdot \sum_w \frac{d}{dv_c} \exp(u_w^T v_c)$$

partial

$$= \frac{1}{\sum_w \exp(u_w^T v_c)} \cdot \sum_w \exp(u_w^T v_c) \cdot \frac{d}{dv_c} u_w^T v_c$$

u_w

$$= -u_0 + \sum_w \frac{\exp(u_w^T v_c)}{\sum_w \exp(u_w^T v_c)} \cdot u_w$$

$$= -u_0 + \sum_w P(0=w/c=c) \cdot u_w = U(\hat{y} - y)$$

$-U \cdot y$

$U \cdot \hat{y}$

c) $w \neq 0$

$$\frac{dJ}{du_w} = - \frac{d}{du_w} (u_0^T v_c) + \frac{d}{du_w} \log \left(\sum_w \exp(u_w^T v_c) \right)$$

$$= \frac{\sum_w \exp(u_w^T v_c) \cdot v_c}{\sum_w \exp(u_w^T v_c)} = P(O=w / C=c) \cdot v_c$$

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

$$\frac{d\sigma(x)}{dx} = \frac{1}{1+e^{-x}} \left(1 - \frac{1}{1+e^{-x}} \right) = \sigma(x) (1 - \sigma(x))$$

e) $J_{\text{neg-sampling}}(v_c, u, \mathcal{U}) = -\log(\sigma(u_0^T v_c)) - \sum_K \log(\sigma(-u_K^T v_c))$

$$\frac{dJ}{dv_c} = -u_0 (1 - \sigma(u_0^T v_c)) + \sum_K u_K (1 - \sigma(-u_K^T v_c))$$

$$\frac{dJ}{du_0} = -v_c (1 - \sigma(u_0^T v_c))$$

$$\frac{dJ}{du_K} = v_c (1 - \sigma(-u_K^T v_c)) = (1 - \sigma(u_K^T v_c)) v_c$$

f) $J_{\text{skip-gram}}(v_c, w_{k-m}, \dots, w_{k+m}, \mathcal{U}) = \sum_{-m \leq j \leq m} J(v_c, w_{k+j}, \mathcal{U})$

$$\frac{dJ}{d\mathcal{U}} = \sum_{-m \leq j \leq m} \frac{dJ(v_c, w_{k+j}, \mathcal{U})}{d\mathcal{U}}$$

$$\frac{dJ}{dv_c} = \sum_{-m \leq j \leq m} \frac{dJ(v_c, w_{k+j}, \mathcal{U})}{dv_c}$$

$$\frac{dJ}{dv_w} = \sum_{-m \leq j \leq m} \frac{dJ(v_c, w_{k+j}, \mathcal{U})}{dv_w} = 0$$