

Let's learn these derivations

We have  $X_i = U^T V_c$

and  $y = \text{softmax}(X_i) = \frac{e^{x_i}}{\sum_j e^{x_j}}$

we want  $\frac{\partial S}{\partial x_i}$

and  $\mathcal{L}_j = -\log(\hat{y}_j)$

$$\frac{\partial S}{\partial x_i} = \frac{\partial S}{\partial \hat{y}_i} \frac{\partial \hat{y}_i}{\partial x_i}$$

$$\frac{\partial S}{\partial \hat{y}_i} = -\frac{1}{\hat{y}_i}$$

$$\begin{aligned} &\downarrow \\ &= \left(-\frac{1}{\hat{y}_i}\right) (\hat{y}_i (\delta_{ij} - \hat{y}_j)) \\ &= \hat{y}_j - y_j \end{aligned}$$

Now we want  $\frac{\partial S}{\partial u_0}$  and  $\frac{\partial S}{\partial v_c}$

$$\frac{\partial S}{\partial u_0} = \frac{\partial S}{\partial x_i} \frac{\partial x_i}{\partial u_0} = (y_j - y_0) w_c$$

$$\frac{\partial S}{\partial v_c} = (y_j - y) v_c^T$$

$$\frac{\partial S}{\partial v_c} = \sum_{w=1}^V (\hat{y}_j - y) u_w = U^T (\hat{y} - y)$$

$N \times V \quad V \times 1 \quad N \times V \quad V \times 1$

• word vectors are rows in U

when  $w \neq 0$ ,

$$\frac{\partial y}{\partial x_0} = \frac{(0)(\Sigma) - (e^{x_0})(e^{x_w})}{\Sigma^2} = -y_0 y_w$$

but  $\frac{\partial x_0}{\partial u_w} = 0?$

2c)  $\frac{d}{dx} \sigma(x) = \frac{d}{dx} \left( \frac{1}{1+e^{-x}} \right) = \frac{(0)(1+e^{-x}) - (1)(-e^{-x})}{(1+e^{-x})^2} = \frac{e^{-x}}{(1+e^{-x})^2}$

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

25)

$$\text{Jueg-sample } \{V_0, 0, U\} = -\log(\sigma(u_0^T V_c)) - \sum_k \log(\sigma(-u_k^T V_c))$$

$$x = u_0^T V_c \quad y_k = -u_k^T V_c$$

$$z_1 = \sigma(x) \quad z_{2k} = \sigma(y_k)$$

$$J = -\log z_1 - \sum_k \log(z_{2k})$$

$$\frac{\partial J}{\partial V_c} = \frac{\partial J}{\partial z_1} \frac{\partial z_1}{\partial x} \frac{\partial x}{\partial V_c} + \sum_k \frac{\partial J}{\partial z_{2k}} \frac{\partial z_{2k}}{\partial y_k} \frac{\partial y_k}{\partial V_c}$$

$$\frac{\partial J}{\partial z_1} = -\frac{1}{z_1}$$

$$\frac{\partial z_1}{\partial x} = \sigma(x)(1-\sigma(x))$$

$$\frac{\partial x}{\partial V_c} = u_0$$

$$\frac{\partial J}{\partial z_1} \frac{\partial z_1}{\partial x} \frac{\partial x}{\partial V_c} = \left(-\frac{1}{z_1}\right) (\sigma(x)(1-\sigma(x))) (u_0) = -\frac{1}{\sigma(x)} (\sigma(x)(1-\sigma(x))) (u_0)$$

$$= (\sigma(u_0^T V_c) - 1) (u_0)$$

$$\frac{\partial J}{\partial z_{2k}} = \left(-\frac{1}{z_{2k}}\right)$$

$$\frac{\partial z_{2k}}{\partial y_k} = (\sigma(y_k)(1-\sigma(y_k)))$$

$$\frac{\partial y_k}{\partial V_c} = -u_k$$

↓

$$\sum_k (1 - \sigma(-u_k^T V_c)) u_k$$

$$\boxed{\frac{\partial J}{\partial V_c} = (\sigma(u_0^T V_c) - 1) (u_0) + \sum_k (1 - \sigma(-u_k^T V_c)) u_k}$$

2.

$$\frac{\partial J}{\partial u_0} = \frac{\partial J}{\partial z_1} \frac{\partial z_1}{\partial x} \frac{\partial x}{\partial u_0} = \left(-\frac{1}{z_1}\right) (z_1(1-z_1)) (V_c)$$

$$\boxed{\frac{\partial J}{\partial u_0} = (\sigma(u_0^T V_c) - 1) (V_c)}$$

$$\frac{\partial J}{\partial u_k} = \sum_k \frac{\partial J}{\partial z_{2k}} \frac{\partial z_{2k}}{\partial y_k} \frac{\partial y_k}{\partial u_k} = \sum_k \left(-\frac{1}{z_{2k}}\right) (z_{2k}(1-z_{2k})) (-V_c)$$

$$\boxed{\frac{\partial J}{\partial u_k} = (1 - \sigma(u_k^T V_c)) V_c}$$

$$2g) J_{\text{avg-sample}}(V_L; 0, U) = -\log(\sigma(u_0^T V_L)) - \sum \log(\sigma(-u_k^T V_L))$$

$$\frac{\partial J}{\partial u} = N(1 - \sigma(-u_k^T V_L)) (V_L) \quad (N = \text{number of reps})$$

$$2u) \text{ ci) } \frac{\partial J_{\text{clip-gran}}}{\partial U} = \sum_{\substack{-w_k \leq u \\ j \neq 0}} \frac{\partial J(V_L, w_{t+j}, U)}{\partial U}$$

$$\text{cii) } \frac{\partial J_{\text{clip-gran}}}{\partial V_L} = \sum_{\substack{-w_k \leq u \\ j \neq 0}} \frac{\partial J(V_L, w_{t+j}, U)}{\partial V_L}$$

$$\text{ciii) } \frac{\partial J_{\text{clip-gran}}}{\partial V_w} = 0$$