

$$P(0=0 | C=c) = \frac{\exp(u_0^T v_c)}{\sum_{w \in \text{vocab}} \exp(u_w^T v_c)} = \hat{y}_0 \in \mathbb{R}^{1 \times 1}$$

$$u_0 \in \mathbb{R}^{n \times 1}, v_c \in \mathbb{R}^{n \times 1}$$

$$U^T = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_{\text{vocab}} \end{bmatrix}^T \in \mathbb{R}^{n \times \text{vocab}}$$

$$\begin{aligned} J_{\text{min-softmax}}(v_c, 0, U) &= -\log P(0=0 | C=c) = -\sum_{w \in \text{vocab}} y_w \cdot \log \hat{y}_w = -\log(\hat{y}) \\ &= -u_0^T v_c + \log\left(\sum_w \exp(u_w^T v_c)\right) \end{aligned}$$

$$U \in \mathbb{R}^{\text{vocab} \times n}$$

$$\begin{aligned} \frac{\partial J}{\partial v_c} &= -u_0^T + \frac{\sum_w u_w^T \exp(u_w^T v_c)}{\sum_w \exp(u_w^T v_c)} = \underbrace{\exp(U)}_{\text{vocab} \times n} \underbrace{v_c}_{n \times 1} \\ &\in \mathbb{R}^{1 \times n} = -u_0^T + \sum_w \underbrace{P(0=w | C=c)}_{1 \times 1} \cdot \underbrace{u_w^T}_{1 \times n} \\ &= -u_0^T + \sum_w \hat{y}_w \cdot u_w^T \\ &= \sum_w (\hat{y}_w - y_w) \cdot u_w^T = (U(\hat{y} - y))^T \in \mathbb{R}^{1 \times n} \end{aligned}$$

$$J = -v_c u_0^T + \log\left(\sum_w \exp(v_c u_w^T)\right)$$

$$\frac{\partial J}{\partial v_c} = \underbrace{(\hat{y} - y)}_{1 \times \text{vocab}} \underbrace{U}_{\text{vocab} \times n} = -u_0 + \frac{\sum_w u_w \exp(v_c u_w^T)}{\sum_w \exp(v_c u_w^T)}$$

$$(b) - (ii) \quad \frac{\partial J}{\partial v_c} = (U(\hat{y} - y))^T = \vec{0}, \text{ The gradient is zero when the predicted distribution equals the true distribution } (\hat{y} = y)$$

$$(b) - (iii) \quad v_c \leftarrow v_c - \eta \cdot \frac{\partial J}{\partial v_c} \Leftrightarrow v_c \leftarrow v_c - \eta \cdot (U\hat{y} - U y)^T$$

$$(d) \quad i) w=0 \quad \frac{\partial J}{\partial w} = -v_c^T + \frac{v_c^T \cdot \exp(u_0^T v_c)}{\sum_z \exp(u_z^T v_c)} = \left(-1 + \frac{\exp(u_0^T v_c)}{\sum_z \exp(u_z^T v_c)}\right) v_c^T = \underbrace{(\hat{y}_0 - y_0)}_{1 \times 1} \underbrace{v_c^T}_{1 \times n}$$

ii) $w \neq 0$

$$\frac{\partial J}{\partial w} = \frac{v_c^T \exp(u_w^T v_c)}{\sum_z \exp(u_z^T v_c)} = \underbrace{\hat{y}_w}_{1 \times 1} \cdot \underbrace{v_c^T}_{1 \times n} \quad \hat{y} \in \mathbb{R}^{\text{vocab} \times 1}$$

(e)

$$\underbrace{\frac{\partial J}{\partial U}}_{\text{vocab} \times n} = \begin{bmatrix} \frac{\partial J}{\partial u_1} \\ \vdots \\ \frac{\partial J}{\partial u_{\text{vocab}}} \end{bmatrix} = \begin{bmatrix} \frac{\partial J}{\partial u_1} & \dots & \frac{\partial J}{\partial u_{\text{vocab}}} \end{bmatrix}^T$$

$$f \in \mathbb{R}^{1 \times \text{vocab}}$$

(f)

$$\frac{dF}{dx} = \begin{cases} 1, & x > 0 \\ x, & x < 0 \end{cases} \quad (g) \quad \frac{d\sigma(x)}{dx} = \frac{e^{-x}}{(1+e^{-x})^2} = \frac{1}{1+e^{-x}} \cdot \frac{e^{-x}}{1+e^{-x}} = \sigma(x) \cdot (1-\sigma(x))$$

(h) - (i)

$$\text{Let } x = u_0^T v_c, \quad y = -u_{ws}^T v_c, \quad f = \log \Rightarrow J = -f(\sigma(x)) - \sum_{s=1}^k f(\sigma(y))$$

$$u_0, v_c \in \mathbb{R}^{1 \times d}$$

$$\textcircled{1} \quad \underbrace{\frac{\partial J}{\partial v_c}}_{\in \mathbb{R}^{1 \times d}} = - \frac{\partial f}{\partial \sigma} \cdot \frac{\partial \sigma}{\partial x} \cdot \frac{\partial x}{\partial v_c} - \sum_{s=1}^k \frac{\partial f}{\partial \sigma} \cdot \frac{\partial \sigma}{\partial y} \cdot \frac{\partial y}{\partial v_c}$$

$$= - \frac{1}{\sigma(x)} \cdot \cancel{\sigma(x)} \cdot (1-\sigma(x)) \cdot u_0^T + \sum_{s=1}^k \frac{1}{\sigma(y)} \cdot \cancel{\sigma(y)} \cdot (1-\sigma(y)) \cdot (+u_{ws}^T)$$

$$= (\sigma(u_0^T v_c) - 1) u_0^T + \sum_{s=1}^k (1 - \sigma(-u_{ws}^T v_c)) \cdot u_{ws}^T$$

$$(\sigma(u_0 v_c^T) - 1) u_0 + \sum_{s=1}^k (1 - \sigma(-u_{ws} v_c^T)) u_{ws}$$

$$\textcircled{2} \quad \frac{\partial J}{\partial u_0} = - \frac{\partial f}{\partial \sigma} \cdot \frac{\partial \sigma}{\partial x} \cdot \frac{\partial x}{\partial u_0} - \sum_{s=1}^k \frac{\partial f}{\partial \sigma} \cdot \frac{\partial \sigma}{\partial y} \cdot \boxed{\frac{\partial y}{\partial u_0}}$$

$$= - \frac{1}{\sigma(x)} \cdot \cancel{\sigma(x)} \cdot (1-\sigma(x)) \cdot v_c^T - \sum_{s=1}^k \frac{1}{\sigma(y)} \cdot \sigma(y) (1-\sigma(y)) \cdot \textcircled{0}$$

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

$$\textcircled{2} \quad \frac{\partial J}{\partial w_s} = - \frac{\partial A}{\partial \sigma} \cdot \frac{\partial \sigma}{\partial x} \cdot \frac{\partial x}{\partial w_s} - \sum_{s=1}^K \frac{\partial A}{\partial \sigma} \cdot \frac{\partial \sigma}{\partial y} \cdot \frac{\partial y}{\partial w_s}$$

$$u_{cs} \leftarrow u_{ms} - \eta \cdot \frac{\partial J}{\partial w_s}$$

$$= - \frac{1}{\sigma(x)} \cdot \cancel{\sigma(x)} \cdot (1 - \sigma(x)) \cdot 0 + \frac{1}{\sigma(y)} \cdot \cancel{\sigma(y)} \cdot (1 - \sigma(y)) \cdot (+ v_c^T)$$

$$= (1 - \sigma(-u_{ms}^T v_c)) \cdot v_c^T$$

$$(h) - (ii) \quad U_{0, \{w_1, \dots, w_k\}} = [u_0, -w_1, \dots, -w_k] \quad \therefore \sigma(U_{0, \{w_1, \dots, w_k\}}^T v_c)$$

$$(i) \quad J_{\text{neg-sample}} = -\log(\sigma(u_0^T v_c)) - \sum_{s=1}^K \log(\sigma(-u_{w_s}^T v_c))$$

$$= -\log(\sigma(u_0^T v_c)) - \left\{ \sum_{i: w_i = w_s} \log(\sigma(-u_{w_i}^T v_c)) + \sum_{j: w_j \neq w_s} \log(\sigma(-u_{w_j}^T v_c)) \right\}$$

$$\frac{\partial J}{\partial w_s} = - \sum_{i: w_i = w_s} \frac{\partial}{\partial w_s} \left\{ \log(\sigma(-u_{w_i}^T v_c)) \right\} = - \sum_{i: w_i = w_s} \sigma(-u_{w_i}^T v_c) v_c$$

$$J_{\text{neg-sample}} = -\log(\sigma(u_0^T v_c)) - \sum_{s=1}^K \log(\sigma(-u_{w_s}^T v_c)) \quad u_0 \in \mathbb{R}^{1 \times d}, v_c \in \mathbb{R}^{d \times 1}$$

$$= -\log(\sigma(u_0^T v_c)) - \left\{ \sum_{i: w_i = w_s} \log(\sigma(-u_{w_i}^T v_c)) + \sum_{j: w_j \neq w_s} \log(\sigma(-u_{w_j}^T v_c)) \right\}$$

$$(i) - (i) \quad \frac{\partial J_{\text{skip-gram}}(v_c, w_{c-m}, \dots, w_{c+m}, U)}{\partial U} = \sum_{-m \leq j \leq m, j \neq 0} \frac{J(v_c, w_{c+j}, U)}{\partial U}$$

$$(j) - (ii) \quad \frac{\partial J_{\text{skip-gram}}(v_c, w_{c-m}, \dots, w_{c+m}, U)}{\partial v_c} = \sum_{-m \leq j \leq m, j \neq 0} \frac{J(v_c, w_{c+j}, U)}{\partial v_c}$$

$$(j) - (iii) \quad \frac{\partial J_{\text{skip-gram}}(v_c, w_{c-m}, \dots, w_{c+m}, U)}{\partial w} = \sum_{\substack{-m \leq j \leq m, j \neq 0 \\ w_{c+j} = w}} \frac{J(v_c, w_{c+j}, U)}{\partial w}$$