

BPE

aaabcaabdeab

① $A = aa \quad B = ab$

ABcAbdeB

② $\text{vocab} = \{A, B, b, c, d, e\}$

NB inference

$$y = \operatorname{argmax}_{y \in Y} \frac{P_0(x|y) \cancel{P(y)}}{\cancel{P(x)}} \rightarrow 1/2$$

$$= \operatorname{argmax}_{y \in Y} \prod_{i=1}^n P_0(x_i|y)$$

$$= \operatorname{argmax}_{y \in Y} \sum_{i=1}^n \underbrace{\log P_0(x_i|y)}_{\text{score of each word}}$$

MLE for NB

Likelihood function

$$\begin{aligned} L(\theta, \alpha) &= \prod_{i=1}^N \log P(x^{(i)}, y^{(i)}; \theta, \alpha) \\ &= \sum_{i=1}^N \log \underbrace{P(x^{(i)} | y^{(i)}; \theta)}_{\text{const.}} \underbrace{P(y^{(i)}; \alpha)} \end{aligned}$$

$$\theta \in \mathbb{R}^{2 \times |V|} \quad \text{s.t.} \quad \sum_i \theta_{i,y} = 1$$

$$\alpha \in \mathbb{R}$$

$$\underbrace{P(y^{(i)}; \alpha)} = \begin{cases} \alpha & \text{if } y^{(i)} = 1 \\ 1 - \alpha & \text{a.w.} \end{cases}$$

$$= \mathbb{1}(y^{(i)} = 1) \alpha + \mathbb{1}(y^{(i)} = 0) (1 - \alpha)$$

$L(\alpha)$ is concave.

$$\begin{aligned} \frac{\partial L(\alpha)}{\partial \alpha} &= \frac{\partial}{\partial \alpha} \sum_{i=1}^N \log [\mathbb{1}(y^{(i)} = 1) \alpha + \mathbb{1}(y^{(i)} = 0) (1 - \alpha)] \\ &= \frac{\partial}{\partial \alpha} \sum_{i=1}^N \mathbb{1}(y^{(i)} = 1) \log \alpha + \mathbb{1}(y^{(i)} = 0) \log (1 - \alpha) \end{aligned}$$

$$= \sum_{i=1}^N \mathbb{1}(y^{(i)} = 1) \frac{1}{\alpha} - \mathbb{1}(y^{(i)} = 0) \frac{1}{1 - \alpha} = 0$$

$$\hat{\alpha} = \frac{\sum_{i=1}^N \mathbb{1}(y^{(i)} = 1)}{\sum_{i=1}^N \mathbb{1}(y^{(i)} = 1) + \mathbb{1}(y^{(i)} = 0)}$$

$$= \# \text{ pos. ex.} / \# \text{ total}$$

MLE for LR

$$L(\underline{w}) = \sum_{i=1}^N \log P(y^{(i)} | x^{(i)}; \underline{w})$$

$$= \sum_{i=1}^N y^{(i)} \log \frac{1}{1 + e^{-\underline{w} \cdot \underline{\phi}(x^{(i)})}} + (1 - y^{(i)}) \log \left[1 - \frac{1}{1 + e^{-\underline{w} \cdot \underline{\phi}(x^{(i)})}} \right]$$

Is $L(\underline{w})$ concave?

$$\log \frac{1}{1 + e^{-z}}$$

① plot

② $f''(z) \leq 0$

No closed-form solution!

BOW

in for a penny, in for a pound

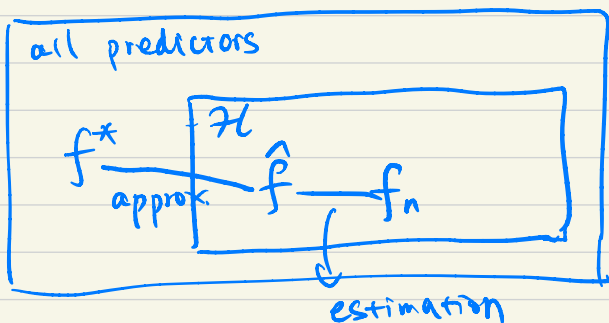
$$\text{in} = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \begin{array}{l} \rightarrow 1(\text{"in"} = w_1) \\ \vdots \\ \rightarrow 1(\text{"in"} = w_{20}) \\ \vdots \\ \rightarrow 1(\text{"in"} = w_{|V|}) \end{array}$$

$|V| \times 1$

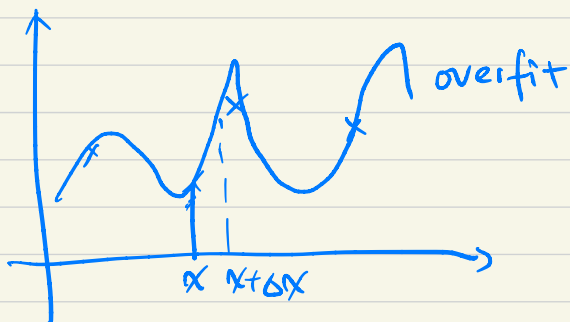
$$\phi_{\text{BOW}}(\text{sentence}) = \begin{bmatrix} 0 \\ 2 \\ 0 \\ 2 \\ 2 \\ 1 \\ 1 \\ 0 \end{bmatrix} \begin{array}{l} \text{the} \\ \text{a} \\ \text{an} \\ \text{in} \\ \text{for} \\ \text{penny} \\ \text{pound} \\ \vdots \end{array}$$

$|V| \times 1$

Error decomposition



Why small norms?



$$|w \cdot x - w \cdot (x + \delta x)|$$

$$= |w \cdot \delta x|$$

$$\leq \underbrace{\|w\|}_{\text{small}} \cdot \|\delta x\|$$