

## Question 1

We would like to ensure that for all  $t, u, v$ ,  $\sum_w q_{BO}(w|t, u, v) = 1$ . Note that the “missing” probability mass is

$$1 - \sum_{w \in \mathcal{A}(t, u, v)} q_{BO}(w|t, u, v) = 1 - \sum_{w \in \mathcal{A}(t, u, v)} \frac{c^*(t, u, v, w)}{c(t, u, v)}$$

If we set  $\alpha(t, u, v) = 1 - \sum_{w \in \mathcal{A}(t, u, v)} \frac{c^*(t, u, v, w)}{c(t, u, v)}$  we can verify that  $\sum_w q_{BO}(w|t, u, v) = 1$ : for any  $t, u, v$

$$\begin{aligned} & \sum_w q_{BO}(w|t, u, v) \\ &= \sum_{w \in \mathcal{A}(t, u, v)} q_{BO}(w|t, u, v) + \sum_{w \in \mathcal{B}(t, u, v)} q_{BO}(w|t, u, v) \\ &= \sum_{w \in \mathcal{A}(t, u, v)} \frac{c^*(t, u, v, w)}{c(t, u, v)} + \sum_{w \in \mathcal{B}(t, u, v)} \frac{\alpha(t, u, v) \times q_{BO}(w|u, v)}{\sum_{w \in \mathcal{B}(t, u, v)} q_{BO}(w|u, v)} \\ &= \sum_{w \in \mathcal{A}(t, u, v)} \frac{c^*(t, u, v, w)}{c(t, u, v)} + \alpha(t, u, v) = 1 \end{aligned}$$

## Questions 2a, 2b

*Maximum* value of perplexity: if for any sentence  $x^{(i)}$ , we have  $p(x^{(i)}) = 0$ , then  $l = -\infty$ , and  $2^{-l} = \infty$ . Thus the maximum possible value is  $\infty$ .

*Minimum* value: if for all sentences  $x^{(i)}$  we have  $p(x^{(i)}) = 1$ , then  $l = 0$ , and  $2^{-l} = 1$ . Thus the minimum possible value is 1.

## Question 2c

An example that gives the maximum possible value for perplexity:

Training corpus consists of the single sentence

the a STOP

Test corpus consists of the single sentence

a the STOP

It can be verified that a bigram language model as described in the question, trained on the single sentence *the a STOP*, gives probability 0 to the sentence *a the STOP*, and hence has infinite perplexity on this test corpus.

## Question 2d

An example that gives the maximum possible value for perplexity:

Training corpus consists of the single sentence

the a STOP

Test corpus consists of the single sentence

the a STOP

It can be verified that a bigram language model as described in the question, trained on the single sentence *the a STOP*, gives probability 1 to the sentence *the a STOP*, and hence has perplexity equal to one on this test corpus.

## Question 3a

Rearranging terms slightly, we have

$$\begin{aligned}q(w|u, v) &= \alpha \times q_{ML}(w|u, v) \\ &\quad + (1 - \alpha) \times \beta \times q_{ML}(w|v) \\ &\quad + (1 - \alpha) \times (1 - \beta) \times q_{ML}(w)\end{aligned}$$

Hence we have  $\lambda_1 = \alpha = 0.5$ ,  $\lambda_2 = (1 - \alpha) \times \beta = 0.25$ , and  $\lambda_3 = (1 - \alpha) \times (1 - \beta) = 0.25$ .

## Question 3b

We have an interpolated model with  $\lambda_1(u, v) = \alpha(u, v)$ ,

$\lambda_2(u, v) = (1 - \alpha(u, v)) \times \beta(u)$ , and

$\lambda_3(u, v) = (1 - \alpha(u, v)) \times (1 - \beta(u))$ .

Define  $\mathcal{V}' = \mathcal{V} \cup \{\text{STOP}\}$ .

$$\sum_{w \in \mathcal{V}'} q(w \mid u, v)$$

$$= \sum_{w \in \mathcal{V}'} [\lambda_1(u, v) \times q_{ML}(w \mid u, v) + \lambda_2(u, v) \times q_{ML}(w \mid v)$$

$$+ \lambda_3(u, v) \times q_{ML}(w)]$$

$$= \lambda_1(u, v) \sum_w q_{ML}(w \mid u, v) + \lambda_2(u, v) \sum_w q_{ML}(w \mid v) + \lambda_3(u, v) \sum_w q_{ML}(w)$$

$$= \lambda_1(u, v) + \lambda_2(u, v) + \lambda_3(u, v)$$

$$= \alpha(u, v) + (1 - \alpha(u, v)) \times \beta(u) + (1 - \alpha(u, v)) \times (1 - \beta(u))$$

$$= 1$$

## Question 3c

As  $\text{Count}(u, v)$  increases,  $\alpha(u, v)$  gets closer to 1, reflecting the intuition that as  $\text{Count}(u, v)$  increases, the estimate  $q_{ML}(w|u, v)$  becomes more reliable, and more weight should be put on it.

A similar argument applies to  $\beta(v)$  and  $\text{Count}(v)$ .

The constants  $C_1$  and  $C_2$  dictate how quickly  $\alpha(u, v)$  and  $\beta(v)$  approach 1 respectively. They can be set by optimization of the perplexity on a held-out corpus.

## Question 3d

Under the assumptions of the question

$$q_{ML}(w) = \text{Count}(w)/N > 0.$$

We have

$$\begin{aligned}q(w|u, v) &= \alpha(u, v) \times q_{ML}(w|u, v) \\ &\quad + (1 - \alpha(u, v)) \times \beta(u) \times q_{ML}(w|v) \\ &\quad + (1 - \alpha(u, v)) \times (1 - \beta(u)) \times q_{ML}(w)\end{aligned}$$

Hence

$$q(w|u, v) \geq (1 - \alpha(u, v)) \times (1 - \beta(u)) \times q_{ML}(w)$$

It can be verified that  $1 - \alpha(u, v) > 0$ ,  $1 - \beta(u) > 0$ , and  $q_{ML}(w) > 0$ . Hence for all  $u, v, w$ ,  $q_{ML}(w|u, v) > 0$ . It follows that for any sentence in the test data  $x^{(i)}$ ,  $p(x^{(i)}) > 0$ . It follows that the perplexity on the test data cannot be infinite.

## Question 4

First consider the statement “for all bigrams  $v, w$ , we have  $q_{BO}(w|v) \geq 0$ ”. For any  $v, w$  such that  $\text{Count}(v, w) = 1$ , we have

$$w \in \mathcal{A}(v)$$

and in addition

$$\text{Count}^*(v, w) = 1 - 1.5 = -0.5$$

It follows that

$$q_{BO}(w|v) = \frac{-0.5}{\text{Count}(v)} < 0$$

So the statement is **false**.

## Question 4 (continued)

Now consider the second statement, *for all unigrams  $v$  we have  $\sum_w q_{BO}(w|v) = 1$ .*

We have for all  $u, v$ ,

$$\begin{aligned}\sum_w q_{BO}(w|v) &= \sum_{w \in \mathcal{A}(v)} q_{BO}(w|v) + \sum_{w \in \mathcal{B}(v)} q_{BO}(w|v) \\ &= \sum_{w \in \mathcal{A}(v)} \frac{\text{Count}^*(v, w)}{\text{Count}(v)} + \sum_{w \in \mathcal{B}(v)} \frac{\alpha(v) \times q_{ML}(w)}{\sum_w q_{ML}(w)} \\ &= \sum_{w \in \mathcal{A}(v)} \frac{\text{Count}^*(v, w)}{\text{Count}(v)} + \alpha(v) \\ &= \sum_{w \in \mathcal{A}(v)} \frac{\text{Count}^*(v, w)}{\text{Count}(v)} + 1 - \sum_{w \in \mathcal{A}(v)} \frac{\text{Count}^*(v, w)}{\text{Count}(v)} + \\ &= 1\end{aligned}$$

Note that this holds even though some values for  $\text{Count}^*$  may be negative. Hence the statement is **true**.

