

(520|600).666  
Information Extraction from Speech and Text

Homework # 4

Due March 28, 2024.

In class, we discussed linear interpolation for smoothing a bigram language model, namely

$$P(w|v) = \gamma f(w|v) + (1 - \gamma) f(w),$$

where  $f(\cdot|\cdot)$  and  $f(\cdot)$  denoted the appropriate relative frequency estimates, and  $\gamma$  was chosen so as to maximize the probability of some held-out data.

This homework considers *alternative* strategies for smoothing a bigram language model by directly modifying the *counts* observed in the training data. In particular, let  $C(v, w)$  denote the count of a bigram  $\langle v, w \rangle$  in the *training text*, and let  $C^*(v, w)$  be the modified count. For some constant  $\theta > 0$ , consider the three cases

- (i)  $C^*(v, w) = C(v, w) + \theta$ ,
- (ii)  $C^*(v, w) = C(v, w) + \theta C(w)$ , and
- (iii)  $C^*(v, w) = C(v, w) + \theta C(v) f(w)$ .

In each case, the smoothed bigram probability is calculated as

$$P^*(w|v) = \frac{C^*(v, w)}{\sum_{w' \in \mathcal{V}} C^*(v, w')}.$$

Let  $N(v, w)$  denote the count of a bigram  $\langle v, w \rangle$  in the *held-out text*  $\mathcal{H}$ .

1. Derive an expression for the  $\theta$  that maximizes the log-probability

$$P(\mathcal{H}) = \sum_{v \in \mathcal{V}} \sum_{w \in \mathcal{V}} N(v, w) \log P^*(w|v)$$

of the held-out text in each of the three cases (i), (ii) and (iii) above.

2. Show that if  $N(v, w) = C(v, w)$  for all bigrams  $\langle v, w \rangle$ , then the optimal value is  $\theta = 0$  in each case. Why is this an expected result?
3. Show, in each case, that  $P^*$  may be written as the linear interpolation of a bigram and a lower order language model, though not necessarily  $f(w)$ .

$$P^*(w|v) = \gamma f_2(w|v) + (1 - \gamma) f_1(w),$$

i.e., identify  $f_1$ ,  $f_2$  and  $\gamma$ , and discuss the merits/drawbacks of each smoothing strategy.

After finishing the homework, carefully review *all sections* of Chapter 4 again.