
A: Theory

We consider a unigram as a single word and a bigram as an ordered pair of words in the form $\{w_1, w_2\}$ separated by space. We have two dictionaries Pw and $Pw2$ that contains unigrams and bigrams respectively along with their frequencies. We use the following notations.

Notations	Meaning
$Pw[w]$	frequency of w in Pw dictionary
$Pw2[w_1, w_2]$	frequency of $\{w_1, w_2\}$ bigram in $Pw2$ dictionary
V	Total number of different unigrams or bigrams
N	Sum of frequencies
$voc[w]$	Number of different bigrams started by w

There can be different cases that might happen when we analyze any bigram. However, when we analyze the probability of the very first word of the document or when we can not find the first word w_1 of an ordered pair $\{w_1, w_2\}$ in the dictionary Pw , we only use the unigram model to assign the probability of the word.

According to the unigram model, The probability of a word w is given by:

$$P(w) = \frac{Pw[w] + 1}{Pw.N + Pw.V + 1} \quad (1)$$

if it is found in Pw or

$$P(w) = \frac{1}{Pw.N + Pw.V + 1} \quad (2)$$

if it is not found in Pw

However, we use bigram model for every ordered pair $\{w_1, w_2\}$ when w_1 is found in Pw . According to the bigram model, the conditional probability of w_2 given that we already observed w_1 is calculated. There can be any of the following three cases.

(i) Bigram $\{w_1, w_2\}$ in $Pw2$

In this case there is a connection between two known words. We consider conditional probability for w_2 as:

$$P(w_2|w_1) = \frac{Pw2[w_1, w_2] + 1}{Pw[w_1] + Pw2.voc[w_1] + 1} \quad (3)$$

(ii) Bigram not in $Pw2$, w_2 in Pw

Although there is no observed connection, both are known words. In this case, we assign the conditional probability based on the frequency of w_2 in Pw .

$$P(w_2|w_1) = \frac{Pw[w_2] + 1}{(Pw[w_1] + Pw2.voc[w_1] + 1)(Pw.N + Pw.V + 1)} \quad (4)$$

(iii) Bigram not in $Pw2$, w_2 is not in Pw

In this case, w_2 is completely unknown. Instead of assigning a zero probability, we assign probability of w_2 as

$$P(w_2|w_1) = \frac{1}{(Pw[w_1] + Pw2.voc[w_1] + 1)(Pw.N + Pw.V + 1)} \quad (5)$$

B: Data Structure

Data Structure	Description
Pw,Pw2	Dictionaries with frequencies of unigrams and bigrams respectively
chart	Dynamic programming Table to store the words
entry	A structure {word, start position, log probability, back pointer}
pq	priority queue of entries, priority= lowest starting

C: Algorithm

The algorithm is given below:

Algorithm 1 Word Segmentation

Input: File with non segmented words *Input*,
 unigram dictionary Pw ,
 bigram dictionary Pw_2

Output: File with segmented words

```

1: Initialize  $chart = \{\}$  and  $maxlen \leftarrow$  longest bigram length
2: Initialize priority queue with starting position as priority
3: for  $l$  in range(1,...,maxlen) do
4:    $w_1 \leftarrow$  substring input[0,...,l-1]
5:   if  $w_1$  is found in  $Pw$  then
6:     calculate  $P(w_1)$  according to unigram model
7:     add  $\{w_1, 0, P[w_1], None\}$  into pq
8:   if nothing inserted in last for loop then
9:      $w_1 \leftarrow$  input[0]
10:    calculate  $P(w_1)$  according to unigram model
11:    add  $\{w_1, 0, Pw[w_1], None\}$  into pq
12: while pq not empty do
13:    $item \leftarrow$  highest priority item from pq
14:    $end \leftarrow item.length + item.start$ 
15:   if  $item.probability > chart[end].probability$  then
16:      $chart[end] \leftarrow item$ 
17:    $nextstart \leftarrow end + 1$ 
18:   if  $w_1$  is found in  $Pw$  then
19:     for  $l$  in range(1,...,maxlen) do
20:        $w_2 \leftarrow$  substring input[nextstart,...,nextstart+l-1]
21:       calculate  $P(w_2|w_1)$  according to bigram model
22:       add  $\{w_2, nextstart, P(w_1) * P(w_2|w_1), w_1\}$  into pq
23:     if nothing inserted in last for loop then
24:        $w_1 \leftarrow$  input[0]
25:       calculate  $P(w_1)$  according to unigram model
26:       add  $\{w_2, nextstart, P(w_1) * P(w_2|w_1), w_1\}$  into pq
27:   else
28:     for  $l$  in range(1,...,maxlen) do
29:        $w_2 \leftarrow$  substring input[nextstart,...,nextstart+l-1]
30:       calculate  $P(w_2)$  according to unigram model
31:       add  $\{w_2, nextstart, P(w_1) * P(w_2), w_1\}$  into pq
32:     if nothing inserted in last for loop then
33:        $w_2 \leftarrow$  input[nextstart]
34:       calculate  $P(w_2)$  according to unigram model
35:       add  $\{w_2, nextstart, P(w_1) * P(w_2), w_1\}$  into pq

```
