Discover Chinese Words

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Problem

- No explicit word boundaries (i.e., spaces) in Chinese text.
- But, we need the words for
- indexing in information retrieval (IR),
- natural language understanding.

Solution

- Use word probabilities to segment sentences into words
- Discover the words and their algorithm text using the probabilities from raw, unsegmented Expectation-Maximization (EM)

Words are running together ...

No explicit word boundaries (i.e., spaces) in a Chinese sentence. E.g.,

在此间正式开始谈判。 今天, 墨西哥与欧盟就签署自由贸易协定问题

- The words in the above Chinese sentence:
- ^ 在 ^ 此间 ^ 正式 ^ 开始 ^ 谈判 今天、墨西哥、与、欧盟、就、签署、自由、贸易、协定、问题
- Compare the English sentence WITHOUT / WITH the spaces between words:
- to establish azone of free trade between them."``Today, the European Union (EU) and Mexico started negotiations
- negotiations to establish a zone of free trade between them." "Today, the European Union (EU) and Mexico started

Segment the sentence into words: the maximum-likelihood approach

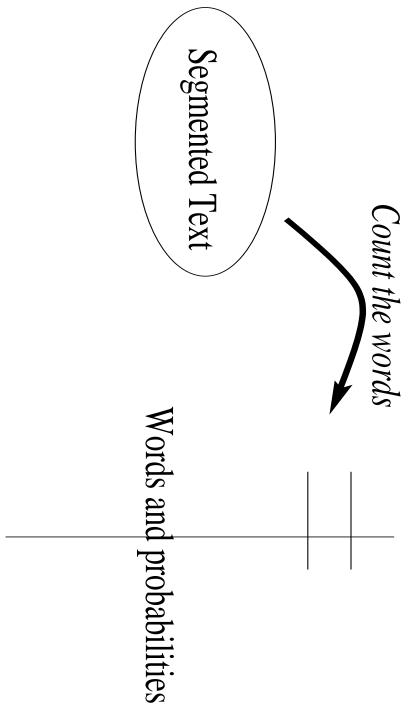
Segment the sentence $C_1C_2C_3...C_n$ into words $W_1W_2...W_m$ to maximize the likelihood $P(W_1)P(W_2)...P(W_m)$.

0.03	$C_1C_2C_3$	
0.001	$C_1 \wedge C_2 C_3$	
0.09	$C_1C_2 \wedge C_3$	
0.01	$C_1 \wedge C_2 \wedge C_3$	
likelihood	Segmentation	

This can be done by dynamic programming if we knew the words $\{W_i\}$ and their probabilities $P(W_i)$.

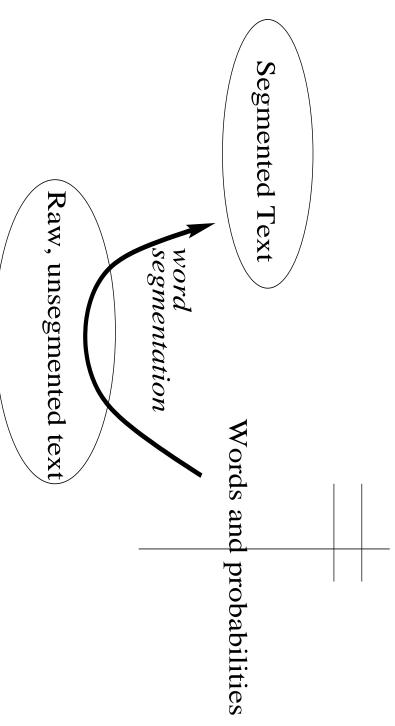
If we had a training corpus of segmented text ...

then we can easily get the words and their probabilities!



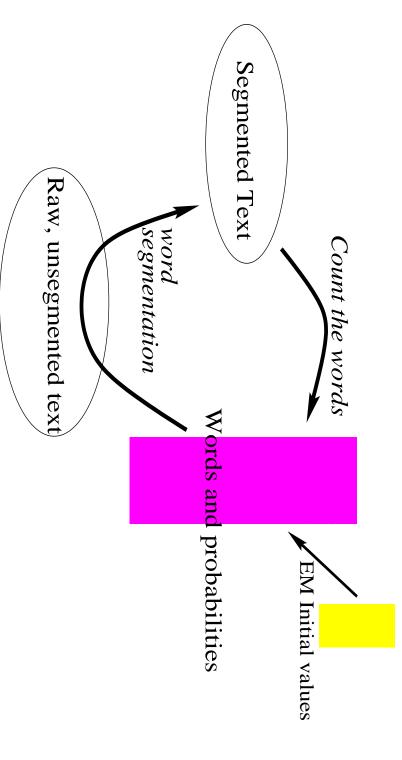
If we knew the words and their probabilities ...

• then we can get the segmented text!



Expectation-Maximization (EM) algorithm Combine them together:

Solves this problem of "Which came first, the chicken or the word probabilities. egg?" by providing the first "egg", i.e., the initial values for the

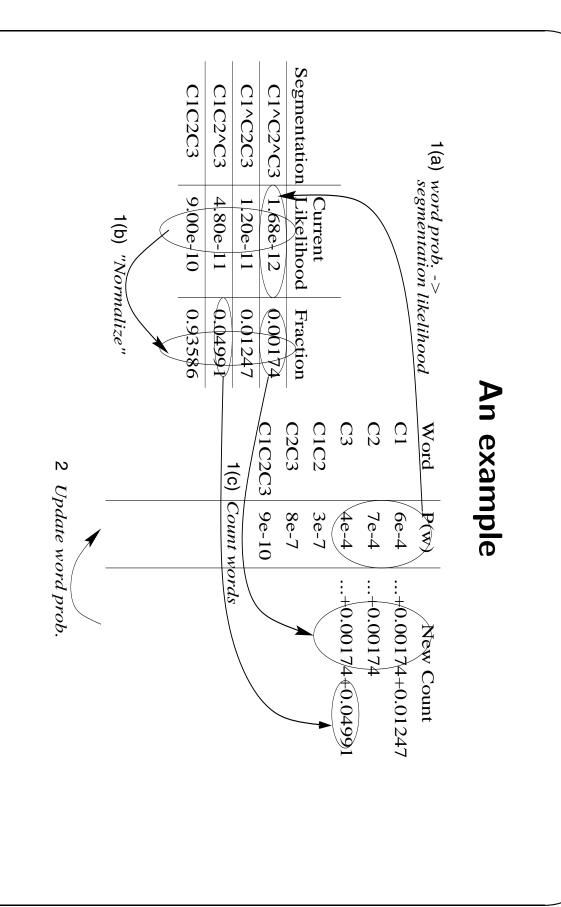


The main procedure

- 1. For each sentence in the unsegmented text,
- (a) Compute the likelihood of each possible segmentation using

the current estimated values of the word probabilities

- (b) The segmentation likelihood is normalized as "fraction" that sums to 1.
- (c) Count the words in each segmentation. I.e., add the "fraction" of the segmentation to the word count.
- 2. Update the word probabilities using the word counts.
- 3. Repeat until convergence.



It works!

- The algorithm correctly discovers most words from the unsegmented text
- 但[^] 艺术家[^] 要尊重[^] 艺术[^] 规律
- 香港、特区、是、祖国、的、一个、组成、部分

城市、是、先进、生产力、的、集聚、地、和、辐射源

- 信用社、体制、改革和、业务、发展
- Recall/Precision=65.65%/71.91%
- words, Recall/Precision can be boosted up to 97.72%/91.05% After splitting single-character "stop" words from content

Future work

Some single-character stop words (like "and", "of", "should", etc.) tend to cling to content words.

$\mathbf{S}_{\mathbf{J}}$	
houldrespect	要尊重
should respec	要^尊重

content words? A more principled way of splitting these stop words from

- Incorporate prior knowledge:
- Distribution of word lengths
- Existing word lists
- Part of speech information
- Other applications

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