

Background

Evaluating Contextual Dependency of Paraphrases using a Latent Variable Model

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- Problem: *contextual dependency* of paraphrases
 - applying
 - collecting
 - etc.
- How to cope with contextual dependency?

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Example

paraphrasing pair:
A: *I want to buy a pair of sandals.*
B: *I'm looking for sandals.*

Can we **always** paraphrase them?
from A to B: Yes
from B to A: No

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Objective

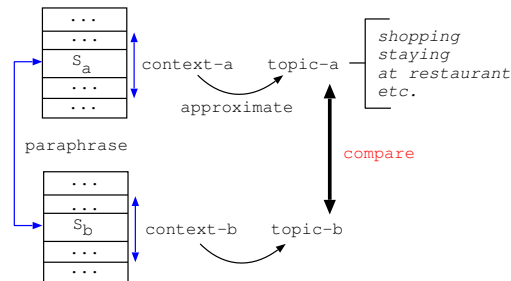
To construct an evaluation method for contextual dependency of paraphrases

If we can evaluate ...

- mis-paraphrasing
 - mis-collecting paraphrases
- will be avoided

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Idea



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Approach

Using latent variable text model

- modeling a text (unsupervised)
- each latent variable represents a *topic*

Context: sentence and surrounding sentences = window

Approximation of context: topic indicated by a latent variable

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Latent variable text models

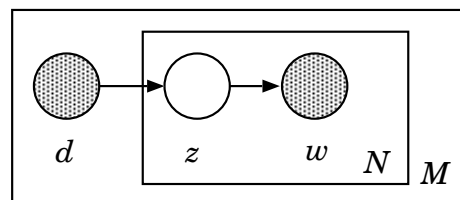
- PLSI $P(d, w)$
Probabilistic replacement of LSI
- LDA $P(d|\alpha, \beta)$
Bayesian replacement of pLSI

Latent (hidden) variable represents *topic*

Note: # of latent variables (topics) is given

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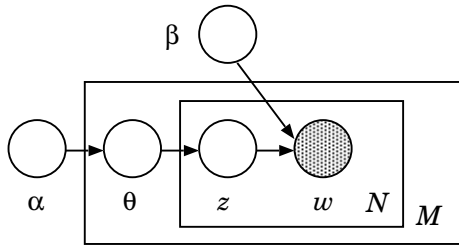
Latent variable model: pLSI



d : document, z : latent variable, w : word,
 N : vocabulary size, M : # of documents

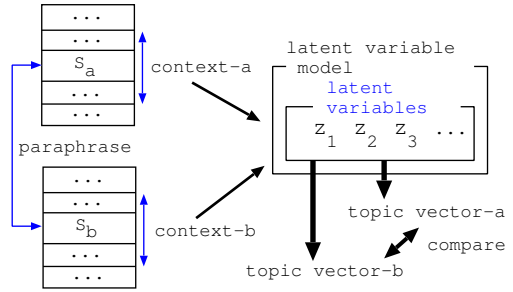
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Latent variable model: LDA



θ : Dirichlet random variable
 α : parameter for θ , β : parameter for w

Evaluating method



Comparing topic vectors

- **Based on the largest element**
Whether the largest element of topic vectors are the same
- **Cosine**
Whether $\cos\theta$ between topic vectors greater than threshold

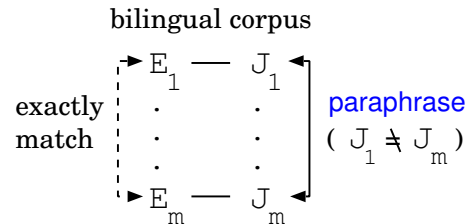
Overview of Experiments

- **Comparing with labels by human**
Matching the results of our method with the result based on the topic labels by human
- **Evaluation for paraphrasing**
Evaluating our method based on collecting situation for paraphrases

Data set

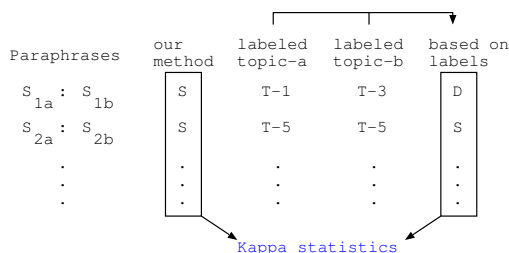
Bilingual corpus of travel conversation
 (162,000 sentence pairs)
 Manually and roughly labeled with topics
 (hierarchical; level-1: 19, level-2: 218)
 Textual cohesion
 → Used fixed window to clip a context
 Data format: bag-of-words

Extracting paraphrases



Obtained 944,547 Japanese paraphrasing pairs

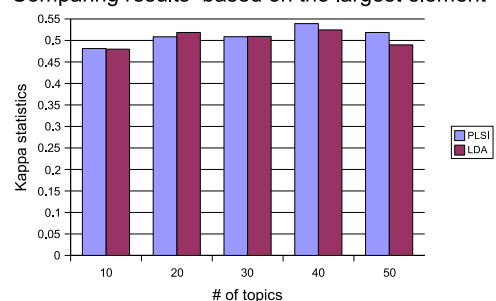
Comparing with labels by human



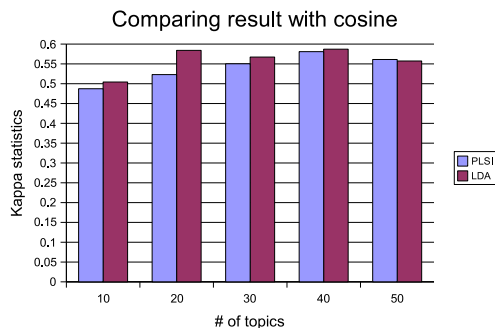
Measurement: Kappa statistics
 Two comparing method: largest, cosine

Result 1/2

Comparing results based on the largest element



Result 2/2



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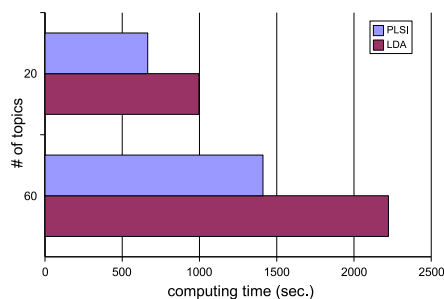
Discussion

- Kappa statistics: 0.59 (highest)
moderate: 0.4-0.6
 substantial: 0.6-0.8
 perfect: over 0.8
- No major performance difference
- LDA was good at with cosine
- PLSI sometimes outperformed LDA

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Computing time

Computing time to evaluate 944,547 paraphrases



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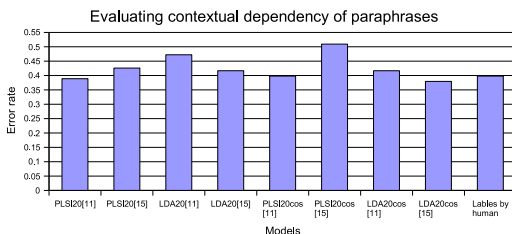
Evaluation for paraphrasing

1. Fed randomly sampled 108 paraphrasing pairs into our method
2. Manually evaluated whether they were contextually independent

Measurement: error rate

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Result



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Discussion

- Almost the same as the result based on labels by human
- *25 unavoidable errors*
 Potential upper bound based on topic information: 77% (0.23 error rate)

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Conclusion

- Proposed evaluation method for contextual dependency of paraphrases using pLSI and LDA
- No major performance difference between pLSI and LDA
- Potential upper bound using *only* topic: 77%
 Achieved: 62%

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Future works

- Introducing a topic-boundary detection technique
- Employing more complicated data (e.g., dependency structure) not b.o.w.
- Investigating difference of paraphrase:
 - What makes contextual dependency?
 - What contexts are possible?

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Thank you very much.