

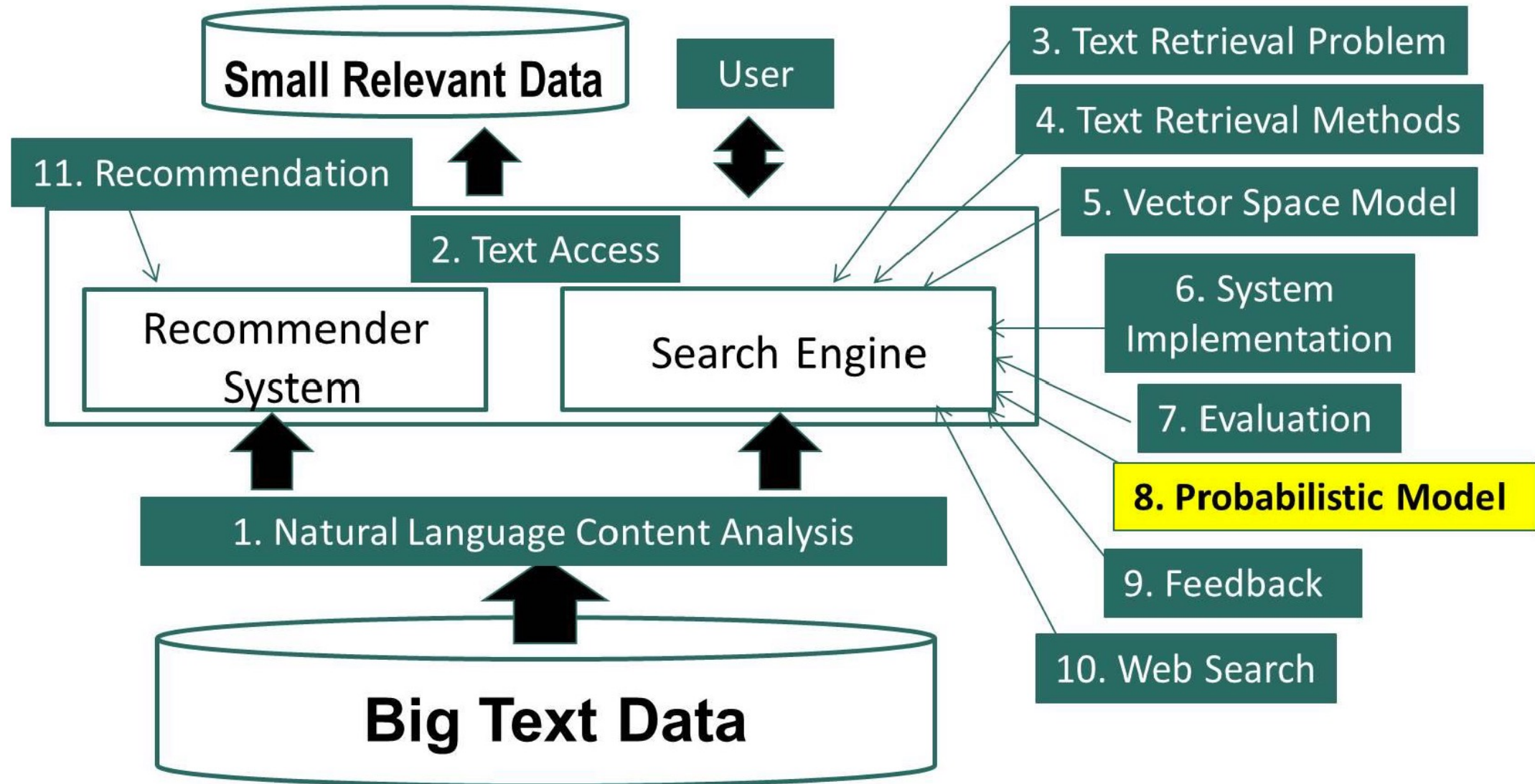


Information Retrieval

Probabilistic Retrieval Model: Statistical Language Model

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Probabilistic Retrieval Model: Basic Idea



Overview

- What is a Language Model?
- Unigram Language Model
- Uses of a Language Model

What is a Statistical Language Model (LM)?

- A probability distribution over word sequences
 - $p(\textit{"Today is Wednesday"}) \approx 0.001$
 - $p(\textit{"Today Wednesday is"}) \approx 0.0000000000000001$
 - $p(\textit{"The eigenvalue is positive"}) \approx 0.00001$
- Context-dependent!



This may be a reasonable language model for describing general conversations, but it may be inaccurate for describing conversations happening at a mathematics conference, where the sequence *The eigenvalue is positive* may occur more frequently than *Today is Wednesday*.

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- Context-dependent! Given a language model, we can sample word sequences according to the distribution to obtain a text sample. In this sense, we may use such a model to “generate” text. Thus, a language model is also often called a generative model for text.



Three arrows point from the dice to the following text samples:

Today is Wednesday
Today Wednesday is
The eigenvalue is positive

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 - Given that a user is interested in sports news, how likely would the user use “baseball” in a query? (information retrieval)

The Simplest Language Model: Unigram LM

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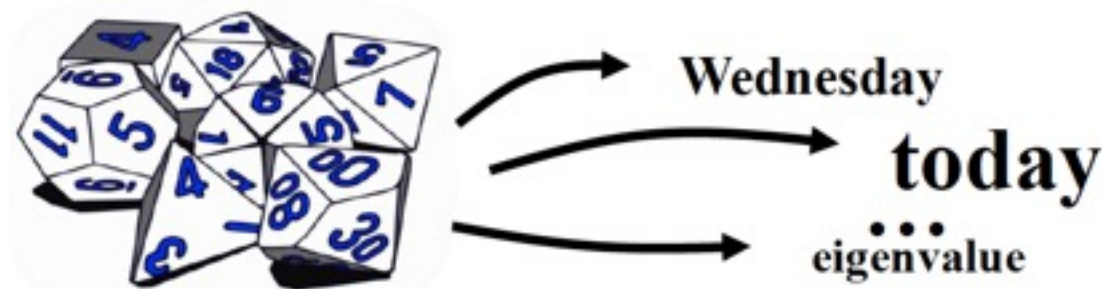
Given probabilities of each word, the sum of all is = 1

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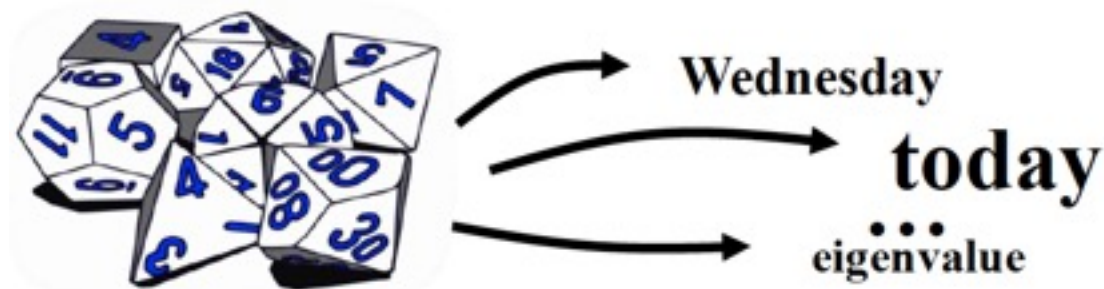
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$$\begin{aligned} p(\text{"today is Wed"}) \\ &= p(\text{"today"})p(\text{"is"})p(\text{"Wed"}) \\ &= 0.0002 \times 0.001 \times 0.000015 \end{aligned}$$

Text Generation with Unigram LM

Unigram LM $p(w|\theta)$  Document =?

Topic 1:
Text mining

...
text 0.2
mining 0.1
association 0.01
clustering 0.02
food 0.00001
...

Topic 2:
Health

...
food 0.25
nutrition 0.1
healthy 0.05
diet 0.02
...

Text Generation with Unigram LM

Unigram LM $p(w|\theta)$

Sampling →

Document =?

Topic 1:
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...



Text mining
paper

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Health

...
food 0.25
nutrition 0.1
healthy 0.05
diet 0.02
...



Food nutrition
paper

Estimation of Unigram LM

Unigram LM $p(w|\theta)=?$

Estimation

Text Mining Paper d

Total #words=**100**

text ?
mining ?
association
?
database ?
query ?



text 10
mining 5
association 3
database 3
algorithm 2

query 1
efficient 1

Estimation of Unigram LM

Unigram LM $p(w|\theta)=?$

Estimation

Text Mining Paper d

Total #words=**100**

10/100



text ?

5/100



mining ?

3/100



association

3/100



?

1/100



database ?

query ?



text 10
mining 5
association 3
database 3
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efficient 1

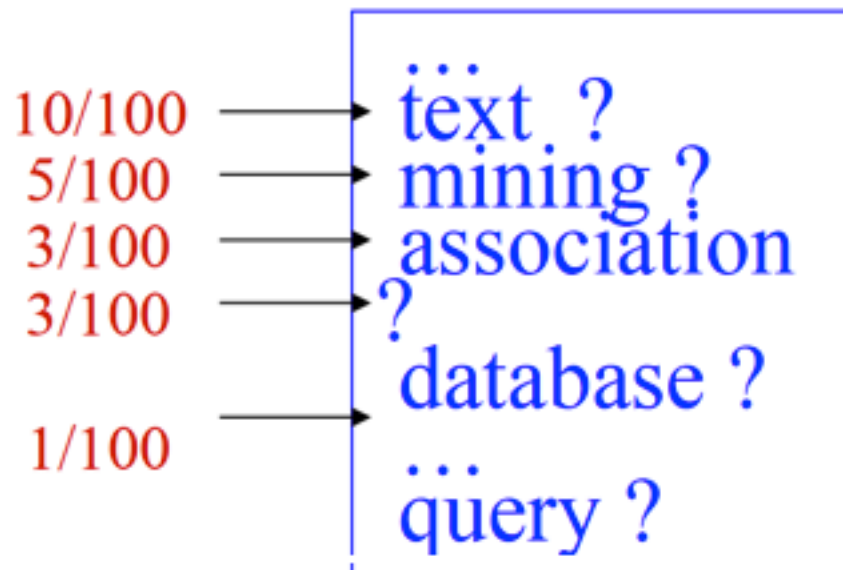
Estimation of Unigram LM

Unigram LM $p(w|\theta)=?$

Estimation

Text Mining Paper d

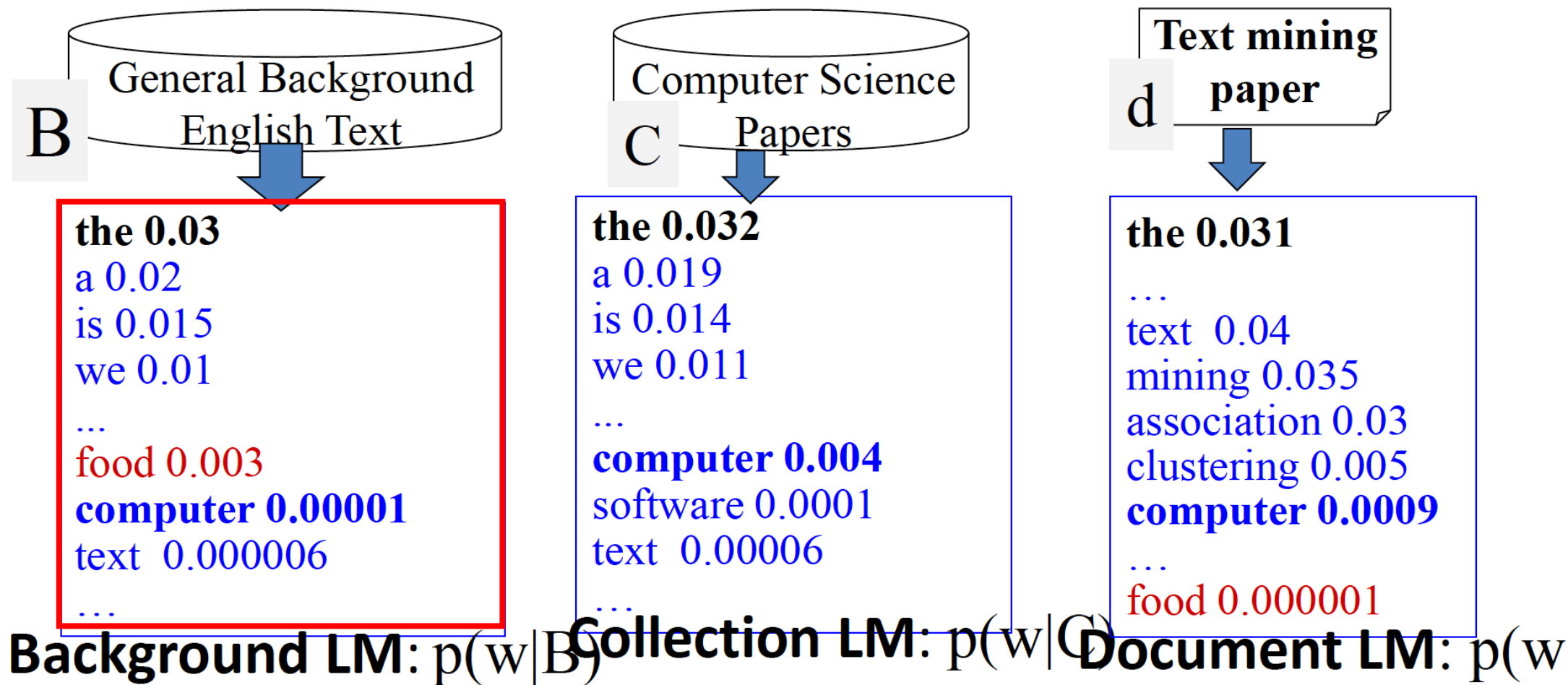
Total #words=**100**



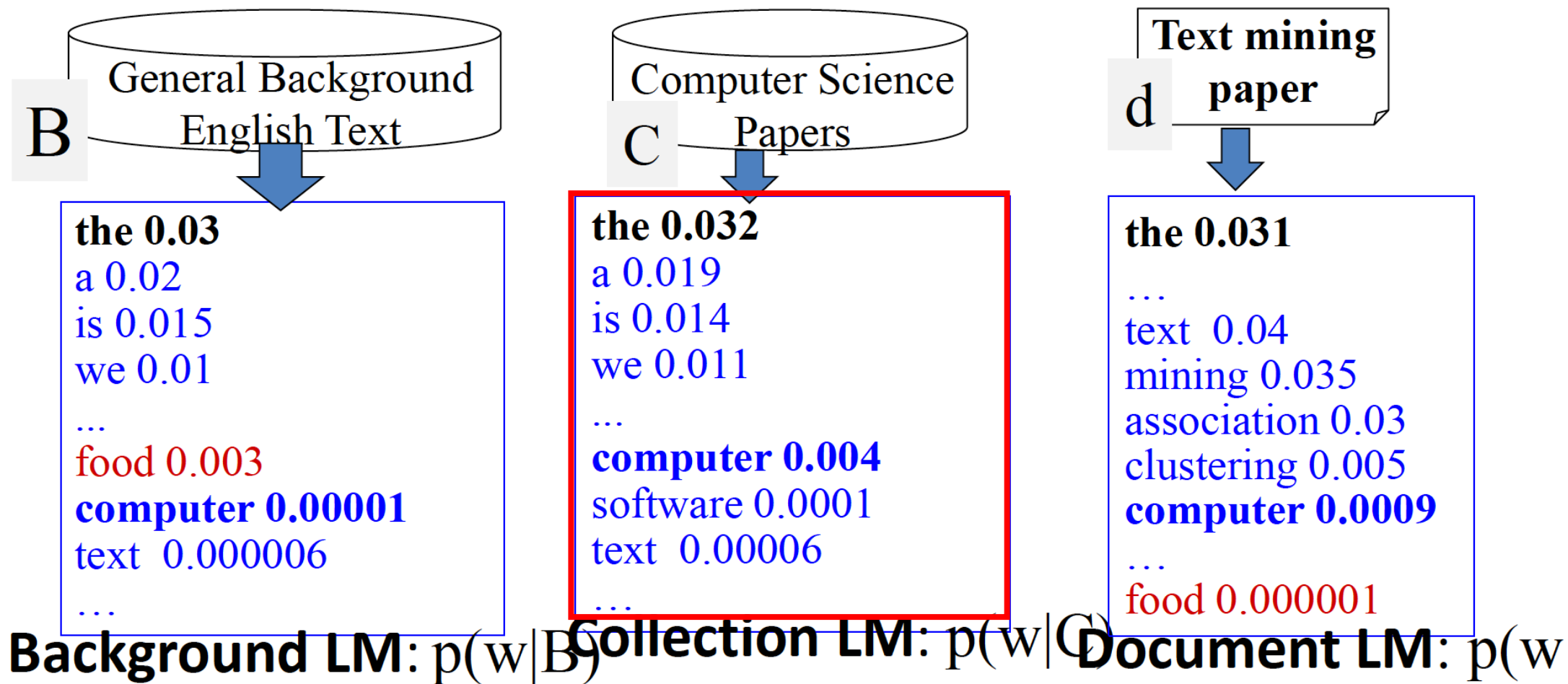
Maximum Likelihood (ML) Estimator:

$$p(w | \theta) = p(w | d) = \frac{c(w, d)}{|d|}$$

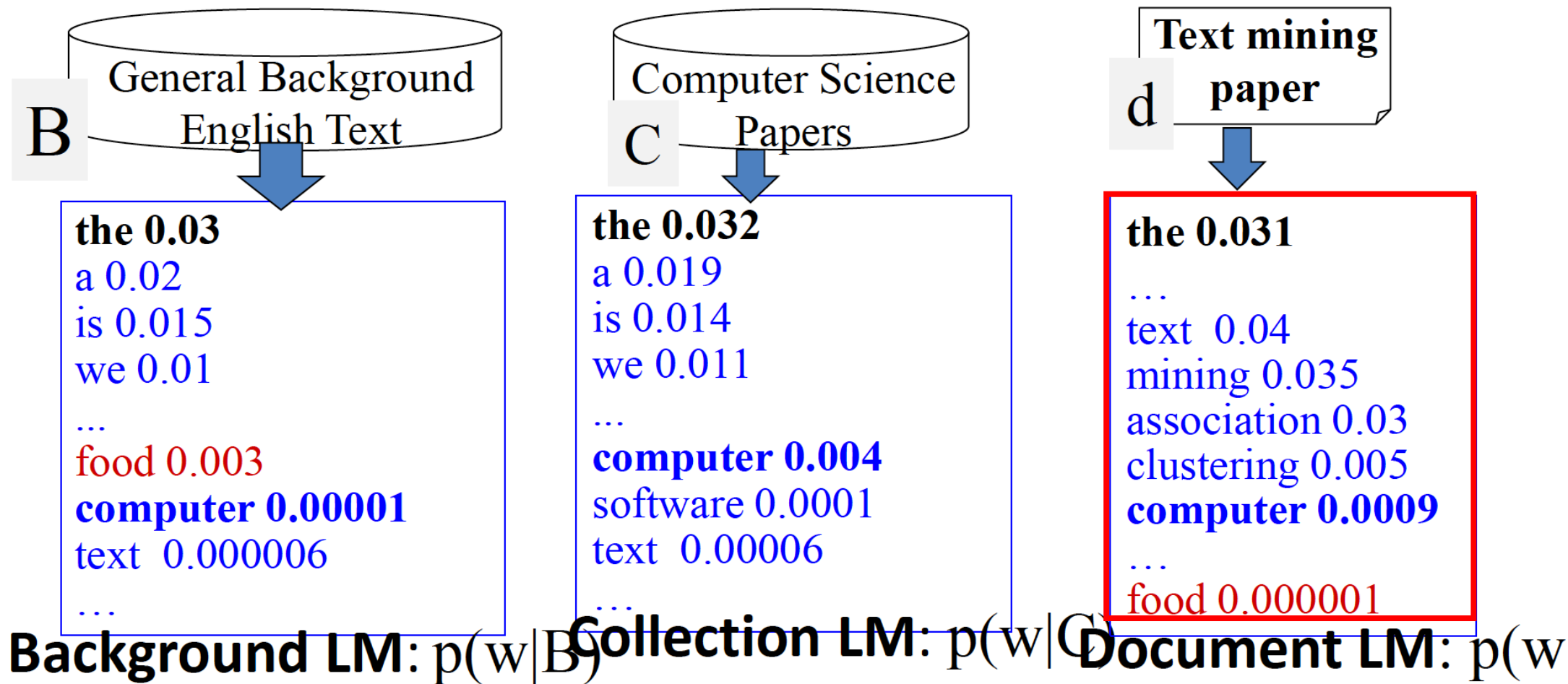
LMs for Topic Representation



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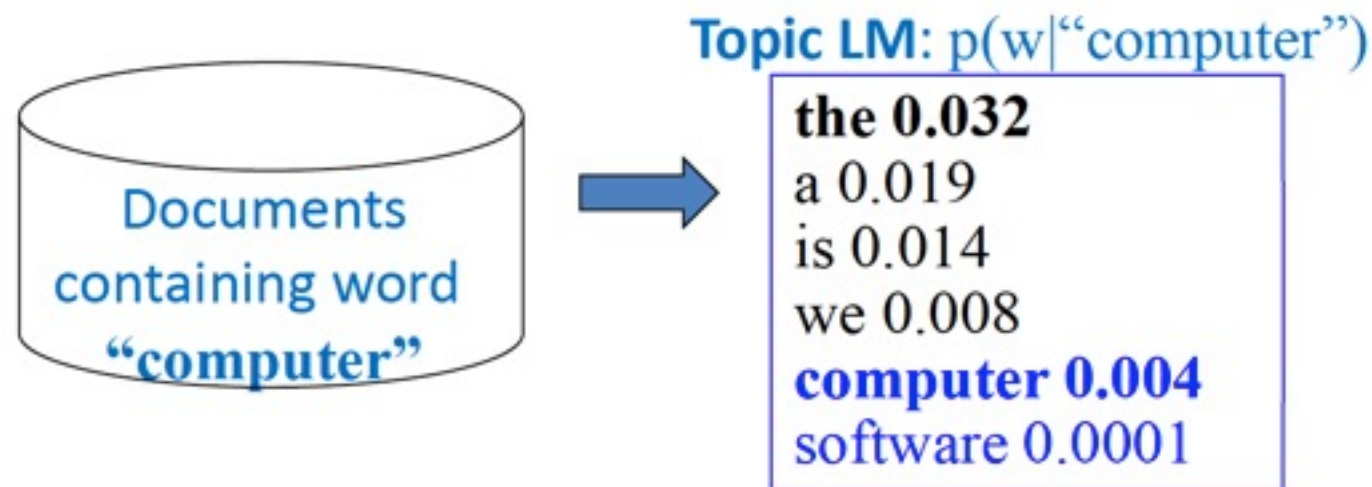


LMs for Topic Representation



LMs for Association Analysis

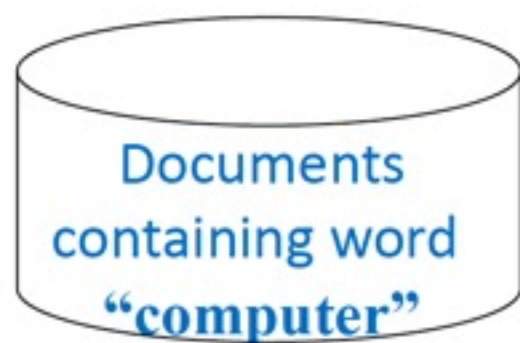
What words are semantically related to “computer”?



LMs for Association Analysis

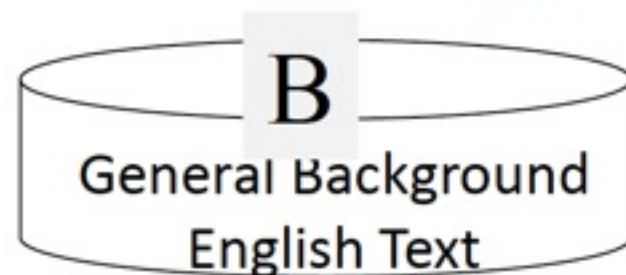
What words are semantically related to “computer”?

Topic LM: $p(w|\text{“computer”})$



the 0.032
a 0.019
is 0.014
we 0.008
computer 0.004
software 0.0001

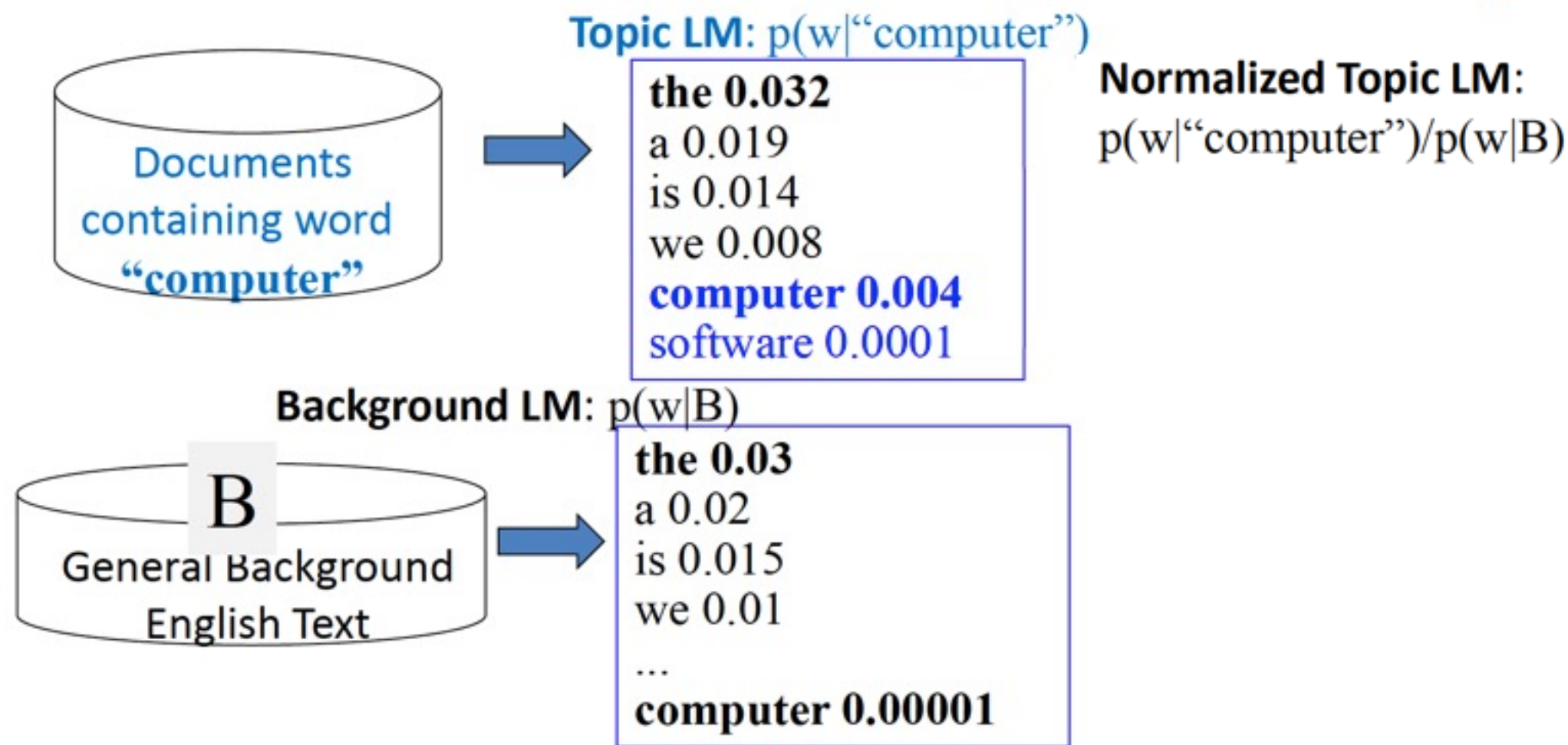
Background LM: $p(w|B)$



the 0.03
a 0.02
is 0.015
we 0.01
...
computer 0.00001

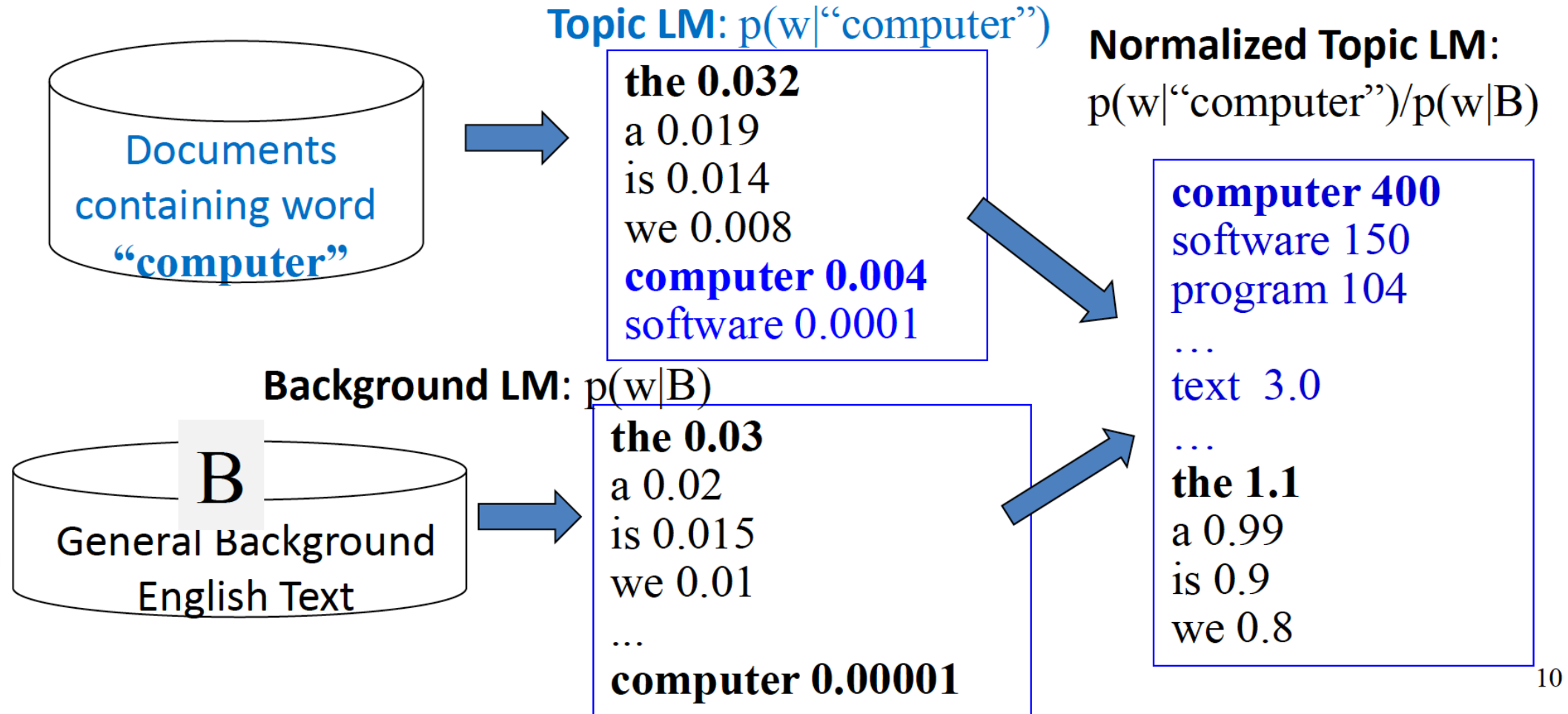
LMs for Association Analysis

What words are semantically related to “computer”?



LMs for Association Analysis

What words are semantically related to “computer”?



Summary

- Language Model = probability distribution over text
- Unigram Language Model = word distribution
- Uses of a Language Model
 - Representing topics
 - Discovering word associations

Additional Readings

- Chris Manning and Hinrich Schütze, Foundations of Statistical Natural Language Processing, MIT Press. Cambridge, MA: May 1999.
- Rosenfeld, R., "Two decades of statistical language modeling: where do we go from here?," *Proceedings of the IEEE* , vol.88, no.8, pp.1270,1278, Aug. 2000