Latent Topic Models for Hypertext

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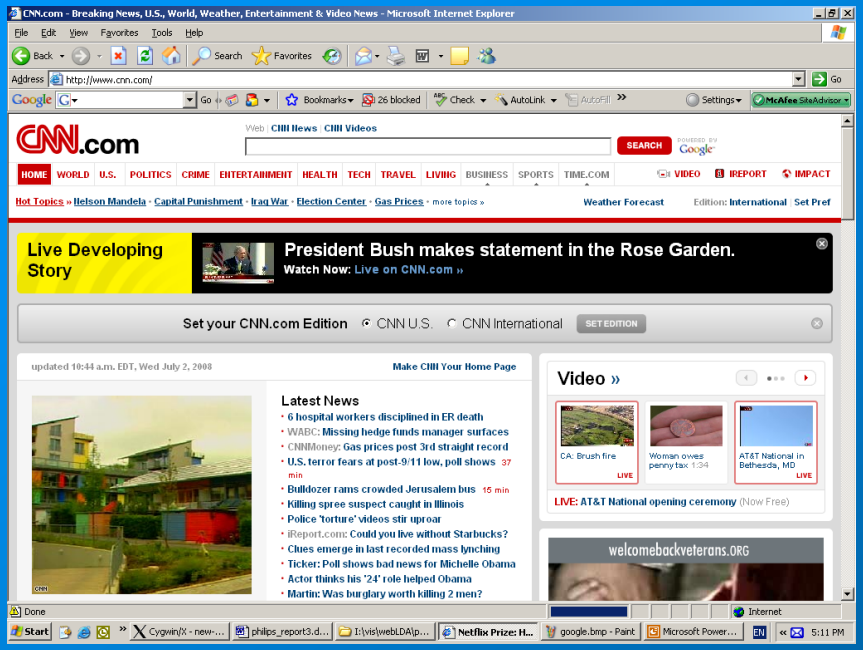
Introduction



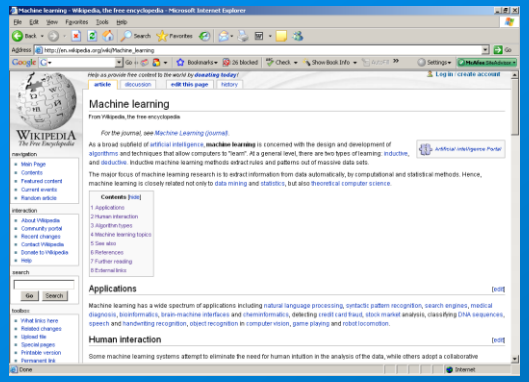
0.3 sports

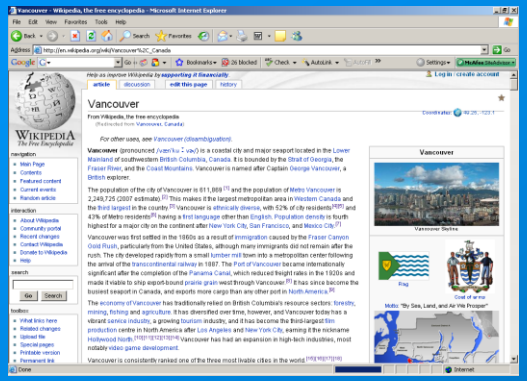
0.4 crime

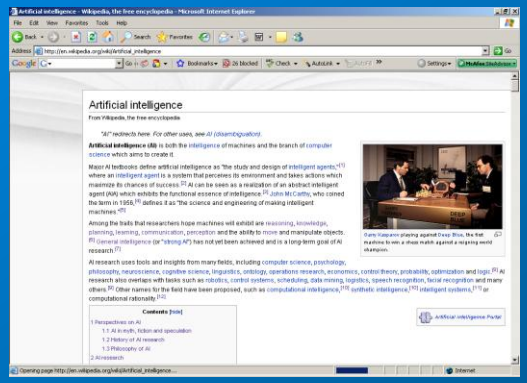
0.3 politics

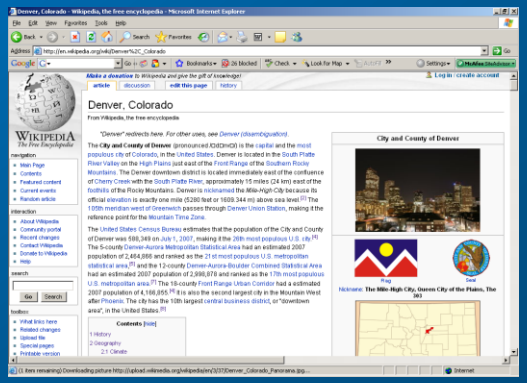
- In this work we focus on hypertext documents, i.e. documents with links

Hypertext Documents

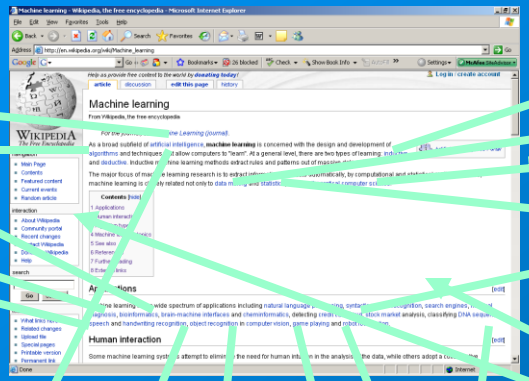


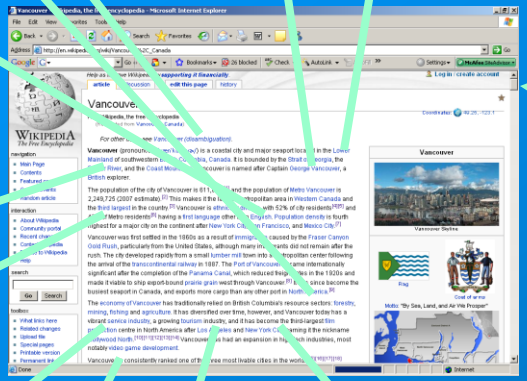


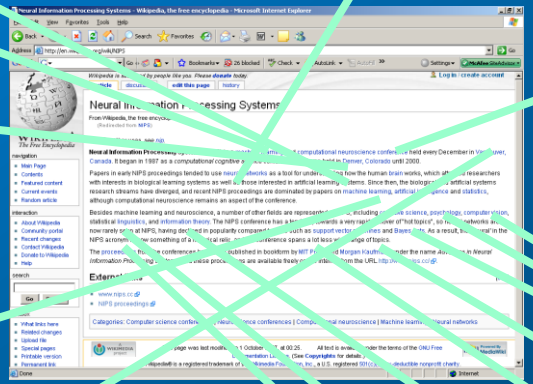


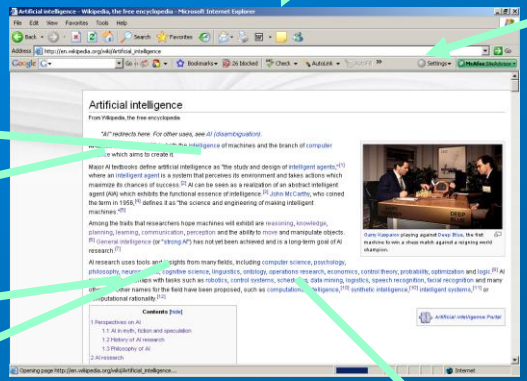


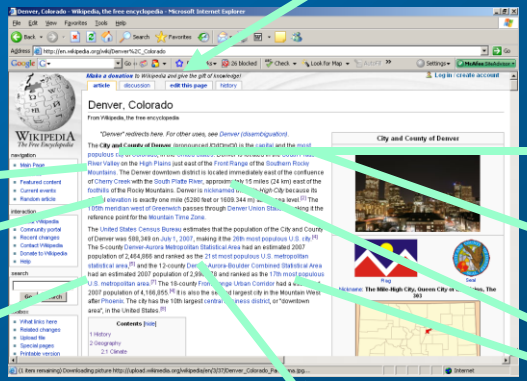
Hypertext Documents





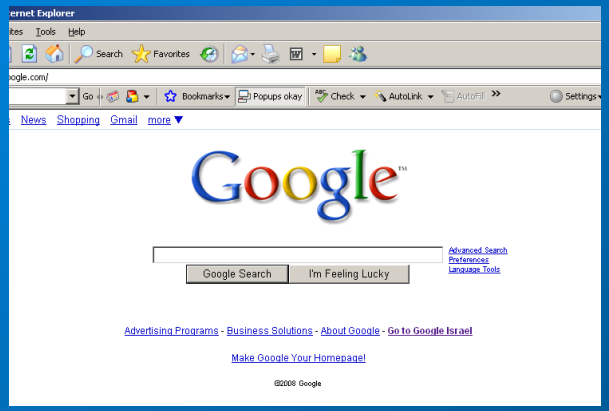






Introduction

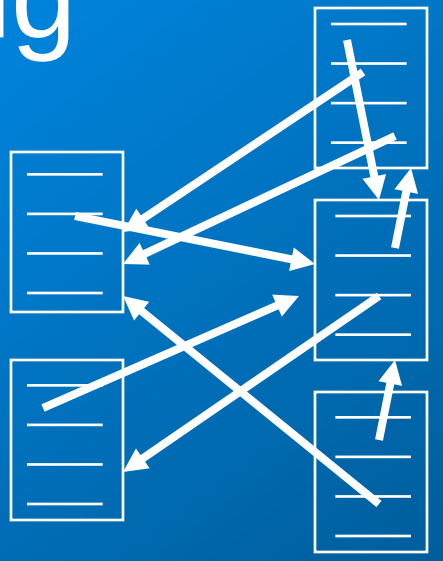
- Hypertext is everywhere !
  - web pages, refs. in scientific publications
- Connectivity is important
  - PageRank



- Topology of the WWW is complicated

Problem Setting

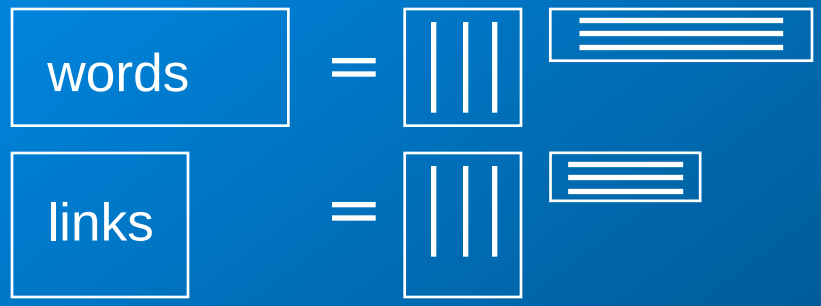
- Input: documents and links



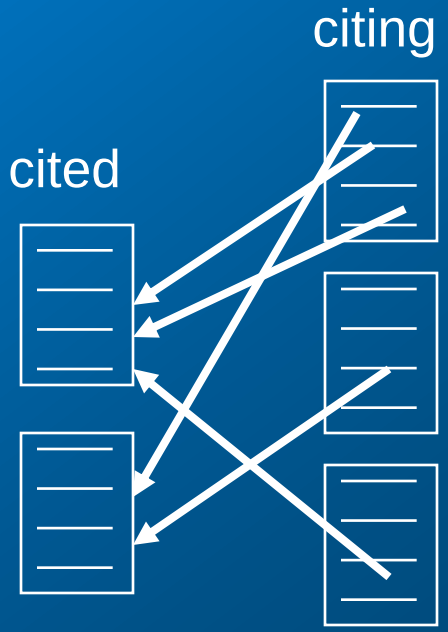
- Estimate:
  - Document topic mixture
  - Pr(word | topic)
  - Document importance
- Unsupervised

Previous Work: Topic Models for Hypertext

- Cohn and Hofmann, ’01.
- Erosheva et al. ’04.
  - Links are modeled similar to words
  - Links are not associated with words



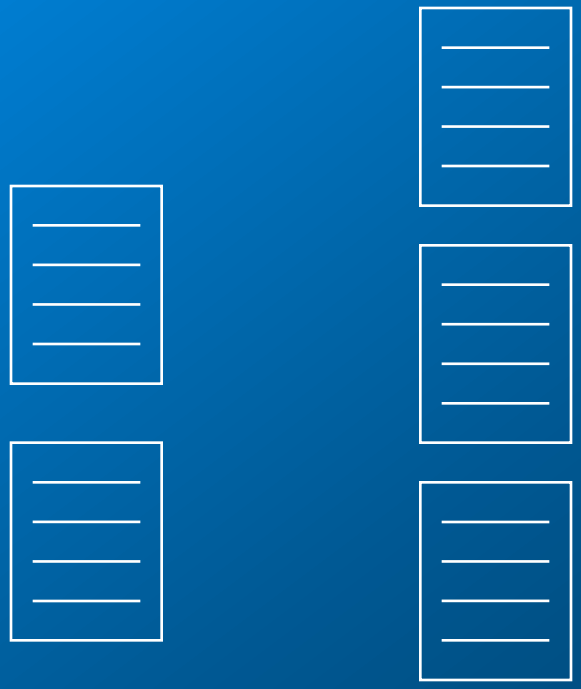
- Dietz et al. ’07.
- Nallapati and Cohen, ’08.
  - Distinguish between citing and cited docs



The Latent Topic Hypertext Model (LTHM)

LTHM: Generative Model

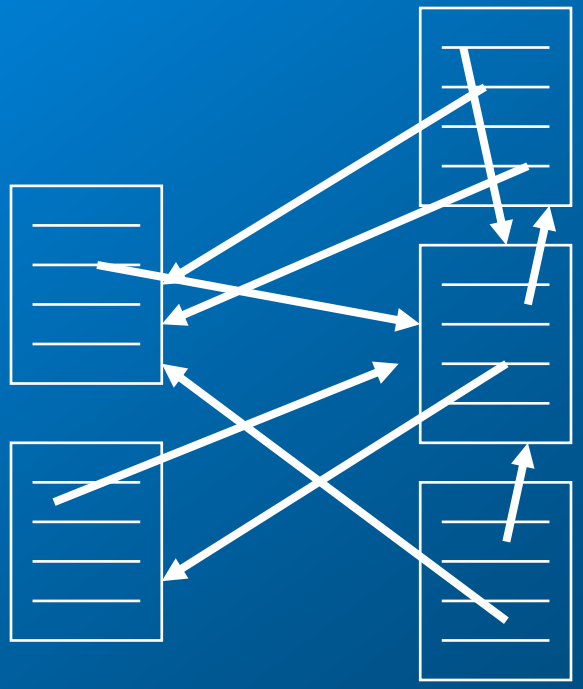
1. Words are created (by LDA)



LTHM: Generative Model

1. Words are created (by LDA)

2. Links are created (our contribution)



LTHM: Modeling Links

- Allows for arbitrary topology of the citation graph (including self links)
- A link points from a word to a document

LTHM: Link Generation

- Depends on:
  - The topic of the anchor word
  - The topic mixture of the target document
  - The importance of the target document

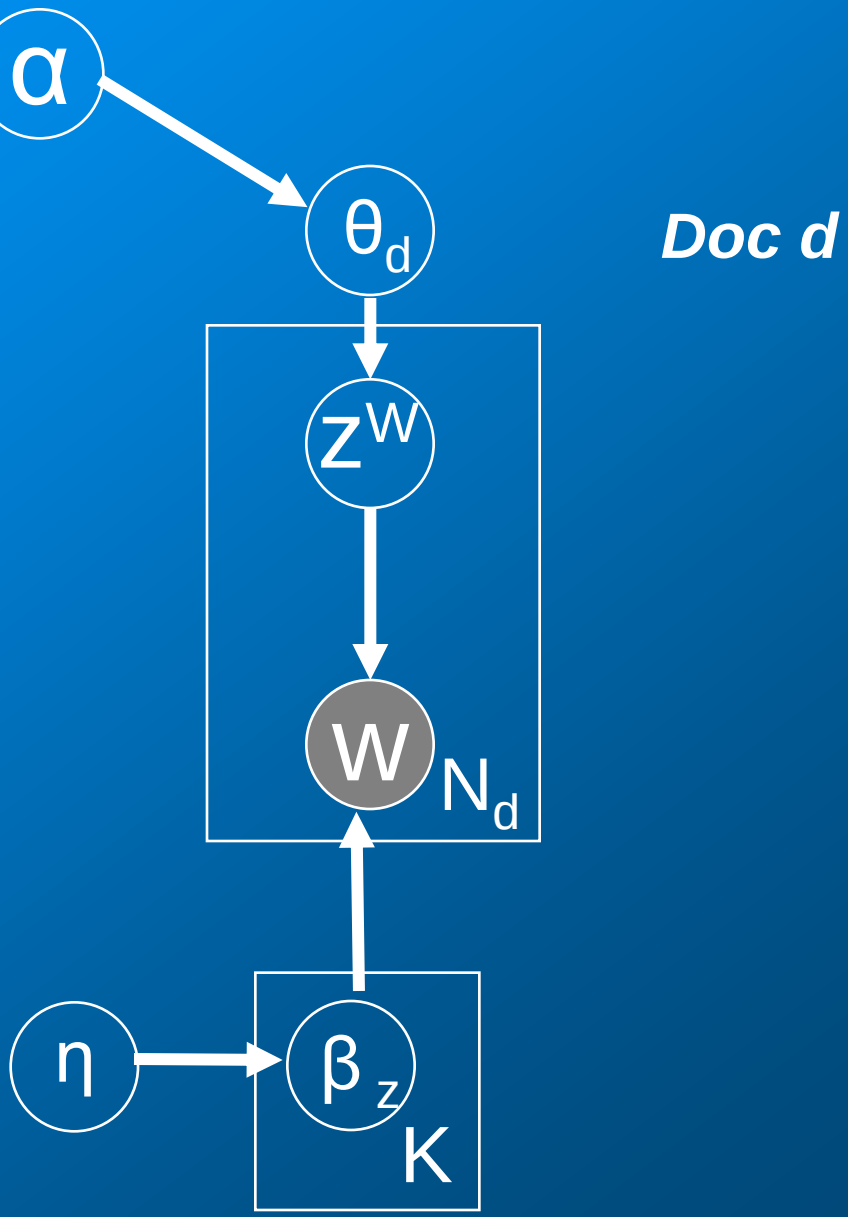
$$\Pr(\text{link} = d \mid \text{topic} = z) = \lambda_d \theta_d(z)$$

importance of d

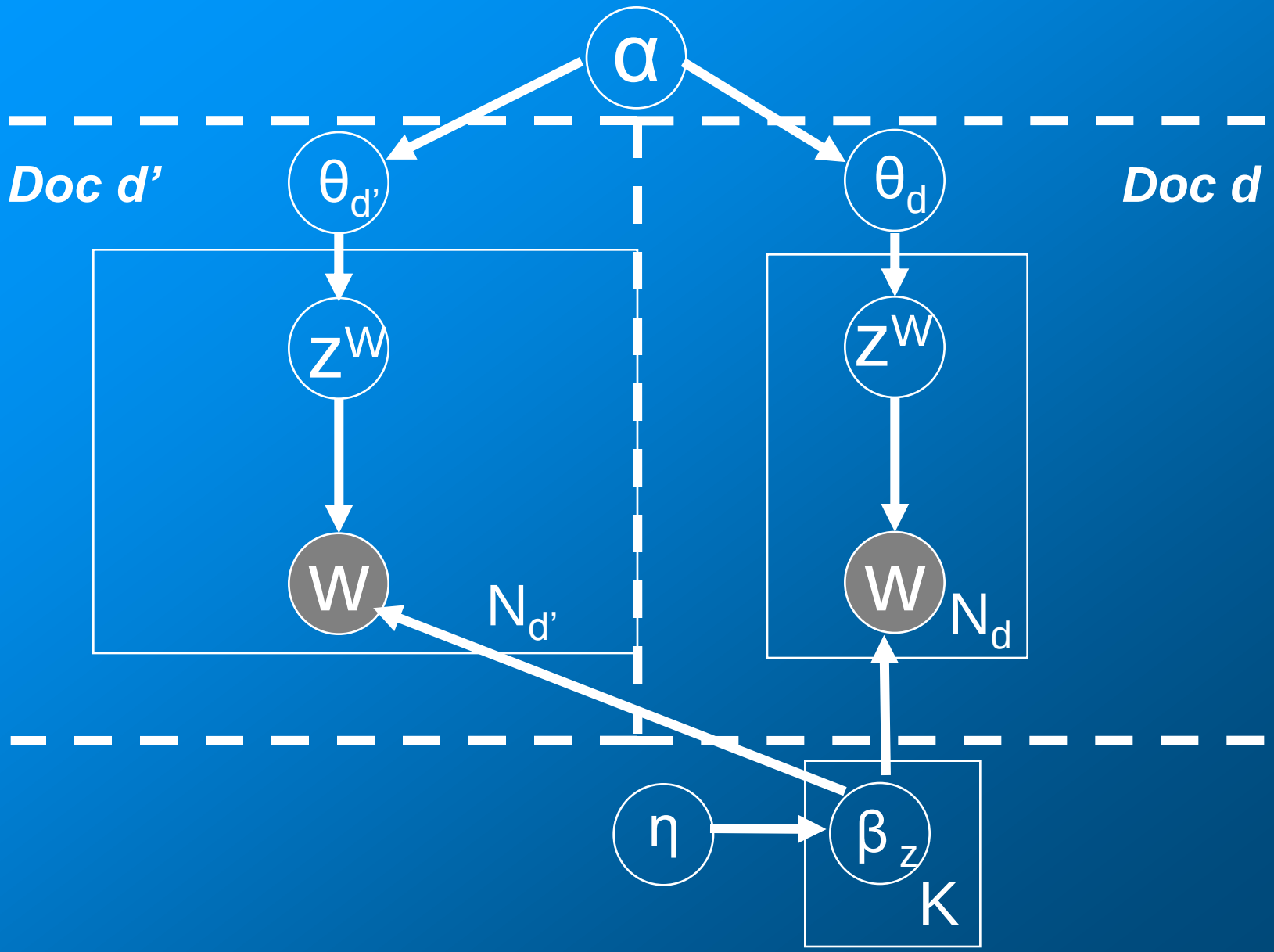
prevalence of z in d

Generating a single document d

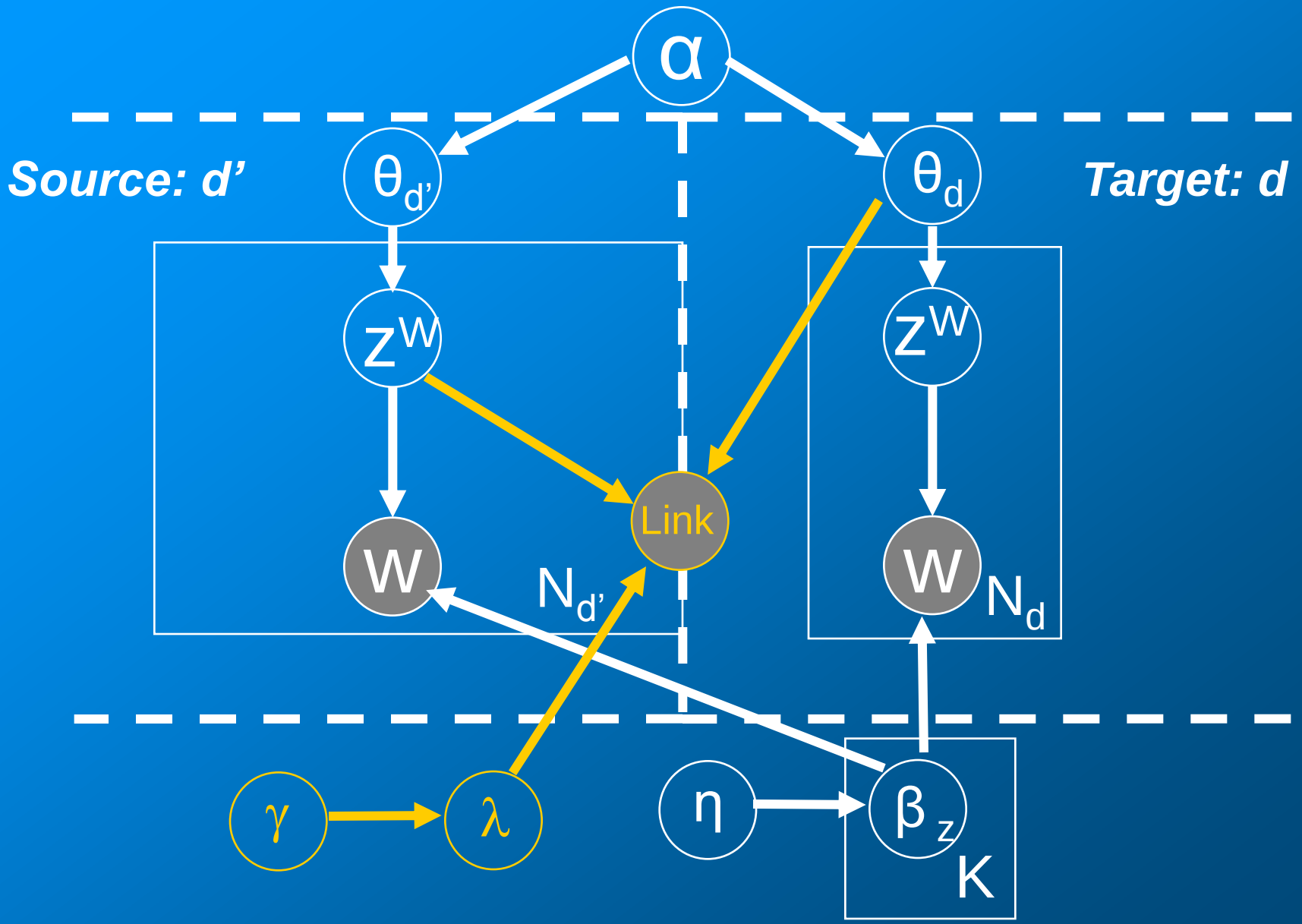
d is generated by Latent Dirichlet Allocation



Generating two documents d' and d



Generating Links from d' to d



Properties of the Model

- D additional parameters (λ) for links vs. $D_x K$ parameters in previous models
- The existence (or non-existence) of a link is an observation
- A link shares the same topic with the word
- Link affects topic estimation in both the source and target documents

Approximate Learning

- Exact inference is intractable in hierarchical models such as LDA
- Approximate inference in LTHM is even more challenging as non-links are also observations
- Using symmetries, we derived an O(K*corpus size) EM algorithm

Experiments

- WebKB dataset
  - 8282 documents
  - 12911 links

- Wikipedia
  - A new data set, collected by crawling from the NIPS Wikipedia page
  - 105 documents
  - 790 links

Experiments: Wikipedia

Topic 1	
neural	0.067
network	0.047
networks	0.039
learning	0.027
artificial	0.017
data	0.015
models	0.014
function	0.014

Artificial neural network 0.004

Neural network 0.003

Experiments: Wikipedia

Topic 2

recognition	0.058
speech	0.033
language	0.015
pattern	0.012
handwriting	0.011
evaluation	0.010
robots	0.010
systems	0.009

Speech recognition

0.004

Pattern recognition

0.004

Experiments: Wikipedia

Topic 3	
vancouver	0.051
denver	0.043
city	0.041
retrieved	0.024
colorado	0.011
area	0.009
population	0.009
canada	0.008

Denver, Colorado

0.0008

Vancouver

0.0002

Experiments: Wikipedia

Topic 4

brain	0.047
cognitive	0.026
science	0.016
press	0.011
neurons	0.010
mind	0.010
systems	0.010
human	0.010

Cognitive science 0.003

Neuroscience 0.002

Journal of Machine Learning Research

LDA

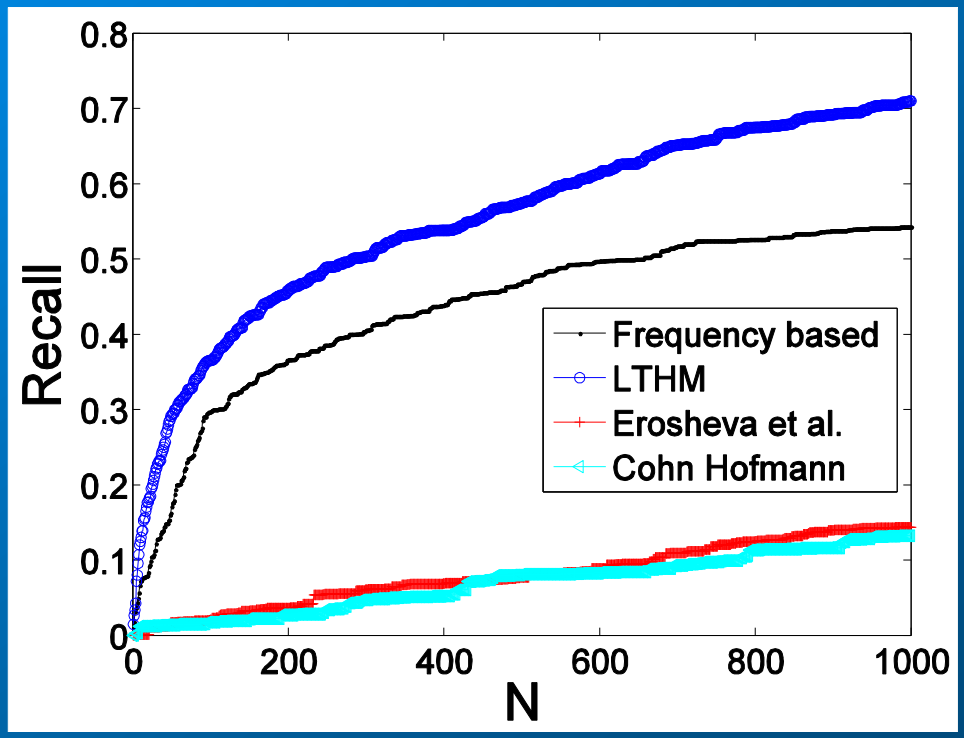
Topic prob	Top words
0.1504	search, article, navigation
0.0798	press, university, new
0.0652	learning, machine, algorithms
0.0594	fixes, skins, import
0.0533	model, regression, reasoning

LTHM

Topic prob	Top words
0.4136	learning, machine, engineering
0.0943	card, conference, credit

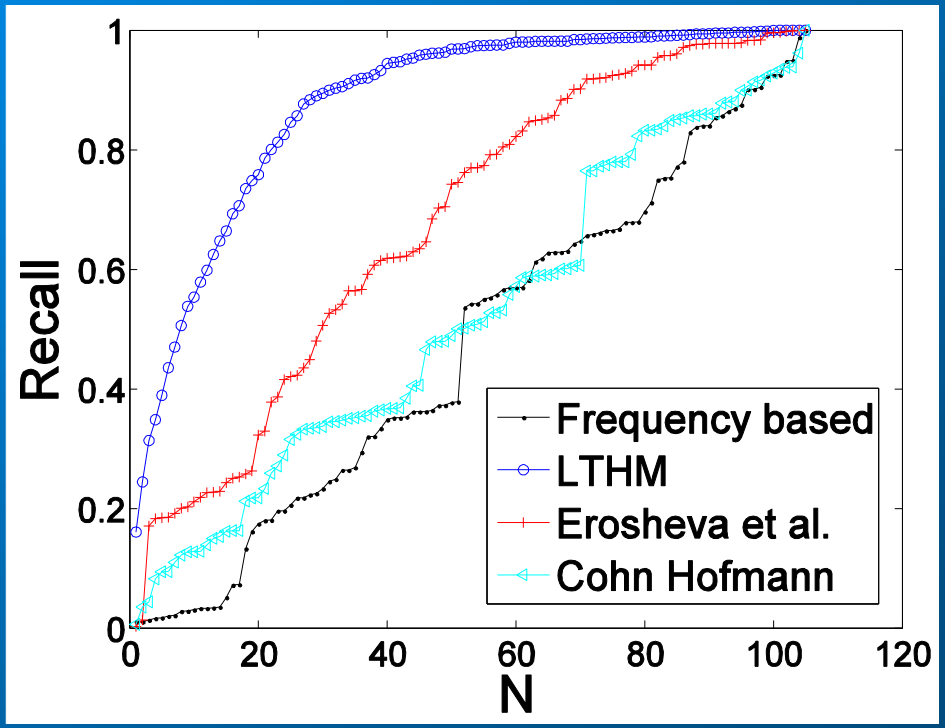
Experiments – link prediction on test set

- Wikipedia corpus: 105 documents with 790 links
- 20 hidden aspects
- Test set: 11 documents, outgoing links are invisible



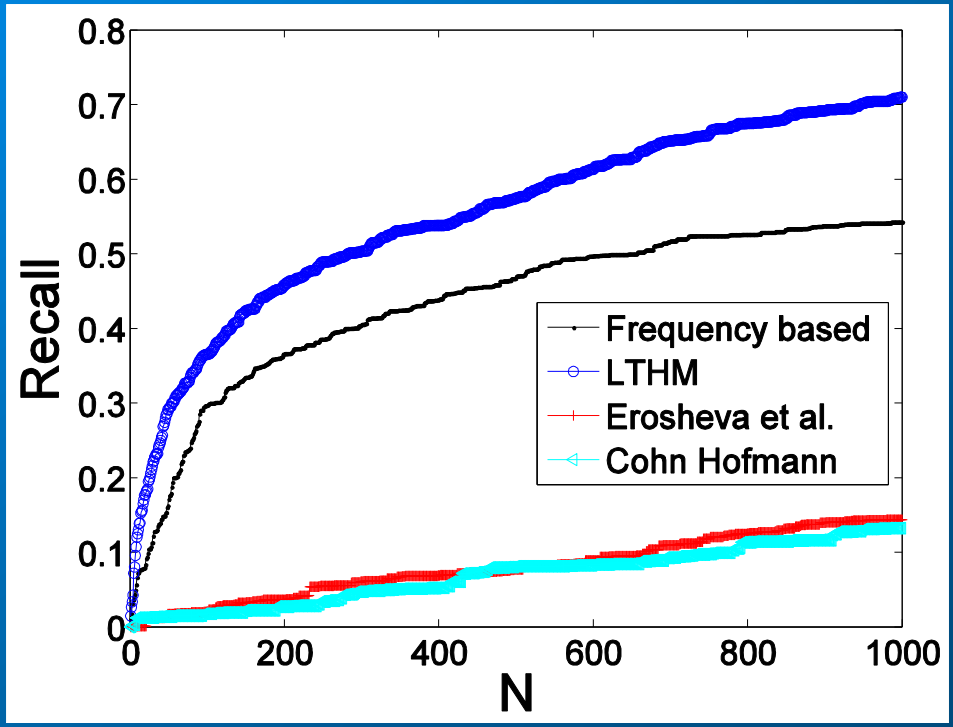
Experiments – link prediction on train set

- Wikipedia corpus: 105 documents with 790 links
- 20 hidden aspects



Experiments – link prediction

- Webkb corpus: 8282 documents with 12911 links
- 20 hidden aspects
- Test set: 10%



Summary

- Explicit modeling of link generation in an LDA like model
- Efficient approximate inference algorithm
- Performs better than previous topic models in link recommendation

- Code and data available online at http://www.cs.huji.ac.il/~amitg/lthm.html