### Clustering the Tagged Web





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WSDM 2009

Images from del.icio.us, Ibaumann.com, www.hometrainingtools.com

halances and more Or make it



CART | MY ACCOUNT | WISH LIST

Information about catalog pricing changes in 2008,-800-860-6272





Words: information about catalog pricing changes in 2008 welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals

molecular models, dial



Monthly Specials

Words: information about catalog pricing changes in 2008 welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals

Anchor Text: home science tools hometrainingtools.com links click follow supplies training experiments other pages



Science Kits for Curriculum

MORE SHOPPING

Science Kits Great Gift Ideas

Bestsellers New Products

Monthly Specials

Nature Backpack Kits

Words: information about catalog pricing changes in 2008 welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals

Anchor Text: home science tools hometrainingtools.com links click follow supplies training experiments other pages

Supp.

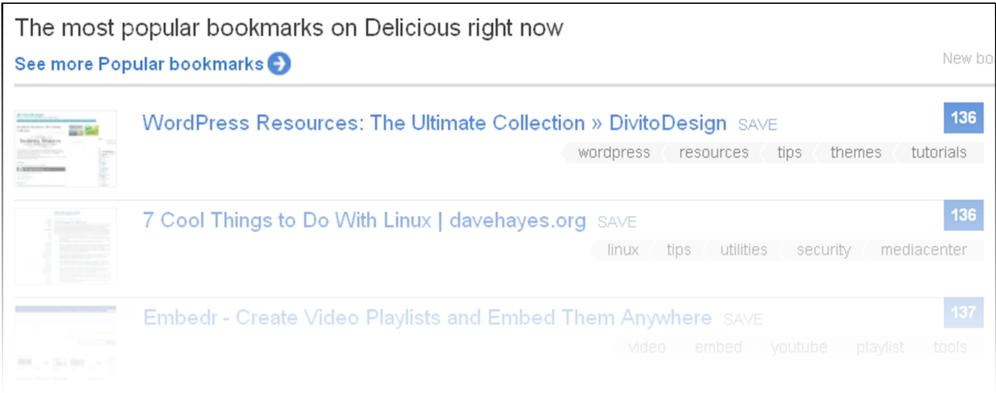
Make to class movith sall experiments stock you with temperature the sall experiments of the sa

**Tags:** science homeschool education shopping curriculum homeschooling experiments tools chemistry supplies

# Why tags? – del.icio.us



# Why tags? – del.icio.us



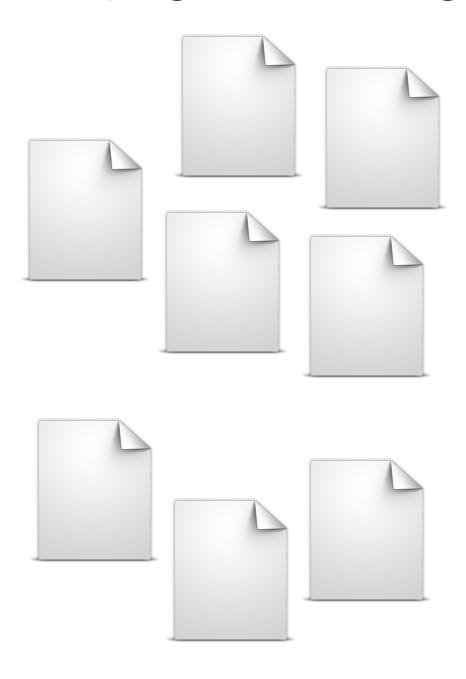
≈120,000 posts / day 12-75 million (≈10 $^{7}$ –10 $^{8}$ ) unique URLs (versus 10 $^{9}$ –10 $^{11}$  total URLs)

Disproportionately the web's most useful URLs (and those URLs have many tags)

### Using tags to understand the web

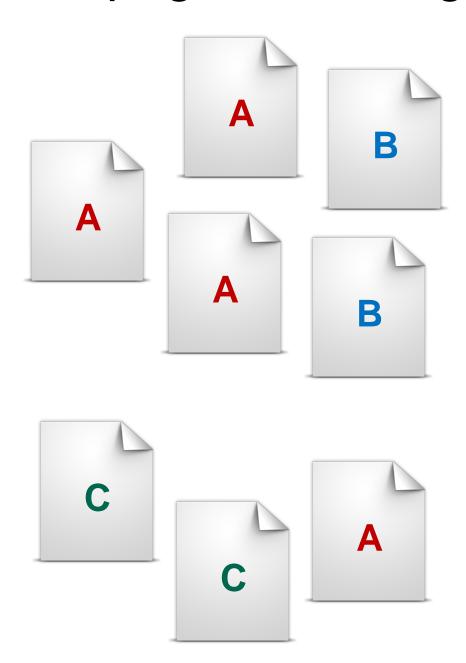
- The web is large and growing: anything that helps us understand high level structure is useful
- Tags encode semantically meaningful labels
- Tags cover much of the web's best content
- How can we use tags to provide high-level insight?

# Web page clustering task



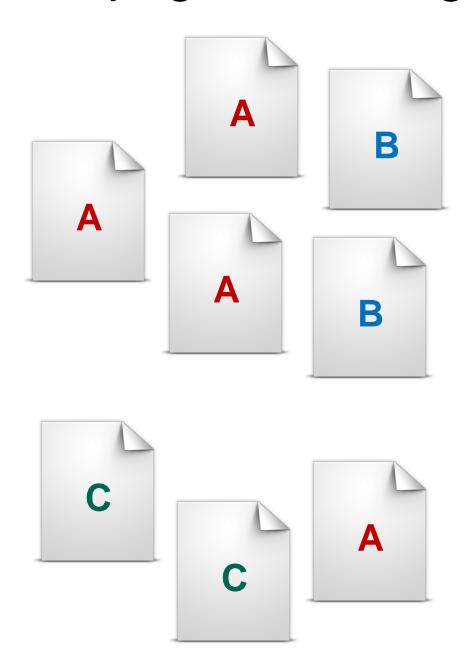
Given a collection of web pages

### Web page clustering task



- Given a collection of web pages
- Assign each page to a cluster, maximizing similarity within clusters

# Web page clustering task



- Given a collection of web pages
- Assign each page to a cluster, maximizing similarity within clusters
- Applications: improved user interfaces, collection clustering, search result diversity, language-model based retrieval

### Structure of this talk

#### **Features**

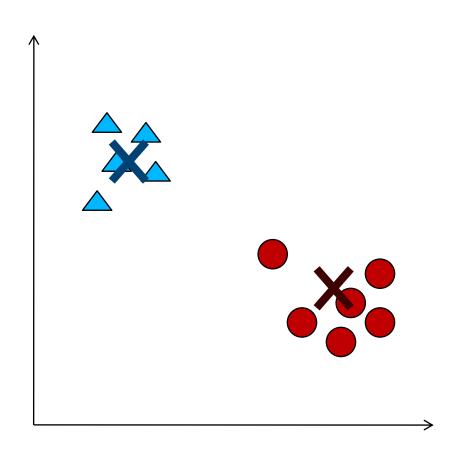
	Words	Tags	Anchors
Vector Space Model: K-means			
Generative Model: MM-LDA			

#### Models: K-means and MM-LDA

#### Features

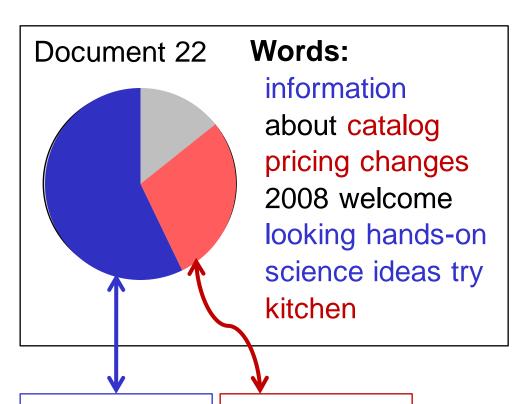
	Words	Tags	Anchors
Vector Space Model: K-means			
Generative Model: MM-LDA			

# Model 1: K-means clustering



- K-means assumes the standard Vector Space Model: documents are Euclidean normalized real-valued vectors
- Algorithm: iteratively
  - Re-assign documents to closest cluster centroid
  - Update cluster centroids from document assignments

#### Model 2: Latent Dirichlet Allocation



 LDA assumes each document's words generated by some topic's word distribution

# Topic 5 science experiment learning

ideas

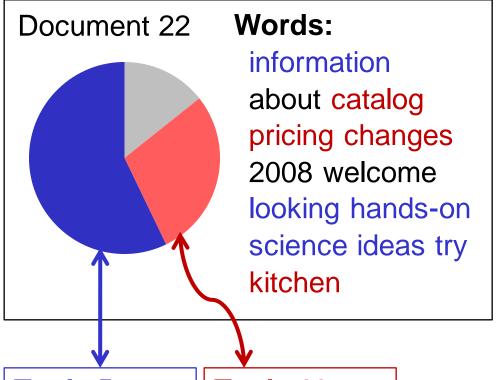
practice information

#### Topic 12

catalog shopping buy Internet checkout cart

. . .

### Model 2: Latent Dirichlet Allocation



#### **Topic 5**

science experiment learning ideas practice information

#### Topic 12

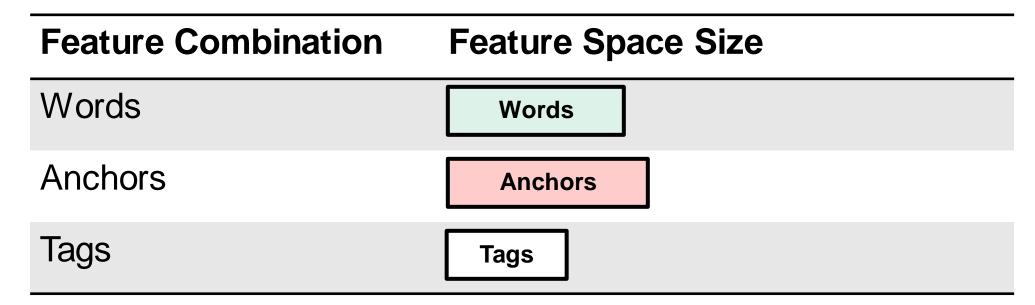
catalog shopping buy Internet checkout cart

- LDA assumes each document's words generated by some topic's word distribution
- Paired with an inference mechanism (Gibbs sampling), learns perdocument distributions over topics, per-topic distributions over words

# Features: words, anchors, and tags

#### **Features**

	Words	Tags	Anchors
Vector Space Model: K-means			
Generative Model: MM-LDA			



Feature Combination	Feature Space Size
Words	Words
Anchors	Anchors
Tags	Tags
Tags as Words	Tags as Words  Anchors as Words
	Tags & Anchors as Words

Feature Combination	Feature Space Size
Words	Words
Anchors	Anchors
Tags	Tags
Tags as Words	Tags as Words
Tags as New Words	Words Tags
	Words Anchors
	Words Tags Anchors

#### **Feature Combination**

#### **Feature Space Size**

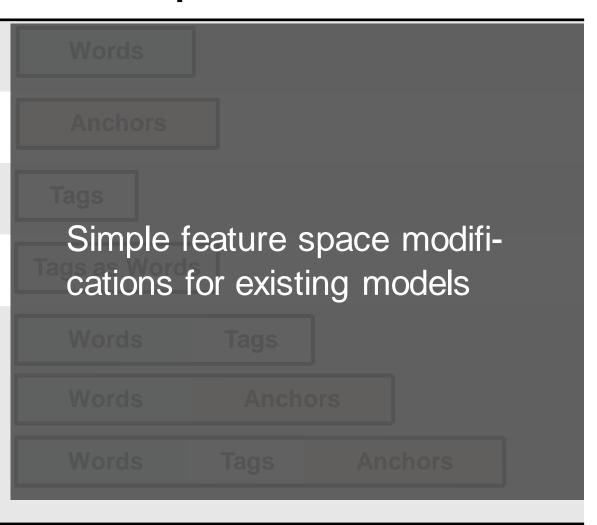
Words

**Anchors** 

Tags

Tags as Words

Tags as New Words



<b>Feature Combination</b>	Feature Space Size
Words	Words
Anchors	Anchors
Tags	Tags
Tags as Words	Tags as Words
Tags as New Words	Words Tags
Words + Tags	Words Tags
	Words Anchors
	Words Tags Anchors

Feature Combination	Feature Space Size	
Words	Words	
Anchors	Anchors	
Tags	Tags	
Tags as Words	Tags as Words	
Tags as New Words	Words Tags	
Words + Tags	K-means: normalize feature input vectors independently LDA: multiple parallel sets of observations via MM-LDA	

#### **Features**

		Words	Tags	Anchors
	Vector Space Model: K-means	1. Combine words a sin the V	and tags	
) ) ) :	Generative Model: MM-LDA			

#### **Features**

Vector Space
Model:
K-means

Generative
Model:
MM-LDA

Words

Tags

Anchors

2. Comparing
models, at
multiple levels of
specificity

#### **Features**

Words Tags Anchors

Vector Space
 Model:
 K-means
 Generative
 Model:
 MM-LDA

Words Tags Anchors

and tags
 complement or substitute
 for anchor text?

### Experimental Setup

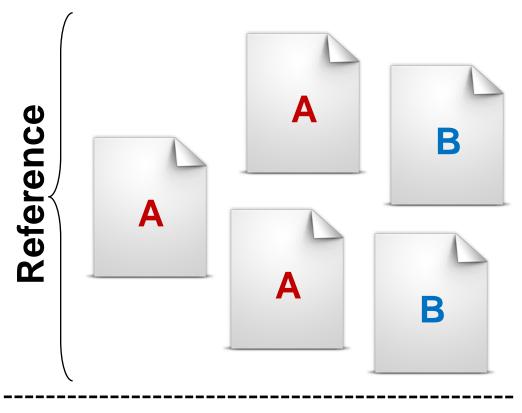
- Construct surrogate "gold standard" clustering using Open Directory Project
- Reflects a (problematic) consensus clustering, with known number of clusters

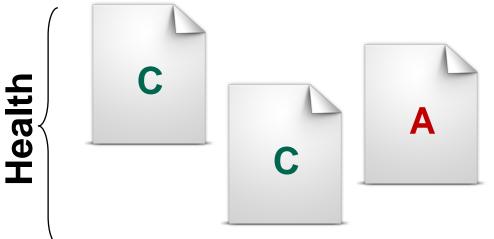
ODP Category	# Documents	Top Tags
Computers	5361	web css tools software programming
Health	434	parenting medicine healthcare medical
Reference	1325	education reference time research dictionary

# Experimental Setup

- Score predicted clusterings with ODP, but not trying to predict ODP
- Useful for relative system performance

ODP Category	# Documents	Top Tags
Computers	5361	web css tools software programming
Health	434	parenting medicine healthcare medical
Reference	1325	education reference time research dictionary





Intuition: balance

pairwise precision

(place only similar

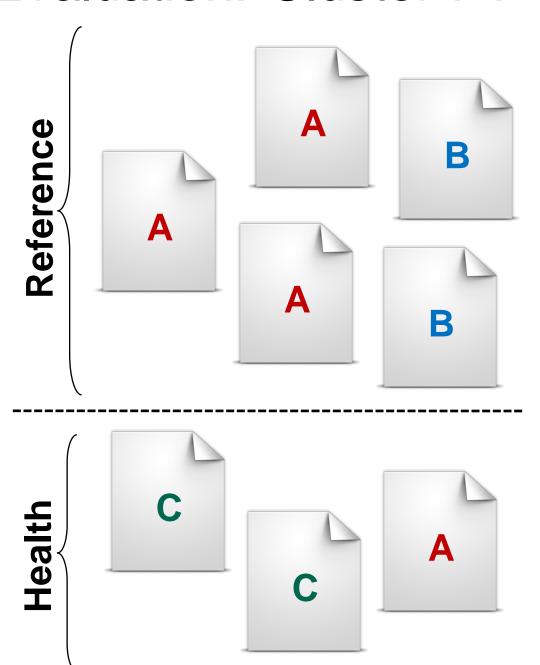
documents together)

with

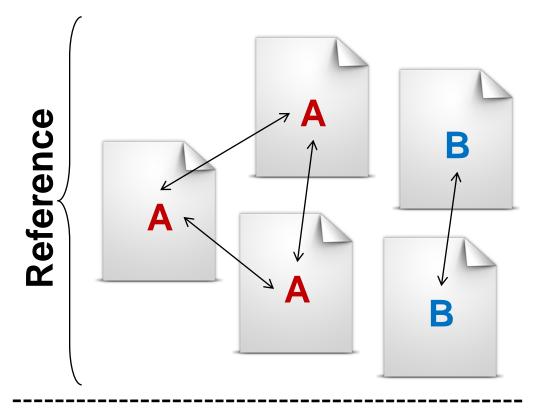
pairwise recall (keep

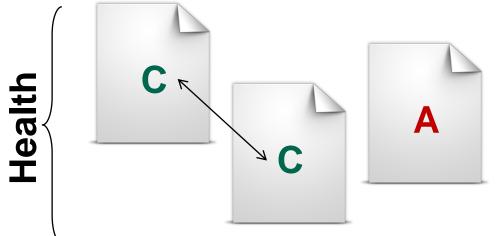
all similar documents

together)

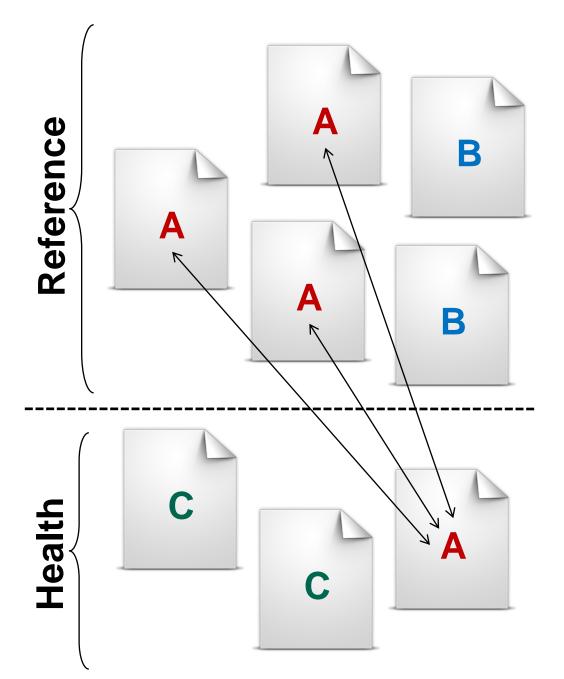


	Same Label	Different Label
Same Cluster		
Different Cluster		



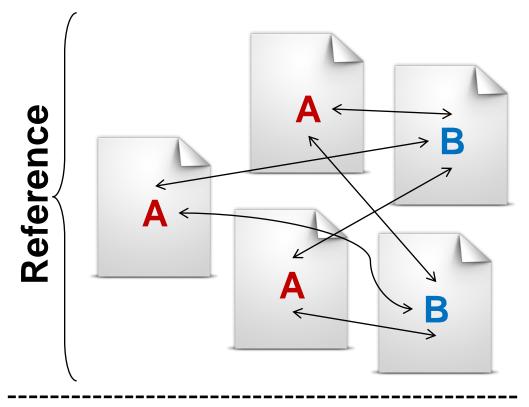


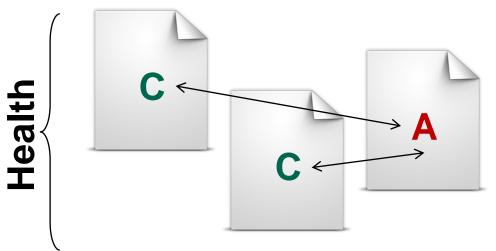
	Same Label	Different Label
Same Cluster	5	
Different Cluster		



	Same Label	Different Label
Same Cluster	5	3
Different Cluster		

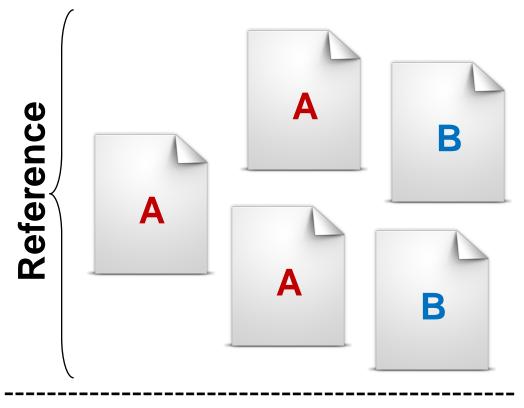
Cluster Precision: 5/8

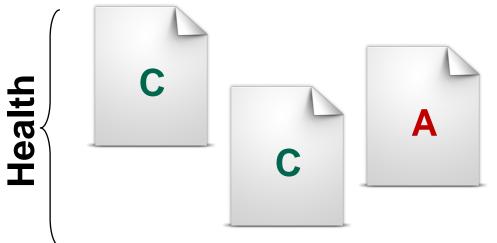




	Same Label	Different Label
Same Cluster	5	3
Different Cluster	8	

Cluster Precision: 5/8
Cluster Recall: 5/13





	Same Label	Different Label
Same Cluster	5	3
Different Cluster	8	

Cluster Precision: 5/8
Cluster Recall: 5/13
Cluster F1: .476

#### **Features**

		Words	Tags	Anchors
	Vector Space Model: K-means	1. Combining words and tags in the VSM		
) ) ) :	Generative Model: MM-LDA			

# Result: normalize words and tags independently in the Vector Space Model

Features				K-means	
Words	Words		.139		
Tags	Tags		.219		
Words+Tags	Words		Tags	.225	

Possible utility for other applications of the VSM

# Result: normalize words and tags independently in the Vector Space Model

Features K-means					
Words	Words			.139	
Tags	Tags			.219	
Words+Tags	Words		Tags	.225	
Tags as Words (x1)	Tags as Words			.158	
Tags as Words (x2)	Tags as Words			.176	
Tags as New Words	Words		Tags	.154	

Possible utility for other applications of the VSM

## Experiments

#### **Features**

Vector Space
Model:
K-means

Generative
Model:
MM-LDA

Words

Tags
Anchors

2. Comparing
models, at
multiple levels of
specificity

Models

# Result: MM-LDA outperforms K-means on top-level ODP categories

Features				K-means	(MM-)LDA
Words	Words	<b>S</b>		.139	.260
Tags	Tags		•	.219	.270
Words+Tags	Words		Tags	.225	.307

# Tagging at multiple basic levels

People use tags to help find the same page later, often at a "natural" level of specificity

## Programming/Languages Society/Social Sciences (1094 documents)

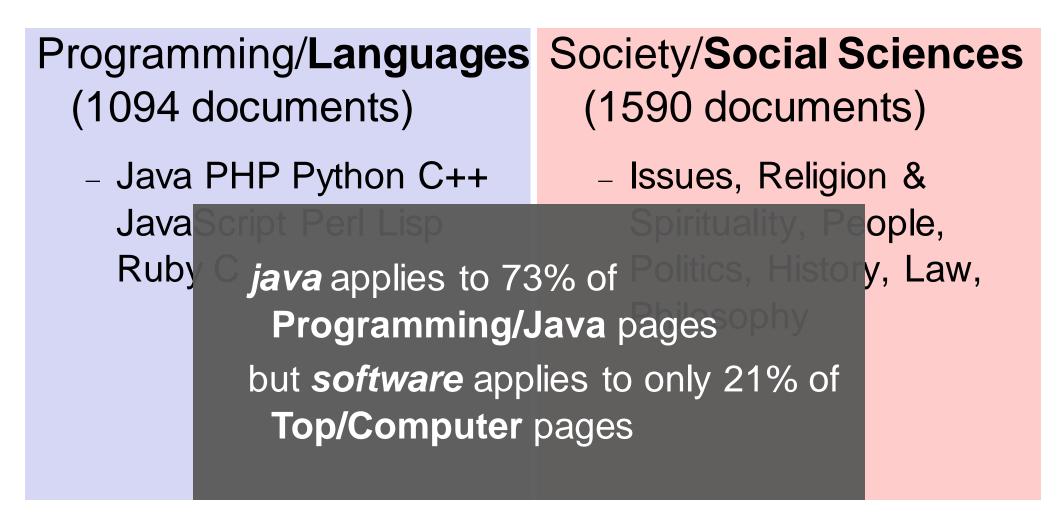
Java PHP Python C++ JavaScript Perl Lisp Ruby C

# (1590 documents)

Issues, Religion & Spirituality, People, Politics, History, Law, Philosophy

# Tagging at multiple basic levels

People use tags to help find the same page later, often at a "natural" level of specificity



# Result: Sometimes, tags tell you more about cluster membership than words do

	Features	K-means	(MM-)LDA
Programming Languages	Words Tags Words+Tags	.189 .567 .556	.288 .463 .297
Social Sciences	Words Tags Words+Tags	.196 .307 .308	.300 .310 .302

- Tags are very discriminating in subcategories
- K-means wins when the feature space is cleaner

## Experiments

#### **Features**

Words Tags Anchors

Vector Space
 Model:
 K-means
 Generative
 Model:
 MM-LDA

Words Tags Anchors

and tags
 complement or substitute
 for anchor text?

Models

# Result: Tags complement anchor text

Features	K-means	(MM-)LDA
Words	.139	.260
Words+Anchors	.128	.248
Words+Anchors+Tags	.224	.306

Anchors can depress performance, but adding tags brings to within delta of Words+Tags.

#### Conclusions

- Tags add real value when high-level semantic information is needed
- Tags act differently than words, anchor text
- At the right level of specificity, tags describe pages better than anything else
- Treat tags and words as separate information channels to maximize utility

#### **Thanks! Questions?**

# Backup material

# Result: Tags complement anchor text

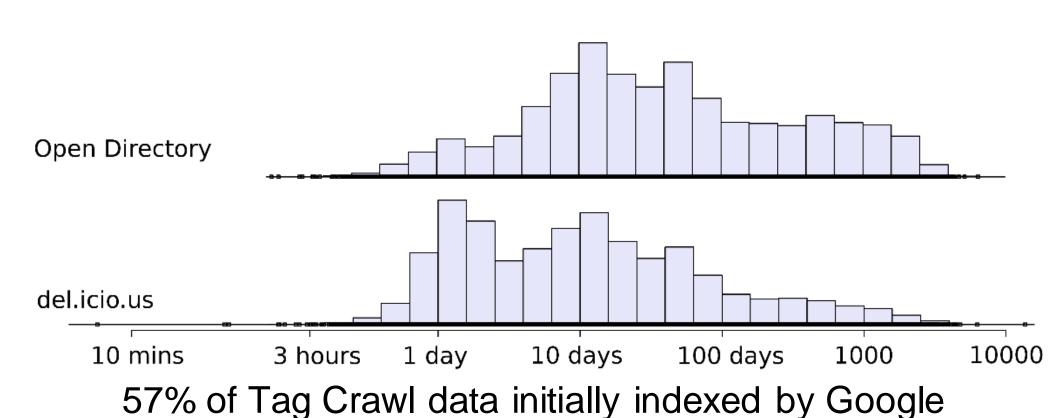
	(MM-)LDA	K-means
Words	.260	.139
Anchors as Words	.270	.120
(Anchors as Words)+Tags	.281	.214
Words+Anchors	.248	.128
Words+Anchors+Tags	.306	.224

Anchor text acts as annotations from another web author. Noisier than words and tags, but can be usfeully integrated into a joint model.

#### Future directions

- More targeted graphical models
  - Individual users with individual vocabularies
  - Time series
- Direct evaluation in retrieval / browsing
- More types of annotated documents
  - Product reviews; academic papers; blog posts

## Content age: ODP versus del.icio.us



# Clustering (flat, parametric)

#### Input

- Number of clusters K
- Set of documents: <words,tags,anchors>

#### Output

Assignment of documents to clusters

#### Evaluation

Comparison to a gold standard

#### **Outline**

- The tagged web
- Dataset and methodology
- Clustering with tags and words
  - K-Means in tag-augmented vector space
  - Multi-Multinomial LDA
- Experiments
- Discussion

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#### Automatic cluster evaluation

- Pick a slice of ODP with k subtrees
- Cluster relevant documents into k sets
- Compare inferred assignments to ODP labeling

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### **Advantages**

- Scalable, automatic, reflects "consensus" clustering

#### **Drawbacks**

- May not translate to performance gains in task
- Does not address choosing best k

# F-measure of cluster quality

# Pairs of Examples	Same cluster	Different cluster
Same ODP Category	A	C
Different ODP Category	В	D

$$P = \frac{A}{A+B} \qquad R = \frac{A}{A \square C} \qquad F_1 = \frac{2 P R}{P \square R}$$

## A tagged document

**Tags** curriculum education(2) homeschool imported learning science(4) shopping slinky teachers teaching tools



#### **ODP Label:** Top/Reference

Top/Reference/Education/K\_through\_12/Home\_Schooling/Curriculum/Science

## MM-LDA implementation

- Collapsed Gibbs-sampler with hard assignments
  - Repeatedly samples new z for each word
  - Usually converges within several dozen passes
  - Could be parallelized

#### • Runtime:

22 min (MM-LDA) versus 6 min (K-means) on 2000 documents

## K-means generated clusters

#### tags

linux security php opensource vpn unix games go game sports firefox gaming music research finance audio mp3 lyrics news business newspaper politics media magazine politics activism travel movies law government

#### words

linux ircd php beware kernel exe dmg munsey ballparks suppes racer game music research redirect nottingham meta laboratory v business leadership d news j aquaculture terrapass geothermal anarchist wwoof cpsc

## MM-LDA generated clusters

#### tags

web2.0 tools online editor photo office guitar scanner chemistry military earthquake groupware health medical medicine healthcare process gardening bible christian space astronomy religion christianity politics activism environment copyright law government

#### words

icons uml powerpoint lucid dreams dreaming grub outlook bittorrent rendering recovery boot exe health openpkg okino dll polytrans gaelic bible nt bone scottish english war shall power prisoners their article

# K-means term weighting

	tf	tf-idf
Words	.131	.152
Tags	.201	.154
Words+Tags	.209	.168

$$\vec{d}_{j}^{(i)} = \begin{cases} f(w^{(i)}, W, j) & if \ 1 \leq j \leq |W| \\ f(t^{(i)}, T, j - |W|) & otherwise \end{cases}$$

$$f_{tf}(\vec{w}^{(i)}, W_{j}) = \frac{1}{2N_{w}} \sum_{k=1}^{N} I[\vec{w}_{k}^{(i)} = W_{j}]$$

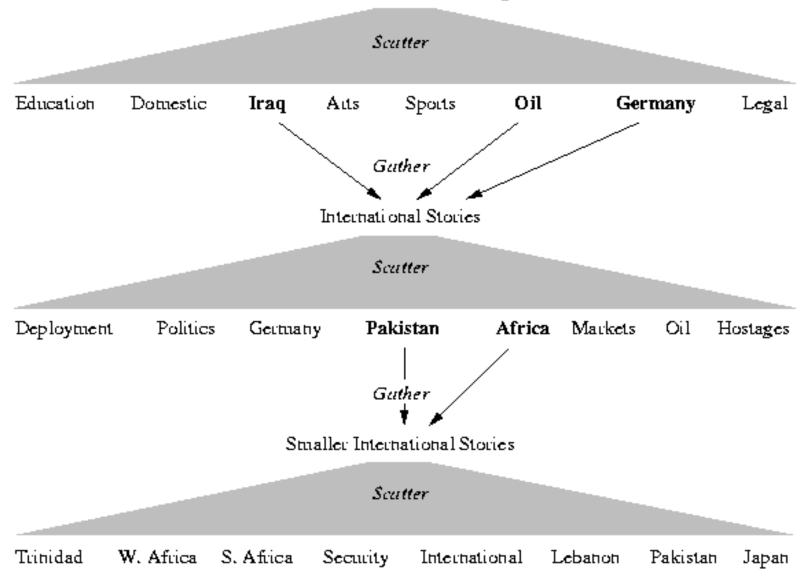
$$f_{tfidf}(\vec{w}^{(i)}, W_{j}) \propto \log f_{tf}(\cdot) \log \frac{D}{\sum_{l=1}^{D} I[W_{j} \in \vec{w}^{(l)}]}$$

## **Impact**

- Social bookmarking is big and getting bigger
- Tags hold promise of specific, relevant indexing vocabulary for the web
  - Not quite full-text indexing
  - Not quite controlled pre-coordinate indexing
- Tagging data improves web clustering performance, which promises better IR
  - How else will tagging impact IR?

## Scatter/Gather [Cutting et al 1992]

New York Times News Service, August 1990



# Stanford tag crawl dataset

Heymann, et. al 2008

#### **Bookmarks/Posts**

paul: news, uk  $\rightarrow$  bbc.co.uk 08:33:25

mary: recipes, food  $\rightarrow$  food.com 08:33:23

dave: tv, cnn, news  $\rightarrow$  cnn.com 08:33:21

#### **Triples**

(paul, news, bbc.co.uk) (paul, uk, bbc.co.uk)

(mary, recipes, food.com) (mary, food, food.com)

(dave, tv, cnn.com) (dave, cnn, cnn.com) (dave, news, cnn.com)

# Stanford tag crawl dataset

#### Back Link Text

... He is also a CNN Contributor, appearing on a variety of shows, including The Situation Room, Anderson Cooper 360, Lou Dobbs Tonight, and many others...

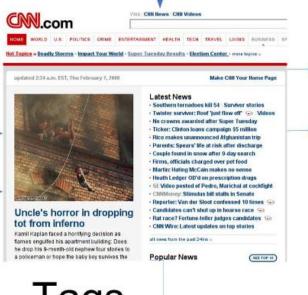
#### Page Text

CNN.com is among the world's leaders in online news and information delivery. Staffed 24 hours, seven days a week by a dedicated staff in CNN's world headquarters in Atlanta, Georgia, ...

#### Forward Link Text

CNN.com is among the world's leaders in online news and information delivery. Staffed 24 hours, seven days a week by a dedicated staff in CNN's world headquarters in Atlanta, Georgia, ...







Tags news cnn

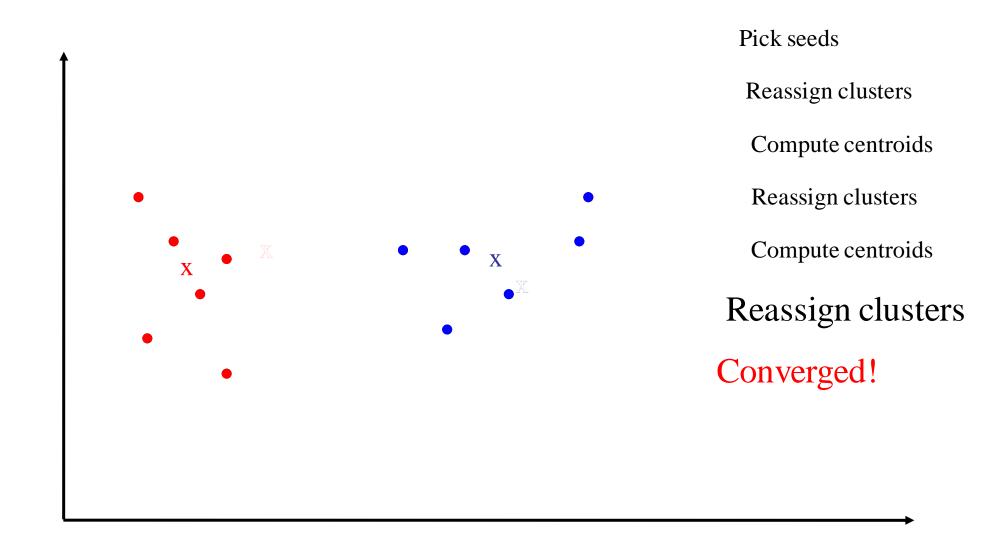
### K-means [CS276]

- Assumes documents are real-valued vectors
- Clusters based on centroids (aka the center of gravity or mean) of points in a cluster, c:

$$\mu c = \frac{1}{|c|} \sum_{x \in c} x$$

 Reassignment of instances to clusters is based on distances to the current cluster centroids

## K-means example (K=2) [CS276]

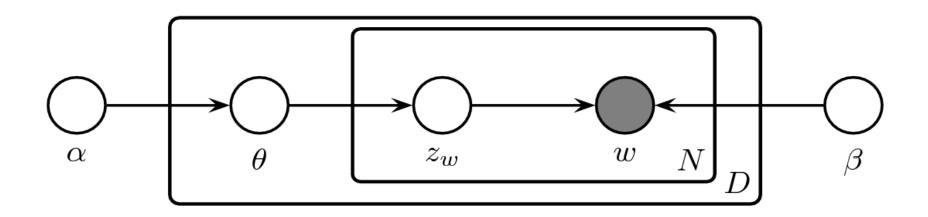


## MM-LDA outperforms K-means

### On top-level ODP categories

	LDA	K-means
Words	0.260	.139
Tags	0.270	.219
Words+Tags	0.307	.225

## Latent Dirichlet Allocation (LDA)



D – number of documents

N – number of words in document

alpha – symmetric Dirichlet prior

theta – per document topic multinomial

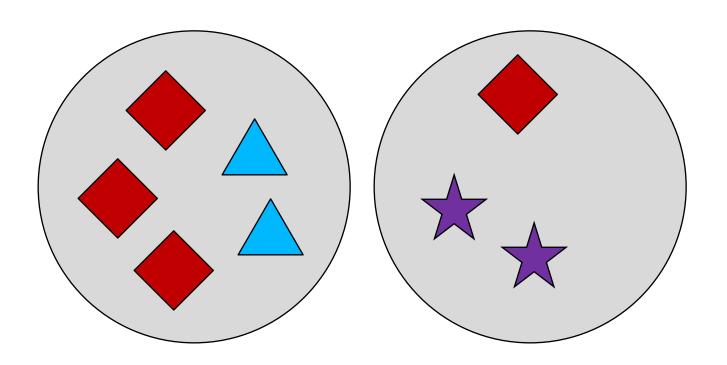
 $z_w$  — per word topic assignment

*w* – word observation

beta – per topic word multinomial

## MM-LDA Properties

- Natural extension of LDA
- Jointly models multiple types of observations
  - Similar to Blei et al.'s GM-LDA for images with captions
- Words and tags counted independently, contribute jointly to document topic model



### A web document collection

Stanford Tag Crawl Dataset:

One month of del.icio.us posts in May/June 2007

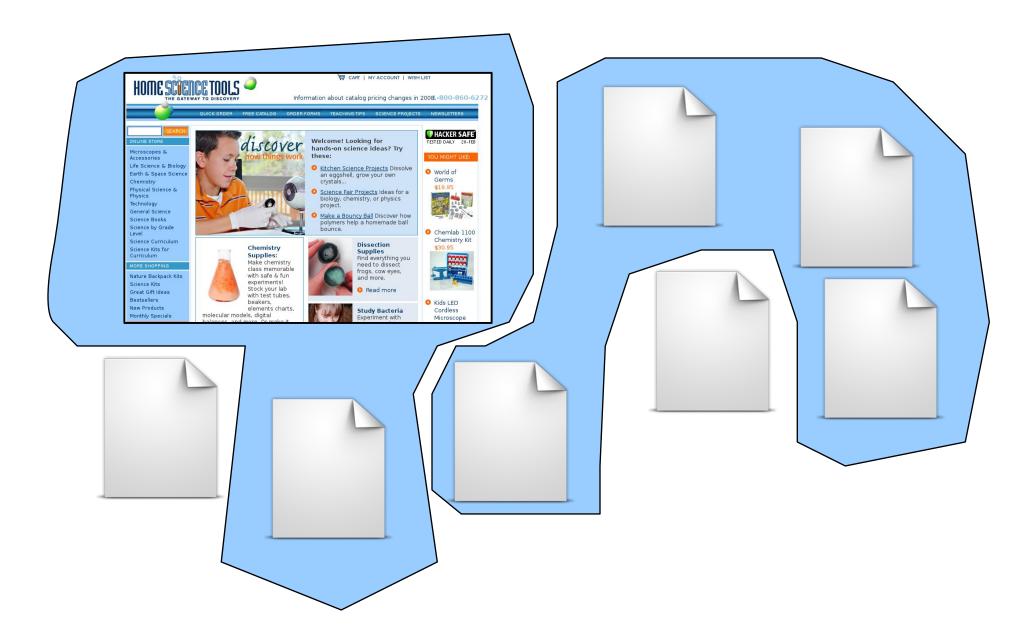


# Most web pages come with words



Words: welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals ...

## Words can be used to cluster



# Text surrounds links from other pages



#### **Anchor Text:**

tools home science links click buy supplies experiments ...

Words: welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals ...

# Social bookmarking websites add tags



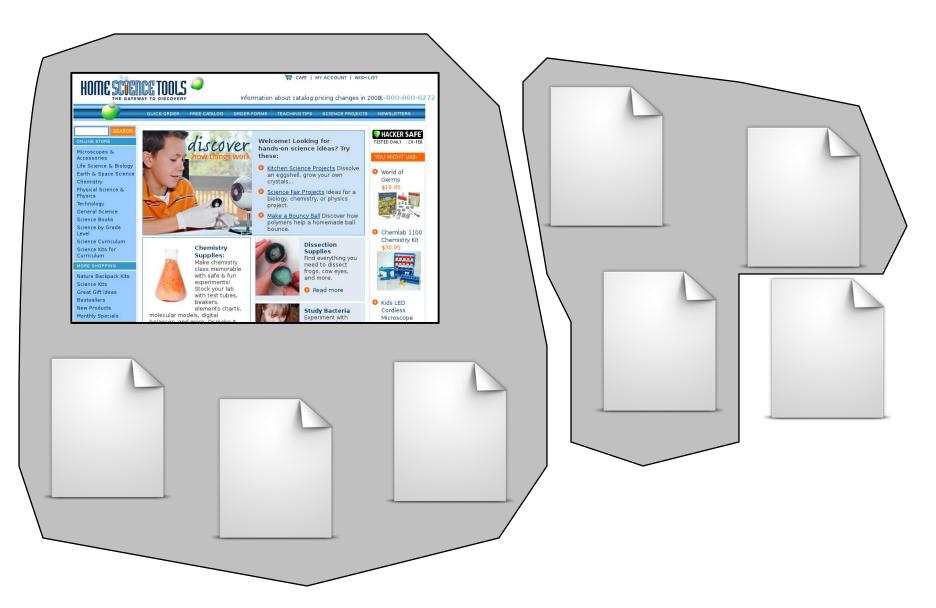
Tags: curriculum education homeschool imported learning science shopping slinky teachers teaching tools

#### **Anchor Text:**

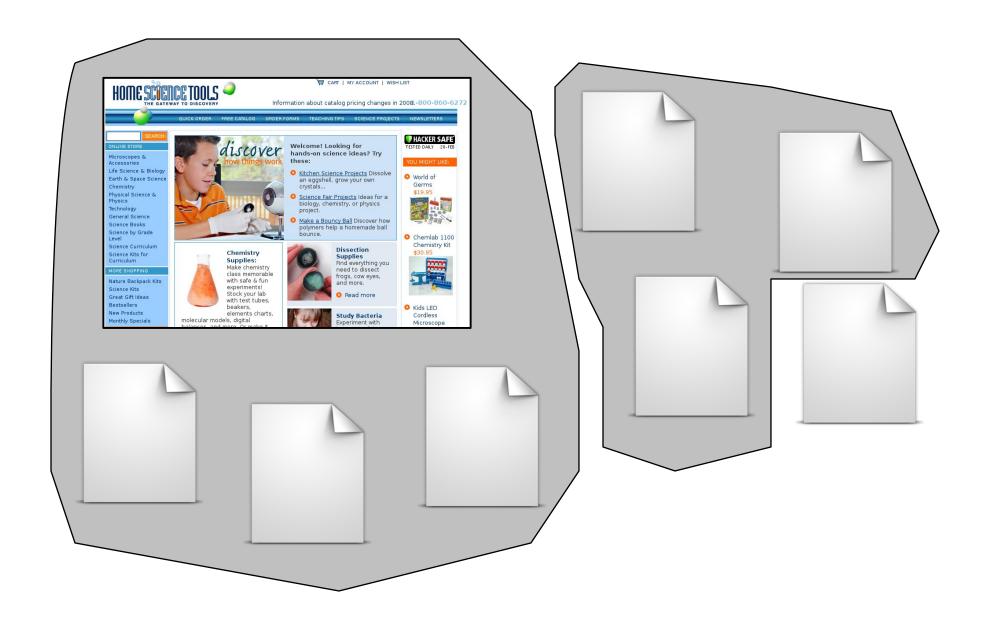
tools home science links click buy supplies experiments ...

Words: welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals ...

# How do we use words, anchor text, and tags together to most improve clustering?



# How do we test if clustering improves?



# Many pages have a "gold standard" label

#### Reference/Education



#### Tags:

curriculum education homeschool imported learning science shopping slinky teachers teaching tools

#### **Anchor Text:**

tools home science links click buy supplies experiments ...

Words: welcome looking hands-on science ideas try kitchen projects dissolve eggshell grow crystals ...

# Open Directory Project

dmoz open directory project

about dmoz | dmoz blog | suggest URL | help | link | editor login

Search

Search

AOL → search

Arts Business Computers

Movies, Television, Music... Jobs, Real Estate, Investing... Internet, Software, Hardware...

Games Health Home

Video Games, RPGs, Gambling... Fitness, Medicine, Alternative... Family, Consumers, Cooking...

Kids and Teens News Recreation

Arts, School Time, Teen Life... Media, Newspapers, Weather... Travel, Food, Outdoors, Humor...

Reference Regional Science

Maps, Education, Libraries... US, Canada, UK, Europe... Biology, Psychology, Physics...

Shopping Society Sports

Clothing, Food, Gifts... People, Religion, Issues... Baseball, Soccer, Basketball...

World

<u>Català</u>, <u>Dansk</u>, <u>Deutsch</u>, <u>Español</u>, <u>Français</u>, <u>Italiano</u>, <u>日本語</u>, <u>Nederlands</u>, <u>Polski</u>, <u>Pyccкий</u>, <u>Svenska</u>...

Become an Editor Help build the largest human-edited directory of the web



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## Cluster evaluation

#### Reference

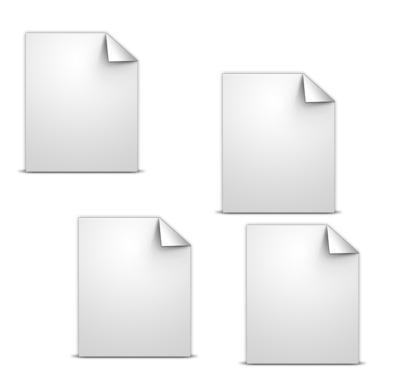
#### HOME SCIENCE TOOLS Information about catalog pricing changes in 2008.-800-860-627 HACKER SAFE Welcome! Looking for discover hands-on science ideas? Try Microscopes & Accessories Life Science & Biology O Kitchen Science Projects Dissolve World of Earth & Space Science an eggshell, grow your own crystals... Germs \$19.95 Chemistry Physical Science & Physics O Science Fair Projects Ideas for a biology, chemistry, or physics project. Technology General Science Make a Bouncy Ball Discover how Science Books polymers help a homemade ball bounce. Science by Grade O Chemlab 1100 Chemistry Kit Science Curriculum Dissection Chemistry Science Kits for Curriculum Supplies Find everything you Supplies: Make chemistry need to dissect class memorable frogs, cow eyes, with safe & fun experiments! Stock your lab with test tubes, Read more Bestsellers beakers, elements charts, Study Bacteria Cordless molecular models, digital Microscope



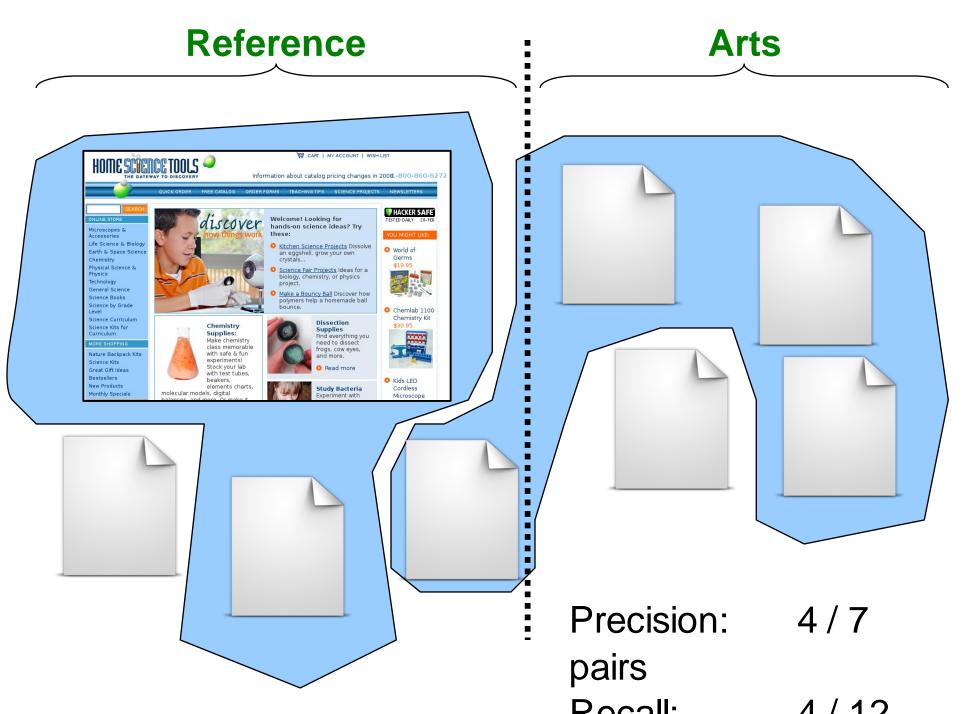




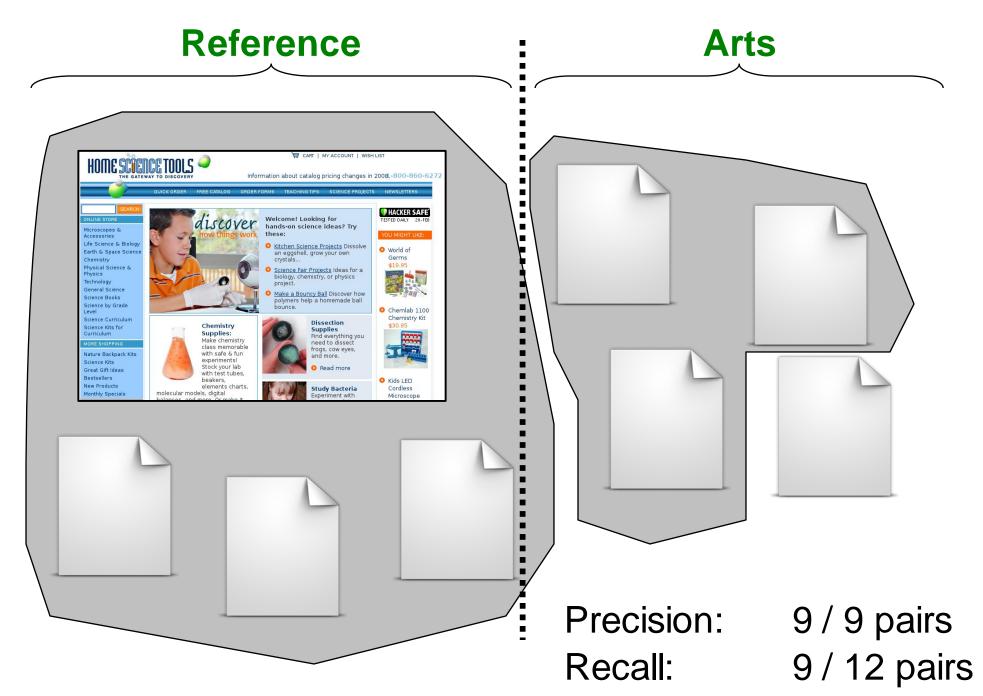
#### **Arts**



## Cluster evaluation



## Cluster evaluation



## **Outline**

- The tagged web
- Dataset and methodology

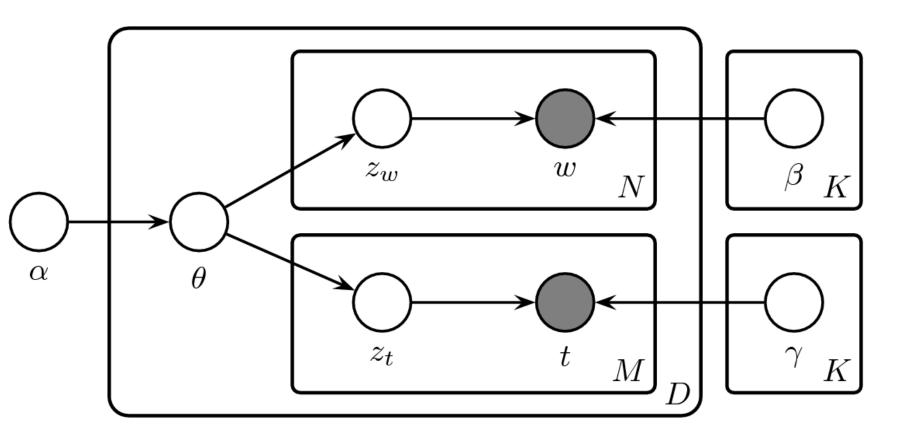
## Algorithms for clustering with tags and words

- K-Means in tag-augmented vector space
- Multi-Multinomial LDA

### Results

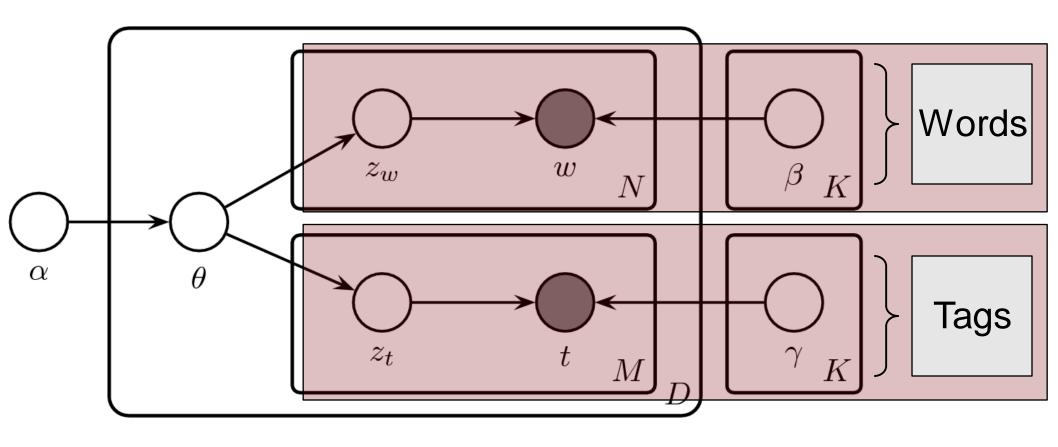
- Tag and word normalization
- Clustering at varying levels of specificity
- Incorporating anchor text

# Multi-Multinomial LDA (MM-LDA)



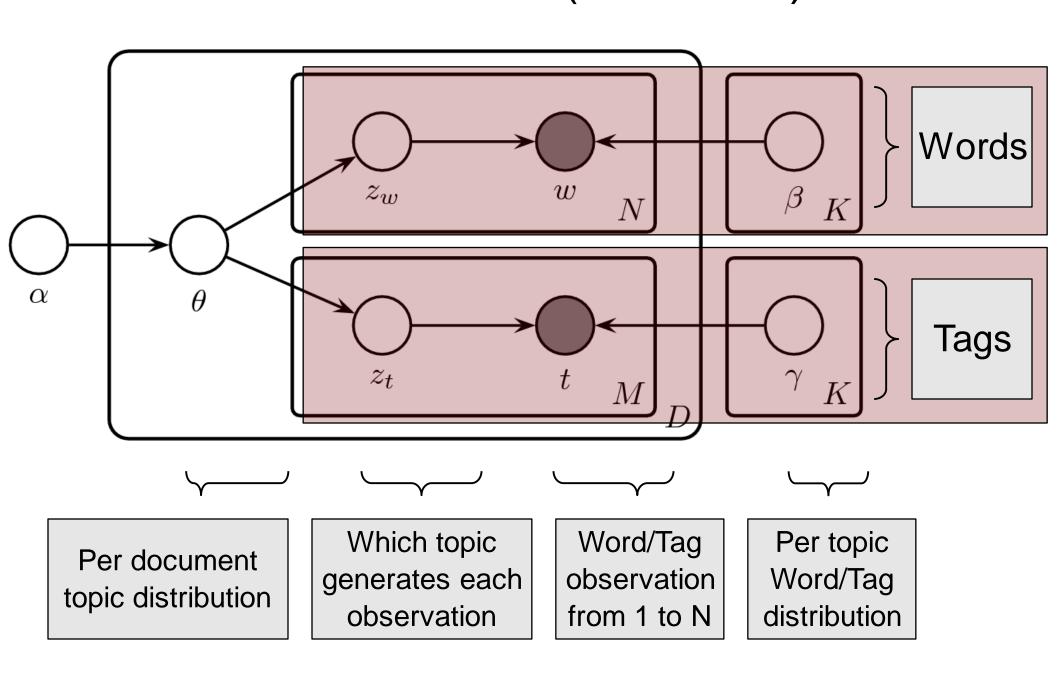
Extends Latent Dirichlet Allocation: Words and tags (and anchors, etc.) are counted independently, contribute jointly to topic probabilities

# Multi-Multinomial LDA (MM-LDA)



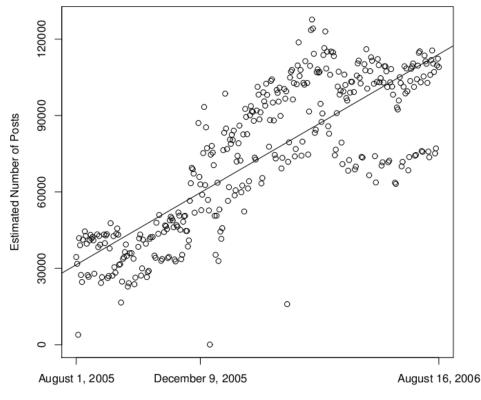
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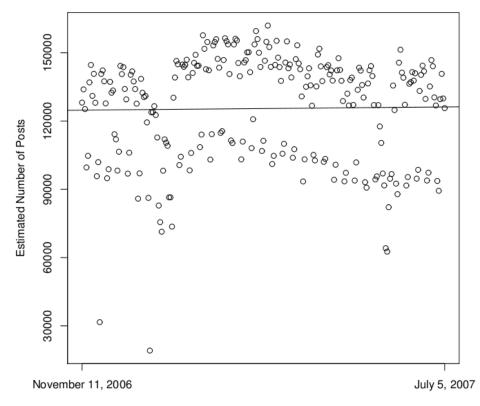


## The tagged web (Heymann, et al., WSDM 2008)

pprox 120 thousand (pprox 10<sup>5</sup>) posts/day (versus pprox 10<sup>6</sup> blog posts/day) 60–150 million posts 12–75 million (pprox 10<sup>7</sup>–10<sup>8</sup>) unique URLs (versus pprox 10<sup>9</sup>–10<sup>11</sup> total URLs)

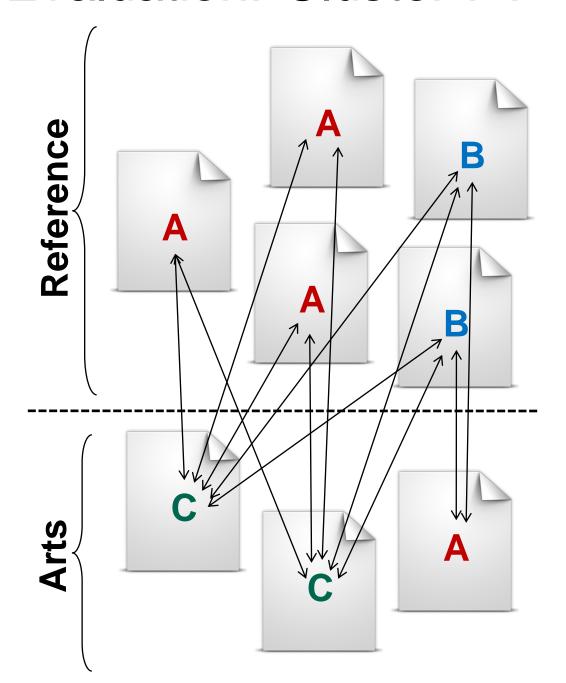


Date



Date

## **Evaluation: Cluster F1**



	Same Label	Different Label
Same Cluster	5	3
Different Cluster	8	12

Cluster Precision: 5/8
Cluster Recall: 5/13

# Goal: clustering for information retrieval

- Better user interfaces
  - e.g. Clusty, Vivisimo, Scatter/Gather, and friends
- Collection clustering
  - e.g. Columbia Newsblaster, Google News
- Improved language models for better retrieval
  - e.g. Liu and Croft 2004; Wei and Croft 2006
- Better cluster based-retrieval
  - e.g. Salton 1971

# Stanford tag crawl / ODP intersection

ODP Name	#Docs	Top Tags by PMI
Adult	36	blog illustration art erotica sex
Arts	1446	lost recipes knitting music art
Business	908	accounting business lockpicking agency
Computers	5361	web css tools software programming
Games	291	un rpg fallout game games
Health	434	parenting medicine healthcare medical
Home	654	recipes blog cooking coffee food
Kids	669	illusions anatomy kids illusion copyright
News	373	system-unfiled daily cnn media news
Recreation	411	humor vacation hotels reviews travel
Reference	1325	education reference time research dictionary
Science	1574	space dreams psychology astronomy science
Shopping	310	custom ecommerce shop t-shirts shopping
Society	1852	buddhism christian politics religion bible
Sports	146	sport cycling nfl football sports
World	756	speed bandwidth google speedtest maps

## K-means feature vectors

Feature Space Size Strategy Words Words Tags **Tags** Tags as Weighted Words **Tags as Words** Tags as New Words **Tags Words** Tags+Words Words Tags

# Experiments

#### **Features**

Vector Space
Model:
K-means

Generative
Model:
MM-LDA

2. Extending LDA
for multiple
feature types

Models

# Result: Sometimes, tags tell you more about cluster membership than words do

	Features	K-means	(MM-)LDA	_
All	Words	.139	.260	
	Tags	.219	.270	
	Words+Tags	.225	.307	
Programming	Words	.189	.288	
Languages	Tags	.567	.463	
	Words+Tags	.556	.297	
Social	Words	.196	.300	
Sciences	Tags	.307	.310	
	Words+Tags	.308	.302	
// 4/				

"software" applies to only 21% of Computer pages

K-means wins when the feature space is cleaner