# Web Mining: Blogspace and Folksonomies

A Guide to Web Research: Lecture 3

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## Outline

- Introduction to Blogspace
- Introduction to Folksonomies
- Algorithmic Challenges
  - Personal News Aggregation
  - Structure Discovery in Folksonomies

# Talk Objective

### **Today:**

- Short description of technology
- Technological challenges
- Algorithmic problems

#### To do:

- Adding assumptions to the problems
- Constructing (approximate) algorithms

Part I Blogspace

What is blogspace?

What related technologies are supposed to appear in nearest future?

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## Blogspace: Overview

**Blogspace** (Blogosphere) is a set of all weblogs

### **Every blog consists of:**

- Profile
- Posts: title, content, time-stamp, comments, tags
- Subscribers

### Prominent technologies in the field:

Blogger, Livejournal, Wordpress, Technorati

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# Part II Folksonomies

What is folksonomy?

What related technologies are supposed to appear in nearest future?

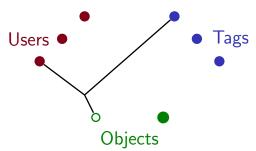
## Technological Challenges in Blogspace

- Blog search and blog ranking
- Personal newspaper: every user every day receives personal digest of all posts in the world
- Advertising in blogspace, in particular, understanding information propagation in blogspace
- Tracking blogspace reflections of real life events

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## Folksonomy: Overview

**Folksonomy** is a set of triples < *user*, *object*, *tag* > **Primary purpose:** memory assistance



## Prominent technologies in the field:

Del.icio.us, Flickr.com, tags in blogspace, GMail labels

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# Technological Challenges in Folksonomies

- Tag-based file system
- Utilizing folksonomies in web search
- Tag subscriptions and other folksonomy-based recommendations
- Second layer challenge: discover and visualize relations between tags
- Automatic labelling

# Part III Algorithmic Challenges

Personal news aggregation

Structure discovery in folksonomies

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## Personal News Aggregation: Informally

### Personal news aggregation:

Every user has a preference profile: specified information sources, keywords, tags(topics), popularity, references to the preferences of others

Every news item has its own description: text, votes and recommendations, tags, author reputation, comments

## All-to-all filtering:

To find, say, ten most appropriate news items for every user

# Personal News Aggregation: Solutions

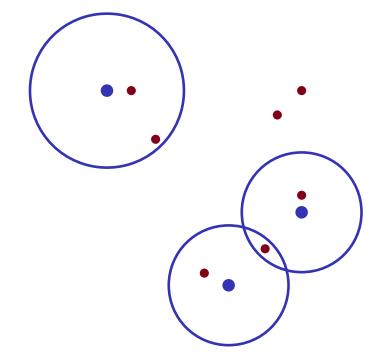
### Personalized news delivery:

Google News
Google Reader
Bloglines
Livejournal Friends
Feedburner

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## Formalization

- Every news is represented by a sparse vector
- Every user profile is represented by a sparse vector
- Similarity is defined as cosine between two vectors
- $\bullet$  Simplification: 0/1 vectors, similarity proportional to the size of intersection



# Large Scale All-to-All Nearest Neighbors

- N blue vectors in d-dimensional space, every vector has at most k nonzero components
- M red vectors in d-dimensional space, every vector has at most k nonzero components
- To find 10 nearest (according to cosine similarity)
   red vectors to every blue vector
- Desired time complexity

$$(N + M)$$
 polylog $(N + M)$  poly $(k)$ 

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# All-to-All Nearest Neighbors in Set Notation

- N blue sets, each of size at most k
- *M* red sets, each of size at most *k*
- To find 10 nearest (according to intersection-size similarity) red sets to every blue one in time

$$(N + M)$$
 polylog $(N + M)$  poly $(k)$ 

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## Structure Discovery in Folksonomies

Problem: finding similar tags in folksonomy

#### **Evidences of similarity:**

- Inner co-occurrence: some user applied both tags to some object
- Outer co-occurrence: one user applied the first tag, another user applied the second tag to the same object

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## Discovering Related Tags

- Bipartite graph tags-objects, *F* edges
- Task 1: for every tag find 10 nearest tags
- Task 2: for a given  $\alpha$  find all tag pairs that have similarity above  $\alpha$
- Desired time complexity: F · polylog(F)

## Tag similarity

### Projection to bipartite graph:

Removing users from folksonomy Notation: Q(t) is the set of all objects tagged by t

### Three formulas for tag similarity:

$$Sim(t_1,t_2)=|Q(t_1)\cap Q(t_2)|$$

$$\mathit{Sim}(t_1, t_2) = rac{|Q(t_1) \cap Q(t_2)|}{|Q(t_1)| + |Q(t_2)|}$$

$$\mathit{Sim}(t_1,t_2) = rac{|Q(t_1) \cap Q(t_2)|}{\min\{|Q(t_1)|,|Q(t_2)|\}}$$

## Discussion

Which of these two problems do you like more?

Changes to presented formalization?

Ideas and approaches?

Relevant work?

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## Call for participation

Know a relevant reference?

Have an idea?

Find a mistake?

Solved one of these problems?

- Knock to my office 1.156
- Write to me yura@logic.pdmi.ras.ru
- Join our informal discussions
- Participate in writing a follow-up paper

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# References (1/2)

#### Course homepage

http://logic.pdmi.ras.ru/~yura/webguide.html

R. Kumar, J. Novak, P. Raghavan, A. Tomkins
On the Bursty Evolution of Blogspace

http://cui.unige.ch/tcs/cours/algoweb/2005/articles/p568-kumar.pdf

D. Gruhl, R. Guha, D. Liben-Nowell, A. Tomkins
Information diffusion through blogspace
http://www.conf.ecs.soton.ac.uk/archive/00000597/01/p491-gruhl.pdf

E. Adar, L.A. Adamic

Tracking Information Epidemics in Blogspace

http://www.hpl.hp.com/research/idl/papers/blogs2/trackingblogepidemics.pdf

## Highlights

#### Three problems to think about:

- All-to-all nearest neighbors in sparse vector model
- All-to-all nearest neighbors in set notation with intersection-size similarity
- Finding all over-threshold similarities between tags in folksonomy

Vielen Dank für Ihre Aufmerksamkeit! Fragen?

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# References (2/2)

A. Mathes

Folksonomies-Cooperative Classification and Communication Through Shared Metadata http://www.adammathes.com/academic/computer-mediated-communication/folksonomies.pdf

A. Hotho, R. Jaschke, C. Schmitz, G. Stumme

Information retrieval in folksonomies: Search and ranking

http://www.kde.cs.uni-kassel.de/stumme/papers/2006/hotho2006information.pdf

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