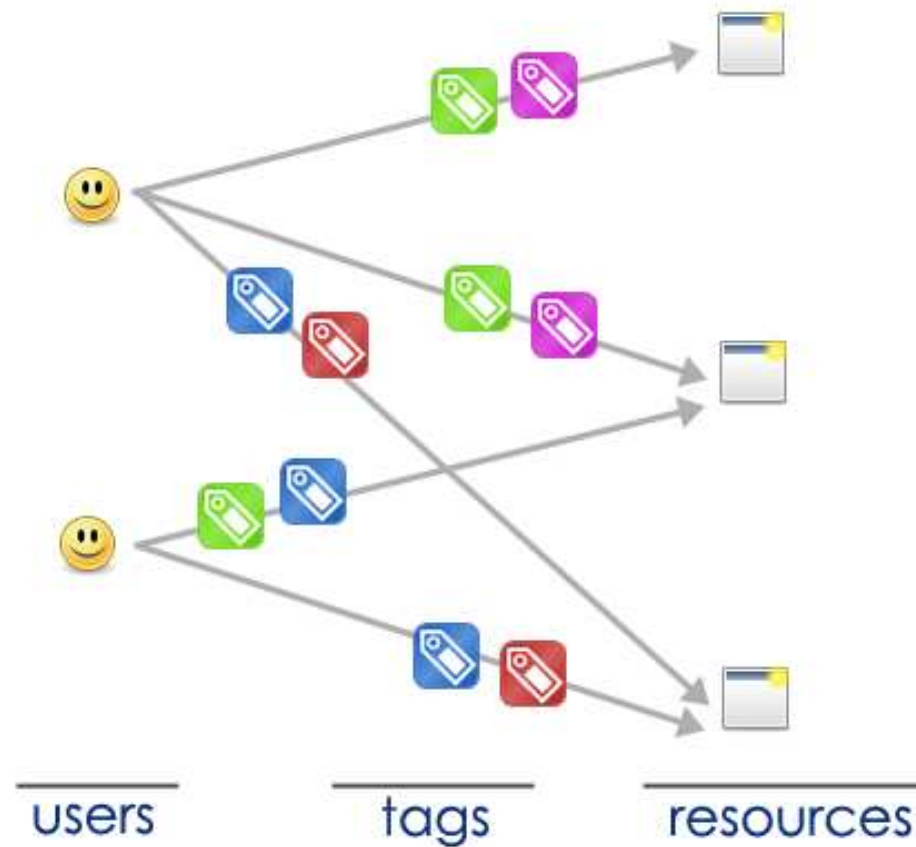


Understanding the Semantics of Ambiguous Tags in Folksonomies

Ching-man Au Yeung, Nicholas Gibbins, Nigel Shadbolt

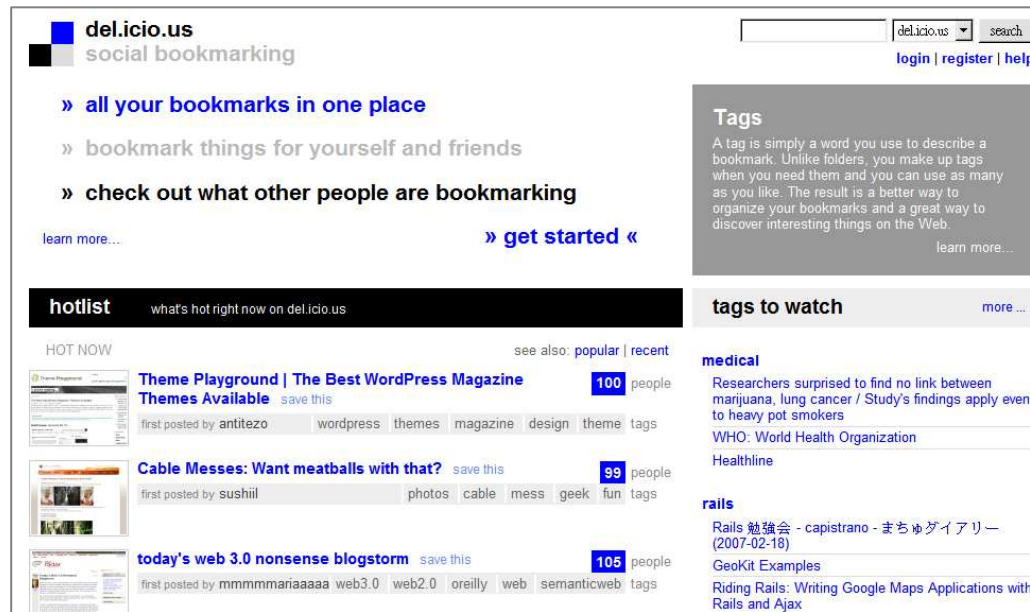
- Background (Collaborative tagging systems, folksonomies)
- Mutual contextualization in folksonomies
- Semantics of tags
- Discussions
- Conclusion and Future Work
- Recent Development

- *Collaborative tagging systems* and *folksonomies*



Background

- Examples of collaborative tagging systems



<http://del.icio.us/>

<http://b.hatena.ne.jp/>



- **Advantages** [Adam 2004, Wu et al. 2006]
 - Freedom and flexibility
 - Quick adaptation to changes in vocabulary (e.g. ajax, youtube)
 - Convenience and serendipity
- **Disadvantages** [Adam 2004, Wu et al. 2006]
 - Ambiguity (e.g. apple, sf, opera)
 - Lack of format (e.g. how multiword tags are handled)
 - Existence of synonyms (e.g. semweb, semanticweb, semantic_web)
 - Lack of semantics

*Are folksonomies really so **chaotic**?*

- Folksonomies are actually associations between the three types of entity - users, tags and resources [Mika 2005]
- Associations between these entities are not randomly made
- There is always a reason why a particular user uses a particular tag to describe a particular Web resources
- Semantics embedded in folksonomies → ***mutual contextualization*** between the entities

Folksonomy (A hypergraph)

$$F = \langle U, T, D, A \rangle; A \subseteq U \times T \times D$$

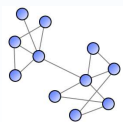


A User

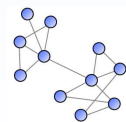
Bipartite graph TD_u

$$TD_u = \langle T \cup D, E_{TD} \rangle$$
$$E_{TD} = \{ \{t, d\} \mid \{u, t, d\} \in A \}$$

adj matrix multiplication



**tag
network**



**document
network**

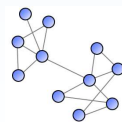


A Tag

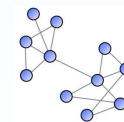
Bipartite graph UD_t

$$UD_t = \langle U \cup D, E_{UD} \rangle$$
$$E_{UD} = \{ \{u, d\} \mid \{u, t, d\} \in A \}$$

adj matrix multiplication



**user
network**



**document
network**

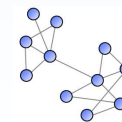


A Document

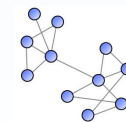
Bipartite graph UT_d

$$UT_d = \langle U \cup T, E_{UT} \rangle$$
$$E_{UT} = \{ \{u, t\} \mid \{u, t, d\} \in A \}$$

adj matrix multiplication



**user
network**



**tag
network**

Mutual contextualization in folksonomies



A Tag

Bipartite graph UD_t

$$UD_t = \langle U \cup D, E_{UD} \rangle$$

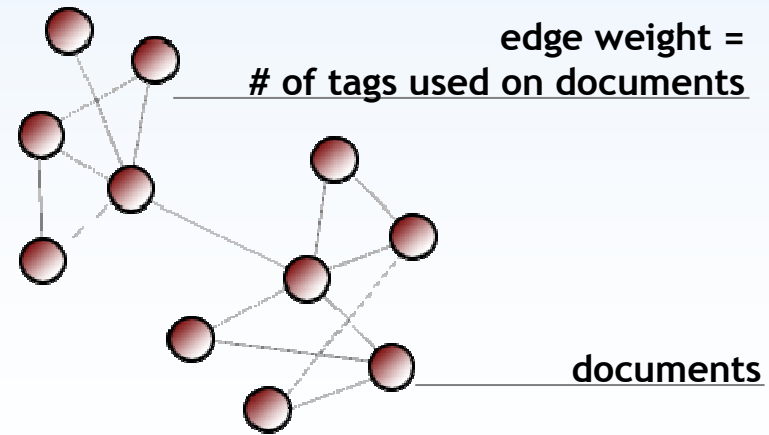
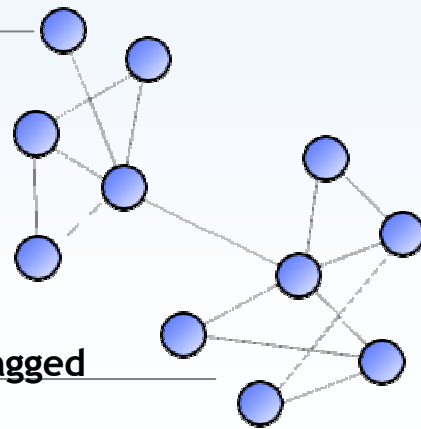
$$E_{UD} = \{ \{u, d\} \mid \{u, t, d\} \in A \}$$

adjacency matrix multiplication

user

edge weight =
of documents tagged

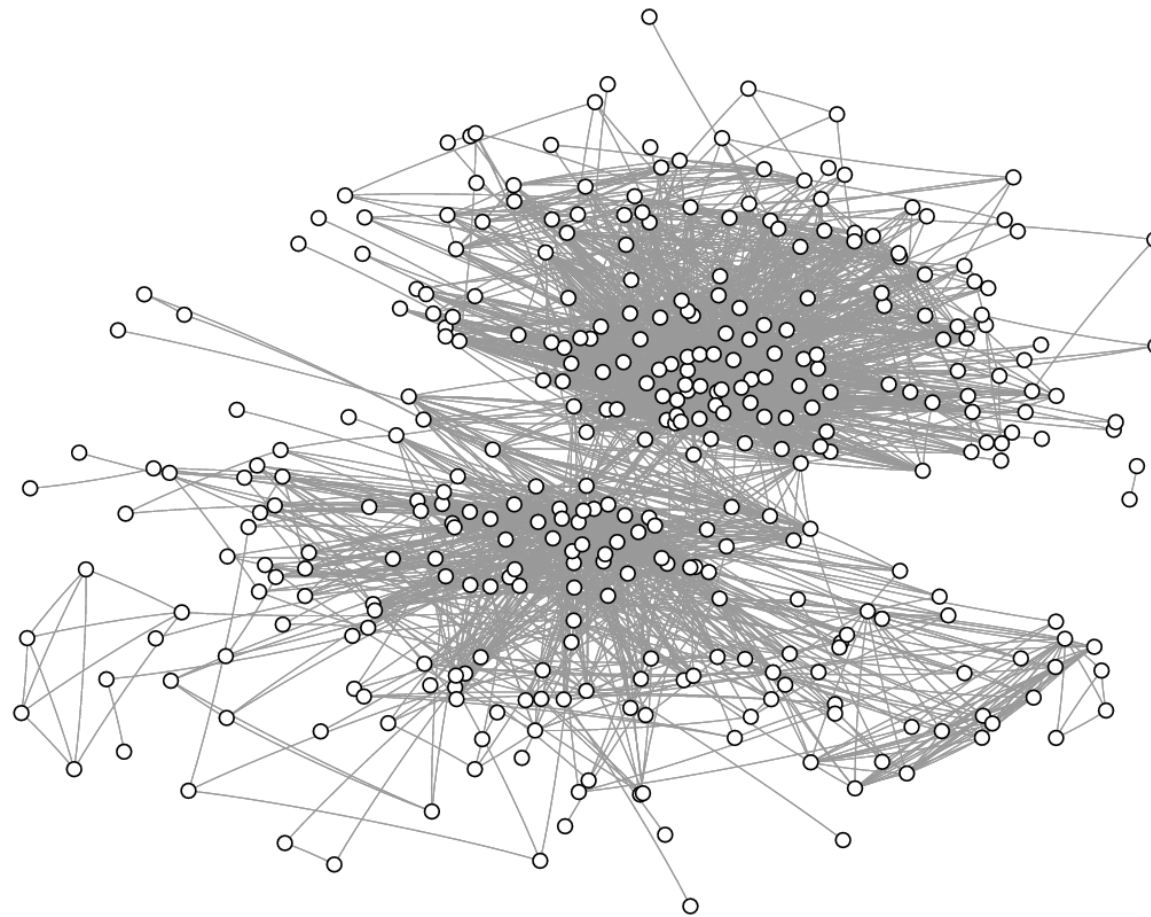
A weighted network of users



A weighted network of documents

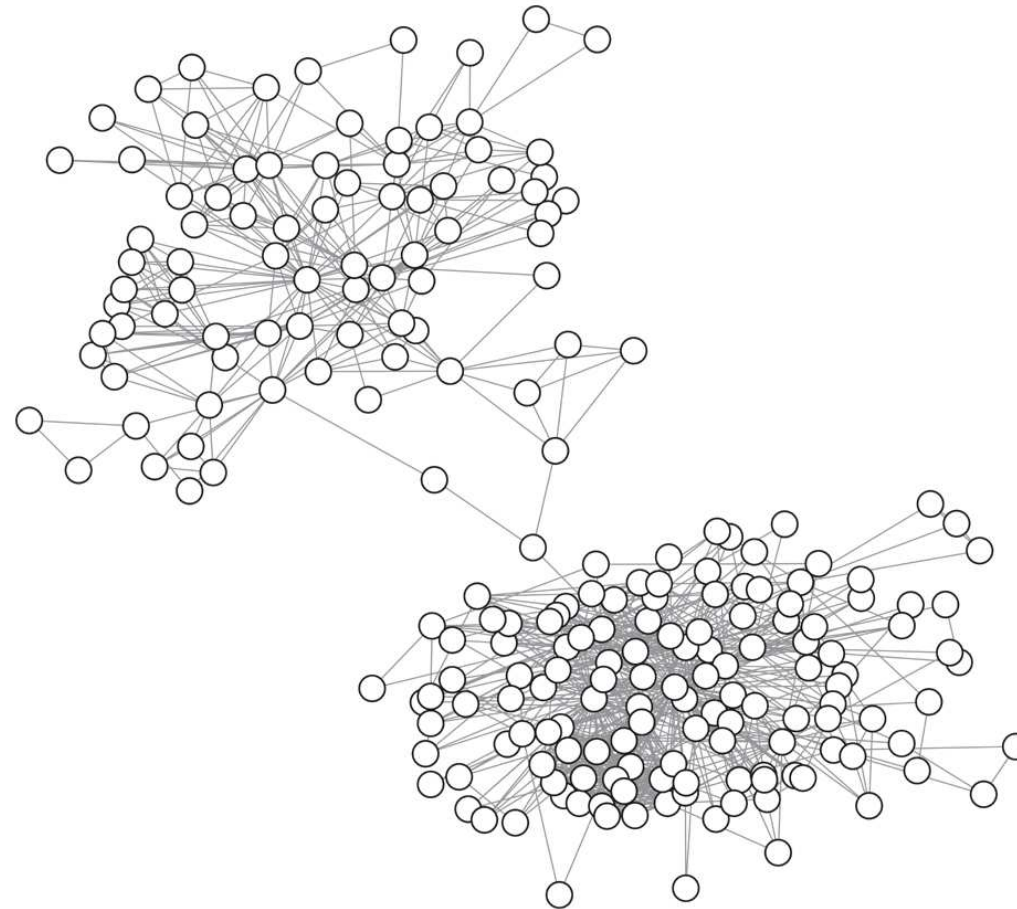
- A case study: *sf* in *del.icious*
 - *sf* is a popular tag in delicious (427 URLs, 19979 users, 5852 triples)
 - *sf* is ambiguous (*Science fiction* or *San Francisco*?)
 - Are users using the same tag to refer to two different concepts?
(Can the users/documents be divided into two groups?)
 - What would be the characteristics of the networks constructed around such ambiguous tag?

Understanding a single tag



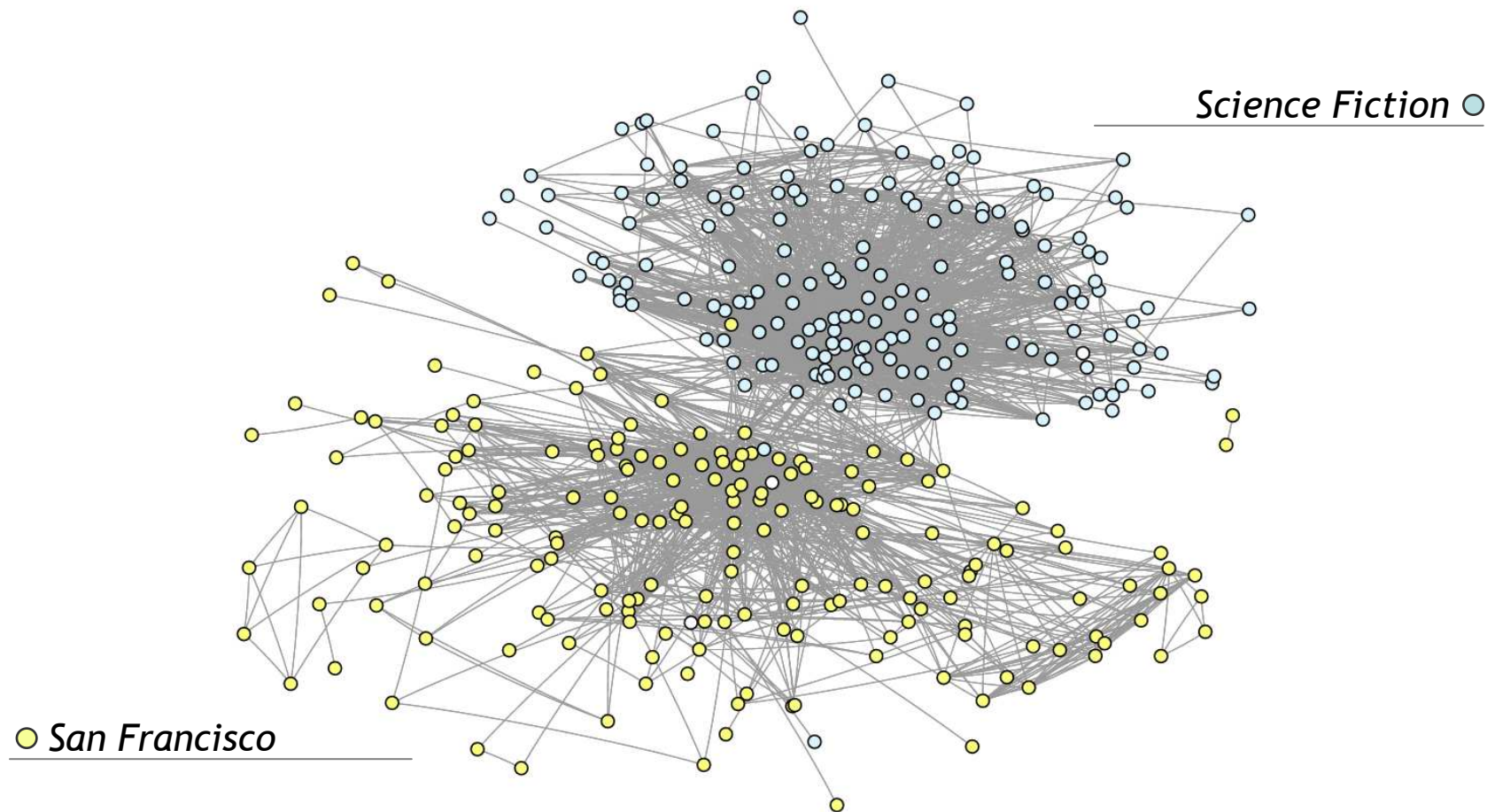
Network of Documents

Understanding a single tag



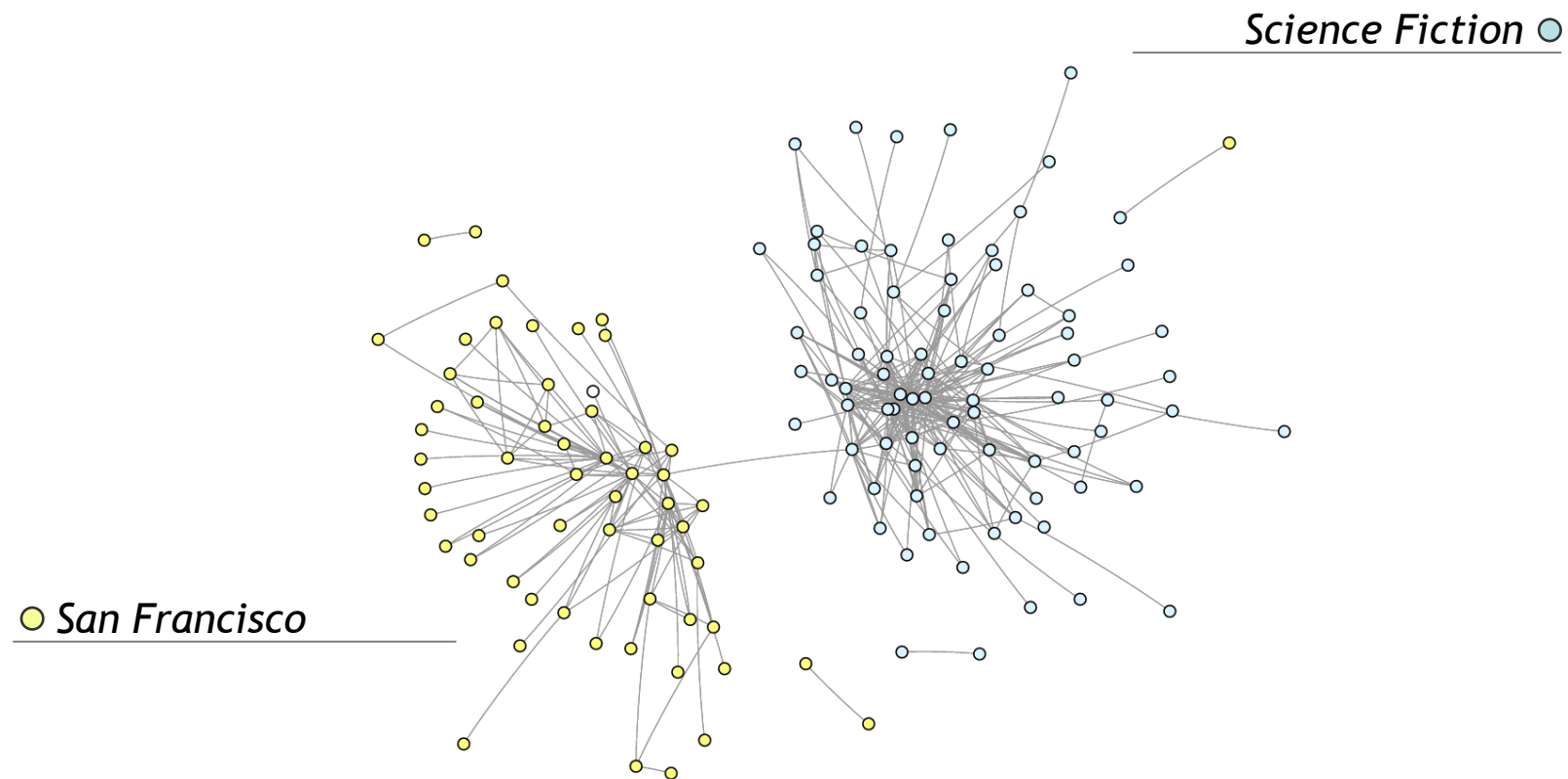
Network of Users

Understanding a single tag



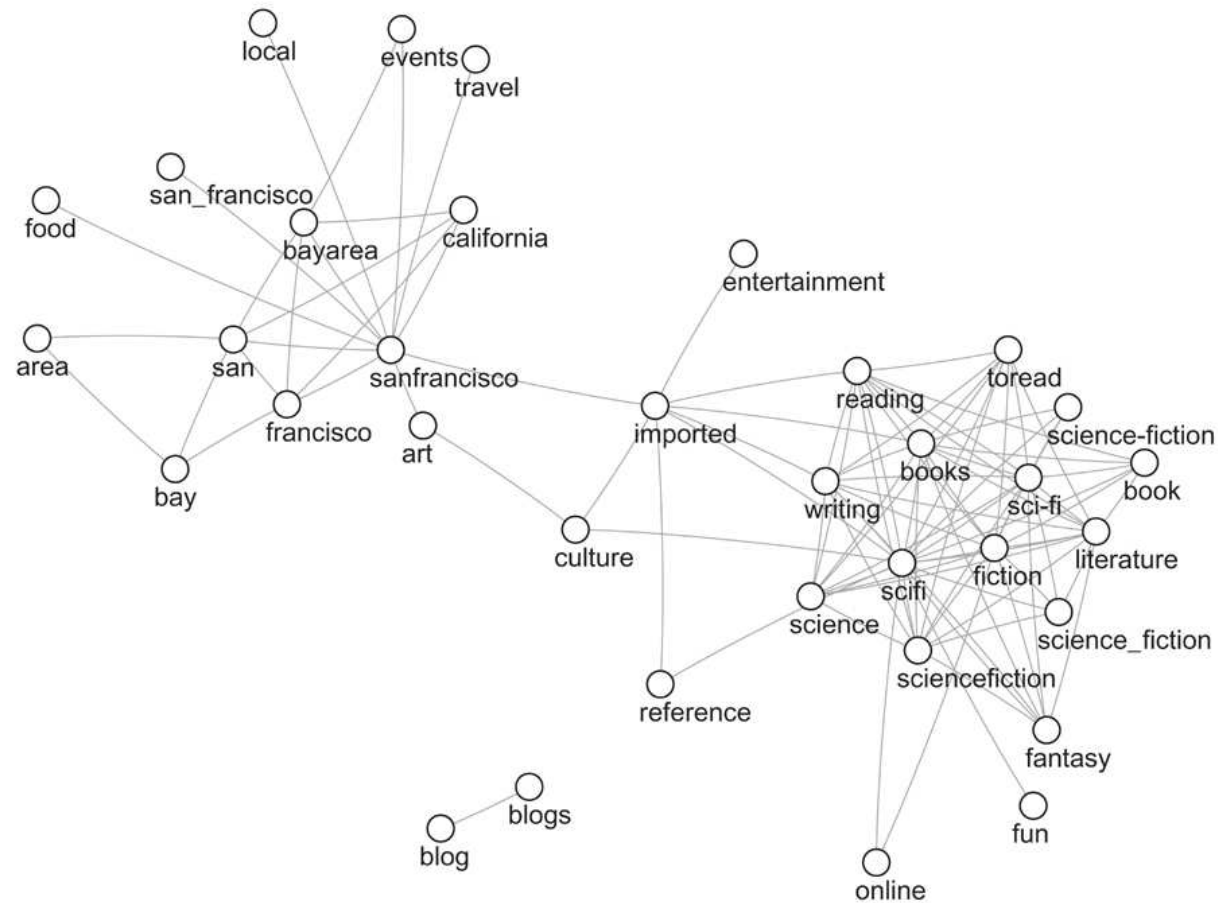
Network of Documents (Classified)

Understanding a single tag



Network of Documents (Removing edges with $w < 2$)

Understanding a single tag



Network of Tags (35 most frequently used)

- Users' behaviour: majority of users tend to use the tag to refer to one concept only
- Possibility of automatic tag disambiguation by examining the network topology
- Possibility of identifying sub-topics (e.g. restaurant-related or arts-related under “San Francisco”)
- Classification of documents which are not tagged with enough tags

- Conclusions

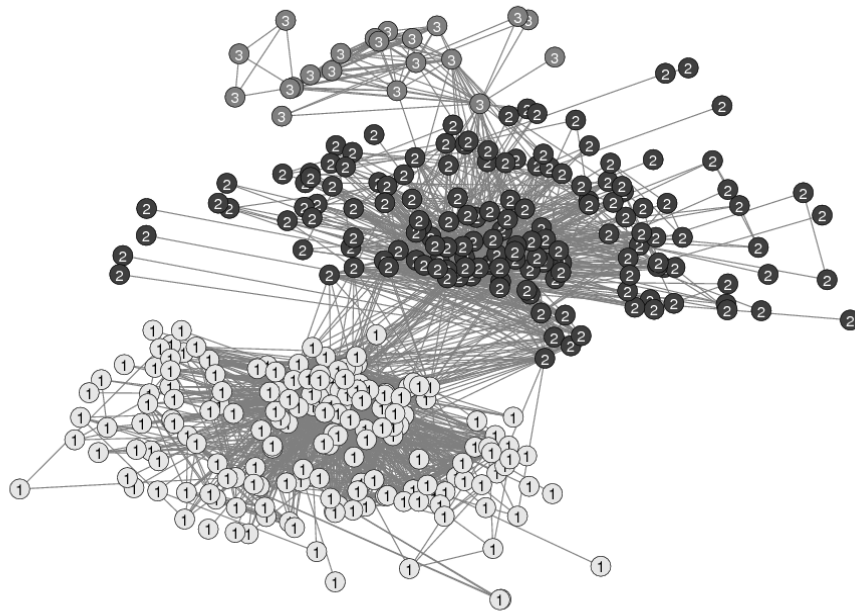
- The semantics of a tag can be understood by studying the associations between users and documents
- Automatic tag disambiguation is possible by exploring the topology of networks of users and documents around a tag

- Future Work

- Develop automatic algorithms for tag disambiguation
- Look for an appropriate representation for tag meanings
- Apply similar techniques on a user or a document (e.g. to understand a user's interest/expertise; to study the social network and annotations of a document)

- Applying community-discovery algorithms on the networks (e.g. modularity optimization [Newman & Girvan 2004])
- Attempt to break down the networks into communities (clusters of documents with similar contents/tags)
- Extract the most frequently used tags from each cluster
- Automatic tag meaning disambiguation
- A few case studies (Published in WI-IAT'07 [Au Yeung et al. 2007])

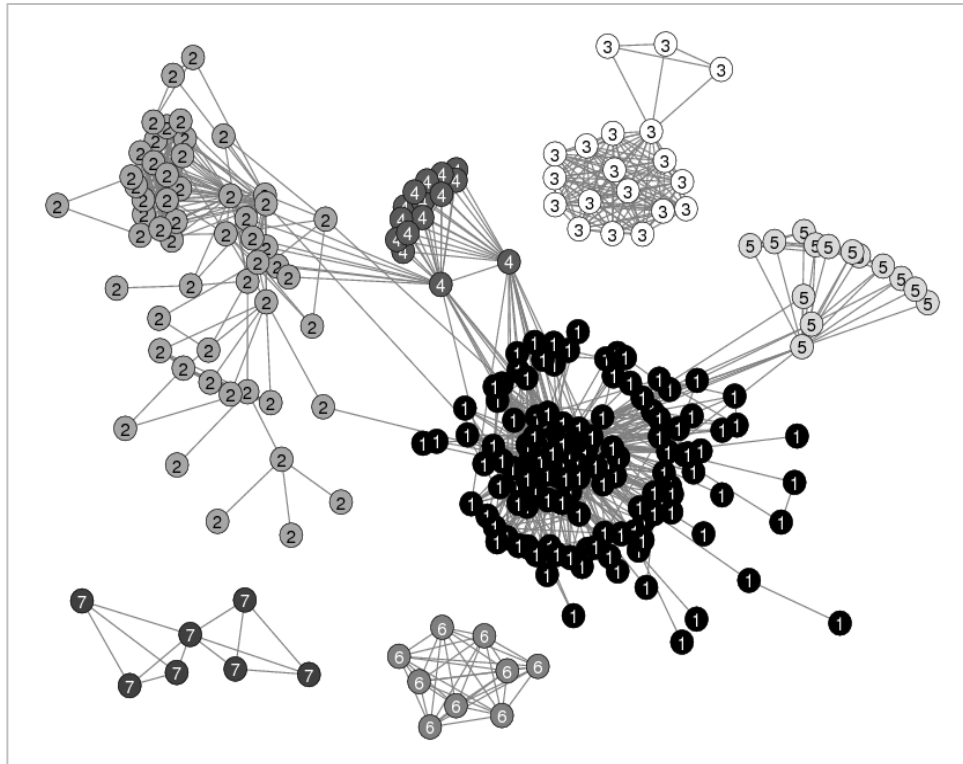
Recent development



Cluster	Tags
1	sf, scifi, fiction, books, sci-fi, writing, literature, science, sciencefiction, fantasy
2	sf, sanfrancisco, bayarea, san, francisco, california, travel, events, art, san_francisco
3	sf, sanfrancisco, design, bayarea, blog, food, todo, california, shopping, san

Automatic disambiguation of the tag “sf”

Recent development



Cluster	Tags
1	tube, london, underground, travel, transport, maps, uk, map, subway, reference
2	tube, diy, audio, electronics, amp, amplifier, amps, tubes, guitar, music
3	tube, video, web, internet, tv, online, web2.0, media, videos, imported
4	tube, video, youtube, videos, funny, cool, interesting, sport, fun, humor
5	tube, video, videos, online, web2.0, youtube, free, media, movie, fun
6	tube, youtube, video, videos, cool, feel.good, fun, funny, flash, music
7	tube, radio, electronics, tubes, antique, amplifier, data, audio, info, incarnate

Automatic disambiguation of the tag “tube”

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