



# Cross-domain Collaboration Recommendation

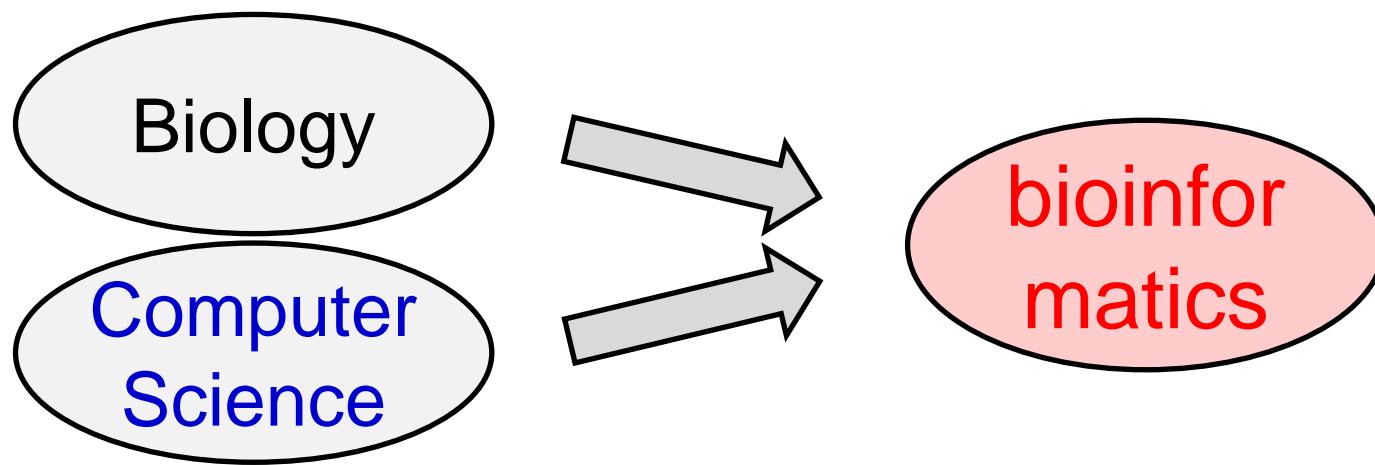
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**<sup>2</sup>IBM TJ Watson Research Center**

# Cross-domain Collaboration

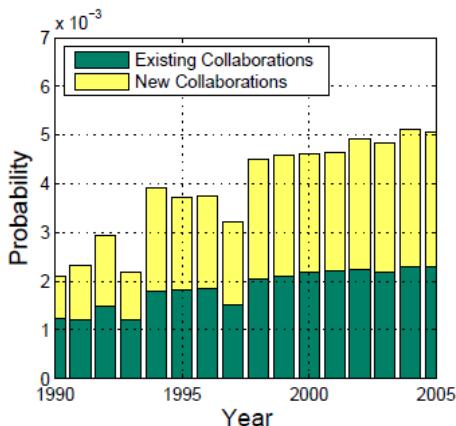
- Interdisciplinary collaborations have generated huge impact, for example,
  - 51 (>1/3) of the KDD 2012 papers are result of cross-domain collaborations between graph theory, visualization, economics, medical inf., DB, NLP, IR
  - Research field evolution



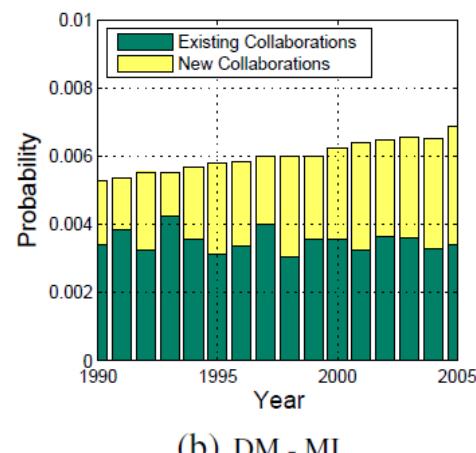
# Cross-domain Collaboration (cont.)



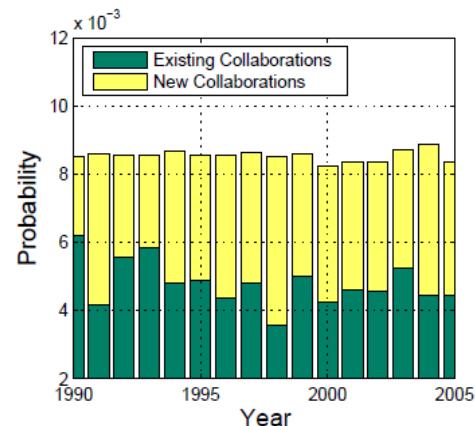
- Increasing trend of cross-domain collaborations



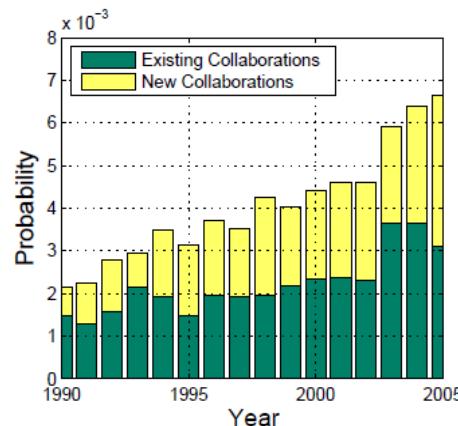
(a) DM - TH



(b) DM - MI



(c) DM - VIS



(d) MI - DB

Data Mining(DM), Medical Informatics(MI), Theory(TH), Visualization(VIS)



# Challenges



# Related Work

-Collaboration recommendation

- Collaborative topic modeling for recommending papers
  - C. Wang and D.M. Blei. [2011]
- On social networks and collaborative recommendation
  - I. Konstas, V. Stathopoulos, and J. M. Jose. [2009]
- CollabSeer: a search engine for collaboration discovery
  - H.-H. Chen, L. Gou, X. Zhang, and C. L. Giles. [2007]
- Referral web: Combining social networks and collaborative filtering
  - H. Kautz, B. Selman, and M. Shah. [1997]
- Fab: content-based, collaborative recommendation
  - M. Balabanovi and Y. Shoham. [1997]



# Related Work

-Expert finding and matching

- Topic level expertise search over heterogeneous networks
  - J. Tang, J. Zhang, R. Jin, Z. Yang, K. Cai, L. Zhang, and Z. Su. [2011]
- Formal models for expert finding in enterprise corpora
  - K. Balog, L. Azzopardi, and M.de Rijke. [2006]
- Expertise modeling for matching papers with reviewers
  - D. Mimno and A. McCallum. [2007]
- On optimization of expertise matching with various constraints
  - W. Tang, J. Tang, T. Lei, C. Tan, B. Gao, and T. Li. [2012]

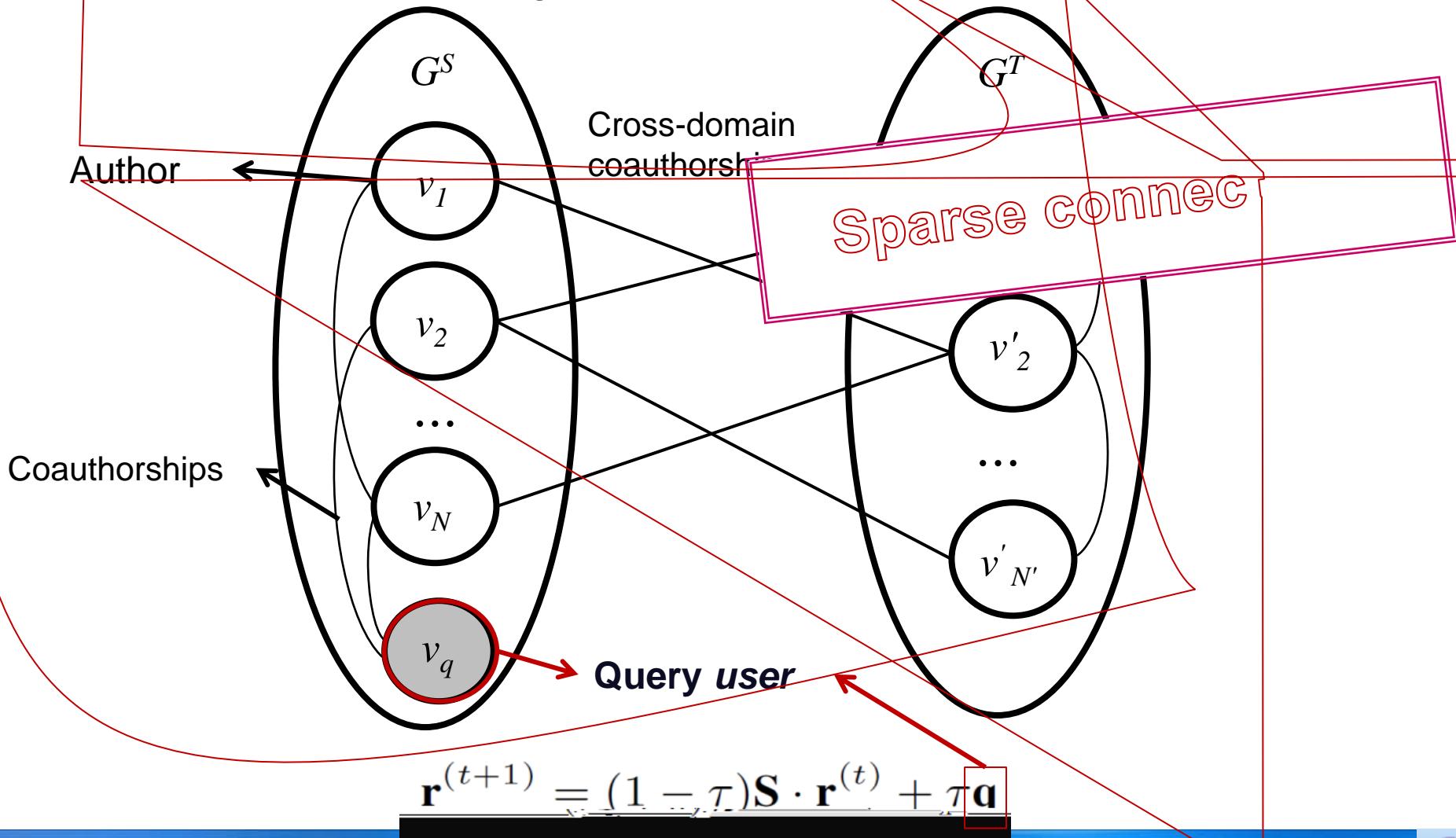


# Approach Framework

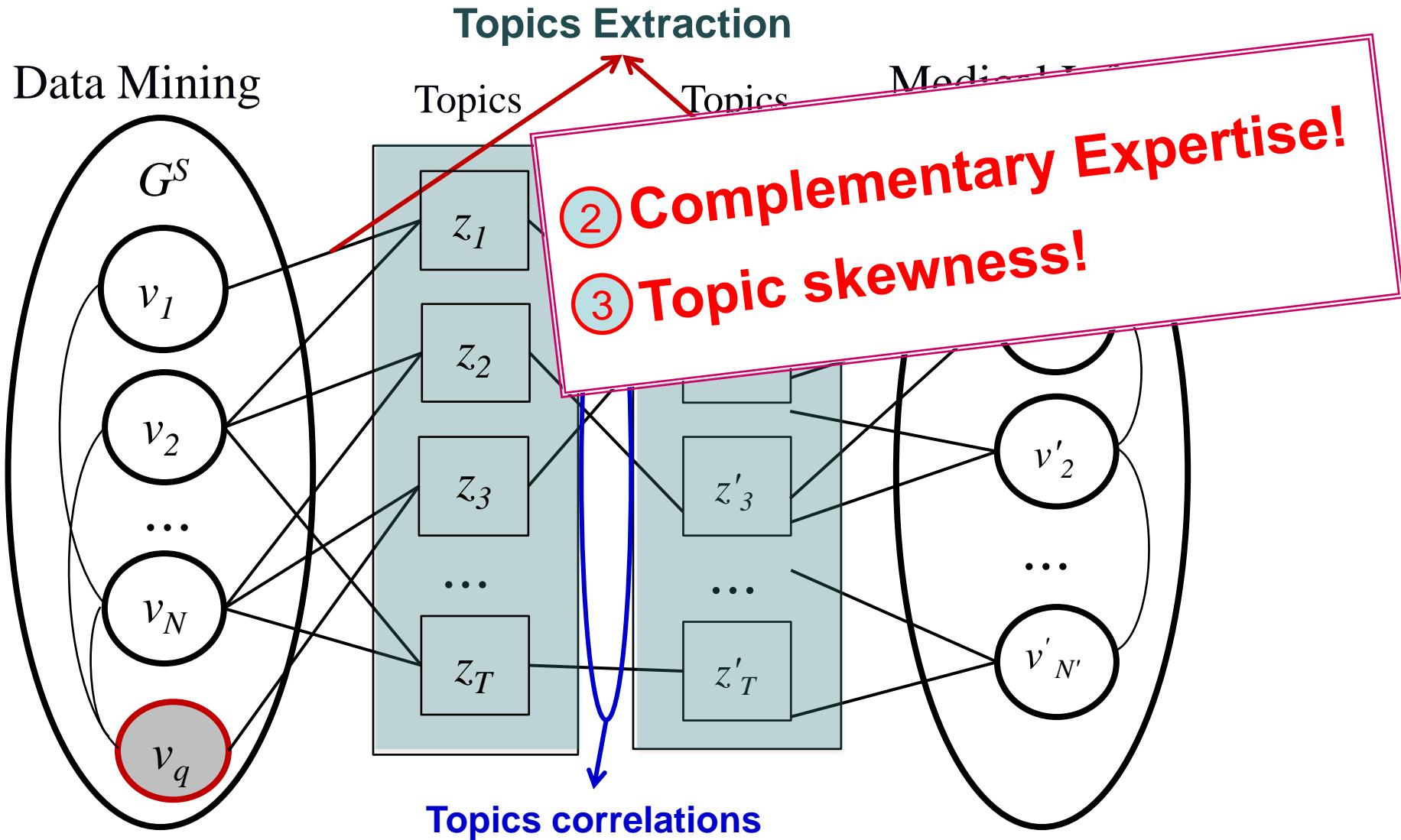
## —Cross-domain Topic Learning

# Author Matching

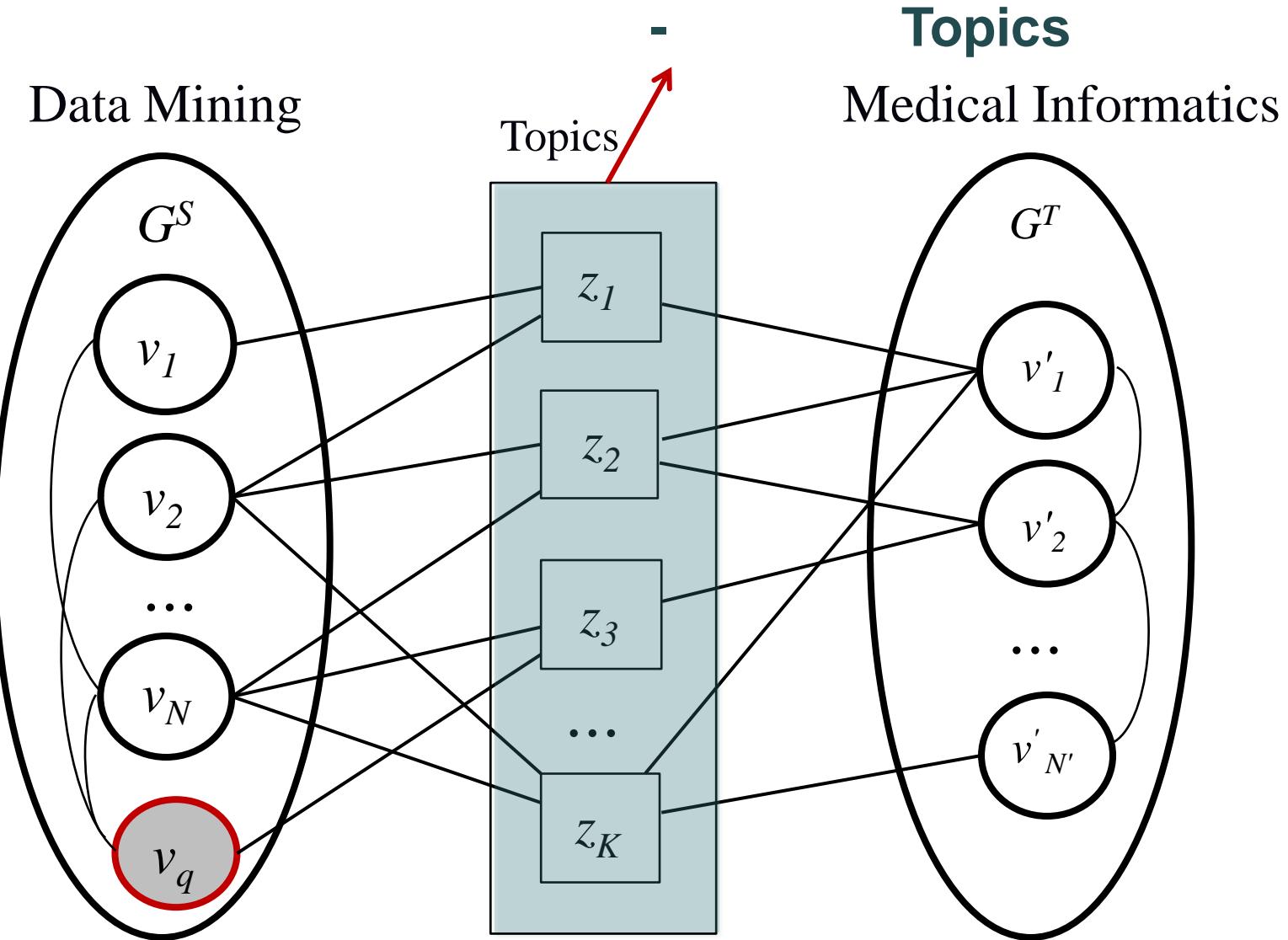
Data Mining      Medical Informatics



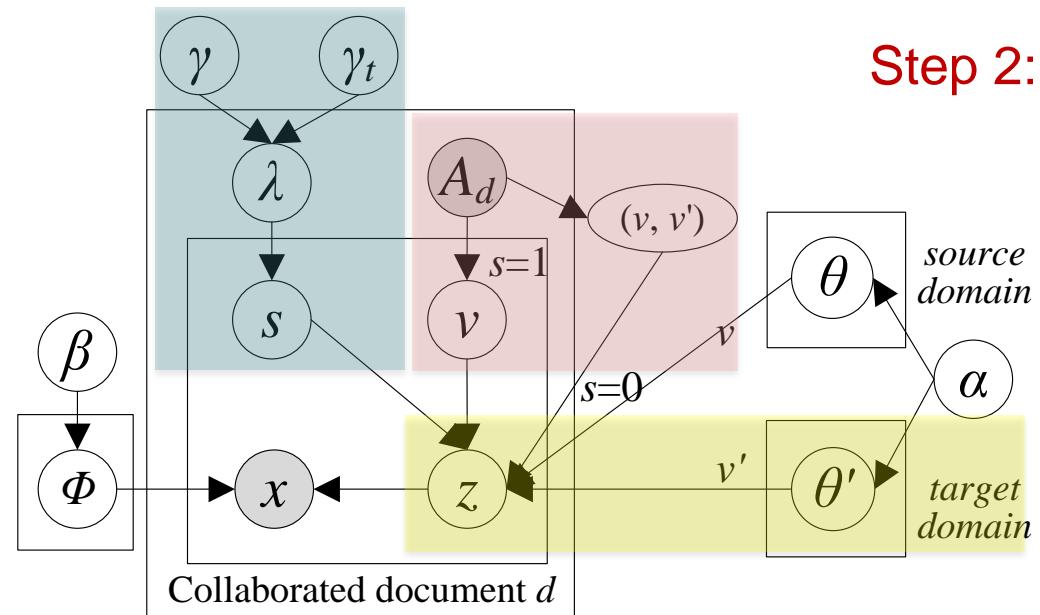
# Topic Matching



# Cross-domain Topic Learning



# Collaboration Topics Extraction



Step 1:

Step 2:

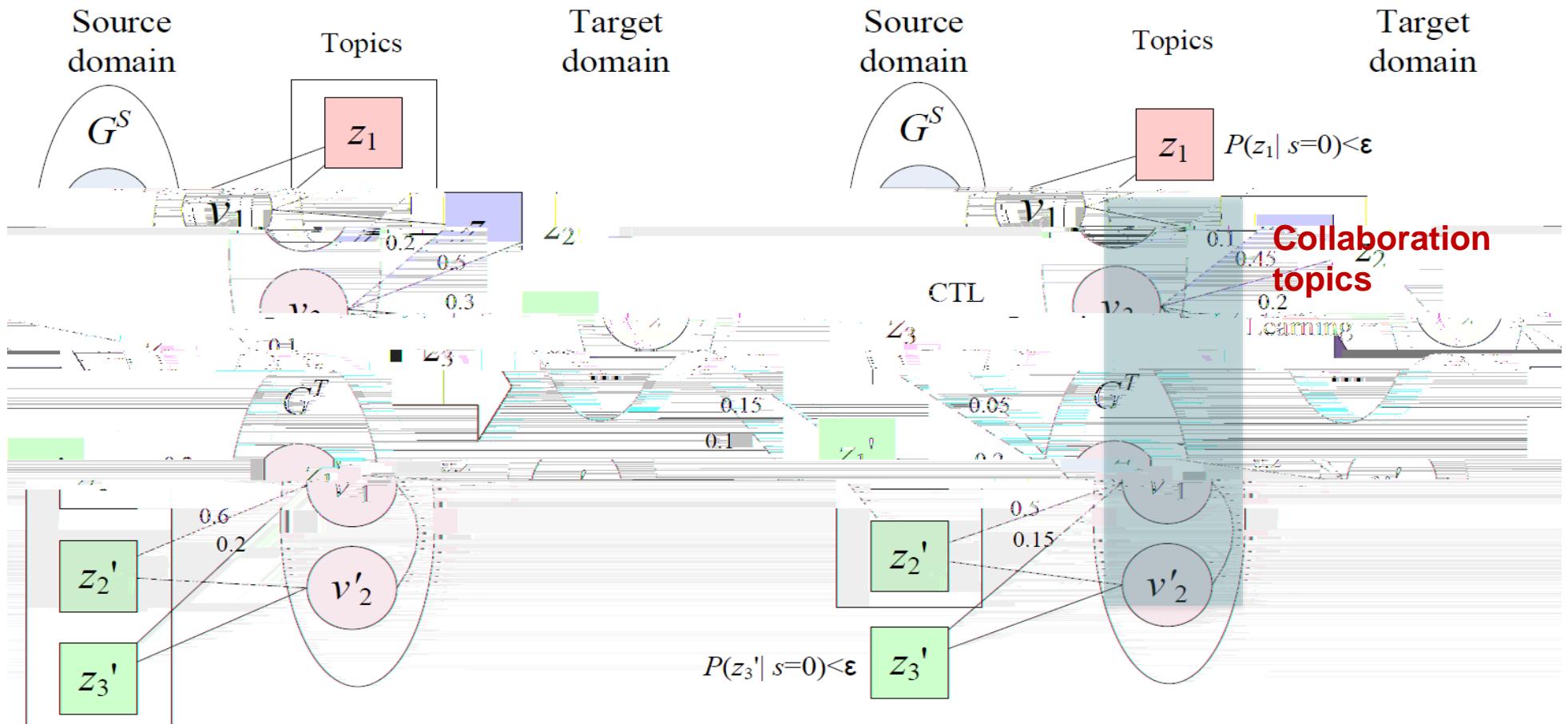
**Input:** a source domain  $G^S$  and a target domain  $G^T$   
**Output:** estimated parameters  $\theta, \theta', \phi, \vartheta$ , and  $\lambda$

```

Initialize an ACT model in  $G^S$  by learning from documents written by
authors only from  $G^S$ ;
Similarly, initialize an ACT model for target domain  $G^T$ ;
foreach collaborated document  $d$  do
    foreach word  $x_{di} \in d$  do
        Toss a coin  $s_{di}$  according to  $bernolli(s_{di}) \sim beta(\gamma_t, \gamma)$ ,
        where  $beta(\cdot, \cdot)$  is a beta distribution and  $\gamma$  and
         $\gamma_t$  are parameters;
        if  $s_{di} = 0$  then
            Randomly select a pair  $(v, v')$  from  $d$ 's aut-
            tors, where  $v$  is an author from  $G^S$  and  $v'$  from  $G^T$ ;
            Draw a topic  $z_{di} \sim multi(\vartheta_{vv'})$  from the
            mixture  $\vartheta_{vv'}$  specific to  $(v, v')$ ;
        end
        if  $s_{di} = 1$  then
            Randomly select a user  $v$ ;
            Draw a topic  $z_{di} \sim multi(\theta_v)$  from the topic
            model of user  $v$ ;
        end
        Draw a word  $x_{di} \sim multi(\phi_{z_{di}})$  from  $z_{di}$ -specific
        distribution;
    end

```

# Intuitive explanation of Step 2 in CTL





# Experiments



# Data Set and Baselines

- Arnetminer (available at <http://arxiv.org/collaboration>)

| Domain              | Authors | Relationships | Source                     |
|---------------------|---------|---------------|----------------------------|
| Data Mining         | 6,282   | 22,862        | KDD, SDM, ICDM, WSDM, PKDD |
| Medical Informatics | 9,150   | 31,851        | JAMIA, JBI, AIM, TMI, TITB |
|                     |         |               |                            |
|                     |         |               |                            |
|                     |         |               |                            |

- Baselines
  - Content Similarity(Content)
  - Collaborative Filtering(CF)
  - Hybrid
  - Katz
  - Author Matching(Author), Topic Matching(Topic)



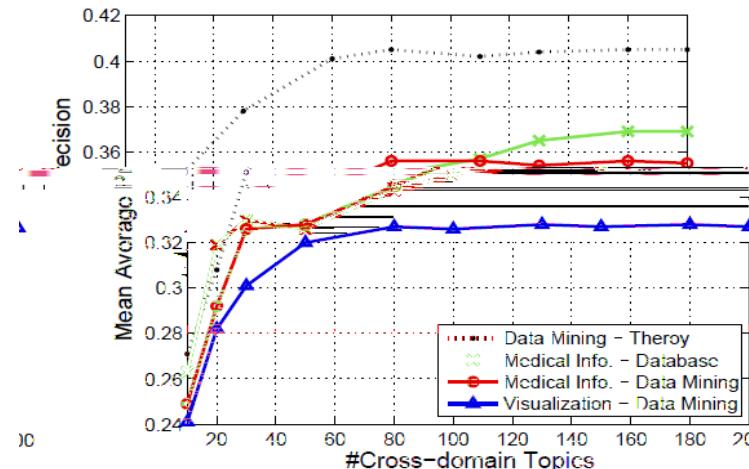
# Performance Analysis

Training: collaboration before 2001      Validation: 2001-2005

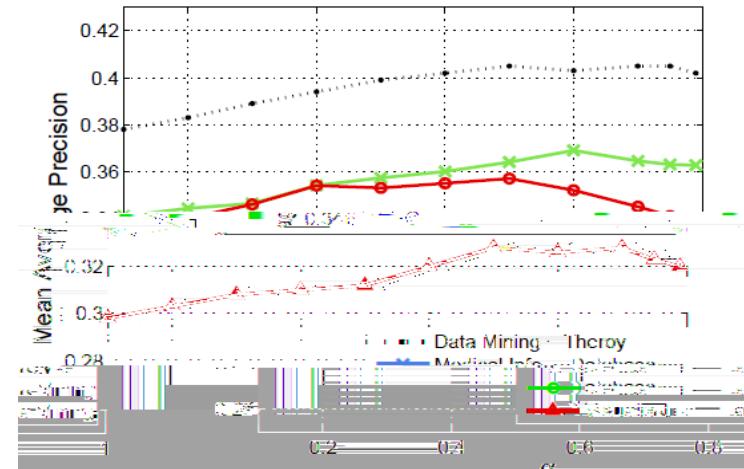
| Cross Domain                | ALG     | P@10 | P@20 | MAP  | R@100 | ARHR -10 | ARHR -20 |
|-----------------------------|---------|------|------|------|-------|----------|----------|
| Data Mining(S) to Theory(T) | Content | 10.3 | 10.2 | 10.9 | 31.4  | 4.9      | 2.1      |
|                             | CF      | 15.6 | 13.3 | 23.1 | 26.2  | 4.9      | 2.8      |
|                             | Hybrid  | 17.4 | 19.1 | 20.0 | 29.5  | 5.0      | 2.4      |
|                             | Author  | 27.2 | 22.3 | 25.7 | 32.4  | 10.1     | 6.4      |
|                             | Topic   | 28.0 | 26.0 | 32.4 | 33.5  | 13.4     | 7.1      |
|                             | Katz    | 30.4 | 29.8 | 21.6 | 27.4  | 11.2     | 5.9      |
|                             | CTL     | 37.7 | 36.4 | 40.6 | 35.6  | 14.3     | 7.5      |



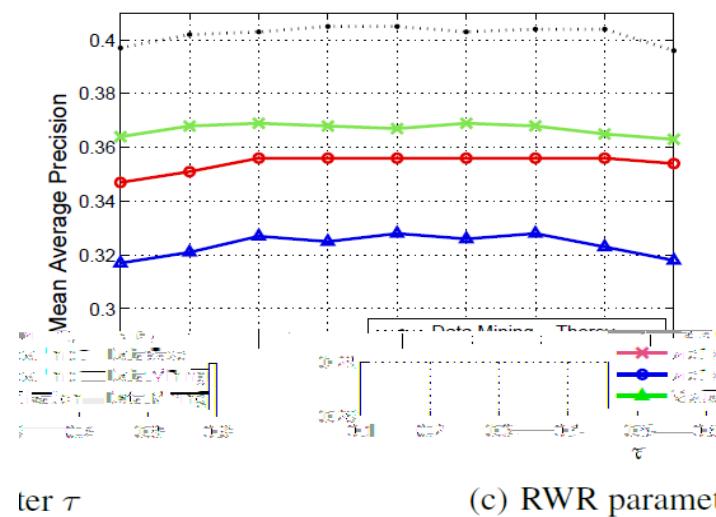
# Parameter Analysis



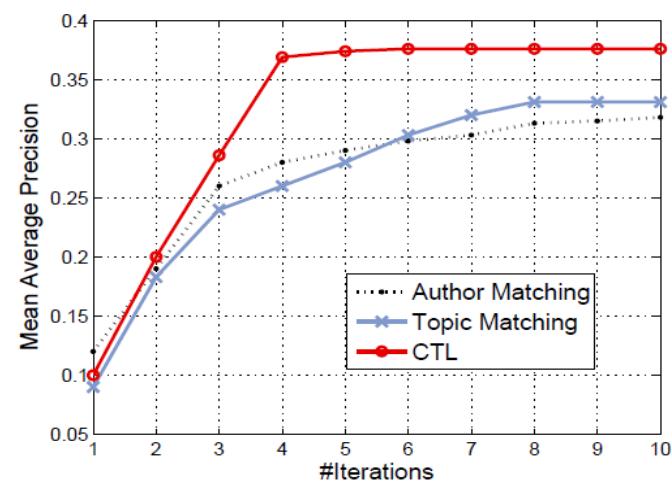
(a) number of topics  $T$



(b) Hyperparameter  $\alpha$



(c) RWR parameter  $\tau$



(d) Convergence analysis

(a) varying the number of topics  $T$

(c) varying the restart parameter  $\tau$  in the random walk

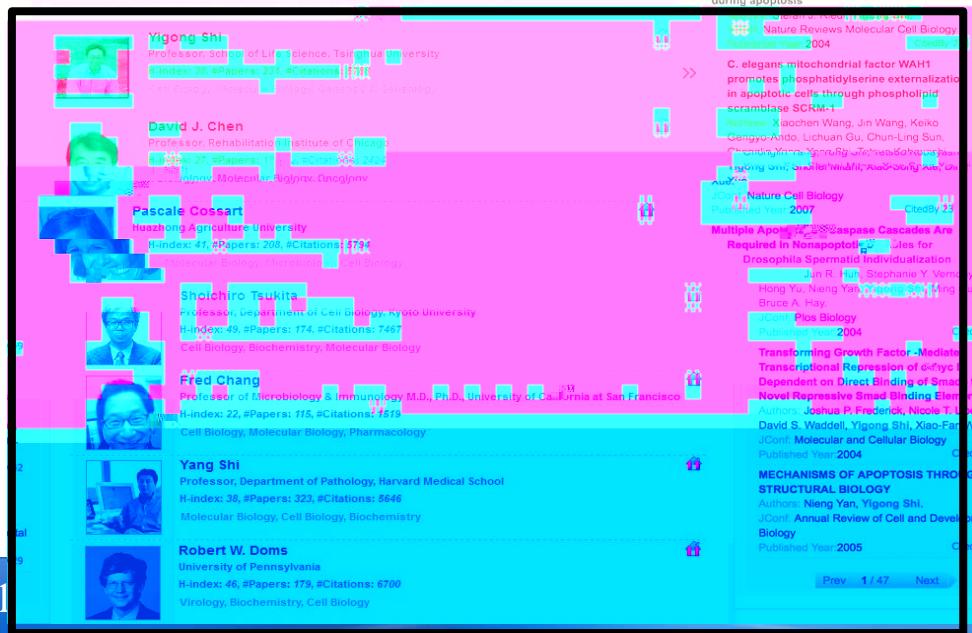
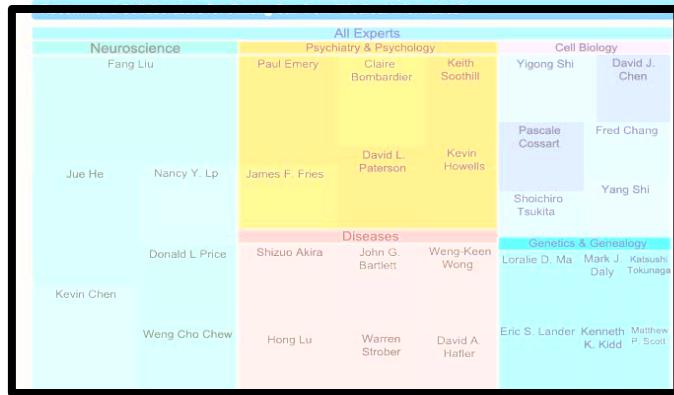
(b) varying  $\alpha$  parameter

(d) Convergence analysis

# Prototype System

<http://arneMiner.org/collaborator>

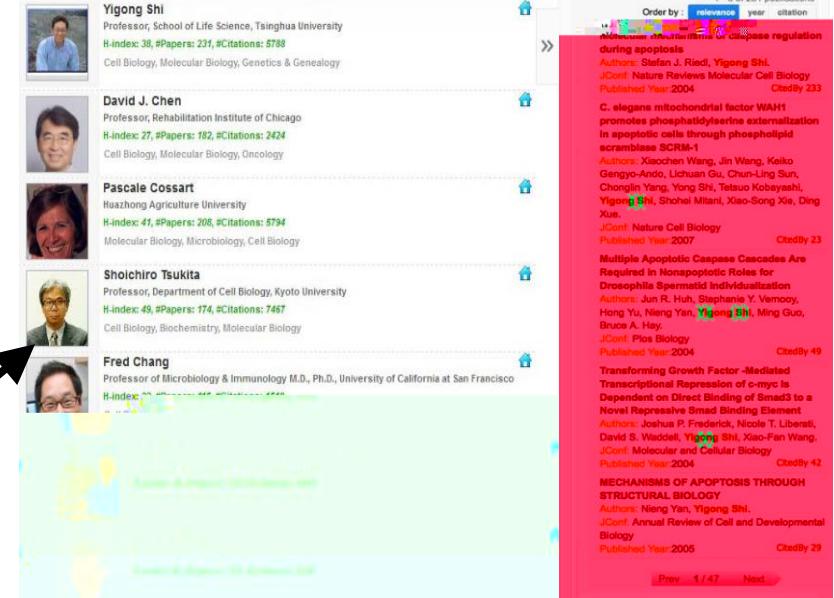
## Cross-Domain Collaboration Recommendation



Treemap: representing subtopic in the target domain

Welcome | Sign In | Sign Out

### Recommend collaborators for Jimeng Sun from "Cell Biology"



Recommend Collaborators &  
Their relevant publications



# Conclusion

- Study the problem of cross-domain collaboration recommendation
- Propose the cross-domain topic model for recommending collaborators
- Experimental results in a coauthor network demonstrate the effectiveness and efficiency of the proposed approach



# Future work

- Connect cross-domain collaborative relationships with social theories (e.g. social balance, social status, structural hole)
- Apply the proposed method to other networks



# Thanks!

System: <http://arxivminer.org/collaborator>

Code&Data: <http://arxivminer.org/collaboration>

# Challenge always be side with opportunity!



- Sparse connection:
  - cross-domain collaborations are rare;
- Complementary expertise:
  - cross-domain collaborators often have different expertise and interest;
- Topic skewness:
  - cross-domain collaboration topics are focused on a subset of topics.



# Performance Analysis

| Cross Domain                     | ALG     | P@10 | P@20 | MAP  | R@100 | ARHR -10 | ARHR -20 |
|----------------------------------|---------|------|------|------|-------|----------|----------|
| Medical Info.(S) to Database(T ) | Content | 10.1 | 10.9 | 12.5 | 45.9  | 3.6      | 2.1      |
|                                  | CF      | 18.3 | 20.2 | 21.4 | 47.6  | 5.3      | 3.9      |
|                                  | Hybrid  | 25.0 | 26.5 | 28.4 | 59.1  | 6.4      | 4.2      |
|                                  | Author  | 26.2 | 29.6 | 32.2 | 54.8  | 10.5     | 5.4      |
|                                  | Topic   | 29.4 | 26.3 | 34.7 | 59.3  | 11.5     | 5.2      |
|                                  | Katz    | 27.5 | 28.3 | 30.7 | 57.2  | 10.5     | 5.0      |
|                                  | CTL     | 32.5 | 30.0 | 36.9 | 59.8  | 11.4     | 5.4      |



# Performance Analysis

| Cross Domain                       | ALG     | P@10 | P@20 | MAP  | R@100 | ARHR -10 | ARHR -20 |
|------------------------------------|---------|------|------|------|-------|----------|----------|
| Medical Info.(S) to Data Mining(T) | Content | 5.8  | 5.7  | 9.5  | 19.8  | 1.9      | 0.9      |
|                                    | CF      | 13.7 | 17.8 | 18.9 | 34.3  | 2.7      | 1.3      |
|                                    | Hybrid  | 18.0 | 19.0 | 19.8 | 36.7  | 3.4      | 1.3      |
|                                    | Author  | 20.1 | 23.8 | 29.3 | 64.4  | 5.3      | 2.1      |
|                                    | Topic   | 26.0 | 25.0 | 33.9 | 48.1  | 10.7     | 5.6      |
|                                    | Katz    | 21.2 | 23.8 | 32.4 | 48.1  | 10.2     | 4.8      |
|                                    | CTL     | 30.0 |      |      |       |          |          |



# Performance Analysis

| Cross Domain                 | ALG     | P@10 | P@20 | MAP  | R@100 | ARHR -10 | ARHR -20 |
|------------------------------|---------|------|------|------|-------|----------|----------|
| Visual.(S) to Data Mining(T) | Content | 9.6  | 11.8 | 13.2 | 18.9  | 3.1      | 1.8      |
|                              | CF      | 14.0 | 20.8 | 26.4 | 29.4  | 6.9      | 4.3      |
|                              | Hybrid  | 16.0 | 20.0 | 27.6 | 30.1  | 6.3      | 4.4      |
|                              | Author  | 22.0 | 25.2 | 27.7 | 31.1  | 11.9     | 6.7      |
|                              | Topic   | 26.3 | 25.0 | 32.3 | 31.4  | 13.2     | 8.8      |
|                              | Katz    | 23.0 | 25.1 | 29.3 | 30.2  | 10.4     | 5.4      |
|                              | CTL     | 28.3 | 26.0 | 32.8 | 36.3  | 14.0     | 9.1      |