# **Cross-domain Collaboration Recommendation**

a<sup>†</sup>, U<sup>†</sup>, U<sup>‡</sup>, a a U<sup>†</sup> <sup>†</sup> a u a <sup>+</sup>, <sup>+</sup> ua <sup>‡</sup> a a<sup>+</sup>, a <sup>+</sup> u., a a., u.., b<sup>+</sup> a .uaa.u.

# ABSTRACT

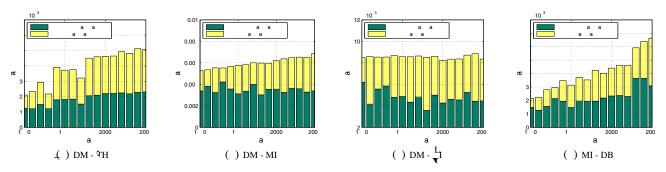


Figure 1: The comparison of existing collaboration and new collaboration trends over years. DM - Data Mining domain; MI - Medical Informatics domain; TH - Theory domain; VIS - Visualization domain; DB - Database domain. The trends of cross-domain collaborations in all but one case are growing (The exception between DM and VIS remain roughly constant over time). Newly formed cross-domain collaborations are significantly in all cases.

B.I. n.I.I. . n., w • ,Ç.- . -. . . u n ff 2 n\_ 1 ..., ...-\_\_\_\_nf.\_\_\_\_ M\_ 2 CVL w 1 1\_ \_ 1 ff\_ ÷. n\_ \* J., f. ff. In u. n. CVL , J.f. a. . n \_ \_ <u>, n \_ n \_ n</u> n, \_ f C L, n \_ n \_ n \_ i \_ . \_\_\_\_\_ \_\_\_\_\_

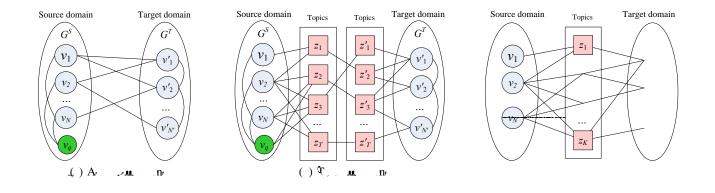
# 2. PROBLEM DEFINITION

 $D \qquad \dots, \quad I. \text{ Source/Target domain. } \begin{array}{c} \mathbb{T}^{\dagger} \\ \mathbb{T}^{\bullet} \\ \mathbb{T}^{\bullet}$ 

 $\begin{array}{c} \begin{pmatrix} \mathbf{i} & \mathbf{j} & \mathbf{i} & \mathbf{j} &$ 

Tifrn\_win\_\_\_invif, W \_ \_ -\_ . \_ n\_ . 1 \_ n\_ \_ . . ff 1n **u** n. n Nnf A ) 🖣 n W . H w ....n. ∠..-, щ п ¶ n, n\_\_\_\_, \_\_\_\_ n' \_\_n' ff 🗦 n . 🤄 n 111 . . . <u>.</u>n , I ∮ n\_ n ' **...** n . In \_ / \_ \_ W \_ n 🖣 ר א ה ח, ח אי . כ \_\_\_\_\_\_ ח, ח אי . כ أممه ار . . C \_ n w\_ w . . 4 **"** n? \_

#### 3. CROSS-DOMAIN TOPIC LEARNING



Algorithm 1:  $n \in \mathbb{N}$   $n \in \mathbb{N}$ 

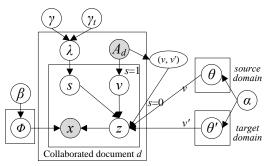


Figure 3: Graphical representation of CTL model.

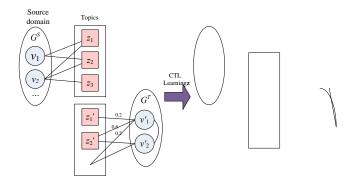
#### **3.3** Cross-domain Topic Learning (CTL)

Table 17 Notations in the CTL model.

MB L	DE C <b>/</b> I <sup>(</sup> T					
T	n m f ,					
d	$= f = - \sqrt{1 + 1}}}}}}} } } } } } } } } } } } } } }$					
$A_d$	$= \int \mathbf{f} \cdot \mathbf{f} \cdot \mathbf{f} \cdot \mathbf{M} \mathbf{n} d$					
di	i i					
di	di					
di	f di W ~ f ~ M n, " M n , ~ ~ M n					
$\theta_v$	$\mathbf{M}_{1,\ldots}\mathbf{n}_{1,\ldots}\mathbf{n}_{n,\ldots} = 1_{n,\ldots}\mathbf{n}_{n,\ldots}\mathbf$					
$\vartheta_{vv'}$	M. n. M					
	$ = - \mathcal{A}( , ') $ , ,					
$\phi_z$	<b>M</b> ( <b>, , n, m_, , , , , e</b> ), <b>n</b> <sup>1</sup> <i>w</i> ( <i>w</i> , , , <sup>1</sup> , . , , , .					
$\alpha, \beta$	$\mathbf{D}_{\mathbf{a}}$ , $\mathbf{n}_{\mathbf{a}}$					
$\lambda$	$[n_{i}, n_{i}] = n_{i} \cdot [n_{i}, n_{i}] + n_{i} \cdot [n_{i}, n_{i}]$					
$\gamma, \gamma_t$	$\mathbf{B}_{-} = \mathbf{M} [\mathbf{A}_{-}] \mathbf{M} [\mathbf{A}_{-}] \mathbf{\lambda}$					

CTL M 🕴 ... M. J. L. ~ f~ M., Ľ Tel in wenter to fee not w n n (s = 1), s = 1, s = 1, s = 1, s = 0; n = 1, s = 0; n = 1, s = 0; s = 0; s = 1, s = 0; s = 0; s = 1, s = 0; . n, 5-м Ť.

 $\begin{array}{c} F & \textbf{m} & \textbf{m} & \textbf{n} & \textbf{f} & \textbf{n} & \textbf{f} & \textbf{m} & \textbf{m} & \textbf{m} & \textbf{n} & \textbf{f} & \textbf{m} & \textbf{m} & \textbf{n} & \textbf{f} & \textbf{m} & \textbf{m} & \textbf{n} & \textbf{f} & \textbf{m} & \textbf{m} & \textbf{m} & \textbf{m}$ 



- - Medical Informatics: (1, n), (1, n $\mathbf{M}^{*} = \mathbf{M}^{*}, \mathbf{H}^{*} = \mathbf{M}^{*}, \mathbf{H}^{*}, \mathbf{$
  - Theory: (1 n + 1) = (1 1) +27,712 <u>-</u> <u>n</u> <u>n</u> <u>n</u>
  - Visualization:  $f = n + f = f = \frac{1}{L}$  (IEEE)2 =  $n = n = \frac{1}{T}$

 Table 2: Recommendation performance by different methods

 on the four cross-domain test cases (%).
 Content- Content

 Similarity; CF- Collaborative Filtering; Author- Author Matching;
 Topic- Topic Matching.

Cross domain	ALG	P@10	P@20	MAP	R@100	ARHR -10	ARHR -20
D_ Mnn ( ) T ~ (T)	C n n	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
	Н. 2	17.4	19.1	20.0	29.5	5.0	2.4
	A ~	27.2	22.3	25.7	32.4	10.1	6.4
	¢.	28.0	26.0	32.4	33.5	13.4	7.1
	K	30.4	29.8	31.6	27.4	11.2	5.9
	CĹL	37.7	36.4	40.6	35.6	14.3	7.5
M <sup>2</sup>	C n <sup>∎</sup> n	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
	Н. 2	25.0	26.5	28.4	59.1	6.4	4.2
	A .	26.2	29.6	32.2	54.8	10.5	5.4
D ()	¢	29.4	26.3	34.7	59.3	11.5	5.2
	K .	27.5	28.3	30.7	57.2	10.5	5.0
	CĹĹ	32.5	30.0	36.9	59.8	11.4	5.4
,	C n <sup>∎</sup> n	5.8	5.7	9.5	19.8	1.9	0.9
Mi	CF	13.7	17.8	18.9	34.3	2.7	1.3
Inf . ( )	H . z	18.0	19.0	19.8	36.7	3.4	1.3
	A ~	20.1	23.8	29.3	64.4	5.3	2.1
D_	£.	26.0	25.0	33.9	48.1	10.7	5.6
M_n_n/ (\mathcal{T})	K	21.2	23.8	32.4	48.1	10.2	4.8
	CŢL	30.0	24.0	35.6	49.6	12.2	6.0
L ▼ └ , ( )	C n n	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
	Н . 🛃	16.0	20.0	27.6	30.1	6.3	4.4
D_	A ~	22.0	25.2	27.7	31.1	11.9	6.7
Mnn/(T)	¢.	26.3	25.0	32.3	31.4	13.2	8.8
	K .	23.0	25.1	29.3	30.2	10.4	5.4
	C¢L	28.3	26.0	32.8	36.3	14.0	9.1

• f ~ **n** n u (< 80), n 🦾 a n' n n "m n n ģ. м - † м п. (° † Jn. w ъð n ո ա՝ 2. 1. . . . CTL M f W/ 1 1

'<del>`</del>, Hyperparameter analysis. \_ n -W The man in the second 4 4 .a. n f CJL  $\mathbf{w} = \mathbf{w} + \mathbf{w} +$ f ÇTL w \_ -, п " n 1 1 .f .a. n 1 ff In \_ n 0.03 \ 🗐 \_ n , † n n .a. C\Lui '\_~ f

~ [ h. ~ n , n , In ، أ سريد , أ , أ بر أ ، مـ أ أ ، , أ ม.ณี n\_ • n 1. - A. n n j j\_n nw W . /  $\begin{array}{c} n, w \quad = \\ n & w \quad = \\ n & w \quad = \\ n & w \quad = \\ \end{array}$ \_ff n, .--л  $\vec{P}_{n,n} = \mathbf{n} \cdot \mathbf{n} \cdot \mathbf{n} = \mathbf{n} \cdot \vec{P}_{n,n} \cdot \mathbf{n}$ 

Convergence analysis. I for infinite i

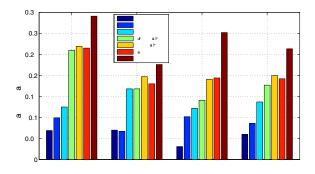


Figure 6: Performance on new collaboration prediction of all algorithms.

New Collaboration Prediction [T]n n w n 🚛 , .nn , 🏓 , w. мÌ. n In, 1 n ... 1 2 2 W <u>ա</u> ո\_ ոn, 🖣 n, →, n\_n, , n/n<sup>†</sup>w w \_, \_, \_, n. C \_ n, n. Ť fr., ≯?, ∜ n.Cn , J w i i n w W -์ - .....ี่ ก faunt f. 1, \_ . . n In <u>n \_</u> n • f ∠n w n 🦻 лŤ if i 2001 f n. In\_ - \_ \_ n ∠\_n\_n\_\_n \_ † \_  $\mathbf{n}, \mathbf{w}', \mathbf{n}, \mathbf{n}, \mathbf{n}''$ w. J.J.f. au nJ fn w n, 🕏 n 2 M. n. 1. 1. f . . . n £ M\_ ~ n\_ / í. .u. 2<sup>1</sup>-**M**, **, , , , , , , , ,** 2 M \_ n, 17  $\mathbf{f} = \mathbf{f} =$ \_ C(TL\_ n \_ \_ nnТ , / n

### 4.3 Prototype System

w nf ~ <sup>5</sup>. ℃ • CULM ( , -1,932,442 n \_\_\_ n \_\_\_ n n h wing n - **n** n j. n٩ .) , м, м, , A.में. म.में २. n, \_ n 1 ..... n 2 1 ..... 1. T1n м 🏓 2 n 1 w n 'n , n m w , w . I CUL m I . . . 2 - - 1  $\mathbf{M} (Cf. 3.3) \quad \mathbf{n} \cdot \mathbf{n} = \mathbf{n}$ ¶ n\_ ~ n 1 **.u.** n.

# 5. RELATED WORK

الم م n م С n 5 .**....** n\_ n. \_ \_ n \_,**u**, n u n  $n^{1}$   $\mathcal{L}$   $C^{1}$   $n^{1}$ . 7 fĴ ∮ņ-С 1 1 2n n T f 13 n 11 n 1 . প n W n 1.-W \_. Ž. . n . . \_ n , Ž . . . . . . . . . . 44

<sup>5 ... : # .# . . . . . . . . . .</sup> 



- 22 L.L = (n + n + n + c), (A + c), C = b + a c, 2(1):1?6, 1993.
- n/
- *E* a dJ a M dc , 354:2463 2472, J n<sup>2</sup> 2006. 25 D. , C an n<sup>2</sup> J J n n  $_{-}$  n n'. In *KDD'10*, , ,
- 979 988, 2010.

- 418 425, 2005.

- 31 J.  $\mathfrak{L}$  n<sup>r</sup>, J. \_ n<sup>r</sup>, L. \_ , J. L., L. \_ n<sup>r</sup>, n . . . . A.  $\mathfrak{n}$  . . .  $\mathfrak{n}$   $\mathfrak{n$
- In KDD'09, \_ + 817 826, 2009.
- 33 ,  $\mathfrak{T}\mathfrak{n}$ ,  $\mathfrak{J}.\mathfrak{T}\mathfrak{n}$ ,  $\mathfrak{T}.\mathfrak{D}$ ,  $\mathfrak{C}.\mathfrak{T}\mathfrak{n}, \mathfrak{B}.\mathfrak{G}$ ,  $\mathfrak{n}$ ,  $\mathfrak{T}.\mathfrak{L}$ ,  $\mathfrak{n}$ ,  $\mathfrak{m}$ ,  $\mathfrak{n}$ 76(1):71 83, 2012.

- 36. J.  $\underline{n}$ , J.  $\underline{m}$ , J.  $\underline{m}$ ,  $\underline{n}$  J. L. E.  $\underline{f}$ ,  $\underline{n}$ , \underline{n},  $\underline{n}$ ,  $\underline{n}$ , \underline{n},  $\underline{n}$ ,  $\underline{n}$ ,  $\underline{n}$ ,  $\underline{n}$ , \underline{n},  $\underline{$ DASFAA'07, \_\_\_\_ 1066 1069, 2007.

# 8. APPENDIX