ACTIVELY DISAMBIGUATING PERSON NAMES WITH USER INTERACTION

Xuezhi Wang*, Jie Tang*, Hong Cheng[‡], Philip S. Yu[†] *Tsinghua University, ‡The Chinese University of Hong Kong, †University of Illinois at Chicago

Cheng Chang 🗞 v 💩

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Bin Yu is a Postdoctoral Fellow in the School of Computer Science at Carnegie Mellon University and he is working with Dr. Katia Sycara. Prior to that, he was a ...

EXISTING METHODS FOR NAME DISAMBIGUATION

- Supervised-based approach:
 - Learn a specific classification model from training data
 - Use model to predict the assignment of each paper
- Unsupervised-based approach:
 - Clustering algorithms to find paper partitions.
 - Papers in different partitions are assigned to different persons.
- Constraint-based approach:
 - Utilizes the clustering algorithms.
 - User-provided constraints are used to guide the clustering towards better data partitioning.

EXISTING METHODS WITH INTERACTION

G F:\WebAssistant\Data Files\award show.txt	🔽 🔛 🔝 Font Size 🧊 🗕 Highlight — 🤇
ORLANDO, Fla Texas quarterback Vince Young and Southern California ta best play,")in the nation as ")one for being the best at his position, as colled for the university of Texas at Austin	ilback <u>Reggie Bush</u> each took home two awards Thursday night, one as ball honored its stars. y of Southern California
Young won the Maxwell Award as the nation's top player and the Davey O'Brier	Award given to the nation's top quarterback.
Bush came away with the Walter Camp player of the year award and the Doak Walter Camp Award "Tm Reggie Bush "Tm Just Chipoying the moment right now, Young said." All of these awards mean a teammates have put in this year. I guess we've "t another one to go (the Heisma Vince Young	<u>Walker Awand</u> given to the nation's best running back. a great deal to me, because of all of the hard work that me and my <u>n</u>) on Saturday."
Young and Bush will be in <u>New York</u> on Saturday night as finalists for the <u>Heiler</u> <u>Young bear out Bush</u> and <u>Leinart</u> for the <u>Maxwell</u> , and <u>Leinart</u> and <u>Notre Dame's</u> to win the award. <u>Matt Leinart</u> <u>Maxwell Award</u>	isman Trophy USC quarterback <u>Matt Leinart</u> is the other <u>Heisman</u> finalist. s <u>Brady Quinn</u> for <u>University of Southern California</u> the first <u>Texas</u> quarterback of Notre Dame

• Several problems:

- User has to check every result to see if it is correct
- No propagation, correction only based on user input

ALGORITHM DESIGN

• How to combine features, relations and user feedback?

Feature, between document pair and label
Relation, between label and label
User Feedback, constraint on partial labels

We need a model to elegantly combine these altogether
Inference on the model can give us the answer to paper assignment



LEARNING ALGORITHM FOR PFG

Input: number of iterations; **Output**: learned configuration for *Y*;

```
2.1 Initialize all \theta = (\{w_k\}, \{\mu\}, \{\alpha_l\}) as 1;
 2.2 Initialize all hidden variables Y = \{y_{ij}\} with y_{ij} = 0;
 2.3 repeat
           % sample a new configuration Y' based on q(Y'|Y);
 2.4
           Y' \leftarrow q(Y'|Y);
 2.5
          \tau \sim \min(\frac{p(Y', X|\theta)}{p(Y, X|\theta)}, 1);
 2.6
           toss a coin s according to a Bernoulli(\tau, (1 - \tau));
 2.7
           if (s = 1) then
 2.8
                % accept the new configuration Y';
 2.9
                Y \leftarrow Y':
2.10
          end
2.11
2.12 until convergence;
2.13 return Y;
```

2: The MH-based learning algorithm for PFG.

Metropolis-Hasting
 Algorithm for
 Approximate Inference

WHY ACTIVE NAME DISAMBIGUATION?

Wei Wang

University of North Carolina at Chapel Hill)

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Order by "Wei Wang".

Wei Wang

)thers)

Dubious

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Are they correct?

How to find document pairs that are most likely to be wrongly classified?

UNCERTAINTY-BASED ACTIVE SELECTION





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MODEL REFINEMENT

 \bigcirc Maximizing the conditional probability P(Y | X)

- SampleRank algorithm
- o for θ ∈ {*w*_{*k*}, *μ*, *α*_{*l*}}, parameters in our PFG model
- *y* : original configuration; *y*': new configuration
- $\theta = \theta \begin{cases} -\eta \cdot \phi_{y',y} & \text{if } y \text{ is preferred } and M(y',y) > 0 \\ +\eta \cdot \phi_{y',y} & \text{if } y' \text{ is preferred } and M(y',y) \le 0 \end{cases}$
- where η is the learning rate
- $M(y', y) = \theta \cdot \phi_{y', y}$ is the unnormalized log probability ratio according to the Metropolis-Hastings Model

Improving Efficiency by Atomic Cluster

- In practice, enumerating all possible document pairs can be really time-consuming and infeasible for an online system
- Atomic cluster-based method
 - Atomic cluster: in this cluster every paper has very high probability that they belong to the same person
 - Bias-classifier——AdaboostM1, aiming to minimize the number of false positives, thus obtaining very high precision



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DATA SET

• Publication Data Set

- From ArnetMiner.org, manually labeled 6,730 papers for 100 author names
- CALO Set
 - Email Directory, labeled data set of 1,085 webpages for 12 names
- News Stories
 - 755 ambiguous entities appearing in 20 web pages

Dataset	#Names	#Persons	#Documents
Publication	100	1,382	6,730
CALO	12	187	1,085
News Stories	380	755	20

EXPERIMENT

Publication Data Set (Average)

Precision 95.4%Recall 85.6%F1-score 89.2%



Performance(F1-score) of the comparison methods.

	Method	Recall	Precision	F1-score	
CALO Set	LS+A/CDC ^[5]	0.745	0.869	0.803	
	Our Approach	0.761	0.878	0.815	

News Data Set	Method	Baseline in [23]	Approach in [23]	Our Approach	
	Accuracy	0.517	0.914	0.973	

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• Result of active name disambiguation (MR: the model refinement)

- UB: Uncertainty-based active selection
- IM: Influence Maximization-based active selection

Method	Ran	dom Selectio	on-MR	Random Selection+MR			Active Selection (with UB)+MR			Active Selection (with IM)+MR		
#Query	Recall	Precision	F1-score	Recall	Precision	F1-score	Recall	Precision	F1-score	Recall	Precision	F1-score
0	0.856	0.954	0.892	0.856	0.954	0.892	0.856	0.954	0.892	0.856	0.954	0.892
2	0.857	0.954	0.893	0.867	0.953	0.899	0.896	0.953	0.915	0.892	0.955	0.921
5	0.855	0.954	0.891	0.873	0.953	0.904	0.922	0.952	0.930	0.976	0.953	0.959
10	0.863	0.956	0.897	0.885	0.951	0.909	0.937	0.953	0.939	0.994	0.952	0.969
20	0.889	0.963	0.917	0.905	0.959	0.926	0.958	0.953	0.952	0.996	0.951	0.969
30	0.903	0.964	0.927	0.915	0.961	0.933	0.965	0.953	0.955	0.997	0.951	0.969

• How F1-score varies with number of queries



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Thank you!