

#### PGT: Measuring Mobility Relationship using Personal, Global and Temporal Factors

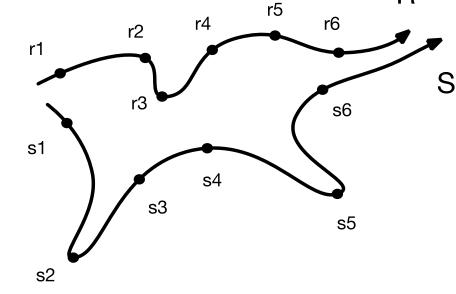
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ICDM 2014 Shenzhen

Measure the mobility relationship strength

Given trajectories of two users, measure their relationship strength

ID	Location	Time-stamp
R	40.812, -77.856	2014-11-22 13:00:00
R	40.770, -77.855	2014-11-22 13:30:40
R	40.774, -73.975	2014-12-27 10:00:00



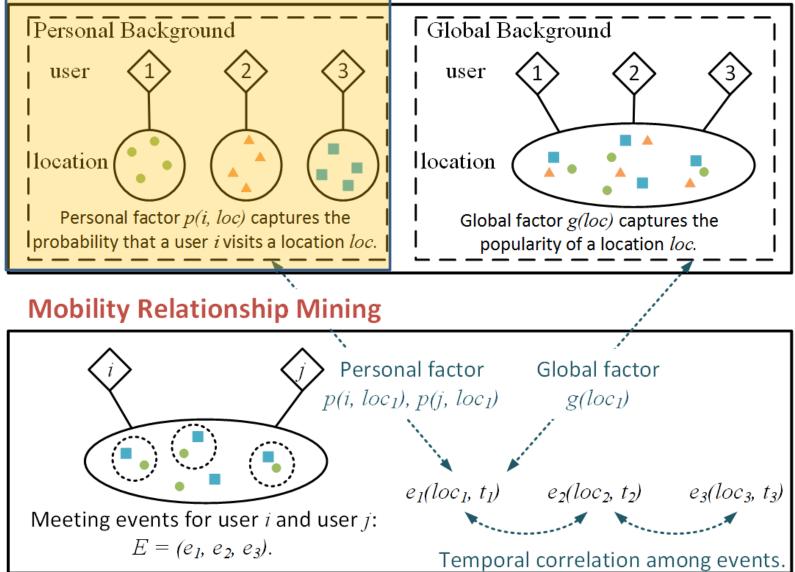
- Application
  - Recommendation
  - Crime investigation

#### **Baseline Method -- Meeting Frequency**

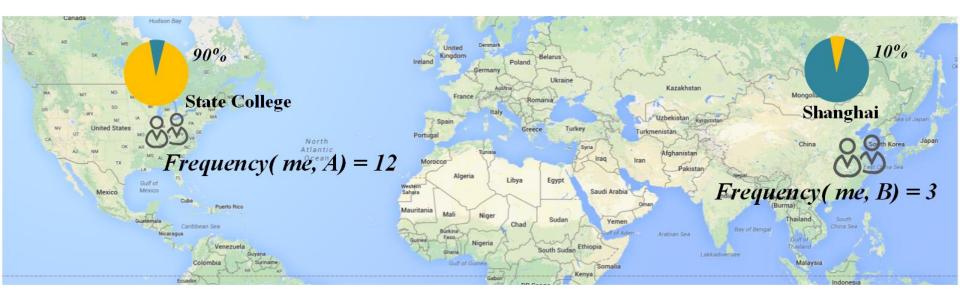
the more frequently you co-locate with another person, less frequently

the stronger the mobility relationship is. weaker

#### **Background Modeling**



### Personal Background is important

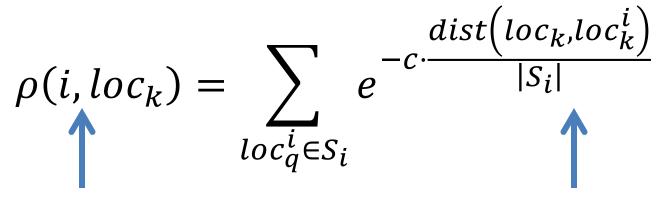


Shanghai has a lower probability to be visited. Co-location in Shanghai is less likely, but it happens.

Co-location event in Shanghai should carry higher weight.

# Personal Background Formulation

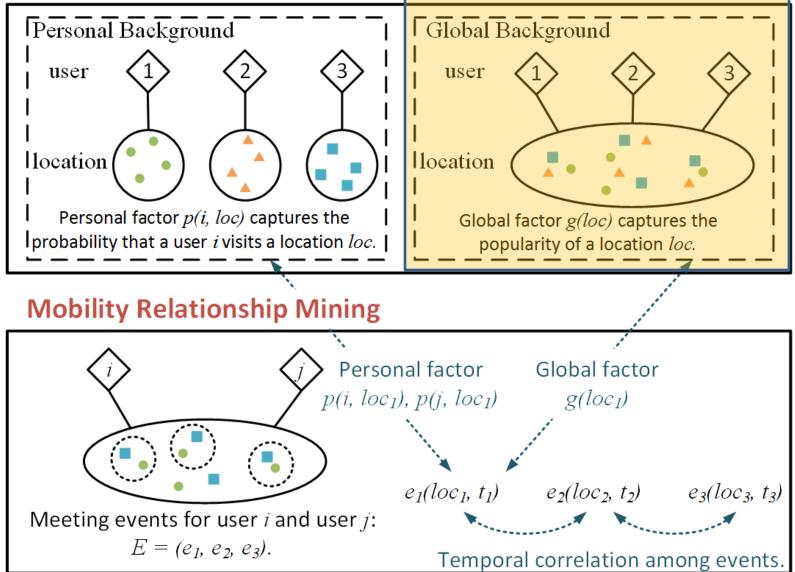
For given user *i*, the probability of visiting location *loc<sub>k</sub>* is



The visited location is far from others, the probability is low.

Judge whether visited location is close to others.

#### **Background Modeling**



### **Global Background Matters**

- A and B meet in downtown for 10 times.
- C and D meet in D's house for 10 times.

#### Relationship(A,B) = Relationship(C,D)

# **Global Background Formulation**

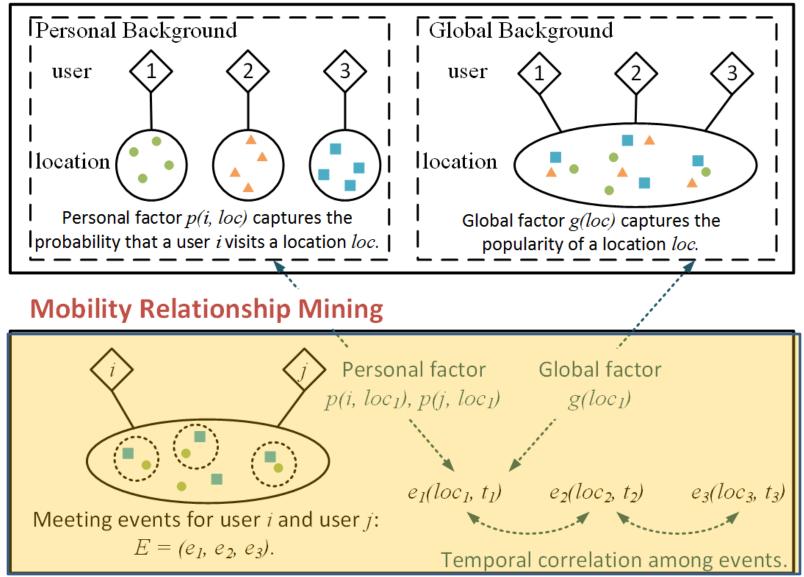
$$P(i, loc_k) = \frac{|S_i(loc_k)|}{\sum_i |S_i(loc_k)|}$$

At  $loc_k$ , the probability of observing different use *i*.

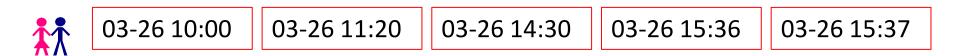
$$g(loc_k) = -\sum_{i:P(i,loc_k)\neq 0} P(i,loc_k) \cdot \log P(i,loc_k)$$

Entropy of  $loc_k$ . Less users visited -> lower entropy -> more private location

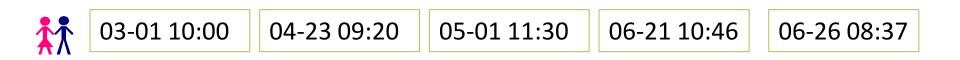
#### **Background Modeling**



#### **Temporal Correlation Between Events**



Continuous meeting events  $\rightarrow$  probably one-time trip?



Sporadic meeting events  $\rightarrow$  a stronger relationship indication

# **Related Work**

- Co-location frequency as measure (without considering background):
  - Kalnis et al. SSTD, 2005
  - Jeung et al. VLDB, 2008
  - Li et al. VLDB, 2010
  - Cranshaw et al. Ubicomp, 2010.
  - Zheng et al. ICDE, 2013
- Global factors: *Pham et al. SIGMOD, 2013.*
- Personal factors: None
- Temporal factors: None

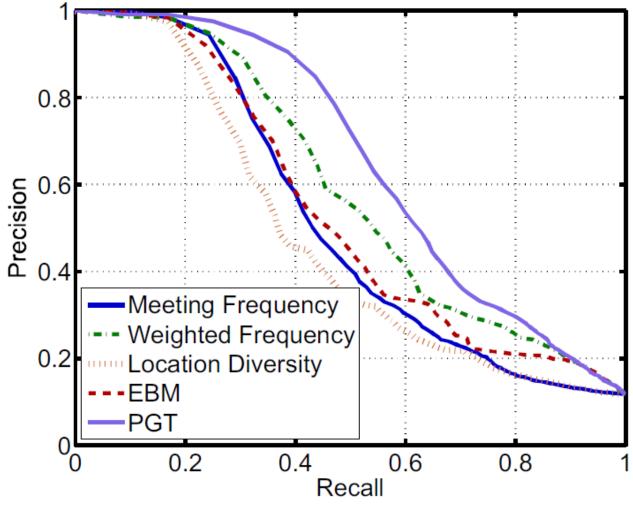
#### Experiments

- Datasets two location-based social networks check-in data\*
  - Gowalla (Feb, 2009 Oct, 2010)
  - Brightkite (Apr, 2008 Oct, 2010)

	Gowalla	Brightkite
No. of users	107,092	58,228
No. of friend pairs	950,327	214,078
No. of check-ins	6,442,890	4,491,143
Average check-ins per user	60	78

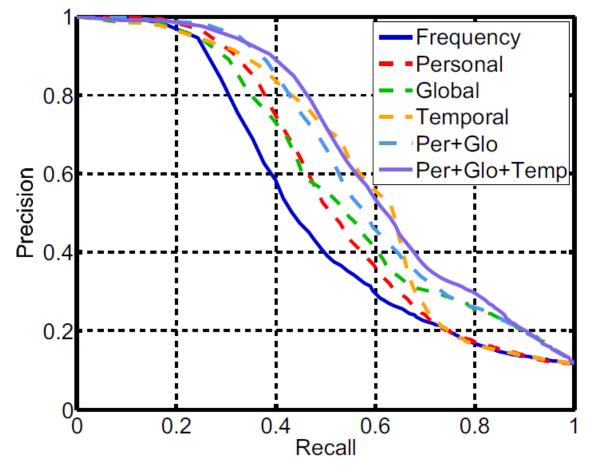
\* E. Cho, S. A. Myers, and J. Leskovec, "Friendship and mobility: user movement in location-based social networks," in Proc. KDD, 2011.

# Experiments: Compare with the State of the Art on Gowalla



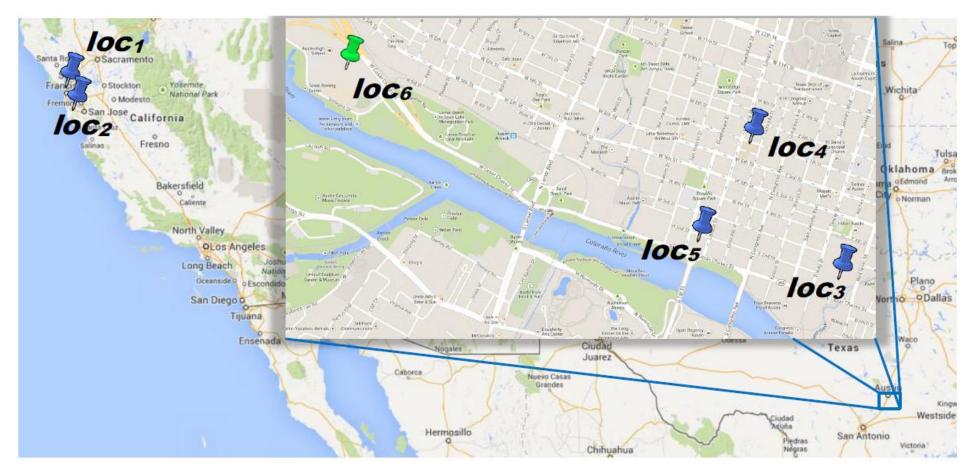
Measuring Mobility Relationship Hongjian Wang, Penn State University

#### **Experiments: Compare Various Factors**



The precision-recall curves on top 5000 users from Gowalla.

### Case Study: Personal Factor works



#### Both pairs meet 5 times in total. Blue Pair are friends. Green not.

#### Personal Profile of the Four Users



(a) User #267

(b) User #510



(c) User #350



(d) User #6138

# **Results using Different Measures**

User Pair	Friends / Not	Frequency	Personal Factor
#267, #510	Yes	5	22.03
#350, #6138	No	5	9.72

#### First Pair is more likely to be friends.

# Summary

- We propose a unified framework to measure the strength of relationship based on two users' mobility.
- Our model is simple and deterministic, which considers:
  - Personal probability visiting a location
  - Location popularity from general public
  - Temporal correlation among co-locations

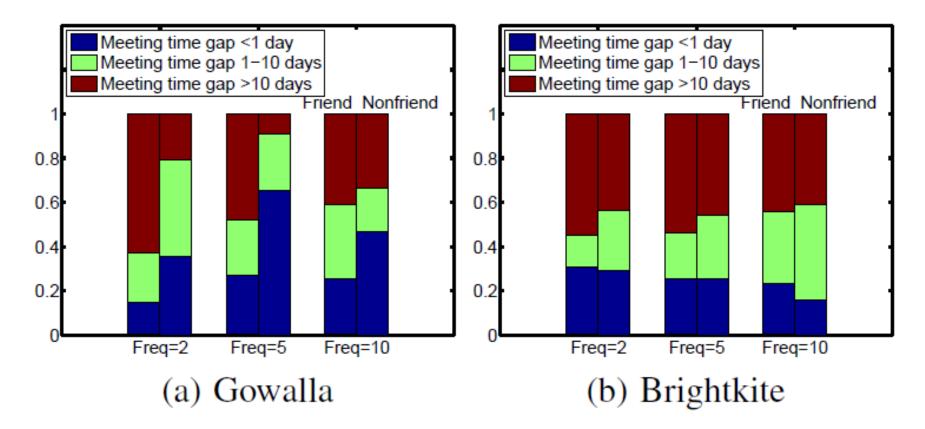
### Future work

- Extend this work from identifying pairwise relationships to discovering common interest groups.
- Further combine the context at each location, such as the activity at that location.

### **Dataset Properties**

- The Gowalla users tend to check-in at featured spots, and recommend places and trips for others.
- The Brightkite users tend to check-in with acquaintance to maintain personal social circle.
- As a result, check-ins in Gowalla are mostly made on popular places.

### **Datasets Have Different Properties**



# The distribution of time gaps between consecutive meeting events for three representative groups (meeting frequency = 2; 5; 10).

#### Social Relation From Geospatial Data

• Diversity of co-locations

Co-occurrence Vector	Shannon Entropy	$D_{ij}$ Value	Diversity	Likelihood of Coincidences	Prob. of a Friendship
$C_{12} = (1, 1, 1, 1, 1, 0, 0, 0, 0, 0)$	1.609	5.000	High	Low	High
$C_{23} = (1, 2, 1, 1, 0, 0, 0, 0, 0, 0)$	1.332	3.789	Medium	Medium	Medium
$C_{13} = (0, 0, 4, 0, 0, 0, 0, 0, 0, 0)$	0.000	1.000	Low	High	Low

#### Table 1: Example of Diversities

#### High diversity -> high probability of friendship

H. Pham, C. Shahabi, and Y. Liu, "Ebm: An entropy-based model to infer social strength from spatiotemporal data," in Proc. SIGMOD, 2013.