

Importance of Friend Recommendation

- With our limited time and so many sites to choose from, we often face a dilemma of choosing a few sites over others.
- Users prefer more engaging sites, where they can find familiar faces such as friends, relatives, or colleagues.
- Users leave sites easily when they cannot find friends: 60% of Twitter/Facebook Users quit within the first month

Solution: Recommend Friends Using Link or Content Information

<u>Why Traditional Methods Fail!</u>

- There is no link or content!
- No Content: right after a user joins a new site,
- there is no profile information or site activity. • No Link: right after a user joins a new site, the user
- is a disconnected singleton in the friendship graph.
- This is a **universal** problem for all sites and for **each** and every user, right after she joins a new site.
- Isn't that the cold start problem?
- In cold-start, one often assumes that either link or content is available.



 Then what do sites do? In Twitter: Recommending Celebrities or political figures in the United States or the world.

Finding Friends on a New Site Using Minimum Information

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Finding Friends with Minimum Information

- What is the minimum information available: Username
- Example: What Twitter is doing: • The search space for finding
 - Friends is the whole Twitter
- What we propose: • Reduce the search space
- Let β for a new user denote the search space reduction ratio.
- For example, when search space is 1/10 of Twitter, $\beta = 10$
- Because we have many users, we are interested in the

expected Value of β

- How can we find these search spaces?
- •**Observation 1**: we have *assortative mixing* (friends are similar)
- **Observation 2**: we can partition Twitter and each new user is assigned to a partition, i.e., no overlapping search spaces

• How do we partition?

- We can partition based on attribute values • For example, language: partitions representing users speaking English/French/etc.
- What attributes should we select?
- We can select attributes that represent the three
- general social forces that result in friendships:
- **Homophily** (attribute selected: age)
- Influence (Fitting well within the friends crowd)

Data Mining and Machine Learning Lab



• **Confounding** (attributes selected: Language and Location)

<u>Predicting the Attributes From Usernames</u>

- We predict the language, age, and the location from the usernames
- The age and location influence the username
- usernames with their age.







1. To train an age classifier, we use a set of 80,000

2. For predicting location, we train over a set of 36 million geo-located usernames from Twitter. 3. For language prediction, we train an n-gram language detector over 40 million words.

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