# Users Joining Multiple Sites: Distributions and Patterns

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### **Users Join Multiple Sites!**

- Our social media life is no longer limited to a single site.
  We post on Reddit, like on Facebook, tweet on Twitter, watch on YouTube, listen on Pandora, among many other activities exhibited by social media users.
- Users prefer more engaging sites, where they can find familiar faces such as friends, relatives, or colleagues.
- On average, popular sites with more members are expected to contain more friends for an average individual.

#### **Question:**

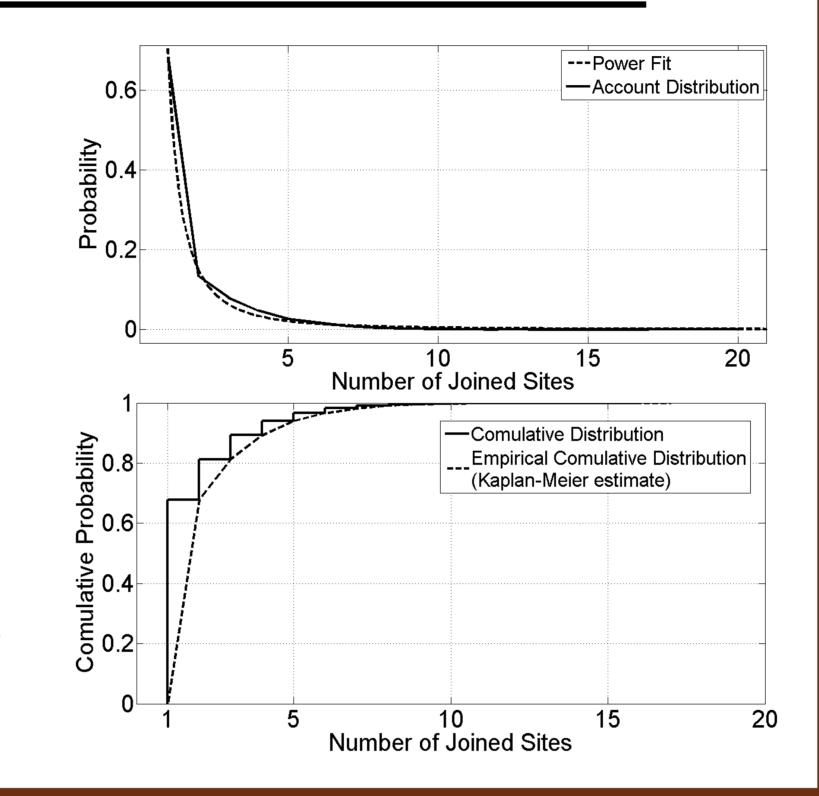
Does this fully explain users' site selections?

# **Data Preparation**

- We can survey individuals for their accounts
  Not Scalable + Expensive
- We can also utilize automatic approaches to connect corresponding identities of users across sites.
- Users often list their accounts on social networking sites, blogging and blog advertisement portals, and forums.
- We collected 96,194 users having accounts on a subset of 20 social media sites:
  - BlogCatalog, BrightKite, Del.icio.us, Digg, Flickr, iLike, IntenseDebate, Jaiku, Last.fm, LinkedIn, Mixx, MySpace, MyBlogLog, Pandora, Sphinn, StumbleUpon, Twitter, Yelp, YouTube, and Vimeo

#### **User Membership Distribution across Sites**

- More than 97% of users have joined at 1 to 5 sites.
- A power function  $(g(x) = 0.6761x^{-2.157})$  with 95% confidence fits to the curve with  $R^2 = 0.9978$
- A maximum likelihood estimation shows that the distribution is *Power-Law*



## <u>User Membership Patterns across Sites</u>

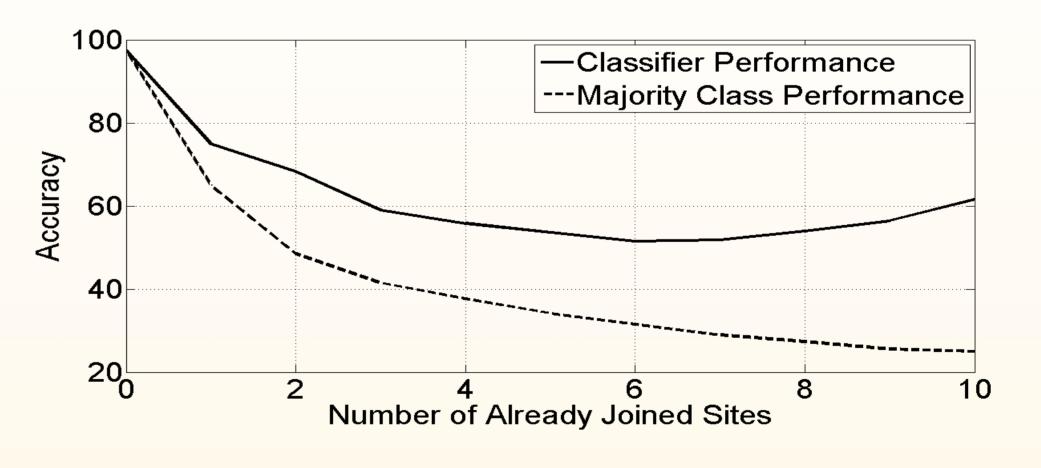
- We find sites that users join together
- If users join sites with a probability that is proportional to their popularity
- The expected overlap between two sites is  $\frac{d_i d_j}{2m}$ .
- $\succ$  Given the actual overlap between the two sites,  $O_{ij}$ , we can compute how non-random joining both sites is.
- > The problem is reduced to weighted modularity.
- There are sites that users join all to be able to access the content that becomes available on each one of them.
- There are popular sites that users join all (or most) to satisfy their basic needs (average user behavior).
- > There are [unknown/new] sites that early adopters join.

### **Evaluating via Recommending Sites to Users**

- By identifying the type of site selection patterns a user has exhibited in the past, we recommend new sites to the user
- For users that have joined n sites:
- ➤ We assume that given the category of *n-1* of them, the category of the *n*th site should be predictable.
- We generate all the possible combinations of n-1 sites and use the number of sites in each category as features (4 features) and the category of the *n*th site as the label.

Technique	AUC	Accuracy
J48 Decision Tree Learning	0.880	79.25%
Random Forest	0.895	79.17%
Logistic Regression	0.886	79.14%
SMO (Sequential Minimal Optimization)	0.728	78.92%
Naive Bayes	0.869	76.66%

- When users haven't joined any sites, they join popular sites: majority prediction is as accurate.
- As users join more sites, preference play an important role: majority prediction = random.



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