# Analysis of Twitter unfollow: How often and why

SocInfo 2011 Singapore Management University

> Haewoon Kwak Hyunwoo Chun Wonjae Lee Sue Moon





# Two basic processes in network evolution





#### Building a relationship



Breaking a relationship







#### **People You May Know**





Jihae Kim 15 mutual friends 纪 Add as friend

#### Building a relationship

#### Breaking a relationship

#### Friends You Are To Break Up Soon



#### Your Dad

"Dark side of the force"

????

# Lack of relationship breakup data

- Can you capture the breakup from
  - A paper coauthorship network?
  - A mobile phone call network?
  - An e-mail network?
  - A wall message network?

#### Very hard to define and capture breakup





#### Canceling the friendship online?



### Unfortunately (?), people do take online etiquette seriously.







#### Click on a friend to begin the sacrifice.

Adam Sachs



Adam Gilad



Adriana C.

Alex Argrow





Alex Panelli



Amiee Parco





Amy Alex Andersen







Adam Zand

Amanda Wheeler







Use this application and be rewarded with a free flame-broiled WHOPPER® when you sacrifice 10 of your Facebook friends. Each friend will be notified so choose wisely.





Share The Sacrifice

# A proxy for relationship breakup

• Disappearance of e-mail exchange

"Alice & Bob had exchanged e-mails frequently.

At some point they didn't do any more." But they were doing instant messaging.

A proxy is not always accurate!





#### How about Twitter?



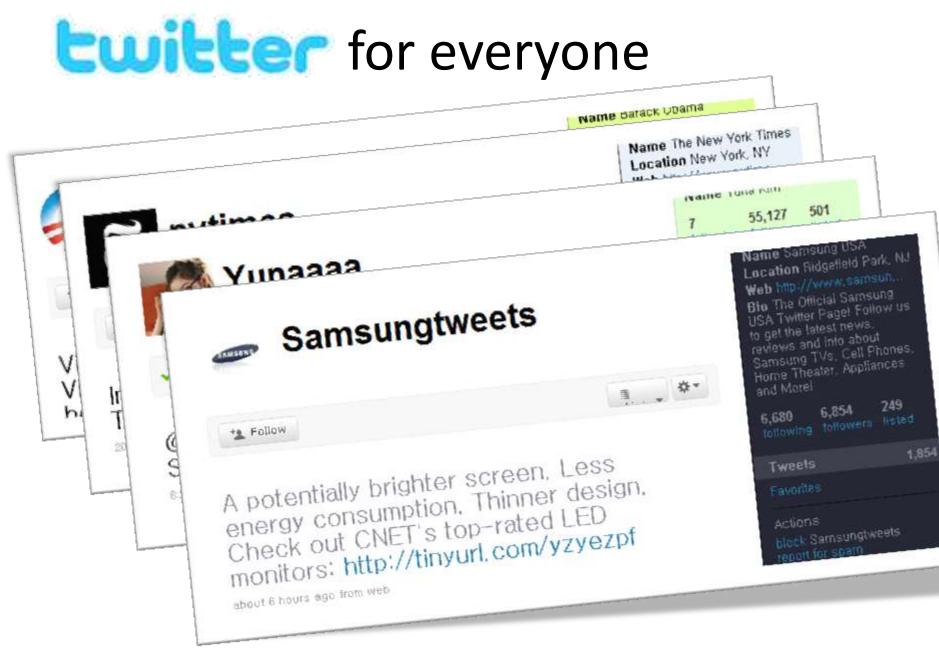


### Different from other OSNs













HI.

19th International World Wide Web Conference (WWW2010)



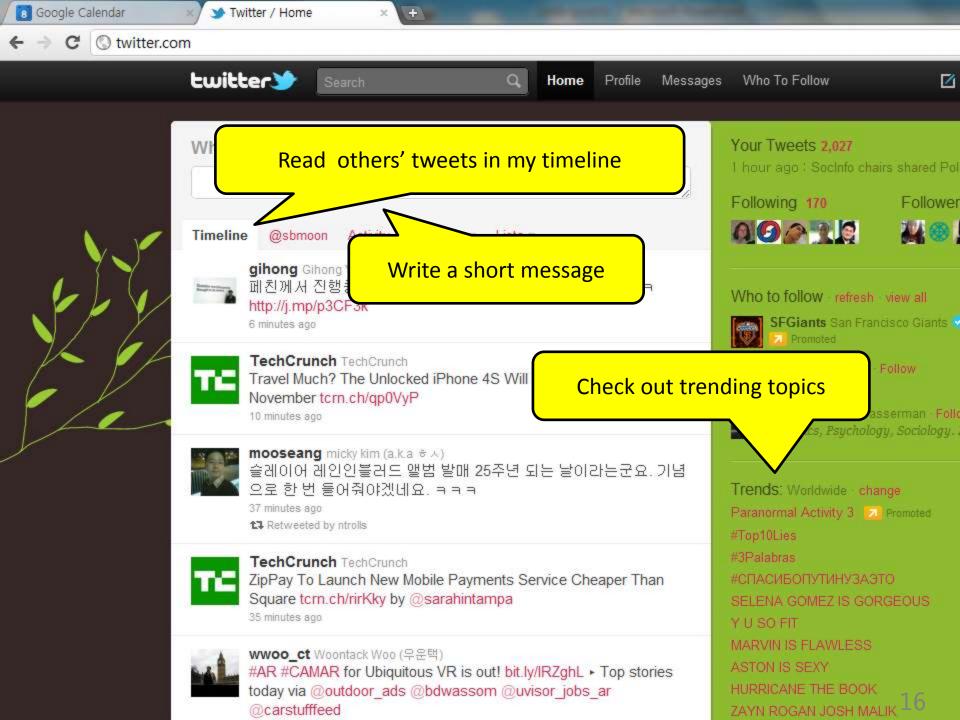
# Unfollow in Twitter

- Intentional action to break a relationship
- No need for an approval
- No notification
  - to the unfollowed



KAIS





## Four Types of Tweets

#### Tweet

Last Day of SocInfo!

Reply

@EePeng Thanks for having me!

Mention

I am attending SocInfo 2011 organized by @EePeng

Retweet

Wow lots of fun talks and great people! RT @sbmoon Last Day of SocInfo!





### Recent changes in Twitter API







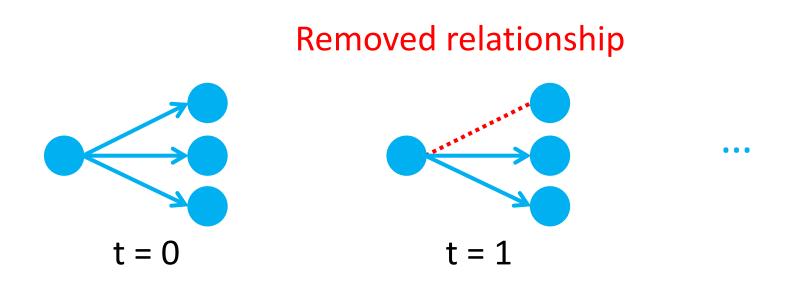
# Our Unfollow Study

- Macroscopic statistics
- User interviews
- Explanatory model





# No official records for unfollow



Compare two follow networks and detect removed relationships





### Twitter now too big to crawl

- In 2009 it took 3 months with 20+ hosts to crawl and get the entire 40 million user profiles
- Now in 2011 it has more than 400 million users
- How to sample?





# Identify a group with common culture

• Collect 1.2M Korean-speaking users identified by Korean in tweets, bio, location, or name







#### Data collection

- Collect daily snapshots of follow networks
  - G(I): June 25<sup>th</sup> to July 15<sup>th</sup>, 2010
  - G(II): August 2<sup>nd</sup> to August 31<sup>st</sup>, 2010

• Time resolution = a single day



# Korean follow network grew fast

- Increasing # of users
  - -G(I): 718,077 → 870,057
  - − G(II): 956,261  $\rightarrow$  1,203,196

+7,599/day +8,515/day

- Increasing (high) reciprocity
  - $G(I): 56 \rightarrow 58\%$
  - $-G(II): 61 \rightarrow 62\%$
- Increasing avg. # of followees  $-59.7 \rightarrow 75.7$





#### **Macroscopic Statistics**





# People unfollow frequently

- 43% of active users unfollow at least once during 51 days
- Average number of unfollows per person
  15.4 in G(I)
  - 16.1 in G(II)

#### Link removal frequently occurs in a 'growing' network

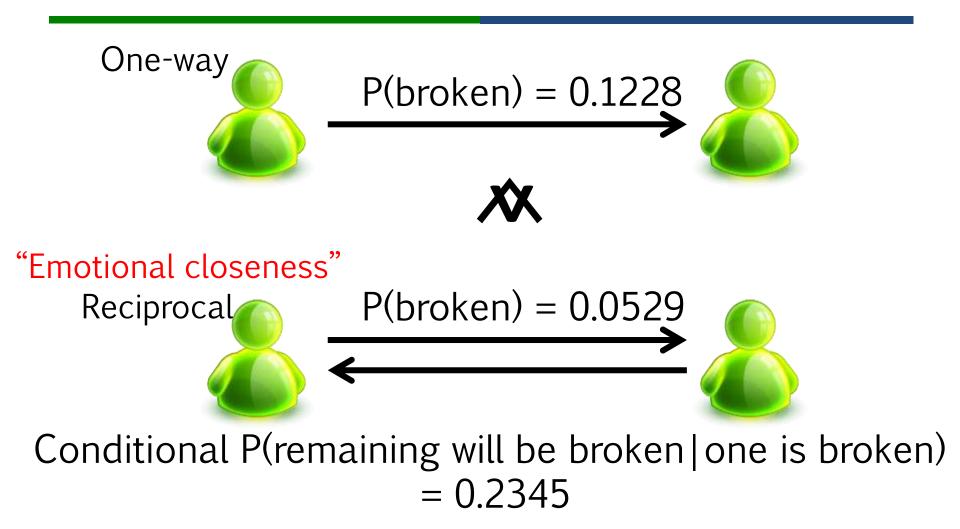


#### Factors that correlate with unfollow

- Reciprocity of relationships
- Duration of a relationship
- Followee's informativeness
- Overlap of relationships

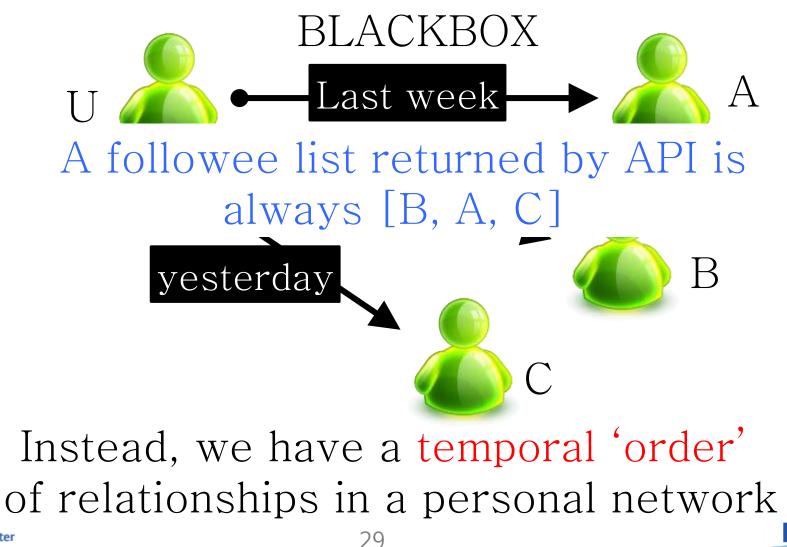


# One-way relationships are fragile



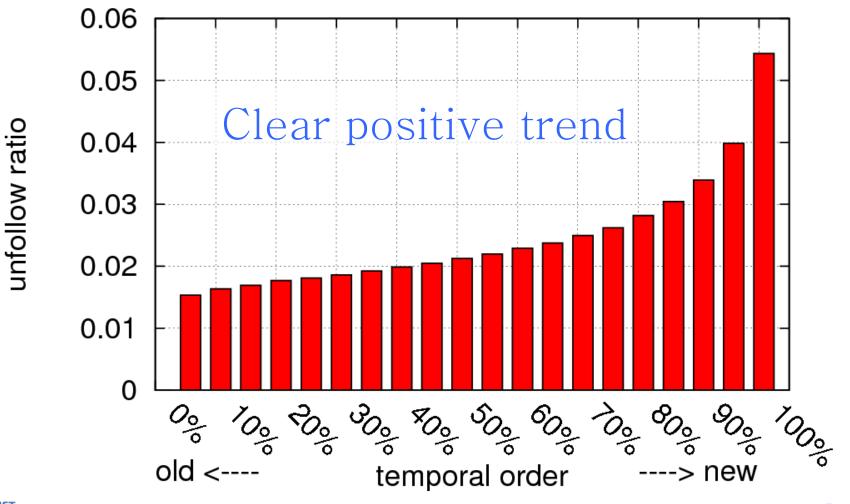


## No knowledge when following started





#### Newer relationships are more fragile



30

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### Informativeness of users



# Next year's rebuttals will be limited to 140 characters, counting spaces.

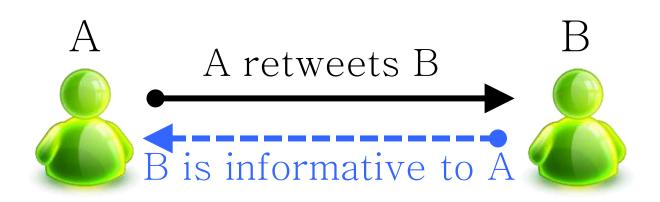
19 Nov via TweetDeck 🏠 Favorite 📭 Retweet 👆 Reply

Retweeted by JochenHuber and 15 others







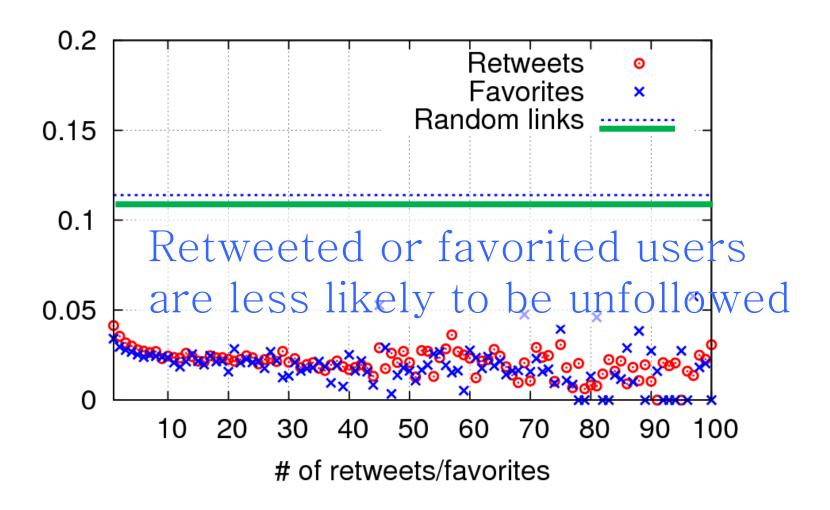




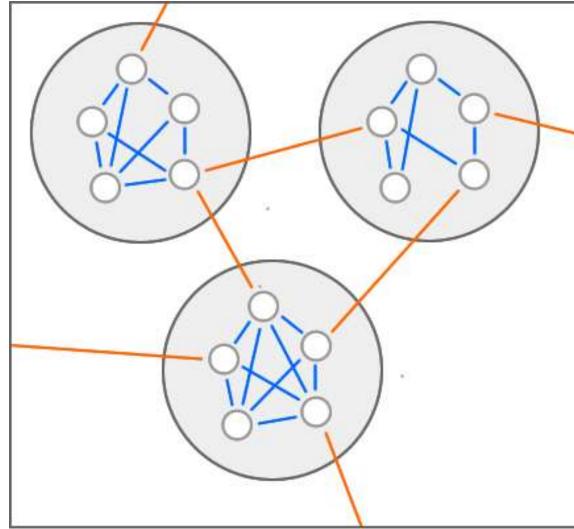


#### Non-informative relationships are fragile

unfollow ratio



#### Strong ties & weak ties



#### Group/Network

Group members, because of their frequent interaction, tend to think alike over time. This reduces the diversity of ideas, and in worst-case scenarios leads to "groupthink"

#### Weak Ties

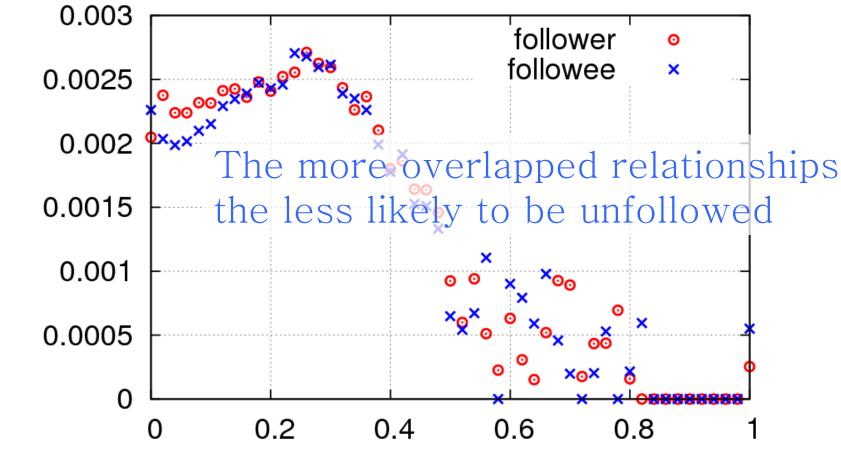
Weak ties are relationships between members of different groups. They are utilized infrequently and therefore don't need a lot of management to stay healthy. They lead to a diversity of ideas, as they tie together disparate modes of thought.

#### **Strong Ties**

Strong ties are relationships between people who work, live, or play together. They are utilized frequently and need a lot of management to stay healthy. Over time, people with strong ties tend to think alike, as they share their ideas all the time.

KAIST <u>http://bokardo.com/archives/weak-ties-and-diversity-in-social-networks</u> 34

### Weak ties are fragile



overlapping ratio of one's follower/followee (binsize 0.02)



unfollow ratio



### No interaction ≠ breakup

85.6% of relationships do not involve any single reply, mention, or retweet
96.3% involve 3 or fewer

 People just subscribe to others' tweets passively

#### Why our study of `unfollow' is important for the study of breakup



### **User Interviews**





# 22 online & face-to-face interviews

- Recruited by word-of-mouth
- Semi-structured
- Logging & camera recording



	Mean	Median	Min	Max	Std. dev.	Distribution
Age	27.3	27	22	36	3.7	ու ումենությո
Favorites	80.7	1	0	851	199.0	I
Followers	846.7	164.5	5	8,772	2,053.9	I
Followings	600.4	144.5	5	7,103	1,562.7	I
Tweets	3,325.8	583.5	5	30,639	7,220.5	<b>b</b>
Registered days	449.2	471	14	766	179.1	11.11



# Top reasons in unfollow

- 1. Burst tweets
- 2. Tweets about uninteresting topics
- 3. Tweets about mundane details of daily life
  - Automatically generated tweets (e.g., 4sq)
- 4. Tweets about political issues





## **Explanatory Model**





### Fragile Online Relationship: A First Look at Unfollow Dynamics in Twitter

### Haewoon Kwak, Hyunwoo Chun, and Sue Moon

Department of Computer Science, KAIST 291 Daehak-ro, Yuseong-gu, Daejeon, Korea (haewoon, hyunwoo)@an kaist.ac.kr, shmoon@kaist.edu

### ABSTRACT

We analyze the dynamics of the behavior known as 'unfollow' in Twitter. We collected daily snapshots of the online relationships of 1.2 million Korean-speaking users for 51 days as well as all of their tweets. We found that Twister users frequently unfollow. We then discover the major factors, including the reciprocity of the relationships, the duration of a relationship, the followees' informativeness, and the overlap of the relationships, which affect the decision to unfollow. We conduct interview with 22 Korean respondents to supplement the qui followed those who left many tweets within a short time

online relationships, often referred to as a "fittend". By contrast, research on the topic of online relationship dissolution has not been extensively conducted due to the lack of data; an online friend relationship remains rigid regardless of the actual relationship [28]. Researchers thus use proxies to represent the state of relationship dissolution. For example, a study of relationship dissolution in email networks assumes that the disannearance of online activities (the exchange of emails) reflects this type of dissolution [16]. However, a disappearance of communication cannot be directly translated

### The Impact of Network Structure on Breaking Ties in **Online Social Networks: Unfollowing on Twitter**

### Funda Kivran-Swaine, Priya Govindan and Mor Naaman Rutgers University funda@rutgers.edu, privagn@cs.rutgers.edu, mor@rutgers.edu

### ABSTRACT

We investigate the breaking of ties between individuals in the online social network of Twitter, a hugely popular social media service. Building on sociology concepts such as strength of ties, embeddedness, and status, we explore how network structure alone influences tie breaks - the common phenomena of an individual ceasing to "follow" another in Twitter's directed social network. We examine these relationships using a dataset of 245,586 Twitter "follow" edges, and the persistence of these edges after nine months. We show that structural properties of individuals and dvade at Time 1 have a confideant effect on the observed over time, are far from random, and depend on various factors that affect the relationship between the users. Most work in the Computer Science and HCI fields has focused on the dynamics and models of the creation [5,9,13]. In particular, researchers considered the structural aspects of the social network that predict formation of ties [5,10,15]. Here, we also focus on the social network structure, but examine breaking and persistence of existing ties, rather than tie creation.

The topic of breaking and persistence of ties is exceedingly important. Tie breaks impact the dynamics and activity in online services over time, and, as we show below, are The manufact the start the manufact the start the manufactor in the start the start the manufactor in the start the manufactor in the start the manufactor in the start the star ervices like Twitter. Moreover, the act of "deed, and their nonexistence could be detected.

### Author Keywords

### Author Keywords Lindulow, composer-metaled of Internactional propertie Structural and interactional propertie (1). This get is ACM Classification Kywords

H 5.0 Information Systems: Info sentation - General

General Terms Human Factors

### INTRODUCTION

Relationship formation and dissolution are two basic processes of relationship change and evolution in personal networks. Studies of relationship formation and dissolution mostly rely on surveys and interviews, both of which require. considerable effort in terms of time and labour. Online social networks (OSN) ald researchers in at least two ways, such as (1) they contain a huge archive of human behavior related to online relationships, and (ii) they allow easy access. Studies of online relationship formation are straightforward, as most OSNs offer simple means of establishing

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CMI 2011, May 7-12, 2011, Vancouver, BC. Canada. Copyright 2011 ACM 978-1-480540267-8011/05. \$10.00. into consideration

a timeline. A user can easily stop following (aufoliow) and needs no confirmation from the followee to do so. Unfollow, thus, is not a proxy but a verifiable action of breaking an online relationship. In the rest of the paper, we use unfollow both as a noun and a verb.

In this work, we analyze the dynamics of the unfollow behavior to understand online relationship dissolution. The two research questions explored here ase: (1) what are the characteristics of the unfollow behavior? and (ii) why do people unfollow others? To address the first research question, we collected daily snapshots of the follow relationships of 1.2 million Korean-speaking users over the course of 51 days as well as their tweets. By comparing the daily snapshots, we confirm that unfollow is prevalent in Twitter. We have found that the reciprocity of the relationship, the duration of the relationships, the followees' informativeness, and the overlap of relationships are critical in the decision to un-

their contacts when they log in). These contacts reflect various types of relationships, including friendship, kinship, common interests, attention, or information exchange. The aggregates of contact lists in each of these services result in immense, articulated online social networks. As these services shift the communication and information fabric of our society, the dynamics of the networks they support are important to understand and reason about in depth.

In particular, the articulated online social networks in these services change and evolve as individuals form new ties, or break existing ties to others. These structural changes,

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social ties often decay rather than "formally" break, making measurement of the phenomena harder. The initial findings from sociology research, however, suggest that the structural aspects of individuals' social networks are strong determinants for persistence and decay of ties [1,11]. We now have an opportunity to examine these theories in online social networks, and at a scale that was not available before. The dynamics of these online networks may help us, then, to reason about other types of networks and social ties, and discover patterns that may shed light on social phenomena in different contexts.

We use two snapshots of the Twitter social network to study the breaking of ties. Twitter is founded on an articulated online network, and allows users to read updates from others that they "follow". Thus, users create a directed social network that reflects attention and transfer of information [5,7,12,13]. We borrow from sociology theory to frame our investigation of tie breaks and network structures using the concepts of tie strength, embeddedness as well as power and status. Our large-scale analysis uses information from 245,586 dyadic relationships on Twitter at Time 1, and the persistence or break of edges in these dyads by Time 2. The analysis aims to answer the following research question: What structural social network properties of nodes and dyads predict the breaking of ties?





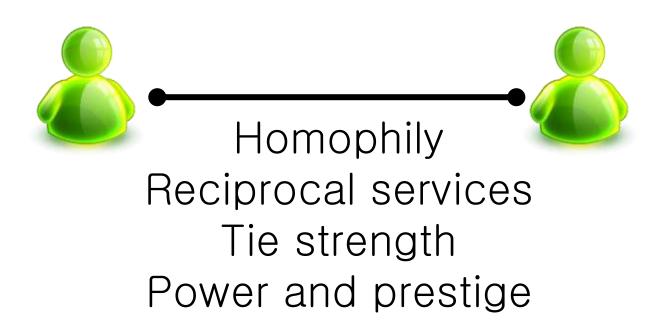
## Interactional properties describe dynamics

- Has a followee sent a reply to a follower?
- Has a user mentioned a follower in any of one's tweet?
- Do a user and a follower share common topics of interest?





### Sociological concepts in persisting ties



+ Twitter-specific feature: Informativeness





# Multiple logistic regression model

- Binary dependent variable
  - Whether a relationship at  $t_0$  will be broken (1) or persisted (0) at  $t_1$
- 78 Independent variables
  - From structural properties
    - # of followers, # of followees, # of common followers, ...
  - From interactional properties
    - # of replies, # of retweets, # of communication partners, ...





# **Filtered variables**

- Removing multicollinearity
  - # of common followers & followees & neighbors
  - # of follower & those who reply to ego
  - # of those who exchange replies & all replies
  - # of received replies & mentions

- We filter out 36 variables and 42 remained
- With stepwise regression, we further winnowed down to 39 variables





	Dependent variables	Coefficient	Odds ratio
	Structure	properties	
u	Followees	-4.13e-04 (1.50e-06)	1.000***
u	Followers	5.59e-05 (8.81e-07)	1.000***
24	Followees/Followers	-1.71e-03 (1.98e-05)	0.998***
u	Follow-back ratio	-0.21 (4.02e-03)	0.810***
f	followees	2.67e-05 (2.53e-07)	1.000***
f	followers	-2.68e-06 (8.01e-08)	1.000***
f	Followees/Followers	1.68e-03 (1.28e-04)	1.002***
f	Follow-back ratio	0.82 (4.68e-3)	2.282***
$i \rightarrow f$	Order of follow	-1.57e-08 (8.41e-11)	1.000***
$\iota \rightarrow f$	<sup>N</sup> Order of follow	2.58e-08 (6.71e-10)	1.000***
$\iota \leftrightarrow f$	Prestige	-1.22e-06 (2.37e-07)	1.000***
$\iota \leftrightarrow f$	Reciprocity of follows	-0.80 (2.50e-03)	0.451***
$\iota \leftrightarrow f$	Common followees	-7.40e-05 (9.12e-06)	1.000***
$\iota \leftrightarrow f$	<sup>N</sup> Common followees	-2.89 (2.78e-02)	0.056***
	Activity	properties	
u	Tweets	2.76e-04 (9.40e-07)	1.000***
24	URL tweets	-1.40e-04 (7.24e-06)	1.000***
u	Auto-generated tweets	5.26e-04 (7.32e-05)	1.001**
u	Popularity	-1.44e-04 (1.07e-06)	1.000***
ſ	Tweets	-4.27e-05 (1.11e-06)	1.000***
ſ	URL tweets	-4.32e-05 (8.74e-06)	1.000***
f	Auto-generated tweets	-2.86e-04 (1.14e-04)	1.000*
ſ	Popularity	-6.04e-06 (1.02e-07)	1.000***
$i \xrightarrow{f} f$	Replies	2.85e-03 (1.32e-04)	1.003***
$i \rightarrow f$	N Replies	-0.67 (2.22e-02)	0.516***
$i \rightarrow f$	Mentions	-9.23e-04 (3.70e-04)	0.999*
$i \rightarrow f$	Retweets	-1.21e-02 (9.65e-04)	0.988***
$i \rightarrow f$	Favorites	-5.06e-02 (1.95e-03)	0.951***
$i \rightarrow f$	<sup>N</sup> Favorites	-3.95e-04 (2.53e-02)	1.000***
$i \leftarrow f$	Replies	-1.58e-02 (3.91e-04)	0.984***
$i \leftarrow f$	N Replies	-1.49 (3.52e-02)	0.225***
	<sup>N</sup> Mentions		
$i \leftarrow f$		-0.66 (5.19e-02)	0.516***
$i \leftarrow f$	Retweets	-5.17e-02 (2.37e-03)	0.950***
$\iota \leftarrow f$	<sup>N</sup> Retweets	-0.76 (4.92e-02)	0.469***
$i \leftarrow f$	Favorites	-6.11e-02 (5.01e-03)	0.941***
$i \leftarrow f$	<sup>N</sup> Favorite	-0.35 (4.88e-02)	0.706***
$\iota \leftrightarrow f$	Common hashtags	-0.12 (1.29e-03)	0.883***
$\iota \leftrightarrow f$	<sup>N</sup> Common hashtags	-2.2 (4.20e-02)	0.109***
$\iota \leftrightarrow f$	Days since first comm.	-1.95e-03 (4.8e-05)	0.998***
$\iota \leftrightarrow f$	Days since last comm.	-1.08e-04 (3.38e-06)	1.000***
		l variables	
$\iota \rightarrow f$	<sup>N</sup> Mentions	3.49e-02 (3.46e-06)	1.000
$\iota \rightarrow f$	<sup>N</sup> Retweets	6.25e-03 (2.28e-02)	1.000
$\iota \leftarrow f$	Mentions	1.09 (5.68e-04)	1.001

\* Odds ratio is rounded to thousandths.

# Interpretation of models

**Emotional closeness**times likely to unfollow **Homophily** eff. of hashtags decreases 0.2, 2.75x **Tié strength**followees decrease 0.2, 2.12x

### **Receiving actions rather than giving ones** more important variable than the opposite

- People appreciate receiving feedback
- Consistent with the study of social capital at Facebook



# Summary

- Unfollow is a clear sign of breaking a relationship
  - No interaction is NOT a good indicator of breaking a relationship due to user's passivity in Twitter
- Quantitative and qualitative studies of unfollow
  - Reveal its characteristics and motivations
  - Discover important structural and interactional properties





# **Application: Practical uses**

### Burst tweets, automatically generated tweets, ...

📶 olleh 奈	오전 <b>1:36</b>			olleh	<ul><li>오후 9:4</li></ul>	I3 🕞 🛙
	haewoon				haewo	on
that ha	alf ended?				that half ended?	
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	d Jered Weaver his ason tonight: es.pn				Whoa the king	
	ky_F0ntane Don't w	13 hours orry, we			#scilifelab	
espn ESFIT We ha	tell on you. ve a winner RT @ scored!	14 hours lindacohn:			owlcity DARTH VADER: " you some toast." bit on the dark sid	
	<b>Ihuss</b> the king of Swec d into our office spa		•	S.	Top Tweet RealWizKhallifa Lol has gone from	1 o n meaning, "laugh
9 @	) 📩 Q	•••	40		@	Q •••
Computer			49			_

# Application: Theoretical models

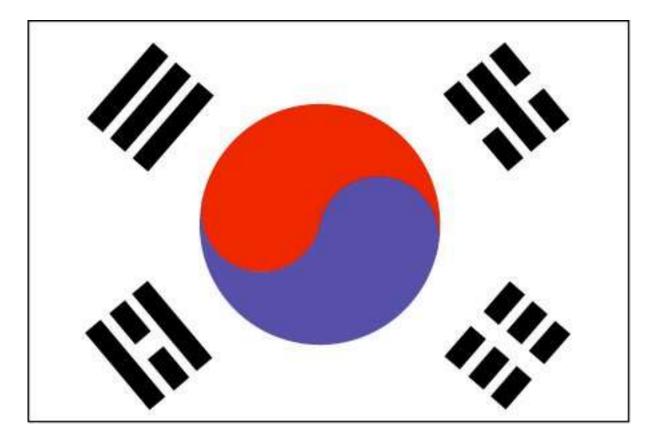
Existing models for growing networks deal with link additions only

Our studies reveal frequent link removals in a growing network





## **One population - Korean**







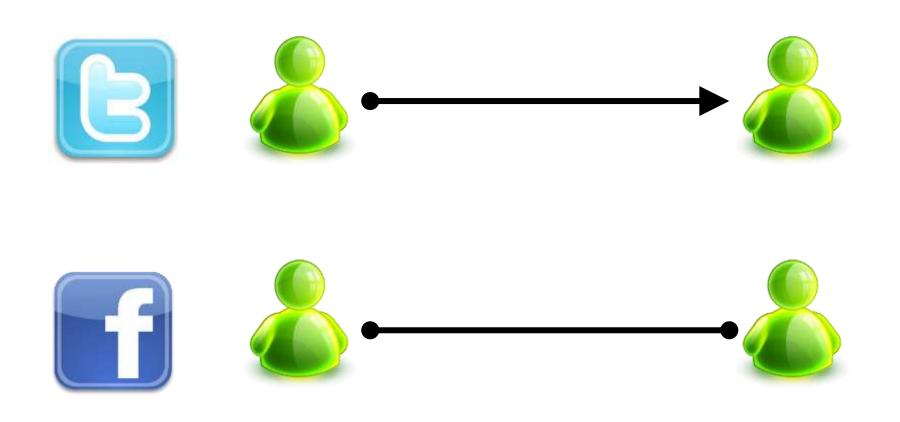
## How can we generalize our results?







### Follow ≠ Friend







# Thanks! Questions?



