

# Emotion, Tie Persistence, and Network Structure on Twitter

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social media information lab?

social media research:

1. what are people doing  
(and why)?

social media research:

2. understanding social  
systems at scale

social media research:

3. creating new experiences

S  
O  
C  
I  
A  
L

media

awareness streams

networks

today's big story

generate a better understanding of the  
social dynamics

validate theories from social sciences in  
these new and important settings



today's more specific story

Twitter and networks:

**Part 1.** social sharing of emotion and networks on Twitter

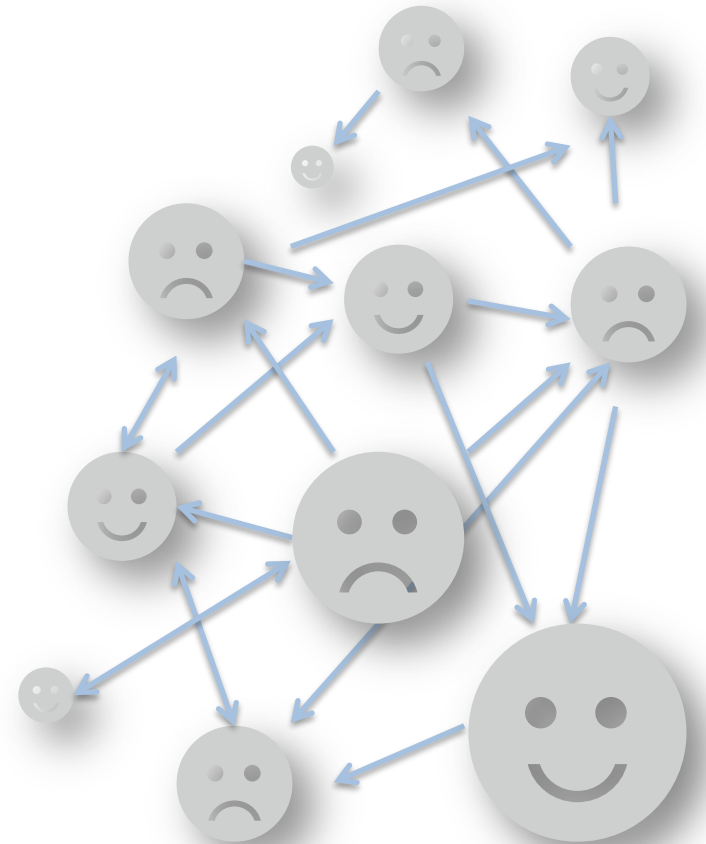
**Part 2.** unfollowing on Twitter

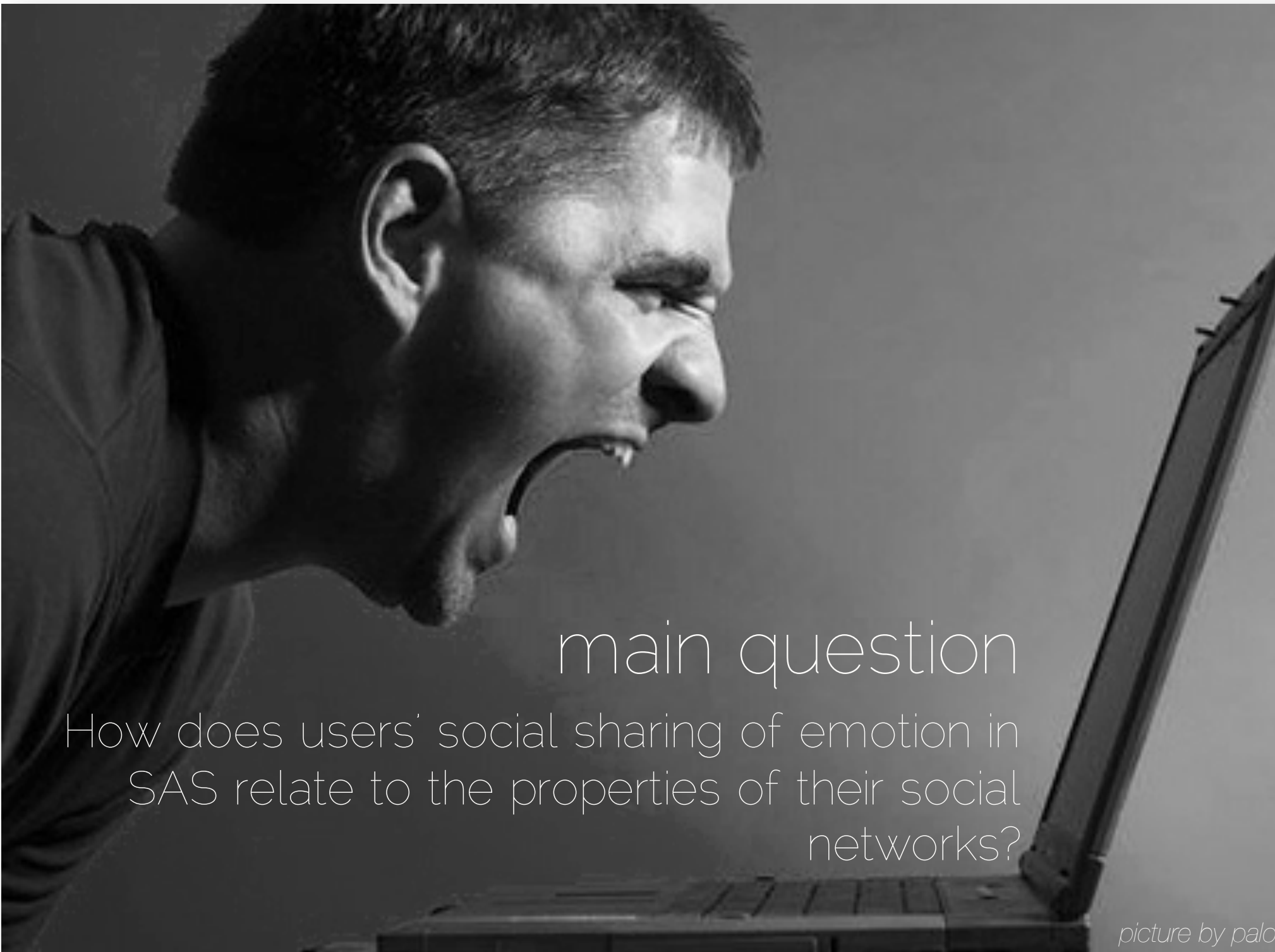


# study 1

## emotion & social networks

Kivran-Swaine & Naaman. Network Properties and Social Sharing of Emotions in Social Awareness Streams. (CSCW 2011).





## main question

How does users' social sharing of emotion in SAS relate to the properties of their social networks?

# research questions

RQ1

What is the association between people's tendency to express emotion (joy, sadness, other) in their posts (updates or interactions) and their number of followers?

# research questions

RQ2

What is the association between people's tendency to express emotion (joy, sadness, other) in their posts (updates or interactions) and their network characteristics like density and reciprocity rate?



# David A. Shamma

**@ayman** iPhone: 47.563553,-122.363365

research scientist. media artist. instructions: place in direct sunlight, water daily

<http://shamurai.com>

**Following**

Message

Timeline

Favorites

Following

Followers

Lists



**AymanM** Ayman Mohyeldin by ayman

The Birth of a New Egypt... <http://fb.me/HN1Lit3Y>

6 hours ago



**ayman** David A. Shamma

[@landay](#) 1password - slick, nice integration, and uses dropbox to autosync.

13 Feb



**RawyaRageh** Rawya Rageh by ayman

Dawn prayers in [#Tahrir](#) sq. I came so close to crying on air. Good morning from a new [#Egypt](#)

11 Feb



**ayman** David A. Shamma

Obama's gonna speak about [#egypt](#) [#jan25](#) any minute now [whitehouse.gov/live/president...](http://whitehouse.gov/live/president...) [#fb](#)

11 Feb



## About @ayman

**2,124**

Tweets

**268**

Following

**1,211**

Followers

**53**

Listed

## Connections

Also followed by [@chkofler](#), [@uxrick](#), [@dwmcphd](#), and more.



You both follow [@dwmcphd](#), [@lyndonkennedy](#), [@cscw2011](#)



## Following **268**



## Similar to @ayman · [view all](#)



**dmrussell** · [Follow](#)

dmrussell



**chloester**

Chloe Fan



**bederson** · [Follow](#)

Ben Bederson

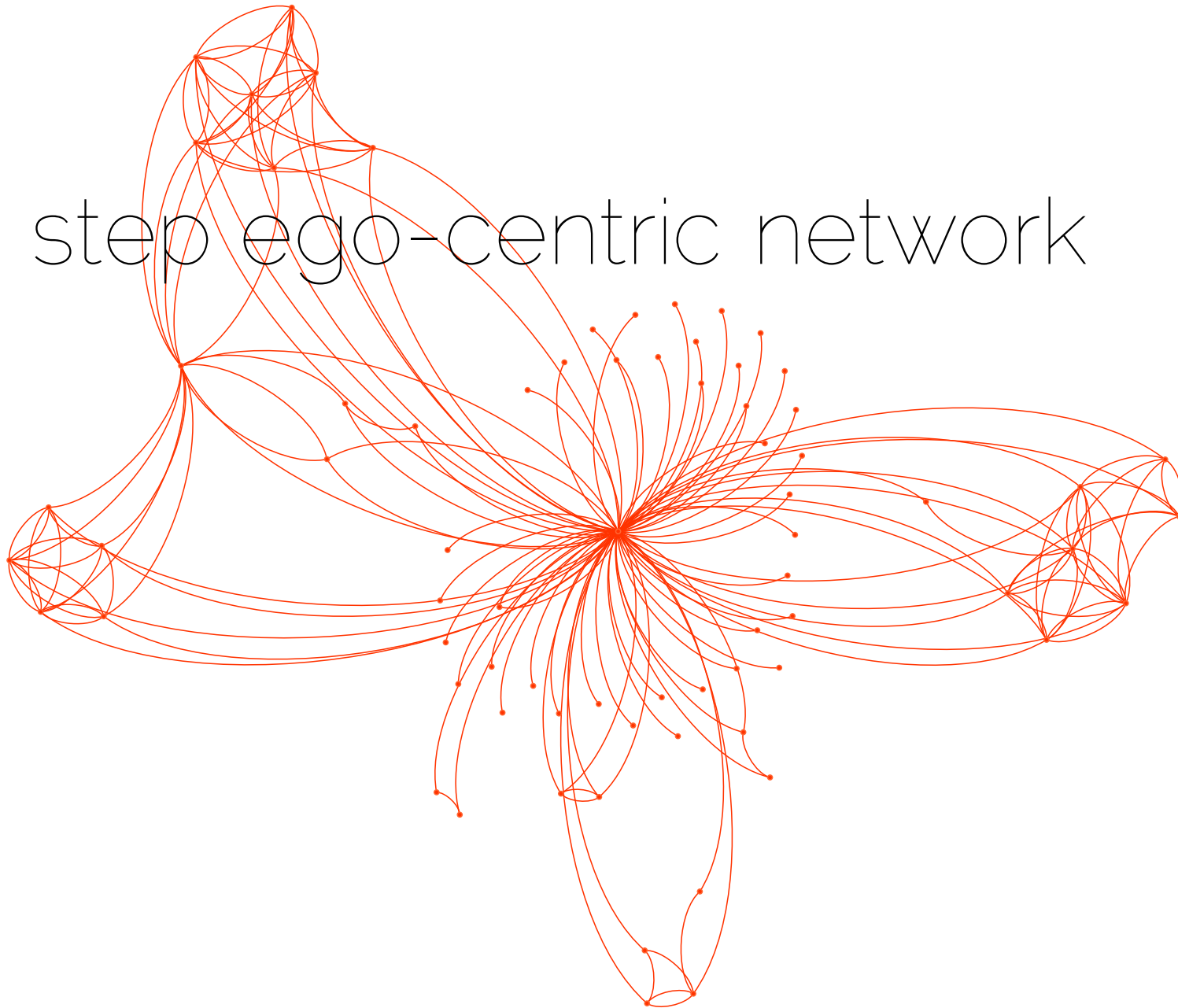


**drewww**

Drew

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1.5 step ego-centric network



# theory background

expression of emotion ↔ number of followers

( - ) people who mostly post about themselves have significantly lower number of followers\*

( + ) emotional broadcaster theory

\* Naaman, Boase, Lai (CSCW 2010)

# theory background

expression of emotion ↔ network density

expression of emotion ↔ reciprocity rate

( + ) intimacy

( - ) curbing



# data

content dataset from Naaman, Boase, Lai (2010)

social network dataset from Kwak et al. (2010)

105,599 messages from 628 users who:

- had no more than 5,000 followers or followees

- posted at least one Twitter update in July 2009 in English

- still had public profile in April 2010

# pilot study



## joy

on average 23% of a user's updates

*"First time at the last remaining fairgrounds reverts Yay!"  
awesome. Sophia had a blast. Lucy said, "oooooh,"  
over and over. Good times with my family.!"*



## sadness

on average 10% of a user's updates

*"RIP Kathy. Live life for today. You never know how  
long you have.!"*

# study details

automated analysis of the users' tweets based on LIWC

"expression of emotion" => "existence of emotive words"

# some gender differences



joy



sadness



other emotions

# analysis

independent variables:

joy (interactions-updates),

sadness (interactions-updates),

emo (interactions-updates)

3 linear regression models for dependent variables:

number of followers

network density

reciprocity rate

# results

... explaining number of followers ( $R^2 = .22$ )

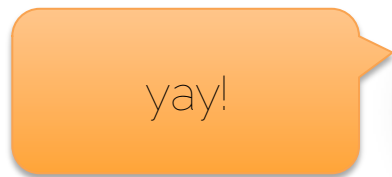
@follower ...  joy-interactions .35 \*\*

@follower ...  sadness-interactions .20 \*\*

\*\*  $p < .01$

# results

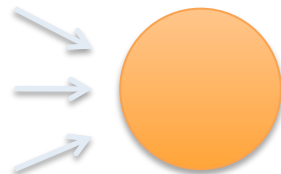
... explaining network density ( $R^2 = .33$ )



joy-updates  $-.10^{**}$



sadness-interactions  $-.18^{**}$



number of followers  $-.50^{**}$

$^{**} p < .01$

# limitations & future work

better emotion classifier

improve sampling, increase dataset

culture dependent

dyad-level analysis





today's more specific story

Twitter and networks:

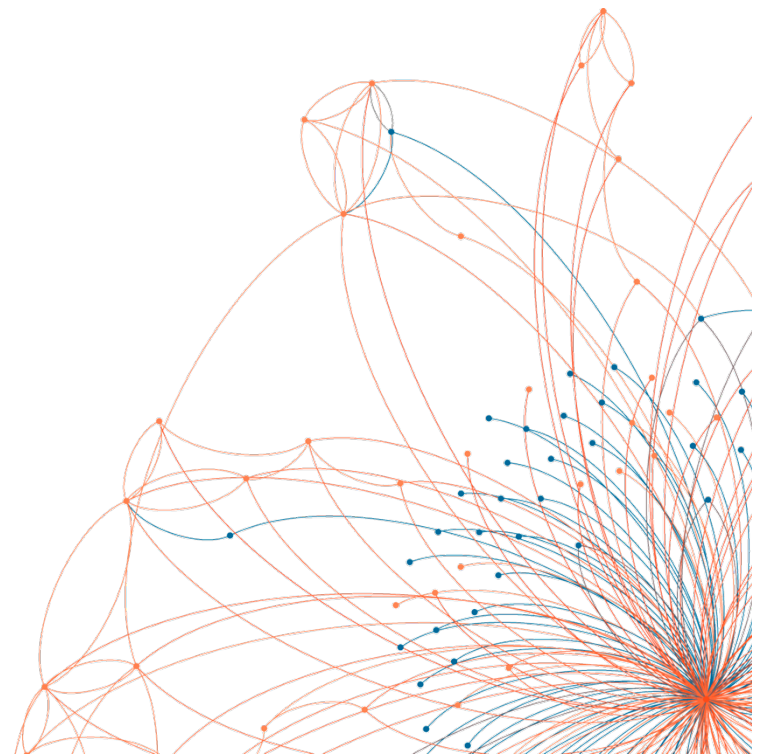
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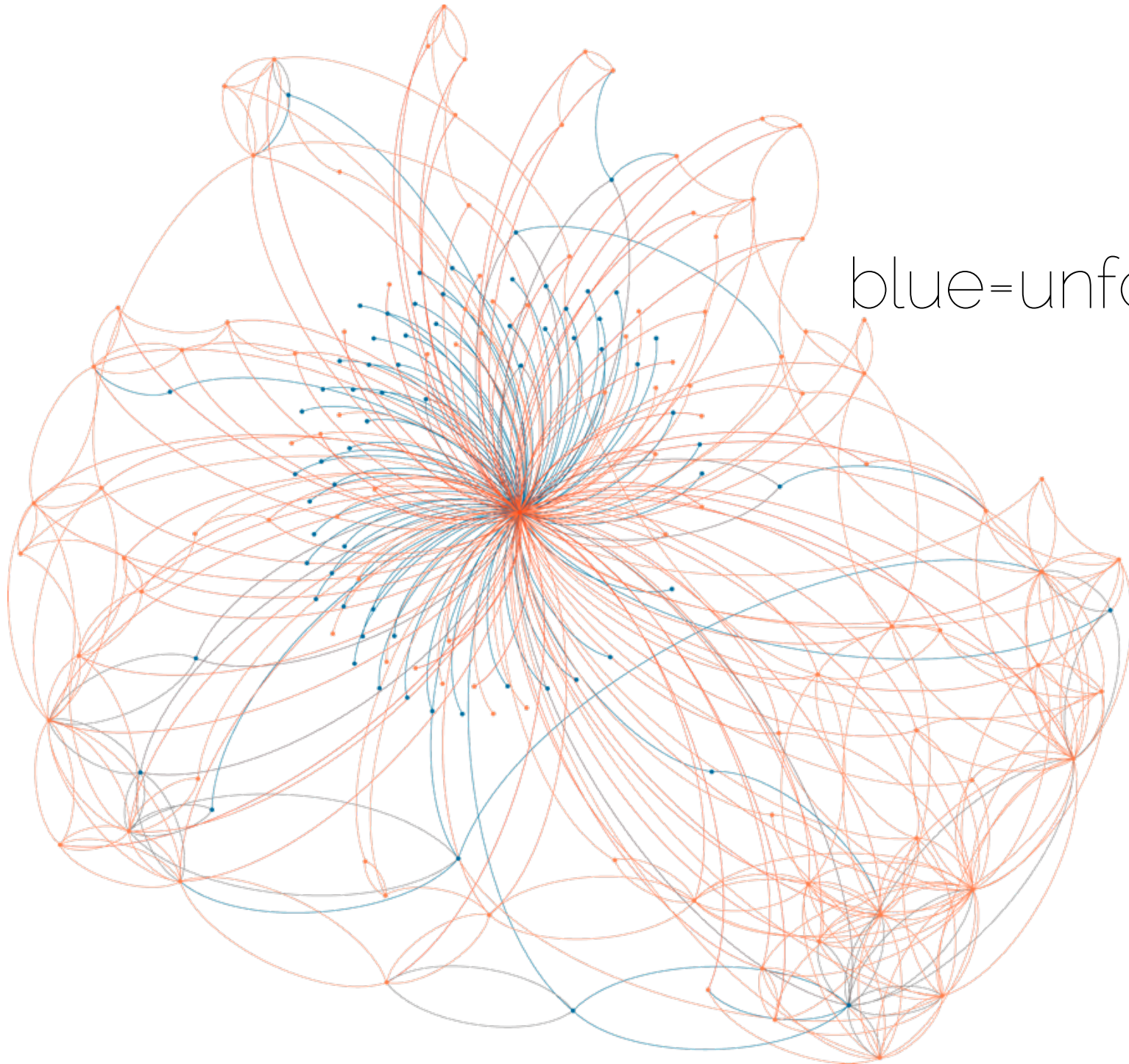
**Part 2.** unfollowing on Twitter

# study 2

## unfollowing on Twitter

Kivran-Swaine, Govindan & Naaman.  
The Impact of Network Structure on  
Breaking Ties in Online Social  
Networks: Unfollowing on Twitter.  
(CHI 2011).





blue=unfollow





main question:

what structural properties of the social network of nodes and dyads predict the breaking of ties (unfollows) on Twitter?

theory background

tie strength

embeddedness within networks

power & status

# data

content dataset from Naaman, Boase, Lai (2010)

social network dataset from Kwak et al. (2010)

Twitter API – connections still exist 9 months later?

715 seed nodes

245,586 “following” connections to seed nodes

30.6% dropped between 07/2009 & 04/2010

# analysis

\* independent variables (computed for our 245K dyads)

## *seed properties*

follower-count, follower-to-followee ratio, network density, reciprocity rate, follow-back rate

## *follower properties*

follower-count, follower-to-followee ratio

## *dyad properties*

reciprocity, common neighbors, common followers, common friends, right transitivity, left transitivity, mutual transitivity, prestige ratio

<disclaimer>

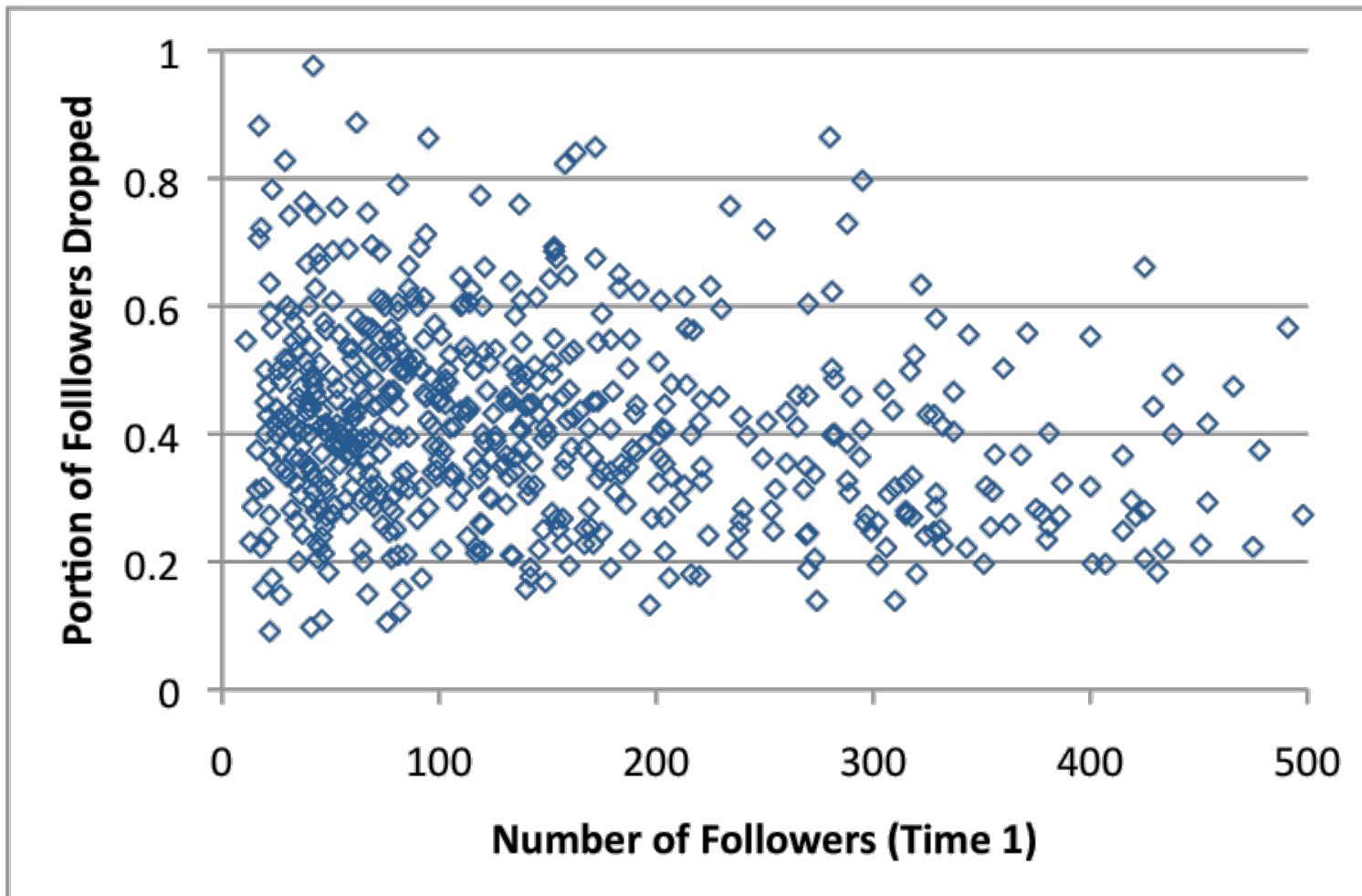
the following figures are NOT scientific evidence  
and are shown here for illustration purposes

no control for intra-seed effects; no inter-variable  
effects

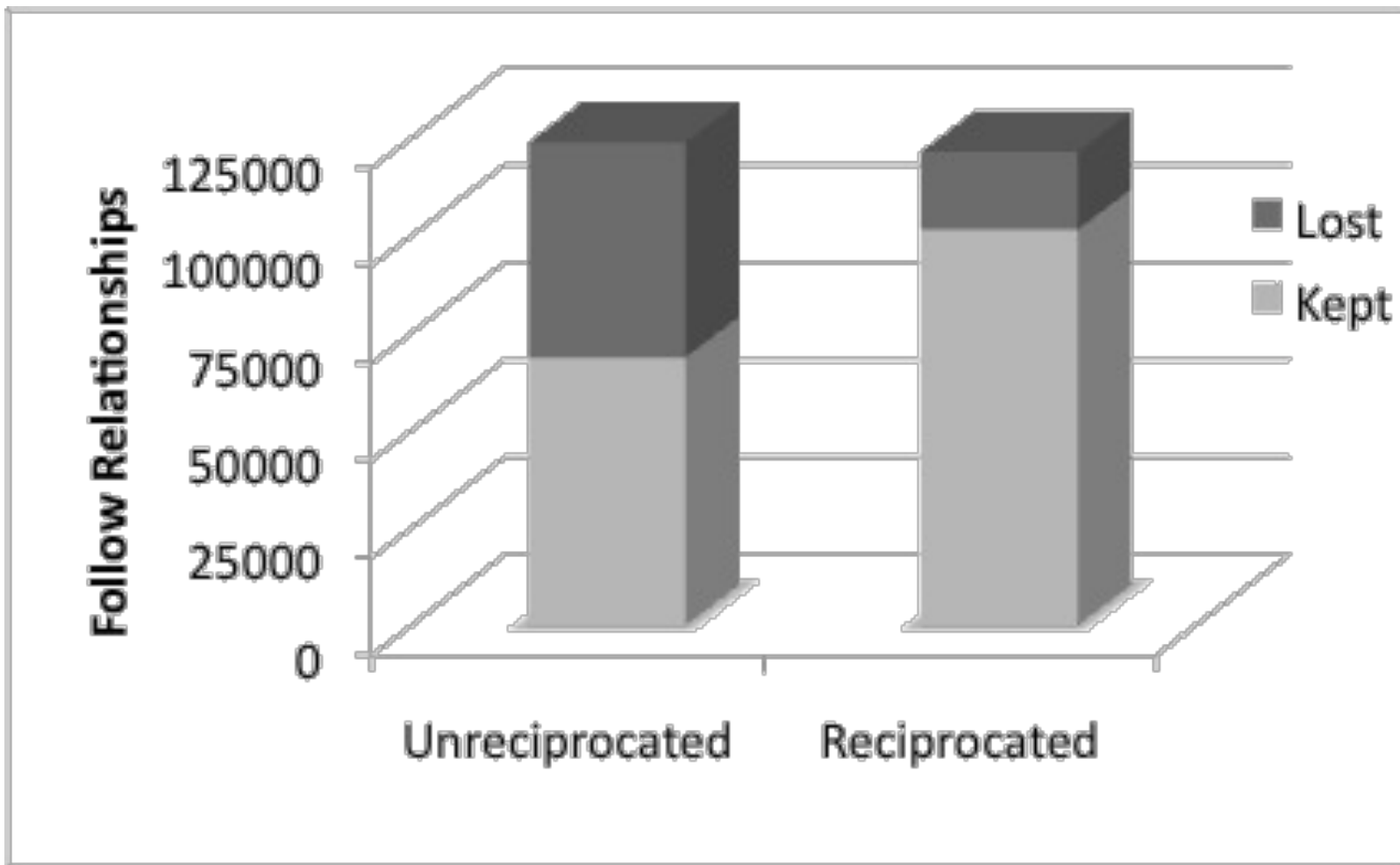
no R installation was harmed in the making of the  
following figures



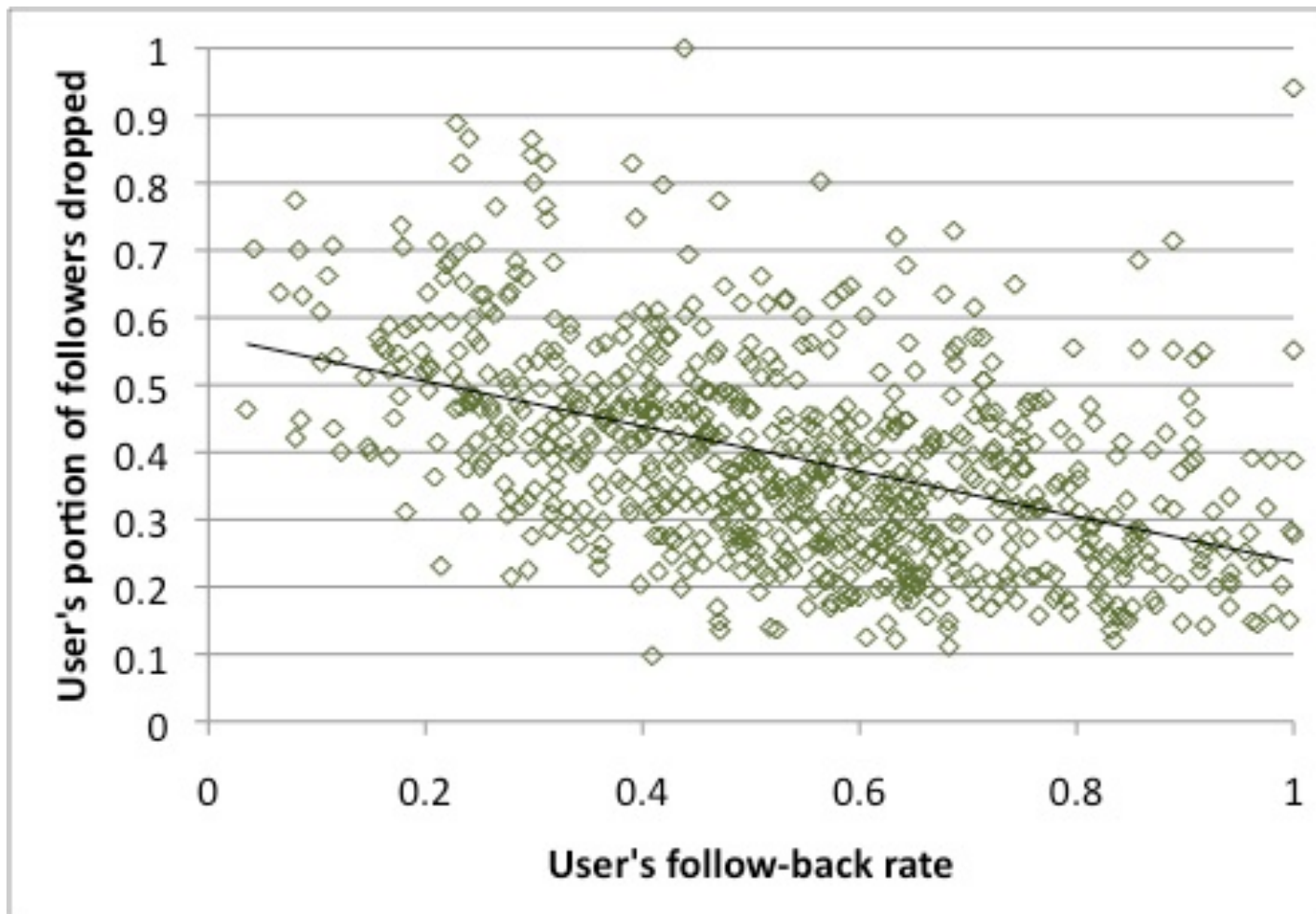
effect of number of followers (none):



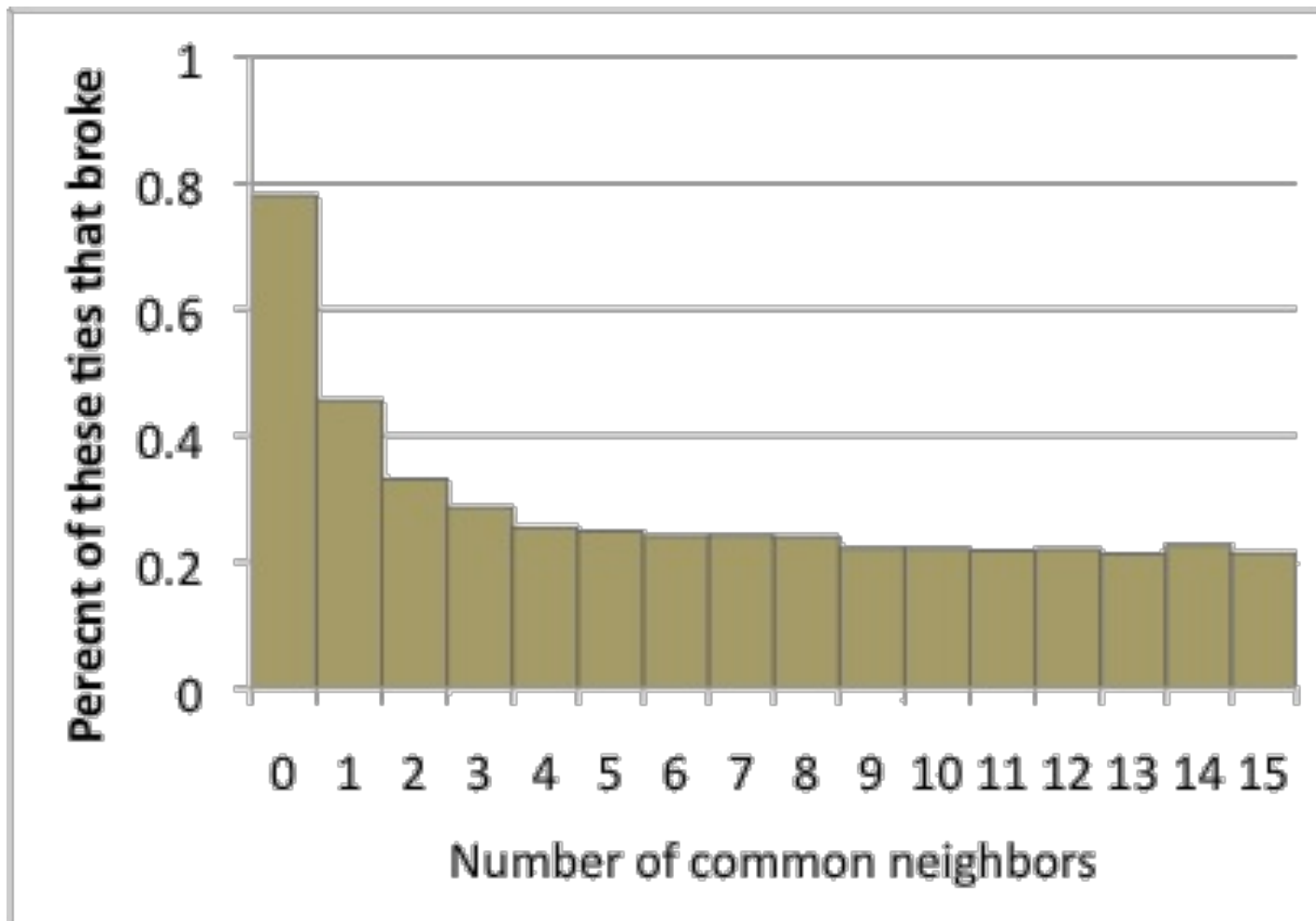
effect of reciprocity (large):



effect of follow-back rate



effect of common neighbors



</disclaimer>

back to scientific results (made R break sweat)  
sparing you the details, though

# in-depth analysis

the details you do not want to hear (now)

multi-level logistic regression (dyads/edges nested within seed nodes)

three models; full one includes seed, follower, and dyadic/edge variables

complete details: in the paper

# some results

effect of tie strength on breaking of ties

\*\*\* dyadic reciprocity (-)

\*\*\* network density (-)

\*\*\* highly statistically significant

# some results

effect of power & status on breaking of ties

\*\*\* prestige ratio (+)

\*\*\* follow-back rate (-)

\*\*\* f's follower-to followee ratio (-)

\*\*\* dyadic reciprocity (-)

\*\*\* highly statistically significant



some results

effect of embeddedness on breaking of ties

\*\*\* common neighbors (-)

\*\*\* highly statistically significant

# limitations & future work

only two snapshots: add more

additional (non-structural) variables (e.g.,  
frequency of posting!)

emotion and tie breaks

...and even broader

what can we learn from social dynamics on  
Twitter (and Facebook) about:

our relationships?

our language?

our society and culture?

our interests and activities?

for more details

<http://bit.ly/MorInfoSeminar>

thank you

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mor@rutgers.edu

@informor

<http://bit.ly/MorInfoSeminar>

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