

The topology of a discussion: The #occupy case

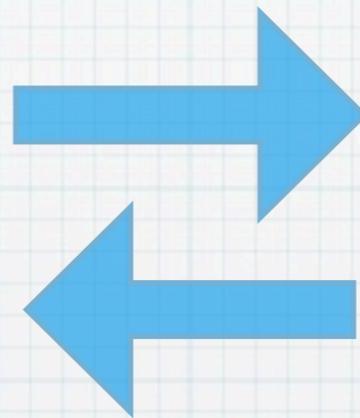
J. Bindi, F. Gargiulo, A. Apolloni



The general framework

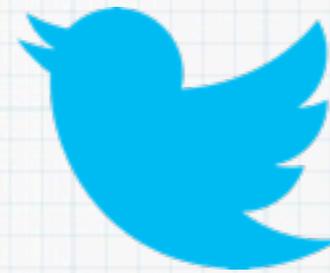
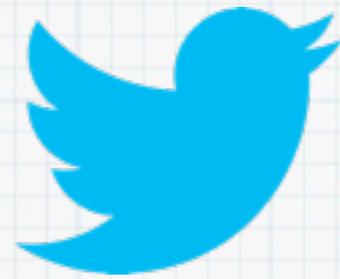
HUMAN COMMUNICATION PATTERNS:

- * Opinion dynamics
- * Temporal activity
- * Interaction rules



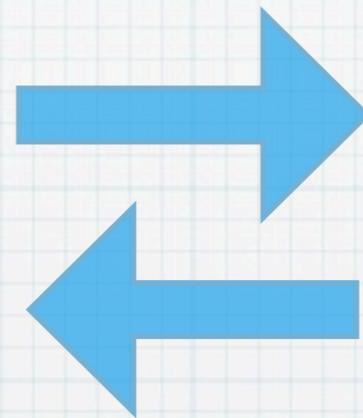
SEMANTIC CONTENT OF A DISCUSSION:

- * Topic of discussion
- * Interest peaks



HUMAN COMMUNICATION PATTERNS:

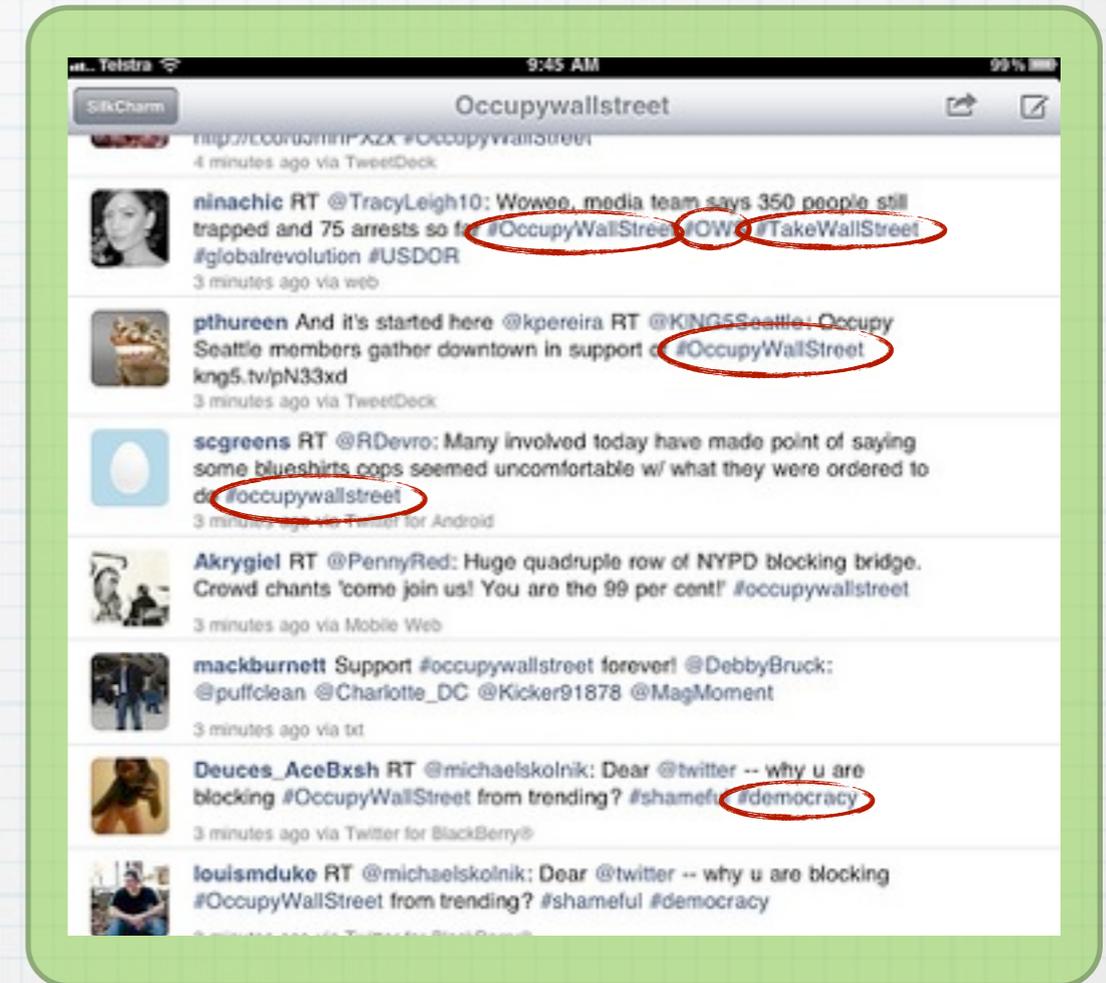
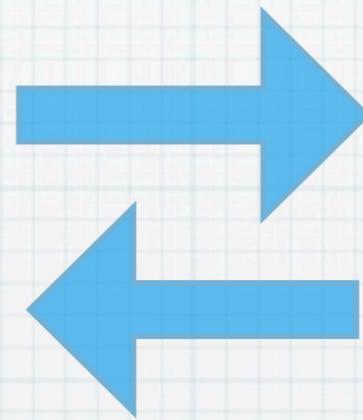
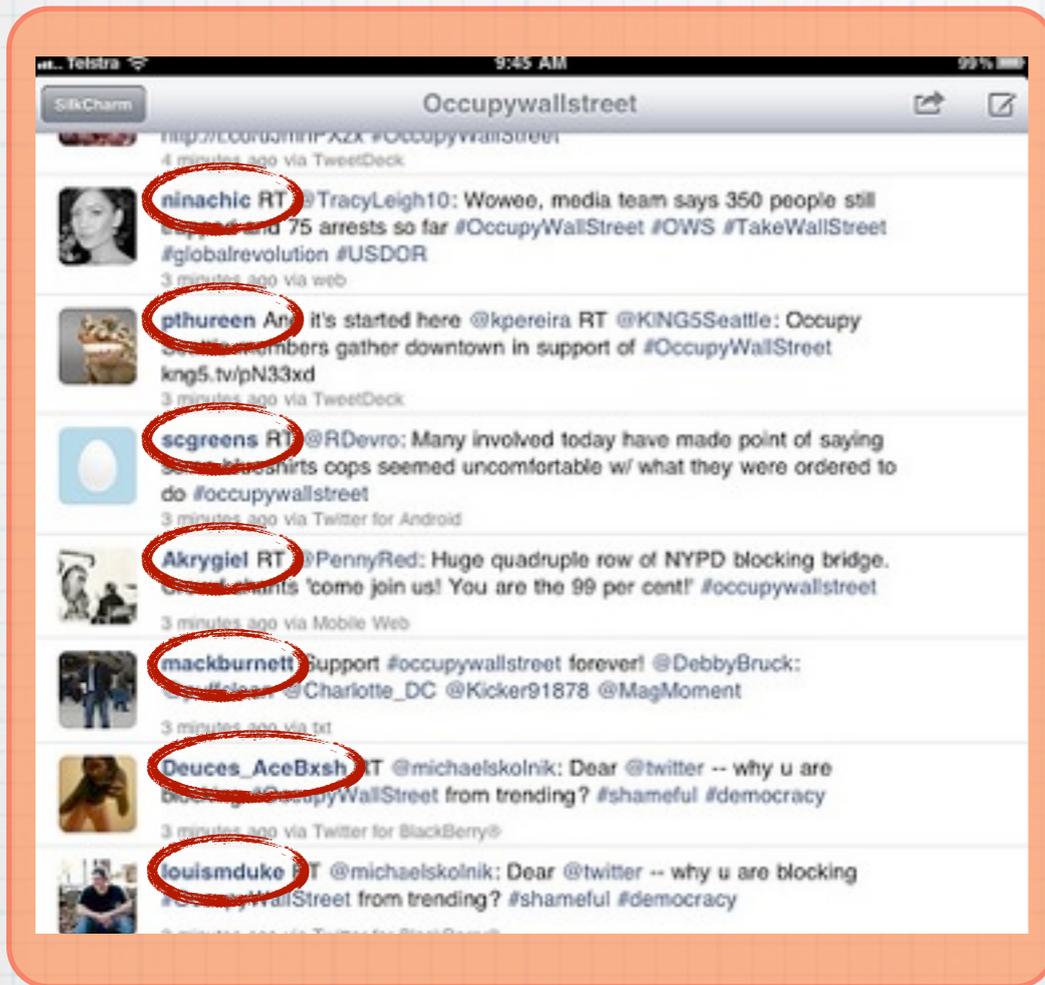
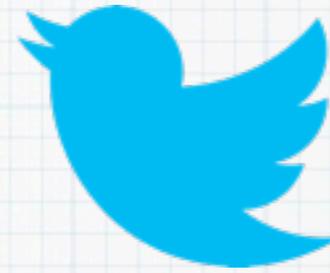
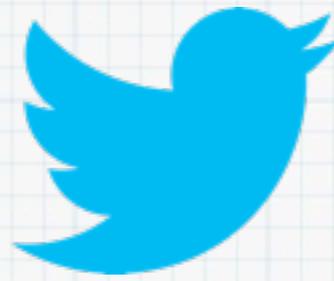
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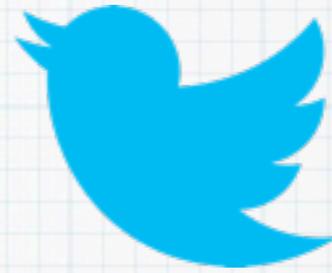
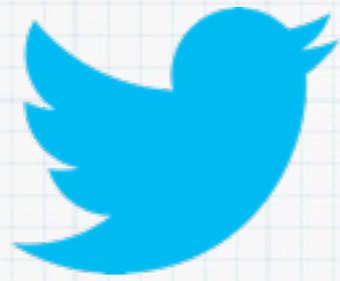
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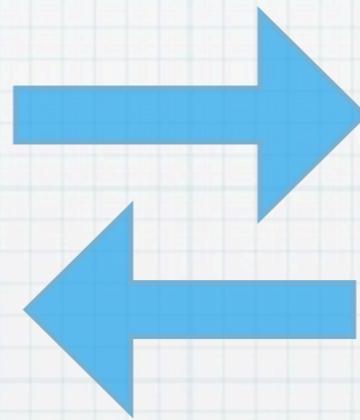
The Twitter platform



The Twitter platform



@USERS



#HASHTAGS

The Twitter platform

Social movements 2.0



The arab
spring

18-12-2010



15-M
Indignados
(Spain)

15-05-2011



Occupy
Wall Street

17-09-2011

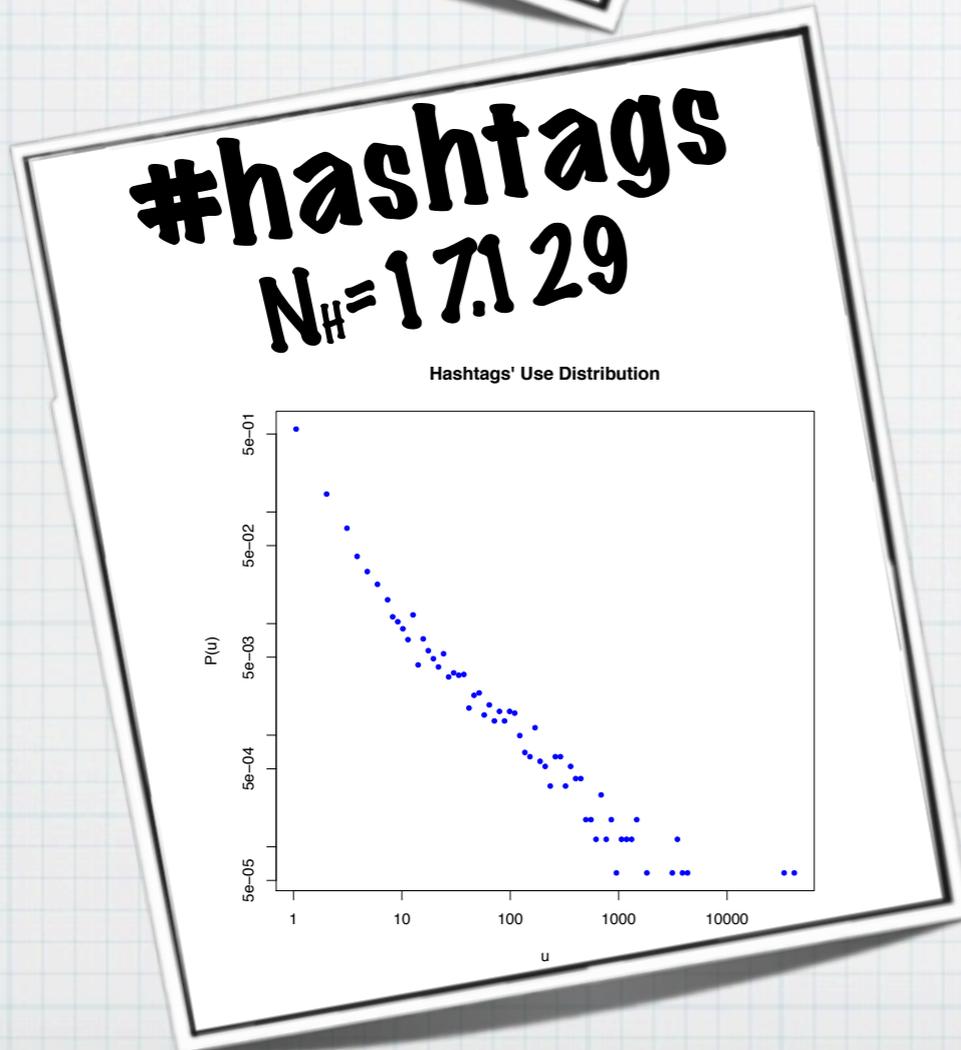
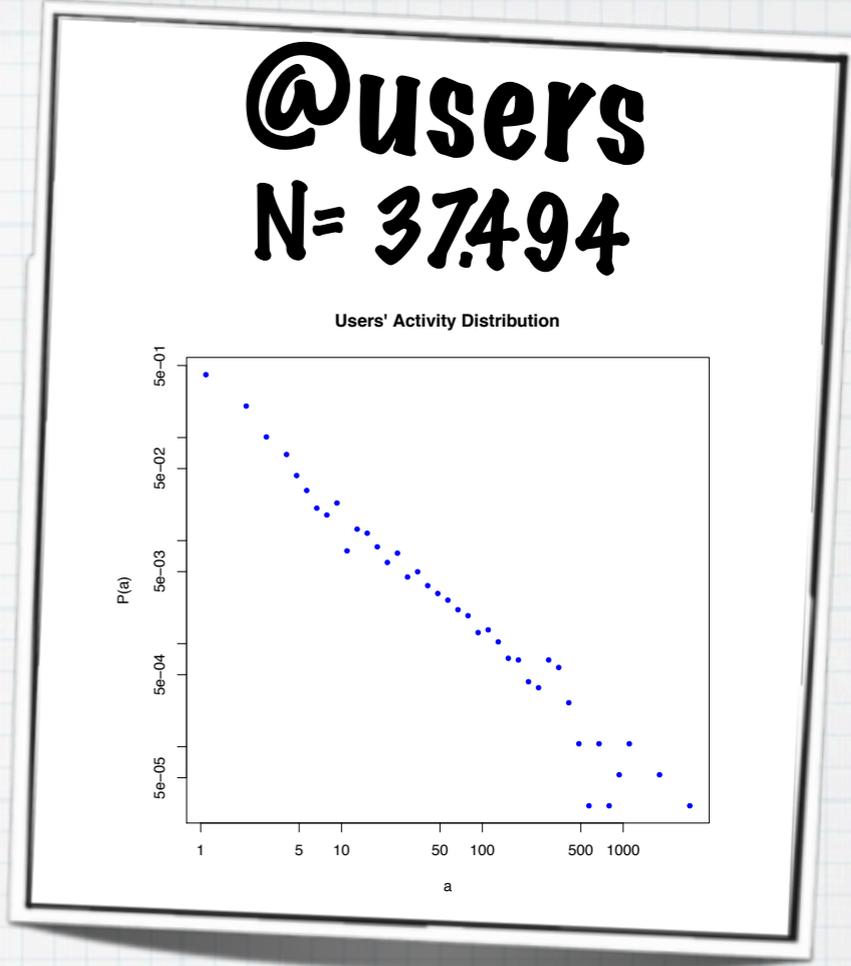
...

- * Spread all around the world
- * Individual participation (no political organizations)
- * Virtual squares on social networks (coordination and discussion)

The dataset

**Number of
Tweets
180.888**

**Dates:
26-10-2011
18-04-2012**
175 days

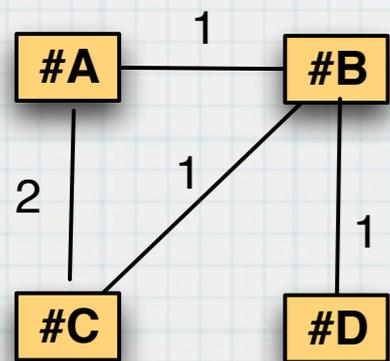
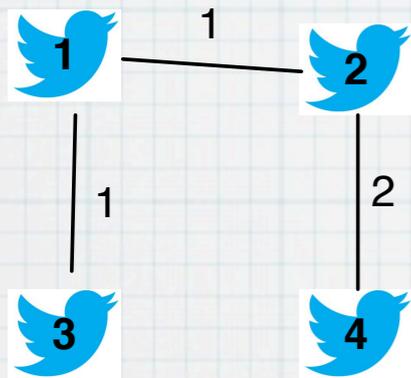


**600 Tweets with
geographical
information**



Network Topology (1)

Social Network

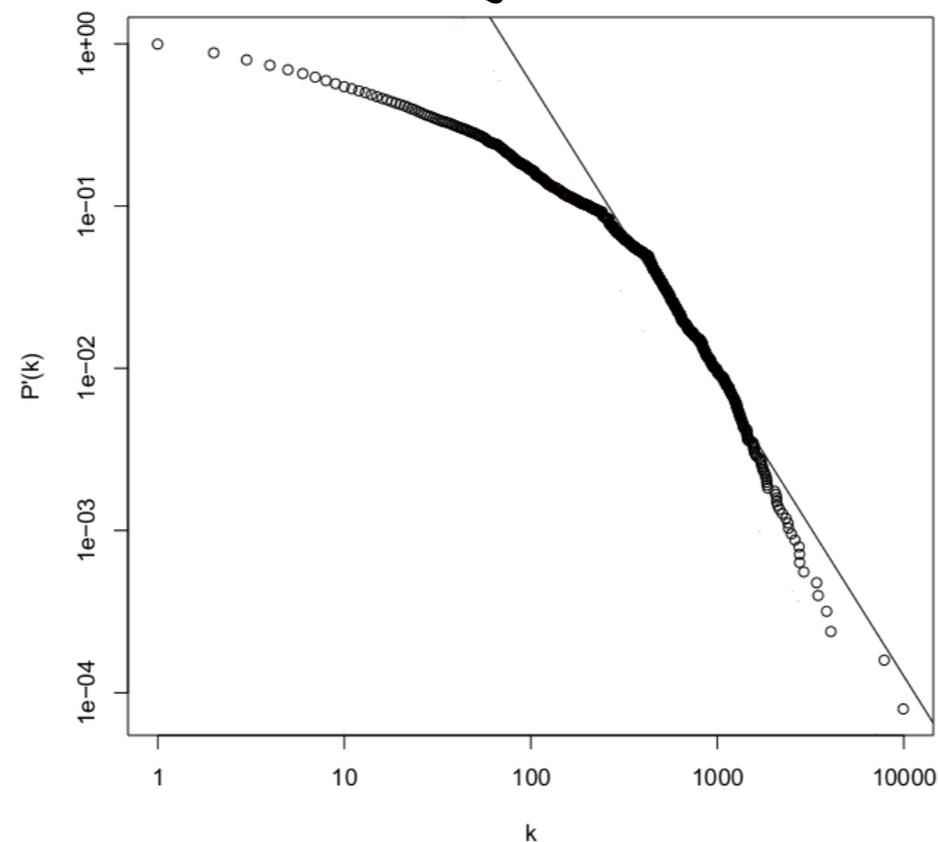


Semantic Network

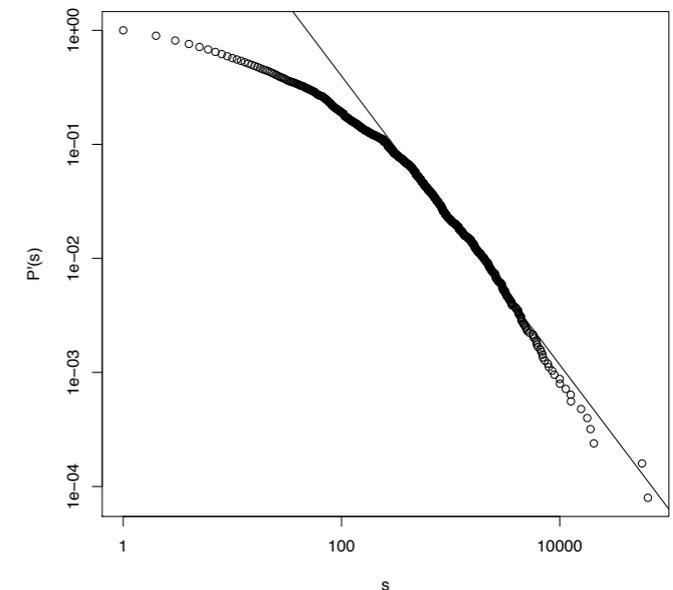
nodes = 17129
links = 497054

$\langle k \rangle = 79.19$
 $k_{MAX} = 9952$
 $\langle l \rangle = 2.21$

Degree

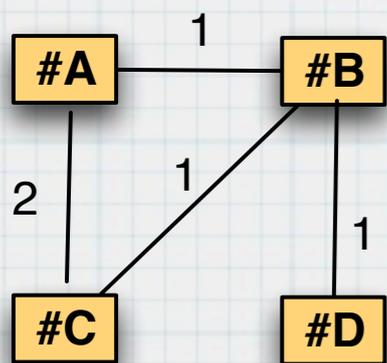
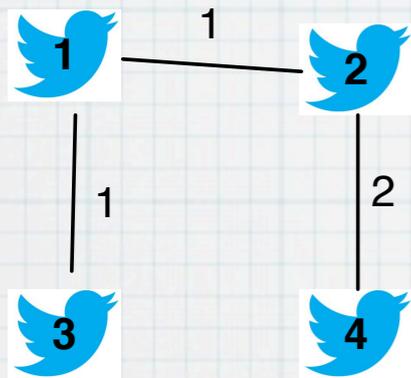


Strength



Network Topology (1)

Social Network

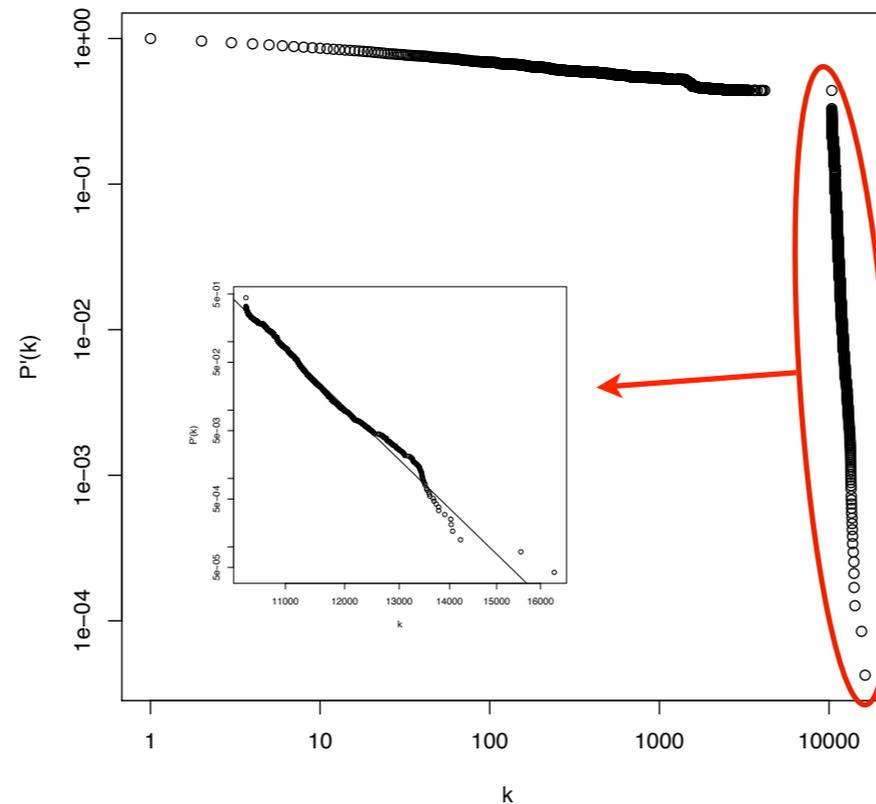


Semantic Network

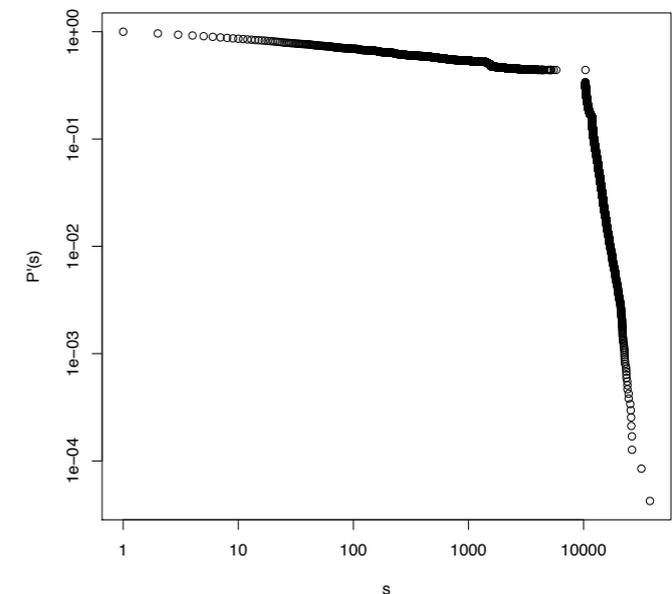
nodes = 37494
links = 58.062.071

$\langle k \rangle = 4919$
 $k_{MAX} = 16.325$
 $\langle l \rangle = 1.9$

Degree

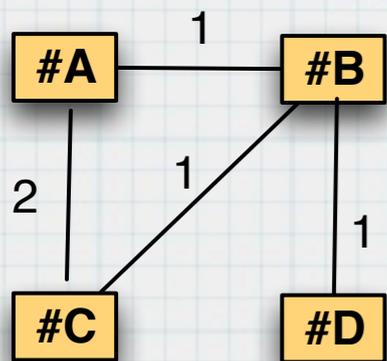
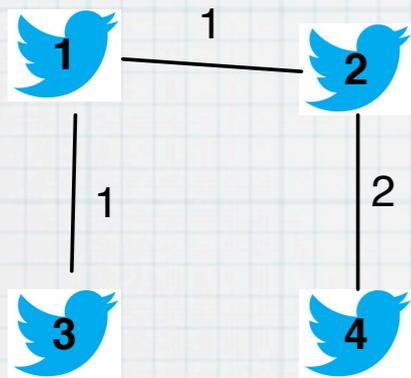


Strength



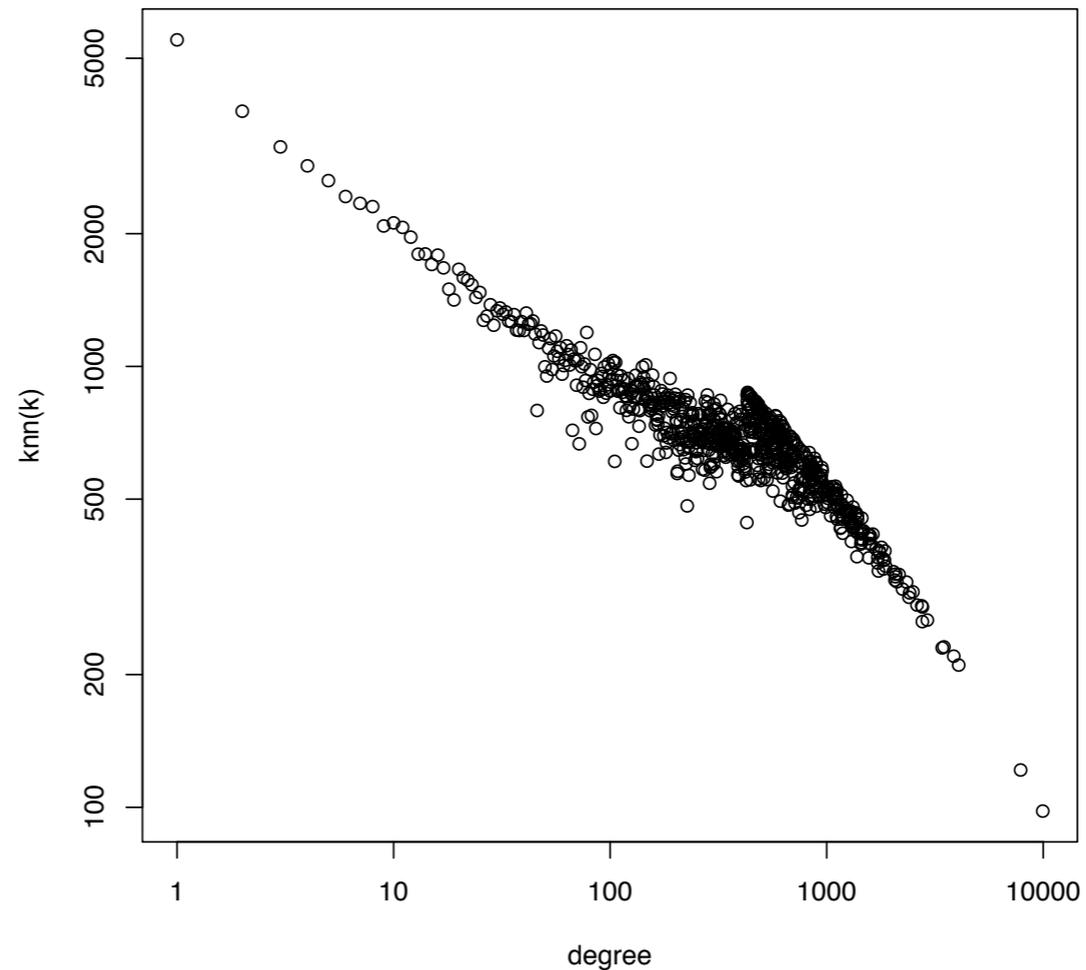
Network Topology (2)

Social Network



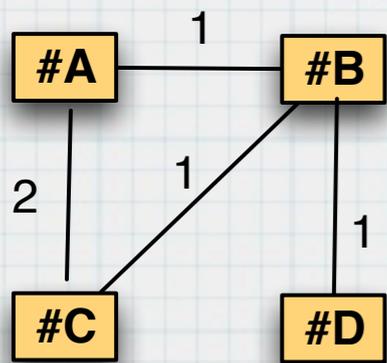
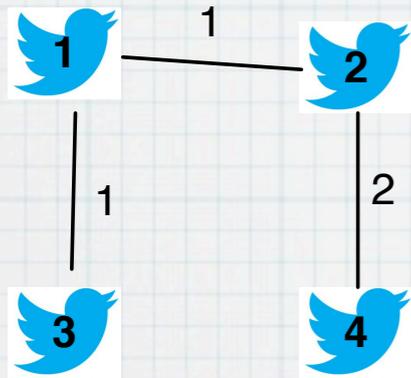
Semantic Network

Degree Assortativity



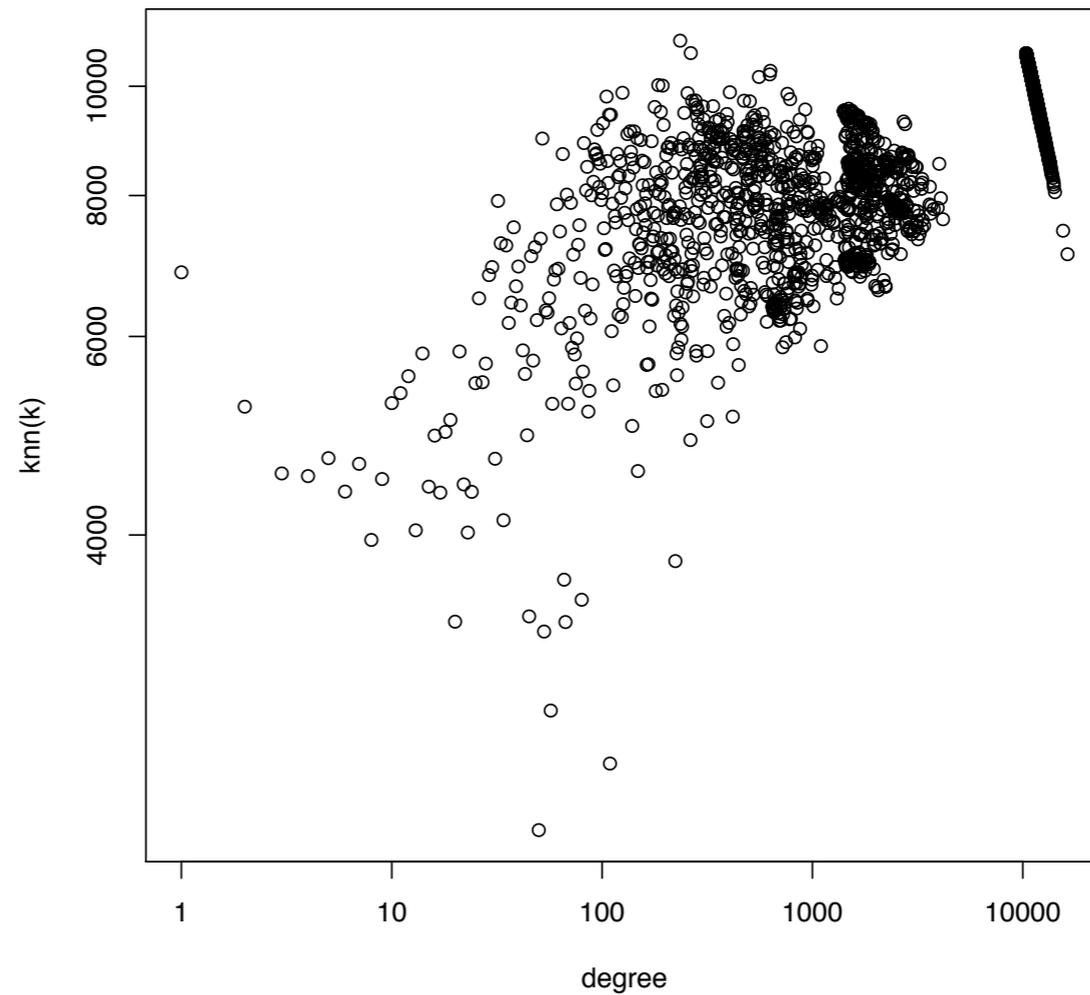
Network Topology (2)

Social Network



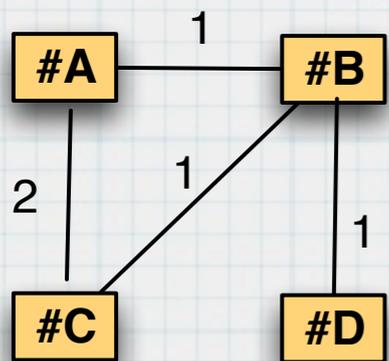
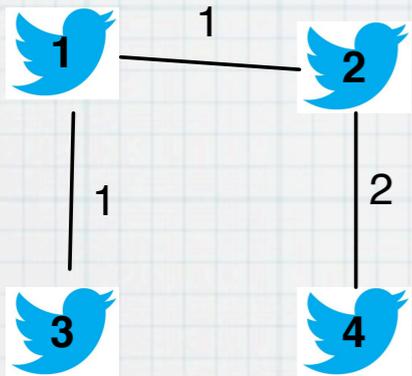
Semantic Network

Degree Assortativity



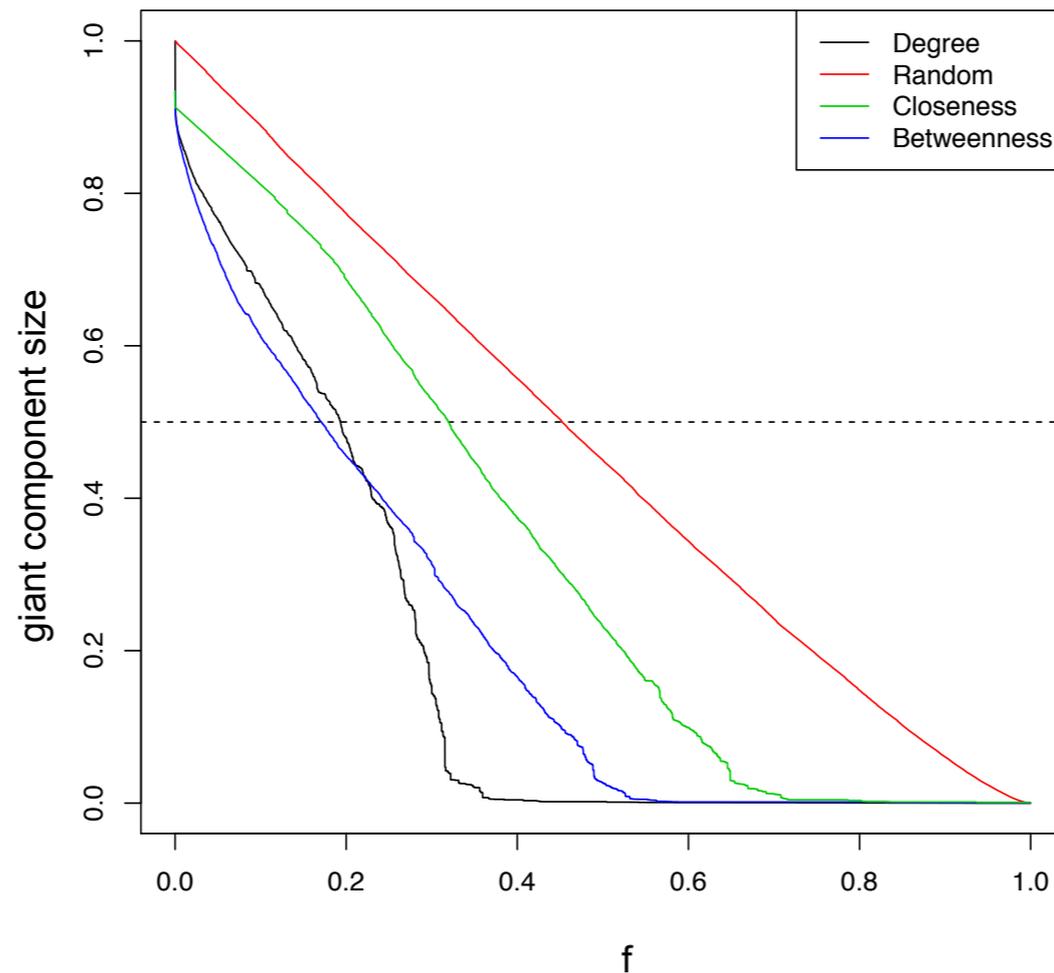
Network robustness

Social Network



Semantic Network

How many hashtags should be censored to break discussion?



Random damage

$$f_c = 0.995 \pm 0.002$$
$$\approx$$

$$f_c^t = 0.998 \pm 0.001$$

Degree preferential damage

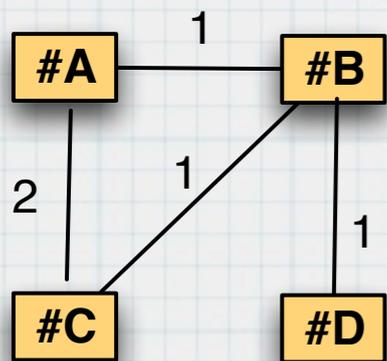
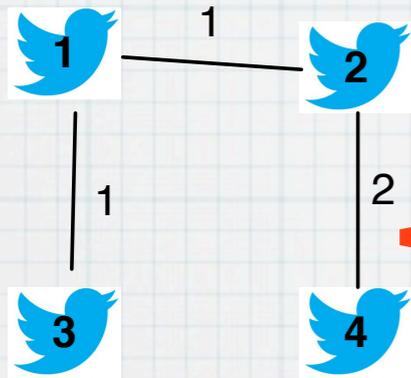
$$f_c = 0.664 \pm 0.009$$

\gg

$$f_c^t = 0.037 \pm 0.006$$

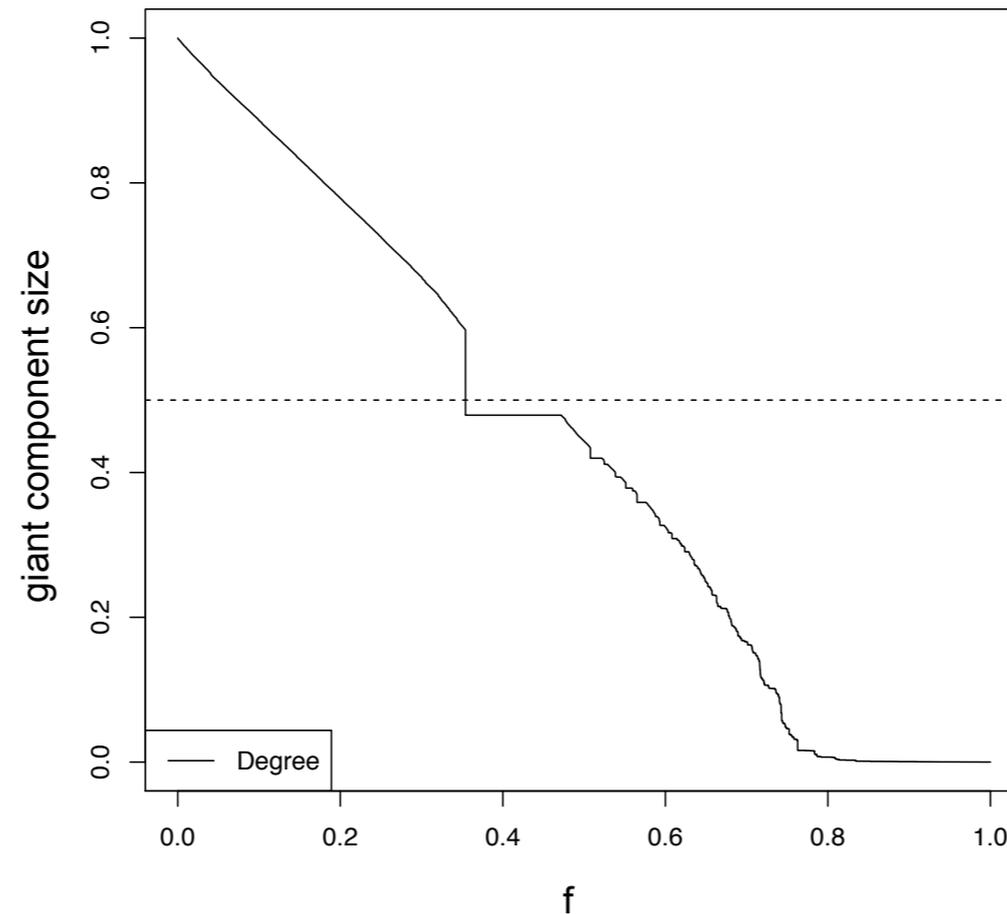
Network robustness

Social Network



Semantic Network

How many account should be blocked to stop the communication?

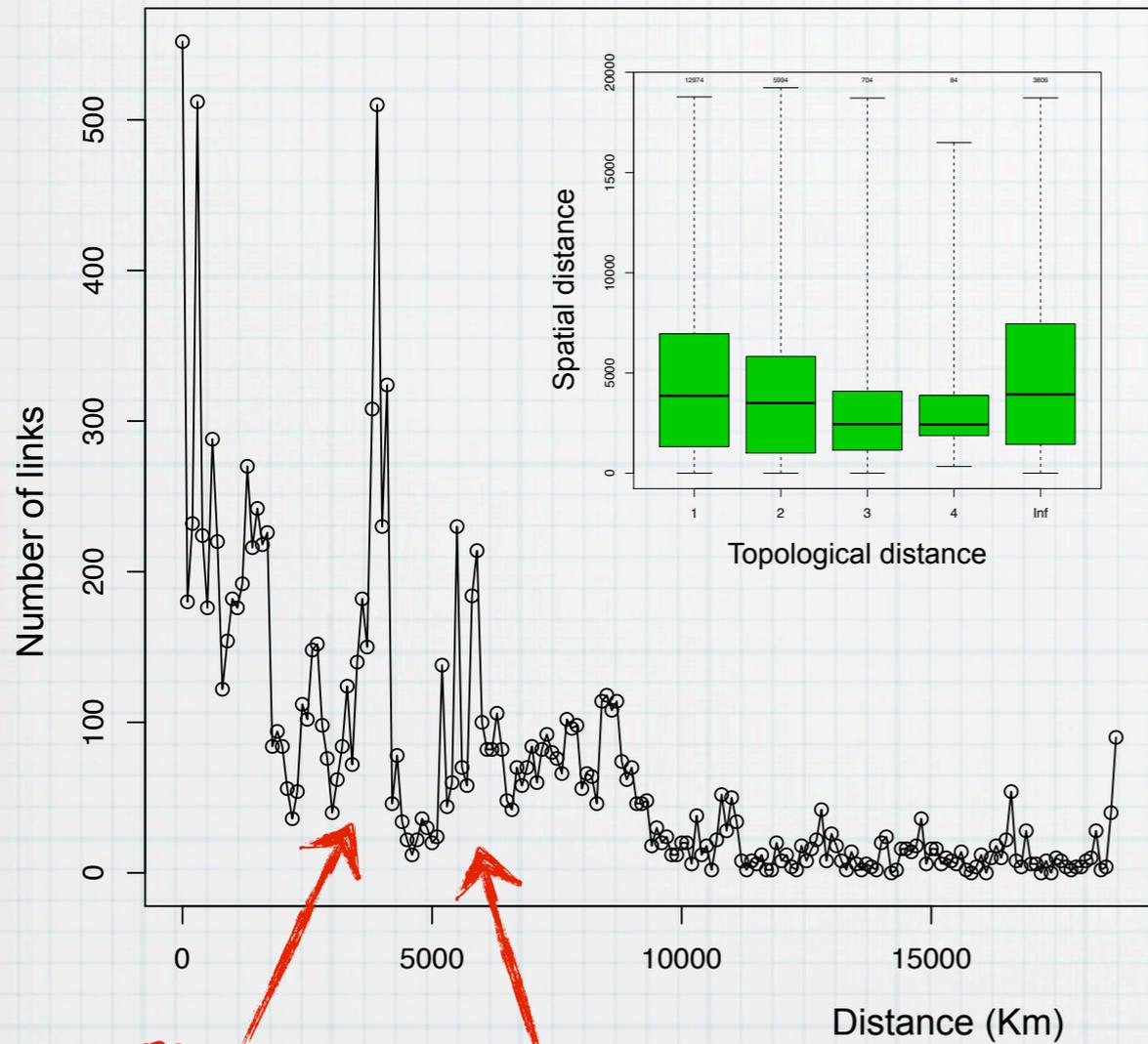


$$f_c = 0.948 \pm 0.001$$

**Extremely robust
due to redundancy
of hubs**

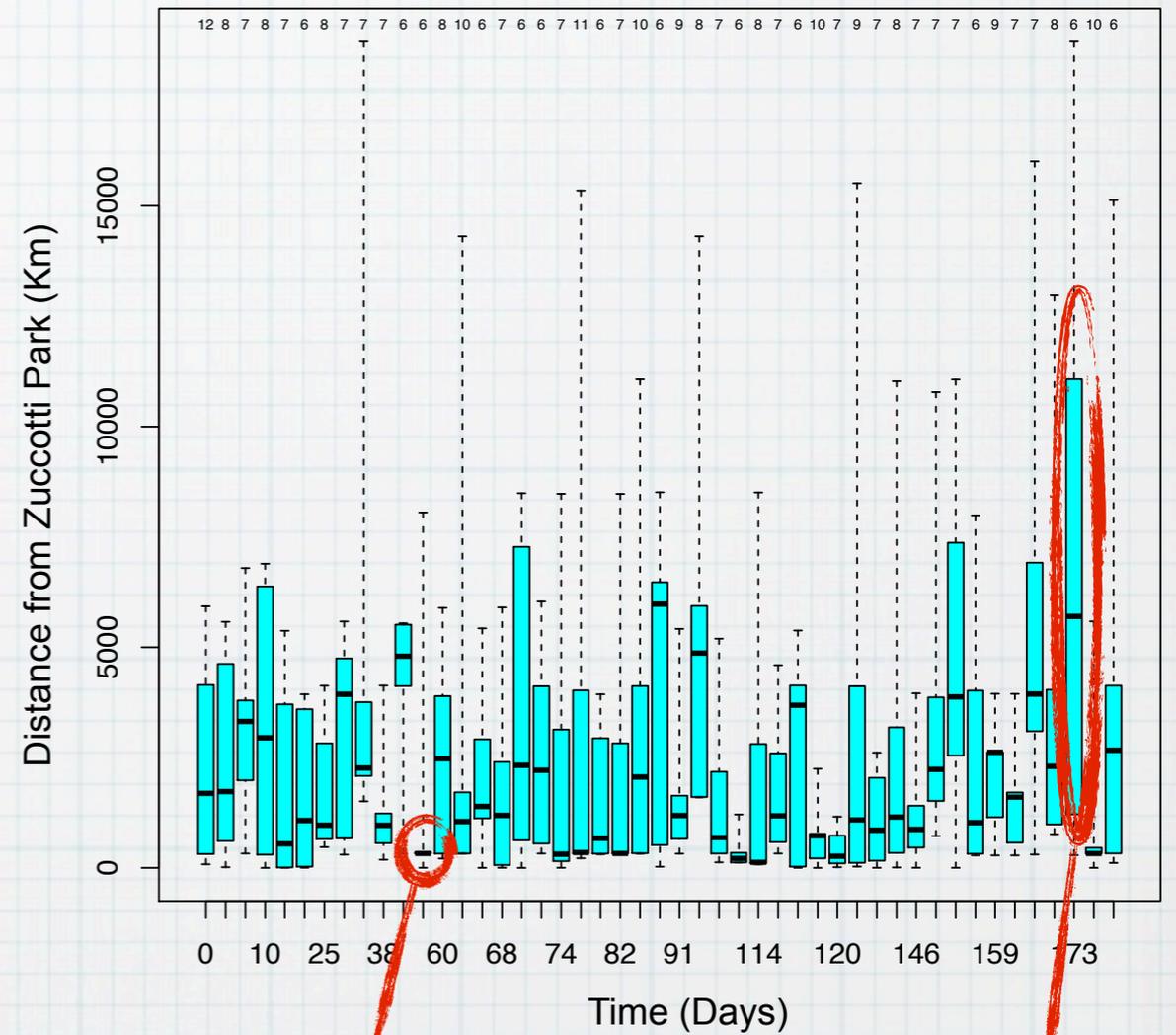
**The two groups
 $k < 4177$
 $k > 10380$
break separately**

Spatial features



Distance
East Cost-
West Cost

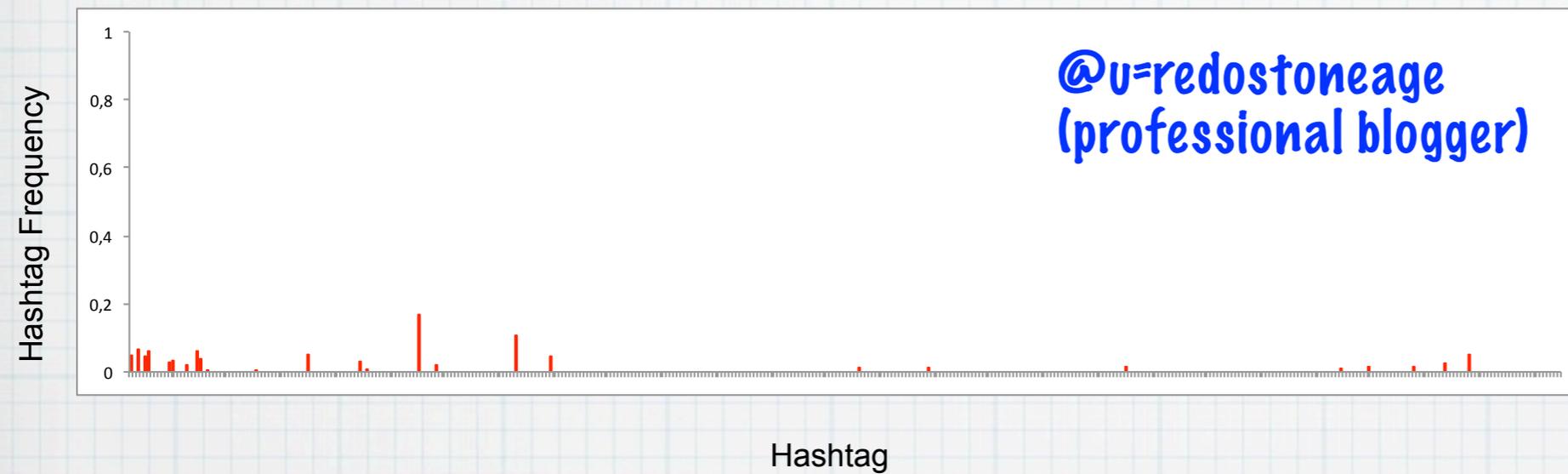
Distance
NY-
Madrid and UK



Comments on a
NYT article

Arrests
in NY

Users' Activities



Shannon entropy

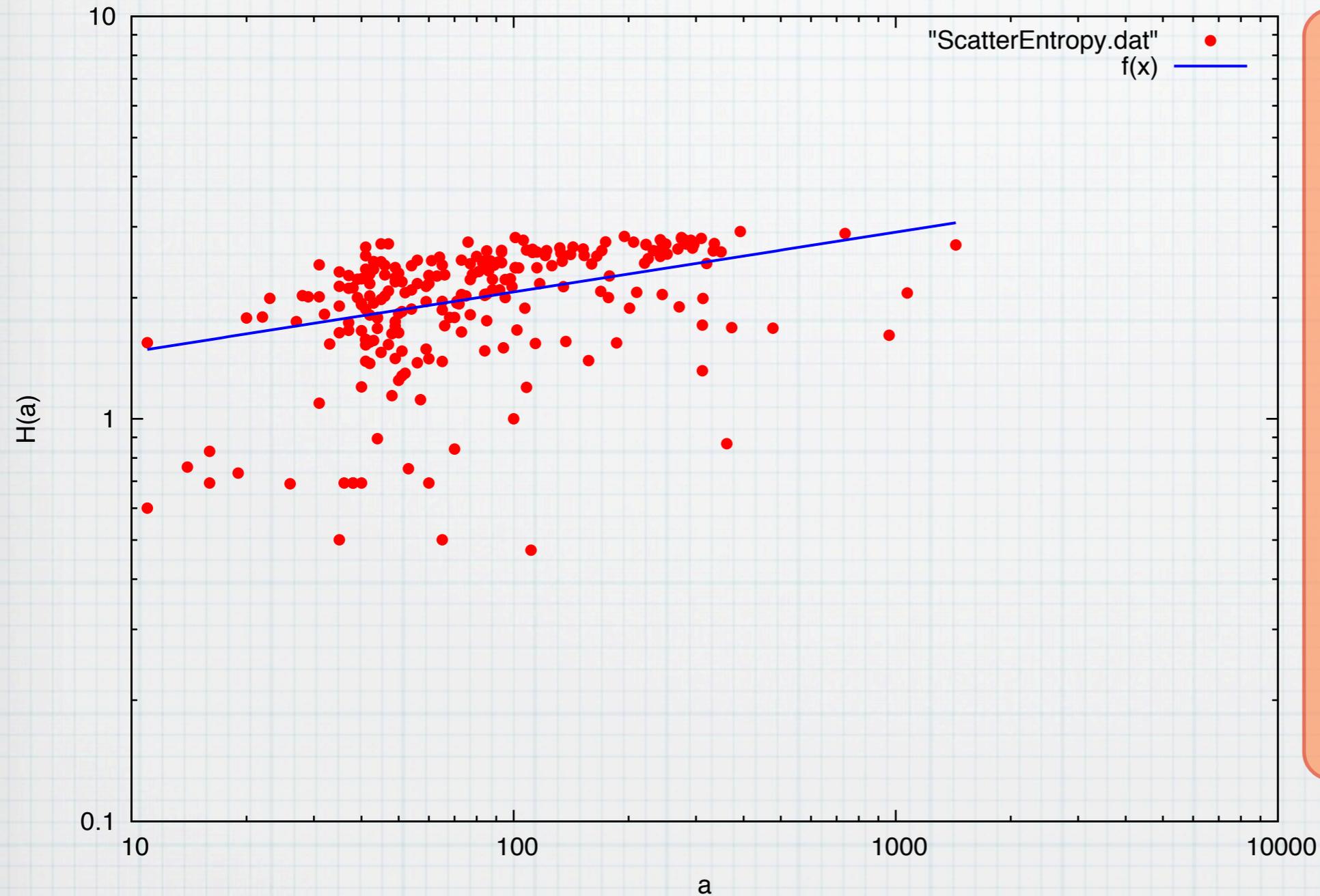
$$H_{user} = - \sum_{HT} p_{HT} \log p_{HT}$$

Limits:

$$H_{user} = 0 \rightarrow \text{Strong Hashtag preference}$$

$$H_{user} = \log(N_{HT}) \rightarrow \text{Uniform Hashtag use}$$

Users' Activities



Shannon entropy

$$H_{user} = - \sum_{HT} p_{HT} \log p_{HT}$$

Limits:

$H_{user} = 0$ → Strong Hashtag preference

$H_{user} = \log(N_{HT})$ → Uniform Hashtag use

Next steps...

- * Mutual information rate
(communication patterns)
- * Temporal behaviour:
Temporal networks
Historical series
- * More geographical data
- * Motifs in temporal network
(opinion paths)
- * Bias in data collection

