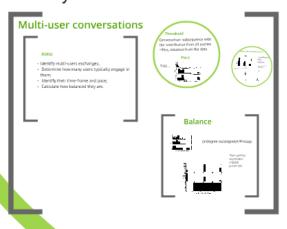
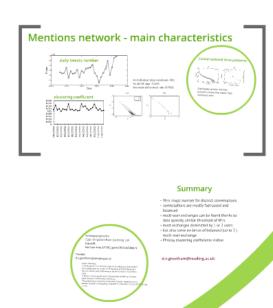


### On Twitter conversations

Danica Greetham, University of Reading

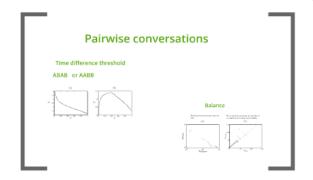
joint work with Jonathan Ward, University of Leeds





d.v.greetham@reading.ac.uk

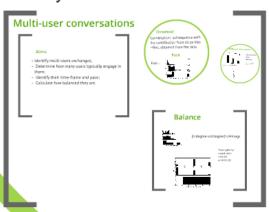


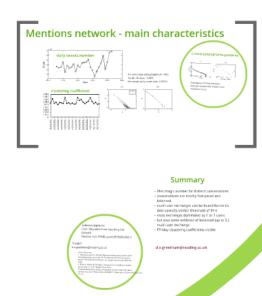


## On Twitter conversations

Danica Greetham, University of Reading

joint work with Jonathan Ward, University of Leeds





d.v.greetham@reading.ac.uk

# data

We looked at the tweets from UK to UK having @ sign

If person A puts "@B" it designates that A is addressing the tweet to B specifically, but B is not the only one who can see the message (even people who do not follow A could see this message).

numbers

4.4M, 730k,

daily:

vertices: 66877 --- 109213 (87624),

edges: 123398 -- 200295 (158380)

The data was collected by Datasift.

```
December 2011 < >
S M T W T F S
27 28 29 30 1 2 3
4 5 6 7 8 9 10
11 12 13 14 15 16 17
18 19 20 21 22 23 24
25 26 27 28 29 30 31
1 2 3 4 5 6 7
```

#### **Motivation**

"everything is about conversation and not about broadcasting"

Most of the previous work about information flow and retweeting...

Most of the previous network analysis focused on followers...

#### "A visit from the Goon Squad" by Jennifer Egan, 2011

"Those pauses, like the spaces between PowerPoint slides, become a metaphor for the gaps between what we mean and what we say"



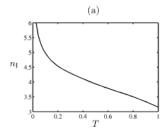
#### Aim

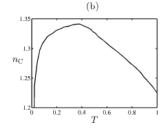
to describe main characteristics of Twitter "chatter": pace, timing, balance

### **Pairwise conversations**

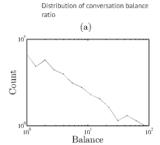
#### Time difference threshold

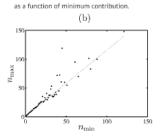
#### ABAB or AABB





#### **Balance**

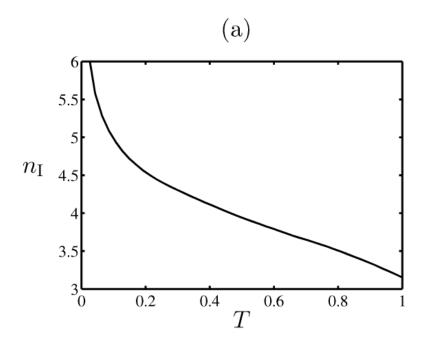


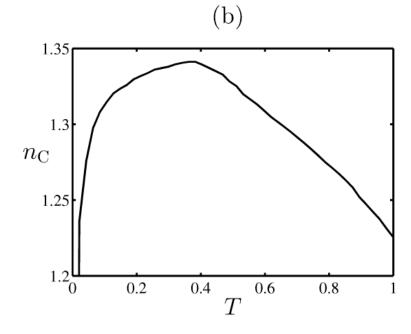


Mean maximum conversation contribution

# Time difference threshold

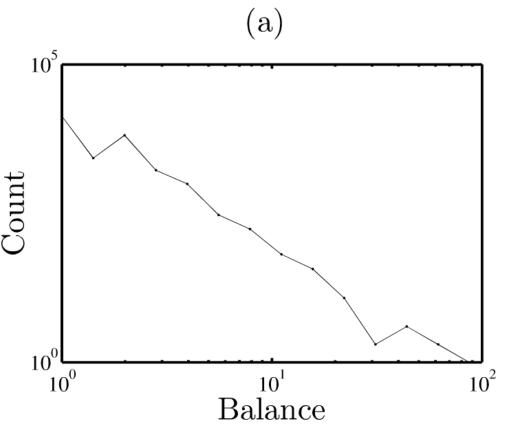
## **ABAB** or **AABB**



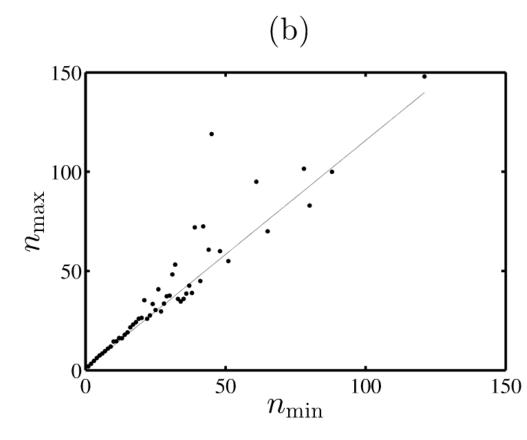


# Balance

Distribution of conversation balance ratio



Mean maximum conversation contribution as a function of minimum contribution.



### **Multi-user conversations**

#### Aims:

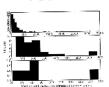
- · Identify multi-users exchanges;
- Determine how many users typically engage in them;
- · Identify their time-frame and pace;
- Calculate how balanced they are.

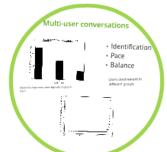
#### Threshold

Conversation- subsequence with the contribution from all parties ~9hrs, obtained from the data

#### Pace

Fast...

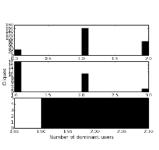








(indegree-outdegree)+/#mssgs



'floor-gaining' round robin r=|E|/|V| p=100\*r/|E|

#### Aims:

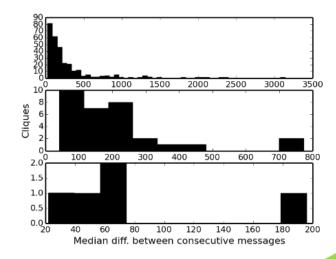
- Identify multi-users exchanges;
- Determine how many users typically engage in them;
- Identify their time-frame and pace;
- Calculate how balanced they are.

### **Threshold**

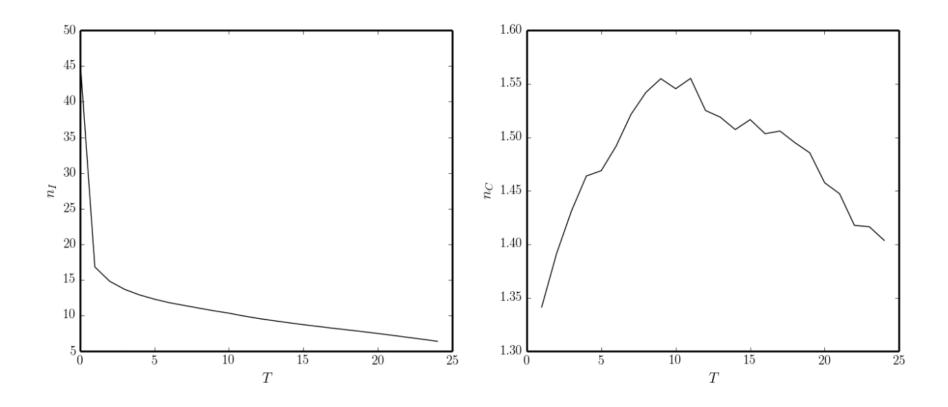
Conversation- subsequence with the contribution from all parties ~9hrs, obtained from the data

#### **Pace**

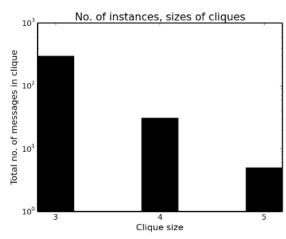
Fast...



# Idld



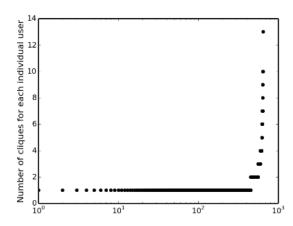
### Multi-user conversations



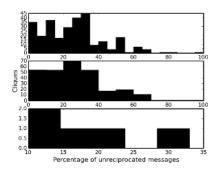
Determine how many users typically engage in them

- Identification
- Pace
- Balance

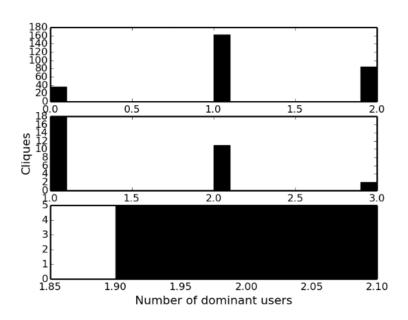
Users involvement in different groups



# Balance

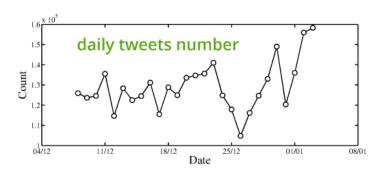


(indegree-outdegree)+/#mssgs

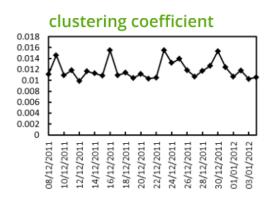


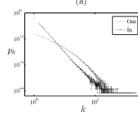
'floor-gaining' round robin r=|E|/|V| p=100\*r/|E|

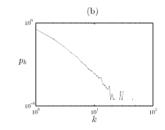
### Mentions network - main characteristics



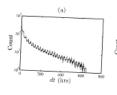
An individual daily maximum -590; for all 28 days - 5,604; the mean daily tweet rate -0.9410.





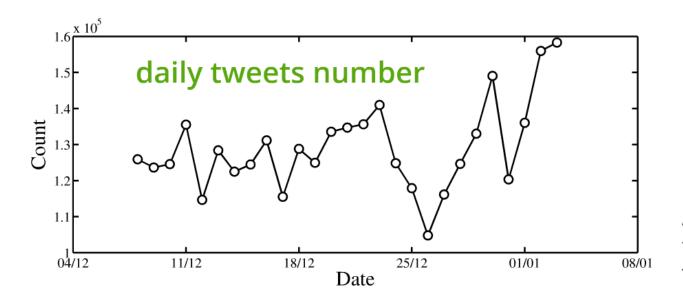


#### Conversational time-patterns

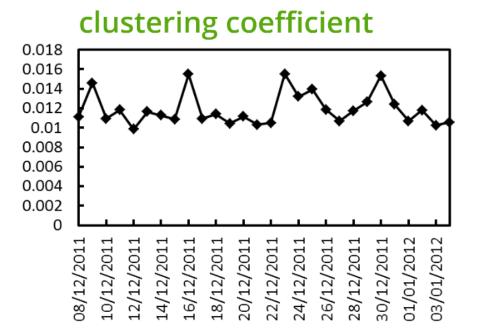


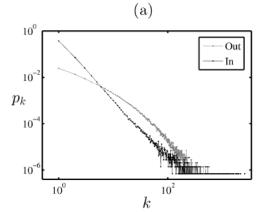


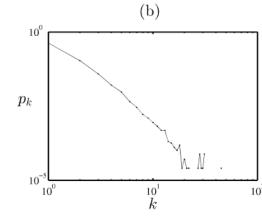
Distribution of time intervals between consecutive tweets from individual users



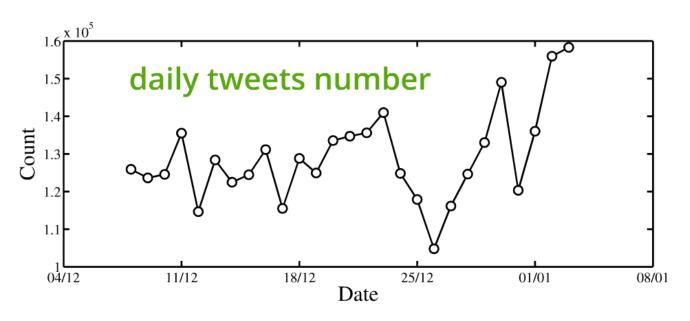
An individual daily maximum -590; for all 28 days - 5,604; the mean daily tweet rate -0.9410.



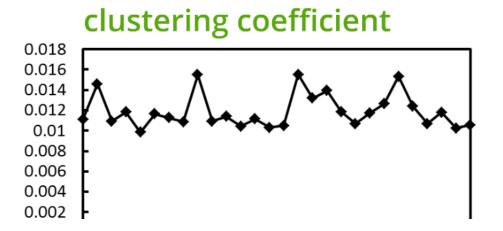


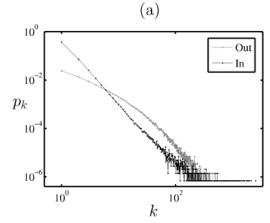


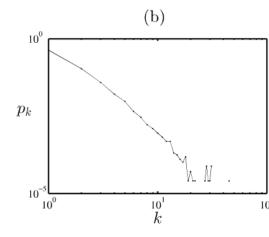
# ntions network - main char



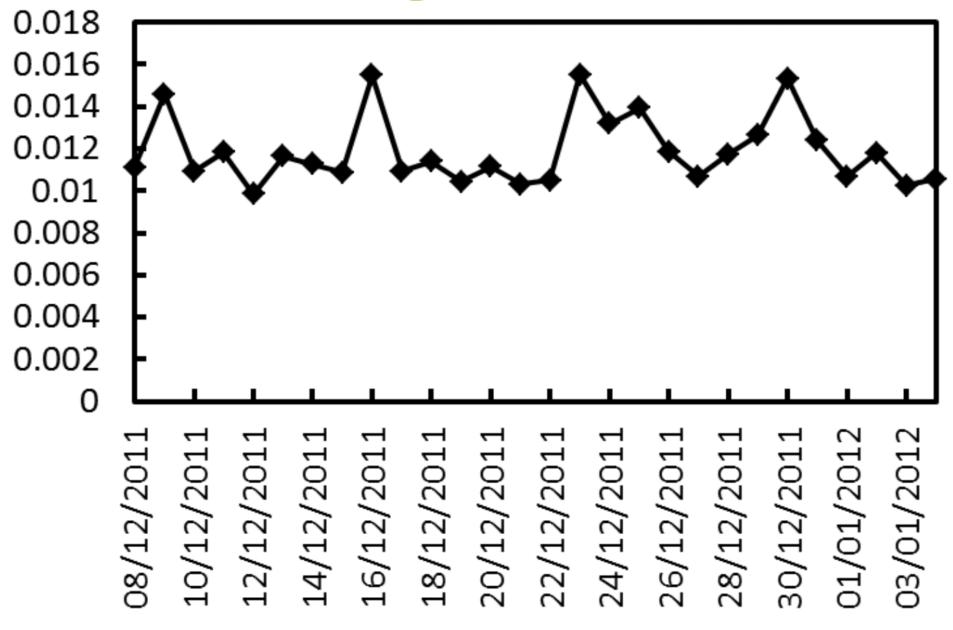
An individual daily maximum -590; for all 28 days - 5,604; the mean daily tweet rate -0.9410.

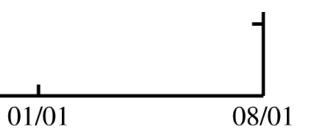




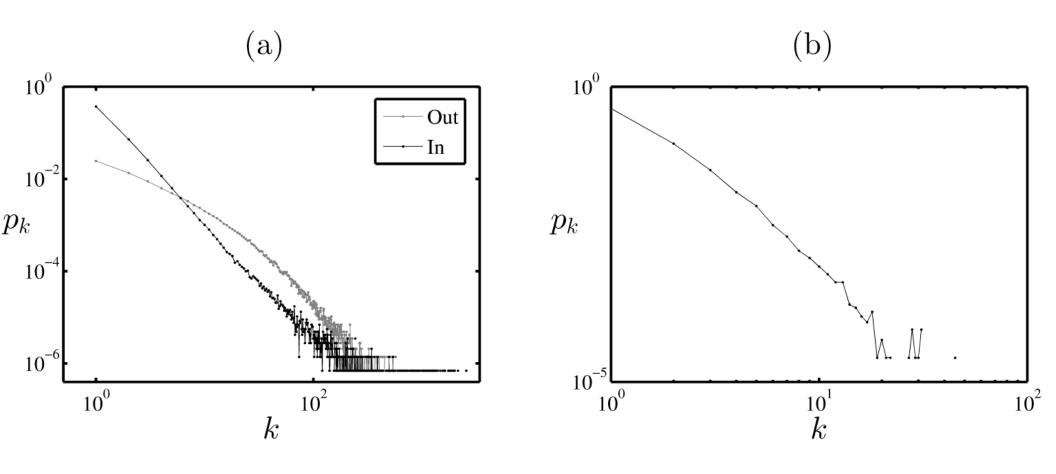


# clustering coefficient

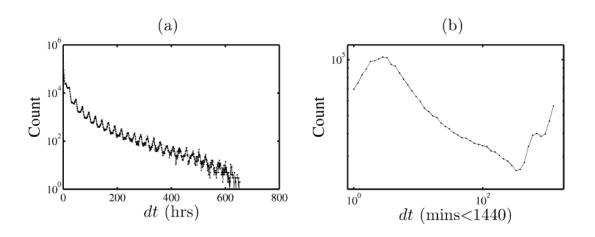




An individual daily maximum -590; for all 28 days - 5,604; the mean daily tweet rate -0.9410.



### Conversational time-patterns



Distribution of time intervals between consecutive tweets from individual users

# Summary

- 9hrs magic number for distinct conversations
- conversations are mostly fast-paced and balanced
- multi-user exchanges can be found thanks to data sparsity, similar threshold of 9hrs
- most exchanges dominated by 1 or 2 users
- but also some evidence of balanced (up to 5) multi-user exchange
- FFriday clustering coefficients visible

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Datasift
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# Thanks! d.v.greetham@reading.ac.uk

#### Some references:

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A. Ritter, C. Cherry, and B. Dolan, "Unsupervised modeling of twitter conversations," in HLT-NAACL 2010, 2010.

"Characterizing and curating conversation threads: expansion, focus, volume, re-entry," in Proceedings of WSDM '13. New York, NY, USA: ACM, 2013, pp. 13–22.