Raphtory: A new tool for large temporal networks applied to the far right social network Gab

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Aims of this talk (unordered list)

Aim A: Give insight into data

What drives the evolution of the alt-right social network gab?

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Aim α : Sell you a tool

Raphtory is an open-source, big-data platform developed at QMUL. It is unique in its ability to perform flexible temporal analysis on batch or streamed graph data.

- 95GB raw data (19 million posts) from gab platform (medium data).
 - Data is user, posts (threaded), timestamps and other metadata.
 - NB our research on structure not content.
- First eighteen months of data available.
 - September 2016 May 2018.
- Largely complete data from this period.
 - "free speech" focus means "everything" is public.
- Alt-right focus:
 - Racism, fake news, hate speech, radicalisation.

Gab Is the Alt-Right Social Network Racists Are Moving to

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- BBC Trending Gab: Free speech haven or alt-right safe space?

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Social $\overline{\text{network}} \to \overline{\text{graph } G}$

We can form a graph from a social network by creating an edge between two users if they have a relationship, interact, etc.

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Social network o many graphs G(t, au)

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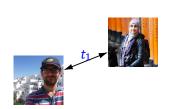




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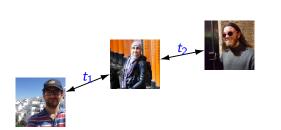




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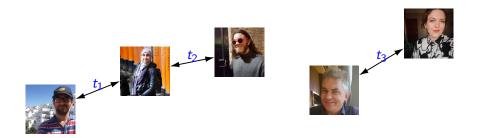




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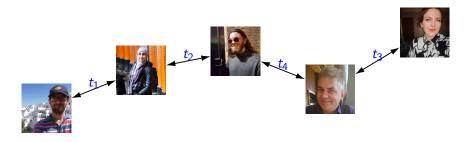
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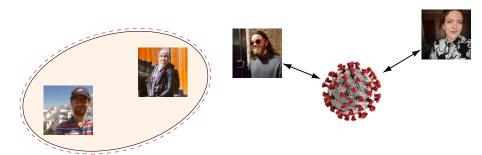




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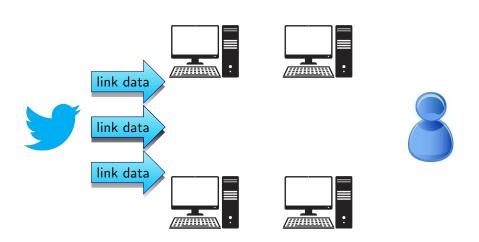


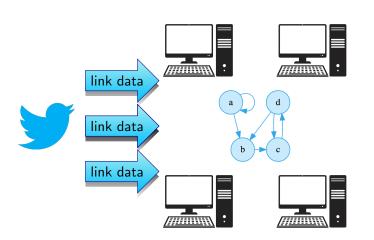




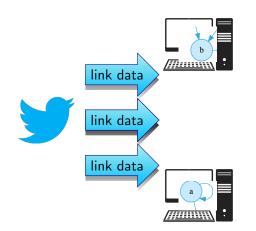








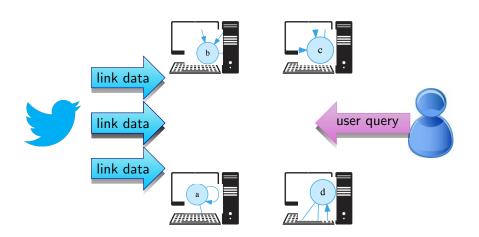


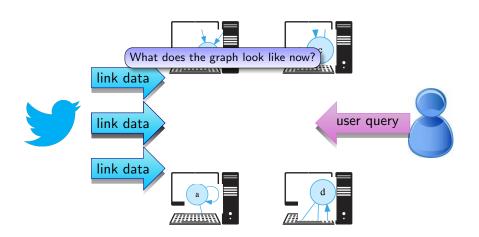


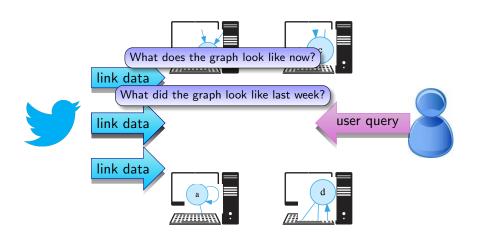


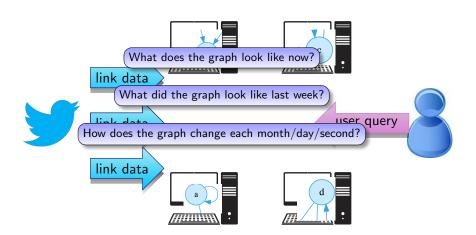


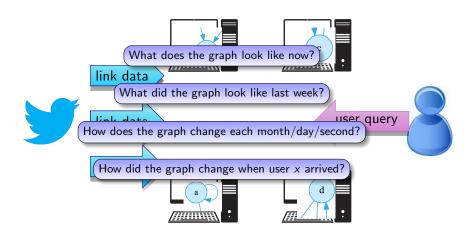








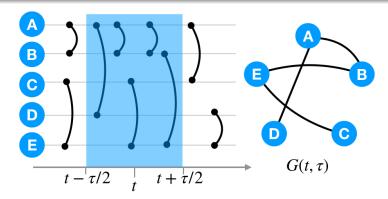




Temporal graphs – interaction graphs

Interaction graph for time t and a window length au

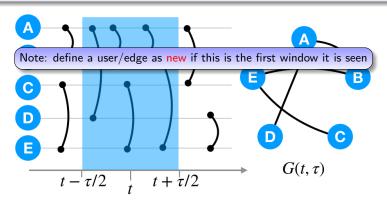
The graph $G(t,\tau)$ is defined by the set of all edges i,j where i and j interact at a time T such that $t-\tau/2 \le T \le t+\tau/2$.



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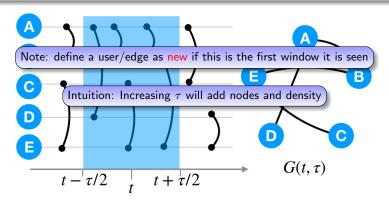
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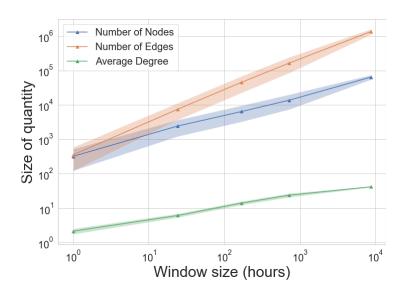
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Checking our intuition

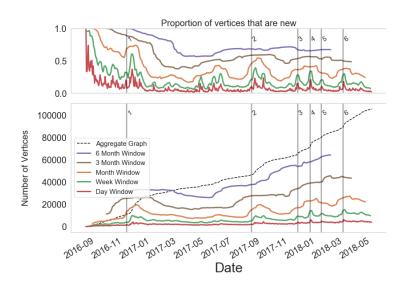


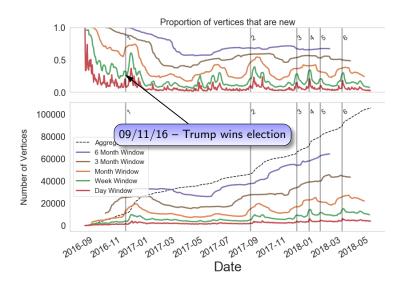
Some research questions for gab data

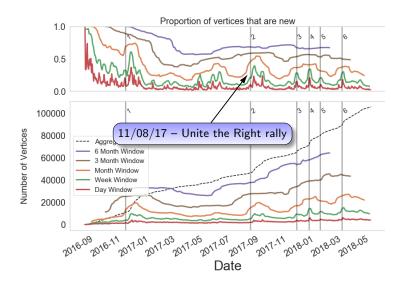
What simple questions can we ask of the data

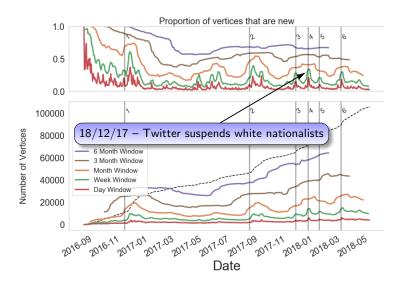
Remember our focus is structure not content. We are not (we cheat a little here at one point) digging into the content of messages.

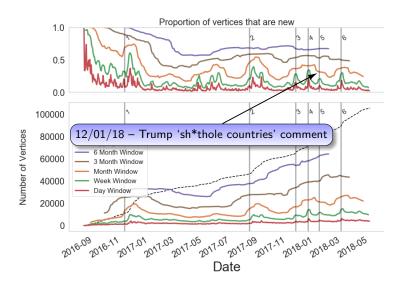
- Is gab growing, what drives the growth?
- Is gab a "social" network (friends interacting)?
- Is gab a "community" (in a loose sense)?
- Is gab controlled by an "elite"?

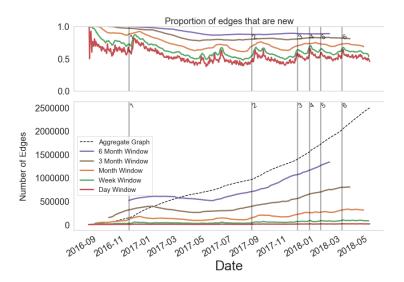


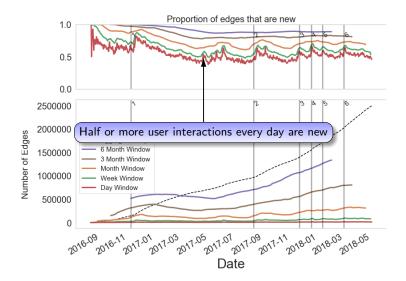


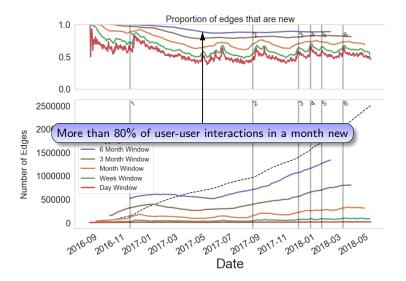


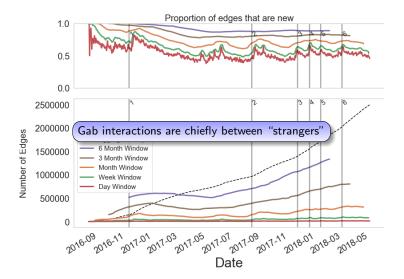












Is gab a community? Largest Connected Component

Definition: Connected component

A connected component is a sub graph of a graph within which all nodes can trace a path to each other. The largest connected component (LCC) is the one with the most nodes.

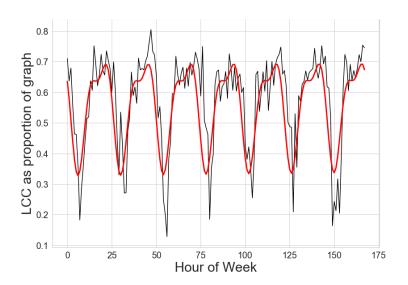
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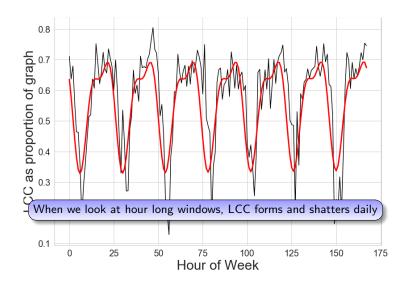
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- LCC is size of the largest "community" (in loosest sense).
- Remember: only count people active within the window.
- Expectation: in a "large" window most users within the LCC.
- But what happens as we look at smaller and smaller windows?

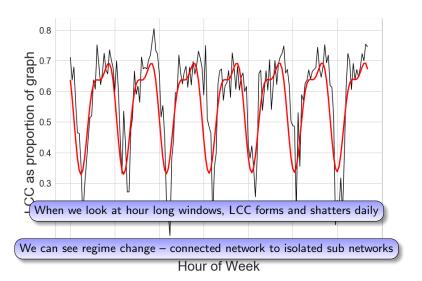
Is gab a community? LCC (hour window)



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Jaccard Similarity (for top N users)

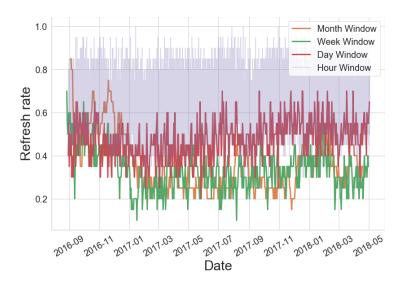
Let A, B be set of top N users in windows W_A, W_B .

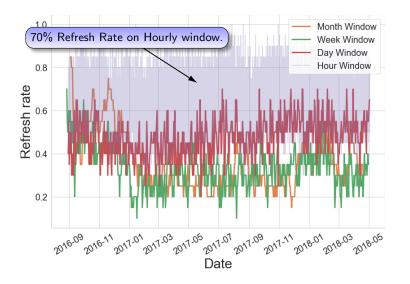
$$J(A,B)=\frac{A\cap B}{A\cup B}.$$

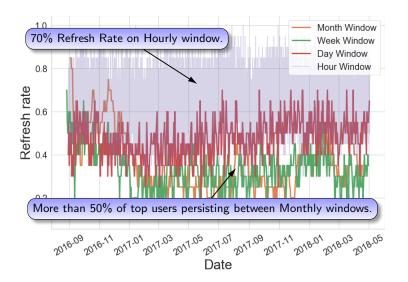
Refresh rate top N users, windows W_A, W_B

Refresh rate R = 1 - J(A, B)

 $0 \to W_A$ same users as W_B and $1 \to$ no users the same.







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- Gab is not a "social" network interactions between "strangers" not friends.
- The interactions within Gab do not always form a "connected" community. We observed a daily shattering of the LCC and diurnal change of regime never observed before.
- Are a cadre of elite users controlling users' attention? Not clear: At longer timescales there is a group who receives a lot of attention. At only one timescale you will get a misleading answer.

Conclusions (about Raphtory and temporal networks)

- Temporal networks provide a rich array of techniques that can get more out of your network data.
 - Window based analysis provides many levels of insight.

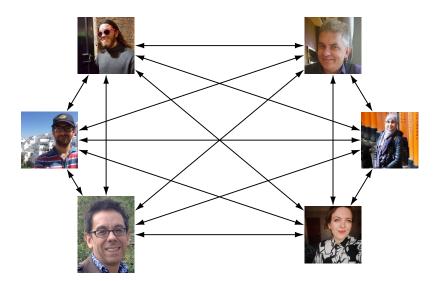
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- Temporal networks provide a rich array of techniques that can get more out of your network data.
 - Window based analysis provides many levels of insight.
- The Raphtory tool is a great way to look at temporal graph data.
 - Urban analytics intervention in social networks.
 - Bitcoin/blockchain tracking "dark markets".
 - Semantic networks changing word meanings.
 - Other social networks compare and learn more.

Our Raphtory social network (a small subgraph)



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