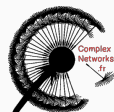


LSCPM: finding communities in Link Streams by Clique Percolation Method

Alexis BAUDIN*, Lionel TABOURIER and Clémence MAGNIEN

June 21st, 2023

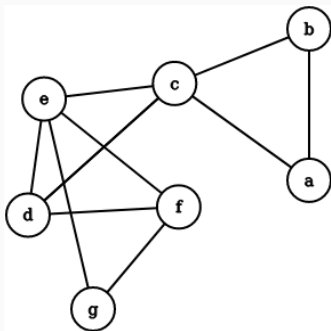
Lip6, Sorbonne Université, CNRS, Complex Networks



- 1 - Link stream formalism to model temporal data
- 2 - Describing temporal data by finding communities in link streams
- 3 - Experiments on real datasets
- 4 - Conclusion: links with BCI ?

1 - Link stream formalism to model temporal data

> Definition – Graph



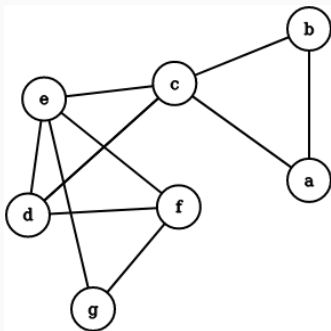
Graph formalism

- Vertices: a, b, \dots, g
- Interactions: edges

Example

Contacts between people, brain networks, ...

> Definition – Graph



Graph formalism

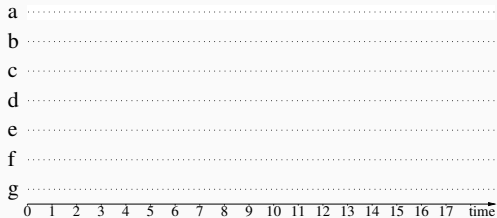
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Example

Contacts between people, brain networks, ...

→ What about temporal interaction ?

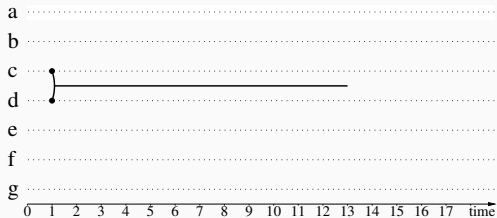
> Definition – Link stream



Link stream formalism

- Vertices: a, b, \dots, g
- Time period: $[0, 18]$
- Interactions: temporal edges

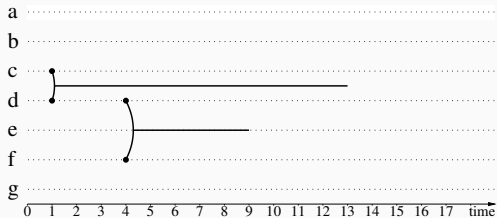
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Link stream formalism

- Vertices: a, b, \dots, g
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 - c, d linked over $[1, 13]$

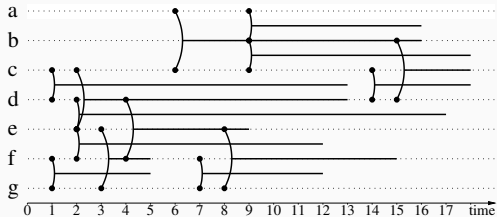
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Link stream formalism

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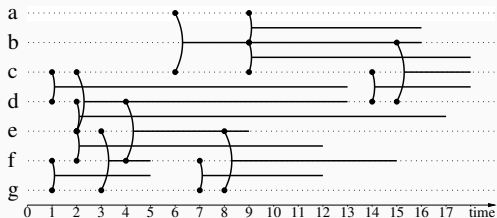
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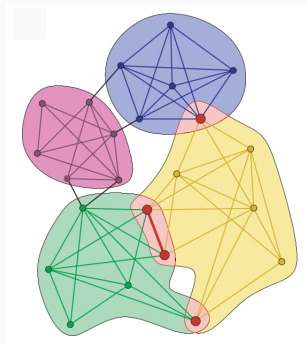
Advantages

- deals directly with the stream of interactions
- no arbitrary choice of time scale
- time is continuous

2 - Describing temporal data by finding communities in link streams

Communities: sets of vertices

- Densely connected inside
- Sparsely connected outside



Palla et al. 2005

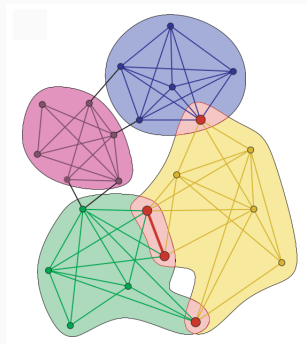
> Communities in static networks

Communities: sets of vertices

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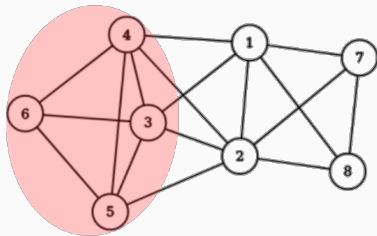
Interest:

- Locate areas of high interaction density
- Understanding the organizational structure of interactions
- Zoom in / out



Palla et al. 2005

> Clique Percolation Method in static networks (CPM)

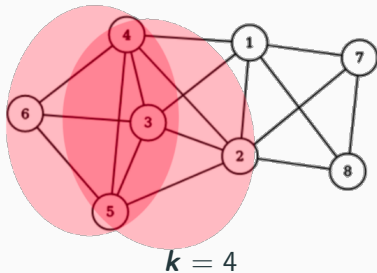


$$k = 4$$

k-clique

Set of k nodes all connected to each other.

> Clique Percolation Method in static networks (CPM)



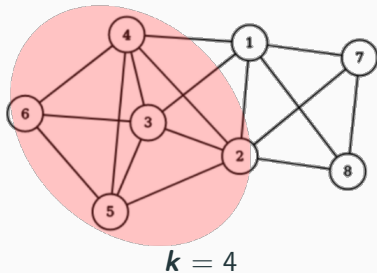
***k*-clique**

Set of k nodes all connected to each other.

Grouping rule

Two k -cliques are adjacent if they share $k - 1$ nodes.

> Clique Percolation Method in static networks (CPM)



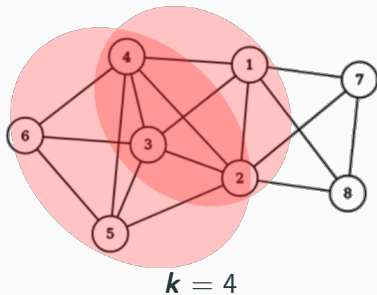
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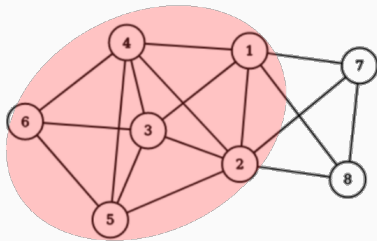
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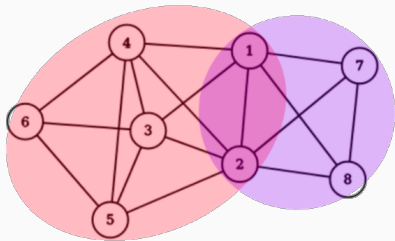
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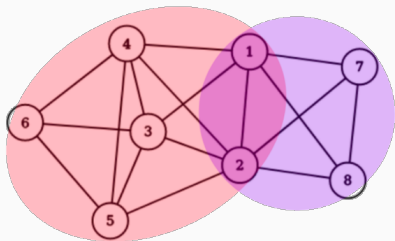
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Advantages of the definition

- Definition local
- Deterministic; no need of heuristic function
- Communities can overlap

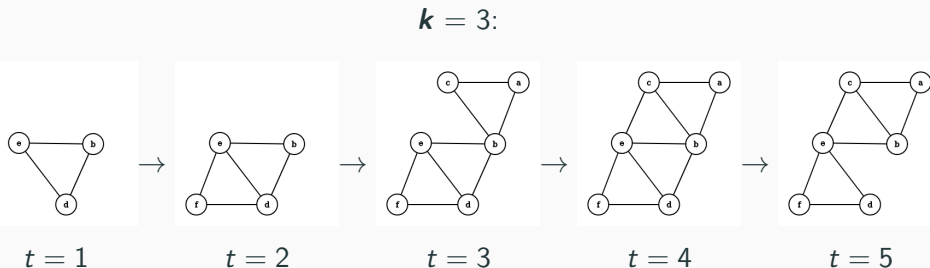
> First extension of CPM to Temporal Graphs (DCPM)

Question

How to extend CPM communities to temporal networks ?

CPM communities in **temporal graphs**: *Palla et. al 2007*

⇒ Communities that evolve from one time to the next



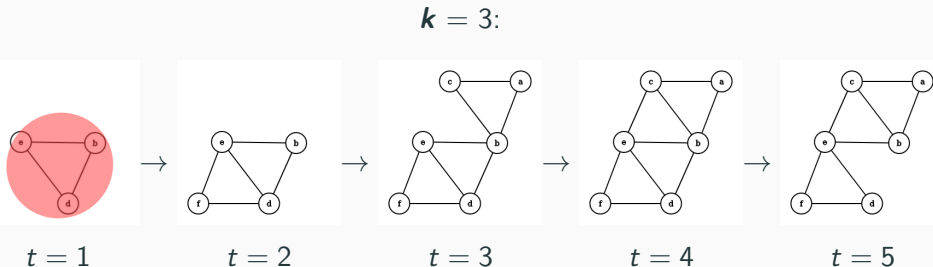
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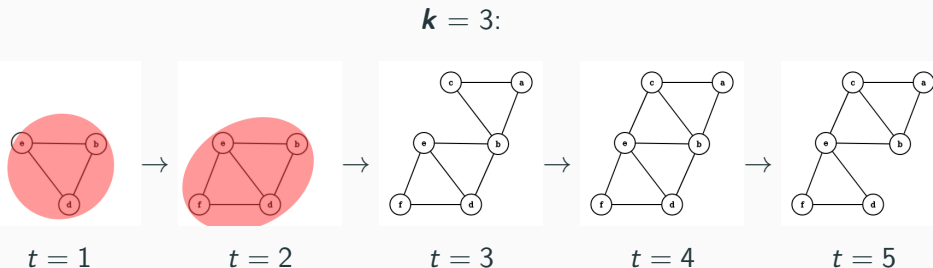
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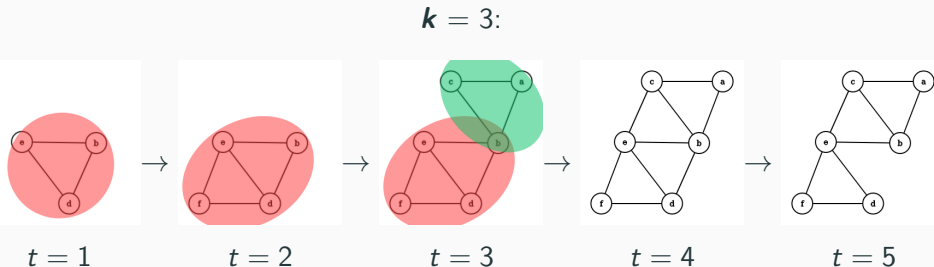
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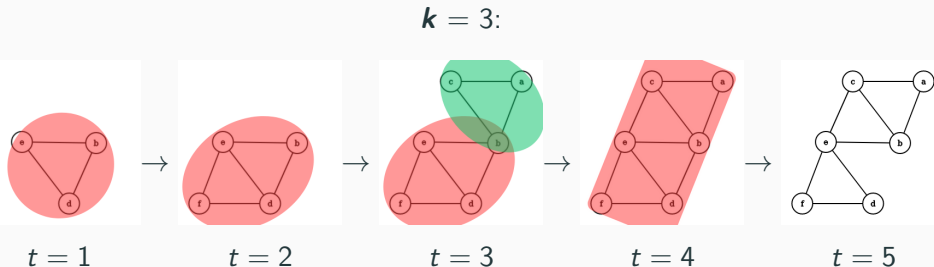
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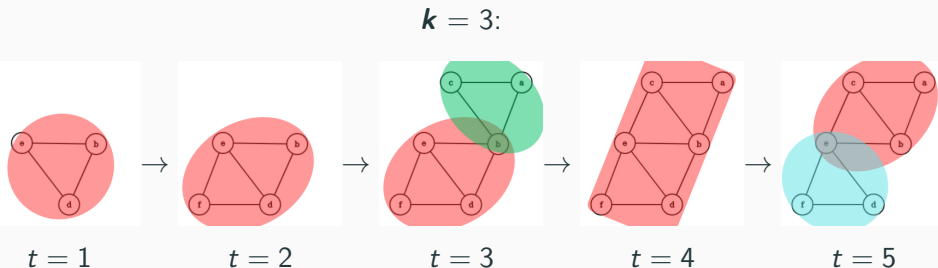
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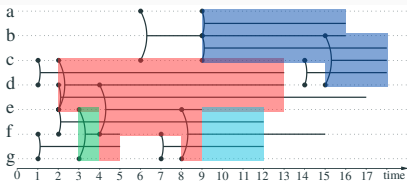
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> LSCPM: CPM in Link Streams

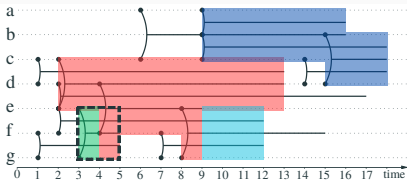
CPM in temporal graphs



- Computing communities at each time step: time consuming;
- Some temporal data expected to be grouped are not.

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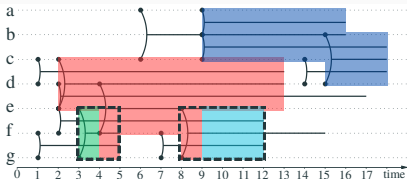
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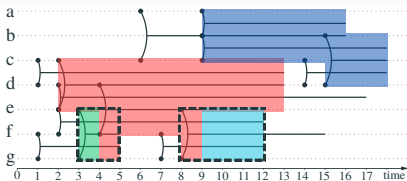
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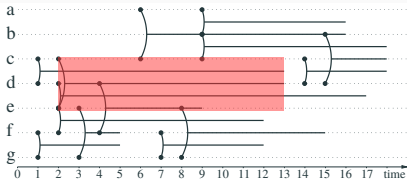
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CPM in link stream (LSCPM)

k -clique in link stream

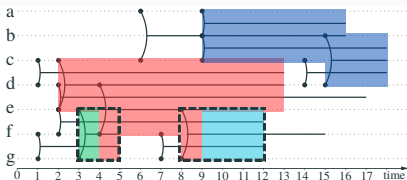
k nodes and $[t_0, t_1]$ such that all nodes are connected to each other over $[t_0, t_1]$.

\Rightarrow Grouping rule with $k = 3$:



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CPM in temporal graphs



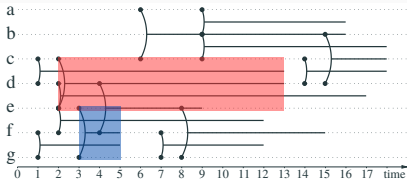
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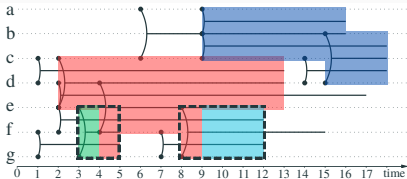
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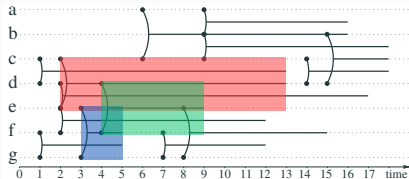
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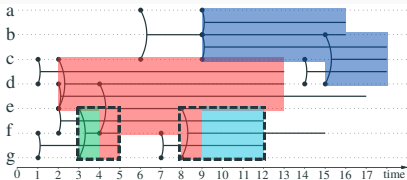
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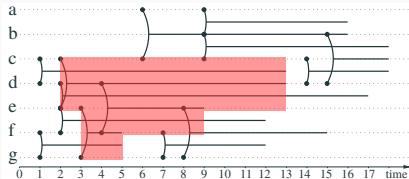
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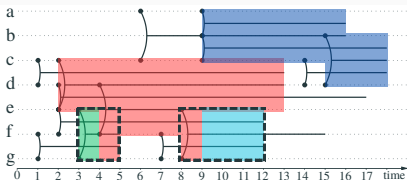
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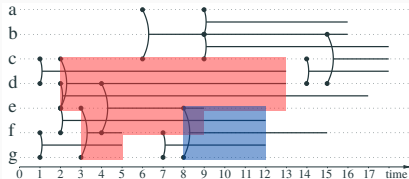
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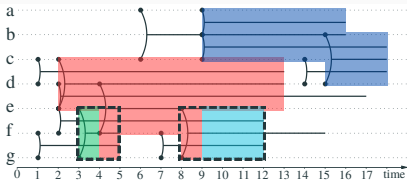
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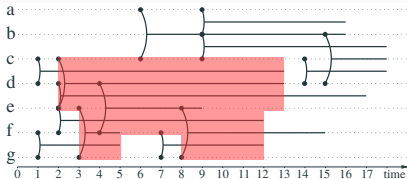
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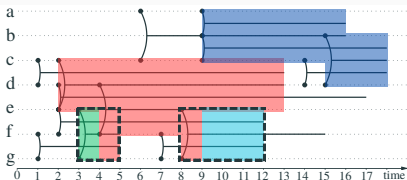
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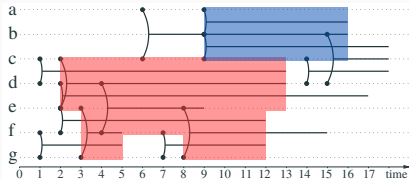
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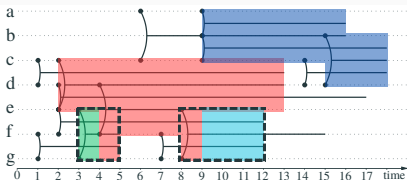
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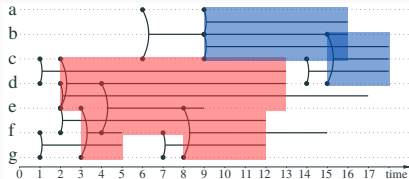
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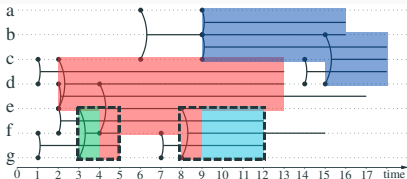
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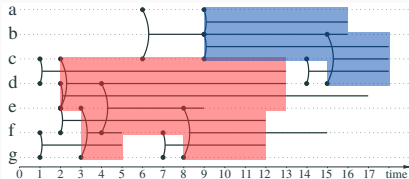
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3 - Experiments on real datasets

> An algorithm efficient and consistent

Efficiency – computation time

Link stream	Nb links	$k = 3$		$k = 4$	
		S. of art	LSCPM	S. of art	LSCPM
<i>Households</i>	2,136	1.5s	0.1s	1.0s	0.1s
<i>Highschool</i>	5,528	3.6s	0.1s	1.9s	0.1s
<i>Infectious</i>	44,658	10min49s	1.4s	6min12s	3.3s
<i>Foursquare</i>	268,472	3h01min	9.2s	2h28min	43s
<i>Wikipedia</i>	39,953,380	-	13min44s	-	15min29s

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Efficiency – computation time

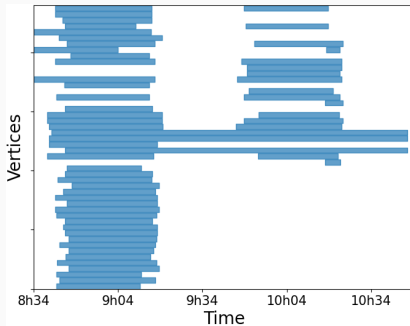
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Consistency with metadata

Highschool: 70% of communities are on one class, 23% on two classes, 6% on three classes, 1% on four classes.

A highschool link stream community

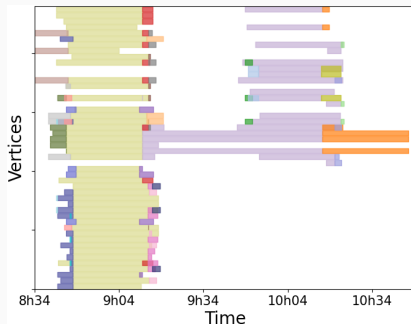
LSCPM community



> Insights on temporal communities

A highschool link stream community

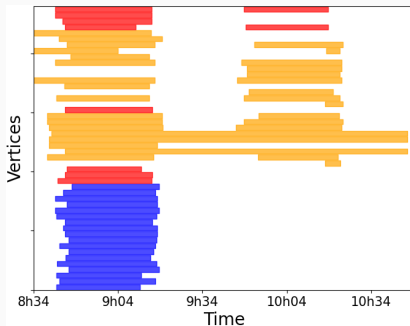
State-of-the-art communities



→ Gather more information over time

A highschool link stream community

Metadata (classes)



- Gather more information over time
- Relate metadata information

4 - Conclusion: links with BCI ?

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→ **Communities in link streams**

- (Re)organization of interactions over time
- Target communities that provide temporal information
- Target vertices that play a central role (or not)

4 - Conclusion: links with BCI ?

→ **Communities in link streams**

- (Re)organization of interactions over time
- Target communities that provide temporal information
- Target vertices that play a central role (or not)

→ **Link stream**

- Study at different time scales
- Online interactions
- Multilayer link stream ?...

Thanks for your attention! Any questions?

Code available at:

`https://gitlab.lip6.fr/baudin`

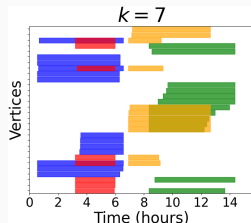
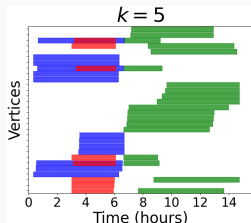
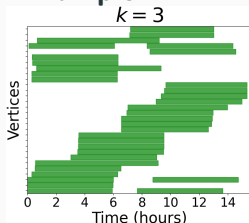
Alexis Baudin – `alexis.baudin@lip6.fr`

Supplementary material

Observation

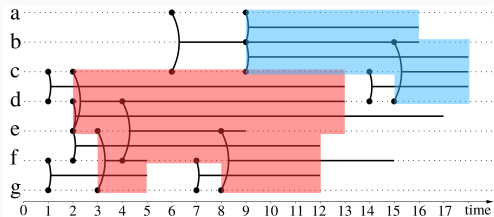
As k increases, communities split.

Example:



⇒ Refining communities; “core”

⇒ Sub-categorizing data



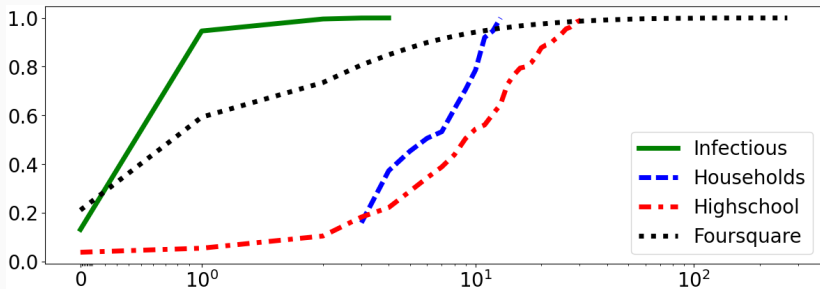
Question

What is a community in a link streams ?

→ Sets of **temporal nodes** that are

- Densely connected inside
- Sparsly connected outside

Importance of vertices

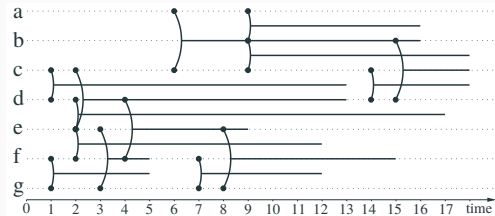


Number of communities each vertex belongs to.

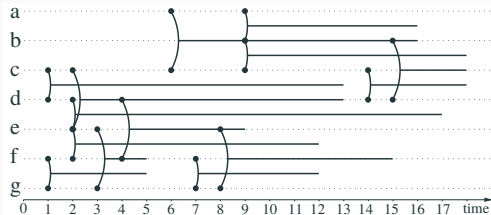
Foursquare dataset

- Nodes very central (Pennsylvania train station, ...)
- Nodes not in any community (offices, ...)

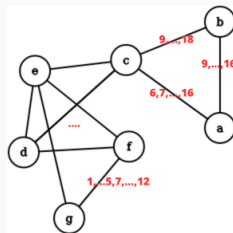
Link stream



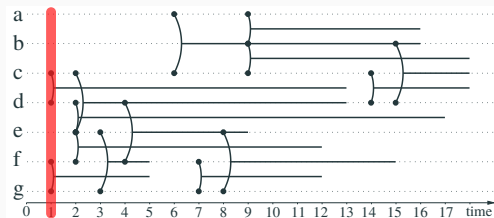
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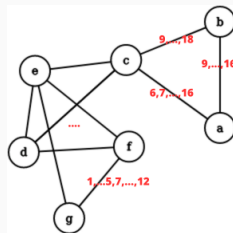
Time varying graphs



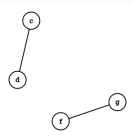
Link stream



Time varying graphs

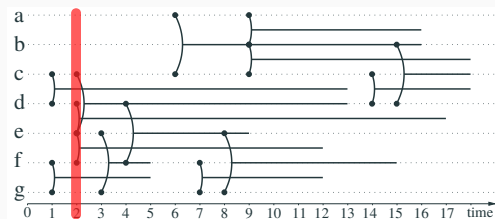


Temporal graphs

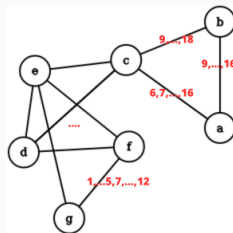


$t = 1$

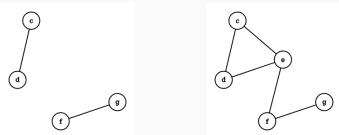
Link stream



Time varying graphs



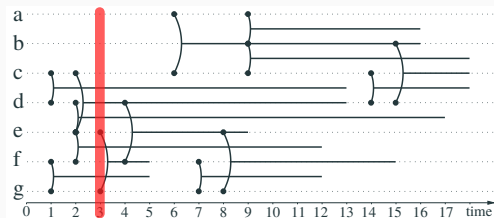
Temporal graphs



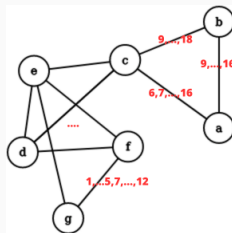
$t = 1 \longrightarrow t = 2$

Supplementary material

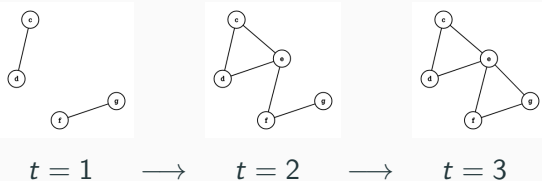
Link stream



Time varying graphs

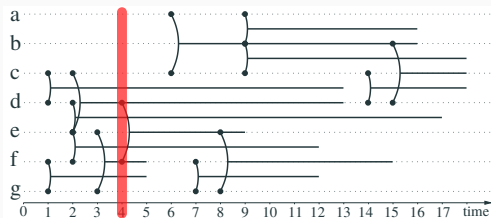


Temporal graphs

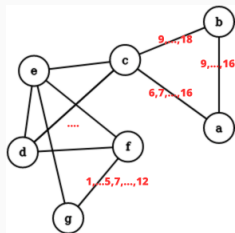


Supplementary material

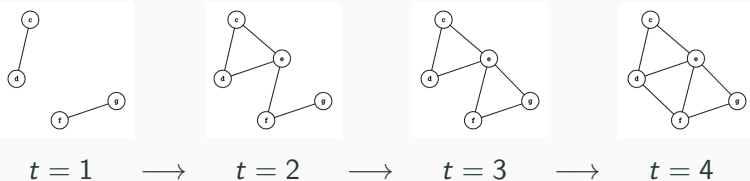
Link stream



Time varying graphs

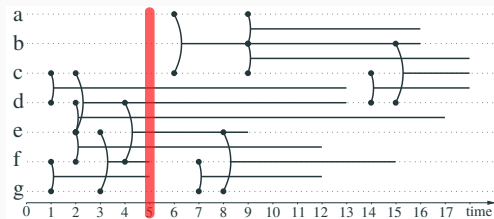


Temporal graphs

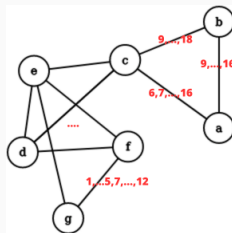


Supplementary material

Link stream



Time varying graphs



Temporal graphs

