

Mining (maximal) span-cores from temporal networks

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Agenda

- 1 Temporal networks
- 2 Background and related work
- 3 Span-core decomposition
- 4 Maximal span-cores
- 5 Experiments
- 6 Applications
- 7 Conclusions

Temporal networks

Temporal networks

- a temporal network is a representation of
 - ▶ **entities** (vertices)
 - ▶ their **relations** (links)
 - ▶ how these relations are **established/broken along time**

Temporal networks

- a temporal network is a representation of
 - ▶ **entities** (vertices)
 - ▶ their **relations** (links)
 - ▶ how these relations are **established/broken along time**
- extracting **dense structures** together with their **temporal span** is a key mining primitive
 - ▶ quantify the transmission opportunities of respiratory infections
 - ▶ identify events and buzzing stories
 - ▶ understand the dynamics of collaboration in successful professional teams

Temporal graphs

Definition

A **temporal graph** is a triplet $G = (V, T, \tau)$, where

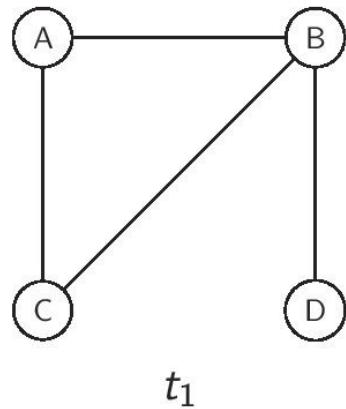
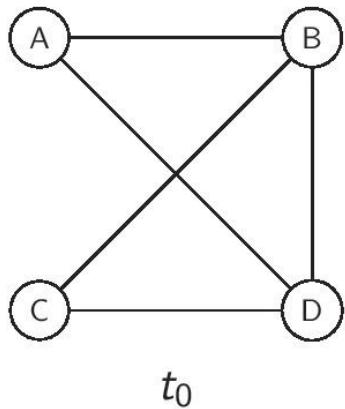
- V is a set of vertices,
- $T = [t_0, t_1, \dots, t_{max}] \subseteq \mathbb{N}$ is a discrete time domain,
- $\tau : V \times V \times T \rightarrow \{0, 1\}$ is a function defining for each $u, v \in V$ and each $t \in T$ whether edge (u, v) exists in t .

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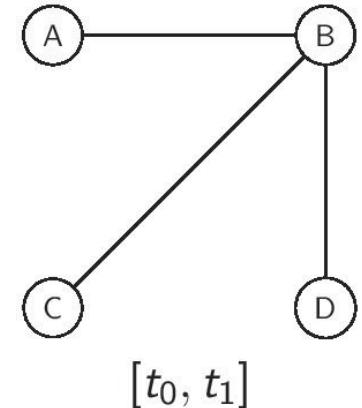
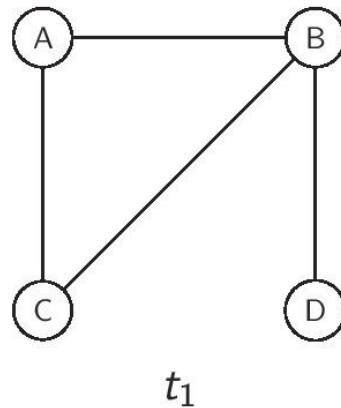
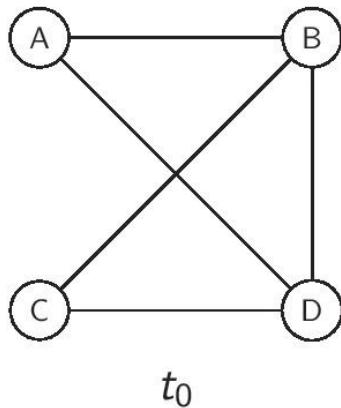


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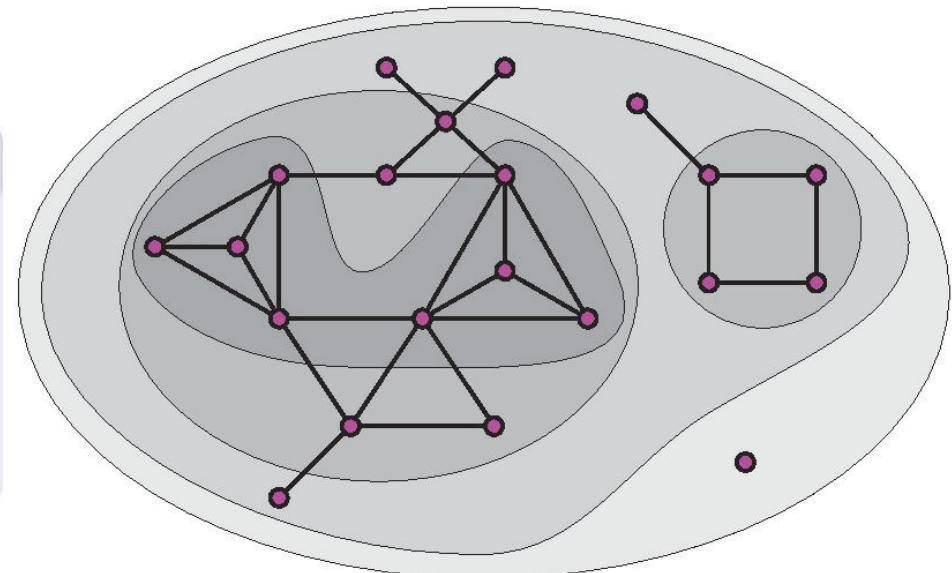
Background and related work

Core decomposition

Definition

The **k-core** (or core of order k) of a simple graph $G = (V, E)$ is a maximal set of vertices $C_k \subseteq V$ such that $\forall u \in C_k : \deg(C_k, u) \geq k$.

The set of all k -cores $V = C_0 \supseteq C_1 \supseteq \dots \supseteq C_{k^*}$ is the **core decomposition** of G .



Core decomposition in multilayer networks [Galimberti *et al.* 2017]

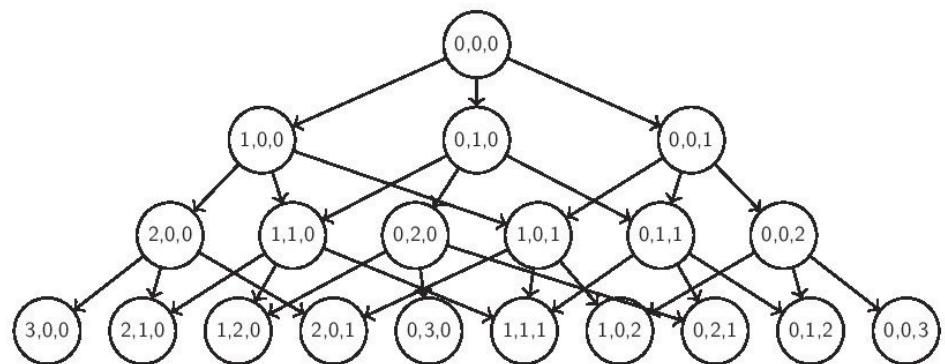
Definition

Given a multilayer graph $G = (V, E, L)$ and an $|L|$ -dimensional integer vector $\vec{k} = [k_\ell]_{\ell \in L}$, the **multilayer \vec{k} -core** of G is a maximal set of vertices $C_{\vec{k}} \subseteq V$ such that $\forall u \in C_{\vec{k}}, \forall \ell \in L : \deg(C_{\vec{k}}, \ell, u) \geq k_\ell$. The set of all \vec{k} -cores is the **multilayer core decomposition** of G .

Core decomposition in multilayer networks [Galimberti et al. 2017]

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- the number of multilayer cores is **exponential** in the number of layer
- layers are not ordered, **there is not sequentiality**

Core decomposition in complex networks

- *Core Decomposition of Uncertain Graphs* [Bonchi et al. 2014]
 - ▶ defined the (k, η) -core as the largest subgraph in which **the probability that every vertex has degree no less than k is greater or equal to η**
- *Core decomposition in large temporal graphs* [Wu et al. 2015]
 - ▶ defined the (k, h) -core as the largest subgraph in which every vertex has at least **k neighbors** and at least **h temporal connections** with each of them
 - ▶ **the sequentiality of connections is not taken into account**: non-contiguous timestamps can support the same core
- *When engagement meets similarity: efficient (k, r) -core computation on social networks* [Zhang et al. 2017]
 - ▶ studied the problem of **enumerating all maximal cores** of a (non-temporal) variant of core decomposition
 - ▶ **the problem is NP-hard**

Span-core decomposition

Span-core decomposition

Definition

The **(k, Δ) -core** of a temporal graph $G = (V, T, \tau)$ is a maximal and non-empty set of vertices $\emptyset \neq C_{k,\Delta} \subseteq V$, such that $\forall u \in C_{k,\Delta} : \deg_{\Delta}(C_{k,\Delta}, u) \geq k$, where $\Delta \sqsubseteq T$ is a temporal interval and $k \in \mathbb{N}^+$.

- $\deg_{\Delta}(C_{k,\Delta}, u)$ represents the **degree of a vertex u in the subgraph induced by $C_{k,\Delta}$ within the temporal interval Δ**

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Problem

Given a temporal graph G , find the set of all (k, Δ) -cores of G .

- the number of span-cores is $\mathcal{O}(|T|^2)$

A naïve approach

Algorithm

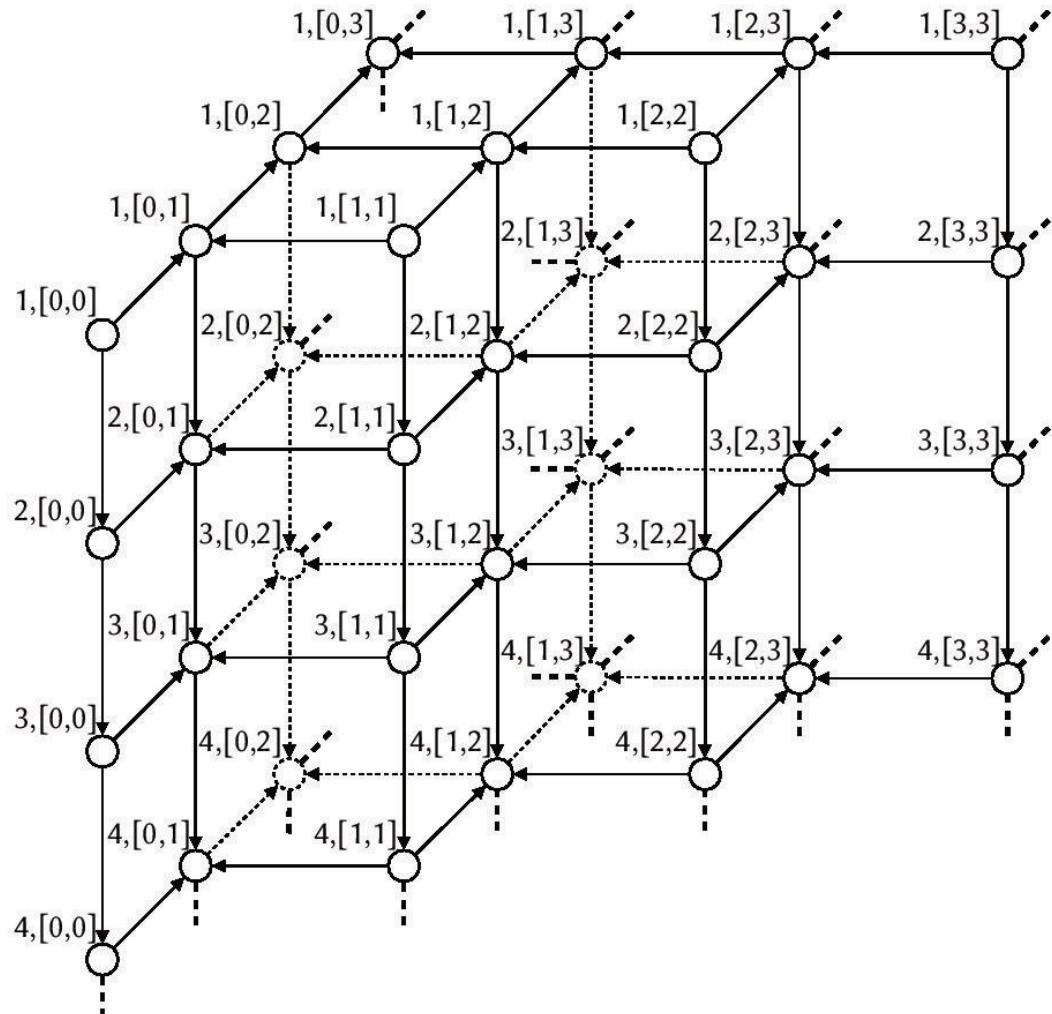
- generate all temporal intervals $\Delta \sqsubseteq T$
- for each $\Delta \sqsubseteq T$, compute the subgraph $G_\Delta = (V, E_\Delta)$
- run a core-decomposition subroutine on each G_Δ

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- the time complexity is $\mathcal{O}(|T|^2 \times |E|)$

Span-core search space



Proposition

For any two span-cores $C_{k,\Delta}, C_{k',\Delta'}$ of a temporal graph G it holds that

$$k' \leq k \wedge \Delta' \sqsubseteq \Delta \Rightarrow C_{k,\Delta} \subseteq C_{k',\Delta'}.$$

Corollary

Given a temporal graph $G = (V, T, \tau)$, and a temporal interval $\Delta = [t_s, t_e] \sqsubseteq T$, let $\Delta_+ = [\min\{t_s + 1, t_e\}, t_e]$ and $\Delta_- = [t_s, \max\{t_e - 1, t_s\}]$. It holds that

$$C_{k,\Delta} \subseteq (C_{k,\Delta_+} \cap C_{k,\Delta_-}) = \bigcap_{\Delta' \sqsubseteq \Delta} C_{k,\Delta'}.$$

A more efficient algorithm

Algorithm

- generate temporal intervals $\Delta \sqsubseteq T$ of **increasing** size
- for each $\Delta \sqsubseteq T$ such that $|\Delta| > 1$, run a core-decomposition subroutine from $(C_{1,\Delta_+} \cap C_{1,\Delta_-})$
- if C_{1,Δ_+} or C_{1,Δ_-} does not exist, skip the core decomposition for Δ

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A span-core $C_{k,\Delta}$ of a temporal graph G is said **maximal** if there does not exist any other span-core $C_{k',\Delta'}$ of G such that $k \leq k'$ and $\Delta \sqsubseteq \Delta'$.

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Problem

Given a temporal graph G , find the set of all maximal (k, Δ) -cores of G .

- the number of maximal span-cores is $\mathcal{O}(|T|^2)$
- experimentally, maximal span-cores are **at least one order of magnitude less** than the overall span-cores

A filtering approach

Algorithm

- equip the algorithm for span-core decomposition with a data structure \mathcal{M} that
 - ▶ stores the span-core of the highest order for every temporal interval $\Delta \sqsubseteq T$
 - ▶ at the storage of a span-core $C_{k,\Delta}$, removes the span-cores dominated by $C_{k,\Delta}$
- return the span-cores retained by \mathcal{M}

Properties of maximal span-cores

Lemma

Given a temporal graph $G = (V, T, \tau)$, let \mathbf{C}_M be the set of all maximal span-cores of G , and $\mathbf{C}_{\text{inner}} = \{C_{k^*}[G_\Delta] \mid \Delta \sqsubseteq T\}$ be the set of innermost cores of all graphs G_Δ . It holds that $\mathbf{C}_M \subseteq \mathbf{C}_{\text{inner}}$.

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Lemma

Given a temporal graph $G = (V, T, \tau)$, and three temporal intervals $\Delta = [t_s, t_e] \sqsubseteq T$, $\Delta' = [t_s - 1, t_e] \sqsubseteq T$, and $\Delta'' = [t_s, t_e + 1] \sqsubseteq T$. The innermost core $C_{k^*}[G_\Delta]$ is a maximal span-core of G if and only if $k^* > \max\{k', k''\}$ where k' and k'' are the orders of the innermost cores of $G_{\Delta'}$ and $G_{\Delta''}$, respectively.

Lemma

Given G , Δ , Δ' , Δ'' , k' , and k'' as in previous Lemma, let $\tilde{V} = \{u \in V \mid \deg_\Delta(V, u) > \max\{k', k''\}\}$, and let $C_{k^*}[G_\Delta[\tilde{V}]]$ be the innermost core of $G_\Delta[\tilde{V}]$. If $k^* > \max\{k', k''\}$, then $C_{k^*}[G_\Delta[\tilde{V}]]$ is a maximal span-core; otherwise, no maximal span-core exists for Δ .

Efficient maximal-span-core finding

Algorithm

- consider intervals $\Delta = [t_s, t_e] \subseteq T$, for increasing values of t_s and decreasing values of t_e
 - ▶ e.g., with $t_{max} = 10$, $\{[0, 10], [0, 9], \dots, [0, 0], [1, 10], [1, 9], \dots, [1, 1], [2, 10], [2, 9], \dots\}$
 - ▶ this guarantees that once we consider Δ , no $\Delta' \sqsupseteq \Delta$ will be considered at later stage
- compute the **lower bound lb on the order** of a span-core in Δ to be recognized as maximal
- build the sets of vertices V_{lb} that have degree in Δ larger than lb
- extract the **innermost** core of the subgraph $(V_{lb}, E_\Delta[V_{lb}])$
- identify such a core as maximal if its order is actually larger than lb

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- identify such a core as maximal if its order is actually larger than lb
- the time complexity is still $\mathcal{O}(|T|^2 \times |E|)$

Experiments

Datasets

dataset	$ V $	$ E $	$ T $	window size (days)	domain
ProsperLoans	89k	3M	307	7	economic
Last.fm	992	4M	77	21	co-listening
WikiTalk	2M	10M	192	28	communication
DBLP	1M	11M	80	366	co-authorship
StackOverflow	2M	16M	51	56	question-and-answer
Wikipedia	343k	18M	101	56	co-editing
Amazon	2M	22M	115	28	co-rating
Epinions	120k	33M	25	21	co-rating

Evaluation

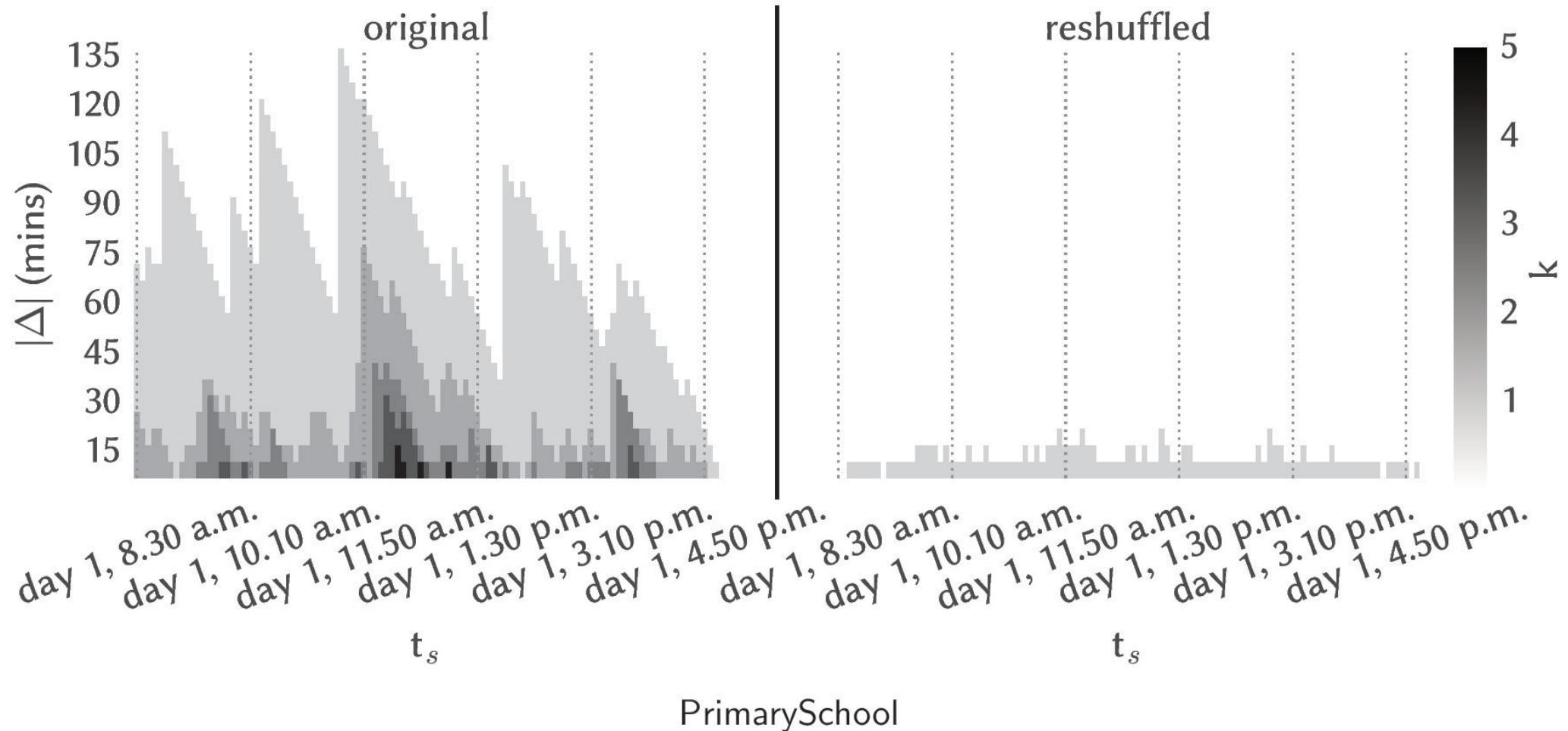
dataset	method	# output span-cores	time (s)	memory (GB)	# processed vertices
WikiTalk	Naïve-span-cores	19 693	322 302	36	25B
	Span-cores		1 084	36	555M
	Naïve-maximal-span-cores	632	1 194	36	555M
	Maximal-span-cores		126	35	2M
Wikipedia	Naïve-span-cores	125 191	17 155	4	1B
	Span-cores		522	4	35M
	Naïve-maximal-span-cores	2 147	537	4	35M
	Maximal-span-cores		201	4	320k
Amazon	Naïve-span-cores	29 318	10 415	18	2B
	Span-cores		409	18	247M
	Naïve-maximal-span-cores	303	580	18	247M
	Maximal-span-cores		123	18	688k

Applications

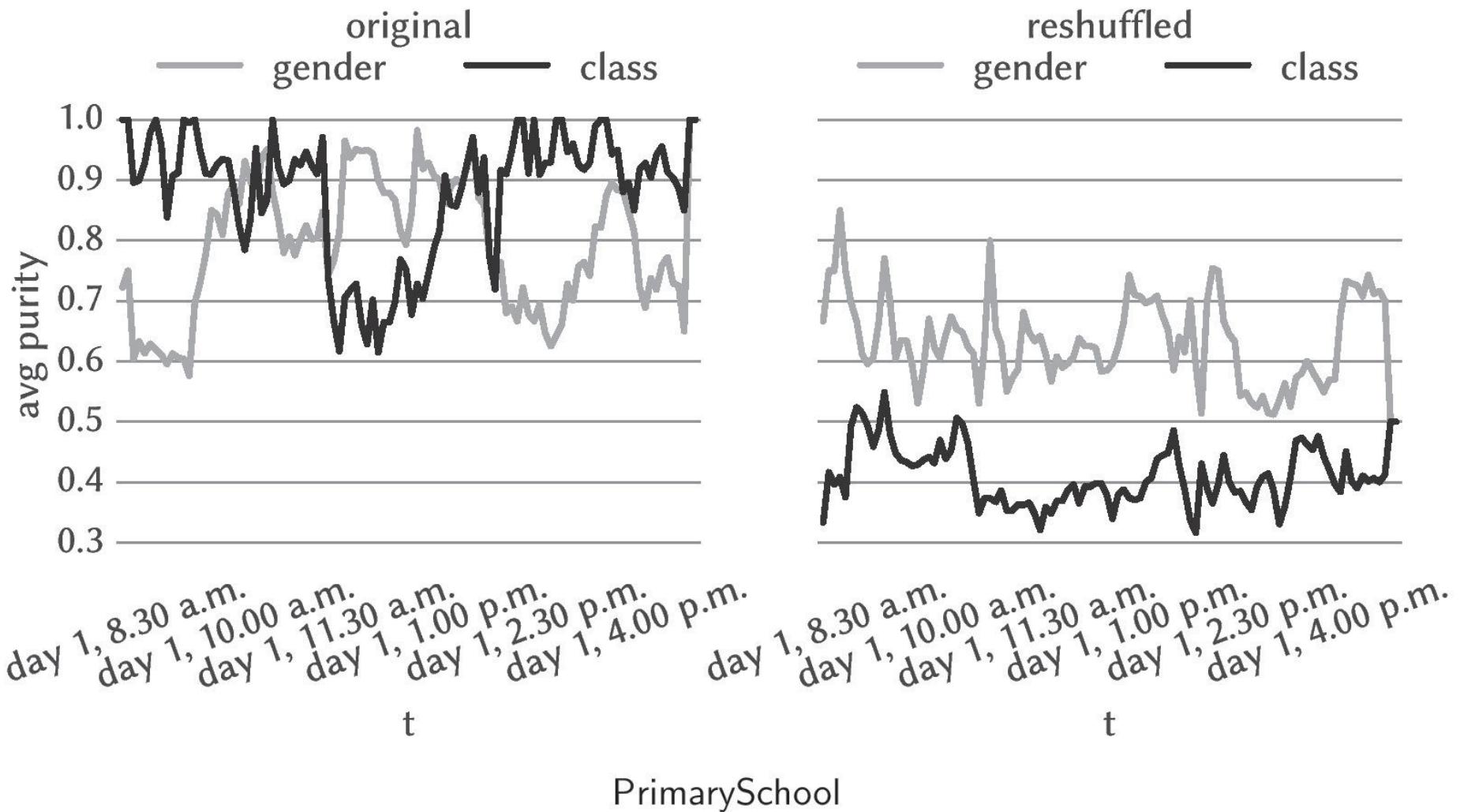
Datasets

- **face-to-face interaction networks** gathered by a proximity-sensing infrastructure in schools
 - ▶ PrimarySchool (242 individuals, 2 days)
 - ▶ HighSchool (327 individuals, 5 days)
 - ▶ HongKong (774 individuals, 11 days)
- window size of 5 minutes
- discarded span-cores of $|\Delta| = 1$

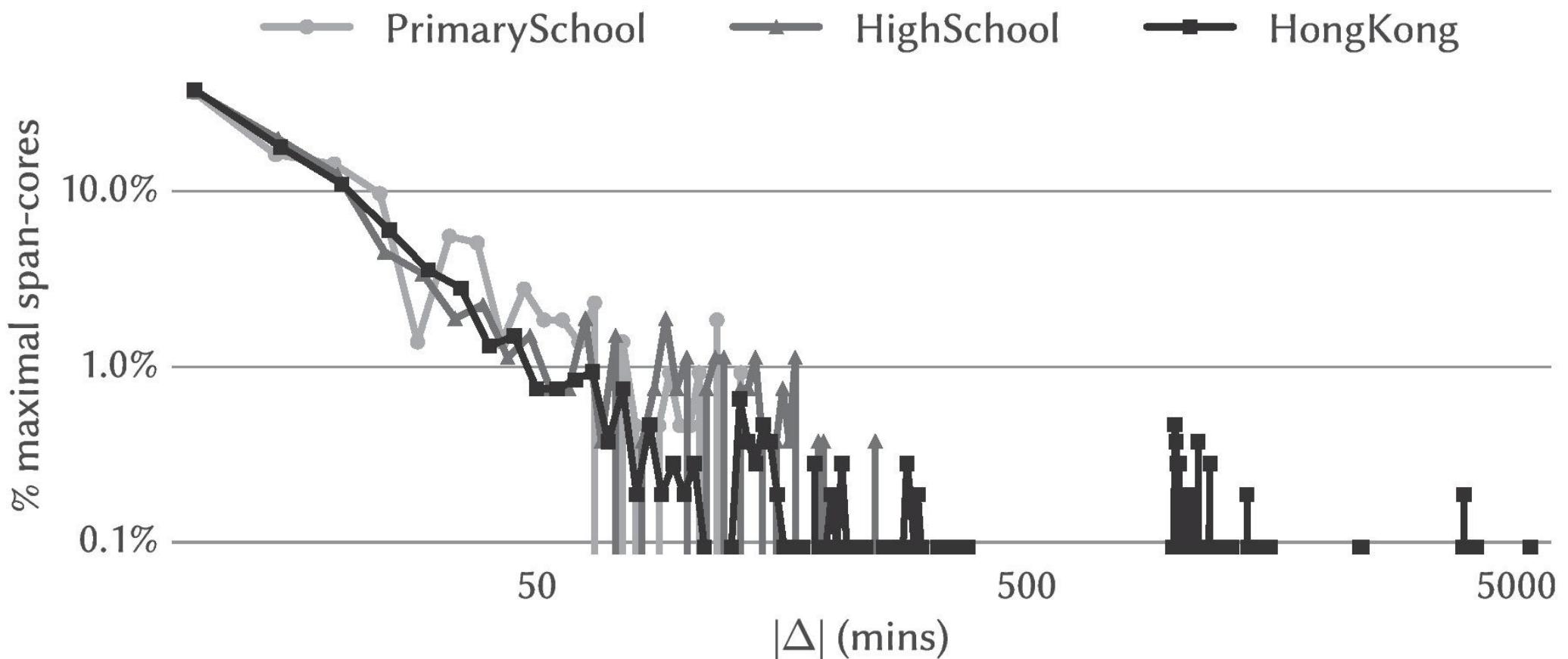
Social activities of groups of students



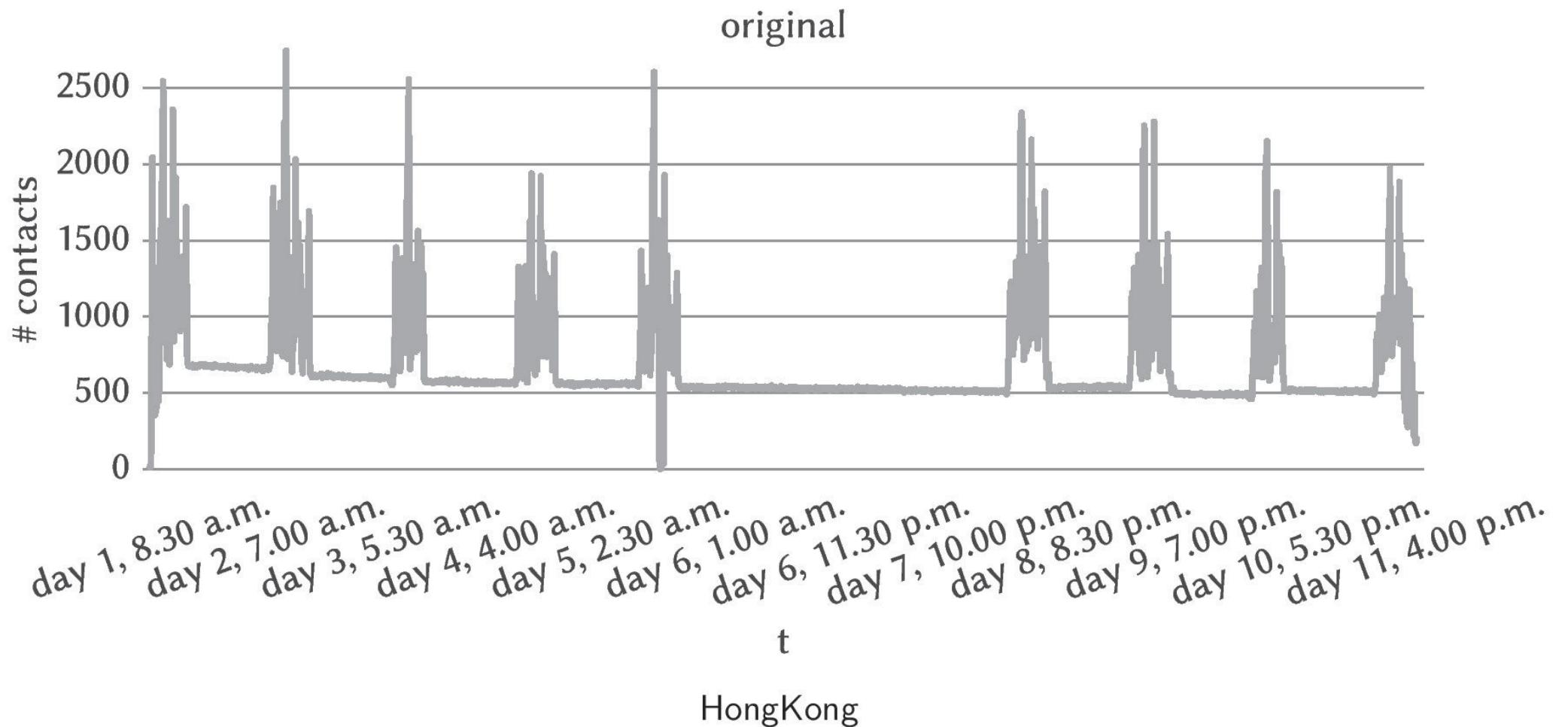
Mixing of gender and class



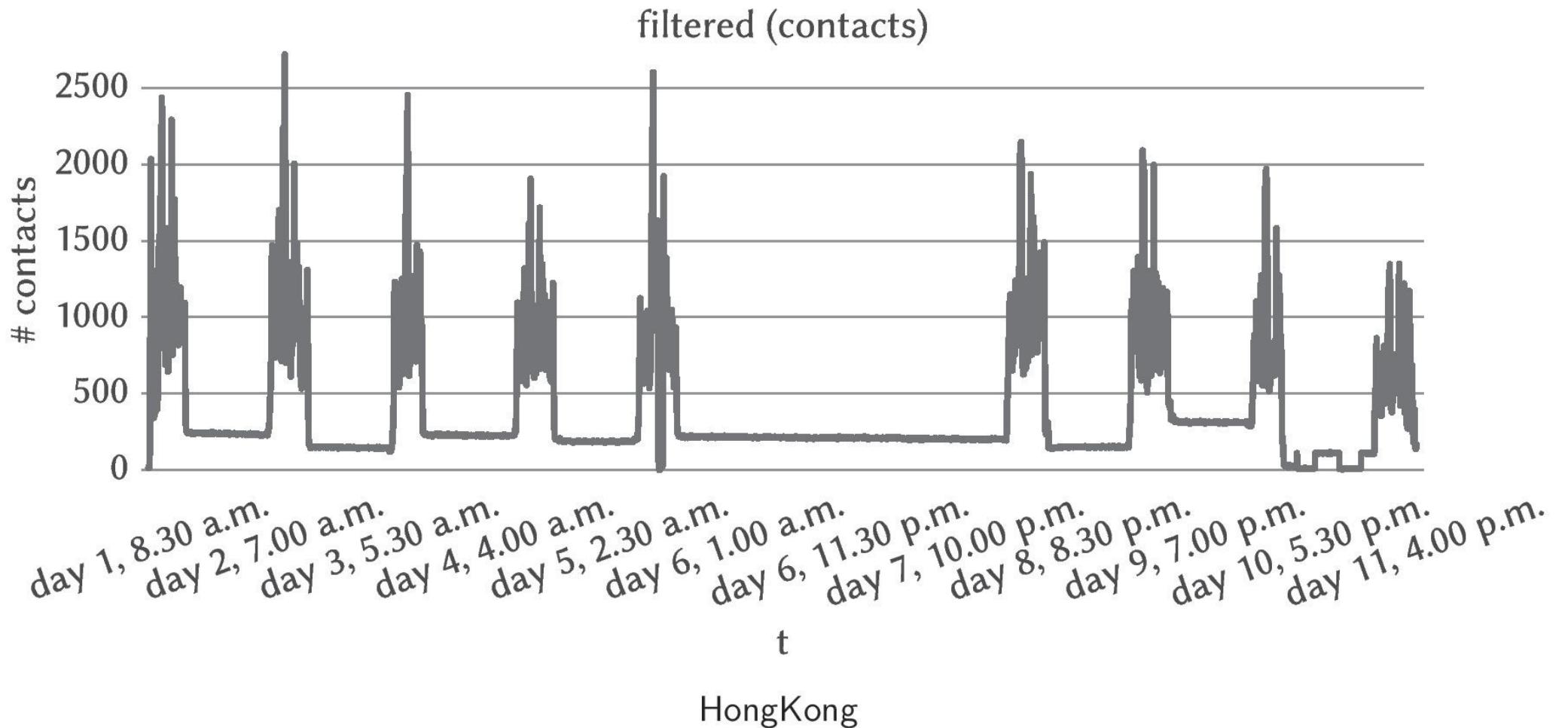
Length of social interactions in groups



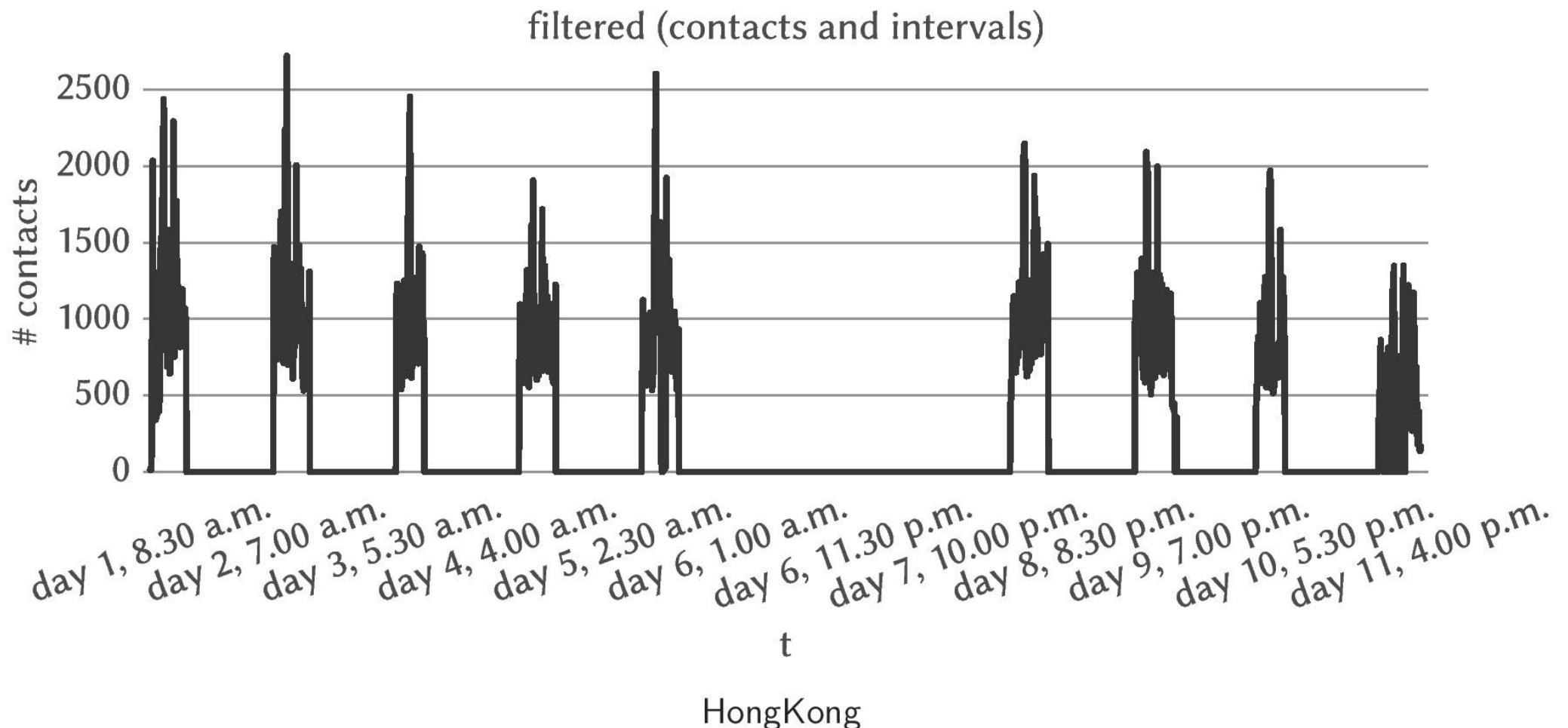
Anomaly detection



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Conclusions

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- introduced a notion of dense pattern in temporal networks that
 - ▶ takes into account the **sequentiality** of connections
 - ▶ is assigned with a clear **temporal collocation**
- developed efficient algorithms for computing all the span-cores, and only the maximal ones
- future work:
 - ▶ spreading processes analysis
 - ▶ temporal community search and temporal densest subgraph
 - ▶ network finger-printing

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