# 18th International Symposium on Parameterized and Exact Computation 

IPEC 2023, September 6-8, 2023, Amsterdam, The Netherlands

Edited by

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## Preface

The International Symposium on Parameterized and Exact Computation (IPEC, formerly IWPEC) is a series of international symposia covering research in all aspects of parameterized and exact algorithms and complexity. It started in 2004 as a biennial workshop and became an annual event in 2009. Previous iterations of the symposium were:

2004 Bergen, Norway

- 2006 Zürich, Switzerland
- 2008 Victoria, Canada
- 2009 Copenhagen, Denmark
- 2010 Chennai, India
- 2011 Saarbrücken, Germany
- 2012 Lubljana, Slovenia
- 2013 Sophia Antipolis, France
- 2014 Wrocław, Poland
- 2015 Patras, Greece
- 2016 Aarhus, Denmark
- 2017 Vienna, Austria
- 2018 Helsinki, Finland
- 2019 Munich, Germany
- 2020 virtual / Hong Kong, China
- 2021 virtual / Lisbon, Portugal
- 2022 Potsdam, Germany

This volume contains the papers presented at IPEC 2023: the 18th International Symposium on Parameterized and Exact Computation. IPEC 2023 was held on September 6-8 (Wed to Fri) as part of ALGO 2023, and took place in Amsterdam, the Netherlands at Centrum Wiskunde \& Informatica (CWI). In response to the call for papers, 85 extended abstracts were registered, of which 10 were withdrawn or otherwise failed to submit a full version. The resulting number of 75 full submissions represents a significant increase in interest in the conference compared to previous years. 34 papers were ultimately selected for presentation at the conference and inclusion in these proceedings. The reviews were performed in a double-blind fashion, and there were 106 external reviews out of a total of 223 reviews.

The Best Paper Award was given to Hans L. Bodlaender (Utrecht University), Isja Mannens (Utrecht University), Jelle Oostveen (Utrecht University), Sukanya Pandey (Utrecht University) and Erik Jan van Leeuwen (Utrecht University) for their paper "The Parameterised Complexity of Integer Multicommodity Flow". The Best Student Paper Award was given to Stefan Kratsch (Humboldt-Universität zu Berlin) and Pascal Kunz (Humboldt-Universität zu Berlin) for their paper "Approximate Turing kernelization and lower bounds for domination problems". The EATCS-IPEC Nerode Prize was given to Marek Cygan (University of Warsaw and Nomagic), Jesper Nederlof (Utrecht University), Marcin Pilipczuk (University of Warsaw), Michał Pilipczuk (University of Warsaw), Johan M. M. van Rooij (Utrecht University) and Jakub Onufry Wojtaszczyk (Google) for their paper "Solving Connectivity Problems Parameterized by Treewidth in Single Exponential Time". IPEC 2023 hosted an award ceremony with a talk given jointly by Michał Pilipczuk and Johan M. M. van Rooij. The Nerode Prize committee consisted of Fedor Fomin (chair; University of Bergen), Thore Husfeldt (IT University of Copenhagen) and Sang-il Oum (Korea Advanced Institute of Science and Technology). Tuukka Korhonen (University of Bergen) presented an invited tutorial on "New methods in FPT algorithms for treewidth". Finally, IPEC 2023 hosted the award ceremony of the eighth Parameterized Algorithms and Computational Experiments (PACE) challenge. These proceedings contain a report on the PACE 2023 challenge and brief communications of the winners about their solvers.

We thank the program committee and the external reviewers for their commitment in the paper selection process. We also thank all the authors who submitted their work. We are grateful to the local organizers of ALGO 2023 for the local arrangements.

Neeldhara Misra and Magnus Wahlström
Gandhinagar and London, October 2022

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[^0]
# An FPT Algorithm for Temporal Graph Untangling 

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#### Abstract

Several classical combinatorial problems have been considered and analysed on temporal graphs. Recently, a variant of Vertex Cover on temporal graphs, called MinTimelineCover, has been introduced to summarize timeline activities in social networks. The problem asks to cover every temporal edge while minimizing the total span of the vertices (where the span of a vertex is the length of the timestamp interval it must remain active in). While the problem has been shown to be NP-hard even in very restricted cases, its parameterized complexity has not been fully understood. The problem is known to be in FPT under the span parameter only for graphs with two timestamps, but the parameterized complexity for the general case is open. We settle this open problem by giving an FPT algorithm that is based on a combination of iterative compression and a reduction to the Digraph Pair Cut problem, a powerful problem that has received significant attention recently.


2012 ACM Subject Classification Theory of computation $\rightarrow$ Parameterized complexity and exact algorithms; Theory of computation $\rightarrow$ Graph algorithms analysis; Mathematics of computing $\rightarrow$ Graph theory; Theory of computation $\rightarrow$ Design and analysis of algorithms

Keywords and phrases Temporal Graphs, Vertex Cover, Graph Algorithms, Parameterized Complexity

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## 1 Introduction

Temporal graphs are emerging as one of the main models to describe the dynamics of complex networks. They describe how relations (edges) change in a discrete time domain $[12,11]$, while the vertex set is not changing. The development of algorithms on temporal graphs has mostly focused on finding paths or walks and on analyzing graph connectivity [12, 20, 21, 7, $22,8,3,17,1,5]$. However, several classical problems in computer science have been recently extended to temporal graphs and one of the most relevant problems in graph theory and theoretical computer science, Vertex Cover, has been considered in this context [2, 10, 19].

In particular, here we study a variant of Vertex Cover, called Network Untangling, introduced in [19]. Network Untangling has applications in discovering event timelines and summarizing temporal networks. It considers a sequence of temporal interactions between entities (e.g. discussions between users in a social network) and aims to explain the observed interactions with few (and short) activity intervals of entities, such that each interaction is covered by at least one of the two entities involved (i.e. at least one of the two entities is active when an interaction between them is observed).

Network Untangling can be seen as a variant of Vertex Cover, where we search for a minimum cover of the interactions, called temporal edges. The size of this temporal vertex cover is based on the definition of span of a vertex, that is the length of vertex activity. In particular, the span of a vertex is defined as the difference between the maximum and minimum timestamp where the vertex is active. Hence, if a vertex is active in exactly one

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timestamp, it has a span equal to 0 . This models the idea that each vertex is present in the network because we know that they interacted at least once, but that sustained periods of interaction are relatively rare.

Four combinatorial formulations of Network Untangling have been defined in [19], varying the definition of vertex activity (a single interval or $h \geq 2$ intervals) and the objective function (minimization of the sum of vertex spans or minimization of the maximum vertex span). Here we consider the formulation, denoted by MinTimelineCover, where vertex activity is defined as a single interval and the objective function is the minimization of the sum of vertex spans. Hence, given a temporal graph, MinTimelineCover asks for a cover of the temporal edges that has minimum span and such that each vertex is active in one time interval.

We focus on this specific problem, since it is not known to be FPT or not, while the variant of the problem where vertex activity is defined as two intervals is known to be NP-hard when the span is equal to 0 [9]. Hence it is unlikely that this problem variant admits an FPT algorithm for parameter the span. The MinTimelineCover problem is known to be NP-hard also in very restricted cases, when each timestamp contains at most one temporal edge [4], when each vertex has at most two incident temporal edges in each timestamp and the temporal graph is defined over three timestamps [4], and when the temporal graph is defined over two timestamps [9]. MinTimelineCover is also known to be approximable within factor $O(T \log n)$, where $n$ is the number of vertices and $T$ is the number of timestamps of the temporal graph [6]. Note that, since the span of a vertex activity in exactly one timestamp is equal to 0 , MinTimelineCover is trivially in P when the temporal graph is defined on a single timestamp, since in this case any solution of the problem has span 0 . Furthermore, deciding whether there exists a solution of MinTimelineCover that has span equal to 0 can be decided in polynomial time via a reduction to 2-SAT [19].

MinTimelineCover has been considered also in the parameterized complexity framework. The definition of span leads to a problem where the algorithmic approaches applied to Vertex Cover cannot be easily extended for the parameter span of the solution. Indeed, in Vertex Cover for each edge we are sure that at least one of the endpoints must be included in the solution, thus at least one of the vertices contributes to the cost of the solution. This leads to the textbook FPT algorithm of branching over the endpoints of any edge. For MinTimelineCover, a vertex with span 0 may cover a temporal edge, as the vertex can be active only in the timestamp where the temporal edge is defined. This makes it more challenging to design FPT algorithms when the parameter is the span of the solution. In this case, MinTimelineCover is known to admit a parameterized algorithm only when the input temporal graph is defined over two timestamps [9], with a parameterized reduction to the Almost 2-SAT problem. However, the parameterized complexity of MinTimelineCover for the span parameter on general instances has been left open [9, 4]. The authors of [9] have also analyzed the parameterized complexity of the variants of Network Untangling proposed in [19], considering other parameters in addition to the span of the solution: the number of vertices of the temporal graph, the length of the time domain, and the number of intervals of vertex activity.

Our contributions. We solve the open question on the parameterized complexity of MinTimelineCover by showing that the problem is FPT in parameter $k$, the span of a solution, even if the number of timestamps is unbounded. Our algorithm takes time $O^{*}\left(2^{5 k \log k}\right)$, where the $O^{*}$ notation hides polynomial factors. Our algorithm is divided into two phases, each using a different technique. First, given a temporal graph $G$, we use a variant of
iterative compression, where we start from a solution $S$ of span at most $k$ on a subgraph of $G$ induced by a subset of vertices (taken across all timestamps), and then try to maintain such a solution after adding a new vertex of $G$ to the graph under consideration. This requires us to reorganize which vertices involved in $S$ should be in the solution or not, and in which timestamps. One challenge is that since the number of such timestamps is unbounded, there are too many ways to choose how to include or not include the vertices that are involved in $S$. We introduce the notion of a feasible assignment, which allows us to compute how the vertices in $S$ can be reorganized (see Def. 8 for the formal definition). There are only $2^{O(k \log k)}$ ways of reorganizing the vertices in $S$. We try each such feasible assignments $X$, and we must then find a temporal cover of the whole graph $G$ that "agrees" with $X$.

This leads to the second phase of the algorithm, which decides if such an agreement cover exists through a reduction to a variant of a problem called Digraph Pair Cut. In this problem, we receive a directed graph and forbidden pairs of vertices, and we must delete at most $k$ arcs so that a specified source vertex does not reach both vertices from a forbidden pair. It is known that the problem can be solved in time $O^{*}\left(2^{k}\right)$. In this work, we need a version where the input specifies a set of deletable and undeletable arcs, which we call Constrained Digraph Pair Cut. The Digraph Pair Cut problem and its variants have played an important role in devising randomized kernels using matroids [16] and, more recently, in establishing a dichotomy in the complexity landscape of constraint satisfaction problems $[13,15]$. Here, the problem is useful since it can model the implications of including a vertex in the solution or not and, in a more challenging way, allows implementing the notion of cost using our definition of span. We hope that the techniques developed for this reduction can be useful for other variants of temporal graph cover.

Overview of the algorithm. Our approach is loosely inspired by some ideas from the FPT algorithm for two timestamps, which is a reduction to Almost 2-SAT [9]. In the latter, one is given a set of clauses with at most two variables each and must delete a minimum number of clauses so that those remaining are satisfiable. We do not use Almost 2-SAT directly, but its usage for two timestamps may help understand the origins of our techniques and the relevance of our reduction to Digraph Pair Cut.

The reduction from MinTimelineCover on two timestamps to Almost 2-SAT associates each vertex $v_{i}$ with a variable $x\left(v_{i}\right)$, which is true when one should include $v_{i}$ in a temporal cover and false otherwise; each edge $u_{i} v_{i}$ is associated with a clause $x\left(u_{i}\right) \vee x\left(v_{i}\right)$ (here, $v_{i}$ represents the occurrence of vertex $v$ at timestamp $i \in\{1,2\}$ ). This corresponds to enforcing the inclusion of $u_{i}$ or $v_{i}$ in our vertex cover, and we can include enough copies of this clause to make it undeletable. Since our goal is to minimize the number of base vertices $v$ with both $v_{1}$ and $v_{2}$ in the cover, we also add a clause $\neg x\left(v_{1}\right) \vee \neg x\left(v_{2}\right)$. Then there is a temporal cover of $G$ of span at most $k$ if and only if one can delete at most $k$ clauses of the latter form to make all remaining clauses satisfiable.

For $T \geq 3$ timestamps, the clauses of the form $x\left(u_{i}\right) \vee x\left(v_{i}\right)$ can still be used to model the vertex cover requirements, but there seems to be no obvious way to model the span of a cover. One would need to devise a set of clauses of size two such that choosing an interval of $t$ vertices in a cover corresponds to deleting $t-1$ negative clauses. Our idea is to extend current FPT algorithms for Almost 2-SAT to accommodate our cost function. In [18], the authors propose an iterative compression FPT algorithm that starts from a solution that deletes $k+1$ clauses, and modifies it into a solution with $k$ clauses, if possible. The algorithm relies on several clever, but complicated properties of the dependency graph of the clauses (in which vertices are literals and arcs are implications implied by the clauses). This
algorithm seems difficult to adapt to our problem. To our knowledge, the only other FPT algorithm for Almost 2-SAT is that of [16]. The algorithm of [16] employs a parameterized reduction to Digraph Pair Cut. At a high level, the idea is to start from an initial guess of assignment for a well-chosen subset of variables, then to construct the dependency graph of the clauses. A certain chain of implications is enforced by our initial guess, the vertex pairs to separate correspond to contradictory literals, and deleting arcs corresponds to deleting clauses. It turns out that, with some work, we can skip the Almost 2-SAT formulation and reduce MinTimelineCover to (a variant of) Directed Pair Cut directly by borrowing some ideas from this reduction. This is not immediate though. The first challenge is that the aforementioned "well-chosen initial guess" idea cannot be used in our context, and we must develop new tools to enumerate a bounded number of initial guesses from a partial solution (which we call feasible assignment). The second challenge is that our reduction to our variant of Directed Pair Cut needs a specific gadget to enforce our cost scheme, while remaining consistent with the idea of modeling the dependency graph of the SAT instance corresponding to the vertex cover problem at hand.

Some of the proofs are omitted due to page limit.

## 2 Preliminaries

For an integer $n$, we denote $[n]=\{1, \ldots, n\}$ and for two integers $i, j$, we denote $[i, j]=$ $\{i, i+1, \ldots, j-1, j\}$ (which is the empty st if $i>j$ ). Temporal graphs are defined over a discrete time domain $\mathcal{T}$, which is a sequence $1,2 \ldots, T$ of timestamps. A temporal graph is also defined over a set of vertices, called base vertices, that do not change in the time domain and are defined in all timestamps, and are associated with vertices, which are base vertices defined in specific timestamps. We use subscripts to denote the timestamp to which a vertex belongs to, so, for a base vertex $v$ and $t \in[T]$, we use $v_{t}$ to denote the occurrence of $v$ in timestamp $t$. A temporal edge connects two vertices, associated with distinct base vertices, that belong to the same timestamp.

- Definition 1. A temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$ consists of

1. A time domain $\mathcal{T}=\{1,2 \ldots, T\}$;
2. A set $V_{B}$ of base vertices; $V_{B}$ has a corresponding set $V(G)$ of vertices, which consists of base vertices in specific timestamps, defined as follows:
$V(G)=\left\{v_{t}: v \in V_{B} \wedge t \in[T]\right\}$.
3. A set $E=E(G)$ of temporal edges, which satisfies:

$$
E \subseteq\left\{u_{t} v_{t}: u, v \in V_{B}, t \in[T] \wedge u \neq v\right\}
$$

For a directed (static) graph $H$, we denote by $(u, v)$ an arc from vertex $u$ to vertex $v$ (we consider only directed static graphs, not directed temporal graphs).

Given a temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$ and a set of base vertices $B \subseteq V_{B}$, we define the set $\tau(B)$ of all vertices of $B$ across all times:

$$
\tau(B)=\left\{v_{t}: v \in B \wedge t \in[T]\right\} .
$$

If $B=\{v\}$, we may write $\tau(v)$ instead of $\tau(\{v\})$.
For a subset $W_{B} \subseteq V_{B}$ of base vertices, we denote by $G\left[W_{B}\right]$ the subgraph induced by $\tau\left(W_{B}\right)$, that is, the graph whose vertex set is $\tau\left(W_{B}\right)$ and whose edge set is $\left\{u_{t} v_{t} \in E\right.$ : $\left.u_{t}, v_{t} \in \tau\left(W_{B}\right)\right\}$. We also use the notation $G-W_{B}=G\left[V_{B} \backslash W_{B}\right]$. Observe that $G\left[W_{B}\right]$ and $G-W_{B}$ are temporal graphs over the same time domain as $G$.

In order to define the problem we are interested in, we need to define the assignment of a set of base vertices.

- Definition 2. Consider a temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$ and a set $W_{B} \subseteq V_{B}$ of base vertices. An assignment of $W_{B}$ is a subset $X \subseteq \tau\left(W_{B}\right)$ such that if $u_{p} \in X$ and $u_{q} \in X$, with $p, q \in[T]$, then $u_{t} \in X$, for each $t \in[p, q]$. For a base vertex $u \in W_{B}$ such that there exists $t \in[T]$ with $u_{t} \in X$, we denote by $\delta(u, X), \Delta(u, X)$, respectively, the minimum and maximum timestamp, respectively, such that $u_{\delta(u, X)}, u_{\Delta(u, X)} \in X$. If $u_{t}$ does not exist, then $\delta(u, X)=\Delta(u, X)=0$.

If $W_{B}$ is clear from the context or not relevant, then we may say that $X$ is an assignment, without specifying $W_{B}$. Note that, given an assignment $X$ and a set $\tau(v)$, for some $v \in V_{B}$, then $X \cap \tau(v)=\left\{v_{t}: v_{t} \in X \wedge v_{t} \in \tau(v)\right\}$ contains vertices for $v$ that belong to a contiguous interval of timestamps. Consider a set $I \subseteq[T]$ of timestamps. An assignment $X$ intersects $I$ if there exists $v_{t} \in X$ such that $t \in I$.

Now, we give the definition of temporal cover.

- Definition 3. Given a temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$ a temporal cover of $G$ is an assignment $X$ of $V_{B}$ such that the following properties hold:

1. For each $v \in V_{B}$ there exists at least one $v_{t} \in X$, for some $t \in \mathcal{T}$.
2. For each $u_{t} v_{t} \in E$, with $t \in[T]$, at least one of $u_{t}$, $v_{t}$ is in $X$.

For a temporal cover $X$ of $G$, the span of $v$ in $X$ is defined as: $s p(v, X)=\Delta(v, X)-\delta(v, X)$. Note that if a temporal cover $X$ contains, for a base vertex $v \in V_{B}$, a single vertex $v_{t}$, then $s p(v, X)=0$. The span of $X$, denoted by $s p(X)$, is then defined as:

$$
s p(X)=\sum_{v \in V_{B}} s p(v, X)
$$

The definition of temporal cover requires that for each base vertex at least one of its associated vertices belongs to the cover. This is not strictly necessary, since it might be possible to cover every temporal edge without this condition. However, this condition simplifies some of the definitions and proofs below. Note that if an assignment of a base vertex is not needed to cover temporal edges, we can assign the vertex to some timestamp without increasing the span.

Now, we are able to define MinTimelineCover (an example is presented in Fig. 1).

- Problem 4. (MinTimelineCover)

Input: A temporal graph $G=\left(V_{B}, \mathcal{T}, E\right)$, an integer $k$.
Question: Does there exist a temporal cover of $G$ of span at most $k$ ?
A temporal cover $S \subseteq V(G)$ of span at most $k$ will sometimes be called a solution. Our goal is to determine whether MinTimelineCover is FPT in parameter $k$.

## 3 An FPT Algorithm

In this section we present our FPT algorithm, which consists of two parts:

1. The iterative compression technique.
2. A reduction to the Constrained Digraph Pair Cut problem.

Before presenting the main steps of our algorithm, we present the main idea and some definitions. Recall that our parameter, that is the span of a solution of MinTimelineCover, is denoted by $k$.

Consider a temporal graph $G$ and assume we have a temporal cover $S$ of span at most $k$ of the subgraph $G-\{w\}$, for some base vertex $w \in V_{B}$. The idea of the iterative compression step is, starting from $S$, to show how to decide in FPT time whether there


Figure 1 An example of MinTimelineCover on a temporal graph $G$ consisting of four base vertices and six timestamps. For each timestamp, we draw the temporal edges of $G$, for example for $t=2$, the temporal edges are $v_{2} u_{2}, v_{2} w_{2}, u_{2} w_{2}, z_{2} w_{2}$. Also note that in $t=1$ and $t=6$ no temporal edge is defined. A temporal cover $X=\left\{v_{5}, u_{2}, u_{3}, u_{4}, z_{3}, z_{4}, w_{2}\right\}$ is represented with grey rectangles. Note that $\delta(v, X)=\Delta(v, X)=5, \delta(u, X)=2, \Delta(u, X)=4, \delta(z, X)=3, \Delta(z, X)=4$, $\delta(w, X)=\Delta(w, X)=2$. It follows that $s p(X)=3$.
exists a solution of MinTimelineCover for $G$. This is done by solving a subproblem, called Restricted Timeline Cover, where we must modify $S$ to consider $w$. A solution to this subproblem is computed by branching on the assignments of base vertices having a positive span in $S$ and on $w$, and then reducing the problem to Constrained Digraph Pair Cut. Restricted Timeline Cover is defined as follows.

- Problem 5. (Restricted Timeline Cover)

Input: A temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$, a vertex $w \in V_{B}$, an integer $k$, a temporal cover $S$ of $G-\{w\}$ of span at most $k$.
Output: Does there exist a temporal cover of $G$ of span at most $k$ ?
For technical reasons that will become apparent later, we will assume that the temporal graph contains no edge at timestamps 1 and $T$, i.e. for every $u_{t} v_{t} \in E$, we have $t \in[2, T-1]$ (as in Fig. 1). In particular, this avoids us to consider different gadget definitions in the reduction to Constrained Digraph Pair Cut, as the cases where a base vertex is assigned the first or the last of its associated vertex behaves somehow differently. It is easy to see that if this is not already the case, we can add two such "dummy" timestamps, where $G$ does not contain any temporal edge. Indeed, since there are no temporal edges in these two timestamps, then $G$ has a temporal cover of span at most $k$ if and only if the same graph with dummy timestamps has a temporal cover of span at most $k$.

Informally, if we are able to solve Restricted Timeline Cover in FPT time, then we can obtain an FPT algorithm for MinTimelineCover as well. Indeed, we can first compute a temporal cover on a small subset of base vertices (for example a single vertex), and then we can add, one at a time, the other vertices of the graph. This requires at most $\left|V_{B}\right|$ iterations, and each time a vertex is added, we compute a solution of Restricted Timeline Cover to check whether it is possible to find a temporal cover of span at most $k$ after the addition of a vertex.

## Iterative Compression

We now present our approach based on iterative compression to solve the RESTRICTED Timeline Cover problem. Given a solution $S$ for $G-\{w\}$, we focus on the vertices of $V_{B}$ that have a positive span in $S$ and vertex $w$. An example of our approach, that illustrates the sets of base vertices and vertices used by the algorithm, is presented in Fig. 2.

Consider the input of Restricted Timeline Cover that consists of a temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$, a vertex $w \in V_{B}$, and a temporal cover $S$ of $G-\{w\}$ of span at most $k$. Define the following sets associated with $S$ :

$$
\begin{aligned}
V_{S} & =\left\{v \in V_{B}: \exists p, q \in[T], p<q, \text { such that } v_{p}, v_{q} \in S\right\} \cup\{w\} \\
V_{S}^{\prime} & =\left\{v_{t}: v_{t} \in S, v \in V_{S} \backslash\{w\}\right\} \cup\left\{w_{t}: t \in[T]\right\} .
\end{aligned}
$$

The set $V_{S}$ is defined as the set of base vertices having span greater than 0 in $S$, plus the vertex $w$. $V_{S}^{\prime}$ contains the vertices in $V(G)$ associated with $V_{S}$, in particular: (1) the vertices corresponding to the base vertices in $V_{S} \backslash\{w\}$ that are included in $S$ and (2) vertices corresponding to the base vertex $w$ in every timestamp.

Define the following set $I_{S}$ of timestamps associated with $V_{S} \backslash\{w\}$ :

$$
I_{S}=\left\{t \in[T]: u_{t} \in V_{S}^{\prime} \text { for some } u \in V_{S} \backslash\{w\}\right\}
$$

Essentially, $I_{S}$ contains those timestamps where the base vertices of $V_{S} \backslash\{w\}$, that is of span greater than zero, have associated vertices in $S$. These timestamps are essential for computing a solution of Restricted Timeline Cover, that is to compute whether there exists a temporal cover of $G$ of span at most $k$ starting from $S$. We define now the sets of base vertices and vertices associated with $S$ having a span equal to 0 :

$$
Z_{S}=V_{B} \backslash V_{S} \quad Z_{S}^{\prime}=S \backslash V_{S}^{\prime}
$$

First, we show two easy properties of $S$ and $I_{S}$ on the temporal graph $G-\{w\}$.

- Lemma 6. Let $S$ be a solution of MinTimelineCover on instance $G-\{w\}$ and let $I_{S}$ be the associated set of timestamps. Then $\left|I_{S}\right| \leq 2 k$.
- Lemma 7. Let $S$ be a solution of MinTimelineCover on instance $G-\{w\}$. Then, $\operatorname{sp}\left(Z_{S}^{\prime}\right)=0$. Moreover, $Z_{S}^{\prime}$ covers each temporal edge of $G-\{w\}$ not covered by $V_{S}^{\prime} \backslash \tau(w)$.

Now, we introduce the concept of feasible assignment, which is used to "guess" how $S$ is rearranged in a solution of Restricted Timeline Cover. Recall that an assignment $X$ intersects a set $I_{S}$ of timestamps if there exists $v_{t} \in X$ such that $t \in I_{S}$.

- Definition 8 (Feasible assignment). Consider an instance of Restricted Timeline Cover that consists of a temporal graph $G=\left(V_{B}, \mathcal{T}, E\right)$, a vertex $w \in V_{B}$, a temporal cover $S$ of $G-\{w\}$ of span at most $k$, and sets $V_{S}, V_{S}^{\prime}$ and $I_{S}$ associated with $S$. We say that an assignment $X \subseteq \tau\left(V_{S}\right)$ of $V_{S}$ is a feasible assignment (with respect to $G, S$, and $I_{S}$ ) if all of the following conditions hold:

1. the span of $X$ is at most $k$;
2. every edge of $G\left[V_{S}\right]$ is covered by $X$;
3. $X \cap \tau(w)$ is a non-empty assignment of $\{w\}$;
4. for every $v \in V_{S} \backslash\{w\}$, at least one of the following holds: (1) $X \cap \tau(v)$ is empty; (2) $X \cap \tau(v)$ is an assignment of $\{v\}$ that intersects with $I_{S}$; or (3) $X \cap \tau(v)$ contains a vertex $v_{t}$ such that $v_{t} w_{t} \in E$ and $w_{t} \notin X \cap \tau(w)$.

Given a feasible assignment $X$, we denote

$$
M_{S}(X)=\left\{v \in V_{S}: X \cap \tau(v) \neq \emptyset\right\} \quad N_{S}(X)=\left\{v \in V_{S}: X \cap \tau(v)=\emptyset\right\}
$$

Informally, point 4 considers the possible cases for a feasible assignment of the vertices of a base vertex $v \in V_{S} \backslash\{w\}$ : none of the associated vertices in $I_{S}$ belongs to the computed solution (case 4.(1)); some of its associated vertices in $I_{S}$ belongs to the solution (case 4.(2)); or some of the $v_{t}$ vertices are forced, since they belong to an edge $v_{t} w_{t}$ with $t \in I_{S}$, that we know is not covered by $w_{t}$ (case 4.(3)). Note that these cases are not necessarily mutually exclusive.

Note that $M_{S}(X)$ and $N_{S}(X)$ form a partition of $V_{S}$. Also note that $G, S$, and $I_{S}$ are fixed in the remainder, so we assume that all feasible assignments are with respect to $G, S$, and $I_{S}$ without explicit mention. We now relate feasible assignments to temporal covers.

- Definition 9. Let $X^{*}$ be a temporal cover of $G$ and let $X$ be a feasible assignment. We say that $X^{*}$ agrees with $X$ if:
- for each $v \in M_{S}(X), X^{*} \cap \tau(v)=X \cap \tau(v)$;
- for each $v \in N_{S}(X)$ and each $t \in I_{S}, X^{*}$ contains every neighbor $u_{t}$ of $v_{t}$ such that $u_{t} \in \tau\left(Z_{S}\right)$.

The intuition of $X^{*}$ agreeing with $X$ is as follows. For $v \in M_{S}(X), X$ "knows" which vertices of $\tau(v)$ should be in the solution, and we require $X^{*}$ to contain exactly those. For $v \in N_{S}(X)$, we interpret that $X$ does not want any vertex $v_{t}$ with $t \in I_{S}$. Thus, to cover the edges incident to $v_{t}$ that go outside of $V_{S}$, we require $X^{*}$ to contain the other endpoint. Note an important subtlety: we act "as if" $X^{*}$ should not contain $v_{t}$ or other vertices of $N_{S}(X)$ with timestamp in $I_{S}$, but the definition does not forbid it. Hence, $X^{*}$ can contain a vertex of $N_{S}(X)$ in some timestamps of $I_{S}$, as long as $X^{*}$ contains also its neighbors (in $I_{S}$ ) outside $V_{S}$.

The main purpose of feasible assignments and agreement is as follows.

- Lemma 10. Let $X^{*}$ be a temporal cover of $G$ of span at most $k$. Then there exists a feasible assignment $X$ such that $X^{*}$ agrees with $X$.

Proof. Construct $X \subseteq X^{*}$ as follows: add $X^{*} \cap \tau(w)$ to $X$, and for $v \in V_{S} \backslash\{w\}$, add $X^{*} \cap \tau(v)$ to $X$ if and only if $X^{*} \cap \tau(v)$ intersects with the set $I_{S}$, or if it contains a vertex $v_{t}$ incident to an edge $v_{t} w_{t} \in E$ such that $w_{t} \notin X^{*} \cap \tau(w)$. Note that since $X^{*}$ is an assignment of $V_{B}, X$ is an assignment of $V_{S}$.

We first focus on arguing that $X$ satisfies each condition of a feasible assignment (Definition 8). For Condition 1, since $X^{*}$ has span at most $k$ and $X \subseteq X^{*}$, it is clear that $X$ also has span at most $k$. For Condition $3, X^{*} \cap \tau(w)$ is non-empty by the definition of a temporal cover, and we added $X^{*} \cap \tau(w)$ to $X$. For Condition 4, we explicitly require in our construction of $X$ that for each $v \in V_{S} \backslash\{w\}$, if $X \cap \tau(v)$ is non-empty, then it is equal to $X^{*} \cap \tau(v)$ and it either intersects with $I_{S}$ or covers an edge not covered by $X \cap \tau(w)=X^{*} \cap \tau(w)$.

Let us focus on Condition 2. Let $u_{t} v_{t} \in E\left(G\left[V_{S}\right]\right)$. If $u=w$, then if we did not add $w_{t}$ to $X, X^{*}$ must contain $v_{t}$ and we added $X^{*} \cap \tau(v)$ to $X$, thereby covering the edge. The same holds if $v=w$. Assume $u \neq w, v \neq w$, and suppose without loss of generality that $X^{*}$ contains $u_{t}$ to cover the edge. Suppose for contradiction that $X$ does not cover $u_{t} v_{t}$. Then we did not add $X^{*} \cap \tau(u)$ to $X$, which implies that $X^{*} \cap \tau(u)$ does not intersect with $I_{S}$. In particular, $t \notin I_{S}$. Recall that $S$, the temporal cover of $G-\{w\}$, only intersects with $\tau(u)$ and $\tau(v)$ in timestamps contained in $I_{S}$. Hence, $S$ cannot cover $u_{t} v_{t}$, a contradiction. We deduce that $X$ covers every edge. Therefore, $X$ is a feasible assignment.


Figure 2 An example of application of iterative compression (timestamps 1 and 6 are not shown as they are edgeless, also vertex $w$ is not shown, its assignment is defined as in Fig. 1). In the left part, we represent solution $S=\left\{v_{2}, v_{3}, u_{3}, u_{4}, z_{4}\right\}$, where the vertices in $S$ are highlighted with grey rectangles. Note that $I_{S}=\{2,3,4\}, V_{S}=\{v, u\}, V_{S}^{\prime}=\left\{v_{2}, v_{3}, u_{3}, u_{4}\right\}, Z_{S}=\{z\}, Z_{S}^{\prime}=\left\{z_{4}\right\}$. In the right part, we represent in grey a feasible assignment $X$ associated with $S$, containing vertices $u_{2}$, $u_{3}, u_{4}$; in light grey we highlight $N_{S}^{\prime}=\left\{v_{2}\right\}$. The sets associated with $S$ and $X$ are: $M_{S}=\{u\}$, $N_{S}=\{v\}, N_{S}^{\prime}=\left\{v_{2}\right\}, N_{S}^{\prime \prime}=\left\{v_{2}, v_{3}, v_{4}\right\}$. The reduction to Constrained Digraph Pair Cut eventually leads to the solution of MinTimeLineCover represented in Fig. 1.

It remains to show that $X^{*}$ agrees with $X$. For $v \in M_{S}(X), X^{*} \cap \tau(v)=X \cap \tau(v)$ by the construction of $X$. For $v \in N_{S}(X)$, there is no $v_{t} \in X^{*}$ with $t \in I_{S}$, as otherwise we would have added $X^{*} \cap \tau(v)$ to $X$. For every such $v_{t}, X^{*}$ must contain all of its neighbors in $\tau\left(Z_{S}\right)$ to cover the edges, as required by the definition of agreement.

It remains to show that the number of feasible assignments has bounded size and can be enumerated efficiently. We first show the latter can be achieved through the following steps. Start with $X$ as an empty set and then apply the following steps (checking that the overall span is at most $k$ ):
(1) Branch into every non-empty assignment $X_{w}$ of $\{w\}$ of span at most $k$. In each branch, add the chosen subset $X_{w}$ to $X$;
(2) For every edge $v_{t} w_{t} \in E\left(G\left[V_{S}\right]\right)$ such that $w_{t} \notin X_{w}$, add $v_{t}$ to $X$;
(3) For every $v \in V_{S} \backslash\{w\}$, such that $X \cap \tau(v)=\emptyset$ at this moment, branch into $\left|I_{S}\right|+1$ options: either add no vertex of $\tau(v)$ to $X$, or choose a vertex $v_{t}$ and add it to $X$, where $t \in I_{S} ;$
(4) For every $v \in V_{S} \backslash\{w\}$ such that $X \cap \tau(v) \neq \emptyset$ at this moment, branch into every assignment $X_{v}$ of $\{v\}$ of span at most $k$ that contains every vertex of $X \cap \tau(v)$ (if no such assignment exists, abort the current branch). For each such branch, add every vertex of $X_{v} \backslash X$ to $X$.

- Theorem 11. The above steps enumerate every feasible assignment in time $O\left(2^{4 k \log k} T^{2} k n\right)$, where $n=\left|V_{B}\right|$.


## Reducing to Constrained Digraph Pair Cut

Our objective is now to list every feasible assignment and, for each of them, to verify whether there is a temporal cover that agrees with it. More specifically, consider a feasible assignment $X \subseteq \tau\left(V_{S}\right)$. Our goal is to decide whether there is a temporal cover $X^{*}$ of span at most $k$ that agrees with $X$. Since we branch over every possible feasible assignment $X$, if there is a temporal cover $X^{*}$ of $G$ of span at most $k$, then by Theorem 11 our enumeration will eventually consider an $X$ that $X^{*}$ agrees with, and hence we will be able to decide of the existence of $X^{*}$.

We show that finding $X^{*}$ reduces to the Constrained Digraph Pair Cut problem, as we define it below. For a directed graph $H$, we denote its set of arcs by $A(H)$ (to avoid confusion with $E(G)$, which is used for the edges of an undirected graph $G$ ). For $F \subseteq A(H)$, we write $H-F$ for the directed graph with vertex set $V(H)$ and arc set $A(H) \backslash F$.

- Problem 12. (Constrained Digraph Pair Cut)

Input: A directed graph $H=(V(H), A(H))$, a source vertex $s \in V(H)$, a set of vertex pairs $P \subseteq\binom{V(H)}{2}$ called forbidden pairs, a subset of arcs $D \subseteq A(H)$ called deletable arcs, and an integer $k^{\prime}$.
Output: Does there exist a set of arcs $F \subseteq D$ of $H$ such that $|F| \leq k^{\prime}$ and such that, for each $\{u, v\} \in P$, at least one of $u, v$ is not reachable from $s$ in $H-F$ ?

It is known that Constrained Digraph Pair Cut can be solved in time $O^{*}\left(2^{k^{\prime}}\right)$ [16], but a few remarks are needed before proceeding. In [16], the authors only provide an algorithm for the vertex-deletion variant, and do not consider deletable/undeletable arcs. It is easy to make an arc undeletable by adding enough parallel paths between the two endpoints, and we show at the end of the section that our formulation of Constrained Digraph Pair Cut reduces to the simple vertex-deletion variant. The vertex-deletion variant also admits a randomized polynomial kernel, and other FPT results are known for weighted arc-deletion variants [14].

So let us fix a feasible assignment $X$ for the remainder of the section. We will denote $M_{S}=M_{S}(X)$ and $N_{S}=N_{S}(X)$. We also consider the following set of vertices associated with $N_{S}$ :

$$
N_{S}^{\prime}=\left\{v_{2}: v \in N_{S}\right\} \quad N_{S}^{\prime \prime}=\left\{v_{t} \in \tau\left(N_{S}\right): t \in I_{S}\right\} .
$$

For each base vertex $v \in N_{S}$, we need $N_{S}^{\prime}$ to contain any vertex of $\tau(v)$ that belongs to the time interval $[2, T-1]$, so we choose $v_{2}$ arbitrarily. Then, $N_{S}^{\prime \prime}$ contains those vertices $v_{t}$, with $t \in I_{S}$, not chosen by the feasible assignment $X$. Note that according to our definition of agreement, a solution $X^{*}$ should contain all the neighbors of $N_{S}^{\prime \prime}$ vertices that are in $Z_{S}$. Recall that we have defined $Z_{S}=V_{B} \backslash V_{S}$ and $Z_{S}^{\prime}=S \backslash V_{S}^{\prime}$. By Lemma 7 we know that $Z_{S}^{\prime}$ covers each temporal edge of $G\left[V_{B} \backslash\{w\}\right]$ not covered by $S \cap V_{S}^{\prime}$, and that $\operatorname{sp}\left(Z_{S}^{\prime}\right)=0$. We may assume that for each $v \in Z_{S}$, there is exactly one $t \in[T]$ such that $v_{t} \in Z_{S}^{\prime}$ (there cannot be more than one since $Z_{S}^{\prime}$ has span 0 , and if there is no such $t$, we can add any $v_{t}$ without affecting the span). Furthermore, we will assume that for each $v \in Z_{S}$, the vertex $v_{t}$ in $Z_{S}^{\prime}$ is not $v_{1}$ nor $v_{T}$. Indeed, since we assume that the first and last timestamps of $G$ have no edges, if $v_{t}=v_{1}$ or $v_{t}=v_{T}$, then $v_{t}$ covers no edge and we may safely change $v_{p}$ to another vertex of $\tau(v)$.

The following observation will be useful for our reduction to Constrained Digraph Pair Cut.

- Observation 13. Let $u_{t} v_{t} \in E(G)$ such that $u \in N_{S}$ and $v \notin M_{S}$. Then $v \in Z_{S}$ and, if $u_{t} \notin N_{S}^{\prime \prime}$, we have $v_{t} \in Z_{S}^{\prime}$.

Now, given a feasible assignment $X \subseteq \tau\left(V_{S}^{\prime}\right)$, sets $M_{S}, N_{S}, N_{S}^{\prime}, N_{S}^{\prime \prime}, Z_{S}$, and $Z_{S}^{\prime}$, we present our reduction to the Constrained Digraph Pair Cut problem. We construct an instance of this problem that consists of the directed graph $H=(V(H), A(H))$, the set of forbidden (unordered) pairs $P \subseteq\binom{V(H)}{2}$, and the deletable $\operatorname{arcs} D \subseteq A(H)$ by applying the following steps. The second step in the construction is the most important and is shown in Figure 3. The intuition of these steps is provided afterwards.

1. add to $H$ the source vertex $s$;
2. for each $v \in Z_{S} \cup N_{S}$, let $v_{i}$ be the vertex of $Z_{S}^{\prime} \cup N_{S}^{\prime}$, where $i \in[2, T-1]$. Add to $H$ the vertices $v_{1}^{+}, \ldots, v_{i-1}^{+}, v_{i}^{-}, v_{i+1}^{+}, \ldots, v_{T}^{+}$, the vertices $b_{v, j}, c_{v, j}, d_{v, j}$, for $j \in[T] \backslash\{i\}$, and the set of arcs shown in Figure 3, that is there are $\operatorname{arcs}\left(v_{j}^{+}, b_{v, j}\right),\left(v_{j}^{+}, c_{v, j}\right),\left(c_{v, j}, d_{v, j}\right)$, $\left(d_{v, j}, v_{j}^{-}\right)$, for each $j \in[T] \backslash\{i\}$ and four directed paths: (1) from $b_{v, i-1}$ to $b_{v, 1}$, (2) from $c_{v, 1}$ to $c_{v, i-1}$, (3) from $b_{v, i+1}$ to $b_{v, T}$ and (4) from $c_{v, T}$ to $c_{v, i+1}$.
Add to $D$ the set of deletable $\operatorname{arcs}\left(c_{v, j}, d_{v, j}\right)$, for $j \in[T] \backslash\{i\}$.
Then add the following pairs to $P$ :
a. $\left\{d_{v, h}, b_{v, j}\right\}$, with $1 \leq h<j \leq i-1$;
b. $\left\{d_{v, h}, b_{v, j}\right\}$, with $i+1 \leq j<h \leq T$;
c. $\left\{c_{v, h}, d_{v, j}\right\}$, with $1 \leq h \leq i-1 \leq i+1 \leq j \leq T$;
d. $\left\{c_{v, h}, d_{v, j}\right\}$, with $1 \leq j \leq i-1 \leq i+1 \leq h \leq T$.

Note that we have created $T+3(T-1)=4 T-3$ vertices in $H$ in this step. The subgraph of $H$ induced by these vertices will be called the gadget corresponding to $v$.
3. for each temporal edge $u_{t} v_{t} \in E(G)$ such that $u_{t}, v_{t} \in \tau\left(Z_{S}\right) \cup\left(\tau\left(N_{S}\right) \backslash N_{S}^{\prime \prime}\right)$, there are three cases. First note that at least one of $u_{t}$ or $v_{t}$ is in $Z_{S}^{\prime}$. Indeed, if $u, v \in Z_{S}$, this is because an element of $Z_{S}^{\prime}$ must cover the temporal edge, and if $u \in N_{S}$, then $v_{t} \in Z_{S}^{\prime}$ by Observation 13 (or if $v \in N_{S}, u_{t} \in Z_{S}^{\prime}$ ). The subcases are then:
a. if $u_{t}, v_{t} \in Z_{S}^{\prime} \cup N_{S}^{\prime}$, add the pair $\left\{u_{t}^{-}, v_{t}^{-}\right\}$to $P$;
b. if $u_{t} \in Z_{S}^{\prime} \cup N_{S}^{\prime}, v_{t} \notin Z_{S}^{\prime} \cup N_{S}^{\prime}$, add the $\operatorname{arc}\left(u_{t}^{-}, v_{t}^{+}\right)$to $H$;
c. if $v_{t} \in Z_{S}^{\prime} \cup N_{S}^{\prime}, u_{t} \notin Z_{S}^{\prime} \cup N_{S}^{\prime}$, add the $\operatorname{arc}\left(v_{t}^{-}, u_{t}^{+}\right)$to $H$;
4. for each temporal edge $u_{t} v_{t} \in E(G)$ such that $u_{t} \in\left(\tau\left(M_{S}\right) \backslash X\right) \cup N_{S}^{\prime \prime}$ and $v_{t} \in \tau\left(Z_{S}\right)$, there are two cases:
a. if $v_{t} \notin Z_{S}^{\prime}$, add the $\operatorname{arc}\left(s, v_{t}^{+}\right)$to $H$;
b. if $v_{t} \in Z_{S}^{\prime}$, add the pair $\left\{s, v_{t}^{-}\right\}$to $P$.

Define $k^{\prime}=k-s p(X)$. This concludes the construction. We will refer to the elements $1,2,3,4$ of the above enumeration as the Steps of the construction. Note that the only deletable arcs in $D$ are the $\operatorname{arcs}\left(c_{v, j}, d_{v, j}\right)$ introduced in Step 2.

From here, the interpretation of $H$ is that if we delete arc set $F$, then
(p1) For $v_{t} \notin Z_{S}^{\prime} \cup N_{S}^{\prime}$ we should include $v_{t}$ in $X^{*}$ if and only if $s$ reaches $v_{t}^{+}$in $H-F$;
(p2) For $v_{t} \in Z_{S}^{\prime} \cup N_{S}^{\prime}$ we should include $v_{t}$ in $X^{*}$ if and only if $s$ does not reach $v_{t}^{-}$in $H-F$.

The idea behind the steps of the construction is then as follows (and is somewhat easier to describe in the reverse order of steps). Step 4 describes an initial set of vertices that $s$ is forced to reach, which correspond to vertices that are forced in $X^{*}$. A vertex $v_{t}$ in $\tau\left(Z_{S}\right)$ is forced in $X^{*}$ if there is in an edge $u_{t} v_{t}$ and $u_{t} \in \tau\left(M_{S}\right)$ but $u_{t} \notin X$. By our definition of agreement, $v_{t}$ is also forced if $u_{t} \in N_{S}^{\prime \prime}$. Step 4 handles both situations: if $v_{t} \notin Z_{S}^{\prime}$, we force $s$ to reach $v_{t}^{+}$with the $\operatorname{arc}\left(s, v_{t}^{+}\right)$, which is not deletable. If $v_{t} \in Z_{S}^{\prime}$, then $v_{t}^{-} \in V(H)$, and $s$ is forced to not reach $v_{t}^{-}$by adding $\left\{s, v_{t}^{-}\right\}$to $P$. By ( p 1 ) and ( p 2 ), both cases correspond to including $v_{t}$ in $X^{*}$. Then, Step 3 ensures that each temporal edge is "covered": for a temporal edge $u_{t} v_{t}$, a pair of the form $\left\{u_{t}^{-}, v_{t}^{-}\right\}$in $P$ requires that $s$ does not reach one of the two, i.e. that we include one in $X^{*}$, and an undeletable arc of the form $\left(u_{t}^{-}, v_{t}^{+}\right)$enforces that if $s$ reaches $u_{t}^{-}$(i.e. $u_{t} \notin X^{*}$ ), then $s$ reaches $v_{t}^{+}$(i.e. $v_{t} \in X^{*}$ ). The reason why $Z_{S}^{\prime}$ is needed in our construction is that each edge has at least one negative corresponding vertex, so that no other case needs to be considered in Step 3.


Figure 3 Gadget for $v_{i} \in Z_{S}^{\prime} \cup N_{S}^{\prime}$, where $i \in[2, T-1]$. We assume that there exist temporal edges $u_{t} v_{t} \in E(G)$, where $t \in\{i-1, i+1\}$, such that $u_{t} \in\left(\tau\left(M_{S}\right) \backslash X\right) \cup N_{S}^{\prime \prime}, v_{t} \in \tau\left(Z_{S}\right)$ and $v_{t} \notin Z_{S}^{\prime}$, thus arcs from $s$ to $v_{t}^{+}$are added. The dashed arcs represent deletable arcs.

Finally, Step 2 enforces the number of deleted arcs to correspond to the span of a solution. That is, it ensures that if we want to add to $X^{*}$ a set of $h$ vertices of base vertex $v \in Z_{S}$ to our solution of Restricted Timeline Cover (so with a span equal to $h-1$ ), then we have to delete $h-1$ deletable arcs of the corresponding gadget of $H$ in order to obtain a solution to Constrained Digraph Pair Cut (and vice-versa). Indeed, consider the gadget in Fig. 3. If $v_{i}$ is not included in $X^{*}$, then in the gadget $s$ reaches $h$ positive vertices $v_{l}^{+}, \ldots, v_{r}^{+}$(and $v_{i}^{-}$). It follows that vertices $b_{v, l}, \ldots, b_{v, r}, c_{v, l}, \ldots, c_{v, r}$ and $d_{v, l}, \ldots, d_{v, r}$ are all reachable from $s$. The pairs $\left\{d_{v, x}, b_{v, y}\right\}$ defined at Step 2, where either $l \leq x \leq y \leq r-1$ if $r<i$, or $l+1 \leq x \leq y \leq r$ if $l>i$, ensures that $\operatorname{arcs}\left(c_{v, j}, d_{v, j}\right)$, with $j \in[l, r-1]$ in the former case or with $j \in[l+1, r]$ in the latter case, are deleted.

If $v_{i}$ is included in $X^{*}$, then in the gadget $s$ reaches $h-1$ positive vertices $v_{l}^{+}, \ldots, v_{r}^{+}$, with $i \in[l, r]$, and must not reach negative vertex $v_{i}^{-}$. It follows that vertices $b_{v, l}, \ldots, b_{v, r}$, $c_{v, l}, \ldots, c_{v, r}$ and $d_{v, l}, \ldots, d_{v, r}$ are all reachable from $s$. Then $h-1 \operatorname{arcs}\left(c_{v, j}, d_{v, j}\right)$, with $j \in[l, r] \backslash\{i\}$, must be deleted, due to the pairs $\left\{d_{v, x}, b_{v, y}\right\},\left\{c_{v, x}, d_{v, y}\right\}$ defined at Step 2.

Note that Step 2 is the reason we added dummy timestamps 1 and $T$. If $v_{1}$ or $v_{T}$ were allowed to be in $Z_{S}^{\prime} \cup N_{S}^{\prime}$, we would need a different gadget for these cases.

- Lemma 14. There exists a solution of Restricted Timeline Cover that agrees with $X$ if and only if there is $F \subseteq D$ with $|F| \leq k^{\prime}$ such that $s$ does not reach a forbidden pair in $H-F$. Moreover, given such a set $F$, a solution of Restricted Timeline Cover can be computed in polynomial time.

Sketch of the proof. ( $\Rightarrow$ ) Suppose that there exists a solution $X^{*}$ of Restricted Timeline Cover that agrees with $X$. By definition of Restricted Timeline Cover, $X^{*}$ has span at most $k$. Note that for $v \in M_{S}$, the agreement requires that $X^{*} \cap \tau(v)=X \cap \tau(v)$, and so the span of $v$ in $X^{*}$ is the same as the span of $v$ in $X$. Thus

$$
\sum_{v \in Z_{S} \cup N_{S}} s p\left(v, X^{*}\right) \leq k-s p(X)=k^{\prime}
$$

We may assume that for every $v \in V_{B}$, at least one of $v_{2}, \ldots, v_{T-1}$ is in $X^{*}$, as otherwise we add one arbitrarily without affecting the span (if only $v_{1}$ or $v_{T}$ is in $X^{*}$, remove it first). For each $v \in Z_{S} \cup N_{S}$, consider the gadget corresponding to $v$ in $H$ and delete some of its dashed arcs as follows (we recommend referring to Figure 3).

First, if only one of $\tau(v)$ is in $X^{*}$, no action is required on the gadget. So assume that $X^{*} \cap \tau(v)$ has at least two vertices; in the following we denote $v_{l}=v_{\delta\left(v, X^{*}\right)}$ and $v_{r}=v_{\Delta\left(v, X^{*}\right)}$ the vertices associated with $v$ having minimum and maximum timestamp, respectively, contained in $X^{*}$. We assume that $l, r \in[2, T-1]$ and $l<r$. Note that $X^{*} \cap \tau(v)=\left\{v_{l}, v_{l+1}, \ldots, v_{r}\right\}$.

Let $v_{i} \in Z_{S}^{\prime} \cup N_{S}^{\prime}$, where $i \in[2, T-1]$. Then

- suppose that $l, r \in[2, i-1]$, then: delete every $\operatorname{arc}\left(c_{v, q}, d_{v, q}\right)$, with $l \leq q \leq r-1$
- suppose that with $l, r \in[i+1, T-1]$, then: delete every $\operatorname{arc}\left(c_{v, q}, d_{v, q}\right)$, with $l+1 \leq q \leq r$
- suppose that $l \in[2, i]$ and $r \in[i, T-1]$, then: delete every arc $\left(c_{v, q}, d_{v, q}\right)$, with $l \leq q \leq i-1$, and delete every $\operatorname{arc}\left(c_{v, q}, d_{v, q}\right)$, with $i+1 \leq q \leq r$.

We see that by construction for all $v \in Z_{S} \cup N_{S}$, the number of arcs deleted in the gadget corresponding to $v$ is equal to the number of vertices in $X^{*} \cap \tau(v)$ minus one, that is the span of $v$ in $X^{*}$. Since these vertices have span at most $k^{\prime}$, it follows that we deleted at most $k^{\prime}$ arcs from $H$. Denote by $H^{\prime}$ the graph obtained after deleting the aforementioned arcs. We argue that in $H^{\prime}, s$ does not reach a forbidden pair. To this end, we claim the following.
$\triangleright$ Claim 15. For $v \in Z_{S} \cup N_{S}$ and $t \in[T]$, if $s$ reaches $v_{t}^{+}$in $H^{\prime}$, then $v_{t} \in X^{*}$, and if $s$ reaches $v_{t}^{-}$in $H^{\prime}$, then $v_{t} \notin X^{*}$.

Now, armed with the above claim, we can prove that in $H^{\prime}, s$ does not reach both vertices of a forbidden pair $q \in P$, thus concluding this direction of the proof.
$(\Leftarrow)$ Suppose that there is a set $F \subseteq D$ with at most $k^{\prime}$ arcs such that $s$ does not reach a forbidden pair in $H-F$. Denote $H^{\prime}=H-F$. We construct $X^{*}$ from $F$, which will also show that it can be reconstructed from $F$ in polynomial time. Define $X^{*} \subseteq V(G)$ as follows: - for each $v \in M_{S}$, add every element of $X \cap \tau\left(M_{S}\right)$ to $X^{*}$;

- for each $v_{t} \in V(G) \backslash \tau\left(M_{S}\right)$, we add $v_{t}$ to $X^{*}$ if and only if one of the following holds: (1) $v_{t}^{+} \in V(H)$ and $s$ reaches $v_{t}^{+}$in $H^{\prime}$; or (2) $v_{t}^{-} \in V(H)$, and $s$ does not reach $v_{t}^{-}$in $H^{\prime}$;
- for each $v_{j}, v_{h} \in X^{*}$ with $j<h$, add $v_{t}$ to $X^{*}$ for each $t \in[j+1, h-1]$.

Note that $X^{*}$ agrees with $X$. Indeed, for $v \in M_{S}$, there is no gadget corresponding to $v$ in the construction and thus we only add $X \cap \tau(v)$ to $X^{*}$. For $u \in N_{S}$, consider $u_{t} \in N_{S}^{\prime \prime}$ and a neighbor $v_{t}$ of $u_{t}$ in $\tau\left(Z_{S}\right)$. If $v_{t} \notin Z_{S}^{\prime}$, Step 4 adds an undeletable arc from $s$ to $v_{t}^{+}$, hence $s$ reaches that vertex and we put $v_{t}$ in $X^{*}$. If $v_{t} \in Z_{S}^{\prime}$, Step 4 adds $\left\{s, v_{t}^{-}\right\}$to $P$, and thus $s$ does not reach $v_{t}^{-}$in $H^{\prime}$, and again we add $v_{t}$ to $X^{*}$. Therefore, we add all the $\tau\left(Z_{S}\right)$ neighbors of $u_{t}$ to $X^{*}$, and so it agrees with $X$. We can prove that $X^{*}$ covers every temporal edge of $G$ and that $s p\left(X^{*}\right) \leq k$.

## Wrapping up

Before concluding, we must show that we are able to use the results of [16] to get an FPT algorithm for Constrained Digraph Pair Cut, as we have presented it. As we mentioned, the FPT algorithm in [16] studied the vertex-deletion variant and does not consider undeletable elements, but this is mostly a technicality. Roughly speaking, in our variant, it suffices to replace each vertex with enough copies of the same vertex, and replace each deletable arc $(u, v)$ with a new vertex, adding arcs from the $u$ copies to that vertex, and arcs from that vertex to the $v$ copies. Deleting $(u, v)$ corresponds to deleting that new vertex. For undeletable arcs, we apply the same process but repeat it $k^{\prime}+1$ times.

- Lemma 16. The Constrained Digraph Pair Cut problem can be solved in time $O^{*}\left(2^{k}\right)$, where $k$ is the number of arcs to delete.

We are able now to prove the main result of our contribution.

- Theorem 17. MinTimelineCover on a temporal graph $G=\left(V_{B}, E, \mathcal{T}\right)$ can be solved in time $O^{*}\left(2^{5 k \log k}\right)$.

Proof. First, we discuss the correctness of the algorithm we presented. Assume that we have an ordering on the base vertices of $G$ and that $v$ is the first vertex of this ordering. A solution $S$ of MinTimelineCover on $G[\{v\}]$ is equal to $S=\emptyset$.

Then for $i$, with $i \in\left[2,\left|V_{B}\right|\right]$, let $G_{i}$ be the temporal graph induced by the first $i$ vertices and let $w$ be the $i+1$-th vertex. Given a solution $S$ of MinTimelineCover on instance $G_{i}$ of span at most $k$, we can decide whether there exists a solution of MinTimelineCover on instance $G_{i+1}$ by computing whether there exists a solution $X^{*}$ of the Restricted Timeline Cover problem on instance $G_{i}, w, S$. By Lemma 10 and by Theorem 11 if there exists such an $X^{*}$, then there exists a feasible assignment $X$ such that $X^{*}$ agrees with $X$. By Lemma 14 we can compute, via the reduction to Constrained Digraph Pair Cut, whether there exists a solution of Restricted Timeline Cover on instance on instance $G_{i}, w, S$, and if so obtain such a solution (if no such solution $X^{*}$ exists, then Lemma 14 also says that we will never return a solution, since every feasible assignment $X$ that we enumerate will lead to a negative instance of Constrained Digraph Pair Cut). Thus the Restricted Timeline Cover subproblem is solved correctly, and once it is solved on $G_{\left|V_{B}\right|}$, we have a solution to MinTimelineCover.

Now, we discuss the complexity of the algorithm. We must solve Restricted Timeline Cover $\left|V_{B}\right|$ times. For each iteration, by Theorem 11 we can enumerate the feasible assignments in $O\left(2^{4 k \log k} T^{3} n\right)$ time. For each such assignment, the reduction from Restricted Timeline Cover to Constrained Digraph Pair Cut requires polynomial time, and each generated instance can be solved in time $O^{*}\left(2^{k}\right)$. The time dependency on $k$ is thus $O^{*}\left(2^{4 k \log k} \cdot 2^{k}\right)$, which we simplify to $O^{*}\left(2^{5 k \log k}\right)$.

## 4 Conclusion

We have presented an FPT algorithm for the MinTimelineCover problem, a variant of VERTEX Cover on temporal graphs recently considered for timeline activities summarizations. We point out some relevant future directions on this topic: (1) to improve, if possible, the time complexity of MinTimelineCover by obtaining a single exponential time algorithm (of the form $\left.O^{*}\left(c^{k}\right)\right)$; (2) to establish whether MinTimelineCover admits a polynomial kernel, possibly randomized (which it might, since Constrained Digraph Pair Cut famously admits a randomized polynomial kernel).

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