Subquadratic Kernels for Implicit 3-HITTING SET and 3-SET PACKING Problems

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Abstract

We consider four well-studied \textsf{NP}-complete packing/covering problems on graphs: Feedback Vertex Set in Tournaments (FVST), Cluster Vertex Deletion (CVD), Triangle Packing in Tournaments (TPT) and Induced $P_3$-Packing. For these four problems kernels with $O(k^2)$ vertices have been known for a long time. In fact, such kernels can be obtained by interpreting these problems as finding either a packing of $k$ pairwise disjoint sets of size 3 (3-Set Packing) or a hitting set of size at most $k$ for a family of sets of size at most 3 (3-Hitting Set). In this paper, we give the first kernels for FVST, CVD, TPT and Induced $P_3$-Packing with a subquadratic number of vertices. Specifically, we obtain the following results.

- FVST admits a kernel with $O(k^3)$ vertices.
- CVD admits a kernel with $O(k^4)$ vertices.
- TPT admits a kernel with $O(k^3)$ vertices.
- Induced $P_3$-Packing admits a kernel with $O(k^2)$ vertices.

Our results resolve an open problem from WorKer 2010 on the existence of kernels with $O(k^{2-\epsilon})$ vertices for FVST and CVD. All of our results are based on novel uses of old and new “expansion lemmas”, and a weak form of crown decomposition where (i) almost all of the head is used by the solution (as opposed to all), (ii) almost none of the crown is used by the solution (as opposed to none), and (iii) if $H$ is removed from $G$, then there is almost no interaction between the head and the rest (as opposed to no interaction at all).

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1 Introduction

Kernelization, a subfield of Parameterized Complexity, provides a mathematical framework to analyze the performance of polynomial time preprocessing. It makes it possible to quantify the degree to which polynomial time algorithms succeed at reducing input instances of an NP-hard problem. More formally, every instance of a parameterized problem \( \Pi \) is associated with an integer \( k \), which is called the parameter, and \( \Pi \) is said to admit a kernel if there is a polynomial-time algorithm, called the kernelization algorithm, that reduces the input instance of \( \Pi \) down to an equivalent instance of \( \Pi \) whose size is bounded by a function \( f(k) \) of \( k \). (Here, two instances are equivalent if both of them are either Yes-instances or No-instances.) Such an algorithm is called an \( f(k) \)-kernel for \( \Pi \). If \( f(k) \) is a polynomial function of \( k \), we say that the kernel is a polynomial kernel. Over the last decade, kernelization has become an active field of study, especially with the development of complexity-theoretic lower bound tools for kernelization. These tools can be used to show that a polynomial kernel \([5, 14, 21, 23]\), or a kernel of a specific size \([10, 11, 24]\) for concrete problems would imply an unlikely complexity-theoretic collapse. We refer to the surveys \([19, 22, 27, 29]\), as well as the books \([9, 13, 17, 32]\), for a detailed treatment of the area of kernelization.

One of the most well known examples of a polynomial kernel is a kernel with \( \mathcal{O}(k^d) \) sets and elements for \( d \)-Hitting Set using the Erdős-Rado Sunflower lemma.\(^1\) In this problem, the input consists of a universe \( U \), a family \( \mathcal{F} \) containing sets of size at most \( d \) over \( U \), and in integer \( k \). The objective is to determine whether there exists a set \( S \subseteq U \) of size at most \( k \) that intersects every set in \( \mathcal{F} \). Abu-Khzam [2] gave an improved kernel for \( d \)-Hitting Set, still with \( \mathcal{O}(k^d) \) sets, but with \( \mathcal{O}(k^{d-1}) \) elements.

The importance of the \( d \)-Hitting Set problem stems from the number of other problems that can be re-cast in terms of it. For example, in the Feedback Vertex Set in Tournaments (FVST) problem, the input is a tournament \( T \) together with an integer \( k \). The task is to determine whether there exists a subset \( S \) of vertices of size at most \( k \) such that the sub-tournament \( T - S \) obtained from \( T \) by removing \( S \) is acyclic. It turns out that FVST is a \( 3 \)-Hitting Set problem, where the vertices of \( T \) are the universe, and the family \( \mathcal{F} \) is the family containing the vertex set of every directed cycle on three vertices (triangle) of \( T \). Indeed, it can easily be shown that for every vertex set \( S \), \( T - S \) is acyclic if and only if \( S \) is a hitting set for \( \mathcal{F} \). Another example is the Cluster Vertex Deletion (CVD) problem. Here, the input is a graph \( G \) and an integer \( k \), and the task is to determine whether there exists a subset \( S \) of at most \( k \) vertices such that every connected component of \( G - S \) is a clique (such graphs are called cluster graphs). Also this problem can be formulated as a \( 3 \)-Hitting Set problem where the family \( \mathcal{F} \) contains the vertex sets of all induced \( P_3 \)’s of \( G \). An induced \( P_3 \) is a path on three vertices where the first and last vertex are non-adjacent in \( G \). The kernel with \( \mathcal{O}(k^2) \) elements for \( d \)-Hitting Set [2] can be adapted to obtain kernels with \( \mathcal{O}(k^2) \) vertices for Feedback Vertex Set in Tournaments [12] and for Cluster Vertex Deletion [25].

The formulation of problems in terms of \( 3 \)-Hitting Set is useful not only in the context of kernelization, but within several paradigms for dealing with NP-hardness. Indeed, the \( 2.076k\mathcal{O}(1) \) time parameterized algorithm of Wahlström [34], the \( \mathcal{O}(1.519n+o(n)) \) time exact exponential time algorithm of Fomin et al. [18], and the folklore factor 3-approximation algorithm for \( 3 \)-Hitting Set, all immediately translate to algorithms with the same performance for FVST and CVD.

Still, as one translates graph problems into \( 3 \)-Hitting Set, some structure is lost. This structure can often be exploited to obtain algorithms with better performance than the corre-

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\(^1\)The origins of this result are unclear. The first kernel with \( \mathcal{O}(k^d) \) sets appeared in the work by Fellows et al. [15], but they do not make use of the Sunflower Lemma. To the best of our knowledge, the first exposition of the kernel based on the Sunflower Lemma appears in the book of Flum and Grohe [17].
Triangle Packing in Tournaments generalizes many “packing” problems. For example, Set Packing is an archetypal “covering” problem that generalizes many such problems, the input is a graph \(G\), and the task is to determine whether \(G\) contains \(k\) pairwise vertex-disjoint triangles. In \textsc{Induced }\textsc{P}_3-\textsc{Packing}, the input is a graph \(G\) and an integer \(k\), and the task is to determine whether \(G\) contains \(k\) pairwise vertex-disjoint induced \(P_3\)’s. These problems are the duals of \textsc{FVST} and \textsc{CVD}, respectively.

Just like the insights that led to a kernel for \textsc{d-Hitting Set} also led to a kernel for \textsc{d-Set Packing}, our insights from the improved kernelization algorithms for \textsc{FVST} and \textsc{CVD} yield improved kernelization algorithms for \textsc{Triangle Packing in Tournaments (TPT)} and \textsc{Induced }\textsc{P}_3-\textsc{Packing}. Specifically, we obtain the following results.

- \textsc{TPT} admits a kernel with \(O(k^{\frac{2}{3}})\) vertices.
- \textsc{Induced }\textsc{P}_3-\textsc{Packing} admits a kernel with \(O(k^{\frac{5}{3}})\) vertices.

We remark that, while the underlying philosophy of the kernels for \textsc{Triangle Packing in Tournaments (TPT)} and \textsc{Induced }\textsc{P}_3-\textsc{Packing} is borrowed from the kernels for \textsc{FVST} and \textsc{CVD}, obtaining the kernels for \textsc{TPT} and \textsc{Induced }\textsc{P}_3-\textsc{Packing} requires significant additional insights. However, for the sake of exposition, we next only focus (in the introduction) on our methods in the context of \textsc{FVST} and \textsc{CVD}.
Overview and Our Methods. Our kernelization algorithms for both FVST and CVD begin by employing trivial factor 3 polynomial time approximation algorithms. We use these algorithms to obtain approximate solutions of size at most 3k, or conclude that no solution of size at most k exists. So, let us now assume that we have solutions S of size at most 3k. In what follows, for both FVST and CVD, we aim to understand which “subpart” of the problem is similar to the Vertex Cover problem.

Let us first focus on our approach to specifically solve FVST. To this end, let (T, k) be an instance of FVST. Given the approximate solution S, our analysis starts by introducing the notion of a strong arc. Formally, an arc xy ∈ E(T) is strong if (i) at least one vertex among x and y belongs to S, and (ii) there are at least k + 2 vertices z ∈ V(T) such that xyz is a triangle. Let F be the set of all the strong arcs of T. Observe that any solution of size at most k + 1 must be a vertex cover of F. Before we analyze F, we need to examine S as described below.

Now, we try to “fit” every vertex s ∈ S into the unique topological ordering, ≺, of X = T − S. Towards this, for s ∈ S and x ∈ V(X), define $f_s^-(x) = |\{y ∈ V(X) : y ⪯ x, sy ∈ E(T)\}|$, and $f_s^+(x) = |\{y ∈ V(X) : y ≻ x, ys ∈ E(T)\}|$. Intuitively, the functions $f_s^-(x)$ and $f_s^+(x)$ measure how many arcs would have been in the “wrong direction” (with respect to the ordering ≺) if we inserted s into the position immediately after x in X. Using a simple “sliding argument”, we show that for each s ∈ S, there exists $x_s ∈ V(X)$ such that $0 ≤ f_s^-(x_s) − f_s^+(x_s) ≤ 1$. Then, for each s ∈ S, the smallest vertex (with respect to ≺) satisfying the property that $0 ≤ f_s^-(x_s) − f_s^+(x_s) ≤ 1$ is denoted by $ϕ(s)$. Observe that if for some s ∈ S and x ∈ X we have that $f_s^-(x), f_s^+(x) ≥ k + 2$, then s participates in k + 1 triangles whose pairwise intersection is exactly s. This implies that s must be part of every solution of size at most k. Thus, $f_{ϕ(s)}^-(ϕ(s)), f_{ϕ(s)}^+(ϕ(s)) ≤ k + 1$.

Next, we separately investigate the structure of triangles that contain a strong arc, and triangles that do not contain any strong arc. Formally, we call a triangle local if it does not contain any strong arc. In particular, we show that the vertices of any local triangle cannot lie “too far apart” in the ordering ≺ (of course, for a vertex s ∈ S, we use $ϕ(s)$ to measure the distance with respect to ≺). Having this claim at hand, FVST can be thought of as the problem of simultaneously hitting local triangles and strong arcs.

To take care of the two sets of objects to be hit simultaneously, we define a variant of Expansion Lemma [33, 9], which we call Double Expansion Lemma. To (roughly) describe it here, let ℓ > 0 and G be a bipartite graph with vertex sets A, S, and ̂S ⊆ S and ̂A ⊆ A. We say that ̂S has an ℓ-expansion into ̂A in G if $|N_G(Y) \cap ̂A| ≥ ℓ|Y|$ for every Y ⊆ ̂S. In addition, we would like to ensure that $N_G( ̂A) \subseteq ̂S$. In Double Expansion Lemma, we consider a scheme where we have one “global” bipartite graph, as well as d vertex-disjoint “local” bipartite graphs, and we would like to find a vertex set that exhibits the expansion and neighborhood containment properties in all of the graphs simultaneously (see Section 3 for details).

To design the subquadratic kernel for FVST, we apply Double Expansion Lemma where one “part” is S, and the other “part” is derived by first defining a set of “carefully selected subintervals” of X, say $Y_1, \ldots, Y_p$, trimming their ends to obtain yet another set of subintervals, $Y'_1, \ldots, Y'_q$, and then further partitioning each trimmed subinterval $Y'_i$ into a more refined set of subintervals, say $Y_{i,1}, \ldots, Y_{i,q}$. To be somewhat more precise, let us note that we have a global graph, G, with vertex bipartition $(\{Y_{i,j} : i ∈ \{1,\ldots,p\}, j ∈ \{1,\ldots,q\}\}, S)$, as well as local bipartite graphs, $H_i$, with vertex bipartition $(\{Y_{i,j} : j ∈ \{1,\ldots,q\}\}, S_i)$, where $S_i$ are those vertices in S that were determined to “fit” $Y_i$. The graphs $H_i$ take care of local triangles, and the global graph G takes care of vertex cover constraints (that is, the edges in F). We apply the Double Expansion Lemma appropriately, and show that if $|V(X)| ≥ ζk^{3/2}$, for some constant
eters unfamiliar with this notion (which we use only in the introduction) are referred to books such as \[9\].

Almost all of the vertices as in a standard crown decomposition. Second, the Crown is a weak form of a crown decomposition, as we explain after its description. Roughly speaking, one run of Mark is executed as follows. Initially, all the vertices in \( S \) are "alive". For \( k + 1 \) stages, Mark examines every vertex \( s \in S \) that is still alive, and attempts to associate an edge of a clique of \( G \setminus S \) to it. Here, the association can be done only if \( s \) is adjacent to exactly one vertex of the edge, and no vertex of that edge belongs to an already associated edge. If the attempt is successful, the vertex remains alive also for the next stage. If there exists a vertex that is alive after stage \( k + 1 \), then this vertex is part of \( k + 1 \) induced \( P_3 \)'s that intersect only at it, and hence we can apply a reduction rule. Supposing that this "lucky" situation does not occur, we say that the procedure was successful if roughly \( k^{2/3} \) vertices were still alive at stage \( k \) (roughly) \( k^{2/3} \). If the run was indeed successful in this sense, we mark all of the vertices alive at stage \( k^{2/3} \), and rerun the procedure on the graph \( G \) from which all marked vertices, which belong to \( S \), are removed (only for the sake of applying Mark again).

Let \( \hat{U} \) denote the set of all the vertices in \( S \) that were marked across all successful runs. Furthermore, denote \( L = S \setminus \hat{U} \). Now, let us explain how the sets \( S, V(G) \setminus (S \cup L) \) and \( L \) can be thought of as a weak form of a crown decomposition. Here, the Head is \( \hat{U} \), and we indeed prove that almost all of the vertices of \( \hat{U} \) should be part of any solution (as opposed to all vertices as in a standard crown decomposition). Second, the Crown is \( V(G) \setminus (S \cup L) \), and as a consequence of the fact that most of \( \hat{U} \) is present in every solution and as \( V(G) \setminus (S \cup L) \) is significantly larger than \( k \) (else we already have a kernel), we can (roughly) say that most of the vertices in \( V(G) \setminus (S \cup L) \) are not present in any solution (as opposed to none). Third, the Rest (or Royal Body) is \( L \), and we prove (in the sense explained below) that the "interaction" between the Head and the Rest is structured (as opposed to non-existent as in a standard crown decomposition). Let us now elaborate on the meaning of our last claim. Here, we compute a "small" subset \( M \subseteq V(G) \setminus (S \cup L) \) (specifically, this is the set of vertices associated to the vertices of \( L \) in the last unsuccessful run of \( \text{Mark} \)) such that every clique in \( G \setminus S \) becomes a module with respect to \( L \) once we remove the vertices in \( M \) from it.

Having the decomposition described above, the situation is more complicated that it usually is when we have a standard crown decomposition. To analyze this situation, we first classify the cliques in \( G \setminus S \) using three definitions. First, we classify these cliques as small, large or huge, and "throw away" the small cliques. Next, we also classify these cliques as either heavy or light, which corresponds to whether the fraction of vertices of the cliques that belong to \( M \) is large or small, respectively; in this step, we also throw away the heavy cliques, which can be done safely as \( M \) is shown to be small. Then, we also classify the cliques as either visible or hidden, corresponding to whether many or few vertices from \( L \) are adjacent to many vertices in these cliques, respectively. We show that not too many cliques can be visible, else a reduction rule can be applied, which allows us to throw away also large (but not huge) visible cliques. Next, we focus on good cliques, which are either large or huge, light, and either hidden or huge.

Our analysis proceeds by defining, for every vertex \( s \in S \), a small and a large side with respect to every clique. Roughly speaking, a side is the set of either all neighbors or all non-neighbors of \( s \) in that clique. Then, in the context of these sides, we prove (using an exchange argument) that good cliques exhibit a vertex cover-like behavior. That is, for any vertex \( s \in S \)

\(^{4}\) A crown decomposition is among the most classical and well-known tools in parameterized complexity. Readers unfamiliar with this notion (which we use only in the introduction) are referred to books such as [9].
and good clique, every solution either picks \( s \) or the entire small side of that clique with respect to \( s \). This claim gives rise to the definition of a bipartite graph where one side is \( S \) and the other side is the set of vertices of the good cliques. Here, there is an edge between \( s \in S \) and a vertex \( v \) in a good clique \( C \) if \( v \) belongs to the small side of \( C \) with respect to \( s \). Using the Expansion Lemma, if we find a large enough expansion in this graph, we prove that it is safe to select the vertices in \( S \) corresponding to that expansion. Let us remark that this proof is non-trivial as the edges of the bipartite graph are not necessarily edges in the input graph \( G \). Finally, if no large expansion can be found, it means that the bipartite graph contains many isolated vertices, which belong to the good cliques. However, because these vertices are isolated, we can observe that they form sets that are modules with respect to the entire graph \( G \) (rather than only with respect to \( L \)), which allows us to employ a reduction rule that decreases their number.

Finally, we say a few words about our kernels for packing problems, that is, for TPT and \textsc{Induced } \( P_3 \)-\textsc{Packing}. In both of these kernels, we start by finding a greedy packing, \( S \) of either triangles or induced paths on 3 vertices, depending on the problem we are dealing with. If the greedy collection is large, then we already have the answer. Else, the vertices present in any set in \( S \), say \( S \), form a hitting set. That is, \( G - S \) is a cluster graph and \( T - S \) is a transitive tournament. We exploit this structure in a manner similar to the way we exploited it to design subquadratic kernels for the hitting problems. Specifically, we make reduction rules that are, in some sense, “dual” to those given for \( \textsc{FVST} \) and \( \textsc{CVD} \), and use the appropriate variants of Expansion Lemma to find an irrelevant vertex to delete. However, as we currently deal with packing problems, there are also major deviations required to design the new kernels. For example, for \textsc{Induced } \( P_3 \)-\textsc{Packing}, the last stage of the kernelization algorithm, which lies at the heart of its correctness, is completely different from the last stage of the kernelization algorithm for \( \textsc{CVD} \). Here, the difference stems from the following crucial observation: in \textsc{Induced } \( P_3 \)-\textsc{Packing}, we need to present structural claims that hold for at least one solution, rather than for all solutions as in \( \textsc{CVD} \), but these structural claims have to be stronger than the ones presented for \( \textsc{CVD} \) as the solution itself has a more complicated structure (being a set of paths rather than a set of vertices). This crucial observation also holds for TPT, posing difficulties of the same nature.

**Additional Related Works.** It is known that unless \( \text{NP} \subseteq \text{co-NP}^{\text{poly}} \), for any \( d \geq 2 \) and for any \( \epsilon > 0 \), \( d \)-\textsc{Hitting Set} and \( d \)-\textsc{Set Packing} do not admit a kernel with \( O(k^{d-\epsilon}) \) sets \([10, 11]\).

In \([10]\), Dell and Marx studied several matching and packing problems, and provided non-trivial lower bounds as well as non-trivial upper bounds for packing some specific graphs such as matchings, \( P_i \)'s (here, the packing need not be induced) and \( K_{1,d} \)'s (stars with \( d \) leaves). Moser et al. \([31]\) studied the problem of packing a fixed connected graph \( H \) on \( \ell \) vertices in an input graph \( G \) (that is, determining whether there exist \( k \) vertex disjoint copies of \( H \) in \( G \)) and designed a kernel with \( O(k^{\ell-1}) \) vertices. In this context, it is also worth to point out the dichotomy result of Jansen and Marx \([26]\) regarding packing a fixed graph \( H \). Finally, very recently Bessy et al. \([3]\) studied \( \text{FVST} \) where the input tournament is restricted to be a sparse tournament, that is, a tournament where the feedback arc set is a matching. For this special case, they presented a linear-vertex kernel, and remarked that their methods do not extend to handle general tournaments.

**Reading Guide.** On the one hand, our kernels for \( \text{FVST} \) and \( \text{CVD} \) are independent of each other. On the other hand, the kernels for TPT and \textsc{Induced } \( P_3 \)-\textsc{Packing} borrow some of their ideas from the corresponding hitting set kernels, and therefore we recommend to read them after reading our kernels for \( \text{FVST} \) and \( \text{CVD} \). Section 3 gives the old and new Expansion Lemmas used in this paper. In Section 4, we give our \( O(k^{5/3}) \)-vertex kernel for \( \text{CVD} \), followed by a kernel of \( O(k^{3/2}) \) vertices for \( \text{FVST} \) in Section 5. In Sections 6 and 7, we give our kernels for \textsc{Induced } \( P_3 \)-\textsc{Packing} and TPT with \( O(k^{5/3}) \) and \( O(k^{3/2}) \) vertices, respectively. We conclude
the paper with some remarks and open problems in Section 8. To get a detailed idea of our techniques, a reader can first read the statements of Expansion Lemmas in Section 3, and then proceed to read our kernels for CVD and FVST. The proofs of the new Expansion Lemmas and the kernels for the packing problems can be read afterwards.

2 Preliminaries

**Graph Theory.** Given a graph $G$ (or digraph $D$), we let $V(G)$ ($V(D)$) and $E(G)$ ($E(D)$) denote its vertex-set and edge-set (arc-set), respectively. We use $\{u, v\}$ to denote an edge in an undirected graph and $uv$ to denote an arc in a digraph. The *open neighborhood*, or simply the *neighborhood*, of a vertex $v \in V(G)$ is defined as $N_G(v) = \{w \mid \{v, w\} \in E(G)\}$. The *closed neighborhood* of $v$ is defined as $N_G[v] = N_G(v) \cup \{v\}$. The *degree* of $v$ is defined as $d_G(v) = |N_G(v)|$. We can extend the definition of neighborhood of a vertex to a set of vertices as follows. Given a subset $U \subseteq V(G)$, $N_G(U) = \bigcup_{u \in U} N_G(u)$ and $N_G[U] = \bigcup_{u \in U} N_G[u]$. The *induced subgraph* $G[U]$ is the graph with vertex-set $U$ and edge-set $\{\{u, u'\} \mid u, u' \in U\}$, and $\{u, u'\} \in E(G)$). Moreover, we define $G \setminus U$ as the induced subgraph $G[V(G) \setminus U]$. We omit subscripts when the graph $G$ is clear from context. We use $P_\ell$ to denote a path in a graph on $\ell$ vertices. A path $P = uvv$ in a graph $G$ is called an *induced path* if there is no edge between $u$ and $v$ in $E(G)$. An induced $P_\ell$-packing is a set of vertex disjoint induced $P_\ell$’s. A subset $X$ of $V(G)$ is called a *module* if every vertex in $X$ has same set of neighbors in $V(G) \setminus X$. For a collection of graph $\mathcal{H}$, by $V(H)$ we denote $\bigcup_{H \in \mathcal{H}} V(H)$.

A tournament is a directed graph $T$ such that for every pair of vertices $u, v \in V(T)$, exactly one of $uv$ or $vu$ is a directed arc of $T$. For any three vertices $x, y, z \in V(T)$, we say that $xyz$ is a triangle if arcs $xy$, $yz$ and $zx$ form a directed cycle. A tournament in which there is no directed cycle is called a transitive tournament.

**Reduction Rules.** Kernelization algorithms often rely on the design of *reduction rules*. The rules are numbered, and each rule consists of a condition and an action. We always apply the first rule whose condition is true. Given a problem instance $(I, k)$, the rule computes (in polynomial time) an instance $(I', k')$ of the same problem where $k' \leq k$. Typically, $|I'| < |I|$, where if this is not the case, it should be argued why the rule can be applied only polynomially many times. We say that the rule safe if the instances $(I, k)$ and $(I', k')$ are equivalent.

3 Tool: Expansion Lemmas

In this section we give the classical Expansion Lemma as well as some two new Expansion Lemmas that we make use of in our kernels. We start with some preliminaries. Let $\ell$ be a positive integer. An $\ell$-*star* is a graph on $\ell + 1$ vertices where one vertex, called the center, has degree $\ell$, and all other vertices are adjacent to the center and have degree one. A *bipartite graph* is a graph whose vertex-set can be partitioned into two independent sets. Such a partition of the vertex-set is called a *bipartition* of the graph. Let $G$ be a bipartite graph with bipartition $(A, S)$. A subset of edges $M \subseteq E(G)$ is called *$\ell$-expansion of $S$ into $A$* if

(i) every vertex of $S$ is incident to exactly $\ell$ edges of $M$, 
(ii) and $M$ saturates exactly $|\ell| S$ vertices in $A$.

Note that an $\ell$-expansion saturates all vertices of $S$, and for each $u \in S$ the set of edges in $M$ incident on $u$ form an $\ell$-star. The following lemma allows us to compute an $\ell$-expansion in a bipartite graph. It captures a certain property of neighborhood sets which is very useful for designing kernelization algorithms.
Lemma 3.1 ([33, 9], Expansion Lemma). Let \( G \) be a bipartite graph with bipartition \((A, S)\) such that there are no isolated vertices in \( A \). Let \( \ell \) be a positive integer such that \(|A| \geq \ell|S|\). Then, there are non-empty subsets \( X \subseteq S \) and \( Y \subseteq A \) such that

- there is a \( \ell \)-expansion from \( X \) into \( Y \),
- and there is no vertex in \( Y \) that has a neighbor in \( S \setminus X \), i.e. \( N_G(Y) = X \).

Further, the sets \( X \) and \( Y \) can be computed in polynomial time.

An alternate but an equivalent view on expansion properties is as follows. Let \( \ell \) \geq 0 and \( G \) be a bipartite graph with vertex sets \( A, S \), and \( \hat{S} \subseteq S \) and \( \hat{A} \subseteq A \). We say that \( \hat{S} \) has an \( \ell \)-expansion into \( \hat{A} \) in \( G \) if \(|N_G(Y) \cap \hat{A}| \geq \ell|Y|\) for every \( Y \subseteq \hat{S} \). We call this (ex1) property. Using the classical Hall’s Theorem one can observe that this condition is equivalent to having for each \( u \in S \), an \( \ell \)-star, and that all of the \( \ell \)-stars are pairwise vertex disjoint. In the next two lemmas, and in Sections 5 and 7 we will use this definition of expansion, while in the rest of the paper, we will use the classical definition of expansion.

Lemma 3.2 (New Expansion Lemma). Let \( \ell \) be a positive integer and \( G \) be a bipartite graph with bipartition \((A, S)\). Then, there exist \( \hat{S} \subseteq S \) and \( \hat{A} \subseteq A \) such that \( \hat{S} \) has an \( \ell \)-expansion into \( \hat{A} \) in \( G \), \( N_G(\hat{A}) \subseteq \hat{S} \) and \(|A \setminus \hat{A}| \leq \ell|S \setminus \hat{S}|\). Moreover, the sets \( \hat{S} \) and \( \hat{A} \) can be computed in polynomial time.

The property that \( N_G(\hat{A}) \subseteq \hat{S} \) will be called (ex2). Lemma 3.2 is slightly different from Lemma 3.1, as it does not require \(|A| \geq \ell|S|\) and that there is no isolated vertex in \( A \), and thus \( \hat{A} \) and \( \hat{S} \) may be empty. However, we still have the bound on the number of removed vertices. That is, \(|A \setminus \hat{A}| < \ell|S \setminus \hat{S}|\), and hence, if \(|A| \geq \ell|S|\), then \( \hat{A} \) is nonempty. The difference between Lemmas 3.1 and 3.2 indeed comes from the viewpoint: in Lemmas 3.1, we obtain \( Y \) by only keeping “desired” vertices in \( A \), while in Lemmas 3.1 we obtain \( \hat{A} \) by only removing “undesired” vertices from \( A \). Thus, in Lemmas 3.1, if \( X \) is empty then \( Y \) is empty, while in Lemmas 3.1, it is possible that \( \hat{S} \) is empty but \( \hat{A} \) is large.

Proof. We first give the formal description of our algorithm in Figure 1. We now analyze one call to Step 1. Clearly, the weight of \( C \) is \(|N_G(S_1) \cap A_2| + \ell|S_2| + |A_1| + |N_G(S_2) \cap A_1|\). If the

**Figure 1:** Algorithm to compute \( \hat{A} \) and \( \hat{S} \) Lemma 3.2.
weight of $C$ is less than $\ell|\hat{S}|$, then

$$|N_{G}(S_{1}) \cap A_{2}| + \ell|S_{2}| + |A_{1}| + |N_{G}(S_{2}) \cap A_{1}| < \ell|\hat{S}|$$

$$\implies |N_{G}(S_{1}) \cap A_{2}| + \ell|S_{2}| + |A_{1}| < \ell|\hat{S}|$$

$$\implies |N_{G}(S_{1}) \cap A_{2}| + |A_{1}| < \ell|\hat{S}| - \ell|S_{2}|$$

$$\implies |N_{G}(S_{1}) \cap (A_{1} \cup A_{2})| < \ell|S_{1}|$$

$$\implies |N_{G}(S_{1}) \cap \hat{A}| < \ell|S_{1}|,$$

which implies $|S_{1}| \geq 1$. Hence, $|\hat{A} \cup \hat{S}|$ is reduced after every call to Step 1, except the last call. The inequality above also implies that after each step, the number of vertices removed from $\hat{A}$ is less than $\ell$ times the number of vertices removed from $\hat{S}$. Note that in the last step, we do not remove anything. Thus, for the output $\hat{A}, \hat{S}$, we have $|A \setminus \hat{A}| \leq \ell|S \setminus \hat{S}|$ (where the equality is achieved only when $|A \setminus \hat{A}| = \ell|S \setminus \hat{S}| = 0$, i.e. the algorithm stops after just one call of Step 1).

If the size of $C$ is at least $\ell|\hat{S}|$, then the algorithm stops. Note that by Max-Flow-Min-Cut Theorem [20], the max-flow, which is equal to the min-cut size, is at most $\ell|\hat{S}|$, which is the total capacity of arcs incident with $s$. Hence, there is a max-flow in $H$ with optimal capacity $\ell|\hat{S}|$. Then for every $Y \subseteq \hat{S}$, the amount of flow passing from $s$ through $Y$ to $t$ in $H$ must be $\ell|Y|$. This gives $|N_{Y}(Y)| \geq \ell|Y|$, and since $V(H) \cap \hat{A} = \hat{A}$, so $|N_{G}(Y) \cap \hat{A}| \geq \ell|Y|$. Therefore, output $\hat{A}, \hat{S}$ satisfy Property (ex1).

Throughout the algorithm we maintain an invariant that $N_{G}(\hat{A}) \subseteq \hat{S}$. Clearly, this holds at the beginning of the algorithm. Since every time we remove a set $B_{i}$, we remove all its neighbors in $\hat{A}$, and so $N_{G}(\hat{A}) \subseteq \hat{S}$ holds at the end of every step, and so Property (ex2) is maintained during the algorithm. Thus Property (ex2) holds for the output $\hat{A}, \hat{S}$.

The algorithm runs in polynomial time, since $\hat{A} \cup \hat{S}$ is reduced each time we call Step 1, except the last call, and that each call to Step 1 runs in polynomial time.

As we discussed before, the viewpoint in the New Expansion Lemma is how many vertices are removed, rather than how many vertices remains as in the classical Expansion Lemma. This viewpoint enables us to generalize the Expansion Lemma to the Double expansion Lemma, where we can simultaneously achieve expansions in many graphs. In the following lemma, we consider a scheme where we have a “global” bipartite graph and $d$ vertex-disjoint “local” bipartite graphs and we would like to achieve the expansion in each of them simultaneously.

**Lemma 3.3 (Double Expansion Lemma).** Let $\ell$ be a positive integer, and $G, H_{1}, \ldots, H_{d}$ be bipartite graphs with bipartition $(A, S), (A_{1}, S_{1}), \ldots, (A_{d}, S_{d})$, respectively, such that $A_{i} \cap A_{j} = \emptyset, S_{i} \cap S_{j} = \emptyset$ for every $i \neq j$, and $\bigcup_{i=1}^{d} A_{i} = A, \bigcup_{i=1}^{d} S_{i} \subseteq S$. We can in polynomial time find $\hat{A} \subseteq A, \hat{S} \subseteq S, \hat{A}_{i} \subseteq A_{i}, \hat{S}_{i} \subseteq S_{i}$ for every $i$, satisfying the following

- $\hat{A} = \bigcup_{i=1}^{d} \hat{A}_{i}$,
- $|A \setminus \hat{A}| \leq 2\ell|S|$,
- $\hat{S}$ has an $\ell$-expansion into $\hat{A}$ in $G$, and for every $i$, $\hat{S}_{i}$ has an $\ell$-expansion into $\hat{A}_{i}$ in $H_{i}$,
- $N_{G}(\hat{A}) \subseteq \hat{S}$, and for all $1 \leq i \leq d$, $N_{H_{i}}(\hat{A}_{i}) \subseteq \hat{S}_{i}$.

Roughly speaking, the lemma asserts that we can find a set $\hat{A}$ such that $\hat{A}$ is the “image” of an expansion in the global graph, and the set of vertices $\hat{A}_{i}$ in every local graph is the image of another expansion in that local graph. Since, $\hat{A} = \bigcup_{i=1}^{d} \hat{A}_{i}$, we achieve simultaneous expansion. Since $|A \setminus \hat{A}| < 2\ell|S|$, we again have the property that if $|A| \geq 2\ell|S|$, then $\hat{A}$ is non-empty. Note that, unlike $\hat{A}$ and $\hat{A}_{i}$, we do not have $\hat{S} = \bigcup_{i=1}^{d} \hat{S}_{i}$, or even $\hat{S}_{i} \subseteq \hat{S}$.

To prove Lemma 3.3, we repeatedly apply Lemma 3.2, alternately to the global graph and then to local graphs, and refine $\hat{A}$ and $\bigcup_{i=1}^{d} \hat{A}_{i}$ until they are equal.
Input: $G, H_i$ for every $i$.

Step 0: Initialize $\hat{A} \leftarrow A, \hat{S} \leftarrow S, \hat{A}_i \leftarrow A_i, \hat{S}_i \leftarrow S_i$ for every $i$.

Stage 1: It consists of the following two steps.

Step 1: Apply the New Expansion Lemma (Lemma 3.2) on $G[\hat{A} \cup \hat{S}]$ and get $S^* \subseteq \hat{S}$ and $A^* \subseteq \hat{A}$ satisfying the Expansion Lemma. Set $\hat{S} \leftarrow S^*, \hat{A} \leftarrow A^*$ and $\hat{A}_i \leftarrow A^* \cap \hat{A}_i$ for every $i$ (we do not update $\hat{S}_i$). [This is to ensure that (ex2) still holds.]

Step 2: For every $i$, apply the New Expansion Lemma (Lemma 3.2) on $H_i[A_i \cup \hat{S}_i]$ and get $S_i^* \subseteq \hat{S}_i$ and $A_i^* \subseteq \hat{A}_i$ satisfying the New Expansion Lemma (Lemma 3.2). Set $\hat{S}_i \leftarrow S_i^*, \hat{A}_i \leftarrow A_i^*$ for every $i$, and $\hat{A} \leftarrow \bigcup_i A_i^*$ (we do not update $\hat{S}$). [Similarly, this is to ensure that (ex2) still holds.]

If at least one of $\hat{A}, \hat{S}, \hat{A}_i, \hat{S}_i$ changes, repeat Stage 1. Otherwise, stop the algorithm.

Output: $\hat{A}, \hat{S}, \hat{A}_i, \hat{S}_i$ for every $i$.

Figure 2: Algorithm to compute $\hat{A}, \hat{S}, \hat{A}_i, \hat{S}_i$ for every $i$ in Lemma 3.3.

Proof. We first give the formal description of our algorithm in Figure 2 and an illustration in Figure 3.

Observe that each call of Stage 1 runs in polynomial time. Towards this, note that the size of at least one of $\hat{A}, \hat{S}, \hat{A}_i, \hat{S}_i$ reduces after each call of Stage 1 (except the last call), and since each step itself can be carried out in polynomial time, the algorithm itself runs in polynomial time. We will show that the output satisfies all the properties stated in the lemma.

The first property, $\hat{A} = \bigcup_i A_i$ is vacous, since it is always maintained as an invariant during the algorithm. To prove the second property, observe that each time we call Step 1, we remove some vertices from $\hat{S}$ and $\hat{A}$. The number of vertices removed from $\hat{A}$ at Step 1 is at most $\ell$ times the number of vertices removed from $\hat{S}$ at the same step (guaranteed by Lemma 3.2). Besides, initially $|\hat{S}| = |S|$, so there are at most $|S|$ vertices removed from $\hat{S}$ in all calls to Step 1. This implies that there are at most $\ell|S|$ vertices removed from $\hat{A}$ in all calls to Step 1. Similarly, each time we call Step 2, we remove some vertices from $\bigcup_i \hat{S}_i$ and $\bigcup_i \hat{A}_i$. The number of vertices removed from $\bigcup_i \hat{A}_i$ at Step 2 is at most $\ell$ times the number of vertices removed from $\bigcup_i \hat{S}_i$ at the same step. Besides, initially $|\bigcup_i \hat{S}_i| \leq |S|$, so there are at most $|S|$ vertices removed from $\bigcup_i \hat{S}_i$ in all calls to Step 2. This implies that there are at most $\ell|S|$ vertices removed from $\bigcup_i \hat{A}_i$ in all calls to Step 2, which is also exactly the number of vertices removed from $\hat{A}$ in all calls to Step 2. In conclusion, there are at most $2\ell|S|$ vertices removed from $\hat{A}$ during the algorithm, and so $|A \setminus \hat{A}| < 2\ell|S|$.

To prove that $\hat{S}$ and $\hat{A}$ satisfies (ex1) and (ex2), we first observe that $\hat{S}$ and $\hat{A}$ satisfies (ex1) after every Step 1 of the algorithm, and so $\hat{S}$ and $\hat{A}$ satisfies (ex1) after Step 2 if no vertex of $\hat{A}$ is removed in that step. This means that the output $\hat{S}$ and $\hat{A}$ satisfies (ex1) since $\hat{S}$ and $\hat{A}$ are unchanged in the last stage. It remains to show that output $\hat{S}$ and $\hat{A}$ satisfies (ex2). To do so, we prove by induction that $N_G(\hat{A}) \subseteq \hat{S}$ at the end of every stage. Clearly it is true at the beginning of the algorithm. Suppose that it is true at after Stage $j$ (i.e. the $j$th call of Stage 1), then there are no edge between $\hat{A}$ and $\hat{S} \setminus \hat{S}$ in $G$. At Step 1 of Stage $j + 1$, we apply Lemma 3.2 on $G[\hat{A} \cup \hat{S}]$ and get $S^*$ and $A^*$ such that $N_G(\hat{A} \cup \hat{S})(A^*) \subseteq S^*$, then there is no edge between $A^*$ and $\hat{S} \setminus S^*$ in $G$. Thus there is no edge between $A^*$ and $(\hat{S} \setminus \hat{S}) \cup (\hat{S} \setminus S^*)$ in $G$, i.e., $N_G(A^*) \subseteq S^*$. We then set $\hat{A} \leftarrow A^*, \hat{S} \leftarrow S^*$, and so $N_G(\hat{A}) \subseteq \hat{S}$ holds at the end of Step 1 of Stage $j + 1$. At Step 2 of Stage $j + 1$, some vertices are removed from $\hat{A}$ while $\hat{S}$ is unchanged, and hence $N_G(\hat{A}) \subseteq \hat{S}$ holds at the end of Stage $j + 1$. This means that output
Figure 3: Illustration of the Double Expansion Algorithm – original vertices are red; a vertex turns blue if it is removed by a call of Step 1, and turns green if it is removed by a call of Step 2. Note that changing colors in $S$ of $G$ does not affect colors in $S_i$ of $H_i$ and vice versa. (Vertices in this figure are “well-ordered” to illustrate the algorithm.)

$\hat{S}$ and $\hat{A}$ satisfies (ex2).

Fix and integer $i \leq d$. To prove that $\hat{S}_i$ and $\hat{A}_i$ satisfies (ex1) and (ex2), we first observe that $\hat{S}_i$ and $\hat{A}_i$ satisfies (ex1) after every execution of Step 2 of the algorithm, and so the output $\hat{S}_i$ and $\hat{A}_i$ satisfies (ex1). It remains to show that output $\hat{S}_i$ and $\hat{A}_i$ satisfies (ex2). To do so, we prove by induction that $N_{H_i}(\hat{A}_i) \subseteq \hat{S}_i$ at the end of every stage. Clearly it is true at the beginning of the algorithm. Suppose that it is true after Stage $j$, then there are no edge between $\hat{A}_j$ and $\hat{S}_j \setminus \hat{S}_i$ in $H_i$. At Step 1 of Stage $j + 1$, some vertices are removed from $\hat{A}_j$ while $\hat{S}_i$ is unchanged, then obviously $N_{H_i}(\hat{A}_i) \subseteq \hat{S}_i$ holds at the end of Step 1 of Stage $j + 1$. At Step 2 of Stage $j + 1$, we apply Lemma 3.2 on $H[\hat{A}_i \cup \hat{S}_i]$ and get $S^*_i$ and $A^*_i$ such that $N_{H_i}([\hat{A}_i \cup \hat{S}_i])(A^*_i) \subseteq S^*_i$, then there is no edge between $A^*_i$ and $\hat{S}_i \setminus S^*_i$ in $H_i$. Thus there are no edge between $A^*_i$ and $(S_i \setminus \hat{S}_i) \cup (\hat{S}_i \setminus S^*_i)$ in $H_i$, i.e., $N_{H_i}(A^*_i) \subseteq S^*_i$. We then set $\hat{A}_i \leftarrow A^*_i$, $\hat{S}_i \leftarrow S^*_i$, and so $N_{H_i}(\hat{A}_i) \subseteq \hat{S}_i$ holds at the end of Stage $j + 1$. This means that output $\hat{S}$ and $\hat{A}$ satisfies (ex2). This concludes the proof of the lemma.

We would like to remark that the Double Expansion Lemma can be generalized to the Triple Expansion Lemma (or $\eta$-levels Expansion Lemma), where the system contains a global bipartite graph $G_i$, local bipartite graphs $H_i$, and super-local bipartite graphs $H_{i,j}$. The proofs of these generalized version are similar to that of the Double Expansion Lemma. The idea of the Double Expansion Lemma (or its generalizations) is that one tries to capture different properties using different bipartite graphs at the same time.

4 Kernel for Cluster Vertex Deletion

In this section, we prove the following theorem.

**Theorem 1.** CVD admits a kernel with $O(k^{1.5})$ vertices.
Let \((G, k)\) be an instance of CVD. Recall that CVD admits a polynomial-time 3-approximation algorithm. We call this algorithm with \(G\) as input, and thus we obtain a 3-approximate solution \(S\). If \(|S| > 3k\), then we conclude that \((G, k)\) is a No-instance. Thus, we next assume that \(|S| \leq 3k\). Notice that \(G \setminus S\) is a collection of cliques, which we denote by \(C\).

In what follows, we denote \(\alpha = 2, \beta = 1, \gamma = 10, \delta = 3, \lambda = 1\), and \(\eta = 1\), so that \((1 - \frac{1}{7})\gamma \geq 2\eta\) (used in the proof of Lemma 4.11), \((\frac{1}{3} - \frac{1}{7})\gamma > (\frac{1}{(\alpha-1)^2} + \lambda)\) (used in the proof of Lemma 4.13), and \(\gamma \geq \frac{\delta}{\delta-1}(\frac{1}{(\alpha-1)^2} + \lambda)\) (used in the proof of 4.14).

### 4.1 Bounding the Number of Cliques

First, we have the following simple rule, whose safeness is obvious.

**Reduction Rule 4.1.** If there exists \(C \in C\) such that no vertex in \(C\) has a neighbor in \(S\), then remove \(C\) from \(G\). The new instance is \((G \setminus C, k)\).

Now, we define the bipartite graph \(B\) by setting one side of the bipartition to be \(S\) and the other side to be \(C\),\(^5\) such that there exists an edge between \(s \in S\) and \(C \in C\) if and only if \(s\) is adjacent to at least one vertex in \(C\). Note that by Reduction Rule 4.1, no clique in \(C\) is an isolated vertex in \(B\). We thus proceed by presenting the following rule, where we rely on the Expansion Lemma (Lemma 3.1). It should be clear that the conditions required to apply the algorithm provided by this lemma are satisfied.

**Reduction Rule 4.2.** If \(|C| \geq 2|S|\), then call the algorithm provided by Lemma 3.1 to compute sets \(X \subseteq S\) and \(Y \subseteq C\) such that \(X\) has a 2-expansion into \(Y\) in \(B\) and \(N_B(Y) \subseteq X\). The new instance is \((G \setminus X, k - |X|)\).

We now argue that this rule is safe.

**Lemma 4.1.** Reduction Rule 4.2 is safe.

*Proof.* In one direction, it is clear that if \(S^*\) is a solution to \((G \setminus X, k - |X|)\), then \(S^* \cup X\) is a solution to \((G, k)\). For the other direction, let \(S^*\) be a solution to \((G, k)\). We denote \(S' = (S^* \setminus V(Y)) \cup X\). Notice that for all \(s \in X\), there exists an induced \(P_3\) in \(G\) of the form \(u - s - v\) where \(u\) is any vertex in one clique associated to \(s\) by the 2-expansion that is adjacent to \(s\) and \(v\) is any vertex in the other clique associated to \(s\) by the 2-expansion that is adjacent to \(v\). The existence of such \(u\) and \(v\) is implied by the definition of the edges of \(B\). Thus, as \(S^*\) is a solution to \((G, k)\), we have that \(|X \setminus S'| \leq |S^* \setminus V(Y)|\), and hence \(|S'| \leq |S^*| \leq k\). Note that \(G \setminus S'\) is a collection of isolated cliques together with a subgraph of \(G \setminus S^*\). Thus, as \(G \setminus S^*\) does not contain any induced \(P_3\), we derive that \(G \setminus S'\) also does not contain any induced \(P_3\). We conclude that \(S'\) is a solution to \((G, k)\), and as \(X \subseteq S'\), we have that \(S' \setminus X\) is a solution to \((G \setminus X, k - |X|)\). Thus, \((G \setminus X, k - |X|)\) is a \(Yes\)-instance. \(\Box\)

Due to Reduction Rule 4.2, from now on \(|C| \leq 6k\).

### 4.2 The Specification of the Marking Procedure

We proceed by presenting a procedure called \textbf{Mark}. Clearly, every vertex in \(S\) that has both a neighbor and a non-neighbor in a clique in \(C\) is a vertex due to which that clique in \(C\) is not a module. The procedure \textbf{Mark} accordingly associates vertices \(s \in S\) with sets \(\text{mark}(s)\) of edges that belong to cliques in \(C\). In particular, we would ensure that for all \(s \in S\), there would not exist two distinct edges \(e, e' \in \text{mark}(s)\) that have a common endpoint, as well as that for all distinct \(s, s' \in S\), there would not exist two distinct edges \(e \in \text{mark}(s), e' \in \text{mark}(s')\) that have a common endpoint.

\(^5\)Here, we slightly abuse notation. Specifically, we mean that each clique in \(C\) is represented by a unique vertex in \(V(B)\), and we refer to both the clique and the corresponding vertex identically.
**Specification.** The procedure Mark first initializes $M \leftarrow \emptyset$, $T \leftarrow S$, and for all $s \in S$, \( \text{mark}(s) \leftarrow \emptyset \). At each stage $i, i = 1, 2, \ldots, k+1$, Mark executes the following process. For each $s \in T$, if there exist $C \in C$ and $\{u, v\} \in E(C)$ such that $\{s, u\} \in E(G)$ but $\{s, v\} \notin E(G)$ and $\{u, v\} \cap M = \emptyset$, then insert $u, v$ into $M$ and $\{u, v\}$ into mark$(s)$, and otherwise remove $s$ from $T$. The order in which the process examines the vertices in $T$ is immaterial given that it examines each vertex in $T$ exactly once. Moreover, if $i = \lceil \beta k^{2/3} \rceil$, then the process sets $U$ to be equal to $T$. If $T$ is updated in subsequent stages, $U$ is not updated as well.

We say that Mark succeeded if $|U| \geq \lceil \alpha k^{2/3} \rceil$, and otherwise we say that Mark failed. Moreover, if there exists $s \in S$ such that $|\text{mark}(s)| \geq k+1$, then we say that Mark was lucky. Let us begin the analysis of Mark with the following simple lemma.

**Lemma 4.2.** For any solution $S^*$ to $(G, k)$ and vertex $s \in S \setminus S^*$, it holds that $S^* \cap \{u, v\} \neq \emptyset$ for all $\{u, v\} \in \text{mark}(s)$.

**Proof.** Let $S^*$ be a solution to $(G, k)$. Consider some vertex $s \in S$ and edge $\{u, v\} \in \text{mark}(s)$. Note that $\{s, u, v\}$ is the vertex set of an induced $P_3$ in $G$. Therefore, $S^* \cap \{s, u, v\} \neq \emptyset$. We thus have that if $s \notin S^*$, then $S^* \cap \{u, v\} \neq \emptyset$. \( \Box \)

In light of Lemma 4.2, we employ the following rule.

**Reduction Rule 4.3.** If there exists $s \in S$ such that $|\text{mark}(s)| \geq k+1$ (i.e., Mark was lucky), then remove $s$ from $G$ and decrement $k$ by $1$. The new instance is $(G \setminus s, k-1)$.

**Lemma 4.3.** Reduction Rule 4.3 is safe.

**Proof.** In one direction, it is clear that if $S^*$ is a solution to $(G \setminus s, k-1)$, then $S^* \cup \{s\}$ is a solution to $(G, k)$. For the other direction, let $S^*$ be a solution to $(G, k)$. Observe that for all $s' \in S$ and $\{u, v\}, \{u', v'\} \in \text{mark}(s)$, it holds that $\{u, v\} \cap \{u', v'\} = \emptyset$. Thus, by Lemma 4.2 and since $|\text{mark}(s)| \geq k+1$, if $s \notin S^*$ then $|S^*| \geq k+1$, which is not possible as $|S^*| \leq k$. We derive that $s \in S^*$, and therefore $S^* \setminus \{s\}$ is a solution to $(G \setminus s, k-1)$. \( \Box \)

The main purpose of Mark is to derive information on $(G, k)$ also when it is not coincidentally lucky. More precisely, we have the following simple but useful lemma.

**Lemma 4.4.** For any solution $S^*$ to $(G, k)$, $|U \setminus S^*| \leq \frac{1}{\beta} k^{1/3}$.

**Proof.** Let $S^*$ be a solution to $(G, k)$. Again, observe that for all $s \in S$ and $\{u, v\}, \{u', v'\} \in \text{mark}(s)$, it holds that $\{u, v\} \cap \{u', v'\} = \emptyset$. In addition, observe that for all $s, s' \in S$, $\{u, v\} \in \text{mark}(s)$ and $\{u', v'\} \in \text{mark}(s')$, it holds that $\{u, v\} \cap \{u', v'\} = \emptyset$. Thus, by Lemma 4.2, $|S^*| \geq \sum_{s \in U \setminus S^*} |\text{mark}(s)| \geq \lceil \beta k^{2/3} \rceil |U \setminus S^*|$. Since $|S^*| \leq k$, we conclude that $|U \setminus S^*| \leq \frac{1}{\beta} k^{1/3}$. \( \Box \)

We also need to derive an upper bound on the number of marked vertices, namely $|M|$.

**Lemma 4.5.** If Mark was neither lucky nor successful, then $|M| \leq 6(\alpha + \beta)k^{1/2}$.

**Proof.** Since Mark was unlucky, $|\text{mark}(s)| \leq k$ for all $s \in S$. Thus, $|M| \leq 2|U||k + 2|S \setminus U|([\beta k^{2/3}]-1)$. Since Mark failed, we further have that $|M| \leq 2([\alpha k^{2/3}]-1)k + 6k([\beta k^{2/3}]-1) \leq 6(\alpha + \beta)k^{1/2}$. \( \Box \)
4.3 Multiple Calls to the Marking Procedure

Let us now explain how we employ Mark. We initialize \( \hat{U} = \emptyset \) and \( \hat{G} = G \). Then, we call Mark with \((\hat{G}, k)\) as input. If Mark was lucky, then we execute Reduction Rule 4.3 and restart the entire process (including the initialization phase). Else, if Mark succeeded, then for the set \( U \) computed by the current call, we update \( \hat{U} = \hat{U} \cup U \) and \( \hat{G} = \hat{G} \setminus U \), and then we proceed to execute another call. Otherwise, Mark was unlucky and also failed, and we let \( M \) denote the same set \( M \subseteq V(G) \setminus S \) as computed by the current call to Mark, after which we terminate the process. Note that after each call to Mark, either Reduction Rule 4.3 is executed or the size of \( \hat{U} \) increases, and therefore it is clear that the process eventually terminates. We denote \( L = S \setminus \hat{U} \).

By relying on Lemma 4.4, we have the following lemma.

**Lemma 4.4.** Let \( i \) be the number of calls to Mark that succeeded but were unlucky. For any solution \( S^* \) to \((G, k)\), \(|\hat{U} \setminus S^*| \leq i \cdot \frac{1}{3} k^{1/3} \) and \(|S^* \cap \hat{U}| \geq i \cdot (\alpha [k^{2/3}] - \frac{1}{3} k^{1/3})\).

**Proof.** First, note that \(|S^* \cap \hat{U}| \geq i \cdot \alpha [k^{2/3}] - |\hat{U} \setminus S^*|\) as the sets \( U \) computed at distinct iterations are pairwise disjoint and the size of each one of them is at least \( \alpha [k^{2/3}]\). Thus, it is sufficient to prove that \(|\hat{U} \setminus S^*| \leq i \cdot \frac{1}{3} k^{1/3}\). However, this inequality follows from Lemma 4.4. \( \square \)

As a consequence of the two bounds in Lemma 4.6, we have the following corollary.

**Corollary 4.1.** For any solution \( S^* \) to \((G, k)\), \(|\hat{U} \setminus S^*| \leq \frac{1}{(\alpha - 1) \beta} k^{2/3}\).

**Proof.** First, note that \( k \geq |S^* \cap \hat{U}|\). Thus, by the second bound in Lemma 4.6, \( k \geq i \cdot (\alpha [k^{2/3}] - \frac{1}{3} k^{1/3}) \geq i \cdot (\alpha k^{2/3} - \frac{1}{3} k^{1/3})\), which implies that \( i \leq \frac{k}{\alpha k^{2/3} - \frac{1}{3} k^{1/3}} = \frac{k^{2/3}}{\alpha k^{1/3} - \frac{1}{3}} \leq \frac{1}{(\alpha - 1) \beta} k^{2/3}\).

By the first bound in Lemma 4.6, we thus derive that indeed \(|\hat{U} \setminus S^*| \leq \frac{1}{(\alpha - 1) \beta} k^{2/3}\). \( \square \)

The usefulness of Corollary 4.1 stems from the observation that it implies that we have found a (possibly large) set \( \hat{U} \subseteq S \) such that not only any \( S^* \) to \((G, k)\) contains almost all the vertices in \( \hat{U} \), but also that the removal of \( \hat{U} \) from \( G \) significantly simplifies \( G \) as described by the following lemma.

**Lemma 4.7.** For every clique \( C \in \mathcal{C} \), \( C[V(C) \setminus M] \) is a module in \( G \setminus \hat{U} \).

**Proof.** Let \( C \) be a clique in \( \mathcal{C} \). By the specification of Mark, for every vertex \( s \in L \), it holds that there do not exist \( u, v \in V(C) \setminus M \) such that \( u \in N_G(s) \) and \( v \notin N_G(s) \) (since \( \{u, v\} \notin \text{mark}(s) \)). Furthermore, every vertex in \( C \) is adjacent to both \( u \) and \( v \), and every vertex in a clique in \( \mathcal{C} \setminus \{C\} \) is adjacent to neither \( u \) nor \( v \). Thus, \( C[V(C) \setminus M] \) is indeed a module in \( G \setminus \hat{U} \). \( \square \)

4.4 Sieving Bad Cliques

We sieve cliques based on three classifications. First, we say that a clique \( C \in \mathcal{C} \) is big if \( |V(C)| > \gamma k^{2/3} \), and otherwise it is small. Furthermore, we say that a clique \( C \in \mathcal{C} \) is huge if \( |V(C)| > 3k \). Recall that by Reduction Rule 4.2, \( |C| \leq 6k \). Thus, we directly have the following observation.

**Observation 4.1.** The total number of vertices in small cliques in \( \mathcal{C} \) is upper bounded by \( 6 \gamma k^{1/2} \).

Second, we say that a clique \( C \in \mathcal{C} \) is heavy if \( |V(C) \cap M| \geq \frac{1}{3} |V(C)| \), and otherwise it is light. It is clear that the total number of vertices in heavy cliques in \( \mathcal{C} \) is upper bounded by \( \delta |M| \). Thus, by Lemma 4.5, we have the following observation.
Observation 4.2. The total number of vertices in heavy cliques in $C$ is upper bounded by $6d(\alpha + \beta)k^{1+\frac{3}{8}}$.

Third, for a clique $C \in C$ and a vertex $s \in S$, we say that $C$ is visible to $s$ if $|N_G(s) \cap V(C)| \geq 2\eta k^{2/3}$, and otherwise we say that $C$ is hidden from $s$. For a clique $C \in C$, we let $\text{vis}(C)$ denote that set of vertices in $S$ to which $C$ is visible. Moreover, we say that a clique $C \in C$ is visible if $|\text{vis}(C)| \geq \lambda k^{2/3}$, and otherwise we say that it is hidden. To bound the number of visible cliques, we need the following rule.

Reduction Rule 4.4. If there exists a vertex $s \in S$ with at least $\frac{1}{2\eta}k^{1/3} + 2$ cliques in $C$ visible to $s$, then remove $s$ from $G$ and decrement $k$ by 1. The new instance is $(G \setminus s, k - 1)$.

Lemma 4.8. Reduction Rule 4.4 is safe.

Proof. In one direction, it is clear that if $S^*$ is a solution to $(G \setminus s, k - 1)$, then $S^* \cup \{s\}$ is a solution to $(G, k)$. For the other direction, let $S^*$ be a solution to $(G, k)$. Let $A$ denote the set of cliques in $C$ that are visible to $s$. Since $|S^*| \leq k$, $|A| \geq \frac{1}{2\eta}k^{1/3} + 2$ and by the definition of visibility, we have that there necessarily exist two distinct cliques $A, A' \in A$ such that each clique among $A, A'$ has a vertex that is a neighbor of $s$ and does not belong to $S^*$. Since these two vertices together with $s$ form an induced $P_3$ in $G$, we derive that necessarily $s \in S^*$. Therefore, $S^* \setminus \{s\}$ is a solution to $(G \setminus s, k - 1)$. \hfill $\Box$

After we exhaustively apply Reduction Rule 4.4, for every vertex $s \in S$ there exist at most $\frac{1}{2\eta}k^{1/3} + 1 \leq \frac{1}{\eta}k^{1/3}$ cliques in $C$ visible to $s$. Since $|S| \leq 3k$, we derive that there are at most $\frac{|S|\frac{1}{\eta}k^{1/3}}{\lambda k^{2/3}} = \frac{3}{\lambda\eta}k^{2/3}$ visible cliques. Thus, we have the following observation.

Observation 4.3. The total number of vertices in non-huge visible cliques in $C$ is upper bounded by $\frac{9}{\lambda\eta}k^{1+\frac{3}{8}}$.

Altogether, we say that a clique $C \in C$ is good if it is (i) big, (ii) light and (iii) hidden or huge (or both), and otherwise we say that it is bad. We denote the set of all good cliques in $C$ by $D$. By Observations 4.1, 4.2 and 4.3, we derive the following lemma.

Lemma 4.9. The total number of vertices in bad cliques in $C$ is upper bounded by $9(\gamma + \delta(\alpha + \beta) + \frac{1}{\lambda\eta})k^{1+\frac{3}{8}}$.

4.5 Properties of Clique Sides

For all $C \in C$ and $s \in S$, denote $N_C(s) = N_G(s) \cap V(C)$ and $\overline{N}_C(s) = V(C) \setminus N_C(s)$. Notice that for all $C \in C$, $s \in S$, $u \in N_C(s)$ and $v \in \overline{N}_C(s)$, it holds that $s - u - v$ is an induced $P_3$ in $G$. Thus, we have the following observation.

Observation 4.4. Let $S^*$ be a solution to $(G, k)$. Then, for all $C \in C$ and $s \in S$, at least one of the following three conditions holds: (i) $s \in S^*$; (ii) $N_C(s) \subseteq S^*$; (iii) $\overline{N}_C(s) \subseteq S^*$.

For all $C \in C$ and $s \in S$, let $M_C(s)$ denote the set of minimum size among $N_C(s)$ and $\overline{N}_C(s)$ (if they have equal sizes, the choice is arbitrary). We first need to apply the following simple rule.

Reduction Rule 4.5. If there exist $C \in C$ and $s \in S$ such that $|M_C(s)| > k$, then remove $s$ from $G$ and decrement $k$ by 1. The new instance is $(G \setminus s, k - 1)$.

Lemma 4.10. Reduction Rule 4.5 is safe.
Proof. In one direction, it is clear that if \( S^* \) is a solution to \( (G \setminus s, k - 1) \), then \( S^* \cup \{s\} \) is a solution to \( (G, k) \). For the other direction, let \( S^* \) be a solution to \( (G, k) \). Since \( |S^*| \leq k \) and \( |M_C(s)| > k \), we have that both \( N_C(s) \setminus S^* \neq \emptyset \) and \( N_C(s) \setminus S^* \neq \emptyset \). Thus, by Observation 4.4, we have that necessarily \( s \in S^* \). Therefore, \( S^* \setminus \{s\} \) is a solution to \( (G \setminus s, k - 1) \).

Specifically, since for every \( s \in S \) and huge clique \( C \in C \), \( |M_C(s)| \leq k \), we have the following corollary, which exhibits a “vertex cover-like” interaction between \( S \) and huge cliques.

Observation 4.5. Let \( S^* \) be a solution to \( (G, k) \). Then, for every \( s \in S \) and huge clique \( C \in C \), at least one of the following two conditions holds: (i) \( s \in S^* \); (ii) \( M_C(s) \subseteq S^* \).

Next, we prove that a similar result holds also for non-huge cliques given that they are good. To this end, we first prove the following simple lemma.

Lemma 4.11. For all \( s \in L \) and \( C \in \mathcal{D} \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \), it holds that \( C \) is visible to \( s \).

Proof. Let \( s \in L \) and \( C \in \mathcal{D} \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \). Then, by Lemma 4.7, we have that \( V(C) \setminus M \subseteq N_G(s) \). Thus, to prove that \( C \) is visible to \( s \), it is sufficient to show that \( |V(C) \setminus M| \geq 2\eta k^{2/3} \). Since \( C \in \mathcal{D} \), we have that \( C \) is light, and therefore \( |V(C) \setminus M| > (1 - \frac{1}{\delta})|V(C)| \). Moreover, since \( C \) is big, \( |V(C)| > \gamma k^{2/3} \), and hence \( |V(C) \setminus M| > (1 - \frac{1}{\delta})\gamma k^{2/3} \). Since \( (1 - \frac{1}{\delta})\gamma \geq 2\eta \), the proof is completed.

Lemma 4.12. Let \( S^* \) be a solution to \( (G, k) \) of minimum size. Then, for every non-huge clique \( C \in \mathcal{D} \), it holds that \( |V(C) \cap S^*| \leq |V(C) \cap M| + \frac{1}{\alpha - 1}\beta \lambda k^{2/3} \).

Proof. Let \( C \in \mathcal{D} \) be a non-huge clique. Suppose, by way of contradiction, that \( |V(C) \cap S^*| > |V(C) \cap M| + \frac{1}{\alpha - 1}\beta \lambda k^{2/3} \). Define \( S' = (S^* \setminus V(C)) \cup \hat{U} \cup (V(M) \cap V(C)) \cup \hat{\text{vis}}(C) \). By Corollary 4.1 and since \( C \) is a non-huge clique in \( \mathcal{D} \), \( |\hat{U} \setminus S'| \leq \frac{1}{\alpha - 1}\beta \lambda k^{2/3} \) and \( |\hat{\text{vis}}(C)| \leq \lambda k^{2/3} \). Thus, \( |S'| < |S^*| \leq k \). Next we show that \( (V(C)) \setminus S' \) is an isolated clique. Towards this we will show that it has no neighbor in the approximate solution \( S \). The only possible neighbors of \( (V(C)) \setminus S' \) in \( S \) are in \( L \). However, if there exists a vertex \( s \in L \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \), then by Lemma 4.11, it holds that \( C \) is visible to \( s \). This implies that a vertex \( s \in L \) is either in \( \text{vis}(C) \) or \( N(s) \cap V(C) \subseteq M \cap V(C) \). Since, \( S' \) contains \( M \cap V(C) \) \cup \text{vis}(C) \) we have that \( (V(C)) \setminus S' \) is an isolated clique. Thus, by Lemma 4.11, the graph \( G \setminus S' \) consists of an isolated clique on the vertex set \( (V(C)) \setminus S' \) and a subgraph of \( G \setminus S^* \). Therefore, as \( G \setminus S^* \) does not contain any induced \( P_3 \), so does \( G \setminus S' \). This implies that \( S' \) is a solution to \( (G, k) \), but since \( |S'| < |S^*| \), we obtain a contradiction to the choice of \( S^* \).

Lemma 4.13. Let \( S^* \) be a solution to \( (G, k) \) of minimum size. Then, for every \( s \in S \) and non-huge clique \( C \in \mathcal{D} \), at least one of the following two conditions holds: (i) \( s \in S^* \); (ii) \( M_C(s) \subseteq S^* \).

Proof. Let \( s \) be a vertex in \( S \), and let \( C \in \mathcal{D} \) be a non-huge clique. Suppose, by way of contradiction, that neither \( s \in S^* \) nor \( M_C(s) \subseteq S^* \). By Observation 4.4, we necessarily have that \( V(C) \setminus M \subseteq S^* \). Thus, \( |V(C) \cap S^*| \geq |V(C) \setminus M_C(s)| \geq \frac{1}{\beta} |V(C)| \). Therefore, to obtain a contradiction to Lemma 4.12, it is sufficient to show that \( \frac{1}{2} |V(C)| > |V(C) \cap M| + \frac{1}{\alpha - 1}\beta \lambda k^{2/3} \). Since \( C \) is light, we have that \( |V(C) \cap M| < \frac{1}{\beta} |V(C)| \), and therefore it remains to show that \( \left( \frac{1}{2} - \frac{1}{\beta} \right) |V(C)| > \frac{1}{\alpha - 1}\beta \lambda k^{2/3} \). Since \( C \) is big, \( |V(C)| > \gamma k^{2/3} \). Thus, we only need to show that \( \left( \frac{1}{2} - \frac{1}{\beta} \right) \gamma > \frac{1}{\alpha - 1}\beta \lambda \), which follows from the definition of \( \alpha, \beta, \gamma, \delta \) and \( \lambda \).
4.6 Expansion with Respect to Clique Sides

We construct the bipartite graph $B'$ by setting one side of the bipartition to be $S$ and the other side $Q'$ to be the set of vertices in good cliques (i.e., $Q' = \bigcup_{C \in D} V(C)$), such that there exists an edge between $s \in S$ and $v \in Q'$ if and only if $v \in M_D(s)$ where $D$ is the clique in $D$ containing $v$. Let $I$ denote the set of isolated vertices in $B'$ that belong to $Q'$, and denote $Q = Q' \setminus I$. Moreover, define $B = B' \setminus I$. Clearly, no clique in $Q$ is an isolated vertex in $B$. We thus proceed by presenting the following rule, where we rely on the Expansion Lemma (Lemma 3.1). It should be clear that the conditions required to apply the algorithm provided by this lemma are satisfied.

**Reduction Rule 4.6.** If $|Q| \geq \left(\frac{1}{(\alpha-1)\beta}k^{2/3} + 1\right)|S|$, then call the algorithm provided by Lemma 3.1 to compute sets $X \subseteq S$ and $Y \subseteq Q$ such that $X$ has a $\left(\frac{1}{(\alpha-1)\beta}k^{2/3} + 1\right)$-expansion into $Y$ in $B$ and $N_B(Y) \subseteq X$. The new instance is $(G \setminus X, k - |X|)$.

We now argue that this rule is safe.

**Lemma 4.14.** Reduction Rule 4.6 is safe.

**Proof.** In one direction, it is clear that if $S^*$ is a solution to $(G \setminus X, k - |X|)$, then $S^* \cup X$ is a solution to $(G, k)$. For the other direction, let $S^*$ be a solution to $(G, k)$ of minimum size. Define $S' = (S^* \setminus Y) \cup X \cup \hat{U}$. First, due to Corollary 4.1, note that $|\hat{U} \setminus U^*| \leq \frac{1}{(\alpha-1)\beta}k^{2/3}$. Moreover, by Observation 4.5 and Lemma 4.13, for every vertex $s \in S \setminus S^*$, it holds that $N_B(s) \subseteq S^*$. Thus, since $X$ has a $\left(\frac{1}{(\alpha-1)\beta}k^{2/3} + 1\right)$-expansion into $Y$ in $B$, we have that $|Y \setminus S^*| \leq \left(\frac{1}{(\alpha-1)\beta}k^{2/3} + 1\right)|X \setminus S^*|$. This implies that $|S'| \leq |S^*| \leq k$.

Notice that if $G \setminus S'$ does not contain any induced $P_3$, then since $X \subseteq S^*$ and we have shown that $|S'| \leq k$, this would imply that $S'$ is a solution to $(G \setminus X, k - |X|)$. Suppose, by way of contradiction, that $G \setminus S'$ contains some induced $P_3$, which we denote by $W$. Note that $V(W) \cap (X \cup \hat{U}) = \emptyset$. Since $G \setminus S'$ does not contain any induced $P_3$ and since $S$ is an approximate solution, we also derive that $V(W) \cap X = \emptyset$ and $V(W) \cap S = \emptyset$. Accordingly, the following case analysis is exhaustive.

- **Case 1:** $W = s - u - v$ where $s \in S \setminus (X \cup \hat{U})$, $u, v \in V(C)$ for some $C \in D$ and $\{u, v\} \nsubseteq X$. In this case, let $y \in \{u, v\}$ denote some vertex in $\{u, v\} \cap Y$ and let $x$ denote the other vertex in $\{u, v\}$ (which might also be in $Y$). Since $N_B(Y) \subseteq X$ and $x \notin X$, we have that $y \notin N_B(s)$. Since $u \in N_C(s)$ and $v \notin N_C(s)$, we have that $|M_C(s) \cap \{u, v\}| = 1$. Since $y \notin N_B(s)$, we have that $y \notin M_C(s)$ and $x \in M_C(s)$. In particular, as $N_B(Y) \subseteq X$ and $x \notin X$, we have that $x \in N_B(s) \setminus Y$ (in fact, $N_B(s) \cap Y = \emptyset$). By Observation 4.5 and Lemma 4.13, we derive that $S^* \cap \{s, x\} \neq \emptyset$. However, as $x \notin Y$, this implies that $S' \cap \{s, x\} \neq \emptyset$, which is a contradiction.

- **Case 2:** $W = s - v - s'$ where $s, s' \in S \setminus (X \cup \hat{U})$, and $v \in V(C) \cap Y$ for some $C \in D$. Since $N_B(Y) \subseteq X$ and $s, s' \notin X$, we have that $v \notin M_C(s) \cup M_C(s')$, which means that $N_C(s) = V(C) \setminus M_C(s)$ and $N_C(s') = V(C) \setminus M_C(s')$. Therefore, $|N_C(s) \cap V(C)|, |N_C(s') \cap V(C)| \geq \frac{1}{2}|V(C)|$. Thus, since $|M \cap V(C)| < \frac{1}{3}|V(C)| < \frac{1}{2}|V(C)|$ (because $C \in D$), we have that there exist $w \in N_C(s) \setminus M$ and $w' \in N_C(s') \setminus M$. By Lemma 4.7 and since $s, s' \notin \hat{U}$, we derive that $C$ is huge or both $V(C) \setminus M \subseteq N_C(s)$ and $V(C) \setminus M \subseteq N_C(s')$.

Let us first consider the subcase where $C$ is huge. Due to Reduction Rule 4.5, we have that $|N_C(s)|, |N_C(s')| \leq k$. Since $|V(C)| > 3k$, we derive that $|N_C(s) \cap N_C(s')| \geq k + 1$. Note that any vertex $w \in N_C(s) \cap N_C(s')$, along with $s$ and $s'$, forms the induced $P_3$ in $G$ that is $s - w - s'$. Note that $s, s' \notin S^*$, as otherwise $\{s, s'\} \cap S' = \emptyset$, which contradicts the
choice of \( W \). Thus, as \( S^* \) is a solution to \((G, k)\), it must hold that \( N_C(s) \cap N_C(s') \subseteq S^* \), but as \(|N_C(s) \cap N_C(s')| \geq k + 1\), this is a contradiction.

Let us now consider the subcase where \( C \) is not huge, and in particular \( V(C) \setminus M \subseteq N_C(s) \) and \( V(C) \setminus M \subseteq N_C(s') \). Then, any vertex \( w \in V(C) \setminus M \), along with \( s \) and \( s' \), forms the induced \( P_3 \) in \( G \) that is \( s - w - s' \). Again, note that \( s, s' \notin S^* \). Thus, since \( S^* \) is a solution to \((G, k)\), we have that \( V(C) \setminus M \subseteq S^* \). Now, recall that by Lemma 4.12 and since \( C \in \mathcal{D} \) is not huge, \(|V(C) \cap S|^* \leq |V(C) \cap M| + \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\). Hence, \(|V(C) \setminus M| \leq \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\). As \(|V(C) \cap M| < \frac{1}{\delta} |V(C)|\) (because \( C \in \mathcal{D} \)), we have that \((1 - \frac{1}{\delta})|V(C)| < \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\), and hence \(|V(C)| < \frac{1}{\delta - 1} \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\), which is a contradiction, since \( C \in \mathcal{D} \) implies that \( C \) is in particular big and \( \gamma \geq \frac{\delta}{\delta - 1} \left( \frac{1}{\alpha - 1} + \lambda \right) \).

**Case 3:** \( W = v - s - s' \) where \( s, s' \in S \setminus (X \cup \hat{U}) \), and \( v \in V(C) \cap Y \) for some \( C \in \mathcal{D} \). The analysis of this case is similar to the one of the previous case, and is only given for completeness. Since \( N_B(Y) \subseteq X \) and \( s, s' \notin X \), we have that \( v \notin M_C(s) \cup M_C(s') \), which means that \( N_C(s) = V(C) \setminus M_C(s) \) and \( N_C(s') = V(C) \setminus M_C(s') \). Therefore, \(|N_C(s) \cap V(C)|, |N_C(s') \cap V(C)| \geq \frac{1}{\delta} |V(C)|\) (because \( C \in \mathcal{D} \)), we have that there exist \( w \in N_C(s) \setminus M \) and \( w' \in N_C(s') \setminus M \). By Lemma 4.7 and since \( s, s' \notin \hat{U} \), we derive that \( C \) is huge or both \( V(C) \setminus M \subseteq N_C(s) \) and \( V(C) \setminus M \subseteq N_C(s') \).

Let us first consider the subcase where \( C \) is huge. Due to Reduction Rule 4.5, we have that \(|N_C(s)|, |N_C(s')| \leq k \). Since \(|V(C)| > 3k\), we derive that \(|N_C(s) \cap N_C(s')| \geq k + 1\). Note that any vertex \( w \in N_C(s) \cap N_C(s') \), along with \( s \) and \( s' \), forms the induced \( P_3 \) in \( G \) that is \( w - s - s' \). Note that \( s, s' \notin S^* \), as otherwise \( \{s, s'\} \cap S^* \neq \emptyset \), which contradicts the choice of \( W \). Thus, as \( S^* \) is a solution to \((G, k)\), it must hold that \( N_C(s) \cap N_C(s') \subseteq S^* \), but as \(|N_C(s) \cap N_C(s')| \geq k + 1\), this is a contradiction.

Let us now consider the subcase where \( C \) is not huge, and in particular \( V(C) \setminus M \subseteq N_C(s) \) and \( V(C) \setminus M \subseteq N_C(s') \). Then, any vertex \( w \in V(C) \setminus M \), along with \( s \) and \( s' \), forms the induced \( P_3 \) in \( G \) that is \( w - s - s' \). Again, note that \( s, s' \notin S^* \). Thus, since \( S^* \) is a solution to \((G, k)\), we have that \( V(C) \setminus M \subseteq S^* \). Now, recall that by Lemma 4.12 and since \( C \in \mathcal{D} \) is not huge, \(|V(C) \cap S^*| \leq |V(C) \cap M| + \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\). Hence, \(|V(C) \setminus M| \leq \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\). As \(|V(C) \cap M| < \frac{1}{\delta} |V(C)|\) (because \( C \in \mathcal{D} \)), we have that \((1 - \frac{1}{\delta})|V(C)| < \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\), and hence \(|V(C)| < \frac{1}{\delta - 1} \left( \frac{1}{\alpha - 1} + \lambda \right) k^{2/3}\), which is a contradiction, since \( C \in \mathcal{D} \) implies that \( C \) is in particular big and \( \gamma \geq \frac{\delta}{\delta - 1} \left( \frac{1}{\alpha - 1} + \lambda \right) \).

**Case 4:** \( W = u - s - v \) where \( s \in S \setminus (X \cup \hat{U}) \), \( u \in V(C) \cap Y \) for some \( C \in \mathcal{D} \), and \( v \in V(C') \) for some \( C' \in C \setminus \{C \} \). Since \( N_B(Y) \subseteq X \) and \( s \notin X \), we have that \( u \notin M_C(s) \), which means that \( N_C(s) = V(C) \setminus M_C(s) \). Therefore, \(|N_C(s) \cap V(C)| \geq \frac{1}{2} |V(C)|\). Thus, since \(|M \cap V(C)| < \frac{1}{\delta} |V(C)| < \frac{1}{\delta} |V(C)|\) (because \( C \in \mathcal{D} \)), we have that there exists \( w \in N_C(s) \setminus M \). By Lemma 4.7 and since \( s \notin \hat{U} \), we derive that \( C \) is huge or \( V(C) \setminus M \subseteq N_C(s) \). Symmetrically, we derive that if \( v \in Y \), then \( C' \) is huge or \( V(C') \setminus M \subseteq N_C(s) \).

Note that for all \( w \in N_C(s) \) and \( w' \in N_C(s') \), it holds that \( w - s - w' \) is an induced \( P_3 \) in \( G \). As \( s \notin S^* \) (as otherwise \( s \in S^* \)), we have that \( N_C(s) \subseteq S^* \) or \( N_C(s') \subseteq S^* \). Observe that if \( v \notin Y \), then since \( v \notin S' \), we have that \( v \notin S^* \) and therefore it clearly holds that \( N_C(s) \subseteq S^* \). If \( v \in Y \) (which means that \( C' \in \mathcal{D} \)), then the proof that \( N_C(s) \subseteq S^* \) is symmetric to the proof that \( N_C(s) \subseteq S^* \). Therefore, in what follows, we only show that \( N_C(s) \subseteq S^* \).

Let us first consider the subcase where \( C \) is huge. Due to Reduction Rule 4.5, we have that \(|N_C(s)| \leq k \). Since \(|V(C)| > 3k\), we derive that \(|N_C(s)| \geq 2k + 1 \). Since \(|S^*| \leq k \), it
is then clear that \( N_C(s) \not\subseteq S^* \). Now, let us now consider the subcase where \( C \) is not huge, and in particular \( V(C) \setminus M \subseteq N_C(s) \). Recall that by Lemma 4.12 and since \( C \in \mathcal{D} \) is not huge, \( |V(C) \cap S^*| \leq |V(C) \cap M| + (\frac{1}{(\alpha - 1)\beta} + \lambda)k^{2/3} \). Suppose, by way of contradiction, that \( N_C(s) \subseteq S^* \). Then, \( |V(C) \setminus M| \leq (\frac{1}{(\alpha - 1)\beta} + \lambda)k^{2/3} \), which leads to a contradiction as in the previous two cases.

Since each case led to a contradiction, the proof is complete. \qed

4.7 Reduction of Almost Modules

At this point, it remains to bound the size of \( I \). We first show that the sets of vertices in \( I \), defined according to the cliques in \( C \), are modules also with respect to \( \hat{U} \). More precisely, we prove the following lemma.

Lemma 4.15. For every clique \( C \in \mathcal{D} \), \( C[I \cap V(C)] \) is a module in \( G \).

Proof. Let \( C \) be a clique in \( \mathcal{D} \). Consider two vertices \( u,v \in I \cap V(C) \). Clearly, every vertex in \( C \) is adjacent to both \( u \) and \( v \), and every vertex in a clique in \( C \cap \{C\} \) is adjacent to neither \( u \) nor \( v \). Thus, \( C[I \cap V(C)] \) is indeed a module in \( G \setminus S \). Now, consider some vertex \( s \in S \). Then, as \( u,v \in I \cap V(C) \), we have that \( u,v \in V(C) \setminus M_C(s) \), because otherwise \( u \) or \( v \) would have been adjacent to \( s \) in the bipartite graph \( B \) of Section 4.6. Thus, we have that either both \( u,v \in N_C(s) \) or both \( u,v \in \overline{N}_C(s) \). As the choices of \( u,v \) and \( s \) were arbitrary, we conclude that \( C[I \cap V(C)] \) is indeed a module in \( G \). \qed

We now present a rule that concerns the set \( I \).

Reduction Rule 4.7. If there exists a visible clique \( C \in \mathcal{D} \) such that \( |I \cap V(C)| > k + 1 \) or a hidden clique \( C \in \mathcal{D} \) such that \( |I \cap V(C)| > |M \cap V(C)| + \frac{1}{(\alpha - 1)\beta}k^{2/3} + \lambda k^{2/3} \), then remove an arbitrarily chosen vertex \( v \in V(C) \cap I \) from \( G \). The new instance is \((G \setminus v,k)\).

Lemma 4.16. Reduction Rule 4.7 is safe.

Proof. In one direction, it is clear that if \((G,k)\) has a solution, so does \((G \setminus v,k)\). Now, let \( S^* \) be a solution to \((G \setminus v,k)\). If \( S^* \) is also a solution to \((G,k)\), then the proof is complete. Therefore, we next assume that \( S^* \) is not a solution to \((G,k)\). Then, there exists an induced \( P_3 \), denoted by \( W \), in \( G \setminus S^* \). Since \( S^* \) is a solution to \((G \setminus v,k)\), \( v \in V(W) \). Furthermore, since \( v \in V(C) \cap I \) and \( I \cap V(C) \) is a clique that is a module (by Lemma 4.15), for any vertex \( u \in I \cap V(C) \), the vertex set \( (V(W) \setminus \{v\}) \cup \{u\} \) induces a \( P_3 \) in \( G \setminus v \). As \( (V(W) \setminus \{v\}) \cup \{v\} \) is a solution to \((G \setminus v,k)\), we deduce that \((I \cap V(C)) \setminus \{v\} \subseteq S^* \).

In case \(|I \cap V(C)| > k + 1 \), the conclusion that \((I \cap V(C)) \setminus \{v\} \subseteq S^* \) implies that \(|S^*| > k \), which is a contradiction. Now, suppose that \( C \) is a hidden clique in \( \mathcal{D} \) such that \(|I \cap V(C)| > |M \cap V(C)| + \frac{1}{(\alpha - 1)\beta}k^{2/3} + \lambda k^{2/3} \). Let us denote \( S' = (S^* \setminus (I \cap V(C))) \cup (M \cap V(C)) \cup \hat{U} \cup \text{vis}(C) \). By Corollary 4.1 and since \( C \) is a hidden clique in \( \mathcal{D} \), we have that \(|S'| \leq |S^*| - |I \cap V(C)| + (\frac{1}{(\alpha - 1)\beta}k^{2/3} + \lambda k^{2/3}) \leq |S^*| \leq k \). Moreover, by Lemma 4.11, the graph \( G[S'] \) consists of an isolated clique on the vertex set \( V(C) \setminus S' \) (for a detail argument see the proof of Lemma 4.12) and a subgraph of \((G \setminus v) \setminus S^* \). Therefore, as \((G \setminus v) \setminus S^* \) does not contain any induced \( P_3 \), so does \( G \setminus S' \). This implies that \( S' \) is a solution to \((G,k)\), and therefore \((G,k)\) is a Yes-instance. \qed

Finally, after the exhaustive application of Reduction Rule 4.7, we can bound the size of \( I \).

Lemma 4.17. After the exhaustive application of Reduction Rule 4.7, \(|I| \leq 6(\frac{1}{\alpha \beta} + \alpha + \beta + \frac{1}{(\alpha - 1)\beta} + \lambda)k^{1/2} \).
Thus, the total number of vertices is indeed $O(V_{\text{FVST}})$.

Theorem 2. In this section, we prove the following theorem.

Feedback Vertex Set in Tournaments

(4.16) can be applied in polynomial time, it strictly decreases the size of

Proof of Theorem 1. Let $(G, k)$ be an instance of CVD. Our kernelization algorithm simply applies (exhaustively) Reduction Rules 4.1 to 4.7. The output is the instance obtained once none of these rules is applicable. Let us observe that each rule among Reduction Rules 4.1 to 4.7 applies (exhaustively) Reduction Rules 4.1 to 4.7. The output is the instance obtained once

Proof of Theorem 1. Let $(G, k)$ be an instance of CVD. Our kernelization algorithm simply applies (exhaustively) Reduction Rules 4.1 to 4.7. The output is the instance obtained once none of these rules is applicable. Let us observe that each rule among Reduction Rules 4.1 to 4.7 can be applied in polynomial time, it strictly decreases the size of $G$ and it does not increase $k$. Thus, our kernelization algorithm runs in polynomial time.

For the sake of clarity, let us now abuse notation and denote the outputted instance by $(G, k)$. Let us observe that $V(G)$ consists of the following vertices.

- Vertices in $S$, whose number is at most $3k$.
- Vertices in bad cliques, whose number is at most $9(g + \delta + 1) k^{1/2} = \mathcal{O}(k^{1/2})$ (by Lemma 4.9).
- Vertices in good cliques that are not isolated in $B'$, whose number is at most $\frac{1}{\lambda(\alpha - 1)^2} k^{2/3} + 1 |S| = \mathcal{O}(k^{1/2})$ (due to Reduction Rule 4.6).
- Vertices in the set $I$, whose number is at most $6(\frac{1}{\lambda} + \lambda) k^{1/2} = \mathcal{O}(k^{1/2})$ (by Lemma 4.17).

Thus, the total number of vertices is indeed $\mathcal{O}(k^{1/2})$. This completes the proof.

4.8 Proof of Theorem 1

We are finally ready to present the proof of Theorem 1.

Proof of Theorem 1. Let $(G, k)$ be an instance of CVD. Our kernelization algorithm simply applies (exhaustively) Reduction Rules 4.1 to 4.7. The output is the instance obtained once none of these rules is applicable. Let us observe that each rule among Reduction Rules 4.1 to 4.7 can be applied in polynomial time, it strictly decreases the size of $G$ and it does not increase $k$. Thus, our kernelization algorithm runs in polynomial time.

For the sake of clarity, let us now abuse notation and denote the outputted instance by $(G, k)$. Let us observe that $V(G)$ consists of the following vertices.

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- Vertices in good cliques that are not isolated in $B'$, whose number is at most $\frac{1}{\lambda(\alpha - 1)^2} k^{2/3} + 1 |S| = \mathcal{O}(k^{1/2})$ (due to Reduction Rule 4.6).
- Vertices in the set $I$, whose number is at most $6(\frac{1}{\lambda} + \lambda) k^{1/2} = \mathcal{O}(k^{1/2})$ (by Lemma 4.17).

Thus, the total number of vertices is indeed $\mathcal{O}(k^{1/2})$. This completes the proof.

5 Feedback Vertex Set in Tournaments

In this section, we prove the following theorem.

Theorem 2. FVST admits a kernel with $\mathcal{O}(k^{3/2})$ vertices.

To prove Theorem 2, we will also use the following folklore result.

Proposition 5.1. Let $T$ be a tournament. Then, the following conditions hold.

1. $T$ has a directed cycle if and only if $T$ has a directed triangle.

2. If $T$ is acyclic then it has a unique topological ordering. That is, there exists a unique ordering $\prec$ of the vertices of $T$ such that for every directed arc $uv$, we have $u \prec v$ (that is, $u$ appears before $v$ in the ordering $\prec$).

Let $(T, k)$ be an instance of FVST. By Proposition 5.1, to find a set $S$ such that $T \setminus S$ is a directed acyclic graph, it is sufficient to find a set that intersects all the triangles of $T$. This immediately yields a simple polynomial-time 3-approximation algorithm for FVST. Indeed, start by greedily finding a maximal collection, say $S$, of vertex-disjoint triangles in $T$ and
Definition 5.1. Let we need the following notion of distance.

Observation 5.1. If there exists \( s \in S \) such that there are \( k + 1 \) triangles intersecting pairwise (only) at \( s \), then remove \( s \) from \( T \). The new instance is \((T \setminus \{s\}, k - 1)\).

We apply Reduction Rule 5.1 exhaustively. Note that each application can be performed in polynomial time, as for any vertex \( s \in S \), we can check whether that exist \( k + 1 \) triangles intersecting pairwise (only) at \( s \) as follows: we construct a bipartite graph where one side of the bipartition is the set \( A \) of in-neighbors of \( s \), the other side of the bipartition is the set \( B \) of out-neighbors of \( s \), and there exists an edge between \( a \in A \) and \( b \in B \) if and only if \( a \) is an out-neighbor of \( b \); then, there exist \( k + 1 \) triangles intersecting pairwise (only) at \( s \) and only if the size of a maximum matching in this bipartite graph is at least \( k + 1 \) (which can be checked in polynomial time). Thus, from now onwards, we assume that Reduction Rule 5.1 is no longer applicable. Throughout this section, we work with the unique ordering \( \prec \) of the vertices of \( X \). For example, whenever we will use a phrase such as the vertices are consecutive in \( X \), we mean that the vertices occur consecutively with respect to the ordering \( \prec \). Similarly, we define the notion of the smallest and the largest vertex in \( X \) according to the ordering \( \prec \).

5.1 Exploring the Vertex Cover Structure

Let us now define a notion of vertex cover for a set of arcs of \( T \). Formally, for a subset of arcs \( A \subseteq E(T) \), a subset \( O \subseteq V(T) \) is called a vertex cover for \( A \) if for every arc \( uv \in A \), either \( u \in O \) or \( v \in O \) (or both). An arc \( xy \) of \( T \) is called strong if (i) at least one vertex among \( x \) and \( y \) belongs to \( S \), and (ii) there are at least \( k + 2 \) vertices \( z \in V(T) \) such that \( xyz \) is a triangle.

Let \( F \) be the set of all the strong arcs of \( T \), which can be easily found in polynomial time. We start our analysis with the following simple observation regarding the set \( F \).

Observation 5.1. If \( O \) is a solution to \((T, k + 1)\), then \( O \) is a vertex cover of \( F \).

The proof is simple: if \( O \) does not hit \( xy \in F \), then \( O \) contains all \( z \in V(T) \) such that \( xyz \) is a triangle, i.e. \( |O| \geq k + 2 \), which is a contradiction.

Recall that throughout our kernelization algorithm, we work with the unique topological ordering \( \prec \) of \( X \). Accordingly, we have that if \( xx' \) is an arc in \( E(X) \), then \( x \prec x' \). Furthermore, we need the following notion of distance.

Definition 5.1. Let \( x, x' \in X \) be two vertices such that \( x \prec x' \), and let \( d - 1 \) be the number of vertices \( y \) such that \( x \prec y \prec x' \). Then, the distance between \( x \) and \( x' \) is \( d \). Accordingly, \( x' - x := d \) and \( x - x' := -d \).

In addition, we need the following definition which concerns the relations between the vertices in \( S \) and the vertices in \( X \).

Definition 5.2. For \( s \in S \) and \( x \in V(X) \), define \( f^-(s)(x) = |\{y \in V(X) : y \preceq x, \ sy \in E(T)\}| \), and \( f^+(s)(x) = |\{y \in V(X) : y \succeq x, \ ys \in E(T)\}| \).
Intuitively, the functions $f_s^-(x)$ and $f_s^+(x)$ measure how many arcs would have been in the “wrong direction” (with respect to the ordering $<$) if we inserted $s$ into the position immediately after $x$ in $X$. First, for every $s \in S$ we would like to find $x_s \in X$ such that $f_s^-(x_s)$ and $f_s^+(x_s)$ are almost equal.

**Lemma 5.1.** For each $s \in S$, there exists $x_s \in V(X)$ such that $0 \leq f_s^-(x_s) - f_s^+(x_s) \leq 1$.

**Proof.** Let $x_m$ be the smallest vertex in $V(X)$, and $x_M$ be the largest vertex in $V(X)$. Fix some $s \in S$. In what follows, we omit the subscript $s$. We have the following two inequalities:

- $f^-(x_M) - f^+(x_M) \geq 0$ (since $f^+(x_M) = 0$), and
- $f^-(x_m) - f^+(x_m) \leq 1$ (since $f^-(x_m) \leq 1$).

Let $x, x' \in V(X)$ where $x' = x + 1$. Then, $f^-(x') - f^+(x') = f^-(x) - f^+(x) + 1$. That is, the function $f^-(x) - f^+(x)$ increases by 1 whenever $x$ increases by 1. Indeed, observe that if $sx' \in E(T)$, then $f^-(x') = f^-(x) + 1$ and $f^+(x') = f^+(x)$. Otherwise, $x's \in E(T)$, and so $f^-(x') = f^-(x)$ and $f^+(x') = f^+(x) - 1$. Thus, the two inequalities above, and the fact that the function $f^-(x) - f^+(x)$ increases by 1 whenever $x$ increases by 1, together imply that there exists $x_s \in V(X)$ such that $0 \leq f^-(x_s) - f^+(x_s) \leq 1$. \hfill \Box

For the sake of clarity, we extract the implication of Lemma 5.1 to the following notation.

**Definition 5.3.** For any $s \in S$, define $\varphi(s)$ as the smallest vertex $x_s \in V(X)$ satisfying the inequalities in Lemma 5.1.

We now show that given Reduction Rule 5.1, neither $f_s^-(\varphi(s))$ nor $f_s^+(\varphi(s))$ can be too “large”. Indeed, if there existed $s \in S$ such that $f_s^-(\varphi(s)) \geq k + 2$, then $f_s^+(\varphi(s)) \geq k + 1$, and we could have formed $k + 1$ triangles, each consisting of $s$, a vertex from $\{x \in V(X) : x \preceq \varphi(s), sx \in E(T)\}$, and a vertex from $\{y \in V(X) : y \succ \varphi(s), ys \in E(T)\}$. In this case, Reduction Rule 5.1 is applicable. However, as we assumed that Reduction Rule 5.1 is no longer applicable, we have that for all $s \in S$, $f_s^-(\varphi(s)), f_s^+(\varphi(s)) \leq k + 1$. By using this assumption, we have useful certificates for strong arcs as follows.

**Lemma 5.2.** Let $x \in X$, and $s, s' \in S$. The following statements are true.

1. If $sx \in E(T)$ and $\varphi(s) - x \geq 2k + 3$, then $sx$ is strong.
2. If $xs \in E(T)$ and $x - \varphi(s) \geq 2k + 3$, then $xs$ is strong.
3. If $s's \in E(T)$ and $\varphi(s') - \varphi(s) \geq 3k + 5$, then $s's$ is strong.

**Proof.** We first prove [1]. As $\varphi(s) - x \geq 2k + 3$, there are at least $2k + 2$ vertices between $x$ and $\varphi(s)$. Since $f_s^-(\varphi(s)) \leq k + 1$, we have $|\{y : x < y \leq \varphi(s), sy \in E(T)\}| \leq k + 1$. Hence, the set $R = \{y : x < y \varphi(s), ys \in E(T)\}$ has at least $k + 2$ vertices. Note that $sxy$ is a triangle for each $y \in R$ since $sx \in E(T)$. This shows that $sx$ is strong. The proof of [2] is similar.

To prove [3], we note that $f_s^+(\varphi(s)) \leq k + 1$ and $f_s^-(\varphi(s)) \leq k + 1$. That is, $|\{x : x < \varphi(s), sx \in E(T)\}| \leq k + 1$, and $|\{x : \varphi(s) < x \varphi(s'), xs \in E(T)\}| \leq k + 1$. Since there are at least $3k + 4$ vertices between $\varphi(s)$ and $\varphi(s')$, this implies that $|\{x : \varphi(s) < x \preceq \varphi(s'), sx, xs' \in E(T)\}| \geq 2k + 2$, i.e. $s's$ is strong. \hfill \Box

To proceed, we also need to introduce two terms concerning triangles.

**Definition 5.4.** Let $x_1x_2x_3$ be a triangle of $T$, and $A = \{x_1, x_2, x_3\}$. The span of $x_1x_2x_3$ is the maximum distance between any two vertices in $(A \setminus S) \cup \varphi(A \cap S)$. Moreover, the triangle is called local if none of its arcs belongs to $F$. 21
In the following lemma, we will show that a local triangle is indeed local in the sense that it must have a “short” span.

**Lemma 5.3.** Let \( x_1x_2x_3 \) be a local triangle with at least one vertex from \( X \). Then, its span is at most \( 6k + 8 \).

**Proof.** For \( 1 \leq i \leq 3 \), define

\[
\varphi'(x_i) = \begin{cases} x_i, & \text{if } x_i \in V(X), \\ \varphi(x_i) & \text{otherwise.} \end{cases}
\]

If the claim is false, then

\[
\max\{|\varphi'(x_1) - \varphi'(x_2)|, |\varphi'(x_2) - \varphi'(x_3)|, |\varphi'(x_3) - \varphi'(x_1)|\} \geq 6k + 9.
\]

By symmetry, we may assume that \( |\varphi'(x_1) - \varphi'(x_2)| \geq 6k + 9 \). We first claim that (* there is an index \( i \in [3] \) such that \( \varphi'(x_1) - \varphi'(x_{i+1}) \geq 3k + 5 \) (where the calculation \( i + 1 \) is modulo 3). Indeed, if \( \varphi'(x_1) - \varphi'(x_2) \geq 6k + 9 \), then (*) is true. Therefore, next suppose that \( \varphi'(x_2) - \varphi'(x_1) \geq 6k + 9 \). If \( \varphi'(x_3) > \varphi'(x_2) \), then \( \varphi'(x_3) - \varphi'(x_1) > 6k + 9 \), and then (*) is true. Moreover, if \( \varphi'(x_3) < \varphi'(x_1) \), then \( \varphi'(x_2) - \varphi'(x_3) > 6k + 9 \), and then (*) is true. Hence, we next suppose that \( \varphi'(x_1) \prec \varphi'(x_3) \prec \varphi'(x_2) \). Then, as \( \varphi'(x_3) - \varphi'(x_1) > 6k + 9 \), we have that either \( \varphi'(x_3) - \varphi'(x_1) \geq 3k + 5 \) or \( \varphi'(x_1) - \varphi'(x_3) \geq 3k + 5 \), so (*) is true. This proves (*).

Let \( i \in [3] \) be an index satisfying (*), that is, \( \varphi'(x_i) - \varphi'(x_{i+1}) \geq 3k + 5 \). If \( x_i, x_{i+1} \in V(X) \), then since \( x_i x_{i+1} \in E(T) \), we have that \( x_i \prec x_{i+1} \), which contradicts that \( \varphi'(x_i) - \varphi'(x_{i+1}) \geq 3k + 5 \) is positive. Thus, at least one vertex among \( x_i \) and \( x_{i+1} \) is in \( S \). However, then Lemma 5.2 implies that \( x_i x_{i+1} \) is a strong arc, which contradicts the fact that \( x_1 x_2 x_3 \) is a local triangle. \( \square \)

### 5.2 Applying the Double Expansion Lemma

In what follows, we denote \( \alpha = 3, \beta = 20, \gamma = 7, \mu = 3, \delta = 2 \), and \( \ell = 3 \) so that \( \beta - 13 \geq \mu \delta \) (used in Observation 5.3), \( \gamma \mu > 6 \ell \) (used in Observation 5.5), \( \frac{1}{2} + \frac{1}{3} < 1 \) and \( \ell - 1 \geq \delta \) (used in the proof of Lemma 5.4).

In order to proceed with our analysis, we need to classify “intervals” of vertices from \( X \) as either good or bad, depending on how many vertices from \( S \) are mapped into these intervals. Formally, we have the following definition.

**Definition 5.5.** A set \( Y \subseteq V(X) \) is an interval if it contains all the vertices in \( X \) that lie between the largest and smallest elements in \( Y \) (with respect to the ordering \( \prec \) induced by \( X \)).

We refer to \( |Y| \) as the length of \( Y \). Moreover, \( Y \) is good if the size of \( S_Y = \{ s \in S \mid \varphi(s) \in Y \} \) is at most \( \alpha \sqrt{k} \), and otherwise it is bad.

Note that for two interval \( Y, Y' \subseteq V(X) \), if \( Y \cap Y' = \emptyset \), then \( S_Y \cap S_{Y'} = \emptyset \) as well.

We partition \( V(X) \) into disjoint intervals, each of length \( \beta k \). That is, we follow the vertices of \( V(X) \) from left to right in the ordering \( \prec \), and partition them into disjoint intervals \( Y_1^*, ..., Y_p^* \) such that each \( Y_j^* \), \( 1 \leq j \leq p \), is of length \( \beta k \). Note that among \( Y_1^*, ..., Y_p^* \), at most \( \frac{3k}{\alpha \sqrt{k}} = \frac{3 \sqrt{k}}{\alpha} \) intervals are bad; otherwise, \( |S| > (\frac{3 \sqrt{k}}{\alpha} + 1) \cdot \alpha \sqrt{k} > 3k \), which contradicts our assumption that \( |S| \leq 3k \). Thus, we have the following upper bound on the number of bad intervals among \( Y_1^*, ..., Y_p^* \).

**Observation 5.2.** There are at most \( \frac{3 \sqrt{k}}{\alpha} \) bad intervals among \( Y_1^*, ..., Y_p^* \).

\( ^6 \) That is, the elements of \( Y \) are consecutive with respect to \( \prec \).
Thus, if \( p \geq (\frac{3}{\alpha} + \beta)\sqrt{k} \), there are at least \( p - \frac{3\sqrt{k}}{\alpha} \geq \beta\sqrt{k} \) good intervals. Consider the first \( \gamma\sqrt{k} \) good intervals among \( Y^*_1, \ldots, Y^*_p \), and rename them as \( Y_1, \ldots, Y_{\gamma\sqrt{k}} \) according to the order of the appearance (by \( < \) of their vertices). The fact that the relative order of the intervals is preserved will be used later. For all \( i \in [\gamma\sqrt{k}] \), denote \( S_i = S_{Y_i} \) (recall that \( S_{Y_i} = \{ s \in S \mid \varphi(s) \in Y_i \} \)), and let \( Y'_i \) be the sub-interval of \( Y_i \) excluding the \( 6k + 9 \) largest and the \( 6k + 9 \) smallest vertices of \( Y_i \). The purpose of this exclusion is to ensure that the vertex set of any local triangle hit by \( Y'_i \) (that is, the triangle contains at least one vertex of \( Y'_i \)) is completely contained in \( Y_i \cup S_i \) (see Lemma 5.6).

**Observation 5.3.** For all \( i \in [\gamma\sqrt{k}] \), the length of \( Y'_i \) is at least \( \beta k - 2(6k + 9) > (\beta - 13)k \geq \mu\delta k \).

We now ready apply the Double Expansion Lemma. One naive idea is to construct a bipartite graph \( G \) with vertex set \( (\bigcup Y'_i, S) \) and every \( H_i \) with vertex set \( (Y'_i, S_i) \). However, this attempt does not work out mainly because we have too little information about the edge set of \( H_i \) to exploit. To overcome this, we chop down every \( Y'_i \) into sub-intervals \( Y_{i,j} \)'s, and we merge each \( Y_{i,j} \) into a single “representative” vertex \( a_{i,j} \) and we put an edge between \( a_{i,j} \) and \( s \) in \( H_i \) if the arcs between \( s \) and \( Y_{i,j} \) have different orientations. Precisely, the construction of \( G \) and \( H_i \) is as follows.

We first partition each \( Y'_i \) into \( \mu\gamma\sqrt{k} \) sub-intervals, \( Y_{i,1}, \ldots, Y_{i,\mu\gamma\sqrt{k}} \), each of length \( \delta\sqrt{k} \) such that \( x < x' \) for every \( x \in Y_{i,j}, x' \in Y_{i,j'} \) with \( j < j' \). We now construct the bipartite graphs \( G, H_1, \ldots, H_{\gamma\sqrt{k}} \). To this end, for all \( i, 1 \leq i \leq \gamma\sqrt{k} \), define \( A_i = \{ a_{i,1}, \ldots, a_{i,\mu\gamma\sqrt{k}} \} \), and \( A = \bigcup_{i=1}^{\gamma\sqrt{k}} A_i \). Then, \( |A| = \gamma\sqrt{k} \cdot \mu\sqrt{k} = \gamma\mu k \). It is useful to think of \( a_{i,j} \) as the representative of the sub-interval \( Y_{i,j} \) for every \( i, j \). Let us now define the bipartite graphs (see Figure 4 for an illustration).

![Figure 4: Construction of G from F (upper figure) and construction of H_i from T[S_i ∪ Y'_i] (lower figure). Not all arcs of T[S_i ∪ Y'_i] are shown; for arcs not shown in T[S_i ∪ Y'_i], their corresponding (possibly) edges in H_i are dotted.](image-url)

1. \( G \): The (undirected) bipartite graph with vertex set \( (A, S) \) and edge set \( E(G) = \{ a_{i,j}s : \exists x \in Y'_i \text{ such that } \{xs, sx\} \cap F \neq \emptyset \} \). This is to take care of strong arcs.
2. \( H_i \): The (undirected) bipartite graph with vertex set \( (A_i, S_i) \) and edge set \( E(H_i) = \{ a_{i,j}s : \exists x, x' \text{ such that } sx, sx' \in E(T) \} \). In other words, \( a_{i,j}s \notin E(H_i) \) if and only if either \( sx \in E(T) \) for every \( x \in Y_{i,j} \) or \( xs \in E(T) \) for every \( x \in Y_{i,j} \). This is to take care of local triangles.

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Before applying the Double Expansion Lemma, we mention here an observation for later use, which is the main purpose of our “merging” vertices into representative.

**Observation 5.4.** If \( sa_{i,j}, sa_{i,j'} \in E(H_i) \) for some \( j < j' \), then there is a triangle \( sxx' \) with \( x \in Y_{i,j}, x' \in Y_{i,j'} \).

The proof is trivial: by definition of \( E(H_i) \), there is \( x \in Y_{i,j} \) such that \( sx \in E(T) \), and there is \( x' \in Y_{i,j'} \) such that \( x's \in E(T) \). Since \( i < j' \), \( x < x' \), and so \( xx' \in E(T) \). Thus, \( sxx' \) is a triangle.

By applying the Double Expansion Lemma 3.3, in polynomial time, we find \( \hat{A} \subseteq A \), \( \hat{S} \subseteq S \), as well as \( \hat{A}_i \subseteq A_i \) and \( \hat{S}_i \subseteq S_i \) for all \( 1 \leq i \leq \gamma \sqrt{k} \), such that

- \( \hat{A} = \bigcup_{i=1}^{\gamma \sqrt{k}} \hat{A}_i \);
- \( |A \setminus \hat{A}| \leq 2|S| \);
- \( \hat{S} \) has an \( \ell \)-expansion into \( \hat{A} \) in \( G \), and \( N_G(\hat{A}) \subseteq N_G(\hat{S}) \);
- \( \hat{S}_i \) has an \( \ell \)-expansion into \( \hat{A}_i \) in \( H_i \), and \( N_{H_i}(\hat{A}_i) \subseteq N_{H_i}(\hat{S}_i) \).

Let \( \hat{Y} = \bigcup_{a_{i,j} \in \hat{A}} Y_{i,j} \) and \( \hat{Y}_i = \bigcup_{a_{i,j} \in \hat{A}_i} Y_{i,j} \). Since \( \hat{A} = \bigcup_{i=1}^{\gamma \sqrt{k}} \hat{A}_i \), we have \( \hat{Y} = \bigcup_{i=1}^{\gamma \sqrt{k}} \hat{Y}_i \).

**Observation 5.5.** \( \hat{Y} \) is nonempty.

*Proof.* Recall that \( |S| \leq 3k \) and \( |A| = \gamma \mu k \). Since \( |A \setminus \hat{A}| \leq 2\ell|S| \), we have \( |\hat{A}| \geq |A| - 2\ell|S| \geq \gamma \mu k - 2\ell \cdot 3k > 0 \). Since \( \hat{A} \neq \emptyset \), there exists \( a_{i,j} \in \hat{A} \), and so \( \hat{Y} \supseteq Y_{ij} \neq \emptyset \). \( \square \)

### 5.3 Using Expansion to Detect an Irrelevant Vertex

Let \( O \) be a solution to \( (T, k+1) \), and define

\[
O' = \left( O \setminus \hat{Y} \right) \cup \left( \hat{S} \cup \bigcup_{i=1}^{\gamma \sqrt{k}} S_i' \right),
\]

where

\[
S_i' = \begin{cases} 
\hat{S}_i & \text{if } |O \cap \hat{Y}_i| < \delta \sqrt{k}, \text{ and} \\
S_i & \text{otherwise.}
\end{cases}
\]

In the rest of this subsection, we show that if \( O \cap \hat{Y} \neq \emptyset \), then \( |O'| < |O| \) and \( O' \) is a solution to \( (T, k+1) \).

**Lemma 5.4.** If \( O \cap \hat{Y} \neq \emptyset \), then \( |O'| < |O| \).

*Proof.* Observe that, to obtain \( O' \) from \( O \), we remove \( O \cap \hat{Y} \), and add \( \hat{S} \setminus O \) and \( \bigcup_{i=1}^{\gamma \sqrt{k}} (S_i' \setminus O) \).

We will prove that

\[
\frac{|O \cap \hat{Y}|}{\ell} \geq |\hat{S} \setminus O|, \quad \text{and} \tag{1}
\]

\[
\frac{|O \cap \hat{Y}|}{\delta} \geq \left| \bigcup_{i=1}^{\gamma \sqrt{k}} (S_i' \setminus O) \right|, \quad \text{and} \tag{2}
\]

Combining (1) and (2) with \( \frac{1}{\ell} + \frac{1}{\delta} < 1 \) and the hypothesis of the lemma that \( |O \cap \hat{Y}| > 0 \), we have

\[
|O \cap \hat{Y}| > |\hat{S} \setminus O| + \left| \bigcup_{i=1}^{\gamma \sqrt{k}} (S_i \setminus O) \right|, 
\]

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which implies that $|O'| < |O|$, proving the lemma.

To prove (1), recall that $\hat{S}$ has an $\ell$-expansion into $\hat{A}$ in $G$, so $|N_G(\hat{S} \setminus O) \cap \hat{A}| \geq \ell |\hat{S} \setminus O|$. Thus, it suffices to show that $|O \cap \hat{Y}| \geq |N_G(\hat{S} \setminus O) \cap \hat{A}|$. Suppose for a contradiction that $|O \cap \hat{Y}| < |N_G(\hat{S} \setminus O) \cap \hat{A}|$. Then

$$\sum_{a_{i,j} \in N_G(\hat{S} \setminus O) \cap \hat{A}} |O \cap Y_{i,j}| \leq \sum_{a_{i,j} \in \hat{A}} |O \cap Y_{i,j}| = |O \cap \hat{Y}| < |N_G(\hat{S} \setminus O) \cap \hat{A}|.$$  

(3)

If $|O \cap Y_{i,j}| \geq 1$ for every $a_{i,j} \in N_G(\hat{S} \setminus O) \cap \hat{A}$, then $\sum_{a_{i,j} \in N_G(\hat{S} \setminus O) \cap \hat{A}} |O \cap Y_{i,j}| \geq |N_G(\hat{S} \setminus O) \cap \hat{A}|$, contradicting (3). Thus we conclude that there exists $a_{i,j} \in N_G(\hat{S} \setminus O) \cap \hat{A}$ such that $O \cap Y_{i,j} = \emptyset$. Let $s \in \hat{S} \setminus O$ such that $s a_{i,j} \in E(G)$ (such a vertex $s$ exists, since $a_{i,j} \in N_G(\hat{S} \setminus O)$). By the definition of $E(G)$, there exists $x \in Y_{i,j}$ such that $s x \in F$. Note that $x \notin O$, since $O \cap Y_{i,j} = \emptyset$, and $s \notin O$, since $s \in \hat{S} \setminus O$. As $O$ is a solution to $(T, k + 1)$, and because of Observation 5.1, $O$ must be a vertex cover of $F$. But $x, s \notin O$, which is a contradiction. From this we conclude that (1) holds.

To prove (2), note that

$$|O \cap \hat{Y}| = \sum_{i=1}^{\gamma \sqrt{k}} |O \cap \hat{Y}_i| \quad \text{and} \quad \sum_{i=1}^{\gamma \sqrt{k}} |S'_i \setminus O| = \left| \bigcup_{i=1}^{\gamma \sqrt{k}} (S'_i \setminus O) \right|.$$

Thus, it suffices to show that $|O \cap \hat{Y}_i| \geq \delta |S'_i \setminus O|$ for every $i$. If $S'_i = S_i$, then $|O \cap \hat{Y}_i| \geq \delta \sqrt{k}$ by the definition of $S'$. Since $Y_i$ is a good interval, $|S_i| \leq \sqrt{k}$. Hence, $|O \cap \hat{Y}_i| \geq \delta \sqrt{k} \geq \delta |S_i| \geq \delta |S'_i \setminus O|$. Now suppose $S'_i = \hat{S}_i$. Since $\hat{S}_i$ has an $\ell$-expansion into $\hat{A}_i$ in $H_i$, we have $|N_{H_i}(\hat{S}_i \setminus O) \cap \hat{A}_i| \geq \ell |\hat{S}_i \setminus O|$. Call $a_{i,j} \in N_{H_i}(\hat{S}_i \setminus O) \cap \hat{A}_i$ pure, if $Y_{i,j} \cap O = \emptyset$. Observe that if $s \in \hat{S}_i \setminus O$ is adjacent to two pure vertices in $H_i$, say $a_{i,j}$ and $a_{i,j'}$ with $j < j'$, then by the definition of $E(H_i)$, then by Observation 5.4 there is a triangle $s x x'$ with $x \in Y_{i,j}$ and $x' \in Y_{i,j'}$, and so $s x x'$ is not hit by $O$ by definition of purity, which contradicts the assumption that $O$ is a feedback vertex set for $T$. Thus, each $s \in \hat{S}_i \setminus O$ is adjacent to at most one pure vertex, i.e. there are at most $|\hat{S}_i \setminus O|$ pure vertices. Thus, the number of non-pure vertices is at least (recall that $\ell - 1 \geq \delta$)

$$|N_{H_i}(\hat{S}_i \setminus O) \cap \hat{A}_i| - |\hat{S}_i \setminus O| \geq (\ell - 1) |\hat{S}_i \setminus O| \geq \delta |\hat{S}_i \setminus O|.$$  

For each non-pure $a_{i,j} \in N_{H_i}(\hat{S}_i \setminus O) \cap \hat{A}_i$, we have $|Y_{i,j} \cap O| \geq 1$, by the definition of purity. Recall that $|O \cap \hat{Y}_i| = \sum_{a_{i,j} \in \hat{A}_i} |O \cap Y_{i,j}|$. Thus, $|O \cap \hat{Y}_i|$ is at least the number of non-pure vertices, i.e. $|O \cap \hat{Y}_i| \geq \delta |\hat{S}_i \setminus O|$. As $S'_i = \hat{S}_i$, we have $|O \cap \hat{Y}_i| \geq \delta |S'_i \setminus O|$, and this proves (2).

It remains to show that $O'$ is a solution to $(T, k + 1)$. To do so, we will prove that $O'$ is a vertex cover of $F$ and $O'$ hits all local triangles.

**Lemma 5.5.** $O'$ is a vertex cover of $F$.

**Proof.** By Observation 5.1, $O$ is a vertex cover of $F$, so every $s s' \in F$ with $s, s' \in S$ is hit by $O$, and hence $s s'$ is hit by $O'$, since $O \cap S \subseteq O' \cap S$. Thus, we only need to show that every $x s \in F$ with $x \in V(X)$ and $s \in S$ is hit by $O'$. Suppose for contradiction that $x s \in F$ is not hit by $O'$. Then either $x \in O \setminus O'$ or $s \in O \setminus O'$. Note that since $O \cap S \subseteq O' \cap S$, $s \notin O \setminus O'$. So $x \in O \setminus O'$, which implies that $x \in O \cap \hat{Y}$. Let $x \in Y_{i,j}$. Then $a_{i,j} \in \hat{A}$, and since $x s \in F$, $a_{i,j} s \in E(G)$. Recall (from the list properties obtained after applying the Double Expansion Lemma) that $N_G(\hat{A}) \subseteq \hat{S}$, which implies that $s \in \hat{S}$. However, $\hat{S} \subseteq O'$ by the definition of $O'$. This implies that $s \in O'$, a contradiction to the assumption that $x s$ is not hit by $O'$. This concludes the proof of the claim. \[\square\]
Recall that a triangle is local if it has no strong arcs.

**Lemma 5.6.** If \(xyz\) is a local triangle with \(x \in O \cap \hat{Y}_i\), then \(O'\) hits \(xyz\).

**Proof.** Suppose, for a contradiction, that \(O'\) does not hit \(xyz\). By Lemma 5.3, \(xyz\) has span at most \(6k + 8\). Note that \(x \in \hat{Y}_i \subseteq Y'_i\), while \(Y'_i\) is obtained from \(Y_i\) by excluding \(6k + 9k\) smallest and \(6k + 9k\) largest vertices, so \(\{(x, y, z) \cap X\} \cup \{x, y, z\} \cap S\) is a subset of \(Y_i\). In other words, \(x, y, z \in S_i \cup Y_i\). At least one of \(y, z\) belongs to \(S\) (otherwise, \(xyz\) is transitive), so at least one of \(y, z\) belongs to \(S_i\). We consider two cases.

**Case 1:** \(y \in S_i\). If \(|O \cap \hat{Y}_i| \geq \sqrt{k}\), then \(S_i = S'_i \subseteq O'\), and so \(y \in O'\), a contradiction. We conclude that \(|O \cap \hat{Y}_i| < \sqrt{k}\). If \(y \in \hat{S}_i\), then \(y \in O'\), a contradiction again. Hence \(y \in S_i \setminus \hat{S}_i\). Let \(x \in Y_{i,j}\), then \(a_{i,j} \in \hat{A}\). Recall that \(N_{H_i}(A_i) \subseteq \hat{S}_i\), and so \(ya_{i,j} \notin E(H_i)\). Thus, by definition of \(E(H_i)\), we have \(x'y \in E(T)\) for every \(x' \in Y_{i,j}\), since \(x \in E(T)\).

If \(z \notin O\), then there is \(x' \in Y_{i,j}\) such that \(x'z \in E(T)\). (4)

To prove (4), assume that \(z \notin O\) and \(xz' \in E(T)\) for every \(x' \in Y_{i,j}\). This implies that \(x'yz\) is a triangle for every \(x' \in Y_{i,j}\). Since \(O'\) does not hit \(xyz\), we have \(y \notin O'\), and so \(y \notin O\) (since \(O \cap S \subseteq O' \cap S\)). However, \(O\) is a solution to \((T, k + 1)\), while \(y, z \notin O\), so \(x' \in O\) for every \(x' \in Y_{i,j}\), i.e., \(Y_{i,j} \subseteq O\). Since \(Y_{i,j} \subseteq \hat{Y}_i\), we have \(|O \cap \hat{Y}_i| \geq |Y_{i,j}| = \sqrt{k}\), a contradiction to the observation \(|O \cap \hat{Y}_i| < \sqrt{k}\) (made at the beginning of Case 1), which proves (4).

Note that \(z \in S_i \cup Y_i\). We now consider all possibilities of \(z\).

- If \(z \in \hat{S}_i\), then clearly \(z \in O'\) since \(\hat{S}_i \subseteq O'\), a contradiction.
- If \(z \in S_i \setminus \hat{S}_i\), then recall that \(N_{H_i}(A_i) \subseteq \hat{S}_i\), and so \(a_{i,j} \notin E(H_i)\). Hence by definition of \(E(H_i)\), we have \(z x' \in E(T)\) for every \(x' \in Y_{i,j}\) (since \(z x \in E(T)\)). If \(z \in O\), then \(z \in O'\) (since \(O \cap S \subseteq O' \cap S\)), a contradiction. Then \(z \notin O\), which contradicts (4).
- We conclude that \(z \notin S_i\), i.e., \(z \in \hat{Y}_i\).
- If \(z \in Y_i\) and \(z \notin O\), since \(yz \in E(T)\) while \(x'y \in E(T)\) for every \(x' \in Y_{i,j}\), we have \(z \notin Y_{i,j}\). Since \(z x \in E(T)\), we have \(z \prec x\), and so \(z \prec x'\) for every \(x' \in Y_{i,j}\). In other words, \(z x' \in E(T)\) for every \(x' \in Y_{i,j}\), a contradiction to (4).
- Otherwise, \(z \in Y_i\) and \(z \in O\). Then \(z \in \hat{Y}_i\) since \(z \notin O'\). Let \(z \in Y_{i,j'}\), then \(a_{i,j'} \in \hat{A}\). Observe that \(j \neq j'\) since \(x'y \in E(T)\) for every \(x' \in Y_{i,j}\) while \(y z \notin E(T)\). Since \(y \in S \setminus \hat{S}_i\), and recall that \(N_{H_i}(A_i) \subseteq \hat{S}_i\), we have \(y a_{i,j} \notin E(H_i)\), and so \(y z' \in E(T)\) for every \(z' \in Y_{i,j'}\) (since \(yz \in E(T)\)). If there are \(x' \in Y_{i,j}, z' \in Y_{i,j'}\) such that \(x'z' \notin O\), then \(x'z'\) is not hit by \(O\), a contradiction. Then either \(Y_{i,j} \subseteq O\) or \(Y_{i,j'} \subseteq O\), so

\[
|O \cap \hat{Y}_i| \geq |O \cap (Y_{i,j} \cup Y_{i,j'})| \geq \min(|Y_{i,j}|, |Y_{i,j'}|) = \sqrt{k},
\]

a contradiction to the observation \(|O \cap \hat{Y}_i| < \sqrt{k}\) at the beginning of Case 1.

We conclude that in all cases, \(O'\) always hits \(xyz\).

**Case 2:** \(z \in S_i\). The argument is similar as for Case 1. \(\square\)

Using Lemmas 5.3 to 5.6, we derive the following result.

**Lemma 5.7.** \(O'\) is a solution to \((T, k + 1)\).

**Proof.** Suppose that \(O'\) is not a solution to \((T, k + 1)\). Then there is a triangle \(xyz\) which is not hit by \(O'\). Note that \(O\) is a solution to \((T, k + 1)\), and \(O \setminus O' \subseteq O \cap \hat{Y}_i\), so at least one vertex of \(xyz\) belongs to \(O \cap hy\), say \(x\). Let \(x \in \hat{Y}_i\), i.e., \(x \in O \cap \hat{Y}_i\). If one of the arcs of \(xyz\) belongs to \(F\), then \(O'\) hits \(xyz\), by Lemma 5.5, which is a contradiction. So, \(xyz\) is local, and \(O'\) hits \(xyz\), by Lemma 5.6, which again contradicts our assumption. \(\square\)
From Lemmas 5.4 to 5.7, we now conclude that if \( O \) is a solution to \((T, k + 1)\), and \( O \cap \hat{Y} \neq \emptyset \), then there is another solution \( O' \) to \((T, k + 1)\) with \( |O'| \leq |O| - 1 \). Therefore, we have the following reduction rule to remove an irrelevant vertex.

**Reduction Rule 5.2.** Let \( x \) be an arbitrary vertex in \( \hat{Y} \). Remove \( x \) from \( T \). The new instance is \((T \setminus \{x\}, k)\).

**Lemma 5.8.** Reduction rule 5.2 is safe.

**Proof.** In one direction, it is clear that if \( S^* \) is a solution to \((T, k)\), then \( S^* \) is a solution to \((T \setminus \{x\}, k)\). For the other direction, let \( S^* \) be a solution to \((T \setminus \{x\}, k)\). Then \( O = S^* \cup \{x\} \) is a solution to \((T, k + 1)\). Since \( x \in O \cap \hat{Y} \), we have \( O \cap \hat{Y} \neq \emptyset \), and so there is a solution \( O' \) to \((T, k + 1)\) with \( |O'| \leq |O| - 1 = (|S^*| + 1) - 1 \leq k \). Thus, \( O' \) is a solution to \((T, k)\). \( \square \)

### 5.4 Proof of Theorem 2

We are finally ready to present the proof of Theorem 2.

**Proof of Theorem 2.** Let \((T, k)\) be an instance of \(FVST\). Our kernelization algorithm simply applies (exhaustively) Reduction Rules 5.1 and 5.2. The output is the instance obtained once none of these rules is applicable. Let us observe that each of Reduction Rules 5.1 and 5.2 can be applied in polynomial time, it strictly decreases the size of \( G \) and it does not increase \( k \). Thus, our kernelization algorithm runs in polynomial time.

For the sake of clarity, let us now abuse notation and denote the outputted instance by \((T, k)\). Let us observe that \( V(T) \) consists of the following vertices.

- Vertices in \( S \), whose number is at most \( 3k \).
- Vertices of \( X \), whose number is at most \( p\beta \sqrt{k} = O(k^{3/2}) \) since \( p \leq (\frac{3}{\alpha} + \beta) \sqrt{k} \).

Thus, the total number of vertices is indeed \( O(k^{3/2}) \). This complete the proof. \( \square \)

### 6 Kernel for Induced \( P_3 \)-Packing

In this section, we prove the following theorem.

**Theorem 3.** Induced \( P_3 \)-Packing admits a kernel with \( O(k^{1/2}) \) vertices.

Our kernel for Induced \( P_3 \)-Packing is based on the kernel for CVD. In fact, several of the steps of both the kernelization algorithms are almost the same, but the subtle differences between them are crucial. Specifically, while in CVD we analyze properties that must be satisfied by all solutions, in Induced \( P_3 \)-Packing we analyze properties such that there exists a solution that satisfies them (if there exists a solution at all). As we progress with the description of our kernelization algorithm for Induced \( P_3 \)-Packing, the deviations from the kernelization algorithm for CVD become more palpable; in particular, the later proofs of both algorithms are completely different (for example, here we we do not even construct the bipartite graph \( B' \) as we did in Section 4.6).

Let \((G, k)\) be an instance of Induced \( P_3 \)-Packing. We start by greedily finding a maximal collection, say \( S \), of vertex-disjoint induced \( P_3 \)’s in \( G \). Clearly, this greedy procedure can be run in polynomial time. If \(|S| \geq k\), then we conclude that \((G, k)\) is a \( \text{Yes} \)-instance. Thus, we next suppose that \(|S| < k\). Let \( S \) be the set of vertices that belong to the induced \( P_3 \)'s in \( S \). Since \(|S| < k\), we have that \(|S| \leq 3k\). Notice that \( G \setminus S \) is a collection of cliques, which we denote by \( C \).
In what follows, we denote \( \alpha = 2, \beta = 1, \gamma = 43, \mu = 26, \delta = 3, \lambda = 1 \) and \( \eta = 1 \), so that \((1 - \frac{1}{4})\gamma \geq 6\eta \) (used in the proof of Lemma 6.11), \( \frac{\delta-1}{\sigma} \mu - \frac{(20}{(\sigma-1)\beta} > 3 \) (used in the proof of Lemma 6.13), \( \frac{\delta-1}{\sigma} \gamma \geq \frac{20}{(\sigma-1)\beta} + \lambda \) (used in the proof of Lemma 6.13), \( \frac{\delta}{\sigma} \geq 3 \) (used in the proof of Lemma 6.14), and \( \frac{\delta}{(\sigma-1)\beta} + \lambda + \frac{\gamma}{2} \leq \frac{\gamma}{2} \) (used in the proof of Lemma 6.14).

6.1 Bounding the Number of Cliques

First, as in the case of CVD, we have the following simple rule, whose safeness is obvious.

Reduction Rule 6.1. If there exists \( C \in \mathcal{C} \) such that no vertex in \( C \) has a neighbor in \( S \), then remove \( C \) from \( G \). The new instance is \((G \setminus C, k)\).

Now, also in the case of CVD, we define the bipartite graph \( B \) by setting one side of the bipartition to be \( S \) and the other side to be \( \mathcal{C} \), such that there exists an edge between \( s \in S \) and \( C \in \mathcal{C} \) if and only if \( s \) is adjacent to at least one vertex in \( C \). Note that by Reduction Rule 6.1, no clique in \( \mathcal{C} \) is an isolated vertex in \( B \). We thus proceed by presenting the following rule (which is slightly different than Reduction Rule 4.2), where we rely on the Expansion Lemma (Lemma 3.1). It should be clear that the conditions required to apply the algorithm provided by this lemma are satisfied.

Reduction Rule 6.2. If \(|C| \geq 2|S|\), then call the algorithm provided by Lemma 3.1 to compute sets \( X \subseteq S \) and \( \mathcal{Y} \subseteq \mathcal{C} \) such that \( X \) has a 2-expansion into \( \mathcal{Y} \) in \( B \) and \( N_B(\mathcal{Y}) \subseteq X \). The new instance is \((G \setminus (X \cup V(\mathcal{Y})), k - |X|)\). Here, \( V(\mathcal{Y}) = \bigcup_{C \in \mathcal{Y}} V(C) \).

We now argue that this rule is safe.

Lemma 6.1. Reduction Rule 6.2 is safe.

Proof. For every vertex \( s \in X \), let \( C_s \) and \( C'_s \) be the two clique assigned to \( s \) by the 2-expansion. Notice that for all \( s \in X \), there exists an induced \( P_3 \) in \( G \) of the form \( u_s - s - v_s \), where \( u_s \) is any neighbor of \( s \) in \( C_s \) (as \( s \) and \( C_s \) are neighbors in \( B \), at least one such vertex exists), and \( v_s \) is any neighbor of \( s \) in \( C'_s \) (again, at least one such vertex exists). Let this special collection of induced \( P_3 \)'s be denoted by \( \mathcal{X}^* \), that is \( \mathcal{X}^* = \{u_s - s - v_s : s \in X\} \). In one direction, it is clear that if \( \mathcal{S}^* \) is a solution to \((G \setminus (X \cup V(\mathcal{Y})), k - |X|)\), then \( \mathcal{S}^* \cup \mathcal{X}^* \) is a solution to \((G, k)\). For the other direction, let \( \mathcal{S}^* \) be a solution to \((G, k)\). Let \( \mathcal{W} \) denote the set of every induced \( P_3 \) in \( \mathcal{S}^* \) that contains at least one vertex from \( X \). We denote \( \mathcal{S}' = (\mathcal{S}^* \setminus \mathcal{W}) \cup \mathcal{X}^* \). Observe that since \( N_B(\mathcal{Y}) \subseteq X \), we have that no induced \( P_3 \) in \( \mathcal{S}^* \setminus \mathcal{W} \) contains any vertex from \( V(\mathcal{Y}) \cup X \). Thus, it holds that \( \mathcal{S}' \) is a collection of induced \( P_3 \)'s in \( G \). Since \( |\mathcal{W}| \leq |X| \), we have that \( |\mathcal{S}'| \geq k \).

We conclude that \( \mathcal{S}' \) is a solution to \((G, k)\), and as \( \mathcal{X}^* \subseteq \mathcal{S}' \), we have that \( \mathcal{S}' \setminus \mathcal{X}^* \) is a solution to \((G \setminus (X \cup V(\mathcal{Y})), k - |X|)\). Thus, \((G \setminus (X \cup V(\mathcal{Y})), k - |X|)\) is a \textbf{Yes}-instance. \( \square \)

Due to Reduction Rule 6.2, from now on \(|C| \leq 6k\).

6.2 The Specification of the Marking Procedure

We proceed by presenting a procedure called Mark. The specification of this procedure is similar to one presented in Section 4.2. In particular, let us emphasize one subtle difference: now we mark an additional set \( N \), which will be a crucial component of latter rules and arguments.
Proof. If there exists $s \in S$ such that $|\text{mark}(s)| \geq 3k + 1$, then there exist $3k + 1$ induced $P_3$’s in the graph of the form $s - u_i - w_i$, $i \in \{1, \ldots, 3k + 1\}$, that intersect only at $s$. That is, we have a “flower” whose core is $s$ and whose petals are $\{u_i, w_i\}$. In one direction, let $S^*$ be a solution to $(G \setminus s, k - 1)$. Note that $|V(S^*)| \leq 3(k - 1)$. Thus, the number of induced paths of the form $s - u_i - w_i$ that intersect $V(S^*)$ is also upper bounded by $3(k - 1)$. This implies that there exists an induced path $s - u_j - w_j$ that does not contain any vertex from $V(S^*)$. Then, $S^* \cup \{s - u_j - w_j\}$ is a solution to $(G, k)$. For the other direction, let $S^*$ be a solution to $(G, k)$. Observe that there is at most one induced $P_3$ in $S^*$ that contains the vertex $s$. Let $S'$ be the set of induced $P_3$’s obtained by deleting the induced $P_3$ in $S^*$ that contains $s$ (if it exists). Then, $S'$ is a solution to $(G \setminus s, k - 1)$. \hfill \Box

As in the case of CVD, the main purpose of Mark is to derive information on $(G, k)$ when it is not coincidentally lucky. However, the information we require here is different than the one we require in the case of CVD. Not only do we analyze one solution rather than all solutions, we also need to state explicit relations between $U$ and the set of vertices marked by $U$ (that is, the set $N$).

Lemma 6.3. For any induced $P_3$-packing $S'$ of size at most $k$ there exists an induced $P_3$-packing $S^*$ of size at least $|S'|$ such that the following conditions hold.

- Let $P'$ be the set of induced $P_3$’s in $S'$ that do not contain any vertex from $U$. Then, $P' \subseteq S^*$.
- There exists a set $A \subseteq U$ of size at most $\frac{3}{\beta}k^{1/3}$ such that for all $s \in U \setminus A$, there exist $P \in S^*$ and $u, v \in N$ such that $P = s - u - v$.

Proof. Let $S'$ be an induced $P_3$-packing of $G$ of size at most $k$. Observe that for all $s \in S$ and $\{u, v\}, \{u', v'\} \in \text{mark}(s)$, it holds that $\{u, v\} \cap \{u', v'\} = \emptyset$. In addition, observe that for all $s, s' \in S$, $\{u, v\} \in \text{mark}(s)$ and $\{u', v'\} \in \text{mark}(s')$, it holds that $\{u, v\} \cap \{u', v'\} = \emptyset$. As $|V(S')| \leq 3k$ and for all $s \in U$, $|\text{mark}(s)| \geq \lceil \beta k^{2/3} \rceil$, we derive that there exist at most $3k/\lceil \beta k^{2/3} \rceil \leq \frac{3}{\beta}k^{1/3}$ vertices $s \in U$ such that for all $\{u, v\} \in \text{mark}(s)$, $V(S') \cap \{u, v\} \neq \emptyset$. Let $A$ denote the set of these vertices in $U$. Moreover, let $P^*$ be the set of induced $P_3$’s in $S'$ that do not contain any vertex from $U \setminus A$. Notice that $P' \subseteq P^*$. Moreover, notice that $|S' \setminus P^*| \leq |U \setminus A|$. Now, define $\overline{P}$ as the $P_3$-packing obtained by selecting, for every vertex $s \in U \setminus A$, an induced $P_3$ that consists of $s$ and an arbitrarily chosen edge $\{u, v\} \in \text{mark}(s)$ such that $V(S') \cap \{u, v\} \neq \emptyset$ (there exists at least one such edge). Then, $S^* = P^* \cup \overline{P}$ is an induced
contradicts the choice of $S' \setminus P^*$, we derive that $|S^*| \geq |S'|$. Moreover, it is clear from its construction that $S^*$ satisfied the two properties in the statement of the lemma. This completes the proof.

We also need to derive an upper bound on the number of marked vertices, namely $|M|$.

**Lemma 6.4.** If $\text{Mark}$ was neither lucky nor successful, then $|M| \leq 6(\alpha + \beta)k^{1/3}$.

**Proof.** Since $\text{Mark}$ was unlucky, $|\text{mark}(s)| \leq 3k$ for all $s \in S$. Thus, $|M| \leq 2|U|3k + 2|S \setminus U|([\beta k^{2/3}] - 1)$. Since $\text{Mark}$ failed, we further have that $|M| \leq 6(\alpha k^{2/3} - 1)k + 6k([\beta k^{2/3}] - 1) \leq 6(\alpha + \beta)k^{1/3}$. □

### 6.3 Multiple Calls to the Marking Procedure

We employ $\text{Mark}$ exactly as in the case of CVD, with the exception that now we also compute a set $\hat{M}$. For the sake of readability, let us repeat this short description (with the computation of $\hat{M}$). We initialize $\hat{U} = \emptyset$, $\hat{M} = \emptyset$ and $\hat{G} = G$. Then, we call $\text{Mark}$ with $(\hat{G}, k)$ as input. If $\text{Mark}$ was lucky, then we execute Reduction Rule 6.3 and restart the entire process (including the initialization phase). Else, if $\text{Mark}$ succeeded, then for the sets $U$ and $N$ computed by the current call, we update $\hat{U} = \hat{U} \cup U$, $\hat{M} = \hat{M} \cup N$ and $\hat{G} = \hat{G} \setminus U$, and then we proceed to execute another call. Otherwise, $\text{Mark}$ was unlucky and also failed, and we let $M$ denote the same set $M \subseteq V(G) \setminus S$ as computed by the current call to $\text{Mark}$, after which we terminate the process. (It may hold that $M \cap \hat{M} = \emptyset$.) Note that after each call to $\text{Mark}$, either Reduction Rule 6.3 is executed or the size of $\hat{U}$ increases, and therefore it is clear that the process eventually terminates. We denote $L = S \setminus \hat{U}$.

By relying on Lemma 6.3, we have the following lemma.

**Lemma 6.5.** Let $i$ be the number of calls to $\text{Mark}$ that succeeded but were unlucky. If $(G, k)$ is a $\text{Yes}$-instance, then there exists a solution $S^*$ to $(G, k)$ and a set $A \subseteq \hat{U}$ of size at most $i \cdot \frac{3}{\beta}k^{1/3}$ such that for all $s \in \hat{U} \setminus A$, there exists $P \in S^*$ and $u, v \in \hat{M}$ such that $P = s - u - v$.

**Proof.** Suppose that $(G, k)$ is a $\text{Yes}$-instance, and let $S'$ be a solution to $(G, k)$ that minimizes the number of vertices $s \in \hat{U}$ for which there do not exist $P \in S'$ and $u, v \in \hat{M}$ such that $P = s - u - v$. Let $A$ denote the set of these vertices in $\hat{U}$. Suppose, by way of contradiction, that $|A| > i \cdot \frac{3}{\beta}k^{1/3}$. Then, by the pigeonhole principle, there exists an iteration $j \in \{1, 2, \ldots, i\}$ such that $|A \cap U_j| > \frac{3}{\beta}k^{1/3}$, where $U_j$ denotes the set $U$ computed in iteration $j$. By Lemma 6.3, there exists a solution $S^*$ to $(G, k)$ such that the following conditions hold:

- Let $P'$ be the set of induced $P_3$’s in $S'$ that do not contain any vertex from $U_j$. Then, $P' \subseteq S^*$.
- There exists a set $A^* \subseteq U_j$ of size at most $\frac{3}{\beta}k^{1/3}$ such that for all $s \in U_j \setminus A^*$, there exist $P \in S^*$ and $u, v \in \hat{M}$ such that $P = s - u - v$. In fact, $u, v \in N$, the set computed in round $j$.

By the first condition, we deduce that for every $P \in S'$ such that $P = s - u - v$ for some $s \in \hat{U} \setminus U_j$ and $u, v \in \hat{M}$, it also holds that $P \in S^*$. Furthermore, from the second condition we derive that $S^*$ has fewer vertices $s \in U_j$ than $S'$ for which there do not exist $P \in S'$ and $u, v \in \hat{M}$ such that $P = s - u - v$. However, we thus conclude that $S^*$ has fewer vertices $s \in \hat{U}$ than $S'$ for which there do not exist $P \in S'$ and $u, v \in \hat{M}$ such that $P = s - u - v$. Since this contradicts the choice of $S'$, we have that $|A| < i \cdot \frac{3}{\beta}k^{1/3}$. This completes the proof. □

Before we proceed to proceed to present a consequence of Lemma 6.5, we need to present a new rule that is also necessary to upper bound $|\hat{M}|$. 30
Reduction Rule 6.4. Let $i$ be the number of calls to Mark that succeeded but were unlucky. If $i \geq \frac{1}{a-1}k^{1/3}$, then return a trivial Yes-instance.

Lemma 6.6. Reduction Rule 6.4 is safe.

Proof. Let us consider the following simple procedure. Initialize $S_0 = \emptyset$. Now, for $j = 1, 2, \ldots, i$, perform the following computation: Let $S_j$ be the induced $P_3$-packing whose existence is guaranteed by Lemma 6.3 when applied with $S' = S_{j-1}$. (We implicitly assume that induced $P_3$’s that are not of the form $s - u - v$, for $s \in \hat{U}$ and $u, v \in \hat{M}$, are discarded.) By the two properties of $S^*$ as specified by Lemma 6.3, we have that for all $j \in \{1, 2, \ldots, i\}$, $|S_j| \geq |S_{j-1}| + |U_j| - \frac{3}{3}k^{1/3}$, where $U_j$ is the set $U$ computed in iteration $j$. Since for all $j \in \{1, 2, \ldots, i\}$, $|U_j| = \lceil \alpha k^{2/3} \rceil$, we overall have that $|S_i| \geq i \cdot \lceil \alpha k^{2/3} - \frac{3}{3}k^{1/3} \rceil$. Observe that $\frac{k}{\alpha k^{2/3} - \frac{3}{3}k^{1/3}} = \frac{k^{2/3}}{\alpha k^{2/3} - \frac{3}{3}k^{1/3}} \leq \frac{1}{a-1}k^{1/3}$.

Thus, if $i \geq \frac{1}{a-1}k^{1/3}$, then $|S_i| \geq k$, in which case $S_i$ is a solution to $(G, k)$. This implies that Reduction Rule 6.4 is indeed safe.

For the sake of clarity, let us formally define the solutions that we would like to analyze.

Definition 6.1. We say that a pair $(S^*, A)$ is a nice solution to $(G, k)$ if $S^*$ is a solution to $(G, k)$ and $A \subseteq \hat{U}$ is a set of size at most $\frac{3}{(a-1)3}k^{2/3}$ such that for all $s \in \hat{U} \setminus A$, there exist $P \in S^*$ and $u, v \in \hat{M}$ such that $P = s - u - v$.

Now, as a consequence of Lemma 6.5 and Reduction Rule 6.4, we have the following corollary.

Corollary 6.1. If $(G, k)$ is a Yes-instance, then there exists a nice solution to $(G, k)$.

Proof. Suppose that $(G, k)$ is a Yes-instance. Let $i$ be the number of calls to Mark that succeeded but were unlucky. By Lemma 6.5, there exists a solution $S^*$ to $(G, k)$ and a set $A \subseteq \hat{U}$ of size at most $i \cdot \frac{3}{(a-1)3}k^{1/3}$ such that for all $s \in \hat{U} \setminus A$, there exist $P \in S^*$ and $u, v \in \hat{M}$ such that $P = s - u - v$. By Reduction Rule 6.4, we have that $i < \frac{1}{a-1}k^{1/3}$. Therefore, we have that $|A| \leq \frac{1}{a-1}k^{1/3}, \frac{3}{(a-1)3}k^{1/3} = \frac{3}{(a-1)3}k^{2/3}$. We have thus obtained a nice solution $(S^*, A)$ to $(G, k)$.

The usefulness of Corollary 6.1 stems from the observation that it implies that we have found a (possibly large) set $\hat{U} \subseteq \hat{S}$ such that not only there exists a solution that packs almost all the vertices in $\hat{U}$ in induced $P_3$’s with vertices in $\hat{M}$, but also that the removal of $\hat{U}$ from $G$ significantly simplifies $G$ as described by the following lemma. As the proof (and statement) of this lemma is identical to the proof of Lemma 4.7, it is omitted.

Lemma 6.7. For every clique $C \in \mathcal{C}$, $|V(C) \setminus \hat{M}|$ is a module in $G \setminus \hat{U}$.

Before we proceed to sieve bad clique, let us upper bound $|\hat{M}|$.

Lemma 6.8. $|\hat{M}| \leq \frac{2a \beta}{1-a}k^{1/2}$.

Proof. Due to each call to Mark, at most $2[\alpha k^{2/3}] \cdot [\beta k^{2/3}]$ new vertices are inserted into $\hat{M}$. By Reduction Rule 6.4, Mark was called less than $\frac{1}{a-1}k^{1/3}$ times. Thus, the total number of vertices inserted into $\hat{M}$ is upper bounded by $2(\frac{1}{a-1}k^{1/3} - 1) \cdot [\alpha k^{2/3}] \cdot [\beta k^{2/3}] \leq \frac{2a \beta}{1-a}k^{1/2}$.

6.4 Sieving Bad Cliques

We sieve cliques based on three classifications, similar to the case of CVD. First, we say that a clique $C \in \mathcal{C}$ is big if $|V(C)| > \gamma k^{2/3}$, and otherwise it is small. Furthermore, we say that a clique $C \in \mathcal{C}$ is huge if $|V(C)| > \mu k$. Recall that by Reduction Rule 6.2, $|C| \leq 6k$. Thus, as in the case of CVD, we directly have the following observation.
Observation 6.1. The total number of vertices in small cliques in $C$ is upper bounded by $6\gamma k^{1.3}$.

Second, we say that a clique $C \in \mathcal{C}$ is heavy if $|V(C) \cap (M \cup \hat{M})| \geq \frac{1}{6}|V(C)|$, and otherwise it is light. In particular, heaviness is now measured with respect to $M \cup \hat{M}$, while in the case of CVD it was measure only with respect to $M$. It is clear that the total number of vertices in heavy cliques in $C$ is upper bounded by $\delta |M \cup \hat{M}|$. Thus, by Lemmata 6.4 and 6.8, we have the following observation.

Observation 6.2. The total number of vertices in heavy cliques in $C$ is upper bounded by $6\delta (\alpha + \beta + \frac{\alpha \beta}{1 - \alpha}) k^{1.3}$.

Third, as in the case of CVD (except that the constant 2 is replaced by 6), for a clique $C \in \mathcal{C}$ and a vertex $s \in S$, we say that $C$ is visible to $s$ if $|N_G(s) \cap V(C)| \geq 6\eta k^{2/3}$, and otherwise we say that $C$ is hidden from $s$. For a clique $C \in \mathcal{C}$, we let $\text{vis}(C)$ denote that set of vertices in $S$ to which $C$ is visible. Moreover, we say that a clique $C \in \mathcal{C}$ is visible if $|\text{vis}(C)| \geq \lambda k^{2/3}$, and otherwise we say that it is hidden. To bound the number of visible cliques, we need the following rule.

Reduction Rule 6.5. If there exists a vertex $s \in S$ with at least $\frac{1}{2\eta} k^{1.3} + 2$ cliques in $C$ visible to $s$, then remove $s$ from $G$ and decrement $k$ by 1. The new instance is $(G \setminus s, k - 1)$.

Lemma 6.9. Reduction Rule 6.5 is safe.

Proof. In one direction, let $S^*$ be a solution to $(G \setminus s, k - 1)$. Let $A$ denote the set of cliques in $C$ that are visible to $s$. Since $|V(S^*)| \leq 3(k - 1)$, $|A| \geq \frac{1}{2\eta} k^{1.3} + 2$ and by the definition of visibility, we have that there necessarily exist two distinct cliques $A, A' \in A$ such that each clique among $A, A'$ has a vertex that is a neighbor of $s$ and does not belong to $V(S^*)$. Since these two vertices together with $s$ form an induced $P_3$ in $G$, called $P$, we derive that $S^* \cup \{P\}$ is a solution to $(G, k)$. For the other direction, let $S^*$ be a solution to $(G, k)$. Observe that there is at most one induced $P_3$ in $S^*$ that contains the vertex $s$. Let $S'$ be the set of induced $P_3$'s obtained by deleting the induced $P_3$ in $S^*$ that contains $s$ (if it exists). Then, $S'$ is a solution to $(G \setminus s, k - 1)$. \hfill $\square$

After we exhaustively apply Reduction Rule 6.5, as in the case of CVD, for every vertex $s \in S$ there exist at most $\frac{1}{2\eta} k^{1.3} + 1 \leq \frac{1}{7} k^{1.3}$ cliques in $C$ visible to $s$. Since $|S| \leq 3k$, we derive that there are at most $\frac{|S| k^{1.3}}{\lambda \eta k^{2/3}} = \frac{3}{\lambda \eta} k^{2/3}$ visible cliques. Thus, we have the following observation.

Observation 6.3. The total number of vertices in non-huge visible cliques in $C$ is upper bounded by $\frac{3\mu}{\lambda \eta} k^{1.3}$.

Altogether, we say that a clique $C \in \mathcal{C}$ is good if it is (i) big, (ii) light and (iii) hidden or huge (or both), and otherwise we say that it is bad. We denote the set of all good cliques in $C$ by $D$. By Observations 6.1, 6.2 and 6.3, and that $\mu = 26$, we derive the following lemma.

Lemma 6.10. The total number of vertices in bad cliques in $C$ is upper bounded by $3\mu (\gamma + \delta (\alpha + \beta + \frac{\alpha \beta}{1 - \alpha}) + \frac{1}{\lambda \eta}) k^{1.3}$.

### 6.5 Properties of Clique Sides

For all $C \in \mathcal{C}$ and $s \in S$, denote $N_C(s) = N_G(s) \cap V(C)$ and $\overline{N}_C(s) = V(C) \setminus N_C(s)$. Notice that for all $C \in \mathcal{C}$, $s \in S$, $u \in N_C(s)$ and $v \in \overline{N}_C(s)$, it holds that $s - u - v$ is an induced $P_3$ in $G$. Furthermore, for all $C \in \mathcal{C}$ and $s \in S$, let $M_C(s)$ denote the set of minimum size among $N_C(s)$ and $\overline{N}_C(s)$ (if they have equal sizes, the choice is arbitrary). Let us first verify that Lemma 4.11 also holds in the context of INDUCED $P_3$-PACKING.
Lemma 6.11. For all \( s \in L \) and \( C \in \mathcal{D} \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \), it holds that \( C \) is visible to \( s \).

Proof. Let \( s \in L \) and \( C \in \mathcal{D} \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \). Then, by Lemma 6.7, we have that \( V(C) \setminus M \subseteq N_G(s) \). Thus, to prove that \( C \) is visible to \( s \), it is sufficient to show that \( |V(C) \setminus M| \geq 6n^2k^2/3 \). Since \( C \in \mathcal{D} \), we have that \( C \) is light, and therefore \( |V(C) \setminus M| > (1 - \frac{1}{3})|V(C)| \). Moreover, since \( C \) is big, \( |V(C)| > \gamma k^{2/3} \), and hence \( |V(C) \setminus M| > (1 - \frac{1}{3})\gamma k^{2/3} \). Since \( (1 - \frac{1}{3})\gamma \geq 6\eta \), the proof is completed.

Now, let us also explicitly state the following simple corollary to Lemma 6.11

Corollary 6.2. For all non-huge \( C \in \mathcal{D} \), the number of vertices \( s \in L \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \) is upper bounded by \( \lambda k^{2/3} \).

Proof. Let \( C \in \mathcal{D} \) be a non-huge clique. By Lemma 6.11, \( C \) is visible to every vertex \( s \in L \) such that \( N_G(s) \cap (V(C) \setminus M) \neq \emptyset \). Thus, since \( C \) is hidden, the statement is true.

Let us now argue that for any nice solution to \((G, k)\), it holds that for every clique \( C \in \mathcal{D} \), most of the clique \( C \) is “unused”.

Lemma 6.12. Let \((\mathcal{S}^*, A)\) be a nice solution to \((G, k)\). For all non-huge \( C \in \mathcal{D} \), it holds that 
\[
|V(C) \cap V(\mathcal{S}^*)) \setminus (M \cup \hat{M})| \leq \left(\frac{6}{(a - 1)\beta}\right)k^{2/3}.
\]

Proof. Let \( C \in \mathcal{D} \) be a non-huge clique. Since \( C \) is a clique where \( N_G(C) \subseteq S \), every induced \( P_3 \) in \( \mathcal{S}^* \) that contains at least one vertex from \( V(C) \) must also contain at least one vertex from \( S \). Because \((\mathcal{S}^*, A)\) is a nice solution, every induced \( P_3 \) in \( \mathcal{S}^* \) that contains at least one vertex from \( \hat{U} \setminus A \) cannot contain any vertex from \( V(C) \setminus \hat{M} \). Furthermore, since \( |A| \leq \frac{3}{(a - 1)\beta}k^{2/3} \), there exist at most \( 2|A| \leq \frac{6}{(a - 1)\beta}k^{2/3} \) vertices \( v \in V(\mathcal{S}^*) \) for which there exists an induced \( P_3 \) in \( \mathcal{S}^* \) that contains both \( v \) and at least one vertex from \( A \). Now, let us denote the set of induced \( P_3 \)’s in \( \mathcal{S}^* \) that contain at least one vertex from \( V(C) \setminus M \) and no vertex from \( \hat{U} \) by \( P \). Then, we note that every induced path \( P \in \mathcal{P} \) must contain an edge \( \{s, v\} \in E(G) \) for some \( s \in L \) and \( v \in V(C) \), and that \( |V(P) \cap (V(C) \setminus M)| = 1 \) (by Lemma 6.7). By Corollary 6.2, we derive that 
\[
|(V(C) \cap V(P)) \setminus M| \leq \lambda k^{2/3}.
\] This completes the proof.

In order to proceed with our analysis, we need to refine Definition 6.1 with respect to a set of vertices.

Definition 6.2. Let \( T \subseteq V(\mathcal{D}) \setminus (M \cup \hat{M}) \). We say that a pair \((\mathcal{S}^*, A)\) is a \( T \)-nice solution to \((G, k)\) if \((\mathcal{S}^*, A)\) is a nice solution, and for all \( P \in \mathcal{S}^* \) such that \( V(P) \cap \hat{U} = \emptyset \), it holds that \( V(P) \cap T = \emptyset \).

We now claim that that for any small enough set \( T \), it is possible to focus on seeking nice solutions with respect to \( T \). Formally, we prove the following lemma.

Lemma 6.13. Let \( T \subseteq V(\mathcal{D}) \setminus (M \cup \hat{M}) \) be a set of size at most \( \frac{14}{(a - 1)\beta}k^{2/3} \). If \((G, k)\) is a \( \text{Yes-instance} \), then there exists a \( T \)-nice solution to \((G, k)\).

Proof. Suppose that \((G, k)\) is a \( \text{Yes-instance} \). Then, by Corollary 6.1, there exists a nice solution to \((G, k)\). Let \((\mathcal{S}^*, A)\) be a nice solution to \((G, k)\) that minimizes the number of vertices \( v \in T \) for which there exists \( P \in \mathcal{S}^* \) such that \( V(P) \cap \hat{U} = \emptyset \) and \( v \in V(P) \). We claim that there do not exist \( v \in T \) and \( P \in \mathcal{S}^* \) such that \( V(P) \cap \hat{U} = \emptyset \) and \( v \in V(P) \). Suppose, by way of contradiction, that there exist \( v \in T \) and \( P \in \mathcal{S}^* \) such that \( V(P) \cap \hat{U} = \emptyset \) and \( v \in V(P) \). Let \( C \) denote the clique in \( \mathcal{D} \) such that \( v \in V(C) \). We first observe that due to Lemma 6.7 and because \( v \notin M \) and \( V(P) \cap \hat{U} = \emptyset \), if we replace \( v \) in \( P \) by any other
vertex in $V(C) \setminus M$, we obtain yet another induced $P_3$. Thus, by the choice of $(S^*, A)$, we derive that $V(C) \setminus (T \cup M \cup V(S^*)) = \emptyset$. In other words, $V(C) \setminus (T \cup M) \subseteq V(S^*)$. Hence, $|\{V(C) \cap V(S^*)\} \setminus (M \cup \hat{M})| \geq |V(C) \setminus (T \cup M \cup \hat{M})|$. Then, because $C \in D$ and $|T| \leq \frac{\mu}{(a_1-a_2)^2/3}$, we have that $|V(C) \setminus (T \cup M \cup \hat{M})| \geq |V(C) \setminus (M \cup \hat{M})| - \frac{14}{(a_1-a_2)} k^{2/3} > \frac{2}{\delta} |V(C)| - \frac{14}{(a_1-a_2)} k^{2/3}$. Thus, $\frac{2}{\delta} |V(C)| - \frac{14}{(a_1-a_2)} k^{2/3} < |\{V(C) \cap V(S^*)\} \setminus (M \cup \hat{M})|$. Now, let us consider two cases, corresponding to whether or not $C$ is huge.

- **Suppose that $C$ is huge.** Then, $\frac{2}{\delta} |V(C)| - \frac{14}{(a_1-a_2)} k^{2/3} \geq \frac{2}{\delta} \mu k - \frac{14}{(a_1-a_2)} k^{2/3} > 3k$ (because $\frac{2}{\delta} \mu - \frac{14}{(a_1-a_2)} > 3$). However, $|\{V(C) \cap V(S^*)\} \setminus (M \cup \hat{M})| \leq |V(S^*)| \leq 3k$. Thus, we have reached a contradiction.

- **Suppose that $C$ is not huge.** Then, by Lemma 6.12, this means that $|\{V(C) \cap V(S^*)\} \setminus (M \cup \hat{M})| \leq (\frac{6}{(a_1-a_2)} + \lambda) k^{2/3}$. Since $\frac{2}{\delta} |V(C)| - \frac{14}{(a_1-a_2)} k^{2/3} < |\{V(C) \cap V(S^*)\} \setminus (M \cup \hat{M})|$ and $|V(C)| \geq \gamma k^{2/3}$, we have that $\frac{2}{\delta} - \lambda k^{2/3} - \frac{14}{(a_1-a_2)} k^{2/3} < (\frac{6}{(a_1-a_2)} + \lambda) k^{2/3}$. However, since $\frac{20}{(a_1-a_2)} + \lambda \leq \frac{2}{\delta} \gamma$, we have reached a contradiction.

As both cases led to a contradiction, the proof is complete. □

### 6.6 Assigning Sets of Vertices to Vertices in $\hat{U}$

For every vertex $s \in \hat{U}$, denote $Q'(s) = \bigcup_{C \in D}(M_C(s) \setminus (M \cup \hat{M}))$. Moreover, for every vertex $s \in \hat{U}$, if $|Q'(s)| \leq \frac{6}{(a_1-a_2)} k^{2/3}$, then denote $Q(s) = Q'(s)$, and otherwise let $Q(s)$ be an arbitrarily chosen subset of $Q(s)$ of size exactly $\frac{7}{(a_1-a_2)} k^{2/3}$. Furthermore, we denote $\hat{Q} = \bigcup_{s \in \hat{U}} Q(s)$. Since $|S| \leq 3k$, the following observation is immediate.

**Observation 6.4.** $|\hat{Q}| \leq \frac{21}{(a_1-a_2)} k^{1/2}$.

Now, we proceed to apply the following rule, whose safeness is based on Lemma 6.13.

**Reduction Rule 6.6.** If there exists a vertex $v \in V(D) \setminus (M \cup \hat{M} \cup \hat{Q})$, then remove $v$ from $G$. The new instance is $(G \setminus v, k)$.

**Lemma 6.14.** Reduction Rule 6.5 is safe.

*Proof.* In one direction, it is clear that if $(G \setminus v, k)$ is a Yes-instance, then $(G, k)$ is a Yes-instance. For the other direction, let us suppose that $(G, k)$ is a Yes-instance. By Lemma 6.13 and since $v \notin M \cup \hat{M}$, there exists a $\{v\}$-nice solution $(S^*, A)$. If $(S^*, A)$ is a solution to $(G \setminus v, k)$, then the proof is complete. Thus, we next suppose that $(S^*, A)$ is not a solution to $(G \setminus v, k)$. Because $(S^*, A)$ is a $\{v\}$-nice solution, this means that there exists $P^* \in S^*$ such that $v \in V(P^*)$ and $V(P^*) \cap \hat{U} \neq \emptyset$. Let $s^*$ denote some vertex in $V(P^*) \cap \hat{U} \neq \emptyset$ (if there exist two vertices in $V(P^*) \cap \hat{U}$, we arbitrarily choose one of them). Now, observe that $S^* \setminus \{P\}$ is a solution to $(G \setminus \{v^*, s\}, k - 1)$, and therefore $(G \setminus \{v, s^*\}, k - 1)$ is a Yes-instance. Moreover, note that $|Q(s^*)| \leq \frac{7}{(a_1-a_2)} k^{2/3} = \frac{7}{(a_1-a_2)} (k-1)^{2/3} - \frac{k^{2/3}}{(k-1)^{2/3}} = \frac{7}{(a_1-a_2)} (k-1)^{2/3} - \frac{1}{(a_1-a_2)} (k-1)^{2/3}$. Then, by Lemma 6.13, there exists a $Q(s^*)$-nice solution $(S', A')$ to $(G \setminus \{v, s^*\}, k - 1)$.

Since $(S', A')$ is a $Q(s^*)$-nice solution, we have that for all $u \in V(S') \cap Q(s^*)$, there exists $P \in S'$ such that $V(P) \cap \hat{U} \neq \emptyset$. However, since $(S', A')$ is a nice solution and $Q(s^*) \cap (M \cup \hat{M}) = \emptyset$, we further derive that for all $u \in V(S') \cap Q(s^*)$, there exists $P \in S'$ such that $V(P) \cap A' \neq \emptyset$. Because $|A'| \leq \frac{3}{(a_1-a_2)} (k-1)^{2/3}$, we deduce that $|V(S') \cap Q(s^*)| \leq 2|A'| \leq \frac{6}{(a_1-a_2)} (k-1)^{2/3}$. However, since $Q(s^*) = \frac{7}{(a_1-a_2)} k^{2/3}$ (because $v \in Q'(s^*) \setminus Q(s^*)$), we have that $Q(s^*) \setminus V(S') \neq \emptyset$. Let $v^*$ denote some vertex in $Q(s^*) \setminus V(S')$ (by our previous argument, such a vertex exists), and
let $C^*$ denote the clique in $D$ that contains $v^*$. Then, by the definition of $Q(s^*)$, we have that $v^* \in MC^*(s^*)$. Observe that any vertex in $V(C^*) \setminus MC^*(s^*)$ together with $s^*$ and $v^*$ forms an induced $P_3$. Hence, if $V(C^*) \setminus MC^*(s^*) \cup V(S') \neq \emptyset$, then $S'$ along with an induced $P_3$ consisting of some vertex in $V(C^*) \setminus (MC^*(s^*) \cup V(S'))$, $s^*$ and $v^*$, forms a solution to $(G,k)$, in which case the proof is complete. However, we claim that necessarily $V(C^*) \setminus (MC^*(s^*) \cup V(S')) \neq \emptyset$. For this purpose, it is sufficient to prove that $|V(S')| < |V(C^*) \setminus MC^*(s^*)|$. Let us first observe that $|V(C^*) \setminus MC^*(s^*)| \geq \frac{1}{2} |V(C^*)|$. Hence, it is sufficient to prove that $|V(C^*) \cap V(S')| < \frac{1}{2} |V(C^*)|$. To this end, we consider two cases, corresponding to whether or not $C^*$ is huge.

- Suppose that $C^*$ is huge. In this case, $\frac{1}{2} |V(C^*)| \geq \frac{4}{3} k$. Since $|V(C^*) \cap V(S')| \leq |V(S')| = 3(k - 1)$ and $\frac{4}{3} \geq 3$, indeed $|V(S')| < \frac{1}{2} |V(C^*)|$. 

- Suppose that $C^*$ is not huge. In this case, by Lemma 6.12, $|(V(C^*) \cap V(S')) \setminus (M \cup M^\prime)| \leq \left(\frac{6}{(\alpha - 1)\beta} + \lambda \right) (k - 1)^{2/3}$. Observe that since $C^* \in D$, we have that $|(V(C^*) \cap V(S')) \setminus (M \cup M^\prime)| = |V(C^*) \cap V(S')| - |V(C^*) \cap V(S') \cap (M \cup M^\prime)| \geq |V(C^*) \cap V(S')| - |V(C^*) \cap (M \cup M^\prime)| > |V(C^*) \cap V(S')| - \frac{1}{2} |V(C^*)| > |V(C^*) \cap V(S')| - \frac{1}{2} \gamma k^{2/3}$. Thus, we derive that $|V(C^*) \cap V(S')| - \frac{1}{2} \gamma k^{2/3} \leq \left(\frac{6}{(\alpha - 1)\beta} + \lambda \right) (k - 1)^{2/3}$, and therefore $|V(C^*) \cap V(S')| < \left(\frac{6}{(\alpha - 1)\beta} + \lambda + \frac{1}{2} \gamma \right) k^{2/3}$. However, $\frac{1}{2} |V(C^*)| > \frac{1}{2} k^{2/3}$. Since $\frac{6}{(\alpha - 1)\beta} + \lambda + \frac{1}{2} \gamma \leq \frac{4}{7}$, indeed $|V(S')| < \frac{1}{2} |V(C^*)|$. 

As both cases led to the desired claim, the proof is complete. 

6.7 Proof of Theorem 3

We are finally ready to present the proof of Theorem 3.

Proof of Theorem 3. Let $(G,k)$ be an instance of Induced $P_3$-packing. Our kernelization algorithm simply applies (exhaustively) Reduction Rules 6.1 to 6.6. The output is the instance obtained once none of these rules is applicable. Let us observe that each rule among Reduction Rules 4.1 to 4.16 can be applied in polynomial time, it strictly decreases the size of $G$ and it does not increase $k$. Thus, our kernelization algorithm runs in polynomial time.

For the sake of clarity, let us now abuse notation and denote the outputted instance by $(G,k)$. Let us observe that $V(G)$ consists of the following vertices.

- Vertices in $S$, whose number is at most $3k$.
- Vertices in bad cliques, whose number is at most $3\mu (\gamma + \delta (\alpha + \beta + \frac{\alpha \beta}{1-\alpha}) + \frac{1}{\delta \eta}) k^{1/2}$ (by Lemma 6.10).
- Vertices in $M \cup M^\prime$, whose number is at most $(6(\alpha + \beta) + \frac{2\alpha \beta}{1-\alpha}) k^{1/2}$ (by Lemmata 6.4 and 6.8).
- Vertices in $\hat{Q}$, whose number is at most $\frac{21}{(\alpha - 1)\beta} k^{1/2}$ (by Observation 6.4).

Thus, the total number of vertices is indeed $O(k^{1/2}\bar{\gamma})$. This completes the proof. 

7 Triangle Packing in Tournaments

In this section, we prove the following theorem.

Theorem 4. TPT admits a kernel with $O(k^{3/2})$ vertices.
Let \((T, k)\) be an instance of TPT. There is a simple polynomial-time \(\frac{1}{3}\)-approximation algorithm for TPT: greedily find a maximal collection, say \(S\), of vertex-disjoint triangles in \(T\) and output \(S\). Indeed, if there is a collection \(S^*\) of vertex-disjoint triangles in \(T\) with \(|S^*| > 3|S|\), then there is a triangle in \(S^*\) not hit by \(V(S)\), contradicting the assumption that \(S\) is maximal. If \(|S| < \frac{2}{3}\), then we conclude that \((T, k)\) is a No-instance. If \(|S| \geq k\), then we conclude that \((T, k)\) is a Yes-instance. Hence, we assume that \(\frac{k}{3} \leq |S| \leq k - 1\). Let \(S = V(S)\), then \(|S| \leq 3k - 3\). By maximality of \(S\), \(T - S\) does not have any directed triangle, and so by Proposition 5.1, \(T - S\) does not have any directed cycle. Hence, \(S\) is a feedback vertex set of \(T\). Let \(X = T - S\). Note that since \(S\) is a feedback vertex set, \(X\) is a transitive tournament.

Let \((T, h)\) be an instance of TPT. We call a collection of at least \(h\) vertex-disjoint triangles of \(T\) as a solution to the instance \((T, h)\). First, we have the following reduction rule.

**Reduction Rule 7.1.** If there exists \(s \in S\) such that there are \(3k - 2\) triangles pairwise intersecting only at \(s\), then remove \(s\) from \(T\). The new instance is \((T \setminus \{s\}, k - 1)\).

The safeness of rule 7.1 is simple. In one direction, if \(S^*\) is a solution to \((T, k)\), by removing the triangle (if any) containing \(s\), we obtain a solution to \((T \setminus \{s\}, k - 1)\). In the other direction, suppose that \(S^*\) is a solution to \((T \setminus \{s\}, k - 1)\). If \(|S^*| \geq k\), then \(S^*\) is a solution to \((T, k)\). Hence \(|S^*| = k - 1\). If there is a triangle, say \(sxy\) that is not hit by \(V(S^*)\), then \(S^* \cup \{sxy\}\) is a solution to \((T, k)\). Otherwise, \(V(S^*)\) hits all \(3k - 2\) triangles pairwise intersecting only at \(s\), and so \(|V(S^*)| \geq 3k - 2\), which contradicts \(|S^*| = k - 1\).

We apply Reduction Rule 7.1 exhaustively. By the same argument as for Reduction Rule 5.1, for any vertex \(s \in S\), we can check whether that exist \(3k\) triangles intersecting pairwise only at \(s\) in polynomial time. Thus, from now onwards, we assume that Reduction Rule 7.1 is no longer applicable.

In this section, we reuse the notation used in Section 5. Throughout this section we work with the unique ordering \(\prec\) of vertices of \(X\) and use terms like consecutive vertices in \(X\), smallest and largest vertex in \(X\).

### 7.1 Exploring the Vertex Cover Structure

Recall the notion of vertex cover for a set of arcs of \(T\). Formally, for a subset of arcs \(A \subseteq E(T)\), a subset \(O \subseteq V(T)\) is called a vertex cover for \(A\) if for every arc \(uv \in A\), either \(u \in O\) or \(v \in O\) (or both). However, the definition of strong arc is slightly different from that in Section 5. An arc \(xy\) of \(T\) is called strong if (i) at least one vertex among \(x\) and \(y\) belongs to \(S\), and (ii) there are at least \(3k\) vertices \(z \in V(T)\) such that \(xyz\) is a triangle. Let \(F\) be the set of all the strong arcs of \(T\), which can be easily found in polynomial time.

Recall that throughout our kernelization algorithm, we work with the unique topological ordering \(\prec\) of \(X\). Accordingly, we have that if \(xx'\) is an arc in \(E(X)\), then \(x \prec x'\). Furthermore, we need the following notion of distance.

**Definition 7.1.** Let \(x, x' \in X\) be two vertices such that \(x \prec x'\), and let \(d - 1\) be the number of vertices \(y\) such that \(x \prec y \prec x'\). Then, the distance between \(x\) and \(x'\) is \(d\). Accordingly, \(x' - x := d\) and \(x - x' := -d\).

In addition, we need the following definition which concerns the relations between the vertices in \(S\) and the vertices in \(X\).

**Definition 7.2.** For \(s \in S\) and \(x \in V(X)\), define \(f^-_s(x) = |\{y \in V(X) : y \leq x, sy \in E(T)\}|\), and \(f^+_s(x) = |\{y \in V(X) : y \geq x, ys \in E(T)\}|\).

Similar to Lemma 5.1, we can prove the following.
Lemma 7.1. For every $s \in S$, there is $x \in X$ such that $0 \leq \phi_s(x) - \psi_s(x) \leq 1$.

As in Section 5 we have the following notation.

Definition 7.3. For any $s \in S$, define $\phi(s)$ as the smallest vertex $x_s \in V(X)$ satisfying the inequalities in Lemma 7.1.

We now show that given Reduction Rule 7.1, neither $\phi_s(\phi(s))$ nor $\psi_s(\phi(s))$ can be too “large”. Indeed, if there existed $s \in S$ such that $\phi_s(\phi(s)) \geq 3k - 1$, then $\phi_s(\phi(s)) \geq 3k - 2$, and we could have formed $3k - 2$ triangles, each consisting of $s$, a vertex from $\{x \in V(X) : x \leq \phi(s), sx \in E(T)\}$, and a vertex from $\{y \in V(X) : y \gg \phi(s), ys \in E(T)\}$. In this case, Reduction Rule 7.1 is applicable. However, as we assumed that Reduction Rule 7.1 is no longer applicable, we have that for all $s \in S$, $\phi_s(\phi(s)), \psi_s(\phi(s)) \leq 3k - 2$. By using this assumption, we have useful certificates for strong arcs similar to the one in Lemma 5.2.

Lemma 7.2. Let $x \in X$, and $s, s' \in S$. The following statements are true.

1. If $sx \in E(T)$ and $\phi(s) - x \geq 6k - 2$, then $sx$ is strong.
2. If $xs \in E(T)$ and $x - \phi(s) \geq 6k - 2$, then $xs$ is strong.
3. If $s's \in E(T)$ and $\phi(s') - \phi(s) \geq 9k - 4$, then $ss'$ is strong.

To proceed, as before, we also need to introduce two terms concerning triangles.

Definition 7.4. Let $x_1x_2x_3$ be a triangle of $T$, and $A = \{x_1, x_2, x_3\}$. The span of $x_1x_2x_3$ is the maximum distance between any two vertices in $(A \setminus S) \cup \phi(A \cap S)$. Moreover, the triangle is called local if none of its arcs belongs to $F$.

In the following lemma, we will show that a local triangle is indeed local in the sense that it must have a “short” span. The proof of the following is identical to the one for Lemma 5.3.

Lemma 7.3. Let $x_1x_2x_3$ be a local triangle with at least one vertex from $X$. Then, its span is at most $18k - 8$.

7.2 Applying the New Expansion Lemma

In what follows, we denote $\alpha = 3, \beta = 845, \gamma = 32, \mu = 9, \lambda = 25, \delta = 11$, and $\ell = 3$ so that $\beta - 3\delta - 3\ell \geq \lambda\gamma$ (used in Observation 7.8), $(\mu - 2)\ell > 4$ (used in the proof of Observation 7.10), $(\delta - 9)\ell > 4$ (used in the proof of Lemma 7.6), $\ell^2 > 4$ (used in Lemma 7.7), and $(\lambda - 2\mu - \alpha)(\gamma - 2\delta) - 2\lambda\mu - \ell\alpha > 0$ (used in the proof of Lemma 7.5).

Next we give the definition of intervals.

Definition 7.5. A set $Y \subseteq V(X)$ is an interval if it contains all the vertices in $X$ that lie between the largest and smallest elements in $Y$ (with respect to the ordering $<$ induced by $X$).\footnote{That is, the elements of $Y$ are consecutive with respect to $<$.} We refer to $|Y|$ as the length of $Y$.

We partition $V(X)$ into disjoint intervals, each of length $\beta k$. That is, we follow the vertices of $V(X)$ from left to right in the ordering $<$, and partition them into disjoint intervals $X_1, \ldots, X_p$ such that each $X_i$, $1 \leq i < p$, is of length $\beta k$. Let $S_i := \{s \in S : \phi(s) \in X_i\}$.

Definition 7.6. Let $X_i$ be an interval such that $|S_i| \geq \alpha \sqrt{k}$, then we call $X_i$ bad of Type 1.

Clearly, there are less than $\frac{3}{\alpha} \sqrt{k}$ bad intervals of Type 1, since $|S| < 3k$.

Observation 7.1. There are at most $\frac{3\sqrt{k}}{\alpha}$ bad intervals of Type 1 among $X_1, \ldots, X_p$. 
For each $i$, we call a 3-approximation algorithm to TPT on the tournament $T[X_i \cup S_i]$. If the 3-approximation algorithm returns a solution of size least $\sqrt{k}$, we call $X_i$ bad of Type 2. There are at most $\sqrt{k}$ bad interval of Type 2; otherwise, the (obviously vertex-disjoint) union of 3-approximate solutions of all these bad intervals has size at least $k$, and we conclude immediately that $(T, k)$ is a Yes-instance.

**Observation 7.2.** There are at most $\sqrt{k}$ bad intervals of Type 2 among $X_1, \ldots, X_p$.

Observations 7.1 and 7.2 imply that there are at least $(p - 1) - \frac{2}{3} \sqrt{k} - \sqrt{k}$ non-bad intervals – we do not call them good yet, since we will introduce another type of bad intervals. For every non-bad interval $X_i$, let $Y^*_i$ be the sub-interval of $X_i$ excluding 18$k$ smallest vertices and 18$k$ largest vertices. Then every $Y^*_i$ has length $(\beta - 36)k$. Recall that a triangle is local if it has no arcs in common with $F$. We give here two observations for later use.

**Observation 7.3.** If $xyz$ is a local triangle with $x \in Y^*_i$ for some $i$, then $x, y, z \in X_i \cup S_i$.

*Proof.* By Lemma 7.3, $xyz$ has span at most $18k - 8$. Note that $x \in Y^*_i$, while $Y^*_i$ is obtained from $X_i$ by excluding 18$k$ smallest and 18$k$ largest vertices, so $(\{x, y, z\} \cap X) \cup \varphi(\{x, y, z\} \cap S)$ is a subset of $X_i$. In other words, $x, y, z \in S_i \cup X_i$. □

**Observation 7.4.** If $X_i$ is a non-bad interval, then every collection of vertex-disjoint local triangles contains less than $6\sqrt{k}$ vertices in $Y^*_i$.

*Proof.* Suppose for a contradiction that there is a collection $O$ of local triangles with at least $6\sqrt{k}$ vertices of $Y^*_i$. Since each local triangle contains at most two vertices of $Y^*_i$ ($Y^*_i$ is transitive), the collection has at least $3\sqrt{k}$ local triangles. Let us consider a local triangle $xyz$ with $x \in Y^*_i$. By Observation 7.3, $x, y, z \in S_i \cup X_i$. Hence $O$ contains at least $3\sqrt{k}$ triangles in $X_i \cup S_i$. This implies that a 3-approximation algorithm for TPT when run on $T[X_i \cup S_i]$ returns a solution of size at least $\sqrt{k}$. Therefore $X_i$ is bad of Type 2, a contradiction. □

We remark that Observation 7.4 is very strong since it allows us to upper bound the number of vertex-disjoint triangles intersects a specific interval.

We now apply the New Expansion Lemma the first time to introduce the bad intervals of Type 3; later on, we will apply the New Expansion Lemma the second time to detect a relevant vertex. Let $Y^*$ be the union of all $Y^*_i$ such that the corresponding $X_i$ is non-bad. Let us consider the (undirected) bipartite graph $G$ with vertex bipartition $(S, Y^*)$, and $E(G)$ consists of edges corresponding to those arcs in $F$ which has one endpoint in $S$ and another endpoint in $Y^*$. By applying New Expansion Lemma (Lemma 3.2) on $G$, we obtain $\hat{Y}^*$ and $\hat{S}$ satisfying the following.

**Observation 7.5.** $\hat{S}$ has an $\ell \sqrt{k}$-expansion to $\hat{Y}^*$ in $G$, $N_G(\hat{Y}^*_i) \subseteq \hat{S}$ and $|Y^* \setminus \hat{Y}^*| \leq \ell \sqrt{k}|S|$.

We can now define the third type of bad intervals.

**Definition 7.7.** For every $i$, if $|Y^*_i \setminus \hat{Y}^*| \geq 3\ell k$, then $X_i$ is called bad of Type 3.

Then there are at most $\sqrt{k}$ bad intervals of Type 3 since $|Y^* \setminus \hat{Y}^*| \leq \ell \sqrt{k}|S| < 3\ell k^{3/2}$.

**Observation 7.6.** There are at most $\sqrt{k}$ bad intervals of Type 3 among $X_1, \ldots, X_p$.

Finally, we are ready to define the notion of good interval.

**Definition 7.8.** Let $X_i$ be an interval such that it is not bad of Types 1, 2 or 3, then it is called good.
Observations 7.1, 7.2 and 7.6 imply that there are at least $p - \frac{3}{\alpha} \sqrt{k} - \sqrt{k} - \sqrt{k}$ good intervals. Then we have the following observation.

**Observation 7.7.** If $p \geq \left(\frac{3}{\alpha} + 3\right) \sqrt{k}$ then there are at least $\sqrt{k}$ good intervals.

For every good $X_i$, let $Y_i = Y_i^* \cap \hat{Y}^*$. Then $|Y_i| \geq |Y_i^*| - |Y_i^* \setminus \hat{Y}^*| \geq (\beta - 36 - 3\ell)k$. Since, $\beta - 36 - 3\ell \geq \lambda \gamma$, we have the following observation.

**Observation 7.8.** $|Y_i| \geq \lambda \gamma k$ for every good $X_i$.

Given $x, x' \in Y_i$ with $x < x'$, we say that $x$ and $x'$ are consecutive in $Y_i$ if there is no $y \in Y_i$ such that $x < y < x'$. Note that $Y_i$ is not a sub-interval of $X_i$, so two consecutive vertices in $Y_i$ are not necessarily consecutive in $X_i$. To avoid confusion, we do not introduce the distance notion between two vertices in $Y_i$; however, the order $\prec$ in $Y_i$ is the restriction of $\prec$ on $X$. The following observation is immediate from Observation 7.4.

**Observation 7.9.** If $X_i$ is good, then every collection of vertex-disjoint local triangles contains less than $6\sqrt{k}$ vertices in $Y_i$.

For a vertex $s \in S_i$, $\varphi(s)$ can be thought as a “balanced projection” of $s$ on $X$. However, $\varphi(s)$ may not be a balanced projection of $s$ on $Y_i$. Thus, we wish to find a balanced projection of $s$ on $Y_i$. To do so, we repeat what we did before to find $\varphi(s)$ as follows. For every $s \in S_i$, let $R_s^-(x) = \{y \in Y_i : y \leq x, sy \in E(T)\}$ and $R_s^+(x) = \{y \in Y_i : y \succ x, ys \in E(T)\}$. Note that $R_s^-(x)$ and $R_s^+(x)$ only count arcs between $s$ and $Y_i$.

**Lemma 7.4.** For every $s \in S_i$, there is $x \in Y_i$ such that $0 \leq |R_s^-(x)| - |R_s^+(x)| \leq 1$.

The proof of Lemma 7.4 is similar to that of Lemma 5.1, where note that $|R_s^-(x')| - |R_s^+(x')| = |R_s^-(x)| - |R_s^+(x)| + 1$ for every $x \prec x'$ consecutive in $Y_i$.

**Definition 7.9.** For any $s \in S_i$, define $\theta(s)$ to be the smallest vertex in $Y_i$ satisfying the inequalities in Lemma 7.4.

We denote $R_s^+ = R_s^+(\theta(s)), R_s^- = R_s^-(\theta(s))$ for short. We could not upper bound $|R_s^-|$ and $|R_s^+|$ as we did for $\varphi(s)$; thus, we overcome this by introducing the notions of heavy and light.

**Definition 7.10.** Given $s \in S_i$, if $|R_s^-| \geq \mu \sqrt{k} + 1$, then we call $s$ heavy; otherwise, we call $s$ light.

Thus, if $s$ is light, $|R_s^-|, |R_s^+| \leq \mu \sqrt{k}$. Let

$$R_i = \bigcup_{\{s \in S_i : s \text{ is light}\}} (R_s^- \cup R_s^+)$$

Then $|R_i| \leq 2\mu \sqrt{k}|S_i| \leq 2\alpha \mu k$ since $|S_i| \leq \alpha \sqrt{k}$.

Recall from Observation 7.8 that $|Y_i| \geq \lambda \gamma$. We partition $Y_i$ into subsets $Y_{i,1},...,Y_{i,\lambda \gamma k}$ where $|Y_{i,j}| \geq \gamma \sqrt{k}$ for every $j \leq \lambda \sqrt{k}$, and $x \prec x'$ for every $x \in Y_{i,j}, x' \in Y_{i,j'}$ with $j < j'$ (it is useful to think that $Y_{i,j}$ is a “sub-interval” of $Y_i$; however, we would like to avoid that term since $Y_i$ itself is not an interval).

**Definition 7.11.** A set $Y_{i,j}$ is called fit if $|Y_{i,j} \cap R_i| < 3 \sqrt{k}$ and $\theta(S_i) \cap Y_{i,j} = \emptyset$.

Since $|R_i| \leq 2\alpha \mu k$, there are at most $\frac{2\mu \alpha \gamma}{3} \gamma \sqrt{k}$ sets $Y_{i,j}$ such that $|Y_{i,j} \cap R_i| \geq 3 \sqrt{k}$. Since $|S_i| \leq \alpha \sqrt{k}$, there are at most $\alpha \sqrt{k}$ intervals $Y_{i,j}$ such that $Y_{i,j}$ contains $\theta(s)$ for some $s \in S_i$. Thus, there are at least $(\lambda - \frac{2\mu \alpha \gamma}{3} - \alpha) \gamma \sqrt{k}$ fit subset $Y_{i,j}$ of every $Y_i$. For each fit $Y_{i,j}$, let $Y_{i,j}^-, Y_{i,j}^+$ be
the $\delta \sqrt{k}$ smallest vertices and $\delta \sqrt{k}$ largest vertices in $Y_{i,j}$, respectively, and $Y'_{i,j} \geq Y_{i,j} \setminus (Y_{i,j} \cap Y_{i,j})$. Then $Y'_{i,j} = (\gamma - 2\delta) \sqrt{k}$. Let

$$A_i = \left( \bigcup_{\{j: Y_{i,j} \text{ is fit}\}} Y'_{i,j} \right) \setminus R_i \quad \text{and} \quad A = \bigcup_{\{i: X_i \text{ is good}\}} A_i.$$  

Then

$$|A_i| \geq \left( \lambda - \frac{2\alpha \mu}{3} - \alpha \right) \sqrt{k}(\gamma - 2\delta)10\sqrt{k} - |R_i| \geq \left( \left( \lambda - \frac{2\alpha \mu}{3} - \alpha \right) (\gamma - 2\delta) - 2\alpha \mu \right) k,$$

and by Observation 7.7, we have $|A| \geq \sqrt{k}(\lambda - \frac{2\alpha \mu}{3} - \alpha)(\gamma - 2\delta) - 2\alpha \mu)k$.

Now we apply the New Expansion Lemma (Lemma 3.2) the second time, this time on $G[A \cup \tilde{S}]$, to get $\hat{A}$ and $\tilde{S}$ such that $\tilde{S}$ has an $\ell \sqrt{k}$-expansion into $\hat{A}$ in $G[A \cup \tilde{S}]$, $N_{G[A \cup \tilde{S}]}(\hat{A}) \subseteq \tilde{S}$ and $|A \setminus \hat{A}| \leq \ell \sqrt{k}|\tilde{S}|$.

**Lemma 7.5.** $\tilde{S}$ has an $\ell \sqrt{k}$-expansion into $\hat{A}$ in $G$, $N_{G}(\hat{A}) \subseteq \tilde{S}$ and $\hat{A}$ is nonempty.

*Proof.* Since $G[A \cup \tilde{S}]$ is a induced subgraph of $G$, then clearly $\tilde{S}$ has an $\ell \sqrt{k}$-expansion into $\hat{A}$ in $G$. By Observation 7.5, $N_{G}(\hat{Y}^*) \subseteq \tilde{S}$, so there is no edge between $\hat{Y}^*$ and $S \setminus \tilde{S}$ in $G$. Thus, there is no edge between $\hat{A}$ and $S \setminus \tilde{S}$ in $G$ since $\hat{A} \subseteq \hat{Y}^*$. Since $N_{G[A \cup \tilde{S}]}(\hat{A}) \subseteq \tilde{S}$, there is no edge between $\hat{A}$ and $S \setminus \tilde{S}$ in $G[A \cup \tilde{S}]$. Thus, there is no edge between $\hat{A}$ and $\tilde{S} \setminus \hat{A}$ in $G$. In other words, $N_{G}(\hat{A}) \subseteq \tilde{S}$.

Observe that $|A \setminus \hat{A}| \leq \ell \sqrt{k}|\tilde{S}| \leq \ell \sqrt{k}|S| \leq \ell \alpha k^{3/2}$. Hence, $\hat{A} \geq |A| - |A \setminus \hat{A}| \geq ((\lambda - \frac{2\alpha \mu}{3} - \alpha)(\gamma - 2\delta) - 2\alpha \mu)k^{3/2} - \ell \alpha k^{3/2} > 0$. This proves the lemma. \(\square\)

### 7.3 Using Expansion to Detect an Irrelevant Vertex

Recall that a triangle is local if it contains no strong arc, i.e. it has no arcs in common with $F$. In the next lemma, we will show that given a mixed collection of local triangles and strong arcs, it is possible to exclude a particular vertex of $\hat{A}$ from the collection.

**Lemma 7.6.** Let $x \in \hat{A}$ and assume that there is a vertex-disjoint collection $O$ of local triangles and strong arcs such that $|O| = k$ and $x$ belongs to a local triangle of $O$. Then there is a vertex-disjoint collection $O'$ of local triangles and strong arcs such that $|O'| = k$ and $x$ does not belong to any local triangle or strong arc of $O$.

*Proof.* Let $O$ be the vertex set of $O$, then $|O| \leq 3k$. Assume that the statement of the lemma was false and let $xyz \in O$ and $x \in Y_{i,j}$. By Observation 7.3, we have $y, z \in X_i \cup S_i$ since $Y_{i,j} \subseteq Y_{i,j}^*$. Thus, either $y \in S_i$ or $z \in S_i$; otherwise, $xyz$ is transitive. We first prove the following observation.

**Observation 7.10.** Neither $y$ nor $z$ is heavy.

*Proof.* Suppose that $y$ is heavy. If there are $v \in R^-_y, v' \in R^+_y$ such that $v, v' \notin O$, then $O' : = (O \setminus \{xyz\}) \cup \{yv'v\}$ is a desired collection, a contradiction. Thus we conclude that either $R^-_y \subseteq O$ or $R^+_y \subseteq O$.

If $R^-_y \subseteq O$, we will show that we can exchange some strong arc of $O$ with some strong arc outside to “free” one vertex of $R^+_y$ from $O$. Since $R^+_y \subseteq Y_i$, by Observation 7.9, at most $6 \sqrt{k}$ vertices in $R^+_y$ belong to a local triangle in $O$. Recall that $|R^+_y| \geq \mu \sqrt{k}$ by the definition of heaviness. Thus at least $(\mu - 6) \sqrt{k}$ vertices belong to some strong arcs of $O$; we call that set $Z$. Then $O$ contains a matching of strong arcs from $Z$ into $S$ (since a strong arc must
contain at least one vertex in $S$). Let $W$ be the set of vertices of $S$ in that matching, then $|W| = |Z| \geq (\mu - 2)\sqrt{k}$. Note that $Z \subseteq \hat{Y}^*$ since $R_i \subseteq Y_i \subseteq \hat{Y}^*$.

By Observation 7.5, we have $N_G(\hat{Y}^*) \subseteq \hat{S}$, and so $W \subseteq N_G(Z) \subseteq \hat{S}$. By Observation 7.5 again, we have $|N_G(W)| \geq \ell \sqrt{k}|W| \geq (\mu - 2)\sqrt{k} > 4k = |O| + k$. We choose an arbitrary $u \in N_G(W)$ (among at least $k$ candidates) such that $u \notin O$, let $w \in W$ be a neighbor of $u$ in $G$ (such $w$ always exists since $u \in N_G(W)$). Suppose that $wv \in O$, then $v \in Z \subseteq R_y^+$. Then remove $wv$ from $O$ and add $wu$ to $O$. We still call the new collection $O$. In doing so, we free $v \in R_y^+$ from $O$.

If $R_y^+ \subseteq O$, by repeating the above argument, we can free some $v' \in R_y^+$ from $O$. Note that since we have $k$ candidates to choose to exchange strong arcs, we can avoid “recapturing” $v$ into $O$. Then $O' := (O \setminus \{xyz\}) \cup \{yvv'\}$ is a desired collection, a contradiction. Similarly, we can show that $z$ is not heavy.

We have 3 cases:

Case 1: $y \in S_i$ and $z \in X_i$. Then $y$ is light by Observation 7.10, and since $x \in \hat{A} \subseteq A$ while $A \cap R_i = \emptyset$, we have $x \notin R_i$, and so $x \notin R_y^+ \cup R_y^-$. Combining with $xy \in E(T)$, we have $\theta(y) > x$. Recall that $Y_{i,j}^+$ is the set of $\delta \sqrt{k}$ largest vertices of $Y_{i,j}$, and hence $u < v$ for every $u \in Y_{i,j}^+, v \in Y_{i,j}^-$. Since $x \in Y_{i,j}^+$, we have $x < v$ for every $v \in Y_{i,j}^-$. Note that $\theta(y) > x$, and $\theta(y) \notin Y_{i,j}$ since $Y_{i,j}$ is fit. Thus $\theta(y) > v$ for every $v \in Y_{i,j}^-$. Since $R_y^+ \cup R_y^- \subseteq R_i$, we have $vy \in E(T)$ for every $v \in Y_{i,j}^- \setminus R_i$. Besides, $|Y_{i,j}^+ \cap R_i| < 3\sqrt{k}$ since $Y_{i,j}$ is fit, and so $|Y_{i,j}^+ \setminus R_i| \geq |Y_{i,j}^+| - |Y_{i,j}^- \cap R_i| \geq (\delta - 3)\sqrt{k}$. Note that $z \in X_i$ and $xz \in E(T)$, so $z < x < v$ for every $v \in Y_{i,j}^+$. If there is $v \in Y_{i,j}^+ \setminus R_i$ such that $v \notin O$, then $O' := (O \setminus \{xyz\}) \cup \{vvy\}$ is a desired collection, a contradiction. Thus we conclude that there is no such $v$. In other words, $Y_{i,j}^+ \setminus R_i \subseteq O$. Since $Y_{i,j}^+ \subseteq Y_i$, by Observation 7.9, at most $6\sqrt{k}$ vertices in $Y_{i,j}^+ \setminus R_i$ belong to local triangles in $O$, while we showed above that $|Y_{i,j}^+ \setminus R_i| \geq (\delta - 3)\sqrt{k}$. Thus at least $(\delta - 3 - 6)\sqrt{k} = (\delta - 9)\sqrt{k}$ vertices of $Y_{i,j}^+ \setminus R_i$ belong to some strong arcs of $O$. Then by the same arguments as in the proof of Observation 7.10, combined with the assumption that $(\delta - 9)\ell > 4$, we can exchange strong arcs of $O$ to free some $v \in Y_{i,j}^+ \setminus R_i$ from $O$. Then $O' := (O \setminus \{xyz\}) \cup \{vvy\}$ is a desired collection, a contradiction.

Case 2: $y \in X_i$ and $z \in S_i$. This case is similar to Case 1, but we will consider $Y_{i,j}^-$ (instead of $Y_{i,j}^+$) to employ the fact that $v < x$ for every $v \in Y_{i,j}^-$.

Case 3: $y, z \in S_i$. Then by a similar argument as in Case 1, both $y, z$ are light and $\theta(z) < x < \theta(y)$. Then we have $\theta(z) < v < \theta(y)$ for every $v \in Y_{i,j}^+$ since $\theta(y), \theta(z) \notin Y_{i,j}^-$. Note that $zw, zy \in E(T)$ for every $v \in Y_{i,j}^+ \setminus R_i$, then we repeat the argument in Case 1 to reach the contradiction. This concludes the proof.

We can now strengthen Lemma 7.6 by omitting the assumption that $x$ belongs to a local triangle of $O$.

**Lemma 7.7.** Let $x \in \hat{A}$ and suppose that there is a vertex-disjoint collection $O$ of local triangles and strong arcs with $|O| = k$. Then there is a vertex-disjoint collection $O'$ of local triangles and strong arcs such that $|O'| = k$ and $x$ does not belong to any triangle or strong arc of $O$.

**Proof.** Let $O$ be the vertex set of $O$, then $|O| \leq 3k$ and $x \in O$ (otherwise, the lemma is obvious). If $x$ belongs to a local triangle of $O$, then we apply Lemma 7.6. Otherwise, $x$ belongs to a strong arc of $O$, say $xy$ (note that in this proof, we do not consider the orientation of a strong arc, i.e. when we say $uv$ is a strong arc, we mean either $uv$ or $vu$ is a strong arc).

By Lemma 7.5, $N_G(\hat{A}) \subseteq \hat{S}$, and so $y \in N_G(x) \subseteq \hat{S}$, and $|N_G(y) \cap \hat{A}| \geq \ell \sqrt{k}|y| = \ell \sqrt{k}$. Let $Z = N_G(y) \cap \hat{A}$. If there is $v \in Z$ such that $v \notin O$, then $O' := (O \setminus \{xy\}) \cup \{vy\}$ is a desired collection. Thus, we conclude that $Z \subseteq O$. 

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Suppose that there is \( v \in Z \) such that \( v \) belongs to a local triangle of \( \mathcal{O} \). Since \( v \in \hat{A} \), we apply Lemma 7.6 to \( v \) and obtain a collection \( \mathcal{O}'' \) such that \( |\mathcal{O}''| = k \) and \( v \) does not belong to any triangle or strong arc of \( \mathcal{O}'' \). Note also that according to the proof of Lemma 7.6, \( \mathcal{O}'' \) is obtained from \( \mathcal{O} \) by exchanging some strong arcs and a local triangle. If \( x \) is freed by these exchange, then \( \mathcal{O}'' \) is a desired collection. Note that it is impossible that \( x \) is freed and then recaptured to \( \mathcal{O}' \) in a strong arc, since we can always have \( k \) candidates of strong arcs, and so we can avoid recapturing \( x \). \( x \) is freed and then recaptured to \( \mathcal{O}' \) in a local triangle, then we just apply Lemma 7.6 again to \( x \) to find a desired collection. Thus, we concluded that \( x \) is “untouched” during the swapping procedure above, i.e. \( xy \in \mathcal{O}' \). Then \( \mathcal{O}' := (\mathcal{O}' \setminus \{xy\}) \cup \{vy\} \) is a desired collection.

We conclude that every element of \( Z \) belongs to some strong arc of \( \mathcal{O} \). Then \( \mathcal{O} \) contains a matching of strong arcs from \( Z \) to \( S \). Let \( W \) be the set of vertices of \( S \) in that matching, then \( |W| = |Z| \geq \ell \sqrt{k} \). Note that \( Z \subseteq \hat{A} \). By Lemma 7.5, we have \( N_G(\hat{A}) \subseteq \tilde{S} \), and so \( W \subseteq \tilde{S} \). By Lemma 7.5 again, we have \( |N_G(W)| \geq \ell \sqrt{k}|W| \geq \ell \sqrt{k} \times \ell \sqrt{k} = \ell^2 k > 4k = |\mathcal{O}| + 4 \). Thus, there is \( u \in N_G(W) \) such that \( u \notin \mathcal{O} \). Let \( w \in W \) be a neighbor of \( u \) in \( G \) (such \( w \) always exists since \( u \in N_G(W) \)). Let \( wv \in \mathcal{O} \). Then \( \mathcal{O}' := (\mathcal{O} \setminus \{xy, wv\}) \cup \{vy, wu\} \) is a desired collection. \( \square \)

Finally, we are ready to state the reduction rule that removes an irrelevant vertex.

**Reduction Rule 7.2.** Let \( x \) be an arbitrary vertex in \( \hat{A} \). Remove \( x \) from \( T \). The new instance is \((T \setminus \{x\}, k)\).

**Lemma 7.8.** Reduction Rule 7.2 is safe.

**Proof.** If it is obvious that if \((T \setminus \{x\}, k)\) is a Yes-instance, then \((T, k)\) is a Yes-instance. Conversely, suppose that \((T, k)\) is a Yes-instance with some solution \( \mathcal{O}^* \), while \((T \setminus \{x\}, k)\) is a No-instance. Then \( |\mathcal{O}^*| = k \). Let \( \mathcal{O} \) be the collection of local triangles and strong arcs obtained from \( \mathcal{O}^* \) as follows. For every \( wvw \in \mathcal{O}^* \), if \( wvw \) is local, then \( wvw \in \mathcal{O} \); otherwise, \( wvw \) must contain some strong arcs, then choose an arbitrary strong arc of \( wvw \) to be in \( \mathcal{O} \). Then \( |\mathcal{O}| = k \). Applying Lemma 7.7, we obtain a collection \( \mathcal{O}' \) such that \( x \) does not belong to any local triangle and strong arc of \( \mathcal{O}' \).

We now construct a solution to \((T \setminus \{x\}, k)\) by repeating the following argument sequentially. Pick an arbitrary strong arc of \( \mathcal{O}' \), say \( yz \), we choose a vertex \( w \) such that \( yzw \) is a triangle, \( w \neq x \) and \( w \) does not belong to any element of \( \mathcal{O}' \) and set \( \mathcal{O}' := (\mathcal{O}' \setminus \{yz\}) \cup \{yzw\} \). It is clear that we can always proceed the exchange since the vertex set \( \mathcal{O}' \setminus \{yz\} \) has at most \( 3k – 3 \) vertices, while there are \( 3k \) possible choices for \( w \) since \( yz \) is strong, so we can always find the desired \( w \). At the end of the process \( \mathcal{O}' \) is a solution to \((T \setminus \{x\}, k)\). This concludes the proof. \( \square \)

### 7.4 Proof of Theorem 4

We are finally ready to present the proof of Theorem 2.

**Proof of Theorem 2.** Let \((T, k)\) be an instance of TPT. Our kernelization algorithm simply applies (exhaustively) Reduction Rules 7.1 and 7.2. The output is the instance obtained once none of these rules is applicable. Let us observe that each of Reduction Rules 7.1 and 7.2 can be applied in polynomial time, it strictly decreases the size of \( G \) and it does not increase \( k \). Thus, our kernelization algorithm runs in polynomial time.

For the sake of clarity, let us now abuse notation and denote the output instance by \((T, k)\). Let us observe that \( V(T) \) consists of the following vertices.

- Vertices in \( S \), whose number is at most \( 3k \).
- Vertices of \( X \), whose number is at most \( p \beta \sqrt{k} = O(k^{3/2}) \) since \( p \leq (\frac{3}{\alpha} + 3) \sqrt{k} \).

Thus, the total number of vertices is indeed \( O(k^{3/2}) \). This complete the proof. \( \square \)
8 Conclusion

In this paper we designed first subquadratic vertex kernels for FVST, CVD, TPT, and Induced $P_3$-Packing. All our kernels were based on the classical Expansion Lemma and the two new versions we proved in this article. We believe that our approach of designing kernels will be fruitful for similar implicit packing and covering problems. A most natural open question is whether these problems admit a kernel with $O(k)$ vertices. Another interesting avenue is to find other problems where the methods developed in this paper can be applied.

References


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