Parameterizing edge modification problems above lower bounds

René van Bevern · Vincent Froese · Christian Komusiewicz

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Abstract We study the parameterized complexity of a variant of the *F*-free Editing problem: Given a graph G and a natural number k, is it possible to modify at most k edges in G so that the resulting graph contains no induced subgraph isomorphic to F? In our variant, the input additionally contains a vertex-disjoint packing \mathcal{H} of induced subgraphs of G, which provides a lower bound $h(\mathcal{H})$ on the number of edge modifications required to transform G into an F-free graph. While earlier works used the number k as parameter or structural parameters of the input graph G, we consider instead the parameter $\ell := k - h(\mathcal{H})$, that is, the number of edge modifications above the lower bound $h(\mathcal{H})$. We develop a framework of generic data reduction rules to show fixed-parameter tractability with respect to ℓ for K_3 -Free Editing, Feedback Arc SET IN TOURNAMENTS, and CLUSTER EDITING when the packing \mathcal{H} contains subgraphs with bounded solution size. For K_3 -FREE EDITING, we also prove NP-hardness in case of edge-disjoint packings of K_{3} s and $\ell = 0$, while for K_q -FREE EDITING and $q \ge 6$, NP-hardness for $\ell = 0$ even holds for vertex-disjoint packings of K_q s. In addition, we provide NP-hardness results for F-FREE VERTEX DELETION, were the aim is to delete a minimum number of vertices to make the input graph F-free.

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René van Bevern

Vincent Froese Technische Universität Berlin, Germany, E-mail: vincent.froese@tu-berlin.de

Christian Komusiewicz

Friedrich-Schiller-Universität Jena, Germany, E-mail: christian.komusiewicz@uni-jena.de

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Novosibirsk State University, Novosibirsk, Russian Federation, E-mail: rvb@nsu.ru Sobolev Institute of Mathematics, Siberian Branch of the Russian Academy of Sciences, Novosibirsk, Russian Federation

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1 Introduction

Graph modification problems are a core topic of algorithmic research [10, 34, 47]. Given a graph *G*, the aim is to transform *G* by a minimum number of modifications (like vertex deletions, edge deletions, or edge insertions) into another graph *G'* fulfilling certain properties. Particularly well-studied are *hereditary* graph properties, which are closed under vertex deletions and are characterized by *minimal forbidden induced subgraphs*: a graph fulfills such a property if and only if it does not contain a graph *F* from a property-specific family \mathcal{F} of graphs as induced subgraph. All nontrivial vertex deletion problems and many edge modification and deletion problems for establishing hereditary graph properties are NP-complete [1, 3, 33, 34, 47]. One approach to cope with the NP-hardness of these problems are *fixed-parameter algorithms* that solve them in $f(k) \cdot n^{O(1)}$ time for some exponential function *f* depending only on some desirably small parameter *k*. If the desired graph property has a finite forbidden induced subgraph characterization, then the corresponding vertex deletion, edge deletion, and edge modification problems are *fixed-parameter tractable* parameterized by the number of modifications *k*, that is, solvable in $f(k) \cdot n^{O(1)}$ time [10].

Parameterization above lower bounds. When combined with data reduction and pruning rules, search-tree based fixed-parameter algorithms for the parameter k of allowed modifications can yield competitive problem solvers [26, 38]. Nevertheless, the number of modifications is often too large and smaller parameters are desirable.

A natural approach to obtain smaller parameters is "parameterization above guaranteed values" [13, 22, 36, 37]. The idea is to use a lower bound h on the solution size and to use $\ell := k - h$ as parameter instead of k. This idea has been applied successfully to VERTEX COVER, the problem of finding at most k vertices such that their deletion removes all edges (that is, all K_2 s) from G. Since the size of a smallest vertex cover is large in many input graphs, parameterizations above the lower bounds "size of a maximum matching M in the input graph" and "optimum value L of the LP relaxation of the standard ILP-formulation of VERTEX COVER" have been considered. After a series of improvements [13, 22, 36, 43], the current best running time is $3^{\ell} \cdot n^{O(1)}$, where $\ell := k - (2 \cdot L - |M|)$ [22].

We extend this approach to edge modification problems, where the number k of modifications tends to be even larger than for vertex deletion problems. For example, in the case of CLUSTER EDITING, which asks to destroy induced paths on three vertices by edge modifications, the number of modifications is often larger than the number of vertices in the input graph [8]. Hence, parameterization above lower bounds seems natural and even more relevant for edge modification problems. Somewhat surprisingly, this approach has not been considered so far. We thus initiate research on parameterization above lower bounds in this context. As a starting point, we focus on edge modification problems for graph properties that are characterized by one small forbidden induced subgraph F:

Problem 1.1 (F-FREE EDITING)

Input: A graph G = (V, E) and a natural number k.

Question: Is there an *F*-free editing set $S \subseteq {\binom{V}{2}}$ of size at most k such that $G \triangle S := (V, (E \setminus S) \cup (S \setminus E))$ does not contain F as induced subgraph?

In the context of a concrete variant of F-FREE EDITING, we refer to an F-free editing set as *solution* and call a solution *optimal* if it has minimum size.

Lower bounds from packings of bounded-cost induced subgraphs. Following the approach of parameterizing VERTEX COVER above the size of a maximum matching, we can parameterize F-FREE EDITING above a lower bound obtained from packings of induced subgraphs containing F.

Definition 1.2 A *vertex-disjoint (or edge-disjoint) packing* of induced subgraphs of a graph *G* is a set $\mathcal{H} = \{H_1, \ldots, H_z\}$ such that each H_i is an induced subgraph of *G* and such that the vertex sets (or edge sets) of the H_i are mutually disjoint.

While it is natural to consider packings of *F*-graphs to obtain a lower bound on the solution size, a packing of other graphs that contain *F* as induced subgraph might yield better lower bounds and thus a smaller parameter above this lower bound. For example, a K_4 contains several triangles and two edge deletions are necessary to make it triangle-free. Thus, if a graph *G* has a vertex-disjoint packing of h_3 triangles and $h_4 K_4$ s, then at least $h_3 + 2 \cdot h_4$ edge deletions are necessary to make it triangle-free.¹ Moreover, when allowing arbitrary graphs for the packing, the lower bounds provided by vertex-disjoint packings of *F*. A disjoint union of $h K_4$ s, for example, has *h* edge-disjoint triangles but also *h* vertex-disjoint K_4 s. Hence, the lower bound provided by packing vertex-disjoint K_4 s is twice as large as the one provided by packing edge-disjoint triangles in this graph.

Motivated by this benefit of vertex-disjoint packings of arbitrary graphs, we mainly consider lower bounds obtained from vertex-disjoint packings, which we assume to receive as input. Thus, we arrive at the following problem, where $\tau(G)$ denotes the minimum size of an *F*-free editing set for a graph *G*:

Problem 1.3 (F-FREE EDITING WITH COST-t PACKING)

Input: A graph G = (V, E), a vertex-disjoint packing \mathcal{H} of induced subgraphs of G such that $1 \le \tau(H) \le t$ for each $H \in \mathcal{H}$, and a natural number k.

Question: Is there an *F*-free editing set $S \subseteq \binom{V}{2}$ of size at most k such that $G \triangle S := (V, (E \setminus S) \cup (S \setminus E))$ does not contain F as induced subgraph?

The special case of F-free Editing with Cost-t Packing where only F-graphs are allowed in the packing is called F-free Editing with F-Packing.

From the packing \mathcal{H} , we obtain the lower bound $h(\mathcal{H}) := \sum_{H \in \mathcal{H}} \tau(H)$ on the size of an *F*-free editing set, which allows us to use the excess $\ell := k - h(\mathcal{H})$ over this lower bound as parameter, as illustrated in Figure 1.1. Since *F* is a fixed graph, we

¹ Bounds of this type are exploited, for example, in so-called cutting planes, which are used in speeding up the running time of ILP solvers.





Fig. 1.1 An instance of TRIANGLE DELETION. The packing graphs have gray background. Left: A vertexdisjoint packing of two triangles giving $\ell = 1$. Right: A vertex-disjoint packing of a triangle and a K_4 giving $\ell = 0$. The solution consists of the three dashed edges.

can compute the bound $h(\mathcal{H})$ in $f(t) \cdot |G|^{O(1)}$ time using the generic algorithm [10] mentioned in the introduction for each $H \in \mathcal{H}$. In the same time we can also verify whether the cost-*t* property is fulfilled.

Packings of forbidden induced subgraphs have been used in implementations of fixed-parameter algorithms to prune the corresponding search trees tremendously [26]. By showing fixed-parameter algorithms for parameters above these lower bounds, we hope to explain the fact that these packings help in obtaining fast algorithms.

Our Results. We first state the negative results since they justify the focus on concrete problems and, to a certain extent, also the focus on parameterizing *edge* modification problems above lower bounds obtained from *vertex*-disjoint packings. We show that K_6 -FREE EDITING WITH K_6 -PACKING is NP-hard for $\ell = 0$. This proves, in particular, that a general fixed-parameter tractability result as it is known for the parameter k [10] cannot be expected. Moreover, we show that, if F is a triangle and \mathcal{H} is an *edge-disjoint* packing of h triangles in a graph G, then it is NP-hard to decide whether G has a triangle deletion set of size h (that is, $\ell = 0$). Thus, parameterization by ℓ is hopeless for this packing lower bound. We also consider vertex deletion problems. For these we show that extending the parameterization "above maximum matching" for VERTEX COVER to d-HITTING SET in a natural way leads to intractable problems. This is achieved by showing that, for all $q \ge 3$, P_q -FREE VERTEX DELETION WITH P_q -PACKING is NP-hard even if $\ell = 0$.

Our positive results are fixed-parameter algorithms and problem kernels (a notion for provably effective polynomial-time data reduction, see Section 2 for a formal definition) for three variants of *F*-FREE EDITING WITH COST-*t* PACKING. Namely, these are the variants in which *F* is a triangle (that is, a K_3) or a path on three vertices (that is, a P_3). The first case is known as TRIANGLE DELETION, the second one as CLUSTER EDITING. We also consider the case in which the input is a tournament graph and *F* is a directed cycle on three vertices. This is known as FEEDBACK ARC SET IN TOURNAMENTS. Using a general approach described in Section 3, we obtain fixed-parameter algorithms for these variants of *F*-FREE EDITING WITH COST-*t* PACKING parameterized by *t* and ℓ . This implies fixed-parameter tractability for *F*-FREE EDITING WITH *F*-PACKING parameterized by ℓ . Specifically, we obtain the following positive results:

- (i) For TRIANGLE DELETION, we show an $O((2t+3)^{\ell} \cdot (nm+n \cdot 2.076^{t}))$ -time algorithm and an $O(t \cdot \ell)$ -vertex problem kernel for cost-*t* packings.
- (ii) For FEEDBACK ARC SET IN TOURNAMENTS, we show a $2^{O(\sqrt{(2t+1)\ell})} \cdot n^{O(1)}$ -time algorithm and an $O(t \cdot \ell)$ -vertex problem kernel for cost-*t* packings.
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(iii) For CLUSTER EDITING, we show an $O(1.62^{(2t+1)\cdot\ell} + nm + n \cdot 1.62^t)$ -time algorithm and an $O(t \cdot \ell)$ -vertex kernel for cost-*t* packings, and a $4^{\ell} \cdot n^{O(1)}$ -time algorithm for P_3 -packings.

For the kernelization results, we need to assume that $t \in O(\log n)$ to guarantee polynomial running time of the data reduction.

Organization of this work. In Section 2, we introduce basic graph-theoretic notation and formally define fixed-parameter algorithms and problem kernelization. In Section 3, we present the general approach used in our algorithmic and data reduction results. In Section 4, we present our results regarding TRIANGLE DELETION, in Section 5 regarding FEEDBACK ARC SET IN TOURNAMENTS, and in Section 6 regarding CLUSTER EDITING. Section 7 shows vertex and edge deletion problems that remain NP-hard for $\ell = 0$, where ℓ is the number of modifications that are allowed in addition to a lower bound based on vertex-disjoint packings. We conclude with some open questions in Section 8.

2 Preliminaries

In this section, we introduce basic graph-theoretic notation and formally define fixedparameter algorithms and problem kernelization.

Notation. Unless stated otherwise, we consider undirected, simple, finite graphs G = (V, E), with a *vertex set* V(G) := V and an *edge set* $E(G) := E \subseteq {V \choose 2} := \{\{u, v\} \mid u, v \in V \land u \neq v\}$. Let n := |V(G)| denote the *order* of the graph and m := |E(G)| its number of edges. A set $S \subseteq {V \choose 2}$ is an *edge modification set* for *G*. For an edge modification set *S* for *G*, let $G \triangle S := (V, (E \setminus S) \cup (S \setminus E))$ denote the *graph obtained by applying S* to *G*. If $S \subseteq E$, then *S* is called an *edge deletion set* and we write $G \setminus S$ instead of $G \triangle S$. The *open neighborhood* of a vertex $v \in V$ is defined as $N_G(v) := \{u \in V \mid \{u, v\} \in E\}$. Also, for $V' \subseteq V$, let $G[V'] := (V', E \cap {V \choose 2})$ denote the subgraph of *G* induced by V'. A directed graph (or digraph) G = (V, A) consists of a vertex set V(G) and an *arc set* $A(G) := A \subseteq \{(u, v) \in V^2 \mid u \neq v\}$. A tournament on *n* vertices is a directed graph (V, A) with |V| = n such that, for each pair of distinct vertices *u* and *v*, either $(u, v) \in A$ or $(v, u) \in A$.

Fixed-parameter algorithms. The idea in fixed-parameter algorithms is to accept the exponential running time that seems to be inevitable when exactly solving NP-hard problems, yet to confine it to some small problem-specific parameter. A problem is *fixed-parameter tractable* with respect to some parameter *k* if there is a *fixed-parameter algorithm* solving any instance of size *n* in $f(k) \cdot n^{O(1)}$ time. We will also say that a problem is fixed-parameter tractable with respect to some combined parameter "*k* and ℓ " or "(*k*, ℓ)" if it is fixed-parameter tractable parameter tractable parameterized by $k + \ell$.

Fixed-parameter algorithms can efficiently solve instances in which the parameter k is small, even if the input size n is large. All vertex deletion, edge deletion, and edge modification problems for graph properties characterized by finite forbidden induced subgraphs are fixed-parameter tractable parameterized by the number of modifications k [10].

Problem kernelization. An important technique in fixed-parameter algorithmics is (*problem*) kernelization [31]—a formal approach of describing efficient and correct data reduction. A kernelization is an algorithm that given an instance x with parameter k, yields an instance x' with parameter k' in time polynomial in |x| + k such that (x,k) is a yes-instance if and only if (x',k') is a yes-instance, and if both |x'| and k' are bounded by some functions g and g' in k, respectively. The function g is referred to as the *size* of the *problem kernel* (x',k'). Kernelizations are commonly described by giving a set of *data reduction rules* which when applied to an instance x of a problem yield an instance x'. We say that a data reduction rule is *correct* if x and x' are equivalent.

All vertex deletion problems for establishing graph properties characterized by a finite number of forbidden induced subgraphs have a problem kernel of size polynomial in the parameter k of allowed modifications [30]. In contrast, many variants of F-FREE EDITING do not admit a problem kernel whose size is polynomial in k [11, 24, 32].

3 General Approach

In this section, we describe the general approach of our fixed-parameter algorithms. Recall that $\tau(H)$ is the minimum number of edge modifications required to transform a graph H into an F-free graph. We present fixed-parameter algorithms for three variants of F-free EDITING WITH COST-t PACKING parameterized by the combination of tand $\ell := k - h(\mathcal{H})$, where $h(\mathcal{H}) := \sum_{H \in \mathcal{H}} \tau(H)$. The idea behind the algorithms is to arrive at a classic win-win scenario [17] where we can either apply data reduction or show that the packing size $|\mathcal{H}|$ is bounded. This will allow us to bound k in $t \cdot \ell$ for yes-instances and, thus, to apply known fixed-parameter algorithms for the parameter kto obtain fixed-parameter tractability results for (t, ℓ) .

More precisely, we show that, for each induced subgraph H of G in a given packing \mathcal{H} , we face essentially two situations. If there is an optimal solution for H that is a subset of an optimal solution for G, then we can apply a data reduction rule. Otherwise, we find a certificate witnessing that H itself needs to be solved suboptimally or that a vertex pair containing exactly one vertex from H needs to be modified. We use the following terminology for these pairs.

Definition 3.1 (External vertex pairs and edges) A vertex pair $\{u, v\}$ is an *external pair for a packing graph* $H \in \mathcal{H}$ if exactly one of u or v is in V(H), an edge is an *external edge for* $H \in \mathcal{H}$ if exactly one of its endpoints is in H.

Observe that every pair or edge is an external pair or edge for at most two packing graphs since the packing graphs are vertex-disjoint. Therefore, the modification of an external vertex pair can destroy at most two certificates. This is the main fact used in the proof of the following bound on ℓ .

Lemma 3.2 Let (G, \mathcal{H}, k) be an instance of *F*-free Editing with Cost-*t* Packing and let *S* be a size-*k* solution that contains, for each $H = (W, F) \in \mathcal{H}$,

- (a) at least $\tau(H) + 1$ vertex pairs from $\binom{W}{2}$, or
- (b) at least one external vertex pair $\{v, w\}$ for H.
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Then, $|\mathcal{H}| \leq 2\ell$ and thus, $k \leq (2t+1)\ell$.

Proof Denote by $\mathcal{H}_a \subseteq \mathcal{H}$ the set of all graphs in \mathcal{H} that fulfill property (a) and let $p_a := |\mathcal{H}_a|$. Let $\mathcal{H}_b := \mathcal{H} \setminus \mathcal{H}_a$ denote the set containing the remaining packing graphs (fulfilling property (b)) and let $p_b := |\mathcal{H}_b|$. Thus, $|\mathcal{H}| = p_a + p_b$. Furthermore, let $h_a := \sum_{H \in \mathcal{H}_a} \tau(H)$ denote the lower bound obtained from the graphs in \mathcal{H}_a and let $h_b := h(\mathcal{H}) - h_a$ denote the part of the lower bound obtained by the remaining graphs.

The packing graphs in \mathcal{H}_a cause $h_a + p_a$ edge modifications inside of them. Similarly, the packing graphs in \mathcal{H}_b cause at least h_b edge modifications inside of them, and each packing graph $H \in \mathcal{H}_b$ additionally causes modification of at least one external vertex pair for H. Since every vertex pair is an external pair for at most two different packing graphs, at least $h_b + p_b/2$ edge modifications are caused by the graphs in \mathcal{H}_b . This implies that

$$k \ge h_a + h_b + p_a + p_b/2$$

$$\Leftrightarrow \qquad k - h(\mathcal{H}) \ge p_a + p_b/2$$

$$\Leftrightarrow \qquad 2\ell \ge 2p_a + p_b \ge |\mathcal{H}|.$$

Consequently, $k = \ell + h(\mathcal{H}) \le \ell + t \cdot |\mathcal{H}| \le \ell + t \cdot 2\ell = (2t+1)\ell$.

4 Triangle Deletion

In this section, we study TRIANGLE DELETION, the problem of destroying all *trian*gles (K_3 s) in a graph by at most k edge deletions. In Section 4.1, we apply our framework from Section 3 to show that TRIANGLE DELETION is fixed-parameter tractable parameterized above the lower bound given by a cost-t packing. In Section 4.2, we then show that parameterization above a lower bound given by edge-disjoint packings of triangles does not lead to fixed-parameter algorithms unless P = NP.

4.1 A fixed-parameter algorithm for vertex-disjoint cost-t packings

Before presenting our new fixed-parameter tractability results for TRIANGLE DELETION, let us first summarize the known results concerning the (parameterized) complexity of TRIANGLE DELETION. TRIANGLE DELETION is NP-complete [47]. It allows for a trivial reduction to 3-HITTING SET since edge deletions do not create new triangles [23]. Combining this approach with the currently fastest known algorithms for 3-HITTING SET [5, 46] gives an algorithm for TRIANGLE DELETION with running time $O(2.076^k + nm)$. Finally, TRIANGLE DELETION admits a problem kernel with at most 6k vertices [9]. We show that TRIANGLE DELETION WITH COST-*t* PACKING is fixed-parameter tractable with respect to the combination of *t* and $\ell := k - h(\mathcal{H})$. More precisely, we obtain a kernelization and a search tree algorithm. Both make crucial use of the following generic reduction rule for TRIANGLE DELETION WITH COST-*t* PACKING.

Reduction Rule 4.1 If there is an induced subgraph $H \in \mathcal{H}$ and a set $T \subseteq E(H)$ of $\tau(H)$ edges such that deleting *T* destroys all triangles of *G* that contain edges of *H*, then delete *T* from *G*, *H* from \mathcal{H} and decrease *k* by $\tau(H)$.



Lemma 4.2 Reduction Rule 4.1 is correct.

Proof Let (G, \mathcal{H}, k) be the instance to which Reduction Rule 4.1 is applied and let $(G', \mathcal{H} \setminus \{H\}, k - \tau(H))$ with $G' := G \setminus T$ be the result. We show that (G, \mathcal{H}, k) is a yes-instance if and only if $(G', \mathcal{H} \setminus \{H\}, k - \tau(H))$ is.

First, let *S* be a solution of size at most *k* for (G, \mathcal{H}, k) . Let $S_H := S \cap E(H)$ denote the set of edges of *S* that destroy all triangles in *H*. By definition, $|S_H| \ge \tau(H)$. Since $S_H \subseteq E(H)$, only triangles containing at least one edge of *H* are destroyed by deleting S_H . It follows that the set of triangles destroyed by S_H is a subset of the triangles destroyed by *T*. Hence, $(S \setminus S_H) \cup T$ has size at most *k* and clearly is a solution for (G, \mathcal{H}, k) that contains all edges of *T*. Thus, $(S \setminus S_H)$ is a solution of size $k - \tau(H)$ for $G \setminus T = G'$ and $(G', \mathcal{H} \setminus \{H\}, k - \tau(H))$ is a yes-instance.

For the converse direction, let S' be a solution of size at most $k - \tau(H)$ for $(G', \mathcal{H} \setminus \{H\}, k - \tau(H))$. Since $T \subseteq E(H)$, it holds that every triangle contained in G that does not contain any edge of H is also a triangle in G'. Thus, S' is a set of edges whose deletion in G destroys all triangles that do not contain any edge of H. Since T destroys all triangles containing an edge of H, we have that $T \cup S'$ is a solution for G. Its size is k.

We now show that, if Reduction Rule 4.1 is not applicable to H, then we can find a *certificate* for this, which will allow us later to branch efficiently on the destruction of triangles:

Definition 4.3 (Certificate) A *certificate* for inapplicability of Reduction Rule 4.1 to an induced subgraph $H \in \mathcal{H}$ is a set \mathcal{T} of triangles in *G*, each containing exactly one distinct edge of *H*, such that $|\mathcal{T}| = \tau(H) + 1$ or $|\mathcal{T}| \leq \tau(H)$ and $\tau(H') > \tau(H) - |\mathcal{T}|$, where *H'* is the subgraph obtained from *H* by deleting, for each triangle in \mathcal{T} , its edge shared with *H*.

Lemma 4.4 Let $\Gamma(G, k)$ be the time needed to compute a triangle-free deletion set of size at most k in a graph G if it exists.

In $O(nm + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time, we can apply Reduction Rule 4.1 to all $H \in \mathcal{H}$ and output a certificate \mathcal{T} if Reduction Rule 4.1 is inapplicable to some $H \in \mathcal{H}$.

In the statement of the lemma, we assume that Γ is monotonically nondecreasing in the size of *G* and in *k*. As described above, currently $O(2.076^k + |V(G)| \cdot |E(G)|)$ is the best known bound for $\Gamma(G, k)$.

Proof (of Lemma 4.4) First, in O(nm) time, we compute for all $H \in \mathcal{H}$ all triangles \mathcal{T} that contain exactly one edge $e \in E(H)$. These edges are labeled in each $H \in \mathcal{H}$. Then, for each $H \in \mathcal{H}$, in $\Gamma(H, t)$ time we determine the size $\tau(H)$ of an optimal triangle-free deletion set for H. Let t' denote the number of labeled edges of H.

Case 1: $t' > \tau(H)$. In this case, we return as certificate $\tau(H) + 1$ triangles of \mathcal{T} , each containing a distinct of $\tau(H) + 1$ arbitrary labeled edges.

Case 2: $t' \le \tau(H)$. Let H' denote the graph obtained from H by deleting the labeled edges. All triangles of G that contain at least one edge of H either contain a labeled edge or they are contained in H'. Thus, we now determine in $\Gamma(H', \tau(H) - t')$ time whether H' can be made triangle-free by $\tau(H) - t'$ edge deletions. If this is the

case, then the rule applies and the set *T* consists of the solution for *H'* plus the deleted labeled edges. Otherwise, destroying all triangles that contain exactly one edge from *H* leads to a solution which needs more than $\tau(H)$ edge deletions and thus the rule does not apply. In this case, we return the certificate \mathcal{T} for this $H \in \mathcal{H}$.

The overall running time now follows from the monotonicity of f, from the fact that $|\mathcal{H}| \leq n$, and from the fact that one pass over \mathcal{H} is sufficient since deleting edges in each H does not produce new triangles and does not destroy triangles in any $H' \neq H$.

Observe that Reduction Rule 4.1 never increases the parameter ℓ since we decrease both *k* as well as the lower bound $h(\mathcal{H})$ by $\tau(\mathcal{H})$. After application of Reduction Rule 4.1, we can upper-bound the solution size *k* in terms of *t* and ℓ , which allows us to transfer parameterized complexity results for the parameter *k* to the combined parameter (t, ℓ) .

Lemma 4.5 Let (G, \mathcal{H}, k) be a yes-instance of TRIANGLE DELETION WITH COST-*t* PACKING such that Reduction Rule 4.1 is inapplicable. Then, $k \leq (2t + 1)\ell$.

Proof Since (G, \mathcal{H}, k) is reduced with respect to Reduction Rule 4.1, for each graph H = (W, F) in \mathcal{H} , there is a set of edges between W and $V \setminus W$ witnessing that every optimal solution for H does not destroy all triangles containing at least one edge from H. Consider any optimal solution S. For each graph $H \in \mathcal{H}$, there are two possibilities: Either at least $\tau(H) + 1$ edges inside H are deleted by S, or at least one external edge of H is deleted by S. Therefore, S fulfills the condition of Lemma 3.2 and thus $k \leq (2t+1)\ell$.

Theorem 4.6 Let $\Gamma(G, k)$ be the time used for computing a triangle-free deletion set of size at most k in a graph G if it exists. Then, TRIANGLE DELETION WITH COST-t PACKING

- (i) can be solved in $O((2t+3)^{\ell} \cdot (nm + \sum_{H \in \mathcal{H}} \Gamma(H, t)))$ time, and
- (ii) admits a problem kernel with at most $(12t + 6)\ell$ vertices that can be computed in $O(nm + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time.

Proof We first prove (ii). To this end, let $(G = (V, E), \mathcal{H}, k)$ be the input instance. First, compute in $O(nm + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time an instance that is reduced with respect to Reduction Rule 4.1. Afterwards, by Lemma 4.5, we can reject if $k > (2t + 1)\ell$. Otherwise, we apply the known kernelization algorithm for TRIANGLE DELETION to the instance (G, k) (that is, without \mathcal{H}). This kernelization produces in $O(m\sqrt{m}) = O(nm)$ time a problem kernel (G', k') with at most $6k \leq (12t + 6)\ell$ vertices and with $k' \leq k$ [9]. Adding an empty packing gives an equivalent instance (G', \emptyset, k') with parameter $\ell' = k' \leq (2t + 1)\ell$ of TRIANGLE DELETION WITH COST-*t* PACKING.

It remains to prove (i). To this end, first apply Reduction Rule 4.1 exhaustively in $O(nm + n \cdot \Gamma(H, t))$ time. Now, consider a reduced instance. If $\ell < 0$, then we can reject the instance. Otherwise, consider the following two cases.

Case 1: $\mathcal{H} = \emptyset$. If G is triangle-free, then we are done. Otherwise, pick an arbitrary triangle in G and add it to \mathcal{H} .

Case 2: \mathcal{H} *contains a graph H.* Since Reduction Rule 4.1 does not apply to H, there is a certificate \mathcal{T} of $t' \leq \tau(H) + 1$ triangles, each containing exactly one distinct



edge of *H* such that deleting the edges of these triangles contained in *H* produces a subgraph *H'* of *H* that cannot be made triangle-free by $\tau(H) - t'$ edge deletions. Thus, branch into the following (2t' + 1) cases: First, for each triangle $T \in \mathcal{T}$, create two cases, in each deleting a different one of the two edges of *T* that are not in *H*. In the remaining case, delete the *t'* edges of *H* and replace *H* by *H'* in \mathcal{H} .

It remains to show the running time by bounding the search tree size. In Case 1, no branching is performed and the parameter is decreased by at least one. In Case 2, the parameter value is decreased by one in each branch: in the first 2t' cases, an edge that is not contained in any packing graph is deleted. Thus, k decreases by one while $h(\mathcal{H})$ remains unchanged. In the final case, the value of k decreases by t' since this many edge deletions are performed. However, $\tau(H') \ge \tau(H) - t' + 1$. Hence, the lower bound $h(\mathcal{H})$ decreases by at most t' - 1 and thus the parameter ℓ decreases by at least one. Note that applying Reduction Rule 4.1 never increases the parameter. Hence, the depth of the search tree is at most ℓ .

Corollary 4.7 TRIANGLE DELETION WITH COST-*t* PACKING

- (i) can be solved in $O((2t+3)^{\ell} \cdot (nm+n \cdot 2.076^{t}))$ time, and
- (ii) admits a problem kernel with at most $(12t + 6)\ell$ vertices that can be computed in $O(nm + n \cdot 2.076^t)$ time.

For the natural special case t = 1, that is, for triangle packings, Theorem 4.6(i) immediately yields the following running time.

Corollary 4.8 TRIANGLE DELETION WITH TRIANGLE PACKING *is solvable in O*($5^{\ell}nm$) *time.*

4.2 Hardness for edge-disjoint packing

We complement the positive results of Theorem 4.6 by the following hardness result for the case of edge-disjoint triangle packings:

Theorem 4.9 TRIANGLE DELETION is NP-hard even for $\ell := k - |\mathcal{H}| = 0$ if \mathcal{H} is an edge-disjoint packing of triangles.

Theorem 4.9 shows that parameterizing TRIANGLE DELETION over a lower bound given by edge-disjoint packings cannot lead to fixed-parameter algorithms unless P = NP. We prove Theorem 4.9 using a reduction from 3-SAT.

Problem 4.10 (3-SAT)

Input: A Boolean formula $\phi = C_1 \land \ldots \land C_m$ in conjunctive normal form over variables x_1, \ldots, x_n with at most three variables per clause.

Question: Does ϕ have a satisfying assignment?

Construction 4.11 Given a Boolean formula ϕ , we create a graph *G* and an edgedisjoint packing \mathcal{H} of triangles such that *G* can be made triangle-free by exactly $|\mathcal{H}|$ edge deletions if and only if there is a satisfying assignment for ϕ . We assume that each clause of ϕ contains exactly three pairwise distinct variables. The construction is illustrated in Figure 4.1.

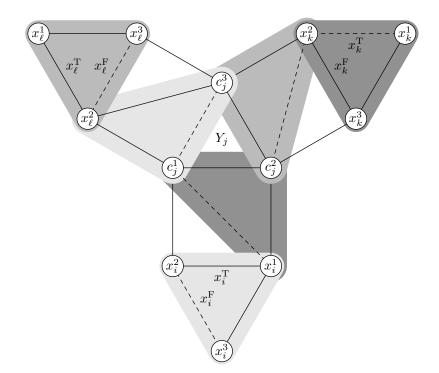


Fig. 4.1 Construction for a clause $C_j = (x_i \vee \neg x_k \vee \neg x_\ell)$. The triangles on a gray background are contained in the mutually edge-disjoint triangle packing \mathcal{H} . Deleting the dashed edges corresponds to setting x_i and x_ℓ to false and x_k to true, thus satisfying C_j . Note that it is impossible to destroy triangle Y_j by $|\mathcal{H}|$ edge deletions if we delete $x_i^{\text{F}}, x_k^{\text{T}}$, and x_ℓ^{T} , which corresponds to the fact that clause C_j cannot be satisfied by this variable assignment.

For each variable x_i of ϕ , create a triangle X_i on the vertex set $\{x_i^1, x_i^2, x_i^3\}$ with two distinguished edges $x_i^T := \{x_i^1, x_i^2\}$ and $x_i^F := \{x_i^2, x_i^3\}$ and add X_i to \mathcal{H} . For each clause $C_j = (l_1, l_2, l_3)$ of ϕ , create a triangle Y_j on the vertex set $\{c_j^1, c_j^2, c_j^3\}$ with three edges $c_j^{l_1}, c_j^{l_2}$, and $c_j^{l_3}$. Connect the clause gadget Y_j to the variable gadgets as follows: If $l_t = x_i$, then connect the edge $c_j^{l_i} =: \{u, v\}$ to the edge $x_i^T = \{x_i^1, x_i^2\}$ via two adjacent triangles $A_{ij} := \{u, v, x_i^1\}$ and $B_{ij} := \{v, x_i^1, x_i^2\}$ sharing the edge $\{v, x_i^1\}$. The triangle A_{ij} is added to \mathcal{H} . If $l_t = \neg x_i$, then connect the edge $c_j^{l_i} =: \{u, v\}$ to the edge $x_i^F = \{x_i^2, x_i^3\}$ via two adjacent triangles $A_{ij} := \{u, v, x_i^3\}$ and $B_{ij} := \{v, x_i^2, x_i^3\}$ sharing the edge $\{v, x_i^3\}$. The triangle A_{ij} is added to \mathcal{H} .

Proof (of Theorem 4.9) First, observe that Construction 4.11 introduces no edges between distinct clause gadgets or distinct variable gadgets. Thus, under the assumption that each clause contains each variable at most once, the only triangles in the constructed graph are the X_i , the Y_j , the A_{ij} and B_{ij} for all variables x_i and the incident clauses C_j .

Now, assume that ϕ allows for a satisfying assignment. We construct a set of edges *S* of size $|\mathcal{H}|$ such that $G' := (V, E \setminus S)$ is triangle-free. For each variable x_i that

is true, add x_i^{T} to *S*. For each variable x_i that is false, add x_i^{F} to *S*. By this choice, the triangle X_i is destroyed in *G'* for each variable x_i . Additionally, for each clause C_j and its *true* literals $l \in \{x_i, \neg x_i\}$, the triangle B_{ij} is destroyed. To destroy A_{ij} , we add to *S* the edge of A_{ij} shared with Y_j , which also destroys the triangle Y_j . For each clause C_j containing a *false* literal $l \in \{x_i, \neg x_i\}$, we destroy B_{ij} and simultaneously A_{ij} by adding to *S* the edge of A_{ij} shared with B_{ij} .

Conversely, assume that there is a set *S* of size $|\mathcal{H}|$ such that $G' = (V, E \setminus S)$ is triangle-free. We construct a satisfying assignment for ϕ . First, observe that, since the triangles in \mathcal{H} are pairwise edge-disjoint, *S* contains exactly one edge of each triangle in \mathcal{H} . Thus, of each triangle X_i , at most one of the two edges x_i^F and x_i^T is contained in *S*. The set *S* contains at least one edge *e* of each Y_j . This edge is shared with a triangle A_{ij} . Since $A_{ij} \in \mathcal{H}$ and, with *e*, *S* already contains one edge of A_{ij} , *S* does not contain the edge shared between A_{ij} and B_{ij} . Since $B_{ij} \notin \mathcal{H}$, *S* has to contain an edge of B_{ij} shared with another triangle in \mathcal{H} . If the clause C_j contains x_i , then the only such edge is x_i^T and we set x_i to true. If the clause C_j contains $\neg x_i$, then the only such edge is x_i^F and we set x_i to false. In both cases, clause C_j is satisfied. Since at most one of x_i^T and x_i^F is in *S*, the value of each variable x_i is well-defined.

5 Feedback Arc Set in Tournaments

In this section, we present a fixed-parameter algorithm and a problem kernel for FEED-BACK ARC SET IN TOURNAMENTS parameterized above lower bounds of cost-*t* packings.

In FEEDBACK ARC SET IN TOURNAMENTS, we are given a directed tournament graph G as input and want to delete a minimum number of arcs to make the graph acyclic, that is, to destroy all directed cycles in G. Due to a well-known observation, one can also view FEEDBACK ARC SET IN TOURNAMENTS as an arc *reversal* problem: After deleting a minimum set of arcs to make the graph acyclic, adding the arc (u, v) for every deleted arc (v, u) does not create any cycle. Since, in tournaments, every pair of vertices is connected by exactly one arc, it follows that that destroying cycles by edge deletions is equivalent to destroying them by arc reversals. Altogether, we arrive at the following problem definition.

Problem 5.1 (FEEDBACK ARC SET IN TOURNAMENTS (FAST))

Input: An *n*-vertex tournament G = (V, A) and a natural number *k*. *Question:* Does *G* have a *feedback arc set* $S \subseteq A$, that is, a set *S* such that reversing all arcs in *S* yields an acyclic tournament, of size at most *k*?

FAST is NP-complete [1] but fixed-parameter tractable with respect to k [2, 14, 16, 20, 27, 42]. The running time of the current best fixed-parameter algorithm is $2^{c \cdot \sqrt{k}} + n^{O(1)}$ where $c \le 5.24$ [20]. Moreover, a problem kernel with $(2 + \epsilon)k$ vertices for each *constant* $\epsilon > 0$ is known [4] as well as a simpler 4k-vertex kernel [40]. It is well-known that a tournament is acyclic if and only if it does not contain a *directed triangle* (a cycle on 3 vertices). Hence, the problem is to find a set of arcs whose reversal leaves no directed triangle in the tournament.

We show fixed-parameter tractability of FAST with Cost-*t* Packing parameterized by the combination of *t* and $\ell := k - h(\mathcal{H})$. Recall that $h(\mathcal{H}) := \sum_{H \in \mathcal{H}} \tau(H) \ge |\mathcal{H}|$,

where $\tau(G)$ is the size of a minimum feedback arc set for a directed graph *G*. The approach is the same as for TRIANGLE DELETION in Section 4, that is, we upper-bound the solution size *k* in *t* and ℓ and apply the fixed-parameter algorithm for *k* [27]. Observe in this context that Lemma 3.2 is also correct if the input graphs are directed and if a solution contains arc reversals, since we observed arc reversals and deletions to be equivalent in the context of FAST.² We use the following reduction rule for FAST analogous to Reduction Rule 4.1 for TRIANGLE DELETION.

Reduction Rule 5.2 If there is a subtournament $H \in \mathcal{H}$ and a feedback arc set $T \subseteq A(H)$ of size $\tau(H)$ such that reversing the arcs in *T* leaves no directed triangles in *G* containing arcs of *H*, then reverse the arcs in *T*, remove *H* from \mathcal{H} , and decrease *k* by $\tau(H)$.

Although Reduction Rule 5.2 is strikingly similar to Reduction Rule 4.1, its correctness proof is significantly more involved.

Lemma 5.3 Reduction Rule 5.2 is correct and, given the tournaments *G* and *H* it can be applied in $O\left(\binom{q}{2}(n-q) + \Gamma(H,t)\right)$ time, where q := |V(H)| and $\Gamma(H,t)$ denotes the running time needed to compute a feedback arc set of size at most *t* in *H* if it exists.

Proof We first show correctness. Let $I' := (G', \mathcal{H} \setminus \{H\}, k - \tau(H))$ be the instance created by Reduction Rule 5.2 from $I := (G, \mathcal{H}, k)$ by reversing a subset T of arcs of a subtournament $H \in \mathcal{H}$ of G. If I' is a yes-instance, then so is I since G' is the graph G with the $\tau(H)$ arcs in T reversed and, thus, adding these arcs to a feedback arc set of size $k - \tau(H)$ for G' gives a feedback arc set of size k for G. It remains to prove that if I is a yes-instance, then so is I'. To this end, we show that there is a minimum-size feedback arc set S for G with $T \subseteq S$.

Let *S* be a minimum-size feedback arc set for G = (V, E). This implies the existence of a linear ordering $\sigma_S = v_1, \ldots, v_n$ of the vertices *V* such that there are |S| backward arcs, that is, arcs $(v_i, v_j) \in A$ such that i > j. Now, let $\sigma_T = w_1, \ldots, w_{|W|}$ be the ordering of the vertices of H = (W, F) corresponding to the local solution *T* for *H* with $\tau(H)$ backward arcs. Let $N^+(w) := \{(w, v) \in A \mid v \in V \setminus W\}$ denote the *out-neighbors* in $V \setminus W$ of a vertex $w \in W$. Analogously, $N^-(w) := \{(v, w) \in A \mid v \in V \setminus W\}$ denotes the set of *in-neighbors*. By the assumption of the rule, for all i < j,

$$N^{+}(w_{i}) \subseteq N^{+}(w_{i}) \text{ and } N^{-}(w_{i}) \subseteq N^{-}(w_{i})$$
 (5.1)

holds since otherwise, after reversing the arcs in T, there exists a directed cycle containing an arc of H (because the arc (w_i, w_j) is present).

If the vertices of *W* appear in σ_S in the same relative order as in σ_T , then we have $T \subseteq S$ and we are done. Otherwise, we show that we can swap the positions of vertices of *W* in σ_S so that their relative order is the same as in σ_T without increasing the number of backward arcs.

First, note that the number of backward arcs between vertices in $V \setminus W$ does not change when only swapping positions of vertices in W. Also, by assumption, the number of backward arcs between vertices in W in any ordering is at least $\tau(H)$,

 $^{^{2}}$ For directed input graphs, we use the term *external arc* instead of *external edge*.

¹³

whereas it is exactly $\tau(H)$ when ordering them according to σ_T . Thus, it remains to show that the number of backward arcs between vertices in W and $V \setminus W$ is not increased. To this end, consider a series of *swaps* of pairs of vertices w_j and w_i such that i < j, where w_j appears before w_i in σ_S , reordering the vertices in W according to σ_T . Let Y denote the set of all vertices that lie between w_j and w_i in σ_S . Note that swapping w_j and w_i removes the backward arcs from w_i to the vertices in $N^+(w_i) \cap Y$ and the backward arcs from vertices in $N^-(w_j) \cap Y$ to w_j , whereas it introduces new backward arcs from w_j to $N^+(w_j) \cap Y$ and from $N^-(w_i) \cap Y$ to w_i . However, by the inclusions in (1), it follows that the overall number of backward arcs does not increase in each swap. Hence, the overall number of backward arcs is not increased by repositioning the vertices in W according to σ_T . It follows that there is an optimal solution containing T.

It remains to show the running time. First, in $\Gamma(H, \tau(H))$ time, we compute the size $\tau(H)$ of an optimal feedback arc set for H = (W, F). Now, for each arc $(u, v) \in F$, we check whether there is a vertex $w \in V \setminus W$ that forms a directed triangle with u and v. If such a vertex exists, then we reverse the arc (u, v). If this arc reversal introduces a new directed triangle with another vertex from $V \setminus W$, then the rule does not apply. Overall, this procedure requires $O(|F| \cdot (n - |W|))$ time. Let T^* denote the set of arcs that are reversed in this process. Clearly, if $|T^*| > \tau(H)$, then the rule does not apply. Otherwise, let H' denote the graph obtained from H by reversing the arcs in T^* and observe that each remaining directed triangle of G that contains at least one arc of H'is contained in H'. Thus, we now compute whether H' has a feedback arc set T' of size $\tau(H) - |T^*|$ in $\Gamma(H', \tau(H) - |T^*|)$ time. If this is the case, then the rule applies and we set $T := T' \cup T^*$ (note that $T' \cap T^* = \emptyset$, since otherwise $|T| < \tau(H)$, which is not possible by definition of $\tau(H)$). Otherwise, removing all directed triangles that contain at least one arc from H requires more than $\tau(H)$ arc reversals and thus the rule does not apply.

Exhaustive application of Reduction Rule 5.2 allows us to show that $k \le (2t + 1)\ell$ holds for any yes-instance (analogous to Lemma 4.5).

Lemma 5.4 Let (G, \mathcal{H}, k) be a yes-instance of FEEDBACK ARC SET IN TOURNAMENTS WITH COST-*t* PACKING such that Reduction Rule 5.2 cannot be applied to any tournament in \mathcal{H} . Then, $k \leq (2t + 1)\ell$.

Proof Since Reduction Rule 5.2 cannot be applied to any tournament in \mathcal{H} , for each tournament H = (W, F) in \mathcal{H} , there is a set of arcs between W and $V \setminus W$ that witness that no optimal feedback arc set for H removes all directed triangles containing at least one arc from F.

Now, for any optimal solution, there are two possibilities for each packing tournament $H \in \mathcal{H}$:

(a) at least τ(H) + 1 arcs in H are reversed, or
(b) at least one external arc of H is reversed.

Therefore, *S* fulfills the condition of Lemma 3.2 and, thus, $|\mathcal{H}| \le 2\ell$.

The bound on k yields the following two fixed-parameter tractability results.

Theorem 5.5 Let $\Gamma(G, k)$ be the running time used for finding a minimum feedback arc set of size at most k for a given tournament G if it exists. Then, FEEDBACK ARC SET IN TOURNAMENTS WITH COST-*t* PACKING

- (i) is solvable in $\Gamma(G, (2t+1)\ell) + n^{O(1)} + \sum_{H \in \mathcal{H}} \Gamma(H, t)$ time, and
- (ii) admits a problem kernel with at most $(8t + 4)\ell$ vertices computable in $n^{O(1)} + \sum_{H \in \mathcal{H}} \Gamma(H, t)$ time.

Proof (i) Given (G, \mathcal{H}, k) , we first apply Reduction Rule 5.2 for each $H \in \mathcal{H}$. This application can be performed in $O(|\mathcal{H}|n^3 + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time by Lemma 5.3 since $|\mathcal{H}| \leq n$. One pass of this rule is sufficient to obtain an instance that is reduced: reversing arcs in some $H \in \mathcal{H}$ does not remove any directed triangles containing arcs of any other $H' \in \mathcal{H}$ with $H' \neq H$. By Lemma 5.4, we can then reject the instance if $k > (2t + 1)\ell$. Otherwise, we can find a solution in $\Gamma(G, (2t + 1)\ell)$ time.

(ii) First, we apply Reduction Rule 5.2 once for each $H \in \mathcal{H}$ in $O(|\mathcal{H}|n^3 + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time. After one pass of this rule, the instance is reduced since reversing arcs in some $H \in \mathcal{H}$ does not remove any directed triangles containing arcs of any other $H' \in \mathcal{H}$ with $H' \neq H$. Afterwards, by Lemma 5.4, we can reject if $k > (2t + 1)\ell$. Otherwise, we apply the kernelization algorithm for FAST by Paul et al [40] to the instance (G, k) to obtain an equivalent instance (G', ℓ') with at most $4k \leq (8t + 4)\ell$ vertices and a solution size $k' \leq k$. Hence, (G', \emptyset, k') is our problem kernel with parameter $\ell' = k' \leq (2t + 1)\ell$ of FEEDBACK ARC SET IN TOURNAMENTS WITH COST-*t* PACKING.

In Theorem 5.5, we again assume that Γ is monotonically nondecreasing in both the size of *G* and in *k*. As mentioned earlier, $2^{O(\sqrt{k})} + |V(G)|^{O(1)}$ is the currently best known running time for $\Gamma(G, k)$ [27].

Corollary 5.6 Feedback Arc Set in Tournaments with Cost-t Packing

- (i) can be solved in $2^{O(\sqrt{(2t+1)\ell})} + n^{O(1)} + n^{2O(\sqrt{t})}$ time, and
- (ii) admits a problem kernel with at most $(8t + 4)\ell$ vertices computable in $n^{O(1)} + n2^{O(\sqrt{t})}$ time.

6 Cluster Editing

We finally apply our framework from Section 3 to CLUSTER EDITING, a well-studied edge modification problem in parameterized complexity [7, 12, 21, 28].

Problem 6.1 (Cluster Editing)

Input: A graph G = (V, E) and a natural number k.

Question: Is there an edge modification set $S \subseteq \binom{V}{2}$ of size at most k such that $G \triangle S$ is a cluster graph, that is, a disjoint union of cliques?

A graph is a cluster graph if and only if it is P_3 -free [45]. Thus, CLUSTER EDITING is the problem of destroying all P_3 s by few edge modifications. For brevity, we refer to the connected components of a cluster graph (which are cliques) and to their vertex sets as *clusters*. The currently fastest algorithm for CLUSTER EDITING parameterized by

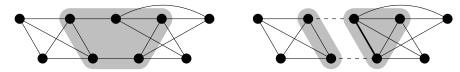


Fig. 6.1 An illustration of Reduction Rule 6.2. Left: An induced subgraph H (highlighted by the gray background) fulfilling the conditions of Reduction Rule 6.2. Right: The result of applying Reduction Rule 6.2. The two clusters in G[W] produced by the optimal solution for H are highlighted by a gray background.

the solution size k runs in $O(1.62^k + n + m)$ time [7]. Assuming the exponential-time hypothesis, CLUSTER EDITING cannot be solved in $2^{o(k)} \cdot n^{O(1)}$ time [21, 28]. CLUSTER EDITING admits a problem kernel with at most 2k vertices [12].

First, in Section 6.1, we present a fixed-parameter algorithm and problem kernel for CLUSTER EDITING parameterized above lower bounds given by cost-*t* packings. Then, in Section 6.2, we present a faster fixed-parameter algorithm for lower bounds given by vertex-disjoint packings of P_{3} s.

6.1 A fixed-parameter algorithm for vertex-disjoint cost-t packings

Several kernelizations for CLUSTER EDITING are based on the following observation: If G contains a clique such that all vertices in this clique have the same closed neighborhood, then there is an optimal solution that puts these vertices into the same cluster [12, 25, 41]. This implies that the edges of this clique are never deleted. The following rule is based on a generalization of this observation.

Reduction Rule 6.2 If G = (V, E) contains an induced subgraph H = (W, F) having an optimal solution *S* of size $\tau(H)$ such that, for all vertices $u, v \in W$,

- $N_G(v) \setminus W = N_G(u) \setminus W$ if u and v are in the same cluster of $H \triangle S$, and - $N_G(v) \cap N_G(u) \subseteq W$ otherwise,

then replace G by $G \triangle S$, remove H from \mathcal{H} , and decrease k by $\tau(H)$.

An example of Reduction Rule 6.2 is presented in Figure 6.1.

Lemma 6.3 Reduction Rule 6.2 is correct.

Proof Let $I' := (G \triangle S, \mathcal{H}', k - \tau(H))$ be the instance obtained by applying Reduction Rule 6.2 to $I := (G, \mathcal{H}, k)$ for some induced subgraph H = (W, F) of G = (V, E). If I' is a yes-instance, then so is I: adding the $\tau(H)$ edges in S to any solution S' of size $k - \tau(H)$ for $G \triangle S$ gives a solution of size k for G since $S \cap S' = \emptyset$ and, thus, $G \triangle (S \cup S') = (G \triangle S) \triangle S'$. It remains to prove that if I is a yes-instance, then so is I'. To this end, we show that (G, \mathcal{H}, k) has an optimal solution S' such that $S \subseteq S'$.

For convenience, let $X := V \setminus W$ denote the set of vertices not in W. Let S^* be any optimal solution for G, and denote by $G^* := G \triangle S^*$ the cluster graph produced by S^* . We show how to transform S^* into an optimal solution $S' \supseteq S$. To this end, partition S^* as follows:

- $S_1^* := \{\{u, v\} \in S^* \mid u, v \in X\}$ containing all edge modifications outside of H, $S_2^* := \{\{u, v\} \in S^* \mid u \in X \land v \in W\}$ containing the edge modifications between Hand the rest of G, and
- $S_3^* := \{\{u, v\} \in S^* \mid u, v \in W\}.$

Moreover, let W_{ext} be the vertices of W that have at least one neighbor in X and let $W_{int} := W \setminus W_{ext}$. Consider the following equivalence relation ~ on W_{ext} : two vertices $u, v \in W_{ext}$ are equivalent with respect to ~ if and only if $N_G(u) \cap X = N_G(v) \cap X$. For each vertex $u \in W_{ext}$, let [u] denote the equivalence class of u in ~.

Fix within each equivalence class [u] of ~ an arbitrary vertex that is incident to a minimum number of edge modifications in S_2^* and, for each vertex u, denote this vertex by \tilde{u} . Furthermore, for each cluster K of G^* containing some vertices of X and some vertices of W_{ext} , fix an arbitrary vertex of W_{ext} that has in G a maximum number of neighbors in $K \cap X$; denote this vertex by u_K . Finally, call a vertex $u \in W_{ext}$ good if there is a cluster K such that $\tilde{u} = u_K$. Now consider the edge modification set $S' := S_1^* \cup \tilde{S} \cup S$, where

$$\widetilde{S} := \left\{ \{u, v\} \mid u \text{ is good and } \{\widetilde{u}, v\} \in S_2^* \right\} \cup \cup \left\{ \{u, v\} \mid u \text{ is not good and } v \in N_G(u) \cap X \right\}.$$

Informally, the modifications in \tilde{S} consider \tilde{u} to determine how to treat all vertices in the equivalence class [u]. If, among the vertices of W_{ext} , \tilde{u} has the most neighbors in its cluster in G^* , then all vertices of [u] are treated like \tilde{u} . Otherwise, all edges between vertices in [*u*] and *X* are deleted.

We first show that S' is a solution, that is, $G' := G \triangle S'$ is a cluster graph: First, $G'[X] = G^*[X]$ is a cluster graph. Second, $G'[W] = H \triangle S$ and thus it is a cluster graph. Third, every vertex $u \in W_{int}$ is contained in a cluster that is a subset of W_{int} in G': In G, there are no edges between W_{int} and X, no edges between W_{int} and X are added by \tilde{S} , and by the first condition on S, no vertex of W_{int} is in the same cluster of $H \triangle S$ as a vertex from W_{ext} . This implies that the connected component of each vertex $u \in W_{\text{int}}$ is completely contained in W and thus it is a clique since G'[W] is a cluster graph. Finally, consider any equivalence class [u] of W_{ext} . By the condition of Reduction Rule 6.2, [u] is contained in a cluster in $H \triangle S$. Now, if all vertices of [u] are good, then there is a cluster K in G'[X] such that in G' all vertices of [u] are adjacent to all vertices of K. Otherwise, no vertex of [u] is good and thus, no vertex of [u] is adjacent to any vertex of X in G'. Finally, if a cluster K of G'[X] has neighbors in W, then these edges are only to the vertex set $[u_K]$. Thus, every vertex in K has neighbors in at most one cluster of $H \triangle S$. Altogether this shows that G' is a cluster graph.

It remains to show that $|S'| \leq |S^*|$. First, S_1^* is a subset of S^* and of S'. Second, $|S| \leq |S_3^*|$ since S is an optimal solution for H. Thus, it remains to show $|\tilde{S}| \leq |S_2^*|$. Since the vertices of W_{int} are not incident to any edge modifications in \hat{S} , we may prove this inequality by showing, for each vertex $u \in W_{ext}$, that

$$|\{e \in \tilde{S} \mid u \in e\}| \le |\{e \in S_2^* \mid u \in e\}|.$$

If u is good, then the number of edge modifications incident with u in \tilde{S} is the same as the number of edge modifications incident with \tilde{u} in S_2^* , because u and \tilde{u} have the

same neighborhood in X by the condition of the rule. By the choice of \tilde{u} , this is at most as large as the number of edge modifications incident with u in S_2^* and the claim holds.

Otherwise, all edges between *u* and *X* are deleted by \tilde{S} . Let *K* denote the cluster in $G \triangle S^*$ containing \tilde{u} . Since *u* is not good, there is a vertex *w* that is not in [u] such that *w* has at least as many neighbors in $K \cap X$ as \tilde{u} . By the condition of Reduction Rule 6.2 and since $[\tilde{u}] \neq [w]$, the neighborhoods of these two vertices in *X* are disjoint, that is, $(N_G(\tilde{u}) \cap X) \cap (N_G(w) \cap X) = \emptyset$. Since *w* has at least as many neighbors in $K \cap X$ as \tilde{u} , this means that

$$|N_G(\tilde{u}) \cap K \cap X| \le |K \cap X|/2.$$

Now, observe that S_2^* contains an edge insertion between \tilde{u} and each nonneighbor of \tilde{u} in $K \cap X$. Thus, at least $|K \cap X|/2$ edges between \tilde{u} and $K \cap X$ are inserted by S_2^* . Moreover, S_2^* contains an edge deletion between \tilde{u} and each neighbor of \tilde{u} in $X \setminus K$. Altogether, this implies

$$|\{e \in S_2^* : \tilde{u} \in e\}| \ge |K \cap X|/2 + |N_G(\tilde{u}) \cap (X \setminus K)| \ge |N_G(\tilde{u}) \cap X|.$$

Therefore, the number of edge modifications incident with u in \tilde{S} (this is exactly the number of edges between u and X) is at most as large as the number of edge modifications incident with \tilde{u} in S_2^* . By the choice of \tilde{u} , this implies the claim.

It remains to analyze the running time for applying Reduction Rule 6.2.

Lemma 6.4 Let $\Gamma(G, k)$ be the running time used for finding an optimal solution of size at most k in a graph G if it exists.

Then, in $O(m + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time, we can apply Reduction Rule 6.2 to all graphs in \mathcal{H} .

Here, we assume that Γ is monotonically nondecreasing in *k* and polynomial in the size of *G*. Currently, $O(1.62^k + |V(G)| + |E(G)|)$ is the best known bound for $\Gamma(G, k)$ [7].

Proof We first show that the rule can be applied in $O(|W| + \sum_{w \in W} \deg_G(w) + \Gamma(H, t))$ time to an arbitrary graph $H = (W, F) \in \mathcal{H}$. For convenience, denote $X := V \setminus W$.

First, observe that a necessary condition for the rule is that, for each pair of vertices *u* and *v* in *H*, their neighborhoods in *X* are the same or disjoint. This can be checked in $O(|W| + \sum_{w \in W} \deg_G(w))$ time as follows. First, build the bipartite graph with parts *W* and *N*(*W*) and those edges between the vertex sets that are also edges of *G* (equivalently, $G[W \cup N(W)]$ minus the edges with both endpoints in *W* or both endpoints in *N*(*W*)). This bipartite graph is a disjoint union of complete bipartite graphs if and only if above condition is fulfilled. Thus, we check in O(|W| + |F|) time, whether the graph is a disjoint union of complete bipartite graphs. If not, then the rule does not apply. Otherwise, in the created bipartite graph, we compute in O(|W| + |F|) time the groups of vertices of *W* whose neighborhood is the same. Afterwards, we can check in O(1) time whether *u* and *v* have the same neighborhood in *X* in *G* by checking whether they belong to the same group. We now compute the set *S'* of edge modifications that is already determined by the conditions of the rule: If *v* and *w* are nonadjacent and have the same nonempty neighborhood in *X*, then the edge {*v*, *w*} needs to be

inserted and is thus added to S'. Similarly, if v and w are adjacent and have different neighborhoods in X, then $\{v, w\}$ needs to be deleted and is thus added to S'. Observe that if S' is a subset of an optimal solution S for H, then $|S| \le |F|$. Hence, at most |F|edges are added to S'. Since we already computed the groups of vertices that have the same or disjoint neighborhoods, we can thus compute S' in O(|W| + |F|) time.

Let W_{ext} denote the vertices of W that have at least one neighbor in N(W) and let $W_{\text{int}} := W \setminus W_{\text{ext}}$. Note that after applying S', that is, in $H \triangle S'$, the vertices of W_{ext} are in separate clusters: $(H \triangle S')[W_{\text{ext}}]$ is a cluster graph and there are no edges between W_{ext} and W_{int} in $H \triangle S'$. Thus, to determine whether S' can be extended to an optimal solution S for H that fulfills the condition of the rule, we compute an optimal solution of $H[W_{\text{int}}]$ in $\Gamma(H[W_{\text{int}}], \tau(H) - |S'|)$ time. Since $|H[W_{\text{int}}]| \le |H|$ and $\tau(H) - |S'| < t$, this can be done in $\Gamma(H, t)$ time. The size of the resulting solution S is compared with the size of an optimal solution of H, which can also be computed in $\Gamma(H, t)$ time.

It remains to show that Reduction Rule 6.2 can be applied to all graphs within the claimed running time. For each graph $H = (W, F) \in \mathcal{H}$, we can check in $O(|W| + \sum_{w \in W} \deg_G(w) + \Gamma(H, t))$ time whether Reduction Rule 6.2 applies. If yes, then we can apply the rule in O(|F|) time by modifying at most |F| edges. Summing up over all graphs in \mathcal{H} gives the claimed running time.

Observe that since *k* is decreased by $\tau(H)$, the parameter ℓ does not increase when Reduction Rule 6.2 is applied. As for the previous problems, applying the rule to each $H \in \mathcal{H}$ is sufficient for bounding *k* in *t* and ℓ and thus, for transferring known fixed-parameter tractability results for the parameter *k* to the combined parameter (*t*, ℓ).

Lemma 6.5 Let (G, \mathcal{H}, k) be a yes-instance of Cluster Editing with Cost-*t* Packing such that Reduction Rule 6.2 does not apply to any $H \in \mathcal{H}$. Then, $k \leq (2t + 1)\ell$.

Proof Since the instance is reduced with respect to Reduction Rule 6.2, for each H = (W, F) in \mathcal{H} and each size- $\tau(H)$ solution S for H, either the vertices of some cluster have different neighborhoods in $V \setminus W$ or two vertices of two distinct clusters have a common neighbor outside of W.

Now, fix an arbitrary optimal solution *S* for *G*. By the observation above, there are the following two possibilities for how *S* modifies each $H \in \mathcal{H}$:

- (a) more than $\tau(H)$ vertex pairs of H are modified by S, or
- (b) at least one external vertex pair for H is modified.

Therefore, *S* fulfills the condition of Lemma 3.2 and thus $k \le (2t + 1)\ell$.

Theorem 6.6 Let $\Gamma(G, k)$ be the running time used for finding an optimal solution of size at most k in a graph G if it exists. Then, CLUSTER EDITING WITH COST-t PACKING

- (i) is solvable in $O(\Gamma(G, (2t+1)\ell) + nm + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time and
- (ii) admits a problem kernel with at most $(4t + 2)\ell$ vertices, which can be computed in $O(nm + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time.

Proof (ii) First, apply Reduction Rule 6.2 exhaustively in $O(m + \sum_{H \in \mathcal{H}} \Gamma(H, t))$ time. Then, by Lemma 6.5, we can either return "no" or have $k \leq (2t + 1)\ell$. In the latter

case, we apply a kernelization algorithm for CLUSTER EDITING to the instance (G, k) (that is, without \mathcal{H}), which produces, in O(nm) time, a problem kernel (G', k') with at most $2k \leq (4t + 2)\ell$ vertices and with $k' \leq k$ [12]. Adding an empty packing gives an equivalent instance (G', \emptyset, k') with parameter $\ell' = k'$ of CLUSTER EDITING WITH COST-*t* PACKING.

(i) First, apply the kernelization. Then, by Lemma 6.5, we can either return "no" or have $k \le (2t + 1)\ell$. We can now apply the algorithm for CLUSTER EDITING that runs in $\Gamma(G, (2t + 1)\ell)$ time.

By plugging in the best known bound for $\Gamma(G, k)$, we obtain the following.

Corollary 6.7 Cluster Editing with Cost-t Packing

- (i) can be solved in $O(1.62^{(2t+1)\cdot \ell} + nm + n \cdot 1.62^t)$ time, and
- (ii) admits a problem kernel with at most $(4t + 2)\ell$ vertices that can be computed in $O(nm + n \cdot 1.62^t)$ time.

6.2 A Search Tree Algorithm for P_3 -Packings

For CLUSTER EDITING WITH P_3 -PACKING, the generic algorithm based on Reduction Rule 6.2 (with t = 1) using the currently best running time for CLUSTER EDITING leads to a running time of $O(4.26^{\ell} + n + m)$. We now show an algorithm that runs in $O(4^{\ell} \cdot \ell^3 + n + m)$ time. The algorithm is based on two special cases of Reduction Rule 6.2, one further reduction rule and a corresponding branching algorithm.

The analysis of the search tree size is done by considering branching vectors and their corresponding branching number. The components of a branching vector denote the decrease of the parameter in each recursive branch. The branching number depends only on the branching vector, the largest branching number gives the base in the upper bound on the search tree size; for further details refer to the relevant monographs [19, 39].

We use the following special cases of Reduction Rule 6.2. In both cases, H is a P_3 ; the correctness is directly implied by Lemma 6.4. In the first rule, adding an edge is a solution which fulfills the condition of Reduction Rule 6.2. For convenience, we denote by uvw a P_3 on the vertices u, v, and w, where v is the degree-two vertex.

Reduction Rule 6.8 If G contains a P_3 uvw such that $N(u) \setminus \{u, v, w\} = N(v) \setminus \{u, v, w\} = N(w) \setminus \{u, v, w\}$, then insert $\{u, w\}$ and decrease k by one.

In the second rule, deleting an edge gives such a solution.

Reduction Rule 6.9 If *G* contains a P_3 *uvw* such that $N(u) \setminus \{u, v, w\} = N(v) \setminus \{u, v, w\}$ and $N(u) \cap N(w) = \{v\}$, then delete $\{v, w\}$ and decrease *k* by one.

The third rule is crucial for showing an improved running time.

Reduction Rule 6.10 If G contains a clique K on at least three vertices such that

- every vertex in K has at most one neighbor in N(K) and

- every vertex in N(K) has exactly one neighbor in K,

then delete all edges between K and N(K) and decrement k by q := |N(K)|.

Lemma 6.11 Reduction Rule 6.10 is correct and can be exhaustively applied in O(nm) time.

Proof First, enumerate the set of all maximal cliques K on at least three vertices such that every vertex of K has at most one neighbor in N(K). This can be done in O(m) time [29]. Note that every vertex is contained in at most one such clique. Now, by scanning through the adjacency lists of all vertices in an enumerated clique, we can identify a vertex that has more than one neighbor in the clique. If there is such a vertex, then the clique can be discarded. Otherwise, the clique fulfills the conditions of the rule and the rule can be applied. Thus, one application of the rule takes O(m) time. Since the rule decreases the number of vertices in G, it can be applied O(n) times.

Since $q \le |K|$ and $|K| \ge 3$, one can construct $q P_3$ s, each containing two vertices from K and one vertex from N(K), such that no two of them share more than one vertex. Thus, at least q edge modifications are needed to destroy all P_3 s that contain at least one vertex $v \in K$. Deleting all q edges between K and N(K) destroys all P_3 s that contain at least one vertex of K. Moreover, since these edge deletions cut Kfrom the rest of the graph, one can safely combine any optimal solution for $G[V \setminus K]$ with these q edge deletions, which are necessary and sufficient to destroy all P_3 s that contain at least one vertex of K, to obtain an optimal solution that deletes all q edges between K and N(K).

The final rule simply removes isolated clusters from G.

Reduction Rule 6.12 If *G* contains a connected component *K* that is a clique, then remove *K* from *G*.

We can now show our improved algorithm for CLUSTER EDITING WITH P_3 -PACKING.

Theorem 6.13 Cluster Editing with P_3 -Packing *can be solved in O*($4^{\ell} \cdot \ell^3 + m + n$) *time*.

Proof We prove the theorem using a branching algorithm, which applies several branching rules. Herein, we assume that \mathcal{H} contains at least one P_3 *uvw*. Otherwise, the graph is either P_3 -free (in this case we are done) or we can add a P_3 to \mathcal{H} , which increases $|\mathcal{H}|$ by one and thus reduces the parameter. Furthermore, we assume that Reduction Rules 6.8 to 6.10 and 6.12 do not apply. First, we take care of P_3 s that do not share an edge with a packing P_3 .

Branching Rule 1: If there is an induced P_3 that contains at most one vertex of each P_3 in \mathcal{H} , then branch into the three cases to destroy this P_3 .

None of the cases destroys a P_3 of \mathcal{H} . Thus, the parameter is decreased by one in each case; the branching number is 3.

The three further rules deal with packing P_{3} s *uvw*.

Branching Rule 2: If there is a vertex x that is adjacent to u and w but not to v, then branch into four cases: delete $\{u, x\}$; delete $\{w, x\}$; add $\{v, x\}$; or delete $\{u, v\}$ and $\{v, w\}$ and add $\{u, w\}$.

In each of the first three branches, k is reduced by one without destroying any P_3 of \mathcal{H} . If none of the first three cases applies, then u, w, and x are in the same cluster

(no edge deletions between *u* or *w* and *x*) and *v* is not in this cluster. This makes the three edge modifications in $G[\{u, v, w\}]$ necessary. Thus, *k* is reduced by three and $|\mathcal{H}|$ is reduced by one in this case. The resulting branching vector is (1, 1, 1, 2), which gives the branching number 3.31.

Branching Rule 3: If there are vertices x and y such that

- -x is adjacent to u and v, and
- y is adjacent to exactly one vertex of $\{u, v\}$,

then branch into four cases: delete $\{u, x\}$; delete $\{v, x\}$; delete the edge between y and its neighbor in $\{u, v\}$; or add the edge between y and its nonneighbor in $\{u, v\}$.

In each case, an edge is modified without destroying any P_3 of \mathcal{H} . If none of the first three cases applies, then u, v, x, and y are in the same cluster, which means that the missing edge between y and either u or v has to be added. Since the parameter is reduced by one in each branch, the branching number is 4.

Branching Rule 4: If there is a vertex x that is adjacent to v and not adjacent to u and w, then branch into four cases: delete $\{v, x\}$; add $\{u, x\}$; add $\{w, x\}$; or delete $\{u, v\}$ and $\{u, w\}$.

If none of the first three cases applies, then any cluster containing v contains neither u nor w, thus the branching is correct. The parameter is reduced by one in each branch, as the last branch destroys a P_3 of \mathcal{H} but reduces k by two. Thus, the branching number is 4.

These are the only branching rules that are performed. We now show, by a case distinction, the following: If none of the branching rules and reduction rules applies, then the remaining graph has maximum degree two and we can solve the problem in polynomial time.

Case I: \mathcal{H} contains a P_3 uvw such that u and w have a common neighbor $x \neq v$. Since Branching Rule 2 does not apply, x is also a neighbor of v. Since Branching Rule 3 does not apply, we have that every other vertex y that is adjacent to u is also adjacent to v and vice versa. Similarly, every other vertex y that is adjacent to w is also adjacent to v. Thus, u, v, and w have the same neighbors in $V \setminus \{u, v, w\}$. This contradicts our assumption that Reduction Rule 6.8 does not apply.

Case II: \mathcal{H} *contains a* P_3 *uvw such that* v *has degree at least three.* Since Branching Rule 4 does not apply, each vertex $x \in N(v) \setminus \{u, w\}$ is a neighbor of u or w and, since Case I does not apply, it is not a neighbor of both. Moreover, since Branching Rule 3 does not apply, v can have common neighbors with at most one of u and w. Thus, without loss of generality, u and v have the same neighborhood in $V \setminus \{u, v, w\}$ and v and w have no common neighbors. This contradicts our assumption that Reduction Rule 6.9 does not apply.

Case III: \mathcal{H} *contains a* P_3 *uvw such that u has degree at least three.* Consider p and q from $N(u) \setminus \{v\}$. Since Case II does not apply, p and q are not middle vertices of a P_3 in \mathcal{H} . Moreover, since Case I does not apply, p and q are not from the same P_3 of \mathcal{H} . Consequently, if $G[\{u, p, q\}]$ is a P_3 , then Branching Rule 1 applies. This implies that $G[N[u] \setminus \{v\}]$ is a clique K of size at least three. We now show the following claim, which contradicts our assumption that Reduction Rule 6.10 does not apply.

Claim: Each vertex of *K* has at most one neighbor in $V \setminus K$ and every vertex in N(K) has at most one neighbor in *K*.

First, observe that no vertices from *K* are middle vertices of a P_3 in \mathcal{H} . Thus, *K* contains at most one vertex *x* from each P_3 of \mathcal{H} and this vertex *x* is not a middle vertex. For each such $x \in K$ from a packing P_3 *xyz*, the same conditions apply as to *u*, thus $G[N[x] \setminus \{y\}]$ is a clique *K'*. This implies K = K': Since $G[N[u] \setminus \{v\}]$ is a clique, *x* is adjacent to every vertex in $N[u] \setminus \{v\}$ and since $G[N[x] \setminus \{y\}]$ is a clique, *u* is adjacent to every vertex in $N[x] \setminus \{y\}$. Summarizing, each vertex from *K* that is in a P_3 of \mathcal{H} has exactly one neighbor outside of *K*, this neighbor is a middle vertex of the packing P_3 containing *x*. These middle vertices have only one neighbor in *K* since they have degree two and *K* contains only one vertex from each packing P_3 .

Now let *x* denote a vertex of *K* that is not contained in any P_3 of \mathcal{H} . We show that N[x] = K, which implies the claim. Since the middle vertices of each P_3 of \mathcal{H} have no neighbors outside of this P_3 and since Branching Rule 2 does not apply, we have that *x* is adjacent to at most one vertex of each packing P_3 . Thus, G[N[x]] is a clique *K'* as otherwise, Branching Rule 1 applies. Again, K' = K as $G[N[u] \setminus \{v\}]$ being a clique implies that *x* is adjacent to every vertex in $N[u] \setminus \{v\}$ and G[N[x]] being a clique implies that *u* is adjacent to every vertex in N[x].

Case IV: otherwise. We show that every vertex has degree at most two. This is true for the middle vertices of P_{3} s in \mathcal{H} as Case II does not apply. This also holds for the endpoints of P_{3} s in \mathcal{H} as Case III does not apply. We now argue that every other vertex *x* cannot have two neighbors.

The vertex x has at least one neighbor u from some P_3 uvw of \mathcal{H} : x is contained in at least one P_3 because the instance is reduced with respect to Reduction Rule 6.12 and this P_3 contains at least two vertices of some P_3 of the packing because Branching Rule 1 does not apply. If x has a further neighbor y, then $y \neq v$ (since Case II does not apply) and $v \neq w$ since (Case I) does not apply. Consequently, u, x, and y form a triangle (since Branching Rule 1 does not apply). Thus, u has two neighbors outside of his P_3 of \mathcal{H} , which means that Case III applies.

Thus, G has maximum degree two and does not contain isolated triangles because the instance is reduced with respect to Reduction Rule 6.12. In this case, an optimal solution can be obtained by computing a maximum matching M and then deleting all edges of G that are not in M.

Altogether, the above considerations imply a search tree algorithm with search tree size $O(4^{\ell})$. After an initial kernelization, which, due to Corollary 6.7, runs in O(m + n) time for t = 1, the instance has $O(\ell)$ vertices. Thus, the steps at each search tree node including the reduction rules can be performed in $O(\ell^3)$ time.

7 Hardness Results for Edge Deletion and Vertex Deletion Problems

In this section, we show edge modification problems and vertex deletion problems that are NP-hard even for small forbidden induced subgraphs and if $\ell = k - |\mathcal{H}|$, where \mathcal{H} is a vertex-disjoint packing of forbidden induced subgraphs. We also show that algorithms for VERTEX COVER parameterized above lower bounds do not generalize to *d*-HITTING SET—the natural generalization of VERTEX COVER to hypergraphs.

7.1 Hard edge deletion problems

Theorem 7.1 For every fixed $q \ge 6$, K_q -free Deletion with K_q -Packing is NP-hard for $\ell = 0$.

We prove Theorem 7.1 by giving a reduction from 3-SAT (Problem 4.10).

Construction 7.2 Let ϕ be a Boolean formula with variables x_1, \ldots, x_n and clauses C_1, \ldots, C_m . We assume that each clause C_j contains exactly three pairwise distinct variables. We create a graph *G* and a vertex-disjoint K_q -packing \mathcal{H} as follows.

For each variable x_i , add a clique X_i on q vertices to G that has two distinguished disjoint edges x_i^{F} and x_i^{T} . For each clause $C_j = (l_1 \land l_2 \land l_3)$ with literals l_1, l_2 , and l_3 , add a clique Y_j on q vertices to G that has three distinguished and pairwise disjoint edges e_{l_1}, e_{l_2} , and e_{l_3} (which exist since $q \ge 6$). Finally, if $l_t = x_i$, then identify the edge e_{l_i} with x_i^{T} and if $l_t = \neg x_i$, then identify the edge e_{l_t} with x_i^{F} . The packing \mathcal{H} consists of all X_i introduced for the variables x_i of ϕ .

Lemma 7.3 Let *G* be the graph output by Construction 7.2 and let *H* be an induced K_q in *G*. Then, *H* is either one of the X_i or one of the Y_j .

Proof First, note that the X_i are pairwise vertex-disjoint since Construction 7.2 only identifies edges of Y_j s with edges of X_i s and no edge in any Y_j is identified with edges in different X_i . For any X_i and Y_j , the vertices in $V(X_i) \setminus V(Y_j)$ are nonadjacent to those in $V(Y_j) \setminus V(X_i)$. Similarly, for Y_i and Y_j , the vertices in $V(Y_i) \setminus V(Y_j)$ are nonadjacent to those in $V(Y_j) \setminus V(Y_i)$ for $i \neq j$. Thus, every clique in *G* is entirely contained in one of the X_i or Y_j .

Lemma 7.3 allows us to prove Theorem 7.1.

Proof (of Theorem 7.1) We show that ϕ is satisfiable if and only if *G* can be made K_q -free by $k = |\mathcal{H}|$ edge deletions (that is, $\ell = 0$).

First, assume that there is an assignment that satisfies ϕ . We construct a K_q -free deletion set *S* for *G* as follows: if the variable x_i is set to true, then put x_i^{T} into *S*. If the variable x_i is set to false, then add x_i^{F} to *S*. Thus, for each X_i , we add exactly one edge to *S*. Since \mathcal{H} consists of the X_i , we have $|S| = |\mathcal{H}|$. Moreover, since each clause C_j contains a true literal, at least one edge of each Y_j is contained in *S*. Thus, $G \setminus S$ is K_q -free, since, by Lemma 7.3, the only K_q s in *G* are the X_i and Y_j and, for each of them, *S* contains at least one edge.

Now, assume that *G* can be made K_q -free by deleting a set *S* of $|\mathcal{H}|$ edges. Then, *S* deletes exactly one edge of each X_i and at least one edge of each Y_j . We can assume without loss of generality that *S* contains either the edge x_i^T or x_i^F for each X_i since deleting one of these edges instead of another edge in X_i always yields a solution by Construction 7.2. Thus, the deletion set *S* corresponds to a satisfying assignment for ϕ .

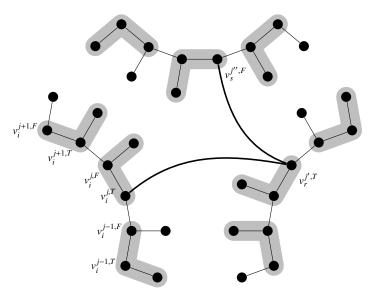


Fig. 7.1 An illustration of Construction 7.6 for q = 3. The figure shows parts of the variable cycles for three variables x_i, x_r, x_s that occur in the clause $C_t = (x_i \lor x_r \lor \neg x_s)$. The packing P_{3s} are highlighted by a gray background.

7.2 Hard vertex deletion problems

In this section, we show NP-hardness of the problem of destroying all induced paths P_q on $q \ge 3$ vertices by at most $|\mathcal{H}|$ vertex deletions if a packing \mathcal{H} vertex-disjoint induced P_q s in the input graph G is provided as input.

Problem 7.4 (P_q -free Vertex Deletion with P_q -Packing)

- *Input:* A graph G = (V, E), a vertex-disjoint packing \mathcal{H} of induced P_q s, and a natural number k.
- *Question:* Is there a vertex set $S \subseteq V$ of size at most k such that $G[V \setminus S]$ does not contain P_q as induced subgraph?

Theorem 7.5 For every fixed $q \ge 3$, P_q -FREE VERTEX DELETION WITH P_q -PACKING is NP-hard even if $\ell = 0$.

The reduction is from *q*-SAT:

Construction 7.6 Let ϕ be a Boolean formula with variables x_1, \ldots, x_n and clauses C_1, \ldots, C_m . We assume that each clause C_j contains exactly q pairwise distinct variables. We construct a graph G and a maximal vertex-disjoint packing \mathcal{H} of P_q s as follows; an illustration of the construction is given in Figure 7.1.

First, we introduce variable gadgets, which will ensure that a solution to P_q -FREE VERTEX DELETION WITH P_q -PACKING corresponds to an assignment of ϕ . In the following, let $\alpha(i)$ denote the number of occurrences of x_i and $\neg x_i$ in clauses of ϕ . For each variable x_i , add $4\alpha(i)$ vertices: $v_i^{j,T}$ and $v_i^{j,F}$, where $1 \le j \le 2\alpha(i)$. Call $v_i^{j,T}$ a *true vertex*



and $v_i^{j,F}$ a *false vertex*. Create an induced cycle on the true and false vertices by adding the edge set

$$E_i := \left\{ \{v_i^{j,\mathrm{T}}, v_i^{j,\mathrm{F}}\} \mid 1 \le j \le 2\alpha(i) \right\} \cup \left\{ \{v_i^{j,\mathrm{F}}, v_i^{j+1,\mathrm{T}}\} \mid 1 \le j < 2\alpha(i) \right\} \cup \left\{ \{v_i^{2\alpha(i),\mathrm{F}}, v_i^{1,\mathrm{T}}\} \right\}$$

Call this cycle the *variable cycle* of x_i . Then, for each even j, attach to $v_i^{j,T}$ an induced P_{q-2} , that is, make one of its degree-one vertices adjacent to $v_i^{j,T}$. Then, again for each even j, attach to $v_i^{j,F}$ an induced P_{a-2} in the same fashion. These paths are called the *attachment paths* of the *j*th segment of the variable cycle of x_i .

Now, for each variable x_i , assign to each clause C_t containing x_i or $\neg x_i$ a unique number $p \in \{1, ..., \alpha(i)\}$. Consider the number j = 2p - 1. We will use vertex $v_i^{j,T}$ or $v_i^{j,F}$ to build the clause gadget for clause C_t . If C_t contains the literal x_i , then attach an induced P_{q-2} to $v_i^{j,F}$. Otherwise, attach an induced P_{q-2} to $v_i^{j,T}$. As above, call the path the *attachment path* of the *j*th segment of the cycle. Now, let $\gamma_i^t := v_i^{j,T}$ if C_t contains x_i , and let $\gamma_i^t := v_i^{i,F}$ if C_t contains $\neg x_i$. Call these vertices the *literal* vertices of clause C_t , denoted Π_t . The construction of *G* is completed as follows. For each Π_t add an arbitrary set of edges to G such that $G[\Pi_t]$ is an induced P_q . The P_q -packing \mathcal{H} contains one (arbitrary) attachment path plus the two segment vertices from each segment of each variable cycle.

Proof (of Theorem 7.5) Let G be the graph output by Construction 7.6 and let \mathcal{H} be the P_q -packing. We show that ϕ has a satisfying assignment if and only if G can be made P_q -free by exactly $|\mathcal{H}|$ vertex deletions (that is, $\ell = 0$).

Assume that ϕ has a satisfying assignment. For each true variable x_i in this assignment, delete all true vertices in its variable gadget, that is, $v_i^{j,T}$ for $1 \le j \le 2\alpha(i)$. For each false variable x_i in this assignment, delete all false vertices in its variable gadget, that is, $v_i^{j,F}$ for $1 \le j \le 2\alpha(i)$. Denote this vertex set by S and observe that $|S| = |\mathcal{H}|$. Moreover, observe that each vertex on the variable cycle for x_i is either deleted or both of its neighbors on the cycle are deleted. Every P_a in G contains at least one vertex from a variable cycle as the attachment paths are too short to induce P_a s. Thus, to show P_q -freeness of $G[V \setminus S]$ it is sufficient to show that no vertex from a variable cycle is in a P_q .

Consider an undeleted vertex in the variable cycle for x_i . Assume, without loss of generality, that this is a true vertex $v_i^{j,T}$. If *j* is even, then $v_i^{j,T}$ is not in a P_q as its neighbors on the cycle are deleted and its only other neighbor is in an attachment path. If j is odd and the clause C_t corresponding to the jth segment of the cycle contains $\neg x_i$, then the only neighbor of $v_i^{j,T}$ in $G[V \setminus S]$ is in an attachment path. It remains to show that $v_i^{j,T}$ is not in a P_q if C_t contains x_i . The only neighbors of $v_i^{j,T}$ in $G[V \setminus S]$ are in Π_t . Observe that $G[\Pi_t]$ is an induced P_q and that, in $G[V \setminus S]$, every vertex on this path is deleted or its neighbors in $V \setminus \Pi_t$ are deleted. Hence, the connected component of $G[V \setminus S]$ containing $v_i^{j,T}$ is an induced subgraph of $G[\Pi_t]$. Since the assignment is satisfying, at least one vertex of Π_t is deleted. Thus, this connected component has at most q - 1 vertices and does not contain a P_q .

Conversely, let $S \subseteq V$ be a size- $|\mathcal{H}|$ vertex set such that $G[V \setminus S]$ is P_q -free. First, observe that, without loss of generality, for each variable cycle either all true or all

false vertices are deleted: No vertex in an attachment path P is deleted since it is always as good to delete the vertex in the variable cycle that has a neighbor in P. Hence, at least one vertex of each segment is deleted since, otherwise, one of the P_q 's in \mathcal{H} is not destroyed. This already requires $|\mathcal{H}|$ vertex deletions and thus *exactly* one vertex for each segment of each variable cycle is deleted. Finally, by construction, every adjacent pair of vertices in the variable cycle forms a P_q with some attachment path. Therefore, one of the two vertices is deleted, which implies that either every even or every odd vertex of the cycle is deleted.

Hence, the vertex deletions in the variable cycle define a truth assignment β to x_1, \ldots, x_n : If all true vertices of the variable cycle of x_i are deleted, then set $\beta(x_i) :=$ true; otherwise, set $\beta(x_i) :=$ false. This assignment is satisfying: Since $G[V \setminus S]$ is P_q -free, for each clause C_t , at least one vertex γ_i^t of Π_t is deleted. Without loss of generality, let $\gamma_i^t = v_i^{j,T}$, that is, C_t contains the literal x_i . Then, $\beta(x_i) =$ true and thus β satisfies clause C_t .

Theorem 7.5 easily transfers to a hardness result for the generalization of VERTEX COVER to *d*-uniform hypergraphs:

Problem 7.7 (d-UNIFORM HITTING SET WITH PACKING)

Input: A hypergraph H = (V, E) with |e| = d for all $e \in E$, a set $\mathcal{H} \subseteq E$ of pairwise vertex-disjoint hyperedges, and an integer k.

Question: Is there a vertex set $V' \subseteq V$ of size at most k such that $\forall e \in E : V' \cap e \neq \emptyset$?

An instance (G, \mathcal{H}, k) of P_q -FREE VERTEX DELETION WITH P_q -PACKING can easily be transformed into an equivalent instance (H, \mathcal{H}, k) of q-UNIFORM HITTING SET WITH PACKING by taking the hypergraph H on the same vertex set as G having a hyperedge e if and only if G[e] is a P_q . The packing \mathcal{H} and k stay unchanged, and so does ℓ . Thus, we obtain the following result:

Corollary 7.8 For every $d \ge 3$, *d*-UNIFORM HITTING SET WITH PACKING is NP-hard even if $\ell = 0$.

Corollary 7.8 shows that the known above-guarantee fixed-parameter algorithms for VERTEX COVER [13, 22, 36, 43] do not generalize to *d*-UNIFORM HITTING SET.

8 Conclusion

It is open to extend our framework to further problems. The most natural candidates appear to be problems where the forbidden induced subgraph has four vertices. Examples are COGRAPH EDITING [35] which is the problem of destroying all induced P_{4} s, K_{4} -FREE EDITING, CLAW-FREE EDITING, and DIAMOND-FREE DELETION [18, 44]. Another direction could be to investigate edge completion problems that allow for subexponential-time algorithms [15]. In the case of vertex-deletion problems, TRIANGLE VERTEX DELETION appears to be the most natural open case. Furthermore, it would be nice to obtain more general theorems separating the tractable from the hard cases for this parameterization. For CLUSTER EDITING and TRIANGLE DELETION improved running times are desirable.

Maybe more importantly, it is open to determine the complexity of CLUSTER EDITING and FEEDBACK ARC SET IN TOURNAMENTS parameterized above the size of *edge-disjoint* packings of forbidden induced subgraphs. Finally, our framework offers an interesting tradeoff between running time and power of generic data reduction rules. Exploring such tradeoffs seems to be a rewarding topic for the future. The generic rules presented in this work can be easily implemented, which asks for subsequent experiments to evaluate their effectiveness.

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