

On Feedback Vertex Set: New Measure and New Structures*

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Abstract

We present a new parameterized algorithm for the feedback vertex set problem (FVS) on undirected graphs. We approach the problem by considering a variation of it, the disjoint feedback vertex set problem (DISJOINT-FVS), which finds a feedback vertex set of size k that has no overlap with a given feedback vertex set F of the graph G . We develop an improved kernelization algorithm for DISJOINT-FVS and show that DISJOINT-FVS can be solved in polynomial time when all vertices in $G \setminus F$ have degrees upper bounded by three. We then propose a new branch-and-search process on DISJOINT-FVS, and introduce a new branch-and-search measure. The process effectively reduces a given graph to a graph on which DISJOINT-FVS becomes polynomial-time solvable, and the new measure more accurately evaluates the efficiency of the process. These algorithmic and combinatorial studies enable us to develop an $O^*(3.83^k)$ -time parameterized algorithm for the general FVS problem, improving all previous algorithms for the problem.

1 Introduction

All graphs in our discussion are undirected and simple, i.e., they contain neither self-loops nor multiple edges. A *feedback vertex set* (FVS) F in a graph G is a set of vertices in G whose removal results in an acyclic graph. The problem of finding a minimum feedback vertex set in a graph is one of the classical NP-complete problems [17]. It has been intensively studied for several decades. The problem is known to be solvable in time $O(1.7548^n)$ for a graph of n vertices [14], and admit a polynomial-time approximation algorithm of ratio 2 [1, 3].

An important application of the feedback vertex set problem is Bayesian inference in artificial intelligence [2, 3], where the *size* k of a minimum FVS F (i.e., the number of vertices in F) of a graph can be expected to be fairly small. This motivated the study of the parameterized version of the problem, which we will name FVS: given a graph G and a parameter k , either construct a FVS of size bounded by k in G or report no such a FVS exists. Parameterized algorithms for FVS have been extensively studied that find a FVS of size k in a graph of n vertices in time $f(k)n^{O(1)}$ for a fixed function f (thus, the algorithms become practically efficient when the value k is small). The existence of such an algorithm for FVS is implied in [13]. The first group of constructive algorithms for this problem was given by Downey and Fellows [10] and by Bodlaender [4]. Since then a chain of improvements has been obtained (see Figure 1).¹

All algorithms summarized in Figure 1 are *deterministic*. There is also an active research line on randomized parameterized algorithms for FVS, based on very different algorithmic techniques. A randomized algorithm of time $O^*(4^k)$ for FVS has been known for more than a decade [2]. More recently, Cygan et al. [7] developed an improved randomized algorithm of time $O^*(3^k)$. As pointed out in [7], however, the techniques employed by this randomized algorithm do not seem to be easily de-randomized.

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¹Following the recent convention in the research in exact and parameterized algorithms, we will denote by $O^*(f(k))$ the complexity $O(f(k)n^{O(1)})$ for a super-polynomial function f .

Authors	Complexity	Year
Downey and Fellows [10]	$O^*((2k+1)^k)$	1992
Bodlaender[4]	$O^*(17(k^4)!)$	1994
Raman et al.[23]	$O^*(\max\{12^k, (4\log k)^k\})$	2002
Kanj et al.[19]	$O^*((2\log k + 2\log\log k + 18)^k)$	2004
Raman et al.[24]	$O^*((12\log k / \log\log k + 6)^k)$	2006
Guo et al.[18]	$O^*(37.7^k)$	2006
Dehne et al.[8]	$O^*(10.6^k)$	2007
Chen et al.[6]	$O^*(5^k)$	2008
This paper	$O^*(3.83^k)$	

Figure 1: The history of parameterized algorithms for FVS.

The main result of the current paper is a deterministic algorithm of time $O^*(3.83^k)$ for FVS.

We give an outline to explain how our algorithm achieves the improvement over previous algorithms. As most recent algorithms, our algorithm is based on the technique of iterative compression [25], which reduces the FVS problem to a closely related DISJOINT FEEDBACK VERTEX SET problem (DISJOINT-FVS). On an instance (G, k, F) , where F is a FVS in the graph G and k is the parameter, the DISJOINT-FVS problem asks whether there is a FVS F' of size k in G such that $F' \cap F = \emptyset$.

The DISJOINT-FVS problem can be solved based on a branch-and-search process on vertices w in $G \setminus F$, whose complexity depends on the number of neighbors of w that are in F [6]. In particular, the more neighbors w has in F , the more effective the branching on w is. A major step of the fastest algorithm [6], before our algorithm, is to show that such a branch-and-search process can always branch on a vertex in $G \setminus F$ that has at least two neighbors in F . Therefore, in order to further speedup this process, we should branch only on vertices in $G \setminus F$ that have more than two neighbors in F . For this, however, two issues must be addressed: (1) during the branch-and-search process, we must be able to continuously maintain the condition that such vertices always exist; and (2) when the branch-and-search process cannot be further applied, we must be able to efficiently solve the problem for the remaining structure.

To address issue (2), we develop a polynomial-time algorithm for the DISJOINT-FVS problem for instances (G, k, F) in which all vertices in $G \setminus F$ have degree upper bounded by three. This algorithm is based on a nontrivial reduction from DISJOINT-FVS to a polynomial-time solvable matroid matching problem, the COGRAPHIC MATROID PARITY problem [22]. This result, however, does not give a direct solution to issue (1): vertices in $G \setminus F$ that have degree larger than three in G do not necessarily have more than two neighbors in F . To resolve this problem, we observe that there are always vertices in $G \setminus F$ on which a branching may not be very effective but will produce structures in $G \setminus F$ that are favored for the polynomial-time algorithm we developed for addressing issue (2). To catch this observation, we use the measure-based method and introduce a new measure to evaluate the effectiveness of our branch-and-search process more accurately. These new techniques, combined with the iterative compression method, yield an improved algorithm for the FVS problem.

The main results of this paper are summarized as follows: (i) a new technique that produces an improved kernelization algorithm for the DISJOINT-FVS problem, which is based on a branch-and-search algorithm for the problem. This, to our best knowledge, is the first time such a technique is used in the literature of kernelization; (ii) a polynomial-time algorithm that solves a restricted version of the DISJOINT-FVS problem; (iii) a new branch-and-search process that effectively reduces an input instance of DISJOINT-FVS to an instance that is solvable by the algorithm developed in (ii); and (iv) a new measure that more accurately evaluates the efficiency of the branch-and-search process in (iii).

2 DISJOINT-FVS and its kernel

We start with a formal definition of our first problem.

DISJOINT-FVS. Given a graph $G = (V, E)$, a FVS F in G , and a parameter k , either construct a FVS F' of size k in G such that $F' \cap F = \emptyset$, or report that no such a FVS F' exists.

The DISJOINT-FVS problem was motivated by the iterative compression method [25] that has become a standard framework for the development of parameterized algorithms for the FVS problem. In this method, a critical step is to construct a solution to an instance (G, F, k) of the DISJOINT-FVS problem in which the FVS F satisfies $|F| = k + 1$ (see, e.g., [6]). However, in the following discussion, we consider a slightly more generalized version in which we do not require $|F| = k + 1$.

Let $V_1 = V \setminus F$. Since F is a FVS, the subgraph induced by V_1 is a forest. Moreover, if the subgraph induced by F is not a forest, then it is impossible to have a FVS F' in G such that $F' \cap F = \emptyset$. Therefore, an instance of DISJOINT-FVS can be written as $(G; V_1, V_2; k)$, and consists of a partition (V_1, V_2) of the vertex set of the graph G and a parameter k such that both V_1 and V_2 induce forests (where $V_2 = F$). We will call a FVS entirely contained in V_1 a V_1 -FVS. Thus, the instance $(G; V_1, V_2; k)$ of DISJOINT-FVS is looking for a V_1 -FVS of size k in the graph G .

For a subgraph G' of G and a vertex v in G' , we will denote by $d_{G'}(v)$ the degree of the vertex v in G' . Thus, $d_G(v)$ is the degree of the vertex v in the original graph G , and $d_{G[V_1]}(v)$ for a vertex $v \in V_1$ is the degree of the vertex v in the induced subgraph $G[V_1]$.

Given an instance $(G; V_1, V_2; k)$ of DISJOINT-FVS, we apply the following two simple rules:

Rule 1. Remove all vertices v with $d_G(v) \leq 1$;

Rule 2. For a vertex v in V_1 with $d_G(v) = 2$,

- if the two neighbors of v are in the same component of $G[V_2]$, then include v into the objective V_1 -FVS, $G = G - v$, and $k = k - 1$;
- else either (2.1) move v from V_1 to V_2 : $V_1 = V_1 \setminus \{v\}$, $V_2 = V_2 \cup \{v\}$; or (2.2) smoothen v : replace v and the two incident edges with a new edge connecting the two neighbors of v .

Note that the second case in Rule 2 includes the cases where the two neighbors of v are both in V_1 , or both in V_2 , or one in V_1 and one in V_2 . In this case, we can pick any of the rules 2.1 and 2.2 and apply it.

The correctness of Rule 1 is trivial: no degree-0 or degree-1 vertices can be contained in any cycle. On the other hand, although Rule 2 is also easy to verify for the general FVS problem [6] (because any cycle containing a degree-2 vertex v must also contain the two neighbors of v), it is much less obvious for the DISJOINT-FVS problem – the two neighbors of the degree-2 vertex v may not be in V_1 and cannot be included in the objective V_1 -FVS. For this, we have the following lemmas.

Lemma 2.1 *For any degree-2 vertex v in V_1 whose two neighbors are not in the same component of $G[V_2]$, if G has a V_1 -FVS of size k , then G has a V_1 -FVS of size k that does not contain the vertex v .*

PROOF. Let F' be a V_1 -FVS of size k that contains v . If one neighbor u_1 of v is in V_1 , then the set $(F' \setminus \{v\}) \cup \{u_1\}$ will be a V_1 -FVS of size bounded by k that does not contain the vertex v . Thus, we can assume that the two neighbors u_1 and u_2 of v are in two different components in $G[V_2]$. Since $G - F'$ is acyclic, there is either no path or a unique path in $G - F'$ between u_1 and u_2 . If there is no path between u_1 and u_2 in $G - F'$, then adding v to $G - F'$ does not create any cycle. Therefore, in this case, the set $F' \setminus \{v\}$ is a V_1 -FVS of size $k - 1$ that does not contain v . If there is a unique path P between u_1 and u_2 in $G - F'$, then the path P must contain at least one vertex w in V_1 (since u_1 and u_2 are in different components in $G[V_2]$). Every cycle C in $G - (F' \setminus \{v\})$ must contain v , thus, also contain u_1 and u_2 . Therefore, the partial path $C \setminus v$ from u_1 to u_2 in C must be the unique path P between u_1 and u_2 in $G - F'$, which contains the vertex w . This shows that w must be contained in all cycles in $G - (F' \setminus \{v\})$. In consequence, the set $(F' \setminus \{v\}) \cup \{w\}$ is a V_1 -FVS of size bounded by k that does not contain v . \square

Lemma 2.2 *Rule 2 is safe. That is, suppose that Rule 2 applied on $(G; V_1, V_2; k)$ produces $(G'; V'_1, V'_2; k')$, then the graph G' has a V'_1 -FVS of size k' if and only if the graph G has a V_1 -FVS of size k .*

PROOF. If the two neighbors of the degree-2 vertex v are contained in the same component in $G[V_2]$, then v and some vertices in V_2 form a cycle. Therefore, in order to break this cycle, the vertex v must be contained in the objective V_1 -FVS. This justifies the first case for Rule 2.

If the two neighbors of the degree-2 vertex v are not in the same component in $G[V_2]$, then $(G'; V'_1, V'_2; k')$ is obtained by applying either Rule 2.1 or Rule 2.2 on $(G; V_1, V_2; k)$. By Lemma 2.1, the graph G has a V_1 -FVS of size k if and only if G has a V_1 -FVS F_1 of size k that does not contain the vertex v . Now it is easy to verify that no matter which of Rule 2.1 and Rule 2.2 is applied, we have $k' = k$, and the V_1 -FVS F_1 for G becomes a V'_1 -FVS of size k for the graph G' . This justifies the second case for Rule 2. \square

Note that the second case of Rule 2 cannot be applied *simultaneously* on more than one vertex in V_1 . For example, let v_1 and v_2 be two degree-2 vertices in V_1 that are both adjacent to two vertices u_1 and u_2 in V_2 . Then it is obvious that we cannot move both v_1 and v_2 to V_2 . In fact, if we first apply the second case of Rule 2 on v_1 , then the first case of Rule 2 will become applicable on the vertex v_2 .

Definition 1 *An instance $(G; V_1, V_2; k)$ of DISJOINT FVS is V_1 -irreducible if none of the Rules 1-2 can be applied on vertices in the set V_1 , or, equivalently, if all vertices in V_1 have degree larger than 2. An instance $(G; V_1, V_2; k)$ is nearly V_1 -irreducible if in the set V_1 there is at most one vertex of degree 2 and all other vertices in V_1 are of degree larger than 2.*

For an instance $(G; V_1, V_2; k)$ that is V_1 -irreducible or nearly V_1 -irreducible, in case there is no ambiguity, we will simply say that the graph G is V_1 -irreducible or nearly V_1 -irreducible, respectively. In the following, we show that a nearly V_1 -irreducible instance is necessarily small.

We start with a simple branch-and-search algorithm for nearly V_1 -irreducible instances of DISJOINT-FVS, as given in Figure 2, which is similar to the one presented in [6], but gives degree-2 vertices a higher priority when selecting a vertex for branching. The basic step of the algorithm is to pick a vertex v in V_1 and branch on either including or excluding v in the objective V_1 -FVS F . Note that in certain situations, the algorithm directly takes one of the two actions in the branching (see the footnotes in the algorithm).

Algorithm FindFVS

INPUT: a nearly V_1 -irreducible instance $(G; V_1, V_2; k)$ of DISJOINT-FVS.

OUTPUT: a V_1 -FVS F of size $\leq k$ in G , or report that no such V_1 -FVS exists.

1. $F = \emptyset$;
2. **while** $|V_1| > 0$ and $k \geq 0$ **do**
3. **if** there are vertices in V_1 that have degree 2 in G
 then let v be a vertex in V_1 that has degree 2 in G
 else let v be a vertex in V_1 that has degree ≤ 1 in the induced subgraph $G[V_1]$
4. **branching**
5. **case 1:** $\setminus \setminus$ v is in the objective V_1 -FVS F .
 add v to F and delete v from G ; $k = k - 1$;[†]
6. **case 2:** $\setminus \setminus$ v is not in the objective V_1 -FVS F .
 move v from V_1 to V_2 ;[‡]
7. **if** $|V_1| = 0$ **then** return F **else** return “no V_1 -FVS of size $\leq k$ ”.

[†] this action will not be taken if $d_G(v) = 2$ and the two neighbors of v are not in the same component of $G[V_2]$.

[‡] this action will not be taken if two neighbors of v are in the same component of $G[V_2]$.

Figure 2: A simple branch-and-search algorithm for DISJOINT-FVS

We will use algorithm FindFVS to count the number of vertices in the set V_1 . Note that Rules 1-2 are *not* applied during the process of the algorithm. Initially, the input graph is V_1 -irreducible. Thus, the selection of the vertex v in step 3 is always possible. In later steps, the selection of the vertex v in step 3 can be argued with the following lemma.

Lemma 2.3 *Each execution of steps 4-8 of algorithm FindFVS results in a nearly V_1 -irreducible instance.*

PROOF. Since the input instance is nearly V_1 -irreducible, it suffices to prove that on a nearly V_1 -irreducible instance, the execution of steps 4-8 of the algorithm produces a nearly V_1 -irreducible instance. Let $(G; V_1, V_2; k)$ be a nearly V_1 -irreducible instance of DISJOINT-FVS before the execution of steps 4-8 of the algorithm, and let v be the vertex in V_1 selected by steps 3 of the algorithm.

Steps 4-8 either deletes the vertex v from the graph (case 1, steps 5-6) or moves v from set V_1 to set V_2 (case 2, steps 7-8). Moving v from V_1 to V_2 does not change the degree of any vertex remaining in V_1 . Therefore, steps 7-8 keep the resulting instance nearly V_1 -irreducible.

Now consider steps 5-6 in the algorithm that delete the vertex v from the graph. If $d_G(v) = 2$ and the two neighbors of v are in the same component of $G[V_2]$, or if v has degree 0 in $G[V_1]$, then deleting v does not affect the degree of any vertex remaining in V_1 . Therefore, in these cases steps 5-6 in the algorithm produce a nearly V_1 -irreducible instance. Note that by the first footnote in the algorithm, if $d_G(v) = 2$ and the two neighbors of v are not in the same component of $G[V_2]$, then steps 5-6 of the algorithm will not be taken. Therefore, the only remaining case we need to examine is that $d_G(v) \geq 3$ and $d_{G[V_1]}(v) \geq 1$. By step 3 of the algorithm, in this case, we must have $d_{G[V_1]} = 1$. Let w be the unique neighbor of v in $G[V_1]$. By the way we picked the vertex v and by our assumption $d_G(v) \geq 3$, no vertex in V_1 has degree 2 in G . In particular, $d_G(w) \geq 3$. Therefore, deleting the vertex v can result in at most one degree-2 vertex in V_1 (i.e., w) and will keep all other vertices in V_1 with degree at least 3. Thus, in this case steps 5-6 of the algorithm again produce a nearly V_1 -irreducible instance.

Finally, note that the second footnote in the algorithm ensures that steps 7-8 will not be taken if the two neighbors of v are in the same component in $G[V_2]$. Moreover, steps 4-8 keep G a simple graph since they never smoothen vertices. These ensure that steps 4-8 produce a valid instance of DISJOINT-FVS. \square

We make some comments on the algorithm FindFVS. First of all, if there is no vertex in V_1 that has degree 2 in G , then the third line in step 3 must be able to find a vertex of degree ≤ 1 in the subgraph $G[V_1]$ since V_1 induces a forest. Now consider the correctness of the actions taken in branching steps 4-8. By the footnotes given in the algorithm FindFVS, if the selected vertex v has degree 2 in G , then no branching is taken and only one of the cases 1-2 is executed: (1) if both neighbors of v are in the same component of $G[V_2]$, then only steps 5-6 for case 1 are executed, i.e., the vertex v is directly included in the objective FVS F ; and (2) if the two neighbors of v are not in the same component of $G[V_2]$, then only steps 7-8 for case 2 are executed, i.e., the vertex v is moved from V_1 to V_2 . The correctness of the algorithm FindFVS for these cases is guaranteed by Lemma 2.2, which ensures the safeness of Rule 2. When the selected vertex v has a degree different from 2, then the branching steps 4-8 are exhaustive and consider both the cases where v is and is not in the objective FVS. Thus, one of these actions must be correct. Therefore, if the graph G has a V_1 -FVS of size k , then one of the computational paths in the search tree corresponding to the algorithm FindFVS must correctly find such a V_1 -FVS.

Theorem 2.4 *Let $(G; V_1, V_2; k)$ be a nearly V_1 -irreducible instance of the DISJOINT-FVS problem, and let τ_1 and τ_2 be the number of components in the induced subgraphs $G[V_1]$ and $G[V_2]$, respectively. Let δ_2 be the number of vertices in V_1 that have degree 2 in G . If $|V_1| > \delta_2 + 2k + \tau_2 - \tau_1 - 1$, then there is no V_1 -FVS of size bounded by k in the graph G .*

PROOF. We prove the theorem by induction on the number $|V_1|$ of vertices in the set V_1 . For $|V_1| = 1$, we have $\tau_1 = 1$, and the condition $|V_1| > \delta_2 + 2k + \tau_2 - \tau_1 - 1$ implies $\delta_2 + 2k + \tau_2 \leq 2$. Let w be the unique vertex in V_1 . If $\tau_2 = 0$, then the vertex w in V_1 would have degree 0 in G (note that by our assumption,

G is a simple graph), contradicting the assumption that the graph G is nearly V_1 -irreducible. Thus, we must have $1 \leq \tau_2 \leq 2$, which implies $k = 0$. If $\tau_2 = 1$, then since the vertex w in V_1 has degree at least 2, two neighbors of w must be in the same (and unique) component of $G[V_2]$. If $\tau_2 = 2$, then from $\delta_2 + 2k + \tau_2 \leq 2$ we have $\delta_2 = 0$, and the vertex w has degree at least 3, which implies again that at least two neighbors of w are in the same component of $G[V_2]$. Thus, for both cases of $\tau_2 = 1$ and $\tau_2 = 2$, the vertex w in V_1 must be included in every V_1 -FVS for G , which concludes that no V_1 -FVS of G can have size bounded by $k = 0$. This verifies the theorem for the case $|V_1| = 1$.

Now consider the general case of $|V_1| > 1$. Let $(G; V_1, V_2; k)$ be a nearly V_1 -irreducible instance of DISJOINT-FVS and suppose that the graph G has a V_1 -FVS of size bounded by k . Since the algorithm FindFVS solves DISJOINT-FVS correctly, there is a computational path \mathcal{P} of the algorithm that returns a V_1 -FVS F with $|F| \leq k$. We consider how the path \mathcal{P} changes the values of an instance when it executes (correctly) the action of one of the cases in steps 4-8 in the algorithm. Let $|V_1|$, δ_2 , k , τ_1 , and τ_2 be the values before the execution of steps 4-8, and let $|V'_1|$, δ'_2 , k' , τ'_1 , and τ'_2 be the corresponding values after the execution of steps 4-8. The relations between these values are summarized in Figure 3, where many are obvious. We give below explanations for some less obvious ones in the figure.

We first consider the case where the computational path \mathcal{P} takes the action of case 2 in the algorithm, i.e., moving the vertex v from set V_1 to set V_2 . See Table I in Figure 3.

If $d_G(v) = 2$ and both neighbors w_1 and w_2 of v are in the set V_2 (see the 3rd line in Table I in Figure 3), then by the second footnote in the algorithm, w_1 and w_2 must belong to two different components of $G[V_2]$. Therefore, moving v from V_1 to V_2 must decrease τ_1 by 1 (because v by itself makes a component in $G[V_1]$) and merge the two components of $G[V_2]$ into one (i.e., $\tau'_2 = \tau_2 - 1$).

If $d_G(v) \geq 3$ and v has no neighbor in V_1 (see the 5th line in Table I in Figure 3), then all neighbors of v (there are at least 3) are in V_2 . Moreover, by the second footnote in the algorithm, no two neighbors of v are in the same component of $G[V_2]$. Therefore, moving v from V_1 to V_2 decreases the value τ_1 by 1 (i.e., $\tau'_1 = \tau_1 - 1$) and merges at least three components of $G[V_2]$ into one (i.e., $\tau'_2 \leq \tau_2 - 2$).

If $d_G(v) \geq 3$ and $N(v) \cap V_1 \neq \emptyset$, then by step 3 of the algorithm, v has exactly one neighbor in V_1 and at least two neighbors in V_2 . Therefore, if v is moved from V_1 to V_2 (see the 6th line in Table I in Figure 3), then the value τ_1 is unchanged (i.e., $\tau'_1 = \tau_1$), and again by the second footnote in the algorithm, the value τ_2 is decreased by at least 1 (i.e., $\tau'_2 \leq \tau_2 - 1$).

Now consider the case where the computational path \mathcal{P} takes the action of case 1 in the algorithm, i.e., deleting the vertex v from the graph G . See Table II in Figure 3. First note that by the first footnote in the algorithm, if v has degree 2 and if the two neighbors of v do not belong to the same component of $G[V_2]$, then the action of case 1 in the algorithm is not taken. In particular, the action of case 1 in the algorithm is not applicable under the conditions of the 2nd line and the 4th line in Table II in Figure 3.

If $d_G(v) \geq 3$ and if v has no neighbors in V_1 (see the 5th line in Table II in Figure 3), then deleting v does not change the number of degree-2 vertices in V_1 (i.e., $\delta'_2 = \delta_2 = 0$) but decreases the value τ_1 by 1 (i.e., $\tau'_1 = \tau_1 - 1$, because v by itself makes a component in $G[V_1]$).

Finally, if $d_G(v) \geq 3$ and $N(v) \cap V_1 \neq \emptyset$ (see the 6th line in Table II in Figure 3), then by the way we picked the vertex v , we must have $|N(v) \cup V_1| = 1$. Let w be the unique neighbor of v in V_1 . Then, deleting v may create at most one degree-2 vertex (i.e., w) in the set V_1 (i.e., $\delta'_2 \leq \delta_2 + 1$), while not changing the values of τ_1 and τ_2 .

This verifies all relations in Tables I and II in Figure 3.

Let $(G'; V'_1, V'_2; k')$ be the instance produced by the computational path \mathcal{P} on the nearly V_1 -irreducible instance $(G; V_1, V_2; k)$. By our assumption, the graph G has a V_1 -FVS of size k . Since we also assume that the computational path \mathcal{P} is correct, the graph G' must have a V'_1 -FVS of size bounded by k' . Since $|V'_1| = |V_1| - 1$ and by Lemma 2.3, the instance $(G'; V'_1, V'_2; k')$ is nearly V'_1 -irreducible, we can apply the induction on the instance $(G'; V'_1, V'_2; k')$, which gives $|V'_1| \leq \delta'_2 + 2k' + \tau'_2 - \tau'_1 - 1$. This gives

$$|V_1| = |V'_1| + 1 \leq \delta'_2 + 2k' + \tau'_2 - \tau'_1 - 1 + 1.$$

Table I. Moving the vertex v from set V_1 to set V_2

degree of v	neighbors of v	δ'_2	k'	τ'_1	τ'_2	V'_1
$d_G(v) = 2$ with neighbors w_1 and w_2	$w_1, w_2 \in V_1$	$\delta_2 - 1$	k	$\tau_1 + 1$	$\tau_2 + 1$	$V_1 - \{v\}$
	$w_1, w_2 \in V_2$	$\delta_2 - 1$	k	$\tau_1 - 1$	$\tau_2 - 1$	$V_1 - \{v\}$
	$w_1 \in V_1, w_2 \in V_2$	$\delta_2 - 1$	k	τ_1	τ_2	$V_1 - \{v\}$
$d_G(v) \geq 3$	$ N(v) \cap V_1 = 0$	δ_2	k	$\tau_1 - 1$	$\leq \tau_2 - 2$	$V_1 - \{v\}$
$d_G(v) \geq 3$	$ N(v) \cap V_1 = 1$	δ_2	k	τ_1	$\leq \tau_2 - 1$	$V_1 - \{v\}$

Table II. Deleting the vertex v in V_1 from the graph G

degree of v	neighbors of v	δ'_2	k'	τ'_1	τ'_2	V'_1
$d_G(v) = 2$ with neighbors w_1 and w_2	$w_1, w_2 \in V_1$					
	$w_1, w_2 \in V_2$	$\delta_2 - 1$	$k - 1$	$\tau_1 - 1$	τ_2	$V_1 - \{v\}$
	$w_1 \in V_1, w_2 \in V_2$					
$d_G(v) \geq 3$	$ N(v) \cap V_1 = 0$	δ_2	$k - 1$	$\tau_1 - 1$	τ_2	$V_1 - \{v\}$
$d_G(v) \geq 3$	$ N(v) \cap V_1 = 1$	$\leq \delta_2 + 1$	$k - 1$	τ_1	τ_2	$V_1 - \{v\}$

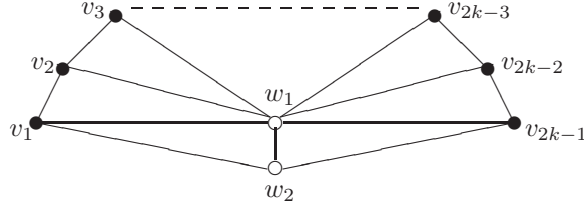
Figure 3: Results of applying the steps 4-8 of algorithm FindFVS on vertex v 

Figure 4: An example showing the tightness of Corollary 2.5.

Using this inequality to examine each situation in Figure 3, we can easily verify that the inequality

$$|V_1| \leq \delta_2 + 2k + \tau_2 - \tau_1 - 1$$

holds true. Therefore, if $|V_1| > \delta_2 + 2k + \tau_2 - \tau_1 - 1$, then the graph G has no V_1 -FVS of size bounded by k . This completes the proof of the theorem. \square

Since a V_1 -irreducible instance is also nearly V_1 -irreducible in which $\delta_2 = 0$, we get immediately

Corollary 2.5 *Let $(G; V_1, V_2; k)$ be a V_1 -irreducible instance of the DISJOINT-FVS problem. If $|V_1| > 2k + \tau_2 - \tau_1 - 1$, then there is no V_1 -FVS of size bounded by k in the graph G .*

The bound given in Corollary 2.5 is in fact *tight*, which can be seen as follows. Consider the graph G in Figure 4, which consists of $2k+1$ vertices $w_1, w_2, v_1, v_2, \dots, v_{2k-1}$, where $k \geq 2$ is an arbitrary positive integer. The vertices of G are partitioned into two sets $V_1 = \{v_1, v_2, \dots, v_{2k-1}\}$ and $V_2 = \{w_1, w_2\}$, and $(G; V_1, V_2; k)$ is a V_1 -irreducible instance of the DISJOINT-FVS problem. Note that $\tau_1 = \tau_2 = 1$. We have $|V_1| = 2k-1 = 2k + \tau_2 - \tau_1 - 1$, while the graph G has a V_1 -FVS F of k vertices: $F = \{v_1, v_3, v_5, \dots, v_{2k-1}\}$.

A particularly interesting class of instances of the DISJOINT-FVS problem was motivated by the iterative compression method for solving the FVS problem, in which each instance $(G; V_1, V_2; k)$ satisfies an additional condition $|V_2| = k+1$. Call this restricted version of DISJOINT-FVS the DISJOINT-SMALLER-FVS problem. For this important version of DISJOINT-FVS, we have the following kernelization result.

Theorem 2.6 *The DISJOINT-SMALLER-FVS problem has a $4k$ -vertex kernel: there is a polynomial-time algorithm that, on an instance $(G; V_1, V_2; k)$ of DISJOINT-SMALLER-FVS, produces an equivalent instance $(G'; V'_1, V'_2; k')$ of DISJOINT-SMALLER-FVS such that $k' \leq k$ and the graph G' contains at most $4k'$ vertices.*

PROOF. On an instance $(G; V_1, V_2; k)$ of DISJOINT-SMALLER-FVS, we apply Rule 1 and Rule 2 on vertices in V_1 . However, for a degree-2 vertex v in V_1 with neighbors u_1 and u_2 not in the same component of $G[V_2]$, we smoothen v *except* in the case where u_1 is in V_1 , u_2 is in V_2 , and $[u_1, u_2]$ is an edge in G . In this case we instead include u_1 in the objective FVS, and remove both u_1 and v . This change can be justified as follows. By Lemma 2.2, we can move v from V_1 to V_2 , which will make u_1 a vertex in V_1 that has two neighbors v and u_2 in the same component in $G[V_2]$. Thus, u_1 can be included directly in the objective FVS, and removed. The removal of u_1 makes v become a degree-1 vertex so can also be removed.

The reason for this change is that we want to keep G a simple graph without changing the vertex set V_2 . Smoothening a degree-2 vertex v in V_1 with neighbors u_1 and u_2 such that $[u_1, u_2]$ is an edge will create multiple edges. Note that in this case, (1) u_1 and u_2 cannot be both in V_1 since V_1 induces a forest; and (2) u_1 and u_2 cannot be both in V_2 because otherwise, v would have two neighbors in the same component of $G[V_2]$ and v would be included in the objective FVS. Thus, the only possibility that this may happen is that one of u_1 and u_2 is in V_1 and the other is in V_2 . Thus, the process in the previous paragraph avoids creating multiple edges, keeps the graph G a simple graph, and keep the vertex set V_2 unchanged (although it may add edges between vertices in V_2 when smoothening degree-2 vertices in V_1).

We repeat this process until it is no longer applicable. Let $(G''; V_1'', V_2''; k'')$ be the resulting instance. By Lemma 2.2 and the above discussion, $(G''; V_1'', V_2''; k'')$ is a YES-instance of DISJOINT-FVS if and only if $(G; V_1, V_2; k)$ is a YES-instance of DISJOINT-SMALL-FVS. Moreover, $k'' \leq k$, $V_2'' = V_2$, and all vertices in V_1'' have degree at least 3 in G'' . Thus, $(G''; V_1'', V_2''; k'')$ is V_1'' -irreducible. By Corollary 2.5, we can assume $|V_1''| \leq 2k'' + \tau_2'' - \tau_1'' - 1$, where τ_1'' and τ_2'' are the number of components in $G''[V_1'']$ and $G''[V_2'']$, respectively, for which we have $\tau_2'' \leq |V_2''| = |V_2| = k + 1$ and $\tau_1'' \geq 1$. Thus, the total number $|G''|$ of vertices in the graph G'' is $|V_1''| + |V_2''| \leq (2k'' + (k + 1) - 2) + (k + 1) = 2(k'' + k)$.

However, $(G''; V_1'', V_2''; k'')$ may not be an instance of DISJOINT-SMALLER-FVS because we may have $|V_2''| = |V_2| = k + 1 > k'' + 1$. If this is the case, let $h = k - k''$, and we add a disjoint simple path $P_{2h} = (w_1, \dots, w_{2h})$ of $2h$ vertices to G'' and let these $2h$ vertices be adjacent to a fixed vertex u in V_2'' . Let the new graph be G' , with the vertex partition (V'_1, V'_2) , where $V'_1 = V_1'' \cup \{w_1, \dots, w_{2h}\}$ and $V'_2 = V_2''$. Now consider the instance $(G'; V'_1, V'_2; k')$ of DISJOINT-FVS, where $k' = k$. It is easy to verify that the graph G' has a V'_1 -FVS of $k' = k$ vertices if and only if the graph G'' has a V_1'' -FVS of $k' - h = k - h = k''$ vertices. Moreover, since $|V'_2| = |V_1''| = |V_2| = k + 1 = k' + 1$, $(G'; V'_1, V'_2; k')$ is a valid instance for DISJOINT-SMALLER-FVS. Therefore, $(G'; V'_1, V'_2; k')$ is a YES-instance of DISJOINT-SMALLER-FVS if and only if $(G; V_1, V_2; k)$ is a YES-instance of DISJOINT-SMALLER-FVS: this holds true because both of these conditions are equivalent to the condition that $(G''; V_1'', V_2''; k'')$ is a YES-instance of DISJOINT-FVS. Finally, the number of vertices in G' is equal to $|G''| + 2h \leq 2(k'' + k) + 2(k - k'') = 4k = 4k'$. \square

Finally, we remark that this kernelization result was obtained based on the branch-and-search algorithm FindFVS for the problem, instead of on an analysis of the resulting structure after applying reduction rules. This technique, to our best knowledge, had not been used in the literature of kernelization.

3 A polynomial-time solvable case for DISJOINT-FVS

In this section we consider a special class of instances for DISJOINT-FVS. This approach is closely related to the classical study on graph maximum genus embeddings [5, 15]. However, the study on graph maximum genus embeddings that is related to our approach is based on general spanning trees of a graph, while our approach must be restricted to only spanning trees that are constrained by the vertex partition (V_1, V_2) of an instance $(G; V_1, V_2; k)$ of DISJOINT-FVS. We start with a simple lemma.

Lemma 3.1 *Let G be a connected graph and let H be a subgraph of G such that H is a forest. There is a spanning tree in G that contains the entire subgraph H , and can be constructed in time $O(m\alpha(n))$, where $\alpha(n)$ is the inverse of Ackermann function.*

PROOF. The lemma can be proved based on a process that is similar to the well-known Kruskal's algorithm for constructing a minimum spanning tree for a given graph, which runs in time $O(m\alpha(n))$ if we do not have to sort the edges. Starting from a structure G_0 that initially consists of the forest H and all vertices in G that are not in H , we repeatedly add each of the remaining edges (in an arbitrary order) to the structure G_0 as long as the edge does not create a cycle. The resulting structure of this process must be a spanning tree that contains the entire subgraph H . \square

Let $(G; V_1, V_2; k)$ be an instance for DISJOINT-FVS. Since the induced subgraph $G[V_2]$ is a forest, by Lemma 3.1, there is a spanning tree T of the graph G that contains $G[V_2]$. Call a spanning tree that contains $G[V_2]$ a $G[V_2]$ -spanning tree.

For a graph H , denote by $E(H)$ the set of edges in H , and for an edge subset E' in H , denote by $H - E'$ the graph H with the edges in E' removed (the end vertices of these edges are not removed).

Let T be a $G[V_2]$ -spanning tree of the graph G . By the construction, every edge in $G - E(T)$ has at least one end in V_1 . Two edges in $G - E(T)$ are V_1 -adjacent if they have a common end in V_1 . A V_1 -adjacency matching in $G - E(T)$ is a partition of the edges in $G - E(T)$ into groups of one or two edges, called 1-groups and 2-groups, respectively, such that two edges in the same 2-group are V_1 -adjacent. A maximum V_1 -adjacency matching in $G - E(T)$ is a V_1 -adjacency matching in $G - E(T)$ that maximizes the number of 2-groups.

Definition 2 *Let $(G; V_1, V_2; k)$ be an instance of the DISJOINT-FVS problem. The V_1 -adjacency matching number $\nu(G, T)$ of a $G[V_2]$ -spanning tree T in G is the number of 2-groups in a maximum V_1 -adjacency matching in $G - E(T)$. The V_1 -adjacency matching number $\nu(G)$ of the graph G is the largest $\nu(G, T)$ over all $G[V_2]$ -spanning trees T in the graph G .*

An instance $(G; V_1, V_2; k)$ of DISJOINT-FVS is V_1 -cubic if every vertex in the set V_1 has degree exactly 3. Let $f_{V_1}(G)$ be the size of a minimum V_1 -FVS for G . Let $\beta(G)$ be the Betti number of G that is the total number of edges in $G - E(T)$ for any spanning tree T in G . Note that the edge set $G - E(T)$ forms a basis of the fundamental cycles for the graph G such that every cycle in G contains at least one edge in $G - E(T)$. In this sense, $\beta(G)$ is the number of fundamental cycles in the graph G [15].

Lemma 3.2 *For any V_1 -cubic instance $(G; V_1, V_2; k)$ of DISJOINT-FVS, we have $f_{V_1}(G) = \beta(G) - \nu(G)$. Moreover, a minimum V_1 -FVS of the graph G can be constructed in linear time from a $G[V_2]$ -spanning tree whose V_1 -adjacency matching number is $\nu(G)$.*

PROOF. First note that a maximum V_1 -adjacency matching in $G - E(T)$ for a $G[V_2]$ -spanning tree T can be constructed in linear time, as follows. Since each vertex in V_1 has degree 3 and T is a spanning tree in G , each vertex in $G - E(T)$ has degree bounded by 2. Thus, each component of $G - E(T)$ is either a simple (possibly trivial) path or a simple cycle. Therefore, a maximum V_1 -adjacency matching in $G - E(T)$ can be constructed trivially by maximally pairing the edges in every component of $G - E(T)$.

Let T be a $G[V_2]$ -spanning tree such that there is a V_1 -adjacency matching M in $G - E(T)$ that contains $\nu(G)$ 2-groups. Let U be the set of edges that are in the 1-groups in M . We construct a V_1 -FVS F as follows: (1) for each edge e in U , arbitrarily pick an end of e that is in V_1 and include it in F ; and (2) for each 2-group of two V_1 -adjacent edges e_1 and e_2 in M , pick the vertex in V_1 that is a common end of e_1 and e_2 and include it in F . Note that every cycle in the graph G contains at least one edge in $G - E(T)$, while now every edge in $G - E(T)$ has at least one end in F . Therefore, F is a FVS. By the above construction, F is a V_1 -FVS. The number of vertices in F is equal to $|U| + \nu(G)$. Since

$|U| = |G - E(T)| - 2\nu(G) = \beta(G) - 2\nu(G)$, we have $|F| = \beta(G) - \nu(G)$. This concludes that

$$f_{V_1}(G) \leq \beta(G) - \nu(G). \quad (1)$$

Now consider the other direction. Let F be a minimum V_1 -FVS for the graph $G = (V, E)$, i.e., $|F| = f_{V_1}(G)$. By Lemma 3.1, there is a spanning tree T in G that contains the entire subgraph $G - F$, which is a forest. We construct a V_1 -adjacency matching in $G - E(T)$ and show that it contains at least $(\beta(G) - |F|)$ 2-groups. Since T contains $G - F$, each edge in $G - E(T)$ has at least one end in F . Let E_2 be the set of edges in $G - E(T)$ that have their both ends in F , and let E_1 be the set of edges in $G - E(T)$ that have exactly one end in F .

Claim. Each end of an edge in E_2 is shared by exactly one edge in E_1 . In particular, no two edges in E_2 share a common end.

To prove the above claim, first note that since T is a spanning tree in G , each vertex in $F \subseteq V_1$, which has degree 3 in G , can be incident to at most two edges in $G - E(T) = E_1 \cup E_2$. In particular, if u is an end of an edge $[u, v]$ in E_2 (i.e., $u, v \in F$), then there is at most one other edge in $E_1 \cup E_2$ that is incident to u . Now assume to the contrary of the claim that the vertex u is not shared by an edge in E_1 . Then for the other two edges e_1 and e_2 in G that are incident to u , either both e_1 and e_2 are in T or exactly one of e_1 and e_2 is in E_2 . If both e_1 and e_2 are in T , then every edge in $G - E(T)$ (including $[u, v]$) has at least one end in $F \setminus \{u\}$. Similarly, if exactly one $[u, w]$ of the edges e_1 and e_2 is in E_2 , where w is also in F , then again every edge in $G - E(T)$ (including $[u, v]$ and $[u, w]$) has at least one end in $F \setminus \{u\}$. Thus, in either case, $F \setminus \{u\}$ would make a smaller V_1 -FVS, contradicting the assumption that F is a minimum V_1 -FVS. This proves the claim.

Suppose that there are m_2 vertices in F that are incident to two edges in $G - E(T)$. Thus, each of the rest $|F| - m_2$ vertices in F is incident to at most one edge in $G - E(T)$. By counting the total number of incidencies between the vertices in F and the edges in $G - E(T)$, we get

$$2|E_2| + |E_1| = 2|E_2| + (\beta(G) - |E_2|) \leq 2m_2 + (|F| - m_2),$$

or equivalently,

$$m_2 - |E_2| \geq \beta(G) - |F|. \quad (2)$$

Now we construct a V_1 -adjacency matching in $G - E(T)$, as follows. For each edge e in E_2 , by the above claim, we can make a 2-group that consists of e and an edge in E_1 that shares an end in V_1 with e (note that this grouping will not put an edge in E_1 in two different 2-groups because if the edge e in E_2 shares an end with an edge e' in E_1 , then e' cannot share an end with any other edges in E_2). Besides the ends of the edges in E_2 , there are $m_2 - 2|E_2|$ vertices in F that are incident to two edges in E_1 . For each v of these vertices, we make a 2-group that consists of the two edges in E_1 that are incident to v . Note that this construction of 2-groups never uses any edges in $G - E(T)$ more than once. Therefore, the construction gives $|E_2| + (m_2 - 2|E_2|) = m_2 - |E_2|$ disjoint 2-groups. We then make each of the rest edges in $G - E(T)$ a 1-group. This gives a V_1 -adjacency matching in $G - E(T)$ that has $m_2 - |E_2|$ 2-groups. By Inequality (2) and by definition, we have

$$\nu(G) \geq \nu(G, T) \geq m_2 - |E_2| \geq \beta(G) - |F| = \beta(G) - f_{V_1}(G). \quad (3)$$

Combining (1) and (3), we conclude with $f_{V_1}(G) = \beta(G) - \nu(G)$.

The first two paragraphs in this proof also illustrate how to construct in linear time a minimum V_1 -FVS from a $G[V_2]$ -spanning tree whose V_1 -adjacency matching number is $\nu(G)$. \square

By Lemma 3.2, in order to construct a minimum V_1 -FVS for a V_1 -cubic instance $(G; V_1, V_2, k)$ of DISJOINT-FVS, we only need to construct a $G[V_2]$ -spanning tree in the graph G whose V_1 -adjacency matching number is $\nu(G)$. The construction of an unconstrained maximum adjacency matching in terms

of general spanning trees has been considered by Furst et al. [15] in their study of graph maximum genus embeddings. We follow a similar approach, based on cographic matroid parity, to construct a $G[V_2]$ -spanning tree in G whose V_1 -adjacency matching number is $\nu(G)$. We start with a quick review on the related concepts in matroid theory. More detailed discussion on this problem can be found in [22].

A *matroid* is a pair (E, \mathfrak{S}) , where E is a finite set and \mathfrak{S} is a nonempty collection of subsets of E that contains the empty set \emptyset and satisfies the following properties (note that the collection \mathfrak{S} may not be explicitly given but is defined in terms of certain subset properties):

- (1) If $A \in \mathfrak{S}$ and $B \subseteq A$, then $B \in \mathfrak{S}$;
- (2) If $A, B \in \mathfrak{S}$ and $|A| > |B|$, then there is an element $a \in A \setminus B$ such that $B \cup \{a\} \in \mathfrak{S}$.

The *matroid parity* problem is stated as follows: given a matroid (E, \mathfrak{S}) and a perfect pairing $\{[a_1, \bar{a}_1], [a_2, \bar{a}_2], \dots, [a_n, \bar{a}_n]\}$ of the elements in the set E , find a largest subset M in \mathfrak{S} such that for all i , $1 \leq i \leq n$, either both a_i and \bar{a}_i are in M , or neither of a_i and \bar{a}_i is in M .

Each connected graph G is associated with a *cographic matroid* (E_G, \mathfrak{S}_G) , where E_G is the edge set of G , and an edge set S is in \mathfrak{S}_G if and only if $G - S$ is connected. It is well-known that matroid parity problem for cographic matroids can be solved in polynomial time [22]. The fastest known algorithm for cographic matroid parity problem is by Gabow and Xu [16], which runs in time $O(mn \log^6 n)$.

In the following, we explain how to reduce our problem to the cographic matroid parity problem. Let $(G; V_1, V_2; k)$ be a V_1 -cubic instance of the DISJOINT-FVS problem. Without loss of generality, we make the following assumptions: (1) the graph G is connected (otherwise, we simply work on each component of G); and (2) for each vertex v in V_1 , there is at most one edge from v to a component in $G[V_2]$ (otherwise, we can directly include v in the objective V_1 -FVS).

Recall that two edges are V_1 -adjacent if they share a common end in V_1 . For an edge e in G , denote by $d_{V_1}(e)$ the number of edges in G that are V_1 -adjacent to e (note that an edge can be V_1 -adjacent to the edge e from either end of e).

We construct a *labeled subdivision* G_2 of the graph G as follows.

1. shrink each component of $G[V_2]$ into a single vertex; let the resulting graph be G_1 ;
2. assign each edge in G_1 a distinguished label;
3. for each edge labeled e_0 in G_1 , suppose the edges V_1 -adjacent to e_0 are labeled by e_1, e_2, \dots, e_d (in arbitrary order), where $d = d_{V_1}(e_0)$; subdivide e_0 into d *segment edges* by inserting $d - 1$ degree-2 vertices in e_0 , and label the segment edges by $(e_0e_1), (e_0e_2), \dots, (e_0e_d)$. Let the resulting graph be G_2 . The segment edges $(e_0e_1), (e_0e_2), \dots, (e_0e_d)$ in G_2 are said to be *from* the edge e_0 in G_1 .

There are a number of interesting properties for the graphs constructed above. First, each of the edges in the graph G_1 corresponds uniquely to an edge in G that has at least one end in V_1 . Thus, without creating any confusion, we will simply say that the edge is in the graph G or in the graph G_1 . Second, because of the assumptions we made on the graph G , the graph G_1 is a simple and connected graph. In consequence, the graph G_2 is also a simple and connected graph. Finally, because each edge in G_1 corresponds to an edge in G that has at least one end in V_1 , and because each vertex in V_1 has degree 3, every edge in G_1 is subdivided into at least two segment edges in G_2 .

Now in the labeled subdivision graph G_2 , pair the segment edge labeled (e_0e_i) with the segment edge labeled (e_ie_0) for all segment edges (note that (e_0e_i) is a segment edge from the edge e_0 in G_1 and that (e_ie_0) is a segment edge from the edge e_i in G_1). By the above remarks, this is a perfect pairing \mathcal{P} of the edges in G_2 . Now with this edge pairing \mathcal{P} in G_2 , and with the cographic matroid $(E_{G_2}, \mathfrak{S}_{G_2})$ for the graph G_2 , we call Gabow and Xu's algorithm [16] for the cographic matroid parity problem. The algorithm produces a maximum edge subset M in \mathfrak{S}_{G_2} that, for each segment edge (e_0e_i) in G_2 , either contains both (e_0e_i) and (e_ie_0) , or contains neither of (e_0e_i) and (e_ie_0) .

Lemma 3.3 *From the edge subset M in \mathfrak{S}_{G_2} constructed above, a $G[V_2]$ -spanning tree for the graph G with a V_1 -adjacency matching number $\nu(G)$ can be constructed in time $O(m\alpha(n))$, where n and m are the number of vertices and the number of edges, respectively, of the original graph G .*

PROOF. Suppose that the edge subset M consists of the edge pairs $\{[(e_1e'_1), (e'_1e_1)], \dots, [(e_h e'_h), (e'_h e_h)]\}$ in G_2 . Since $M \in \mathfrak{S}_{G_2}$, $G_2 - M$ is connected. Thus, for each edge e_i in G_1 , there is at most one segment edge in M that is from e_i . Therefore, the edge subset M corresponds to an edge subset M' of exactly $2h$ edges in G_1 (thus exactly $2h$ edges in G): $M' = \{e_1, e'_1; \dots, e_h, e'_h\}$, where for $1 \leq i \leq h$, the edges e_i and e'_i are V_1 -adjacent. Since $G_2 - M$ is connected, it is easy to verify that the graph $G_1 - M'$ (thus the graph $G - M'$) is also connected. Also note that the graph $G - M'$ contains the induced subgraph $G[V_2]$ because no edge in G_1 has its both ends in V_2 . Therefore, by Lemma 3.1, we can construct, in time $O(m\alpha(n))$, a $G[V_2]$ -spanning tree T_1 for the graph $G - M'$, which is also a $G[V_2]$ -spanning tree for the graph G . Now if we make each pair $[e_i, e'_i]$ a 2-group for $1 \leq i \leq h$, and make each of the rest edges in $G - E(T_1)$ a 1-group, we get a V_1 -adjacency matching with h 2-groups in $G - E(T_1)$.

To complete the proof of the lemma, we only need to show that $h = \nu(G)$. For this, it suffices to show that no $G[V_2]$ -spanning tree can have a V_1 -adjacency matching with more than h 2-groups. Let T_2 be a $G[V_2]$ -spanning tree with q 2-groups $[e_1, e'_1], \dots, [e_q, e'_q]$ in $G - E(T_2)$. Since $G - \bigcup_{i=1}^q \{e_i, e'_i\}$ entirely contains T_2 , it is connected. In consequence, the graph $G_1 - \bigcup_{i=1}^q \{e_i, e'_i\}$ is also connected. From this, it is easy to verify that the graph $G_2 - \bigcup_{i=1}^q \{(e_i e'_i), (e'_i e_i)\}$ is also connected. Therefore, the edge subset $\{(e_1 e'_1), (e'_1 e_1); \dots, (e_q e'_q), (e'_q e_q)\}$ is in \mathfrak{S}_{G_2} . Now since M is the the solution of the matroid parity problem for the cographic matroid $(E_{G_2}, \mathfrak{S}_{G_2})$ and since M consists of h edge pairs, we must have $h \geq q$. This completes the proof of the lemma. \square

Now we are ready to present our main result in this section, which is a nontrivial generalization of a result in [28] (the result in [28] can be viewed as a special case of Lemma 3.2 in which all vertices in the set V_2 have degree 2).

Theorem 3.4 *There is an $O(n^2 \log^6 n)$ -time algorithm that on a V_1 -cubic instance $(G; V_1, V_2; k)$ of DISJOINT-FVS, either constructs a V_1 -FVS of size bounded by k , if such a V_1 -FVS exists, or reports correctly that no such a V_1 -FVS exists.*

PROOF. For the V_1 -cubic instance $(G; V_1, V_2; k)$ of DISJOINT-FVS, we first construct the graph G_1 in linear time by shrinking each component of $G[V_2]$ into a single vertex. Note that since each vertex in V_1 has degree 3, the total number of edges in G_1 is bounded by $3|V_1|$. From the graph G_1 , we construct the labeled subdivision graph G_2 . Again since each vertex in V_1 has degree 3, each edge in G_1 is subdivided into at most 4 segment edges in G_2 . Therefore, the number n_2 of vertices and the number m_2 of edges in G_2 are both bounded by $O(|V_1|) = O(n)$. From the graph G_2 , we apply Gabow and Xu's algorithm [16] on the cographic matroid $(E_{G_2}, \mathfrak{S}_{G_2})$ that produces the edge subset M in \mathfrak{S}_{G_2} in time $O(m_2 n_2 \log^6 n_2) = O(n^2 \log^6 n)$. By Lemma 3.3, from the edge subset M , we can construct in time $O(m\alpha(n))$ a $G[V_2]$ -spanning tree T for the graph G whose V_1 -adjacency matching number is $\nu(G)$. Finally, by Lemma 3.2, from the $G[V_2]$ -spanning tree T , we can construct a minimum V_1 -FVS F in linear time. Now the solution to the V_1 -cubic instance $(G; V_1, V_2; k)$ of DISJOINT-FVS can be trivially derived by comparing the size of F and the parameter k . Summarizing all these steps gives the proof of the theorem. \square

Combining Theorem 3.4 and Lemma 2.2, we have

Corollary 3.5 *There is an $O(n^2 \log^6 n)$ -time algorithm that on an instance $(G; V_1, V_2; k)$ of DISJOINT-FVS where all vertices in V_1 have degree bounded by 3, either constructs a V_1 -FVS of size bounded by k , if such a FVS exists, or reports correctly that no such a V_1 -FVS exists.*

We remark that Corollary 3.5 is the best possible in terms of the maximum vertex degree in the set V_1 . This can be reasoned as follows. It is known that the FVS problem on graphs of maximum degree 4 is NP-hard [26]. Given an instance G of the FVS problem on graphs of maximum degree 4, we add a degree-2 vertex to the middle of each edge in G . Let the new graph be G' . Let V_1 be the set of vertices in G' that correspond to the original vertices in G , and let V_2 be the set of new degree-2 vertices in G' . Now it is rather straightforward to see that a minimum V_1 -FVS in G' corresponds to a minimum FVS in the original graph G . Moreover, the maximum vertex degree in the set V_1 in G' is bounded by 4. This proves that the DISJOINT-FVS problem is NP-hard even when restricted to graphs in which the maximum vertex degree in the set V_1 is 4.

4 An improved algorithm for DISJOINT-FVS

Now we consider DISJOINT-FVS in general. Let $(G; V_1, V_2; k)$ be an instance of DISJOINT-FVS, for which we are looking for a V_1 -FVS of size bounded by k . Our algorithm for solving the DISJOINT-FVS problem is presented in Figure 5.

Algorithm Feedback(G, V_1, V_2, k)

INPUT: an instance $(G; V_1, V_2; k)$ of DISJOINT-FVS.

$\backslash \backslash$ p = the number of nice V_1 -vertices; τ_2 = the number of components in $G[V_2]$.

OUTPUT: a V_1 -FVS F of size bounded by k in G if such a V_1 -FVS exists, or “No” otherwise.

1. **if** $(k < 0)$ or $(k = 0$ and G is not a forest) or $(2p \geq 2k + \tau_2)$ **then** return “No”;
2. **if** $(k \geq 0$ and G is a forest) or $(p = |V_1|)$ **then** solve the problem in polynomial time;
3. **if** a vertex $w \in V_1$ has degree ≤ 1 **then** return **Feedback** $(G - w, V_1 \setminus \{w\}, V_2, k)$;
4. **if** a vertex $w \in V_1$ has two neighbors in the same component in $G[V_2]$ **then** return $\{w\} \cup \text{Feedback}(G - w, V_1 \setminus \{w\}, V_2, k - 1)$;
5. **if** a vertex $w \in V_1$ has degree 2 **then**
 return **Feedback** (G', V_1, V_2, k) , where $G' = G$ with the vertex w smoothened;
6. **if** a leaf w in $G[V_1]$ is not a nice V_1 -vertex and has ≥ 3 neighbors in V_2 **then**
 6.1 $F_1 = \text{Feedback}(G - w, V_1 \setminus \{w\}, V_2, k - 1)$;
 6.2 **if** $F_1 \neq \text{“No”}$ **then** return $F_1 \cup \{w\}$
 6.3 **else** return **Feedback** $(G, V_1 \setminus \{w\}, V_2 \cup \{w\}, k)$;
7. pick a lowest parent w in any tree in $G[V_1]$ and let v be a child of w ;
 7.1 $F_1 = \text{Feedback}(G - w, V_1 \setminus \{w, v\}, V_2 \cup \{v\}, k - 1)$;
 7.2 **if** $F_1 \neq \text{“No”}$ **then** return $F_1 \cup \{w\}$
 7.3 **else** return **Feedback** $(G, V_1 \setminus \{w\}, V_2 \cup \{w\}, k)$.

Figure 5: Algorithm for DISJOINT-FVS

We first give some explanations to the terminologies used in the algorithm. A vertex v in the set V_1 is a *nice V_1 -vertex* if v is of degree 3 and if all its neighbors are in the set V_2 . We will denote by p the number of nice V_1 -vertices in G , and, as before, by τ_2 the number of components in the induced subgraph $G[V_2]$. We have slightly abused the use of the set union operation in step 4 in the sense that when **Feedback** $(G - w, V_1 \setminus \{w\}, V_2, k - 1)$ returns “No,” then the union $\{w\} \cup \text{Feedback}(G - w, V_1 \setminus \{w\}, V_2, k - 1)$ is also interpreted as a “No.” In step 5, by “smoothening” a degree-2 vertex w , we mean replacing the vertex w and the two edges incident to w with a new edge connecting the two neighbors of w . In step 6, by a “leaf” in $G[V_1]$, we mean a vertex w that has at most one neighbor in the set V_1 . Finally, in step 7, we assume that we have picked an (arbitrary) vertex in each tree in $G[V_1]$ and designate it as the root of the tree so that a parent-child relationship is defined in the tree. A “lowest parent” w in a tree in $G[V_1]$ is a vertex in the tree that has children and all its children are leaves.

We start with the following lemma.

Lemma 4.1 *If $2p \geq 2k + \tau_2$, then there is no V_1 -FVS of size bounded by k in the graph G .*

PROOF. Suppose that there is a V_1 -FVS F of size $k' \leq k$. Let V_1' be the set of any $p - k'$ nice V_1 -vertices that are not in F . Then the subgraph $G' = G[V_2 \cup V_1']$ induced by the vertex set $V_2 \cup V_1'$ is a forest. On the other hand, the subgraph G' can be constructed from the induced subgraph $G[V_2]$ and the $p - k'$ isolated vertices in V_1' , by adding the $3(p - k')$ edges that are incident to the vertices in V_1' . Since $k' \leq k$, we have $2(p - k') \geq 2(p - k) \geq \tau_2$. This gives $3(p - k') = 2(p - k') + (p - k') \geq \tau_2 + (p - k')$. This contradicts the fact that G' is a forest—in order to keep G' a forest, we can add at most $\tau_2 + (p - k') - 1$ edges to the structure that consists of the induced subgraph $G[V_2]$ of τ_2 components and the $p - k'$ isolated vertices in V_1' . This contradiction proves the lemma. \square

Now we are ready to analyze the algorithm $\text{Feedback}(G, V_1, V_2, k)$ for the DISJOINT-FVS problem in Figure 5. We first prove the correctness of the algorithm.

Lemma 4.2 *The algorithm Feedback solves the DISJOINT-FVS problem correctly.*

PROOF. The correctness of step 1 follows from Lemma 4.1 and other trivial facts. If $k \geq 0$ and the graph G is a forest, then obviously the empty set \emptyset is a solution to the input instance. If $p = |V_1|$, then by definition, all vertices in the set V_1 have degree 3. By Corollary 3.5, this case can be solved in polynomial time. This verifies the correctness of step 2. The correctness of step 3 follows from the fact that no vertices of degree bounded by 1 can be contained in any cycle. Step 4 is correct because in this case, the vertex w is the only vertex in the set V_1 in a cycle in the graph G , so it must be included in the objective V_1 -FVS. Step 5 follows from Lemma 2.2 and the fact that step 4 does apply to the vertex w .

Step 6 is correct because it simply branches on either including or excluding the vertex w in the objective V_1 -FVS. Note that after passing steps 3-5, all vertices in the set V_1 have degree at least 3, and after passing steps 3-6, each vertex in the set V_1 either is a nice V_1 -vertex or has at least one neighbor in V_1 . In particular, after steps 3-6, if a leaf v in $G[V_1]$ is not a nice V_1 -vertex, then v has exactly two neighbors in V_2 that belong to two different components of $G[V_2]$. Now consider step 7. As remarked above (also noting step 2), at this point there must be a tree with more than one vertex in the induced subgraph $G[V_1]$. Therefore, we can always find a lowest parent w in a tree in $G[V_1]$. Step 7 branches on this lowest parent w . In case w is included in the objective V_1 -FVS, w is deleted from the graph, and the parameter k is decreased by 1. Note that after the vertex w is deleted, the child v of w becomes of degree 2 with its two neighbors in two different components of $G[V_2]$. By Lemma 2.1, the vertex v can be excluded from the objective V_1 -FVS. Thus, it is safe to move the vertex v from set V_1 to set V_2 . This verifies the correctness of steps 7.1-7.2. Step 7.3 is simply to exclude the vertex w from the objective V_1 -FVS.

Observe that before making recursive calls, each of the steps 3-7 decreases the number of vertices in the set V_1 by at least 1. Therefore, the algorithm must terminate in a finite number of steps. Summarizing all the above discussion, we conclude with the correctness of the algorithm $\text{Feedback}(G, V_1, V_2, k)$. \square

Now we analyze the complexity of the algorithm Feedback . The recursive execution of the algorithm can be depicted as a search tree \mathcal{T} , whose complexity can be analyzed by counting the number of leaves in the search tree. For an input instance (G, V_1, V_2, k) , we, as before, let p be the number of nice V_1 -vertices in G , and let τ_2 be the number of components in the induced subgraph $G[V_2]$. To analyze the complexity of the algorithm more precisely, we introduce a new measure, defined as $\mu = 2(k - p) + \tau_2$. Let $T(\mu)$ be the number of leaves in the search tree \mathcal{T} for the algorithm on the input (G, V_1, V_2, k) .

Theorem 4.3 *The algorithm $\text{Feedback}(G, V_1, V_2, k)$ correctly solves the DISJOINT-FVS problem in time $O(2^{k+\tau_2/2} n^2 \log^6 n)$, where n is the number of vertices in the graph G , and τ_2 is the number of components in the induced subgraph $G[V_2]$.*

PROOF. We have verified the correctness of the algorithm in Lemma 4.2. Herein we analyze its complexity, i.e., we consider the value $T(\mu)$.

Each of steps 1-5 of the algorithm proceeds without branching; hence it suffices to verify that neither of them *increases* the value of the measure μ . Step 3 does not change the values of k , p , and τ_2 , thus neither that of μ . Step 4 does not change the value τ_2 , but decreases the value k by 1. Moreover, step 4 *may* also decrease the value p by at most 1 (in case the vertex w is a nice V_1 -vertex). Overall, step 4 does not increase the value $\mu = 2(k - p) + \tau_2$. Step 5 does not change the value of k . Moreover, it will never decrease the value of p or increase the value of τ_2 . Note that step 5 may increase the value of p (e.g., a neighbor of w in V_1 may become a nice V_1 -vertex after smoothening w) or decrease the value of τ_2 (e.g., when the two neighbors of w are in two different components in $G[V_2]$). In any case, step 5 does not increase the value $\mu = 2(k - p) + \tau_2$.

Now we study the branching steps. First consider step 6. The branch of steps 6.1-6.2 decreases the value k by 1 and does not change the value of τ_2 . Moreover, the steps may increase the value of p (e.g., a neighbor of w in V_1 may become a nice V_1 -vertex after deleting w from the graph) but will never decrease the value of p . Therefore, the branch of steps 6.1-6.2 will decrease the value $\mu = 2(k - p) + \tau_2$ by at least 2. On the other hand, because w has at least three neighbors in V_2 , step 6.3 will decrease the value of τ_2 by at least 2, while neither changing the value of k nor decreasing the value of p . Thus, step 6.3 also decreases the value $\mu = 2(k - p) + \tau_2$ by at least 2. In summary, if step 6 is executed in the algorithm, then the function $T(\mu)$ satisfies the recurrence relation $T(\mu) \leq 2T(\mu - 2)$.

Similarly, the branch of steps 7.1-7.2 deletes the vertex w from the graph and decreases the value of k by 1. As we pointed out before, since the algorithm has passed steps 3-6, the leaf v has exactly three neighbors: one is w and the other two are in two different components in $G[V_2]$. Therefore, after deleting w from the graph, moving the degree-2 vertex v from set V_1 to set V_2 decreases the value of τ_2 by 1. Also note that in this branch, the value of p is not changed (because of step 6, the vertex w cannot have a neighbor that is a leaf in $G[V_1]$ but has three neighbors in V_2). In summary, the branch of steps 7.1-7.2 decreases the value $\mu = 2(k - p) + \tau_2$ by at least 3. Now consider step 7.3 that moves the vertex w from set V_1 to set V_2 . We break this case into two subcases:

Subcase 7.3.1. The vertex w has at least one neighbor in V_2 . Then moving w from V_1 to V_2 neither changes the value of k nor increases the value of τ_2 . On the other hand, it creates at least one new nice V_1 -vertex (i.e., the vertex v) thus increases the value of p by at least 1. Therefore, in this subcase, step 7.3 decreases the value of $\mu = 2(k - p) + \tau_2$ by at least 2.

Subcase 7.3.2. The vertex w has no neighbor in V_2 . Because the degree of w is larger than 2 and w is a lowest parent in $G[V_1]$, w has at least two children in V_1 , each is a leaf in $G[V_1]$ with exactly two neighbors that are in two different components of $G[V_2]$. Note that after moving w from V_1 to V_2 , all children of w in $G[V_1]$ will become nice V_1 -vertices. Therefore, moving w from V_1 to V_2 increases the value of τ_2 by 1, and increases the value of p by at least 2, with the value of k unchanged. Therefore, in this subcase, step 7.3 decreases the value of $\mu = 2(k - p) + \tau_2$ by at least 3.

Summarizing the above discussion, we conclude that if step 7 is executed in the algorithm, then the function $T(\mu)$ satisfies the recurrence relation $T(\mu) \leq T(\mu - 2) + T(\mu - 3)$.

Therefore, the function $T(\mu)$, which is the number of leaves in the search tree \mathcal{T} , in the worst case satisfies the recurrence relation $T(\mu) \leq 2T(\mu - 2)$. Also note that Lemma 4.1, if $\mu = 2(k - p) + \tau_2 \leq 0$, then we can conclude immediately without branching that the input instance is a “No.” Therefore, $T(\mu) = 1$ for $\mu \leq 0$. Now the recurrence relation $T(\mu) \leq 2T(\mu - 2)$ with $T(\mu) = 1$ for $\mu \leq 0$ can be solved using the well-known techniques in parameterized computation (see, for example, [11]), as follows. The characteristic polynomial for the recurrence relation $T(\mu) = 2T(\mu - 2)$ is $x^2 - 2$, which has a unique positive root $\sqrt{2}$. From this, we derive $T(\mu) = (\sqrt{2})^\mu = 2^{\mu/2}$. Moreover, it is fairly easy to see that each computational path in the search tree \mathcal{T} has its time bounded by $O(n^2 \log^6 n)$, and $\mu/2 = k - p + \tau_2/2 \leq k + \tau_2/2$. Therefore, the running time of the algorithm **Feedback**(G, V_1, V_2, k) is $O(2^{k+\tau_2/2} n^2 \log^6 n)$ \square

5 An improved algorithm for FVS

The results in previous sections lead to an improved algorithm for the general FVS problem. Following the idea of *iterative compression* proposed by Reed et al. [25], we formulate the following problem:

FVS REDUCTION: given a graph G and a FVS F of size $k + 1$ for G , either construct a FVS of size bounded by k for G , or report that no such a FVS exists.

Lemma 5.1 *The FVS REDUCTION problem can be solved in time $O^*(3.83^k)$.*

PROOF. The proof goes similar to that for Lemma 2 in [3]. Let $G = (V, E)$ be a graph and let F_{k+1} be a FVS of size $k + 1$ in G . Suppose that the graph G has a FVS F'_k of size k , and let the intersection $F_{k+1} \cap F'_k$ be a set F_{k-j} of $k - j$ vertices, for some j , $0 \leq j \leq k$. Let $F_{j+1} = F_{k+1} \setminus F_{k-j}$ and $F'_j = F'_k \setminus F_{k-j}$. Construct the graph $G' = G - F_{k-j}$. Note that both F_{j+1} and F'_j are FVS for G' , and that F_{j+1} and F'_j are disjoint. Thus, if we let $V'_1 = V \setminus F_{k+1}$ and $V'_2 = F_{j+1}$, then F'_j is a solution to the instance (G', V'_1, V'_2, j) of the DISJOINT-FVS problem. On the other hand, it is also easy to see that any solution to the instance (G', V'_1, V'_2, j) of DISJOINT-FVS plus the subset F_{k-j} makes a FVS of no more than k vertices for the original graph G .

Therefore, to solve the instance (G, F_{k+1}) for the FVS REDUCTION problem, it suffices to find the subset $F_{k-j} = F_{k+1} \cap F'_k$ of $k - j$ vertices in F_{k+1} for some integer j , $0 \leq j \leq k$, then to solve the instance (G', V'_1, V'_2, j) for the DISJOINT-FVS problem. To find the subset F_{k-j} of F_{k+1} , we enumerate all subsets of $k - j$ vertices in F_{k+1} for all $0 \leq j \leq k$. To solve the corresponding instance (G', V'_1, V'_2, j) for DISJOINT-FVS derived from the subset F_{k-j} of F_{k+1} , we call the algorithm $\text{Feedback}(G', V'_1, V'_2, j)$. By Theorem 4.3 (note that $\tau_2 \leq |V'_2| = j + 1$), the instance (G', V'_1, V'_2, j) for DISJOINT-FVS can be solved in time $O(2^{j+(j+1)/2} n^2 \log^6 n) = O(2.83^j n^2 \log^6 n)$. Applying this procedure for every integer j ($0 \leq j \leq k$) and all subsets of size $k - j$ in F_{k+1} will successfully find a FVS of size k in the graph G , if such a FVS exists. This algorithm solves the FVS REDUCTION problem in time $\sum_{j=0}^k \binom{k+1}{k-j} \cdot O(2.83^j n^2 \log^6 n) = O^*(3.83^k)$. \square

Finally, by combining Lemma 5.1 with the iterative compression technique [25, 6], we obtain the main result of this paper, which solves the FVS problem, formally defined as follows:

FVS: given a graph G and a parameter k , either construct a FVS of size bounded by k for the graph G , or report that no such FVS exists.

Theorem 5.2 *The FVS problem is solvable in time $O^*(3.83^k)$.*

PROOF. To determine if a given graph $G = (V, E)$ has a FVS of size bounded by k , we start by applying the polynomial-time approximation algorithm of approximation ratio 2 for the MINIMUM FEEDBACK VERTEX SET problem [1]. This algorithm runs in $O(n^2)$ time, and either returns a FVS F' of size at most $2k$, or verifies that no FVS of size bounded by k exists. Thus, if no FVS is returned by the algorithm, then no FVS of size bounded by k exists. In the case of the opposite result, we use any subset V' of k vertices in F' , and put $V_0 = V' \cup (V \setminus F')$. Obviously, the induced subgraph $G[V_0]$ has a FVS V' of size k . Let $F' \setminus V' = \{v_1, v_2, \dots, v_{|F'|-k}\}$, and let $V_i = V_0 \cup \{v_1, \dots, v_i\}$ for $i \in \{0, 1, \dots, |F'|-k\}$. Inductively, suppose that we have constructed a FVS F_i for the graph $G[V_i]$, where $|F_i| = k$. Then the set $F'_{i+1} = F_i \cup \{v_{i+1}\}$ is a FVS for the graph $G[V_{i+1}]$, and $|F'_{i+1}| = k + 1$.

Now the pair $(G[V_{i+1}], F'_{i+1})$ is an instance for the FVS REDUCTION problem. Therefore, in time $O^*(3.83^k)$, we can either construct a FVS F_{i+1} of size k for the graph $G[V_{i+1}]$, or report that no such a FVS exists. Note that if the graph $G[V_{i+1}]$ does not have a FVS of size k , then the original graph G cannot have a FVS of size k . In this case, we simply stop and claim the non-existence of a FVS of size k for the original graph G . On the other hand, with a FVS F_{i+1} of size k for the graph $G[V_{i+1}]$, our induction proceeds to the next graph $G[V_{i+1}]$, until we reach the graph $G = G[V_{|F'|-k}]$. This process runs in time $k \cdot O^*(3.83^k) = O^*(3.83^k)$ since $|F'| - k \leq k$, and solves the FVS problem. \square

6 Concluding remarks

We developed an $O^*(3.83^k)$ -time parameterized algorithm for the FVS problem. Our algorithm was obtained by a nontrivial combination of several known techniques in algorithm research and their generalizations. This includes iterative compression, branch-and-search, and efficient algorithms for graphs of low vertex-degrees. For branch-and-search processes for dealing with the FVS problem, we introduced new branching rules and new branching measures, which allow us to more effectively reduce a general instance into a polynomial-time solvable instance of the problem and to more accurately evaluate the efficiency of the branch-and-search process. For efficient algorithms for graphs of low vertex-degrees, we use a nontrivial reduction that transforms the FVS problem to a polynomial-time solvable version of the matroid matching problem. Note that using matroid matching to solve the FVS problem for 3-regular graphs has been observed previously [27, 28, 15], while we extended the techniques to solve the DISJOINT-FVS problem on a larger graph class in which not all vertices are required to have degree bounded by 3.

Further faster algorithms for FVS have drawn much attention in the recent research in parameterized computation [9]. Following our approach with a new reduction rule introduced, Kociumaka and Pilipczuk [20] have announced a revision of our algorithm that has an improved running time $O^*(3.62^k)$ for the FVS problem. On the other hand, the study on the lower bound of the FVS problem has made significant progress. Based on the *Strong Exponential Time Hypothesis* (see [21]), Cygan et al. [7] have reported a lower bound on the complexity of the FVS problem in terms of the pathwidth pw of a graph, which states that the FVS problem cannot be solved in time $O^*((3 - \epsilon)^{pw})$ for any positive constant $\epsilon > 0$. This result does not yet directly lead to a lower bound for the FVS problem in terms of the parameter k , which is the *size* of the objective FVS (to see this, observe that the ladder graph $P_l \times P_2$ has a pathwidth 2 but its minimum FVS has a size $\lfloor l/2 \rfloor$, where P_i denotes the simple path of i vertices). On the other hand, studying the complexity of the FVS problem in terms of graph pathwidth or treewidth seems to have very interesting connection to the complexity of the original FVS problem. For example, the $O^*(3^{tw})$ -time randomized algorithm for the FVS problem proposed in [7], where tw is the treewidth of the input graph, directly implies an $O^*(3^k)$ -time randomized algorithm for FVS. In particular, this has motivated an interesting open problem whether there is a deterministic $O^*(3^k)$ -time algorithm for the FVS problem [9].

It is interesting to observe that the research on parameterized algorithms and that on approximation algorithms for the FVS problem have undergone a similar process. Early algorithms used the cycle packing-covering duality, and hence ended with $O^*(\log k^{O(k)})$ -time parameterized algorithms [23, 19] and $O(\log n)$ -ratio approximation algorithms [12], respectively. Later algorithms turned to the observation on graph vertex-degrees, which resulted in $O^*(2^{O(k)})$ -time parameterized algorithms [6, 8] and constant-ratio approximation algorithms [1, 3], respectively. However, constant-ratio approximation algorithms for FVS do not seem to rely on a process that is related to the iterative compression process [25], which, on the other hand, seems to have played a critical role in the development of all $O^*(2^{O(k)})$ -time parameterized algorithms for the FVS problem. A parameterized algorithm based on iterative compression for the FVS problem runs in time $O^*((1 + \alpha)^k)$, where α is a constant such that the DISJOINT-FVS problem can be solved in time $O^*(\alpha^k)$. Since the DISJOINT-FVS problem is NP-hard, the constant α has to be larger than 1. In other words, using the iterative compression technique excludes the possibility of solving the FVS problem in time $O^*(2^k)$. An interesting research direction and a possible approach to developing further improved algorithms for the FVS problem is to explore new algorithmic techniques that are *not* based on iterative compression.

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