Node Multiway Cut and Subset Feedback Vertex Set on Graphs of Bounded Mim-width

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January 19, 2022

An extended abstract appeared in the proceedings of 46th International Workshop on Graph-Theoretic Concepts in Computer Science, WG 2020.

Abstract

The two weighted graph problems NODE MULTIWAY CUT (NMC) and SUBSET FEEDBACK VERTEX SET (SFVS) both ask for a vertex set of minimum total weight, that for NMC disconnects a given set of terminals, and for SFVS intersects all cycles containing a vertex of a given set. We design a meta-algorithm that allows to solve both problems in time $2^{O(rw^3)} \cdot n^4$, $2^{O(q^2 \log(q))} \cdot n^4$, and $n^{O(k^2)}$ where $rw$ is the rank-width, $q$ the $Q$-rank-width, and $k$ the mim-width of a given decomposition. This answers in the affirmative an open question raised by Jaffke et al. (Algorithmica, 2019) concerning an XP algorithm for SFVS parameterized by mim-width.

By a unified algorithm, this solves both problems in polynomial-time on the following graph classes: INTERVAL, PERMUTATION, and BI-INTERVAL graphs, CIRCULAR ARC and CIRCULAR PERMUTATION graphs, CONVEX graphs, $k$-POLYGON, DILWORTH-$k$ and CO-$k$-DEGENERATE graphs for fixed $k$; and also on LEAF POWER graphs if a leaf root is given as input, on $H$-GRAPHS for fixed $H$ if an $H$-representation is given as input, and on arbitrary powers of graphs in all the above classes. Prior to our results, only SFVS was known to be tractable restricted only on INTERVAL and PERMUTATION graphs, whereas all other results are new.

Keywords: Subset feedback vertex set, node multiway cut, neighbor equivalence, rank-width, mim-width.

1 Introduction

Given a vertex-weighted graph $G$ and a set $S$ of its vertices, the SUBSET FEEDBACK VERTEX SET (SFVS) problem asks for a vertex set of minimum weight that intersects all cycles containing a vertex of $S$. SFVS was introduced by Even et al. [17] who proposed an 8-approximation algorithm. Cygan et al. [15] and Kawarabayashi and Kobayashi [30] independently showed that SFVS is fixed-parameter tractable (FPT) parameterized by the solution size, while Hols and
Kratsch [22] provide a randomized polynomial kernel for the problem. As a generalization of the classical NP-complete Feedback Vertex Set (FVS) problem, for which $S = V(G)$, there has been a considerable amount of work to obtain faster algorithms for SFVS, both for general graphs [10, 18] where the current best is an $O^*(1.864^n)$ algorithm due to Fomin et al. [20], and restricted to special graph classes [2, 23, 38, 37]. Naturally, FVS and SFVS differ in complexity, as exemplified by split graphs where FVS is polynomial-time solvable [11] whereas SFVS remains NP-hard [20]. Moreover, note that the vertex-weighted variation of SFVS behaves differently than the unweighted one, as exposed on graphs with bounded independent set sizes: weighted SFVS is NP-complete on graphs with independent set size at most four, whereas unweighted SFVS is in XP parameterized by the independent set size [38].

Closely related to SFVS is the NP-hard Node Multiway Cut (NMC) problem in which we are given a vertex-weighted graph $G$ and a set $T$ of (terminal) vertices, and asked to find a vertex set of minimum weight that disconnects all the terminals [8, 21]. NMC is a well-studied problem in terms of approximation [21], as well as parameterized algorithms [3, 9, 10, 14, 16, 20]. It is not difficult to see that SFVS for $S = \{v\}$ coincides with NMC in which $T = N(v)$. In fact, NMC reduces to SFVS by adding a single vertex $v$ with a large weight that is adjacent to all terminals and setting $S = \{v\}$ [20]. Thus, in order to solve NMC on a given graph one may apply a known algorithm for SFVS on a vertex-extended graph. Observe, however, that through such an approach one needs to clarify that the vertex-extended graph still obeys the necessary properties of the known algorithm for SFVS. This explains why most of the positive results on SFVS on graph families [23, 37, 38] can not be translated to NMC.

In this paper, we investigate the complexity of SFVS and NMC when parameterized by structural graph width parameters. Well-known graph width parameters include tree-width [4], clique-width [13], rank-width [33], and maximum induced matching width (a.k.a. mim-width) [11]. These are of varying strength, with tree-width of modeling power strictly weaker than clique-width, as it is bounded on a proper subset of the graph classes having bounded clique-width, with rank-width and clique-width of the same modeling power, and with mim-width much stronger than clique-width. Belmonte and Vatshelle [1] showed that several graph classes, like interval graphs and permutation graphs, have bounded mim-width and a decomposition witnessing this can be found in polynomial time, whereas it is known that the clique-width of such graphs can be proportional to the square root of the number of vertices [24]. In this way, an XP algorithm parameterized by mim-width has the feature of unifying several algorithms on well-known graph classes.

We obtain most of these parameters through the well-known notion of branch-decomposition introduced in [39]. This is a natural hierarchical clustering of $G$, represented as a subcubic tree $T$ with the vertices of $G$ at its leaves. Any edge of the tree defines a cut of $G$ given by the leaves of the two subtrees that result from removing the edge from $T$. Judiciously choosing a cut-function to measure the complexity of such cuts, or rather of the bipartite subgraphs of $G$ given by the edges crossing the cuts, this framework then defines a graph width parameter by a minmax relation, minimum over all trees and maximum over all its cuts. Several graph width parameters have been defined this way, like carving-width, maximum matching-width, boolean-width etc. We will in this paper focus on: (i) rank-width [33] whose cut function is the $GF[2]$-rank of the adjacency matrix, (ii) $\mathbb{Q}$-rank-width [34] a variant of rank-width with interesting algorithmic properties which instead uses the rank over the rational field, and (iii) mim-width [11] whose cut function is the size of a maximum induced matching of the graph crossing the cut. Concerning their computations, for rank-width and $\mathbb{Q}$-rank-width, there are $2^{2k \cdot n^4}$ time algorithms that, given a graph $G$ as input and $k \in \mathbb{N}$, either output a decomposition for $G$ of width at most $3k + 1$ or confirms that the width of $G$ is more than $k$ [33, 36]. However, it is not known whether the mim-width of a graph can be approximated within a constant factor.
in time $n^{O(k)}$ for some function $f$.

Let us mention what is known regarding the complexity of NMC and SFVS parameterized by these width measures. Since these problems can be expressed in MSO$_1$-logic it follows that they are FPT parameterized by tree-width, clique-width, rank-width or Q-rank-width \[12, 34\], however the runtime will contain a tower of 2’s with more than 4 levels. Recently, Bergougnoux et al. \[2\] proposed $k^{O(k)} \cdot n^3$ time algorithms for these two problems parameterized by treewidth and proved that they cannot be solved in time $k^{o(k)} \cdot n^{O(1)}$ unless ETH fails. For mim-width, we know that FVS and thus SFVS are both \$W[1]\$-hard when parameterized by the mim-width of a given decomposition \[29\].

Attacking SFVS seems to be a hard task that requires more tools than for FVS. Even for very small values of mim-width that capture several graph classes, the tractability of SFVS, prior to our result, was left open besides interval and permutation graphs \[37\]. Although FVS was known to be tractable on such graphs for more than a decade \[32\], the complexity status of SFVS still remained unknown.

Our results. We design a meta-algorithm that, given a graph and a branch-decomposition, solves SFVS (or NMC via its reduction to SFVS). The runtime of this algorithm is upper bounded by $2^{O(rw^3)} \cdot n^4$, $2^{O(q^2 \log(q))} \cdot n^4$ and $n^{O(k^2)}$ where $rw$, $q$ and $k$ are the rank-width, the Q-rank-width and the mim-width of the given branch-decomposition. For clique-width, our meta-algorithm implies that we can solve SFVS and NMC in time $2^{O(k^2)} \cdot n^{O(1)}$ where $k$ is the clique-width of a given clique-width expression. However, we do not prove this as it is not asymptotically optimal, indeed Jacob et al. \[26\] show recently that SFVS and NMC is solvable in time $2^{O(k \log k)} \cdot n$ given a clique-width expression.

We resolve in the affirmative the question raised by Jaffke et al. \[29\], also mentioned in \[37\] and \[38\], asking whether there is an XP-time algorithm for SFVS parameterized by the mim-width of a given decomposition. For rank-width and Q-rank-width we provide the first explicit FPT-algorithms with low exponential dependency that avoid the MSO$_1$ formulation. Our main results are summarized in the following theorem:

**Theorem 1.** Let $G$ be a graph on $n$ vertices. We can solve Subset Feedback Vertex Set and Node Multiway Cut in time $2^{O(rw^3)} \cdot n^4$ and $2^{O(q^2 \log(q))} \cdot n^4$, where $rw$ and $q$ are the rank-width and the Q-rank-width of $G$, respectively. Moreover, if a branch-decomposition of mim-width $k$ for $G$ is given as input, we can solve Subset Feedback Vertex Set and Node Multiway Cut in time $n^{O(k^2)}$.

Note it is not known whether the mim-width of a graph can be approximated within a constant factor in time $n^{f(k)}$ for some function $f$. However, by the previously mentioned results of Belmonte and Vatshelle \[1\] on computing decompositions of bounded mim-width, combined with a result of \[27\] showing that for any positive integer $r$ a decomposition of mim-width $k$ of a graph $G$ is also a decomposition of mim-width at most $2k$ of its power $G^r$, we get the following corollary.

**Corollary 2.** We can solve Subset Feedback Vertex Set and Node Multiway Cut in polynomial time on Interval, Permutation, and Bi-Interval graphs, Circular Arc and Circular Permutation graphs, Convex graphs, k-Polygon, Dilworth-k and Co-k-Degenerate graphs for fixed $k$, and on arbitrary powers of graphs in any of these classes.

Previously, such polynomial-time tractability was known only for SFVS and only on Interval and Permutation graphs \[37\]. It is worth noticing that Theorem 1 implies also that we
can solve Subset Feedback Vertex Set and Node Multiway Cut in polynomial time on Leaf Power if an intersection model is given as input (from which we can compute a decomposition of mim-width 1) \cite{11, 27} and on H-Graphs for a fixed H if an H-representation is given as input (from which we can compute a decomposition of mim-width \(2|E(H)|\)) \cite{19}.

Our approach. We give some intuition to our meta-algorithm, that will focus on Subset Feedback Vertex Set. Since NMC can be solved by adding a vertex \(v\) of large weight adjacent to all terminals and solving SFVS with \(S = \{v\}\), all within the same runtime as extending the given branch-decomposition to this new graph increases the width at most by one for all considered width measures.

Towards achieving our goal, we use the \(d\)-neighbor equivalence, with \(d = 1\) and \(d = 2\), a notion introduced by Bui-Xuan et al. \cite{7}. Two subsets \(X\) and \(Y\) of \(A \subseteq V(G)\) are \(d\)-neighbor equivalent w.r.t. \(A\), if \(\min(d, |X \cap N(u)|) = \min(d, |Y \cap N(u)|)\) for all \(u \in V(G) \setminus A\). For a cut \((A, \overline{A})\) this equivalence relation on subsets of vertices was used by Bui-Xuan et al. \cite{7} to design a meta-algorithm, also giving XP algorithms by mim-width, for so-called \((\sigma, \rho)\) generalized domination problems. Recently, Boughougnoux and Kanté \cite{3} extended the uses of this notion to acyclic and connected variants of \((\sigma, \rho)\) generalized domination and similar problems like FVS. An earlier XP algorithm for FVS parameterized by mim-width had been given by Jaffke et al. \cite{29} but instead of the \(d\)-neighbor equivalences this algorithm was based on the notions of reduced forests and minimal vertex covers.

Our meta-algorithm does a bottom-up traversal of a given branch-decomposition of the input graph \(G\), computing a vertex subset \(X\) of maximum weight that induces an \(S\)-forest (i.e., a graph where no cycle contains a vertex of \(S\)) and outputs \(V(G) \setminus X\) which is necessarily a solution of SFVS. As usual, our dynamic programming algorithm relies on a notion of representativity between sets of partial solutions. For each cut \((A, \overline{A})\) induced by the decomposition, our algorithm computes a set of partial solutions \(A \subseteq 2^G\) of small size that represents \(2^A\). We say that a set of partial solutions \(A \subseteq 2^A\) represents a set of partial solutions \(B \subseteq 2^A\), if, for each \(Y \subseteq \overline{A}\), we have \(\text{best}(A, Y) = \text{best}(B, Y)\) where \(\text{best}(C, Y)\) is the maximum weight of a set \(X \in C\) such that \(X \cup Y\) induces an \(S\)-forest. Since the root of the decomposition is associated with the cut \((V(G), \emptyset)\), the set of partial solutions computed for this cut represents \(2^{V(G)}\) and thus contains an \(S\)-forest of maximum weight. Our main tool is a subroutine that, given a set of partial solutions \(B \subseteq 2^A\), outputs a subset \(A \subseteq B\) of small size that represents \(B\).

To design this subroutine, we cannot use directly the approaches solving FVS of any earlier approaches, like \cite{3} or \cite{29}. This is due to the fact that \(S\)-forests behave quite differently than forests; for example, given an \(S\)-forest \(F\), the graph induced by the edges between \(A \cap V(F)\) and \(\overline{A} \cap V(F)\) could be a biclique. Instead, we introduce a notion of vertex contractions and prove that, for every \(X \subseteq A\) and \(Y \subseteq \overline{A}\), the graph induced by \(X \cup Y\) is an \(S\)-forest if and only if there exists a partition of \(X \setminus S\) and of \(Y \setminus S\), satisfying certain properties, such that contracting the blocks of these partitions into single vertices transforms the \(S\)-forest into a forest.

This equivalence between \(S\)-forests in the original graph and forests in the contracted graphs allows us to adapt some ideas from \cite{3} and \cite{29}. Most of all, we use the property that, if the mim-width of the given decomposition is \(mim\), then the contracted graph obtained from the bipartite graph induced by \(X\) and \(Y\) admits a vertex cover \(VC\) of size at most \(4mim\). Note however, that in our case the elements of \(VC\) are contracted subsets of vertices. Such a vertex cover allows us to control the cycles which are crossing the cut.

We associate each possible vertex cover \(VC\) with an index \(i\) which contains basically a representative for the 2-neighbor equivalence for each subset of vertices in \(VC\). Moreover, for each index \(i\), we introduce the notions of partial solutions and complements solutions associated with \(i\) which correspond, respectively, to subsets of \(X \subseteq A\) and subsets \(Y \subseteq \overline{A}\) such that, for some
contractions of $X$ and $Y$, the contracted graph obtained from the bipartite graph induced by $X$ and $Y$ admits a vertex cover $VC$ associated with $i$. We define an equivalence relation $\sim_i$ between the partial solutions associated with $i$ such that $X \sim_i W$, if $X$ and $W$ connect in the same way the representatives of the vertex sets which belongs to the vertex covers described by $i$. Given a set of partial solutions $B \subseteq 2^A$, our subroutine outputs a set $A$ that contains, for each index $i$ and each equivalence class $C$ of $\sim_i$ over $B$, a partial solution in $C$ of maximum weight. In order to prove that $A$ represents $B$, we show that:

- for every $S$-forest $F$, there exists an index $i$ such that $V(F) \cap A$ is a partial solution associated with $i$ and $V(F) \cap \overline{A}$ is a complement solutions associated with $i$.
- if $X \sim_i W$, then, for every complement solution $Y$ associated with $i$, the graph induced by $X \cup Y$ is an $S$-forest if and only if $W \cup Y$ induces an $S$-forest.

The number of indices $i$ is upper bounded by $2^{O(q^2 \log(q))}$, $O(rw^3)$ and $n^{O(mim^2)}$. This follows from the known upper-bounds on the number of 2-neighbor equivalence classes and the fact that the vertex covers we consider have size at most $4mim$. Since there are at most $(4mim)^{4mim}$ ways of connecting $4mim$ vertices and $rw,q \geq mim$, we deduce that the size of $A$ is upper bounded by $2^{O(q^2 \log(q))}$, $2^{O(rw^3)}$ and $n^{O(mim^2)}$.

To the best of our knowledge, this is the first time a dynamic programming algorithm parameterized by graph width measures uses this notion of vertex contractions. Note that in contrast to the meta-algorithms in [3, 7], the number of representatives (for the $d$-neighbor equivalence) contained in the indices of our meta-algorithm are not upper bounded by a constant but by $4mim$. This explains the differences between the runtimes in Theorem [1] and those obtained in [3, 7], i.e. $n^{O(mim^2)}$ versus $n^{O(mim)}$. However, for the case $S = V(G)$, thus solving FVS, our meta-algorithm will have runtime $n^{O(mim)}$, as the algorithms for FVS of [3, 29]. We do not expect that SFVS can be solved as fast as FVS when parameterized by graph width measures. In fact, we know that it is not the case for tree-width as FVS can be solved in $2^{O(k)} \cdot n$ [3] but SFVS cannot be solved in $k^{O(k)} \cdot n^{O(1)}$ unless ETH fails [2].

## 2 Preliminaries

The size of a set $V$ is denoted by $|V|$ and its power set is denoted by $2^V$. We write $A \setminus B$ for the set difference of $A$ from $B$. We let $\min(\emptyset) = +\infty$ and $\max(\emptyset) = -\infty$.

**Graphs** The vertex set of a graph $G$ is denoted by $V(G)$ and its edge set by $E(G)$. An edge between two vertices $x$ and $y$ is denoted by $xy$ (or $yx$). Given $S \subseteq 2^{V(G)}$, we denote by $V(S)$ the set $\bigcup_{S \in S} S$. For a vertex set $U \subseteq V(G)$, we denote by $\overline{U}$ the set $V(G) \setminus U$. The set of vertices that are adjacent to $x$ is denoted by $N_G(x)$, and for $U \subseteq V(G)$, we let $N_G(U) = (\cup_{v \in U} N_G(v)) \setminus U$.

The subgraph of $G$ induced by a subset $X$ of its vertex set is denoted by $G[X]$. For two disjoint subsets $X$ and $Y$ of $V(G)$, we denote by $G[X,Y]$ the bipartite graph with vertex set $X \cup Y$ and edge set $\{xy \in E(G) \mid x \in X \text{ and } y \in Y\}$. We denote by $M_{XY}$ the adjacency matrix between $X$ and $Y$, i.e., the $(X,Y)$-matrix such that $M_{XY}[x,y] = 1$ if $y \in N(x)$ and 0 otherwise. A *vertex cover* of a graph $G$ is a set of vertices $VC \subseteq V(G)$ such that, for every edge $uv \in E(G)$, we have $u \in VC$ or $v \in VC$. A *matching* is a set of edges having no common endpoint and an *induced matching* is a matching $M$ of edges such that $G[V(M)]$ has no other edges besides $M$. The size of an induced matching $M$ refers to the number of edges in $M$.

For a graph $G$, we denote by $cc_G(X)$ the partition $\{C \subseteq V(G) \mid G[C]$ is a connected component of $G[X]\}$. We will omit the subscript $G$ of the neighborhood and components notations whenever there is no ambiguity.
For two graphs $G_1$ and $G_2$, we denote by $G_1 - G_2$ the graph $(V(G_1), E(G_1) \setminus E(G_2))$.

Given a graph $G$ and $S \subseteq V(G)$, we say that a cycle of $G$ is an $S$-cycle if it contains a vertex in $S$. Moreover, we say that a subgraph $F$ of $G$ is an $S$-forest if $F$ does not contain an $S$-cycle. Typically, the SUBSET FEEDBACK VERTEX SET problem asks for a vertex set of minimum (weight) size such that its removal results in an $S$-forest. Here we focus on the following equivalent formulation:

**SUBSET FEEDBACK VERTEX SET (SFVS)**

**Input:** A graph $G$, $S \subseteq V(G)$ and a weight function $w : V(G) \to \mathbb{Q}$.

**Output:** The maximum among the weights of the $S$-forests of $G$.

### Rooted Layout

For the notion of branch-decomposition, we consider its rooted variant called **rooted layout**. A rooted binary tree is a binary tree with a distinguished vertex called the **root**. Since we manipulate at the same time graphs and trees representing them, the vertices of trees will be called **nodes**.

A rooted layout of $G$ is a pair $(T, \delta)$ of a rooted binary tree $T$ and a bijective function $\delta$ between $V(G)$ and the leaves of $T$. For each node $x$ of $T$, let $L_x$ be the set of all the leaves $l$ of $T$ such that the path from the root of $T$ to $l$ contains $x$. We denote by $V_x$ the set of vertices that are in bijection with $L_x$, i.e., $V_x := \{ v \in V(G) \mid \delta(v) \in L_x \}$.

All the width measures dealt with in this paper are special cases of the following one, where the difference in each case is the used set function. Given a set function $f : 2^V(G) \to \mathbb{N}$ and a rooted layout $L = (T, \delta)$, the $f$-width of a node $x$ of $T$ is $f(V_x)$ and the $f$-width of $(T, \delta)$, denoted by $f(T, \delta)$ (or $f(L)$), is $\max\{f(V_x) \mid x \in V(T)\}$. Finally, the $f$-width of $G$ is the minimum $f$-width over all rooted layouts of $G$.

**(Q)-Rank-width** The rank-width and Q-rank-width are, respectively, the $rw$-width and $rwQ$-width where $rw(A)$ (resp. $rwQ(A)$) is the rank over $GF(2)$ (resp. $Q$) of the matrix $M_{A, \overline{A}}$ for all $A \subseteq V(G)$.

**Mim-width** The mim-width of a graph $G$ is the mim-width of $G$ where $\text{mim}(A)$ is the size of a maximum induced matching of the graph $G[A, \overline{A}]$ for all $A \subseteq V(G)$.

Observe that all three parameters $rw$, $rwQ$, and mim-width are symmetric, i.e., for the associated set function $f$ and for any $A \subseteq V(G)$, we have $f(A) = f(\overline{A})$. The following lemma provides upper bounds between mim-width and the other two parameters.

**Lemma 3 ([11]).** Let $G$ be a graph. For every $A \subseteq V(G)$, we have $\text{mim}(A) \leq \text{rw}(A)$ and $\text{rw}(A) \leq \text{rwQ}(A)$.

**Proof.** Let $A \subseteq V(G)$. Let $S$ be the vertex set of a maximum induced matching of the graph $G[A, \overline{A}]$. By definition, we have $\text{mim}(A) = |S \cap A| = |S \cap \overline{A}|$. Observe that the restriction of the matrix $M_{A, \overline{A}}$ to rows in $S \cap A$ and columns in $S \cap \overline{A}$ is a permutation matrix: a binary square matrix with exactly one entry of 1 in each row and each column. The rank of this permutation matrix over $GF[2]$ or $Q$ is $|S \cap A| = \text{mim}(A)$. Hence, $\text{mim}(A)$ is upper bounded both by $\text{rw}(A)$ and $\text{rwQ}(A)$. \qed

6
**d-neighbor-equivalence.** The following concepts were introduced in \[7]. Let \( G \) be a graph. Let \( A \subseteq V(G) \) and \( d \in \mathbb{N}^+ \). Two subsets \( X \) and \( Y \) of \( A \) are *d-neighbor equivalent w.r.t. \( A \), denoted by \( X \equiv_{d}^{A} Y \), if \( \min(d, |X \cap N(u)|) = \min(d, |Y \cap N(u)|) \) for all \( u \in \overline{A} \). It is not hard to check that \( \equiv_{d}^{A} \) is an equivalence relation. See Figure 1 for an example of 2-neighbor equivalent sets.

![Figure 1: We have \( X \equiv_{2}^{A} Y \), but it is not the case that \( X \equiv_{3}^{A} Y \).](image)

For all \( d \in \mathbb{N}^+ \), we let \( \text{necc}_{d} : 2^{V(G)} \to \mathbb{N} \) where for all \( A \subseteq V(G) \), \( \text{necc}_{d}(A) \) is the number of equivalence classes of \( \equiv_{d}^{A} \). Notice that \( \text{necc}_{1} \) is a symmetric function \[31,\, \text{Theorem 1.2.3}\] but \( \text{necc}_{d} \) is not necessarily symmetric for \( d \geq 2 \). To simplify the running times, we will use the shorthand \( s\cdot \text{necc}_{2}(A) \) to denote \( \max(\text{necc}_{2}(A), \text{necc}_{2}(\overline{A})) \) (where \( s \) stands for symmetric). The following lemma shows how \( \text{necc}_{d}(A) \) is upper bounded by the other parameters.

**Lemma 4** (\[11, \ 35, \ 41\]). Let \( G \) be a graph. For every \( A \subseteq V(G) \) and \( d \in \mathbb{N}^+ \), we have the following upper bounds on \( \text{necc}_{d}(A) \):

\[
\begin{align*}
(a) \quad & 2^{\text{draw}(A)^2}, \\
(b) \quad & 2^{\text{rw}_0(A) \log(\text{draw}(A)) + 1}, \\
(c) \quad & |A|^{d \text{min}(A)}.
\end{align*}
\]

In order to manipulate the equivalence classes of \( \equiv_{d}^{A} \), one needs to compute a representative for each equivalence class in polynomial time. This is achieved with the following notion of a representative. Let \( G \) be a graph with an arbitrary ordering of \( V(G) \) and let \( A \subseteq V(G) \). For each \( X \subseteq A \), let us denote by \( \text{rep}_{d}^{A}(X) \) the lexicographically smallest set \( R \subseteq A \) such that \( |R| \) is minimized and \( R \equiv_{d}^{A} X \). Moreover, we denote by \( R_{d}^{A} \) the set \( \{ \text{rep}_{d}^{A}(X) \mid X \subseteq A \} \). It is worth noticing that the empty set always belongs to \( R_{d}^{A} \), for all \( A \subseteq V(G) \) and \( d \in \mathbb{N}^+ \). Moreover, we have \( R_{d}^{V(G)} = R_{d}^{\overline{A}} = \{ \emptyset \} \) for all \( d \in \mathbb{N}^+ \). In order to compute these representatives, we use the following lemma.

**Lemma 5** (\[11\]). Let \( G \) be an \( n \)-vertex graph. For every \( A \subseteq V(G) \) and \( d \in \mathbb{N}^+ \), one can compute in time \( O(\text{necc}_{d}(A) \cdot n^2 \cdot \log(\text{necc}_{d}(A))) \), the sets \( R_{d}^{A} \), and a data structure that, given a set \( X \subseteq A \), computes \( \text{rep}_{d}^{A}(X) \) in time \( O(|A| \cdot n \cdot \log(\text{necc}_{d}(A))) \).

**Vertex Contractions** In order to deal with SFVS, we will use the ideas of the algorithms for *Feedback Vertex Set* from \[3,\ 28\]. To this end, we will contract subsets of \( S \) in order to transform \( S \)-forests into forests.

In order to compare two partial solutions associated with \( A \subseteq V(G) \), we define an auxiliary graph in which we replace contracted vertices by their representative sets in \( R_{d}^{A} \). Since the sets in \( R_{d}^{A} \) are not necessarily pairwise disjoint, we will use the following notions of graphs “induced” by collections of subsets of vertices. We will also use these notions to define the contractions we make on partial solutions.

Let \( G \) be a graph. Given \( A \subseteq 2^{V(G)} \), we define \( G[A] \) as the graph with vertex set \( A \) where \( A, B \in A \) are adjacent if and only if \( N(A) \cap B \neq \emptyset \). Observe that if the sets in \( A \) are pairwise
disjoint, then $G[A]$ is obtained from an induced subgraph of $G$ by vertex contractions (i.e., by replacing two vertices $u$ and $v$ with a new vertex with neighborhood $N((u, v)))$ and, for this reason, we refer to $G[A]$ as a contracted graph. Notice that we will never use the neighborhood notation and connected component notations on contracted graphs. Given $A, B \subseteq 2^{|V(G)|}$, we denote by $G[A, B]$ the bipartite graph with vertex set $A \cup B$ and where $A, B \subseteq A \cup B$ are adjacent if and only if $A \in A, B \in B$, and $N(A) \cap B \neq \emptyset$. Moreover, we denote by $G[A \mid B]$ the graph with vertex set $A \cup B$ and with edge set $E(G[A]) \cup E(G[A, B])$. Observe that both graphs $G[A, B]$ and $G[A \mid B]$ are subgraphs of the contracted graph $G[A \cup B]$. To avoid confusion with the original graph, we refer to the vertices of the contracted graphs as blocks. It is worth noticing that in the contracted graphs used in this paper, whenever two blocks are adjacent, they are disjoint.

The following observation states that we can contract from a partition without increasing the size of a maximum induced matching of a graph. It follows directly from the definition of contractions.

**Observation 6.** Let $H$ be a graph. For any partition $P$ of a subset of $V(H)$, the size of a maximum induced matching of $H \mid P$ is at most the size of a maximum induced matching of $H$.

Let $(G, S)$ be an instance of SFVS. The vertex contractions that we use on a partial solution $X$ are defined from a given partition of $X \setminus S$. A partition of the vertices of $X \setminus S$ is called an $S$-contraction of $X$. We will use the following notations to handle these contractions.

Given $Y \subseteq V(G)$, we denote by $\{1\}$ the partition of $Y$ which contains only singletons, i.e., $\{1\} = \{\{v\} \mid v \in Y\}$. Moreover, for an $S$-contraction $P$ of $X$, we denote by $X_{\mid P}$ the partition of $X$ where $X_{\mid P} = P \cup (X_{\setminus S})$. Given a subgraph $H$ of $G$ and an $S$-contraction $P$ of $V(H)$, we denote by $H_{\mid P}$ the graph $H[V(H)_{\mid P}]$. For example, given two $S$-contractions $P_X, P_Y$ of two disjoint subsets $X, Y$ of $V(G)$, we denote the graph $G[X_{\mid P_X}, Y_{\mid P_Y}]$ by $G[X, Y]_{\mid P_X \cup P_Y}$ and the graph $G[X_{\mid P_X}, Y_{\mid P_Y}]$ by $G[X]_{\mid P_X \cup P_Y}$. It is worth noticing that in our contracted graphs, all the blocks of $S$-vertices are singletons and we denote them by $\{v\}$.

Given a set $X \subseteq V(G)$, we will intensively use the graph $G[X]_{\mid \text{cc}(X \setminus S)}$ which corresponds to the graph obtained from $G[X]$ by contracting the connected components of $G[X \setminus S]$, see Figure 2.

![Figure 2: An S-forest induced by a set $X \subseteq V(G)$, the vertices of $S$ are white. The gray circles represent the blocks of $X_{\mid \text{cc}(X \setminus S)}$. The graph $G[X]_{\mid \text{cc}(X \setminus S)}$ is obtained by contracting each gray circle.](image)

Observe that, for every subset $X \subseteq V(G)$, if $G[X]$ is an $S$-forest, then $G[X]_{\mid \text{cc}(X \setminus S)}$ is a forest. The converse is not true as we may delete $S$-cycles with contractions: take a triangle with one vertex $v$ in $S$ and contract the neighbors of $v$. However, we can prove the following equivalence.

**Fact 7.** Let $G$ be a graph and $S \subseteq V(G)$. For every $X \subseteq V(G)$, $G[X]$ is an $S$-forest if and only if there exists an $S$-contraction $P$ of $X$ satisfying the following two properties:
• $G[X] \downarrow \mathcal{P}$ is a forest, and
• for every $B \in \mathcal{P}$ and $v \in X \cap S$, we have $|N(v) \cap B| \leq 1$.

Moreover, if $G[X]$ is an $S$-forest, then the $\mathcal{S}$-contraction $cc(X \setminus S)$ satisfies these two properties.

Proof. ($\Rightarrow$) Suppose first that $G[X]$ is an $S$-forest. We claim that the $\mathcal{S}$-contraction $cc(X \setminus S)$ satisfies the two properties. Assume towards a contradiction that there is a cycle $C$ in $G[X]_{\downarrow cc(X \setminus S)}$. By definition of $G[X]_{\downarrow cc(X \setminus S)}$, the blocks of this graph are the connected components $cc(X \setminus S)$ and the singletons in $\binom{X \setminus S}{1}$. The blocks in $cc(X \setminus S)$ are pairwise non-adjacent, thus $C$ contains a block $\{s\}$ with $s \in X \cap S$. Observe that for every pair of consecutive blocks $B_1,B_2$ of $C$, there exists a vertex $v_1 \in B_1$ and $v_2 \in B_2$ such that $v_1v_2 \in E(G[X])$. As every block of $G[X]_{\downarrow cc(X \setminus S)}$ induces a connected component in $G$, we can construct an $S$-cycle in $G[X]$ by replacing every block of $C$ by a path in $G[X]$, yielding a contradiction. Hence, $G[X]_{\downarrow cc(X \setminus S)}$ is a forest, i.e. the first property is satisfied. Observe that if there exists $C \in cc(X \setminus S)$ and $v \in X \cap S$ such that $v$ has two neighbors in $C$, then there exists an $S$-cycle in $G[X]$ since $C$ is a connected component. Hence, $cc(X \setminus S)$ satisfies the second property.

($\Leftarrow$) Let $\mathcal{P}$ be a $\mathcal{S}$-contraction of a subset $X \subseteq V(G)$ that satisfies the two properties. Assume for contradiction that there is an $S$-cycle $C$ in $G[X]$. Let $v$ be a vertex of $C$ that belongs to $S$ and let $u$ and $w$ be the neighbors of $v$ in $C$.

Let $B_u$ and $B_w$ be the blocks in $(X \cup Y)_{\downarrow \mathcal{P}}$ that contain $u$ and $w$ respectively. As $v$ is in $S$, it belongs to the block $\{v\}$ of $G[X \cup Y]_{\downarrow \mathcal{P}}$ and thus it is not contained in $B_u$ nor $B_w$. The second property implies that $|N(u) \cap B| \leq 1$ for each $B \in \mathcal{P}$. Thus, $B_u$ and $B_w$ are two distinct blocks both connected to the block $\{v\}$. Since there exists a path between $u$ and $w$ in $C$ that does not go through $v$, we deduce that there is a path between $B_u$ and $B_w$ in $G[X]_{\downarrow \mathcal{P}}$ that does not go through $\{v\}$. Indeed, this follows from the fact that if there is an edge between two vertices $a$ and $b$ in $G[X]$, then either $a$ and $b$ belong to the same block of $G[X]_{\downarrow \mathcal{P}}$ or there exists an edge between the blocks in $G[X]_{\downarrow \mathcal{P}}$ which contain $a$ and $b$. We conclude that there exists a cycle in $G[X]_{\downarrow \mathcal{P}}$, a contradiction with the first property.

3 A Meta-Algorithm for Subset Feedback Vertex Set

In the following, we present a meta-algorithm that, given a rooted layout $(T, \delta)$ of $G$, solves SFVS. We will show that this meta-algorithm will imply that SFVS can be solved in time $2^{O(mn)} \cdot n^4 \cdot 2^{O(mn^2)} \cdot n^4$ and $n^{O(min(T,\delta))}$. The main idea of this algorithm is to use $\mathcal{S}$-contractions in order to employ similar properties of the algorithm for MAXIMUM INDUCED TREE of [3] and the $n^{O(min(T,\delta))}$ time algorithm for FEEDBACK VERTEX SET of [28]. In particular, we use the following lemma which is proved implicitly in [3]. To simplify the following statements, we fix a graph $G$, a rooted layout $(T, \delta)$ of $G$ and a node $x \in V(T)$.

Lemma 8. Let $X$ and $Y$ be two disjoint subsets of $V(G)$. If $G[X \cup Y]$ is a forest, then the number of vertices of $X$ that have at least two neighbors in $Y$ is bounded by $2w$ where $w$ is the size of a maximum induced matching in the bipartite graph $G[X,Y]$.

Proof. Let $X^{2+}$ be the set of vertices in $X$ having at least 2 neighbors in $Y$. In the following, we prove that $F = G[X^{2+}, Y]$ admits a good bipartition, that is a bipartition $\{X_0, X_1\}$ of $X^{2+}$ such that, for each $i \in \{0,1\}$ and, for each $v \in X_i$, there exists $y_i \in Y$ such that $N_F(y_i) \cap X_i = \{v\}$. Observe that this is enough to prove the lemma since if $F$ admits a good bipartition $\{X_0, X_1\}$, then $|X_0| \leq w$ and $|X_1| \leq w$. Indeed, if $F$ admits a good bipartition $\{X_0, X_1\}$, then, for each $i \in \{0,1\}$, the set of edges $M_i = \{vy_i \mid v \in X_i\}$ is an induced matching of $G[X,Y]$. In order to
prove that \( F \) admits a good bipartition it is sufficient to prove that each connected component of \( F \) admits a good bipartition.

Let \( C \) be a connected component of \( F \) and \( x_0 \in C \cap X^{2+} \). As \( G[V \cup Y] \) is a forest, we deduce that \( F[C] \) is a tree. Observe that the distance in \( F[C] \) between each vertex \( v \in C \cap X^{2+} \) and \( x_0 \) is even because \( F[C] \) is bipartite w.r.t. \((C \cap X^{2+}, C \cap Y)\). Let \( X_0 \) (resp. \( X_1 \)) be the set of all vertices \( v \in C \cap X^{2+} \) at distance \( 2 \ell \) from \( x_0 \) with \( \ell \) even (resp. odd). We claim that \{\( X_0, X_1 \)\} is a good bipartition of \( F[C] \).

Let \( i \in \{0,1\} \), \( v \in X_i \) and \( \ell \in \mathbb{N} \) such that the distance between \( v \) and \( x_0 \) in \( F[C] \) is \( 2 \ell \). Let \( P \) be the set of vertices in \((C \setminus \{v\})\) that share a common neighbor with \( v \) in \( F[C] \). We want to prove that \( v \) has a neighbor \( y \) in \( F \) that is not adjacent to \( P \cap X_1 \). Observe that, for every \( v' \in P \), the distance between \( v' \) and \( x_0 \) in \( F[C] \) is either \( 2 \ell - 2 \), \( 2 \ell \) or \( 2 \ell + 2 \) because \( F[C] \) is a tree and the distance between \( v \) and \( x_0 \) is \( 2 \ell \). By construction of \{\( X_0, X_1 \)\}, every vertex at distance \( 2 \ell - 2 \) or \( 2 \ell + 2 \) from \( x_0 \) belongs to \( X_{1-i} \). Thus, every vertex in \( P \cap X_i \) is at distance \( 2 \ell \) from \( x_0 \). If \( \ell = 0 \), then we are done because \( v = x_0 \) and \( P \cap X_i = \emptyset \). Assume that \( \ell \neq 0 \). As \( F[C] \) is a tree, \( v \) has only one neighbor \( w \) at distance \( 2 \ell - 1 \) from \( x_0 \) in \( F[C] \). Because \( F[C] \) is a tree, \( w \) is the only vertex adjacent to \( v \) and the vertices in \( P \cap X_i \). By definition of \( X^{2+} \), \( v \) has at least two neighbors in \( Y \), so \( v \) admits a neighbor that is not \( w \) and this neighbor is not adjacent to the vertices in \( P \cap X_i \). Hence, we deduce that \{\( X_0, X_1 \)\} is a good bipartition of \( F[C] \).

We deduce that every connected component of \( F \) admits a good bipartition and, thus, \( F \) admits a good bipartition. This proves that \( |X^{2+}| \leq 2w \). \(\square\)

The following lemma generalizes Fact 7 and presents the equivalence between \( S \)-forests and forests that we will use in our algorithm.

**Lemma 9.** Let \( X \subseteq V_{S} \) and \( Y \subseteq \overline{V_{S}} \). If the graph \( G[V \cup Y] \) is an \( S \)-forest, then there exists an \( S \)-contraction \( \mathcal{P}_Y \) of \( Y \) that satisfies the following conditions:

1. \( G[X \cup Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \) is a forest,
2. for all block \( P \in \mathcal{cc}(X \setminus S) \cup \mathcal{P}_Y \) and \( v \in (X \cup Y) \cap S \), we have \( |N(v) \cap P| \leq 1 \),
3. the graph \( G[X,Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \) admits a vertex cover \( VC \) of size at most \( 4 \min(|V_{S}|) \) such that the neighborhoods of the blocks in \( VC \) are pairwise distinct in \( G[X,Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \).

**Proof.** Assume that \( G[X \cup Y] \) is an \( S \)-forest. Let us explain how we construct \( \mathcal{P}_Y \) that satisfies Conditions (1)-(3). First, we initialize \( \mathcal{P}_Y = \mathcal{cc}(Y \setminus S) \). Observe that there is no cycle in \( G[X \cup Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \) that contains a block in \( \binom{Y}{1} \) because \( G[X \cup Y] \) is an \( S \)-forest. Moreover, \( \mathcal{cc}(X \setminus S) \) and \( \mathcal{P}_Y \) form two independent sets in \( G[X \cup Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \). Consequently, for all the cycles \( C \) in \( G[X \cup Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \) we have \( C = (X_1, Y_1, X_2, Y_2, \ldots, X_t, Y_t) \) where \( X_1, \ldots, X_t \in \mathcal{cc}(X \setminus S) \) and \( Y_1, \ldots, Y_t \in \mathcal{P}_Y \). We do the following operation, until the graph \( G[X \cup Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \) is a forest: take a cycle \( C = (X_1, Y_1, X_2, Y_2, \ldots, X_t, Y_t) \) in \( G[X \cup Y]_{\mathcal{L}(X \setminus S) \cup \mathcal{P}_Y} \) and replace the blocks \( Y_1, \ldots, Y_t \) in \( \mathcal{P}_Y \) by the block \( Y_1 \cup \cdots \cup Y_t \). See Figures 5 and 6 in Appendix A for an example of \( S \)-contraction \( \mathcal{P}_Y \).

For each \( B \in \mathcal{cc}(X \setminus S) \cup \mathcal{cc}(Y \setminus S) \), the vertices of \( B \) are pairwise connected in \( G[(X \cup Y) \setminus S] \). We deduce by induction that whenever we apply the operation on a cycle \( C = (X_1, Y_1, X_2, Y_2, \ldots, X_t, Y_t) \), it holds that the vertices of the new block \( Y_1 \cup \cdots \cup Y_t \) are pairwise connected in \( G[(X \cup Y) \setminus S] \). Thus, for every block \( B \in \mathcal{cc}(X \setminus S) \cup \mathcal{P}_Y \), the vertices of \( B \) are pairwise connected in \( G[(X \cup Y) \setminus S] \). It follows that for every \( v \in (X \cup Y) \cap S \) and \( B \in \mathcal{cc}(X \setminus S) \cup \mathcal{P}_Y \), since \( G[X \cup Y] \) is an \( S \)-forest, we have \( |N(v) \cap B| \leq 1 \). Thus, Condition (2) is satisfied.
It remains to prove Condition (3). Let VC be the set of blocks of $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ containing:

- the blocks that have at least 2 neighbors in $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$, and
- one block in every isolated edge of $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$.

By construction, it is clear that VC is indeed a vertex cover of $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ as every edge is either isolated or incident to a block of degree at least 2. We claim that $|\text{VC}| \leq 4\text{mim}(V_x)$. By Observation 6, we know that the size of a maximum induced matching in $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ is at most $\text{mim}(V_x)$. Let $t$ be the number of isolated edges in $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$. Observe that the size of a maximum induced matching in the graph obtained from $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ by removing isolated edges is at most $\text{mim}(V_x) - t$. By Lemma 8 we know that there are at most $4(\text{mim}(V_x) - t)$ blocks that have at least 2 neighbors in $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$. We conclude that $|\text{VC}| \leq 4\text{mim}(V_x)$.

Since $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ is a forest, the neighborhoods of the blocks that have at least 2 neighbors must be pairwise distinct. We conclude from the construction of VC that the neighborhoods of the blocks of VC in $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ are pairwise distinct. Hence, Condition (3) is satisfied.

In the following, we will use Lemma 9 to design some sort of equivalence relation between partial solutions. To this purpose, we use the following set of tuples. We call each such tuple an index because it corresponds to an index into a table in a DP (dynamic programming) approach. We do this even though the presentation of our algorithm is not given by a standard DP description.

**Definition 10 ($I_x$).** We define the set $I_x$ of indices as the set of tuples

$$(x_{vc}^S, x_{vc}^S, x_{vc}^S, y_{vc}^S, y_{vc}^S) \in 2R_2^{V_x} \times 2R_1^{V_x} \times R_3^{V_x} \times 2R_2^{V_x} \times 2R_1^{V_x}$$

such that $|x_{vc}^S| + |x_{vc}^S| + |y_{vc}^S| + |y_{vc}^S| \leq 4\text{mim}(V_x)$.

These sets of indices play a major role in our meta-algorithm, in particular, the sizes of these sets of indices appear in the runtime of our meta-algorithm. In fact, to prove the algorithmic consequences of our meta-algorithm for rank-width, Q-rank-width and mim-width, we show (Lemma 23) that the size of $I_x$ is upper bounded by $2O(\text{rw}(V_x)^3), 2O(\text{rw}(V_x)^2 \log(\text{rw}(V_x)))$ and $nO(\text{mim}(V_x)^3)$.

In the following, we will define partial solutions associated with an index $i \in I_x$ (a partial solution may be associated with many indices). In order to prove the correctness of our algorithm (the algorithm itself will not use this concept), we will also define complement solutions (the sets $Y \subseteq \overline{V_x}$ and their $\overline{S}$-contractions $P_Y$) associated with an index $i$. We will prove that, for every partial solution $X$ and complement solution $(Y, P_Y)$ associated with $i$, if the graph $G[X \cup Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ is a forest, then $G[X \cup Y]$ is an S-forest.

Let us give some intuition on these indices by explaining how one index is associated with a solution, figures explaining this association and the purposes of the sets $x_{vc}^S$, $x_{vc}^S$, $x_{vc}^S$, $y_{vc}^S$, $y_{vc}^S$ can be found in Appendix A. Let $X \subseteq V_x$ and $Y \subseteq \overline{V_x}$ such that $G[X \cup Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ is an S-forest. Let $P_Y$ be the $\overline{S}$-contraction of $Y$ and VC be a vertex cover of $G[X, Y]_{\overline{cc}(X\setminus S) \cup P_Y}$ given by Lemma 9. Then, $X$ and $Y$ are associated with $i = (x_{vc}^S, x_{vc}^S, x_{vc}^S, y_{vc}^S, y_{vc}^S) \in I_x$ such that:

- $x_{vc}^S$ (resp. $y_{vc}^S$) contains the representatives of the blocks $\{v\}$ in VC such that $v \in X \cap S$ (resp. $v \in Y \cap S$) w.r.t. the 1-neighbor equivalence over $V_x$ (resp. $\overline{V_x}$). We will only use
the indices where $\mathcal{X}_c^S$ contains representatives of singletons, in other words, $\mathcal{X}_c^S$ is included in $\{\text{rep}_{\mathcal{V}}(\{v\}) \mid v \in V_x\}$ which can be much smaller than $\mathcal{R}_1^S$. The same observation holds for $\mathcal{Y}_c^S$. In Definition \ref{def:auxiliary}, we state that $\mathcal{X}_c^S$ and $\mathcal{Y}_c^S$ are, respectively, subsets of $2^{\mathcal{R}_1^S}$ and $2^{\mathcal{R}_S^V}$, for the sake of simplicity.

- $\mathcal{X}_c^S$ (resp. $\mathcal{Y}_c^S$) contains the representatives of the blocks in $\text{cc}(X \setminus S) \cap \text{VC}$ (resp. $\mathcal{P}_Y \cap \text{VC}$) w.r.t. the 2-neighbor equivalence relation over $V_x$ (resp. $V_x$).

- $\mathcal{X}_{\text{cc}}$ is the representative of $X \setminus V(\text{VC})$ (the set of vertices which do not belong to the vertex cover) w.r.t. the 1-neighbor equivalence over $V_x$.

Because the neighborhoods of the blocks in $\text{VC}$ are pairwise distinct in $G[X, Y]_{|\text{cc}(X \setminus S) \cup \mathcal{P}_Y}$ (Property (3) of Lemma \ref{lemma:neighborhood}), there is a one to one correspondence between the representatives in $\mathcal{X}_c^S \cup \mathcal{X}_c^S \cup \mathcal{Y}_c^S \cup \mathcal{Y}_c^S$ and the blocks in $\mathcal{VC}$.

While $\mathcal{X}_c^S, \mathcal{X}_c^S, \mathcal{Y}_c^S, \mathcal{Y}_c^S$ describe $\mathcal{VC}$, the representative set $\mathcal{X}_{\text{cc}}$ describes the neighborhood of the blocks of $\mathcal{X}_{\text{cc}}(X \setminus S)$ which are not in $\mathcal{VC}$. The purpose of $\mathcal{X}_{\text{cc}}$ is to make sure that, for every partial solution $X$ and complement solution $(Y, \mathcal{P}_Y)$ associated with $i$, the set $\mathcal{VC}$ described by $\mathcal{X}_c^S, \mathcal{X}_c^S, \mathcal{Y}_c^S, \mathcal{Y}_c^S$ is a vertex cover of $G[X, Y]_{|\text{cc}(X \setminus S) \cup \mathcal{P}_Y}$. For doing so, it is sufficient to require that $Y \setminus V(\text{VC})$ has no neighbor in $\mathcal{X}_{\text{cc}}$ for every complement solution $(Y, \mathcal{P}_Y)$ associated with $i$.

Observe that the sets $\mathcal{X}_c^S$ and $\mathcal{Y}_c^S$ contain representatives for the 2-neighbor equivalence. We need the 2-neighbor equivalence to control the $S$-cycles which might disappear after vertex contractions. To prevent this situation, we require, for example, that every vertex in $X \cap S$ has at most one neighbor in $\overline{R}$ for each $R \in \mathcal{Y}_c^S$. Thanks to the 2-neighbor equivalence, a vertex $v$ in $X \cap S$ has at most one neighbor in $\overline{R} \in \mathcal{Y}_c^S$ if and only if $v$ has at most one neighbor in the block of $\mathcal{P}_Y$ associated with $\overline{R}$. This property of the 2-neighbor equivalence is captured by the following fact.

**Fact 11.** For every $A \subseteq V(G)$ and $B, P \subseteq A$, if $B \equiv_A P$, then, for all $v \in A$, we have $|N(v) \cap B| \leq 1$ if and only if $|N(v) \cap P| \leq 1$.

In order to define partial solutions associated with $i$, we need the following notion of auxiliary graph. Given $X \subseteq V_x$ and $i = (\mathcal{X}_c^S, \mathcal{X}_c^S, \mathcal{X}_{\text{cc}}, \mathcal{Y}_c^S, \mathcal{Y}_c^S) \in \mathbb{I}_x$, we write $\text{aux}(X, i)$ to denote the graph

$$G[\mathcal{X}_{\text{cc}}(X \setminus S)|\mathcal{Y}_c^S \cup \mathcal{Y}_c^S].$$

Observe that $\text{aux}(X, i)$ is obtained from the graph induced by $\mathcal{X}_{\text{cc}}(X \setminus S) \cup \mathcal{Y}_c^S \cup \mathcal{Y}_c^S$ by removing the edges between the blocks from $\mathcal{Y}_c^S \cup \mathcal{Y}_c^S$. Figure \ref{fig:example} illustrates an example of the graph $\text{aux}(X, i)$ and the related notions. The figures in Appendix \ref{appendix} explain the relations between an $S$-forest and these auxiliary graphs.

We will ensure that, given a complement solution $(Y, \mathcal{P}_Y)$ associated with $i$, the graph $\text{aux}(X, i)$ is isomorphic to $G[\mathcal{X}_{\text{cc}}(X \setminus S)|\mathcal{Y}_c \cap \mathcal{P}_Y \cap \mathcal{VC}]$. We are now ready to define the notion of partial solution associated with an index $i$.

**Definition 12 (Partial solutions).** Let $i = (\mathcal{X}_c^S, \mathcal{X}_c^S, \mathcal{X}_{\text{cc}}, \mathcal{Y}_c^S, \mathcal{Y}_c^S) \in \mathbb{I}_x$. We say that $X \subseteq V_x$ is a partial solution associated with $i$ if the following conditions are satisfied:

(a) for every $R \in \mathcal{X}_c^S$, there exists a unique $v \in X \cap S$ such that $R \equiv_{V_x} \{v\}$,

(b) for every $R \in \mathcal{X}_c^S$, there exists a unique $C \in \text{cc}(X \setminus S)$ such that $R \equiv_{V_x} C$,

(c) $\text{aux}(X, i)$ is a forest.
Figure 3: An example of a set $X \subseteq V_x$ and its auxiliary graph associated with the index $i = (X_{vc}, X^{S}_{vc}, X_{vc}, Y_{vc}^{S}, Y_{vc}^{S})$. Here, $X_{vc}^{S} = \{R_{2}\}$ with $R_{2} = \text{rep}_{V_{x}}^{i}(X_{2})$, $X_{vc}^{S} = \{R_{1}\}$ with $R_{1} = \text{rep}_{V_{x}}^{i}(\{v_{1}\})$, $X_{vc}$ is the representative of the union of the circular blocks, $Y_{vc}^{S} = \{U_{1}, U_{2}\}$, and $Y_{vc}^{S} = \{U_{3}, U_{4}\}$.  The singletons in $X \cap S$ and $y_{vc}^{S}$ are white filled, whereas the square blocks are $\{v_{1}\}$, $X_{2}$ and the blocks in $Y_{vc}^{S} \cup Y_{vc}^{S}$.

(d) for every $C \in \text{cc}(X \setminus S)$ and $\{v\} \in y_{vc}^{S}$, we have $|N(v) \cap C| \leq 1$,

(e) for every $v \in X \cap S$ and $U \in Y_{vc}^{S} \cup \text{cc}(X \setminus S)$, we have $|N(v) \cap U| \leq 1$,

(f) $X_{vc}^{S} \equiv_{V_{x}}^{i} X \setminus V(\text{VC}_{X})$, where $\text{VC}_{X}$ contains the blocks $\{v\} \in (X \setminus S)$ such that $\text{rep}_{V_{x}}^{i}(\{v\}) \in X_{vc}^{S}$ and the components $C$ of $G[X \setminus S]$ such that $\text{rep}_{V_{x}}^{i}(C) \in X_{vc}^{S}$.

Similarly to Definition 12, we define the notion of complement solutions associated with an index $i \in I_{x}$. We use this concept only to prove the correctness of our algorithm.

**Definition 13 (Complement solutions).** Let $i = (X_{vc}^{S}, X_{vc}^{S}, X_{vc}, Y_{vc}^{S}, Y_{vc}^{S}) \in I_{x}$. We call complement solutions associated with $i$ all the pairs $(Y, \mathcal{P}_{Y})$ such that $Y \subseteq V_{x}$, $\mathcal{P}_{Y}$ is an $S$-contraction of $Y$ and the following conditions are satisfied:

(a) for every $U \in Y_{vc}^{S}$, there exists a unique $v \in Y \cap S$ such that $U \equiv_{V_{x}}^{i} \{v\}$,

(b) for every $U \in Y_{vc}^{S}$, there exists a unique $P \in \mathcal{P}_{Y}$ such that $U \equiv_{2}^{i} P$,

(c) $G[Y]|_{\mathcal{P}_{Y}}$ is a forest,

(d) for every $P \in \mathcal{P}_{Y}$ and $\{v\} \in X_{vc}^{S}$, we have $|N(v) \cap P| \leq 1$,

(e) for every $y \in Y \cap S$ and $R \in X_{vc}^{S} \cup \mathcal{P}_{Y}$, we have $|N(y) \cap R| \leq 1$,

(f) $N(X_{vc}) \cap V(Y_{vc}) = \emptyset$, where $Y_{vc}$ contains the blocks $\{v\} \in (Y \setminus S)$ such that $\text{rep}_{V_{x}}^{i}(\{v\}) \in Y_{vc}$ and the blocks $P \in \mathcal{P}_{Y}$ such that $\text{rep}_{V_{x}}^{i}(P) \notin Y_{vc}$.

Let us give some explanations on the conditions of Definitions 12 and 13. Let $X$ be a partial solution associated with an index $i \in I_{x}$ and $(Y, \mathcal{P}_{Y})$ be a complement solution associated with $i$. Conditions (a) and (b) of both definitions guarantee that there exists a subset $\text{VC}$ of $X_{\text{cc}(X \setminus S)} \cup Y_{1}\mathcal{P}_{Y}$ such that there is a one-to-one correspondence between the blocks of $\text{VC}$ and the representatives in $X_{vc}^{S} \cup X_{vc}^{S} \cup Y_{vc}^{S} \cup Y_{vc}^{S}$.  

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Condition (c) of Definition \[12\] guarantees that the connections between \(X_{\leq c}(X \setminus S)\) and VC are acyclic. As explained earlier, Conditions (d) and (e) of both definitions are here to control the \(S\)-cycles which might disappear with the vertex contractions. In particular, by Fact \[7\] Conditions (c), (d) and (e) together imply that \(G[X]\) and \(G[Y]\) are \(S\)-forests.

Finally, as explained earlier, the last conditions of both definitions ensure that VC the set described by \(X_{\leq cc}^S, X_{\leq cc}^S, Y_{\geq cc}^S, Y_{\geq cc}^S\) and \(Y_{\geq cc}^S\) is a vertex cover of \(G[X,Y]_{\leq cc}(X \setminus S) \cup P_Y\). Notice that \(X \setminus V(VC_X)\) and \(V(Y_{\geq cc})\) correspond the set of vertices in \(X\) and \(Y\), respectively, that do not belong to a block in the vertex cover VC. Such observations are used to prove the following two results.

**Lemma 14.** Let \(X \subseteq V_x\) and \(Y \subseteq \overline{V_x}\) such that \(G[X \cup Y]\) is an \(S\)-forest. There exist \(i \in \mathbb{I}_x\) and an \(S\)-contraction \(P_Y\) of \(Y\) such that (1) \(G[X \cup Y]_{\leq cc}(X \setminus S) \cup P_Y\) is a forest, (2) \(X\) is a partial solution associated with \(i\) and (3) \((Y, P_Y)\) is a complement solution associated with \(i\).

**Proof.** By Lemma \[3\] there exists an \(S\)-contraction \(P_Y\) of \(Y\) such that the following properties are satisfied:

(A) \(G[X \cup Y]_{\leq cc}(X \setminus S) \cup P_Y\) is a forest,

(B) for all \(P \in cc(X \setminus S) \cup P_Y\) and all \(v \in (X \cup Y) \cap S\), we have \(|N(v) \cap P| \leq 1\),

(C) the graph \(G[X,Y]_{\leq cc}(X \setminus S) \cup P_Y\) admits a vertex cover VC of size at most \(4\text{mim}(V_x)\) such that the neighborhoods of the blocks in VC are pairwise distinct in \(G[X,Y]_{\leq cc}(X \setminus S) \cup P_Y\).

We construct \(i = (X_{\leq cc}^S, X_{\leq cc}^S, X_{\leq cc}^S, Y_{\geq cc}^S, Y_{\geq cc}^S) \in \mathbb{I}_x\) from VC as follows:

- \(X_{\leq cc}^S = \{\text{rep}_{V_i}^1(\{v\}) \mid \{v\} \in (X \cap S) \cap VC\},\)
- \(X_{\leq cc}^S = \{\text{rep}_{V_i}^2(P) \mid P \in cc(X \setminus S) \cap VC\},\)
- \(X_{\leq cc}^S = \text{rep}_{V_i}^1(X \setminus V(VC)),\)
- \(Y_{\geq cc}^S = \{\text{rep}_{V_i}^2(P) \mid P \in P_Y \cap VC\},\)
- \(Y_{\geq cc}^S = \{\text{rep}_{V_i}^1(\{v\}) \mid \{v\} \in (Y \cap S) \cap VC\}.

Since \(|VC| \leq 4\text{mim}(V_x)\), we have \(|X_{\leq cc}^S| + |X_{\leq cc}^S| + |Y_{\geq cc}^S| + |Y_{\geq cc}^S| \leq 4\text{mim}(V_x)\) and thus we have \(i \in \mathbb{I}_x\).

We claim that \(X\) is a partial solution associated with \(i\). By construction of \(i\), Conditions (a), (b) and (f) of Definition \[12\] are satisfied. In particular, Condition (a) and (b) follow from Property (C), i.e. the neighborhoods of the blocks in VC are pairwise distinct in \(G[X,Y]_{\leq cc}(X \setminus S) \cup P_Y\).

So, the blocks in \(X_{\leq cc}(X \setminus S) \cap VC\) are pairwise non-equivalent for the \(d\)-neighbor equivalence over \(V_x\) for any \(d \in \mathbb{N}^+\) including 1 and 2. Consequently, there is a one to one correspondence between the blocks of \(X_{\leq cc}(X \setminus S) \cap VC\) and the representatives in \(X_{\leq cc}^S \cup X_{\leq cc}^S\).

It remains to prove Conditions (c), (d) and (e). We claim that Condition (c) is satisfied: \(\text{aux}(X, i)\) is a forest. Observe that, by construction, \(\text{aux}(X, i)\) is isomorphic to the graph \(G[X_{\leq cc}(X \setminus S) \cap Y_{\geq cc}P_Y \cap VC]\). Indeed, for every \(P \in Y_{\geq cc}P_Y \cap VC\), by construction, there exists a unique \(U \in Y_{\geq cc}^S \cup Y_{\geq cc}^S\) such that \(U \equiv V_i^X P\) or \(U \equiv V_i^Y P\). In both case, we have \(N(U) \cap V_x = N(P) \cap V_x\) and thus, the neighborhood of \(P\) in \(G[X_{\leq cc}(X \setminus S) \cap Y_{\geq cc}P_Y \cap VC]\) is the same as the neighborhood of \(U\) in \(\text{aux}(X, i)\). Since \(G[X_{\leq cc}(X \setminus S) \cap Y_{\geq cc}P_Y \cap VC]\) is a subgraph of \(G[X \cup Y]_{\leq cc}(X \setminus S) \cup P_Y\) and this latter graph is a forest, we deduce that \(\text{aux}(X, i)\) is also a forest.

We deduce that Conditions (d) and (e) are satisfied from property (B) and Fact \[11\]. Consequently, \(X\) is a partial solution associated with \(i\).
Let us now prove that \((Y, \mathcal{P}_Y)\) is a complement solution associated with \(i\). From the construction of \(i\) and with the same argument used earlier, we deduce that that Conditions (a) and (b) of Definition 13 are satisfied. By Property (A) \(G[X \cup Y]_{\text{cc}(X \setminus S) \cup \mathcal{P}_Y}\) is a forest, as a subgraph, \(G[Y]_{\mathcal{P}_Y}\) is also a forest and thus Condition (c) is satisfied. Conditions (d) and (e) are satisfied from Property (B) and Fact 11.

It remains to prove Condition (f): \(N(X_{\mathcal{P}_r}) \cap y_{\mathcal{P}_r} = \emptyset\) where \(y_{\mathcal{P}_r}\) contains the blocks \(\{v\} \in \left(Y \cap S\right)\) such that \(\text{rep}_{12}^r(\{v\}) \notin y_{\mathcal{P}_c}\) and the blocks \(P \in \mathcal{P}_Y\) such that \(\text{rep}_{12}^r(P) \notin y_{\mathcal{P}_c}\).

By construction, \(y_{\mathcal{P}_r} = Y \setminus V(\mathcal{C})\). Since, \(\mathcal{C}\) is a vertex cover of the graph \(G[X,Y]_{\text{cc}(X \setminus S) \cup \mathcal{P}_Y}\), there are no edges between \(X \setminus \text{cc}(X \setminus S) \setminus \mathcal{P}_Y\) and \(Y\). We deduce that \(N(X \setminus V(\mathcal{C})) \cap y_{\mathcal{P}_r} = \emptyset\). Since \(X \setminus V(\mathcal{C}) = X_{\mathcal{P}_r}\), we conclude that \(N(X_{\mathcal{P}_r}) \cap y_{\mathcal{P}_r} = \emptyset\). This proves that \((Y, \mathcal{P}_Y)\) is a complement solution associated with \(i\).

\[\square\]

**Lemma 15.** Let \(i = (X_{\mathcal{P}_c}, X_{\mathcal{P}_c}, X_{\mathcal{P}_r}, y_{\mathcal{P}_c}, y_{\mathcal{P}_c}) \in \mathbb{I}_x\), \(X\) be a partial solution associated with \(i\) and \((Y, \mathcal{P}_Y)\) be a complement solution associated with \(i\). If the graph \(G[X \cup Y]_{\text{cc}(X \setminus S) \cup \mathcal{P}_Y}\) is a forest, then \(G[X \cup Y]\) is an \(S\)-forest.

**Proof.** Assume that \(G[X \cup Y]_{\text{cc}(X \setminus S) \cup \mathcal{P}_Y}\) is a forest. By Fact 11, in order to prove that \(G[X \cup Y]\) is an \(S\)-forest, it is enough to prove that for all \(v \in (X \cup Y) \cap S\) and all \(P \in \text{cc}(X \setminus S) \cup \mathcal{P}_Y\), we have \(|N(v) \cap P| \leq 1\). Let us prove this statement for a vertex \(v \in X \cap S\) the proof is symmetric for \(v \in Y \cap S\). Let \(P \in (X \cup Y)_{\text{cc}(X \setminus S) \cup \mathcal{P}_Y}\). If \(P \notin (\text{cc}(X \setminus S) \cup \mathcal{P}_Y)\), then \(P\) is a singleton in \((X \cup Y)\) and we are done. If \(P \in \text{cc}(X \setminus S)\), then Condition (e) of Definition 12 guarantees that we have \(|N(v) \cap P| \leq 1\).

Assume now that \(P \in \mathcal{P}_Y\). Suppose first that \(\text{rep}_{12}^r(P) \notin y_{\mathcal{P}_c}\). From Condition (f) of Definition 13 we have \(N(P) \cap X_{\mathcal{P}_r} = \emptyset\). Let \(r = \text{rep}_{12}^r(\{v\})\). From the definition of \(X_{\mathcal{P}_r}\) in Definition 12 we deduce that if \(r \notin X_{\mathcal{P}_c}\), then \(N(v) \cap V_r \subseteq N(X_{\mathcal{P}_r})\) and thus \(N(v) \cap P = \emptyset\). On the other hand, if \(r \in X_{\mathcal{P}_c}\), then Condition (d) of Definition 13 ensures that \(|N(v) \cap P| \leq 1\) and thus \(|N(v) \cap P| \leq 1\). Now, suppose that \(\text{rep}_{12}^r(P) \in y_{\mathcal{P}_c}\). By Condition (e) of Definition 12 we know that \(|N(v) \cap \text{rep}_{12}^r(P)| \leq 1\). From Fact 11, we conclude that \(|N(v) \cap P| \leq 1\). This concludes the proof of Lemma 15. \(\square\)

For each index \(i \in \mathbb{I}_x\), we will design an equivalence relation \(\sim_i\) between the partial solutions associated with \(i\). We will prove that, for any partial solutions \(X\) and \(W\) associated with \(i\), if \(X \sim_i W\), then, for any complement solution \(Y \subseteq V_r\) associated with \(i\), the graph \(G[X \cup Y]\) is an \(S\)-forest if and only if \(G[W \cup Y]\) is an \(S\)-forest. Then, given a set of partial solutions \(\mathcal{A}\) whose size needs to be reduced, it is sufficient to keep, for each \(i \in \mathbb{I}_x\) and each equivalence class \(C\) of \(\sim_i\), one partial solution in \(C\) of maximal weight. The resulting set of partial solutions has size bounded by \(|\mathbb{I}_x| \cdot (4 \text{ mim}(V_x))^\text{ mim}(V_x)|V_x|^i\) because \(\sim_i\) generates at most \((4 \text{ mim}(V_x))^\text{ mim}(V_x)|V_x|^i\) equivalence classes.

Intuitively, given two partial solutions \(X\) and \(W\) associated with \(i = (X_{\mathcal{P}_c}, X_{\mathcal{P}_c}, X_{\mathcal{P}_r}, y_{\mathcal{P}_c}, y_{\mathcal{P}_c})\), we have \(X \sim_i W\) if the blocks of \(\mathcal{V}\) (i.e., the vertex cover described by \(i\)) are equivalently connected in \(G[X_{\text{cc}(X \setminus S)}] | y_{\mathcal{P}_c}^c \cup y_{\mathcal{P}_c}^c\) and \(G[W_{\text{cc}(W \setminus S)}] | y_{\mathcal{P}_c}^c \cup y_{\mathcal{P}_c}^c\). In order to compare these connections, we use the following notion.

**Definition 16 (cc(X, i)).** Let \(i = (X_{\mathcal{P}_c}, X_{\mathcal{P}_c}, X_{\mathcal{P}_r}, y_{\mathcal{P}_c}, y_{\mathcal{P}_c}) \in \mathbb{I}_x\) and \(X \subseteq V_x\) be a partial solution associated with \(i\). For each connected component \(C\) of \(\text{aux}(X, i)\), we define the set \(C_{\mathcal{P}_c}\) as follows:

- for every \(U \in C\) such that \(U \subseteq y_{\mathcal{P}_c}^c \cup y_{\mathcal{P}_c}^c\), we have \(U \subseteq X_{\mathcal{P}_c}\),
- for every \(\{v\} \in \left(X \cap S\right) \cap C\) such that \(\{v\} \subseteq V_r\) for some \(R \in X_{\mathcal{P}_c}\), we have \(R \subseteq X_{\mathcal{P}_c}\).
for every $U \in \text{cc}(X \setminus S)$ such that $U \supseteq V_z$ for some $R \in \mathcal{X}_v$, we have $R \in C_{vc}$.

We define $\text{cc}(X, i)$ as the collection $\{C_{vc} \mid C$ is a connected component of $\text{aux}(X, i)\}$.

For a connected component $C$ of $\text{aux}(X, i)$, the set $C_{vc}$ contains $C \cap (\mathcal{Y}_{vc} \cup \mathcal{Y}_{vc}^S)$ and the representatives of the blocks in $C \cap \mathcal{X}_{\text{cc}(X \setminus S)} \cap \text{VC}$ with $\text{VC}$ the vertex cover described by $i$. Consequently, for every $X \subseteq V_x$ and $i \in I_x$, the collection $\text{cc}(X, i)$ is partition of $\mathcal{X}_{vc}^S \cup \mathcal{X}_{vc} \cup \mathcal{Y}_{vc} \cup \mathcal{Y}_{vc}^S$. For the example given in Figure 3 observe that $\text{cc}(X, i)$ is the partition that contains $\{R_1, U_1, U_2\}$, $\{R_2, U_3\}$ and $\{U_4\}$ (see also Figure 13).

Now we are ready to give the notion of equivalence between partial solutions. We say that two partial solutions $X, W$ associated with $i$ are $i$-equivalent, denoted by $X \sim_i W$, if $\text{cc}(X, i) = \text{cc}(W, i)$. Our next result is the most crucial step. As already explained, our task is to show equivalence between partial solutions under any complement solution with respect to $S$-forests. Figure 14 gives an example of two $i$-equivalent partial solutions.

**Lemma 17.** Let $i = (\mathcal{X}_{vc}, \mathcal{X}_{vc}^S, \mathcal{X}_{vc}^T, \mathcal{Y}_{vc}, \mathcal{Y}_{vc}^S) \in I_x$. For every partial solutions $X, W$ associated with $i$ such that $X \sim_i W$ and for every complement solution $(Y, P_Y)$ associated with $i$, the graph $G[X \cup Y]_{\text{cc}(X(S)) \cup P_Y}$ is a forest if and only if the graph $G[W \cup Y]_{\text{cc}(W(S)) \cup P_Y}$ is a forest.

**Proof.** Let $X, W$ be two partial solutions associated with $i$ such that $X \sim_i W$ and let $(Y, P_Y)$ be a complement solution associated with $i$. To prove this lemma, we show that if $G[W \cup Y]_{\text{cc}(W(S)) \cup P_Y}$ contains a cycle, then $G[X \cup Y]_{\text{cc}(X(S)) \cup P_Y}$ contains a cycle too. See Figure 16 for some intuitions on this proof. We will use the following notation in this proof.

For $Z \in \{X, W\}$, we denote by $\text{VC}_Z$ the set that contains:

- all $\{v\} \in (Z \cap S)$ such that $\text{rep}_1^V(\{v\}) \in \mathcal{X}_{vc}^S$,

- all $P \in \text{cc}(Z \setminus S)$ such that $\text{rep}_2^V(P) \in \mathcal{X}_{vc}$.

We define also $\text{VC}_Y$ as the set that contains:

- all $\{v\} \in (Y \cap S)$ such that $\text{rep}_1^V(\{v\}) \in \mathcal{Y}_{vc}^S$,

- all $P \in P_Y$ such that $\text{rep}_2^V(P) \in \mathcal{Y}_{vc}$.

The sets $\text{VC}_X, \text{VC}_W$ and $\text{VC}_Y$ contain the blocks in $X_{\text{cc}(X(S))}, W_{\text{cc}(W(S))}$ and $Y_{\text{cc}(Y)}$, respectively, which belong to the vertex cover described by $i$. Finally, for each $Z \in \{X, W\}$, we define the following two edge-disjoint subgraphs of $G[Z \cup Y]_{\text{cc}(Z(S)) \cup P_Y}$:

- $G_Z = G[Z_{\text{cc}(Z(S)) \cup \text{VC}_Y}]$,

- $G_Z^+ = G[Z \cup Y]_{\text{cc}(Z(S)) \cup P_Y} - G_Z$.

As explained in the proof of Lemma 14 for any $Z \in \{X, W\}$, the graph $\text{aux}(Z, i)$ is isomorphic to the graph $G_Z$. Informally, $G_Z$ contains the edges of $G[Z \cup Y]_{\text{cc}(Z(S)) \cup P_Y}$ which are induced by $Z_{\text{cc}(Z(S))}$ and those between $Z_{\text{cc}(Z(S))}$ and $\text{VC}_Y$. The following fact implies that $G_Z^+$ contains the edges of $G[Z \cup Y]_{\text{cc}(Z(S)) \cup P_Y}$ that are induced by $Y_{\text{cc}(Y)}$ and those between $Y_{\text{cc}(Y)} \setminus \text{VC}_Y$ and $\text{VC}_Z$.

**Fact 18.** For any $Z \in \{X, W\}$, the set $\text{VC}_Z \cup \text{VC}_Y$ is a vertex cover of $G[Z \cup Y]_{\text{cc}(Z(S)) \cup P_Y}$. 

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Proof. First observe that \( N(Y \setminus V(VC_Y)) \cap X_{cc} = \emptyset \) thanks to Condition (f) of Definition \[\ref{def:cc}]. Moreover, we have \( X_{cc} \equiv V(P_Y \cup V(C_Z)) \) by Condition (f) of Definition \[\ref{def:aux}]. We conclude that there are no edges between \( Y_{\subseteq P_Y} \setminus V(C_Y) \) and \( Z_{\subseteq cc(Z \setminus S)} \setminus V(C_Z) \) in \( G[Z, Y]_{\subseteq cc(Z \setminus S), P_Y} \). Hence, \( VC_Z \cup VC_Y \) is a vertex cover of \( G[Z, Y]_{\subseteq cc(Z \setminus S), P_Y} \). \( \square \)

Assume that \( G[W \cup Y]_{\subseteq cc(W \setminus S), P_Y} \) contains a cycle \( C \). Our task is to show that \( G[X \cup Y]_{\subseteq cc(X \setminus S), P_Y} \) contains a cycle as well. We first explore properties of \( C \) with respect to \( G_W \) and \( \overline{G_W} \). Since the graph \( aux(W, i) \) is a forest and it is isomorphic to \( G_W \), we know that \( C \) must contain at least one edge from \( \overline{G_W} \). Moreover, \( C \) must go through a block of \( W_{\subseteq cc(W \setminus S)} \) because \( G[Y]_{\subseteq P_Y} \) is a forest. Consequently, (and because from Fact \[\ref{fact:vc} \) \( VC_W \cup VC_Y \) is a vertex cover of \( G[W \cup Y]_{\subseteq cc(W \setminus S), P_Y} \), we deduce that \( C \) is the concatenation of edge-disjoint paths \( P_1, \ldots, P_\ell \) such that for each \( \ell \in [\ell] \) we have:

- \( P_\ell \) is a non-empty path with endpoints in \( VC_W \cup VC_Y \) and internal blocks not in \( VC_W \cup VC_Y \) and \( P_\ell \) is either a path of \( G_W \) or \( \overline{G_W} \).

At least one of these paths is in \( \overline{G_W} \) and, potentially, \( C \) may be entirely contained in \( \overline{G_W} \). Figure 4 presents two possible interactions between \( C \) and the graphs \( G_W \) and \( \overline{G_W} \).

![Figure 4: How cycles in \( G[W \cup Y]_{\subseteq cc(W \setminus S), P_Y} \) may interact with the graphs \( G_W \) and \( \overline{G_W} \). The solid edges belong to \( G_W \) and the dashed edges belong to \( \overline{G_W} \).](image)

Given an endpoint \( U \in VC_W \cup VC_Y \) of one of the paths \( P_1, \ldots, P_\ell \), we define \( U_X \) and \( U_i \) as the analogs of \( U \) in \( VC_X \cup VC_Y \) and \( X_{vc}^S \cup X_{vc}^P \cup Y_{vc}^S \cup Y_{vc}^P \), respectively, as follows:

- if \( U \in cc(W \setminus S) \), then \( U_X \) and \( U_i \) are the unique elements of \( cc(X \setminus S) \) and \( X_{vc}^S \), respectively, such that \( U \equiv \gamma^2 U_X \equiv \gamma^2 U_i \),
- if \( U = \{v\} \in (W \setminus S)^{1} \), then \( U_X \) and \( U_i \) are the unique elements in \( (X \setminus S)^{1} \) and \( X_{vc}^S \), respectively, such that \( U \equiv \gamma^1 U_X \equiv \gamma^1 U_i \),
- if \( U \in P_Y \), then \( U_X = U \) and \( U_i \) is the unique element of \( Y_{vc}^S \) such that \( U \equiv \gamma^2 U_i \),
- otherwise, if \( U = \{v\} \in (Y \setminus S)^{1} \), then \( U_X = U \) and \( U_i \) is the unique element of \( Y_{vc}^S \) such that \( U \equiv \gamma^2 U_i \).

Observe that \( U_X \) and \( U_i \) exist by Conditions (a) and (b) of Definition \[\ref{def:cc} \) and Definition \[\ref{def:aux} \).

For each \( \ell \in [\ell] \), we construct a non-empty path \( P'_\ell \) whose endpoints are the analogs in \( VC_X \cup VC_Y \) of the endpoints of \( P_\ell \) and such that if \( P_\ell \) is a path in \( G_W \) (resp. \( \overline{G_W} \)), then \( P'_\ell \) is a path in \( G_X \) (resp. \( \overline{G_X} \)). This is sufficient to prove the claim because thanks to this, we can
By construction, we formalize this notion of representativity. Let \( \ell \in [t] \) and \( U, T \) be the endpoints of \( P_\ell \). We denote by \( U_X, T_X, U_i \) and \( T_i \) the analogs of \( U \) and \( T \) in \( VC_X \cup VC_Y \) and \( X_{cc}^X \cup X_{cc}^Y \cup Y_{cc}^X \cup Y_{cc}^Y \), respectively.

First, assume that \( P_\ell \) is path of \( G_W \). Observe that \( U_i \) and \( T_i \) belong to the same partition class of \( cc(W, i) \). This follows from the definitions of \( U_i \), \( T_i \) and the fact that \( G_W \) is isomorphic to \( aux(W, i) \). As \( W \sim_i X \), we deduce that \( U_i \) and \( T_i \) belong to the same partition class of \( cc(X, i) \). By construction, \( U_i \) and \( T_i \) are the analogs of \( U_X \) and \( T_X \) in \( X_{cc}^X \cup X_{cc}^Y \cup Y_{cc}^X \cup Y_{cc}^Y \). We conclude that \( U_X \) and \( T_X \) are connected in \( G_X \) via a path \( P_\ell \). We claim that \( P_\ell \) is not empty because we have \( U_X \neq T_X \). As \( P_\ell \) is a non-empty path of \( G_W \) and because \( G_W \) is acyclic, we know that \( U \) and \( T \) are distinct. Hence, by the construction of \( U_X \) and \( T_X \), we deduce that \( U_X \neq T_X \).

Now, assume that \( P_\ell \) is a non-empty path of \( \overline{G_W} \). Since \( VC_W \cup VC_Y \) is a vertex cover of \( G[W, Y]_{|cc(W \setminus S) \cup P_Y} \), the blocks in \( W_{cc(W \setminus S)} \) that do not belong to \( VC_W \) are isolated in \( \overline{G_W} \). As \( P_\ell \) is not empty, the blocks of \( P_\ell \) which belong to \( W_{cc(W \setminus S)} \) are in \( VC_W \). Because the internal blocks of the paths \( P_1, \ldots, P_t \) are not in \( VC_W \cup VC_Y \), we deduce that the internal blocks of \( P_\ell \) belong to \( Y_{cc}P_Y \). We distinguish the following cases:

- **If both endpoints of \( P_\ell \) belong to \( VC_Y \), then \( U_X = U \), \( T_X = T \) and all the blocks of \( P_\ell \) belong to \( Y_{cc}P_Y \). It follows that \( P_\ell \) is a non-empty path of \( \overline{G_X} \) because \( G[Y]_{|P_Y} \) is a subgraph of \( \overline{G_X} \). In this case, we take \( P_\ell' = P_\ell \).

- **Assume now that one or two endpoints of \( P_\ell \) belong to \( VC_W \). Suppose w.l.o.g. that \( U \) belongs to \( VC_W \). Since \( P_\ell \) is non-empty and the internal blocks of \( P_\ell \) are in \( Y_{cc}P_Y \), \( U \) has a neighbor \( Q \in Y_{cc}P_Y \) in \( P_\ell \). We claim that \( Q \) is adjacent to \( U_X \) in \( \overline{G_X} \). By definition of \( U_X \), we have \( U \equiv_{U_X} U_X \) for some \( d \in \{1, 2\} \) and in particular \( N(U) \cap \overline{V_x} = N(U_X) \cap \overline{V_x} \). As \( U \) and \( Q \) are adjacent in \( \overline{G_W} \), we deduce that \( N(U) \cap Q \neq \emptyset \). It follows that \( N(U_X) \cap Q \neq \emptyset \) and thus \( Q \) and \( U_X \) are adjacent in \( \overline{G_X} \). Symmetrically, we can prove that if \( T \in VC_W \), then the neighbor of \( T \) in \( P_\ell \) is adjacent to \( T_X \) in \( \overline{G_X} \).

Hence, the neighbors of \( U \) and \( T \) in \( P_\ell \) are adjacent to \( U_X \) and \( T_X \), respectively in \( \overline{G_X} \). We obtain \( P_\ell' \) from \( P_\ell \) by replacing \( U \) and \( T \) by \( U_X \) and \( T_X \). Since the internal blocks of \( P_\ell \) belong to \( Y_{cc}P_Y \) and \( G[Y]_{|P_Y} \) is a subgraph of \( \overline{G_X} \), we deduce that \( P_\ell' \) is a path of \( \overline{G_X} \). The path \( P_\ell' \) is not empty because it contains \( U_X \) and \( Q \) which are distinct blocks of \( \overline{G_X} \) as \( U_X \in VC_X \) (since \( U \in VC_W \) by assumption) and \( Q \in Y_{cc}P_Y \).

The following theorem proves that, for every set of partial solutions \( A \subseteq 2^{V_x} \), we can compute a small subset \( B \subseteq A \) such that \( B \) represents \( A \), i.e., for every \( Y \subseteq \overline{V_x} \), the best solutions we obtain from the union of \( Y \) with a set in \( A \) are as good as the ones we obtain from \( B \). Firstly, we formalize this notion of representativity.

**Definition 19 (Representativity).** For every \( A \subseteq 2^{V_x} \) and \( Y \subseteq \overline{V_x} \), we define

\[
\text{best}(A, Y) = \max \{w(X) \mid X \in A \text{ and } G[X \cup Y] \text{ is an } S\text{-forest}\}.
\]

Given \( A, B \subseteq 2^{V_x} \), we say that \( B \) represents \( A \) if, for every \( Y \subseteq \overline{V_x} \), we have \( \text{best}(A, Y) = \text{best}(B, Y) \).

We recall that \( s\text{-nec}_2(A) = \max(\text{nec}_2(A), \text{nec}_2(\overline{A})) \).
Lemma 20. There exists an algorithm reduce that, given a set \( A \subseteq 2^V \), outputs in time \( O(|A| \cdot |I_x| \cdot (4\text{min}(V_x))^{4\text{min}(V_x)} \cdot \log(s\text{-nec}_2(V_x)) \cdot n^3) \) a subset \( B \subseteq A \) such that \( B \) represents \( A \) and \( |B| \leq |I_x| \cdot (4\text{min}(V_x))^{4\text{min}(V_x)} \).

Proof. Given \( A \subseteq 2^V \) and \( i \in I_x \), we define \( \text{reduce}(A, i) \) as the operation which returns a set containing one partial solution \( X \in A \) associated with \( i \) of each equivalence class of \( \sim_{i} \) such that \( w(X) \) is maximum. Moreover, we define \( \text{reduce}(A) = \bigcup_{i \in I_x} \text{reduce}(A, i) \).

We prove first that \( \text{reduce}(A) \) represents \( A \), that is \( \text{best}(A, Y) = \text{best}(\text{reduce}(A), Y) \) for all \( Y \subseteq V_x \). Let \( Y \subseteq V_x \). Since \( \text{reduce}(A) \subseteq A \), we already have \( \text{best}(\text{reduce}(A), Y) \leq \text{best}(A, Y) \). Consequently, if there is no \( X \in A \) such that \( G[X \cup Y] \) is an \( S \)-forest, we have \( \text{best}(\text{reduce}(A), Y) = \text{best}(A, Y) = \max(\emptyset) = -\infty \).

Assume that there exists \( X \in A \) such that \( G[X \cup Y] \) is an \( S \)-forest. Let \( X \in A \) such that \( G[X \cup Y] \) is an \( S \)-forest and \( w(X) = \text{best}(A, Y) \). By Lemma 14 there exists \( i \in I_x \) and an \( S \)-contraction \( P_Y \) of \( Y \) such that (1) \( G[X \cup Y] \subseteq (X \cup Y)_{G[\mathcal{X} \cup S] \cup P_Y} \) is a forest, (2) \( X \) is a partial solution associated with \( i \) and (3) \( Y, P_Y \) is a complement solution associated with \( i \).

From the construction of \( \text{reduce}(A, i) \), there exists \( W \in \text{reduce}(A) \) such that \( W \) is a partial solution associated with \( i \), \( X \sim_{i} W \), and \( w(W) \geq w(X) \). By Lemma 17 and since \( G[X \cup Y] \subseteq (X \cup Y)_{G[\mathcal{X} \cup S] \cup P_Y} \) is a forest too. Thanks to Lemma 15 we deduce that \( G[W \cup Y] \) is an \( S \)-forest. As \( w(W) \geq w(X) = \text{best}(A, Y) \), we conclude that \( \text{best}(A, Y) = \text{best}(\text{reduce}(A), Y) \). Hence, \( \text{reduce}(A) \) represents \( A \).

We claim that \( |\text{reduce}(A)| \leq |I_x| \cdot (4\text{min}(V_x))^{4\text{min}(V_x)} \). For every \( i \in I_x \), \( (X, i) \in \mathcal{X}_V \cup \mathcal{X}_S \cup \mathcal{X}_W \cup \mathcal{Y}_V \cup \mathcal{Y}_S \). Since \( |\mathcal{X}_V| + |\mathcal{X}_S| + |\mathcal{Y}_V| + |\mathcal{Y}_S| \leq 4\text{min}(V_x) \), there are at most \( (4\text{min}(V_x))^{4\text{min}(V_x)} \) possible values for \( cc(X, i) \). We deduce that, for every \( i \in I_x \), the relation \( \sim_{i} \) generates at most \( (4\text{min}(V_x))^{4\text{min}(V_x)} \) equivalence classes, so \( |\text{reduce}(A, i)| \leq (4\text{min}(V_x))^{4\text{min}(V_x)} \) for every \( i \in I_x \). By construction, we conclude that \( |\text{reduce}(A)| \leq |I_x| \cdot (4\text{min}(V_x))^{4\text{min}(V_x)} \).

It remains to prove the runtime. As \( \text{nec}_1(V_x) \leq \text{nec}_2(V_x) \), by Lemma 5 we can compute in time \( O(s\text{-nec}_2(V_x) \cdot \log(s\text{-nec}_2(V_x)) \cdot n^3) \) the sets \( R^x_V, R^x_S, R^x_W, R^x_T \) and data structures which compute \( \text{rep}_V^x, \text{rep}_S^x \) and \( \text{rep}_W^x \) in time \( O(\log(s\text{-nec}_2(V_x)) \cdot n^3) \). Given \( R^x_V, R^x_S, R^x_W, R^x_T \), we can compute \( I_x \) in time \( O(|I_x| \cdot n^2) \). Since \( \text{nec}_2(V_x) \leq |I_x| \), the time required to compute these sets and data structures is less than \( O(|A| \cdot |I_x| \cdot (4\text{min}(V_x))^{4\text{min}(V_x)} \cdot \log(s\text{-nec}_2(V_x)) \cdot n^3) \).

For each \( i \in I_x \) and \( X \in A \), we can decide whether \( X \) is a partial solution associated with \( i \) and compute \( \text{aux}(X, i), cc(X, i) \) in time \( O(\log(s\text{-nec}_2(V_x)) \cdot n^3) \). For doing so, we simply start by computing \( \text{rep}_V^x(C) \) and \( \text{rep}_S^x(\{v\}) \) for each \( C \subseteq cc(X \setminus S) \) and \( \{v\} \subseteq \{X \setminus S\} \), this is doable in \( O(\log(s\text{-nec}_2(V_x)) \cdot n^3) \) since \( |cc(X \setminus S)| + |X \cap S| \leq n \). Then with standard algorithmic techniques, we check whether \( X \) satisfies all the conditions of Definition 12 and compute \( \text{aux}(X, i) \) and \( cc(X, i) \).

Given two partial solutions \( X, W \) associated with \( i \), \( cc(X, i) \) and \( cc(W, i) \), we can decide whether \( X \sim_{i} W \) in time \( O(\text{min}(V_x)) \leq O(n) \). We deduce that, for each \( i \in I_x \), we can compute \( \text{reduce}(A, i) \) in time \( O(|A| \cdot (4\text{min}(V_x))^{4\text{min}(V_x)} \cdot \log(s\text{-nec}_2(V_x)) \cdot n^3) \). We deduce the running time to compute \( \text{reduce}(A) \) by multiplying the running time needed for \( \text{reduce}(A, i) \) by \( |I_x| \). □

We are now ready to prove the main theorem of this paper. For two subsets \( A \) and \( B \) of \( 2^V \), we define the merging of \( A \) and \( B \), denoted by \( A \otimes B \), as \( A \otimes B := \{X \cup Y \mid X \in A \text{ and } Y \in B\} \). Observe that \( A \otimes B = \emptyset \) whenever \( A = \emptyset \) or \( B = \emptyset \).
Theorem 21. There exists an algorithm that, given an $n$-vertex graph $G$ and a rooted layout $(T, \delta)$ of $G$, solves Subset Feedback Vertex Set in time

$$O\left(\sum_{x \in V(T)} ||x||^3 \cdot (4\text{min}(V_x))^{12\text{min}(V_x)} \cdot \log(s\text{-nec}_2(V_x)) \cdot n^3\right).$$

Proof. The algorithm is a usual bottom-up dynamic programming algorithm. For every node $x$ of $T$, the algorithm computes a set of partial solutions $A_x \subseteq 2^{V_x}$ such that $A_x$ represents $2^{V_x}$ and $|A_x| \leq ||x|| \cdot (4\text{min}(V_x))^{12\text{min}(V_x)}$. For the leaves $x$ of $T$ such that $V_x = \{v\}$, we simply take $A_x = 2^{V_x} = \{\emptyset, \{v\}\}$. In order to compute $A_x$ for $x$ an internal node of $T$ with $a$ and $b$ as children, our algorithm will simply compute $A_x = \text{reduce}(A_a \otimes A_b)$. Once the the set $A_r$ is computed with $r$ the root of $T$, our algorithm outputs a set $X \in A_r$ of maximum weight.

By Lemma 20 we have $|A_x| \leq ||x|| \cdot (4\text{min}(V_x))^{12\text{min}(V_x)}$, for every node $x$ of $T$. The following claim helps us to prove that $A_x$ represents $2^{V_x}$ for the internal nodes $x$ of $T$.

Claim 22. Let $x$ be an internal of $T$ with $a$ and $b$ as children. If $A_a$ and $A_b$ represent, respectively, $2^{V_a}$ and $2^{V_b}$, then $\text{reduce}(A_a \otimes A_b)$ represents $2^{V_x}$.

Proof. Assume that $A_a$ and $A_b$ represent, respectively, $2^{V_a}$ and $2^{V_b}$. First, we prove that $A_a \otimes A_b$ represents $2^{V_x}$. We have to prove that, for every $Y \subseteq V_x$, we have $\text{best}(A_a \otimes A_b, Y) = \text{best}(2^{V_x}, Y)$. Let $Y \subseteq V_x$. By definition of best, we have the following

$$\text{best}(A_a \otimes A_b, Y) = \max\{w(X) + w(W) \mid X \in A_a \land W \in A_b \land G[X \cup W \cup Y] \text{ is an } S\text{-forest}\}$$

$$= \max\{\text{best}(A_a, W \cup Y) + w(W) \mid W \in A_b\}.$$ 

As $A_a$ represents $2^{V_a}$, we have $\text{best}(A_a, W \cup Y) = \text{best}(2^{V_a}, W \cup Y)$ and we deduce that $\text{best}(A_a \otimes A_b, Y) = \text{best}(2^{V_a} \otimes A_b, Y)$. Symmetrically, as $A_b$ represents $2^{V_b}$, we infer that $\text{best}(2^{V_a} \otimes A_b, Y) = \text{best}(2^{V_a} \otimes 2^{V_b}, Y)$. Since $2^{V_a} \otimes 2^{V_b} = 2^{V_x}$, we conclude that $\text{best}(A_a \otimes A_b, Y)$ equals $\text{best}(2^{V_x}, Y)$. As this holds for every $Y$, it proves that $A_a \otimes A_b$ represents $2^{V_x}$. By Lemma 20 we know that $\text{reduce}(A_a \otimes A_b)$ represents $A_a \otimes A_b$. As the relation “represents” is transitive, we conclude that $\text{reduce}(A_a \otimes A_b)$ represents $2^{V_x}$. \hfill\Box

For the leaves $x$ of $T$, we obviously have that $A_x$ represents $2^{V_x}$, since $A_x = 2^{V_x}$. From Claim 22 and by induction, we deduce that $A_x$ represents $2^{V_x}$ for every node $x$ of $T$. In particular, $A_r$ represents $2^{V(G)}$ with $r$ the root of $T$. By Definition 19, $A_r$ contains a set $X$ of maximum size such that $G[X]$ is an $S$-forest. This proves the correctness of our algorithm.

It remains to prove the running time. Observe that, for every internal node $x$ of $T$ with $a$ and $b$ as children, the size of $A_a \otimes A_b$ is at most $||x||^2 \cdot (4\text{min}(V_x))^{12\text{min}(V_x)}$ and it can be computed in time $O(||x||^2 \cdot (4\text{min}(V_x))^{12\text{min}(V_x)} \cdot n^3)$. By Lemma 20 the set $A_x = \text{reduce}(A_a \otimes A_b)$ is computable in time $O(||x||^3 \cdot (4\text{min}(V_x))^{12\text{min}(V_x)} \cdot \log(s\text{-nec}_2(V_x)) \cdot n^3)$. This proves the running time. \hfill\Box

3.1 Algorithmic consequences

In order to obtain the algorithmic consequences of our meta-algorithm given in Theorem 21 we need the following lemma which bounds the size of each set of indices with respect to the considered parameters.

Lemma 23. For every $x \in V(T)$, the size of $||x||$ is upper bounded by:
\[ 2^{O(rw(V_z)^3)}, \quad 2^{O(rw_q(V_z)^2 \log(rw_q(V_z)))}, \quad n^{O(mim(V_z)^2)}. \]

**Proof.** For \( A \subseteq V(G) \), let \( mw(A) \) be the number of different rows in the matrix \( M_A \). Observe that, for every \( A \subseteq V(G) \), we have \( mw(A) = \{ \text{rep}^4(\{v\}) \mid v \in A \} \). Remember that the \( X_{vc}^S \)'s and \( y_S \)'s are subsets of \( \{ \text{rep}^4(\{v\}) \mid v \in V_z \} \) and \( \{ \text{rep}^4(\{v\}) \mid v \in V_z^I \} \) respectively. From Definition 10 we have \( |X_{vc}^S| + |X_{vc}^I| + |y_S| \leq 4 \cdot \text{mim}(V_z) \), for every \( ( X_{vc}^S, X_{vc}^I, y_S, y_S^M ) \). Thus, the size of \( I_x \) is at most
\[ \text{nc}_1(V_z) \cdot (\text{nc}_2(V_z) + mw(V_z) + \text{nc}_2(V_z) + mw(V_z))^{4 \cdot \text{mim}(V_z)}. \]

**Rank-width.** By Lemma 4, we have \( \text{nc}_1(V_z) \leq 2^{rw(V_z)^2} \) and \( \text{nc}_2(V_z) \leq 2^{rw(V_z)^2} \). Moreover, there is at most \( 2^{rw(V_z)} \) different rows in the matrices \( M_{V_z, V_z} \) and \( M_{V_z, V_z} \), so \( mw(V_z) \) and \( mw(V_z) \) are upper bounded by \( 2^{rw(V_z)} \). By Lemma 3, we have \( 4 \cdot \text{mim}(V_z) \leq 4 \cdot \text{rw}(V_z) \). We deduce from these inequalities that \( |I_x| \leq 2^{rw(V_z)^2} \cdot (2^{2rw(V_z)^2} + 2^{rw(V_z)^2}) \cdot 4^{rw(V_z)}. \)

**\( \mathbb{Q} \)-rank-width.** By Lemma 4, we have \( \text{nc}_1(V_z), \text{nc}_2(V_z), \text{nc}_2(V_z) \in 2^{O(rw_q(V_z) \log(rw_q(V_z)))} \). Moreover, there is at most \( 2^{rw_q(V_z)} \) different rows in the matrices \( M_{V_z, V_z} \) and \( M_{V_z, V_z} \), so \( mw(V_z) \) and \( mw(V_z) \) are upper bounded by \( 2^{rw_q(V_z)} \). By Lemma 3, we have \( 4 \cdot \text{mim}(V_z) \leq 4 \cdot \text{rw}_q(V_z) \). We deduce from these inequalities that \( |I_x| \leq 2^{O(rw_q(V_z) \log(rw_q(V_z)))} \cdot \left( 2^{O(rw_q(V_z) \log(rw_q(V_z)))} + 2^{rw_q(V_z) + 1} \right) \cdot 4^{rw_q(V_z)}. \)

We conclude that \( |I_x| \leq 2^{O(rw_q(V_z) \log(rw_q(V_z)))} \).

**Mim-width.** By Lemma 4, we know that \( \text{nc}_1(V_z) \leq |V_z|^{\text{mim}(V_z)}, \text{nc}_2(V_z) \leq |V_z|^{2 \cdot \text{mim}(V_z)}, \) and \( \text{nc}_2(V_z) \leq |V_z|^{2 \cdot \text{mim}(V_z)} \). We can assume that \( n > 2 \) (otherwise the problem is trivial), so \( \text{nc}_2(V_z) + \text{nc}_2(V_z) \leq |V_z|^{4 \cdot \text{mim}(V_z)} + |V_z|^{2 \cdot \text{mim}(V_z)} \leq n^{2 \cdot \text{mim}(V_z)} \). Moreover, notice that, for every \( A \subseteq V(G) \), we have \( \{ \text{rep}^4(\{v\}) \mid v \in A \} \leq |A| \).

We deduce that \( |I_x| \leq n^{\text{mim}(V_z)} \cdot (n + n^{2 \cdot \text{mim}(V_z)})^{4 \cdot \text{mim}(V_z)} \). As we assume that \( n > 2 \), we have \( |I_x| \leq n^{8 \cdot \text{mim}(V_z)^2 + 5 \cdot \text{mim}(V_z)} \) in \( O(\text{mim}(V_z)^2) \).

Now we are ready to state our algorithms with respect to the parameters rank-width \( rw(G) \) and \( \mathbb{Q} \)-rank-width \( rw_q(G) \). In particular, with our next result we show that Subset Feedback Vertex Set is in FPT parameterized by \( rw_q(G) \) or \( rw(G) \).

**Theorem 24.** There exist algorithms that solve Subset Feedback Vertex Set in time \( 2^{O(rw(G)^3)} \cdot n^4 \) and \( 2^{O(rw_q(G)^2 \log(rw_q(G)))} \cdot n^4 \).

**Proof.** We first compute a rooted layout \( L = (T, \delta) \) of \( G \) such that \( rw(L) \in O(rw(G)) \) or \( rw_q(L) \in O(rw_q(G)) \). This is achieved through a \( (3k + 1) \)-approximation algorithm that runs in FPT time \( O(3^k \cdot n^4) \) parameterized by \( k \in \{ rw(G), rw_q(G) \} \). Then, we apply the algorithm given in Theorem 21. Observe that for every node \( x \in V(T) \), by Lemma 23 \( |I_x|^3 \) lies in \( 2^{O(rw(v_x)^3)} \) and \( 2^{O(rw_q(v_x)^2 \log(rw_q(v_x)))} \) and by Lemma 3 s-\text{nc}_2(V_x) lies in \( 2^{O(rw(v_x)^3)} \) and \( 2^{O(rw_q(v_x)^2 \log(rw_q(v_x)))} \). Moreover, Lemma 3 implies that \( \text{mim}(V_x) \) is upper bounded by \( 2^{rw(G) \log(rw(G))} \) and \( 2^{rw_q(G) \log(rw_q(G))} \). Therefore, we get the claimed runtimes for SFVS since \( T \) contains \( 2n - 1 \) nodes.
Regarding mim-width, our algorithm given below shows that Subset Feedback Vertex Set is in XP parameterized by the mim-width of a given rooted layout. Note that we cannot solve SFVS in FPT time parameterized by the mim-width of a given rooted layout unless FPT = W[1], since Subset Feedback Vertex Set is known to be W[1]-hard for this parameter even for the special case of $S = V(G)$ [29]. Moreover, contrary to the algorithms given in Theorem 24, here we need to assume that the input graph is given with a rooted layout. However, our next result actually provides a unified polynomial-time algorithm for Subset Feedback Vertex Set on well-known graph classes having bounded mim-width and for which a layout of bounded mim-width can be computed in polynomial time [1] (e.g., interval graphs, permutation graphs, circular arc graphs, convex graphs, $k$-polygon, Dilworth-$k$ and Co-$k$-degenerate graphs for fixed $k$).

**Theorem 25.** There exists an algorithm that, given an $n$-vertex graph $G$ and a rooted layout $L$ of $G$, solves Subset Feedback Vertex Set in time $n^{O(mim(L)^2)}$.

**Proof.** We apply the algorithm given in Theorem 21. By Lemma 4, we have $s$-nec$_2(V_x) \leq n^{O(mim)}$ and from Lemma 23 we have $|I_x|^3 \in n^{O(mim(V_x)^2)}$. The claimed runtime for SFVS follows by the fact that the rooted tree $T$ of $L$ contains $2n - 1$ nodes.

Let us relate our results for Subset Feedback Vertex Set to the Node Multiway Cut. It is known that Node Multiway Cut reduces to Subset Feedback Vertex Set [20]. In fact, we can solve Node Multiway Cut by adding a new vertex $v$ with a large weight that is adjacent to all terminals and, then, run our algorithms for Subset Feedback Vertex Set with $S = \{v\}$ on the resulting graph. Now observe that any extension of a rooted layout $L$ of the original graph to the resulting graph has mim-width $mim(L) + 1$. Therefore, all of our algorithms given in Theorems 24 and 25 have the same running times for the Node Multiway Cut problem.

### 4 Conclusion

This paper highlights the importance of the $d$-neighbor-equivalence relation to obtain meta-algorithm for several width measures at once. We extend the range of applications of this relation [3 7 22 35] by proving that it is useful for the atypical acyclicity constraint of the Subset Feedback Vertex Set problem. It would be interesting to see whether this relation can be helpful with other kinds of constraints such as 2-connectivity and other generalizations of Feedback Vertex Set such as the ones studied in [6]. In particular, one could consider the following generalization of Odd Cycle Transversal:

**Subset Odd Cycle Transversal (SOCT)**

**Input:** A graph $G$ and $S \subseteq V(G)$.

**Output:** A set $X \subseteq V(G)$ of minimum weight such that $G[X]$ does not contain an odd cycle that intersects $S$.

Similar to SFVS, we can solve SOCT in time $k^{O(k)} \cdot n^{O(1)}$ parameterized by treewidth and this is optimal under ETH [2]. We do not know whether SOCT is in XP parameterized by mim-width, though it is in FPT parameterized by clique-width or rank-width, since we can express it in MSO$_1$ (with the characterization used in [2]).

For many well-known graph classes a decomposition of bounded mim-width can be found in polynomial time. However, for general graphs it is known that computing mim-width is W[1]-hard and not in APX unless $\text{NP} = \text{ZPP}$ [40], while Yamazaki [42] shows that under the small
set expansion hypothesis it is not in APX unless \( P = NP \). For dynamic programming algorithms as in this paper, to circumvent the assumption that we are given a decomposition, we want functions \( f, g \) and an algorithm that given a graph of mim-width \( OPT \) computes an \( f(OPT) \)-approximation to mim-width in time \( n^{g(OPT)} \), i.e. XP by the natural parameter. This is the main open problem in the field. The first task could be to decide if there is a constant \( c \) and a polynomial-time algorithm that given a graph \( G \) either decides that its mim-width is larger than 1 or else returns a decomposition of mim-width \( c \).

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A Explanations with several figures

The following figures explain the relation between solutions (S-forests) and an index $i$.

Figure 5: Illustration of the $\overline{S}$-contraction $\mathcal{P}_Y$ constructed in Lemma 9. The graph represented is an $S$-forest induced by the union of a set $X \subseteq V_x$ and a set $Y \subseteq V_x$. The white filled disks represent the vertices in $S$. We transform this $S$-forest into a forest (see the next figure) by contracting the circled subsets of vertices. The $\overline{S}$-contraction used on $X$ is $cc(X \setminus S)$. Observe that $Y_1$ is the only block of $\mathcal{P}_Y$ which is not a connected component of $G[Y \setminus S]$. We need $Y_1$ to kill the orange cycle which is not an $S$-cycle.

Figure 6: The contracted graph $G[X \cup Y]_{cc(X \setminus S) \cup \mathcal{P}_Y}$. The vertices of this graph are the blocks of $cc(X \setminus S) \cup \mathcal{P}_Y$ and the singletons $\{v\}$ for every vertex $v$ in $(X \cup Y) \cap S$; these singletons are represented by the white filled disks.
Figure 7: The bipartite graph $G[X, Y]_{\text{cc}(X \setminus S) \cup P_Y}$. The colored squares represent the blocks in a vertex cover $VC$ of $G[X, Y]_{\text{cc}(X \setminus S) \cup P_Y}$ that satisfies the properties of Lemma 9.

Figure 8: The graph $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$. As done in Lemma 14, we construct from the vertex cover $VC$ an index $i = (X^S_{vc}, x^S_{vc}, X^S_{vc}, y^S_{vc}, y^S_{vc}) \in I_x$ such that $X$ is a partial solution associated with $i$ and $(Y, P_Y)$ is a complement solution associated with $i$.

The set $X^S_{vc}$ contains the representatives of the blocks in $\text{cc}(X \setminus S)$ that are in the vertex cover $VC$. In this example, we have $X^S_{vc} = \{R_2\}$ where $R_2 = \text{rep}_x^y(X_2)$. 

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Figure 9: The set $\mathcal{X}_V^S$ contains the representatives of the singletons $\{x\} \in \mathcal{V}_C$ with $x \in S$. In this example, we have $\mathcal{X}_V^S = \{R_1\}$ where $R_1 = \text{rep}_1^{V_C}(\{x_1\})$.

Figure 10: The set $\mathcal{X}_V^\pi = \text{rep}_1^{V_C}(X \setminus V(\mathcal{V}_C))$. It is the representative set of the union of the blocks in $X_{\downarrow \text{cc}}(X \setminus S) \setminus \mathcal{V}_C$. 

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Figure 11: The set $Y_{\text{vc}}$ contains the representatives of the blocks in $\mathcal{P}_Y \cap \text{VC}$. In this example, we have $Y_{\text{vc}} = \{R_1, R_2\}$ where $R_\ell = \text{rep}_{\text{VC}}(Y_\ell)$ for $\ell = 1, 2$.

Figure 12: The set $Y_{\text{vc}}$ contains the representatives of the singletons $\{y\} \in \text{VC}$ with $y \in S$. In this example, we have $Y_{\text{vc}} = \{R_3, R_4\}$ where $R_\ell = \text{rep}_{1}(\{y_\ell\})$ for $\ell = 3, 4$. 
Figure 13: The auxiliary graph $\text{aux}_x(X, i)$. It can be obtained from $G[X \cup Y]_{\downarrow \text{cc}(X \setminus S) \cup P_Y}$ by (1) removing the edges between blocks of $Y_{\downarrow P_Y}$, (2) removing the blocks of $Y_{\downarrow P_Y}$ that do not belong to $\text{VC}$ and (3) replacing each remaining blocks of $Y_{\downarrow P_Y}$ by its representatives.

The blocks of the partition $\text{cc}(X, i)$ are \{$R_1, \overline{R_1}$, $R_2$, $\overline{R_2}$\}, \{$R_2, \overline{R_3}$\} and \{$R_4, \overline{R_4}$\} where $R_1$ and $R_2$ are the representatives of \{$x_1$\} and $X_2$ respectively.
Figure 14: The auxiliary graphs \( \text{aux}_x(X, i) \) and \( \text{aux}_x(W, i) \) with \( W \) a partial solution associated with \( i \). We have \( R_1 = \text{rep}^V_x(\{w_1\}) = \text{rep}^V_x(\{x_1\}) \) and \( R_2 = \text{rep}^V_x(W_2) = \text{rep}^V_x(X_2) \). Observe that \( \text{cc}(X, i) = \text{cc}(W, i) = \{\{R_1, \overline{R_1}, \overline{R_2}\}, \{R_2, \overline{R_3}\}, \{R_4\}\} \). Consequently, \( X \) and \( W \) are \( i \)-equivalent.
Figure 15: The graphs $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$ and $G[W \cup Y]_{\text{cc}(W \setminus S) \cup P_Y}$. As $X$ and $W$ are $i$-equivalent and $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$ is a forest, by Lemma 17, $G[W \cup Y]_{\text{cc}(W \setminus S) \cup P_Y}$ is also a forest. Thus, by Lemma 15, $G[W \cup Y]$ is an $S$-forest. In fact, these two lemmas imply that for every complement solution $(Y', P'_Y)$ associated with $i$, $G[X \cup Y']$ is an $S$-forest if and only if $G[W \cup Y']$ is an $S$-forest.
To prove Lemma 17, we prove that if $G[W \cup Y]_{\text{cc}(W \setminus S) \cup P_Y}$ contains a cycle, then $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$ contains a cycle. We use the following arguments.

The orange paths exist in both graphs because these paths only use blocks in $Y_{\downarrow P_Y}$. Since $\text{cc}(X,i) = \text{cc}(W,i)$, the purple path in $G[W \cup Y]_{\text{cc}(W \setminus S) \cup P_Y}$ implies the existence of the purple path in $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$. Finally, as $X_2 \equiv^W_2 W_2$, the blocks in $Y_{\downarrow P_Y}$ adjacent to $W_2$ are also adjacent to $X_2$. Thus, the green path in $G[W \cup Y]_{\text{cc}(W \setminus S) \cup P_Y}$ implies the existence of the green path in $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$. 

Figure 16: New examples for the graphs $G[X \cup Y]_{\text{cc}(X \setminus S) \cup P_Y}$ and $G[W \cup Y]_{\text{cc}(W \setminus S) \cup P_Y}$. 