A More Fine-Grained Complexity Analysis of Finding the Most Vital Edges for Undirected Shortest Paths

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Abstract

We study the NP-hard Shortest Path Most Vital Edges problem arising in the context of analyzing network robustness. For an undirected graph with positive integer edge lengths and two designated vertices s and t, the goal is to delete as few edges as possible in order to increase the length of the (new) shortest st-path as much as possible. This scenario has been studied from the viewpoint of parameterized complexity and approximation algorithms. We contribute to this line of research by providing refined computational tractability as well as hardness results. We achieve this by a systematic investigation of various problem-specific parameters and their influence on the computational complexity. Charting the border between tractability and intractability, we also identify numerous challenges for future research.

1 Introduction

Finding shortest paths in graphs is arguably among the most fundamental graph problems. We study the case of undirected graphs with positive integer edge lengths within the framework of "most vital edges" or (equivalently) "interdiction" or "edge blocker" problems. That is, we are interested in the scenario where the goal is to delete (few) edges such that in the resulting graph the shortest st-path gets (much) longer. This is motivated by applications in investigating robustness and critical infrastructure in the context of network design. Our results provide new insights with respect to classical, parameterized, and approximation complexity of this fundamental edge deletion problem which is known to be NP-hard [3, 23]. In its decision version, the problem reads as follows.

Shortest Path Most Vital Edges (SP-MVE)

Input: An undirected graph G = (V, E) with positive edge lengths : $E \mathbb{N}$, two vertices s, t V, and integers k, \mathbb{N} .

Question: Is there an edge subset S E, |S| k, such that the length of a shortest st-path in G – S is at least ?

We set $b := -\text{dist}_G(s, t)$ to be the number by which the length of every shortest st-path shall be increased. If all edges have length one, then we say that the graph has unit-length edges. Naturally, SP-MVE comes along with two optimization versions: Either delete as few edges as possible in order to achieve a length increase of at least b (called Min-Cost SP-MVE) or obtain a maximum

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length increase under the constraint that k edges can be deleted (called Max-Length SP-MVE). For an instance of SP-MVE or Max-Length SP-MVE we assume that k is smaller than the size of any st-edge-cut in the input graph. Otherwise, removing all edges of a minimum-cardinality st-edge-cut (which is polynomial-time computable) would lead to a solution disconnecting s and t.

Related work. Due to the immediate practical relevance, e.g. in supply [14, 16] and communication [22] networks, there are numerous studies concerning "most vital edges (and vertices)" and related problems. We focus on shortest paths, but there are further studies for problems such as Minimum Spanning Tree [4, 5, 13, 19, 27] or Maximum FI ow [19, 31, 36], to mention only two. With respect to shortest path computation, the following is known.

First, we mention in passing that a general result of Fulkerson and Harding [14] implies that allowing the subdivision of edges instead of edge deletions as modification operation makes the problem polynomial-time solvable. Notably, it also has been studied to find *one* most vital edge of a shortest path; this can be solved in almost linear time [29].

Bar-Noy et al. [3] showed that SP-MVE is NP-complete. Khachiyan et al. [23] found polynomialtime constant-factor inapproximability results for both optimization versions. For the case of directed graphs, Israeli and Wood [21] provided heuristic solutions based on mixed-integer programming together with experimental results. Pan and Schild [31] studied the restriction of the directed case to planar graphs and again obtained NP-hardness results.

Baier et al. [2] studied a minimization variant of SP-MVE where edges, in addition to a length value, also have a deletion cost associated with them. They refer to this problem as Minimum Length-Bounded Cut (MLBC) and showed that it is NP-hard to approximate within a factor of 1.1377 for 5. Moreover, they developed a polynomial-time algorithm for the special case of b = 1. Further, they showed that MLBC with general edge-costs and edge-lengths remains NP-hard on series-parallel and outerplanar graphs.

Golovach and Thilikos [17] studied SP-MVE with unit-length edges under the name Bounded Edge Undirected Cut (BEUC) from a parameterized complexity point of view. They showed that SP-MVE with unit-length edges is W[1]-hard with respect to k and that it is fixed-parameter tractable with respect to the combined parameter (k_{1}) . Answering an open question of Golovach and Thilikos [17], Fluschnik et al. [12] showed that SP-MVE with unit-length edges does not admit a polynomial-size problem kernel with respect to (k,), unless NP coNP/poly. Moreover, the latter showed that SP-MVE remains NP-hard on planar graphs. Dvořák and Knop [10] also studied SP-MVE with unit-length edges. They showed that the problem is W[1]-hard with respect to pathwidth. On the positive side, they showed that the problem is fixed-parameter tractable with respect to the treedepth of the input graph and with respect to and the treewidth tw of the input graph combined. Upon the latter, they proved that SP-MVE does not admit a polynomialsize problem kernel with respect to (, tw), unless NP coNP/poly. Moreover, they developed an algorithm running in n^{O(tw²)} time, that is, they showed that the problem lies in the complexity class XP when parameterized by tw. Kolman [24] studied SP-MVE and its vertex deletion variant. He proved that both variants on planar graphs are fixed-parameter tractable when parameterized by . Additionally, for the vertex-deletion variant, he developed an O(tw log tw)-approximation algorithm, which improves to a tw-approximation algorithm when the tree decomposition is given.

Our results. We perform an extensive study of multivariate complexity aspects [11, 30] of SP-MVE. More specifically, we perform a refined complexity analysis in terms of how certain problem-specific parameters influence the computational complexity of SP-MVE and its optimization variants. The parameters we study include aspects of graph structure as well as special restrictions on the problem parameters. We also report a few findings on (parameterized) approximability. Let us feature three main conclusions from our work: First, it is known that harming the network only a little bit (that is, b = 1) is doable in polynomial time [2] while we show that harming the network slightly more (that is, b = 2) becomes NP-hard. Second, the "cluster vertex deletion number", advocated by Doucha and Kratochvíl [9] as a parameterization between vertex cover number and clique-width, currently is our most interesting parameter that yields



Figure 1: The parameterized complexity of SP-MVE with unit-length edges with respect to di erent graph parameters. Herein, "distance to X" denotes the number of vertices that have to be deleted in order to transform the input graph into a graph of the graph class X. For two parameters that are connected by a line, the upper parameter is weaker (that is, larger) than the parameter below [25]. In the later sections we will only define the graph parameters that we directly work with. Refer to Sorge and Weller [33] for formal definitions of all parameters.

	k	
related to polynomial time	ХР	NP-hard for b = 2 and = 9 -approximation
related to fpt time	n/2 ^{O(log n)} -approximation for unit-length edges	r(n)-approximation for every increasing r
	fpt with respect to combined parameter (k,)	

Table 1: Overview on the computational complexity classification of SP-MVE on n-vertex graphs.

fixed-parameter tractability for SP-MVE with unit-length edges. Third, with general edge-lengths SP-MVE remains NP-hard even on complete graphs. Figure 1 surveys our current understanding of the parameterized complexity of SP-MVE with respect to a number of well-known graph parameters, identifying numerous open questions. Moreover, towards the goal of spotting further fixed-parameter tractable special cases, it also suggests to look for reasonable parameter combinations. In addition, Section 1 overviews our exact and approximate complexity results for SP-MVE. Figure 2 summarizes our understanding of the complexity of SP-MVE with unit-length edges on several graph classes.

Organization of the paper. After introducing some preliminaries in Section 2, we prove in Section 3 our NP-hardness results. In Section 4, we present our polynomial-time solvable special cases. In Section 5, we provide parameterized and approximation algorithms for SP-MVE. Conclusions and open questions are provided in Section 6.



Figure 2: Computational complexity of SP-MVE with unit-length edges for some graph classes. For SP-MVE with unit-length edges on proper interval graphs, we conjecture that it is solvable in polynomial time.

2 Preliminaries

For an undirected graph G = (V, E) we set n := |V| and m := |E|. A path P of length r - 1 in G is a sequence of distinct vertices $P = v_1 \cdot v_2 \cdot \ldots \cdot v_r$ with $\{v_i, v_{i+1}\}$ E for all $i \{1, \ldots, r-1\}$; the vertices v_1 and v_r are the endpoints of the path. For $1 \quad i < j \quad r$, we set $v_i P v_j$ to be the subpath of P starting in v_i and ending in v_j , formally $v_i P v_j := v_i \cdot v_{i+1} \cdot \ldots \cdot v_j$. For i = 1 or j = r we omit the corresponding endpoint, that is, we set $P v_j := v_1 P v_j$ and $v_i P := v_i P v_r$. For u, v = V, a uv-path P is a path with endpoints u and v. The distance between u and v in G, denoted by dist_G(u, v), is the length of a shortest uv-path. The diameter of G is the length of the longest shortest path in G.

For v V let $N_G(v)$ be the set of neighbors of v and let $N_G[v] = N_G(v)$ {v} be v's closed neighborhood. Two vertices u, v = V are called *true twins* if $N_G[u] = N_G[v]$ and *false twins* if $N_G(u) = N_G(v)$ but $N_G[u] = N_G[v]$; they are called *twins* if they are either true or false twins. We denote by G - S the graph obtained from G by removing the edge subset S E. For s, t V, an edge subset S is called st-cut if G - S contains no st-path. For V V let G[V] denote the subgraph induced by V. For E E let G[E] denote the subgraph consisting of all endpoints of edges in E and the edges in E.

Parameterized complexity. A parameterized problem consisting of input instance I and parameter k is called *fixed-parameter tractable* (fpt) if there is an algorithm that decides any instance (I, k) in $f(k) \cdot |I|^{O(1)}$ time for some computable function f solely depending on k, where |I| denotes the size of I. On the contrary, the parameterized complexity class XP contains all parameterized problems that can be solved in $|I|^{f(k)}$ time; in other words, membership in XP means polynomial-time solvability when the parameter value is a constant.

A core tool in the development of fixed-parameter tractability results is polynomial-time preprocessing by data reduction, called *kernelization* [18, 26]. Here, the goal is to transform a given problem instance (I, k) in polynomial time into an equivalent instance (I, k) whose size is upperbounded by a function of k. That is, (I, k) is a yes-instance if and only if (I, k) with |I|, k g(k) for some function g is a yes-instance. Thus, such a transformation is a polynomial-time selfreduction with the constraint that the reduced instance is "small" (measured by g(k)). If such a transformation exists, then I is called (*problem*) *kernel* of size g(k).



Figure 3: A schematic representation of the graph G constructed from the tripartite graph $G = (V_1 \ V_2 \ V_3, E)$. The vertices are grouped into the described sets. The edges in the picture correspond to edge sets in G and cover the incidence structure of the displayed vertices in G. A bold edge indicates an edge-gadget and the corresponding number denotes its length.

Approximation. Given an NP optimization problem and an instance I of this problem, we use opt(I) to denote the optimum value of I and val(I,S) to denote the value of a feasible solution S of instance I. The *approximation ratio* of S (or *approximation factor*) is $r(I,S) = \max \frac{val(I,S)}{opt(I)}, \frac{opt(I)}{val(I,S)}$. For a function , an algorithm A is a (|I|)-*approximation* if for every instance I of the problem, it returns a solution S such that r(I,S) (|I|). If the problem comes with a parameter k and the algorithm A runs in $f(k) \cdot |I|^{O(1)}$ time, then A is called *parameterized* (|I|)-*approximation*.

3 NP-hardness results

In this section, we provide several hardness results for restricted variants of SP-MVE. We start by adapting a reduction idea due to Khachiyan et al. [23] for the vertex deletion variant of SP-MVE. We prove that SP-MVE is NP-hard even for constant values of b, , and the diameter of the input graph.

Theorem 1. SP-MVE is NP-hard, even for unit-length edges, b = 2, = 9, and diameter 8.

Proof. As Khachiyan et al. [23, Theorems 8 and 11], we reduce from the NP-hard [15, GT1] problem Vertex Cover on tripartite graphs, where the question is, given a tripartite graph $G = (V = V_1 \quad V_2 \quad V_3, E)$ and an integer h 0, whether there is a subset V V with |V| h such that $G[V \setminus V]$ contains no edge. While the fundamental approach remains the same, the technical details when moving their vertex deletion scenario to our edge deletion scenario change to quite some extent. We refrain from a step-by-step comparison. Given a Vertex Cover instance (G, h) with G = (V_1 V_2 V_3, E) being a tripartite graph on n vertices, we construct an SP-MVE instance I = (G, k,) as follows. First, let k := h and := 9. The graph G = (V, E) contains vertices $V = V_1 \quad V_2 \quad V_3 \quad V_2 \quad \{s, t\}$, where s and t are two new vertices, and for each v V_2 we add a copy v V_2 .

Before describing the edge set E, we introduce edge-gadgets. Here, by adding a lengthedge-gadget $e_{u,v}$, 2, from the vertex u to vertex v, we mean to add n vertex-disjoint paths of length -2 and to make u adjacent to the first vertex of each path and v adjacent to the last vertex of each path. If = 2, then each path is just a single vertex which is at the same time the first and last vertex. The idea behind this is that one will never delete edges in an edge-gadget.

We add the following edges and edge-gadgets to G (see Figure 3 for a schematic representation of the constructed graph). For each vertex $v = V_2$ we add the edge {v, v} between v and its copy v. For each vertex $v = V_1$, we add the edge {s, v}, and for each vertex $v = V_3$, we add the edge {v, t}.

We also add the following edge-gadgets: For each edge $\{u, v\}$ $(V_1 \times V_2)$ E we add the edge-gadget $e_{u,v}$ of length two, for each edge $\{u, v\}$ $(V_2 \times V_3)$ E we add the edge-gadget $e_{u,v}$ of length two, where $u = V_2$ is the copy of u, and for each edge $\{u, v\}$ $(V_1 \times V_3)$ E we add the edge-gadget $e_{u,v}$ of length five. Furthermore, we add edge-gadgets of length four between s and every vertex $v = V_2$ and between t and every vertex $v = V_2$. Observe that we have dist_G (s, t) = 7 and thus $b = -dist_G$ (s, t) = 2.

We now show that G has a vertex cover of size at most h if and only if deleting k = h edges in G results in s and t having distance at least = 9.

":" Let V V be a vertex cover of size at most h in G. Consider the edge sets $E_1 := \{\{s, v\} : v \quad V_1 \quad V \}, E_2 := \{\{\{v, v\} : v \quad V_2 \quad V \quad , v \quad V_2 \text{ copy of } v\}, \text{ and } E_3 := \{\{v, t\} : v \quad V_3 \quad V \}.$ We claim that for the set

$$\mathbf{E} = \mathbf{E}_1 \quad \mathbf{E}_2 \quad \mathbf{E}_3$$

it holds that dist_G $_{-E}$ (s, t) 9 and |E| = |V| h. Clearly, |E| = |V| h. Suppose towards a contradiction that dist_G $_{-E}$ (s, t) < 9. Let P be an st-path of length less than nine. Observe that P contains an edge connecting s with some vertex in V₁ or an edge connecting t with some vertex in V₃. We discuss only the first case, as the second follows by symmetry.

Let P contain an edge connecting s with vertex u in V_1 . Path P contains either (i) a subpath of length three to vertex in V_2 , or (ii) a subpath of length five to a vertex in V_3 .

Case (i): Let $u - a_1 - a_2 - v - v$, with $v = V_2$ and $v = V_2$ the copy of v, be a subpath of P, where a_1, a_2 are vertices in an edge-gadget $e_{u,v}$. Then $\{u, v\} = E$ and u, v = V, as $\{s, u\} = E_1$ and $\{v, v\} = E_2$, contradicting that V is a vertex cover of G.

Case (*ii*): Let $u - a_1 - ... - a_5 - v$ with $v = V_3$ be a subpath of P, where $a_1, ..., a_5$ are vertices in an edge-gadget $e_{u,v}$. As P is of length less than 9, it follows that $P = s - u - a_1 - ... - a_5 - v - t$. Then $\{u, v\} = E$ and u, v = V, contradicting that V is a vertex cover of G.

" :" Let E E be a set of edges such that dist_{G -E} (s,t) 9 and |E | h. If E contains edges from an edge-gadget $e_{u,v}$, then it must contain at least n edges from this gadget in order to have a chance to increase the solution value. Therefore, since h < n, we can assume that E does not contain any edge contained in an edge-gadget. Thus, E $({s} \times V_1) (V_2 \times V_2) (V_3 \times {t})$. We construct a vertex cover V for G as follows: For each edge ${s,v}$ E it follows that v V_1 and we add v to V. Similarly, for each edge ${v, t}$ E it follows that v V_3 and we add v to V. Finally, for each edge ${v, v}$ E $(V_2 \times V_2)$, we add v to V.

Suppose towards a contradiction, that V is not a vertex cover in G, that is, there exists an edge {u, v} E with u, v / V. If v V₁ and u V₂, then the st-path s-v-u-u -t of length 8 < is contained in G – E. If v V₁ and u V₃, then the st-path s-v-u-t of length 7 < is contained in G – E. Finally, if v V₂ and u V₃, then the st-path s-v-v -u-t of length 8 < is contained in G – E. Each of the three cases contradicts the assumption that dist_{G – E} (s, t) 9.

Baier et al. [2] showed that SP-MVE is polynomial-time solvable for the special case of b = 1. Theorem 1 shows that this result cannot be extended to larger values of b. Regarding the diameter of the input graph, the statement of Theorem 1 will be strengthened later: Considering the problem with unit-length edges, we show that it remains NP-hard on graphs of diameter three (Theorem 4), while it becomes polynomial-time solvable on graphs of diameter two (Proposition 1). For arbitrary edge lengths, we show that the problem remains NP-hard on graphs of diameter one (Theorem 5).

When allowing length zero edges, Khachiyan et al. [23] stated that it is NP-hard to approximate Max-Length SP-MVE within a factor smaller than two. We consider in this paper only positive edge lengths and, by adapting the construction given in the above proof by considering edge-gadgets of lengths polynomial in n (with high degree), we obtain the following.

Theorem 2. Unless P = NP, Max-Length SP-MVE is not 4/3 - 1/poly(n)-approximable in polynomial time, even for unit-length edges.

Proof. We construct a gap-reduction [1] from Vertex Cover on tripartite graphs to Max-Length SP-MVE. More specifically, we use a gap-reduction from a decision problem to a maximization problem. A decision problem is called *gap-reducible* to a maximization problem



Figure 4: A schematic representation of the graph G constructed from the tripartite graph $G = (V_1 \ V_2 \ V_3, E)$. The vertices are grouped to the used sets. The edges in the picture correspond to edge sets in G and cover the incidence structure of the displayed vertices in G. A bold edge indicates an edge-gadget and the corresponding number denotes its length.

with gap (|I|) > 1 if for any instance I of we can construct an instance I of Q in polynomial time while satisfying the following properties for some function $c : \mathbb{N} \cap \mathbb{R}$ (0, +).

- If I is a yes-instance, then opt(I) c(|I|).
- If I is a no-instance, then opt(I) < $\frac{c(|I|)}{(|I|)}$.

The idea behind a gap-reduction is that if is NP-hard then is not approximable within a factor provided that P = NP.

Starting with an instance (G = (V, E), h) of Vertex Cover on tripartite graphs we construct an instance I = (G = (V, E), k, s, t) of Max-Length SP-MVE as in the proof of Theorem 1. We only change some lengths as follows (see also Figure 4): For each edge {u,v} (V₁ × V₂) E we add the edge-gadget $e_{u,v}$ of length x, for each edge {u,v} (V₂ × V₃) E we add the edgegadget $e_{u,v}$ of length x, where u V₂ is the copy of u, and for each edge {u,v} (V₁ × V₃) E we add the edge-gadget $e_{u,v}$ of length 3x. We add edge-gadgets of length 2x between s and every vertex v V₂ and between t and every vertex v V₂. The value x could be any polynomial function in |V| = n. Observe that we have dist_G (s, t) 3x + 2.

We now show that if G has a vertex cover of size at most h, then opt(I) = 4x + 1, otherwise opt(I) = 3x + 2.

Since V is not a vertex cover in G, there exists an edge $\{u, v\}$ E with u, v / V. If $v V_1$ and $u V_2$, then the st-path s-v-u-u -t of length 3x + 2 is contained in G – E. If $v V_1$ and $u V_3$, then the st-path s-v-u-t of length 3x + 2 is contained in G – E. Finally, if $v V_2$ and $u V_3$, then the st-path s-v-v -u-t of length 3x + 2 is contained in G – E.

Since Vertex Cover is NP-hard on tripartite graphs [15, GT1], Max-Length SP-MVE is not $\frac{4x+1}{3x+2} = 4/3 - 1/$ poly(n)-approximable in polynomial time.

Concerning special graph classes, we can show that the problem remains NP-hard on restricted bipartite graphs. To formulate our result, we need the graph parameter degeneracy. A graph G

has *degeneracy* d if every subgraph of G contains a vertex of degree at most d. By subdividing every edge, we obtain the following.

Theorem 3. SP-MVE is NP-hard, even for bipartite graphs with degeneracy two, unit-length edges, b = 4, = 18, and diameter 8.

Proof. We provide a self-reduction from SP-MVE with unit-length edges with b = 2, = 9, and diameter 8. Let I = (G = (V, E), k, ., s, t) be the given SP-MVE instance. We construct an instance I = (G, k, 2, s, t) where G is obtained from G by subdividing all edges, that is, each edge is replaced by a path of length two. The correctness of the reduction is easy to see as any minimal solution contains at most one edge of each of the introduced induced paths of length two. Clearly, I can be computed in polynomial time. Furthermore, G is bipartite and has degeneracy two.

We next prove that SP-MVE remains NP-hard on split graphs. A split graph is a graph whose vertex set can be partitioned into a clique and an independent set. Observe that a split graph has diameter at most three. Thus, the next theorem also shows NP-hardness on diameter-three graphs.

Theorem 4. SP-MVE is NP-hard on split graphs, even for unit-length edges.

Proof. We reduce from SP-MVE on general graphs. Let I := (G = (V, E), s, t, k,) be an instance of SP-MVE, recall n = |V|. We obtain the graph G = (V, E) from G by subdividing each edge of G, and subsequently turning V (G) into a clique. Formally, the graph G = (V, E) is defined through

$$\begin{split} V &:= V \quad (W := \{w_j^{\{u,v\}} \mid \{u,v\} \quad E,j \quad [n^2]\}), \\ E &= \frac{V}{2} \quad \{u,w_j^{\{u,v\}}\}, \{v,w_j^{\{u,v\}}\} \mid \{u,v\} \quad E,j \quad [n^2] \ . \end{split}$$

Observe that G is a split graph since G [W] forms an independent set and G [V] forms a clique. Let I := (G, s, t, k,) be an instance of SP-MVE on split graphs with $k = \frac{n}{2} + k \cdot n^2$ and := 2. We show that I is a yes-instance if and only if I is a yes-instance.

Let I be a yes-instance. Let S E(G) be such that G - S has no st-path of length smaller than . We claim that G - S with

S :=
$$\frac{V}{2}$$
 {u, w_j^{u,v}} | {u, v} S, j [n²]

does not have an st-path of length smaller than .

Note that $|S| = \frac{n}{2} + k \cdot n^2$. Suppose that there is an st-path P in G - S with |P| < 2. Then the vertices in P alternate between the vertices in V and W. By construction, if $\{v, w\}, \{w, u\} \in (P)$ with u, v V and w W, then the edge $\{u, v\}$ is present in G - S. Hence, consider the st-path P in G - S obtained from P by restricting P to V. It follows that |P| = |P|/2 < ..., a contradiction to the choice of S. Thus I is a yes-instance.

Conversely, let I be a yes-instance. Let S E(G) be minimal such that G - S has no st-paths of length smaller than . We claim that G - S with

$$S := \ \{u,v\} \ | \ \ w_j^{\{u,v\}} \quad \ W,e \quad S \ : w_j^{\{u,v\}} \quad e$$

does not have an st-path of length smaller than . If $\{u, w_j^{\{u,v\}}\}\$ S for some u, v V and j $[n^2]$, then for all i $[n^2]$, $w_i^{\{u,v\}}$ is incident to exactly one edge in S since S is minimal (otherwise S \{u, $w_j^{\{u,v\}}\}\$ is a smaller solution). Together with $|S| = \frac{n}{2} + k \cdot n^2 < (k + 1) \cdot n^2$ it follows that |S| < k + 1. Suppose there is an st-path P in G – S with |P| < . Then for each edge $\{u, v\} = E(P)$, there is a j $[n^2]$ such that $\{u, w_j^{\{u,v\}}\}$, $\{v, w_j^{\{u,v\}}\}\$ S. We construct an st-path P in G – S from P by replacing each edge $\{u, v\} = E(P)$ by two edges $\{u, w_j^{\{u,v\}}\}$, $\{v, w_j^{\{u,v\}}\}\$ S for some j $[n^2]$. Then $|P| = 2 \cdot |P| < 2 \cdot$, a contradiction to the choice of S. Thus I is a yes-instance. \Box

Note that SP-MVE can be solved on complete graphs with unit-length edges in polynomial time. If = 1, then the instance is trivially a yes-instance. If = 2, one edge deletion is necessary to obtain the desired distance. If > 2, then observe that for each vertex $v = V \setminus \{s, t\}$ the path s - v - t has length two and all these paths are edge-disjoint. Hence, to increase the distance between s and t to three, we have to delete n - 1 edges (the edge $\{s, t\}$ and one edge in each of the n - 2 paths of length two). However, with n - 1 edge deletions, one can delete all edges incident to s and disconnect s from t, so this solution works for all > 2. Thus, if > 2, then the instance is a yes-instance if and only if the number of edge-deletion is at least n - 1.

As soon as one deals with arbitrary edge lengths, however, the problem becomes NP-hard even on complete graphs.

Theorem 5. SP-MVE remains NP-hard on complete graphs.

Proof. We reduce from SP-MVE on general graphs. Let I := (G = (V, E), s, t, k,) be an instance of SP-MVE (w.l.o.g. let G not contain isolated vertices). Let G be the graph obtained from G by adding the edge set $E := \{\{v, w\} | \{v, w\} \in E\}$ and assigning length (e) := +1 to each edge e E. Observe that G is a complete graph. We claim that I := (G, s, t, k,) is a yes-instance of SP-MVE if and only if I is a yes-instance of SP-MVE.

By construction, G is isomorphic to G [E(G)]. This implies that for any S E(G), there is a bijection between the set of st-paths in G – S and the set of st-paths in G [E(G)] – S. Observe that every st-path in G using an edge in $E(G) \setminus E(G)$ has length greater than . Hence, if there is an S E(G) such that there is no st-path in G – S of length smaller than , then there is no st-path in G – S of length smaller than . \Box

4 Polynomial-time algorithms

In this section, we present three polynomial-time algorithms for special cases of SP-MVE.

We start with considering instances of SP-MVE on series-parallel graphs with s and t being the natural two terminals of the underlying two-terminal graph. Here, a two-terminal graph is a triplet containing a graph and two distinct vertices of the graph (the terminals). Every twoterminal series-parallel graph can be constructed by a sequence of parallel and serial compositions starting from single edges where the endpoints of an edge are the two terminals. Given two two-terminal series-parallel graphs G_1 and G_2 with terminals s_1 , t_1 and s_2 , t_2 respectively, then

- 1. G is a *serial composition* of G_1 and G_2 with terminals s_1, t_2 if G is the disjoint union of G_1 and G_2 where t_1 is identified with s_2 .
- 2. G is a *parallel composition* of G_1 and G_2 with terminals s, t if G is the disjoint union of G_1 and G_2 where s_1 is identified with s_2 and t_1 is identified with t_2 .

Moreover, we can construct for each two-terminal series-parallel graph G a so-called *sp-tree* in linear time [7, 35], a binary rooted tree representing the serial and parallel composition of two-terminal series-parallel graphs to obtain G. Herein, every leaf of the sp-tree is identified with an edge, and the label () of the leaf is the set of the endpoints of the edge. Moreover, each inner node of the sp-tree is labeled by either () = S or () = P, representing a serial or parallel composition, respectively.

Theorem 6. Min-Cost-SP-MVE can be solved in $O(m \cdot ^2)$ time on two-terminal series-parallel graphs with s and t being the two terminals.

Proof. Let (G = (V, E), s, t) be a two-terminal series-parallel graph with edge lengths specified by : E \mathbb{N} . Let (T,) be an sp-tree for G, where is the labeling of the nodes of T. We identify each node V(T) with a two-terminal series-parallel graph G induced by the subtree rooted at . Recall that if V(T) is the root of T, then G = G.

Let C[, x] denote the minimum number of edges to delete in G such that there is no path of length smaller than x connecting the two terminals. Observe that such an edge deletion set exists for every x \mathbb{N} , and its size is upper-bounded by the size of a minimum cut disconnecting the terminals.

Case 1: If V(T) is a leaf of T with $() = \{v, w\}$, then

$$C[, x] = \begin{cases} 1, & \text{if } (\{v, w\}) < x, \\ 0, & \text{otherwise.} \end{cases}$$

Correctness: In the graph $G = (\{v, w\}, \{\{v, w\}\})$, we have to delete the edge $\{v, w\}$ to increase the distance between v and w to x. This is possible if and only if $(\{v, w\}) < x$.

Case 2: If V(T) is an inner node of T with () = S and children ₁ and ₂, then

$$C[, x] = \min_{x \in \{0,...,x\}} (C[_1, x] + C[_2, x - x]).$$
(1)

Correctness: Let G , G , and G be the graphs corresponding to nodes , 1, and 2 respectively. Let v, w denote the terminals of G , and let v, u and u, w be the terminals of G and G respectively. Recall that G is the serial composition of G and G , thus G is obtained by identifying u with u as u, and setting v := v and w := w.

Let S E(G) be a set of C[, x] edges such that there is no vw-path of length smaller than x in G – S. Since G is the serial composition, $S = S_1 - S_2$ with $S_1 - E(G_1)$ and $S_2 - E(G_2)$. Then there is x $\{0, \ldots, x\}$ with dist_{G 1}-S₁(v, u) x and dist_{G 2}-S₂(u, w) x - x since every vw-path contains u. It follows that

$$C[, x] = |S| = |S_1| + |S_2| \min_{x \in \{0,...,x\}} (C[_1, x] + C[_2, x - x]).$$

Conversely, let x $\{0, ..., x\}$ be such that the expression in Equation (1) is minimum. Let $S_1 = E(G_1)$ and $S_2 = E(G_2)$ with $|S_1| = C[1, x]$ and $|S_2| = C[2, x - x]$ such that there is no v u -path of length smaller than x in $G_1 - S_1$ and no u w -path of length smaller than x - x in $G_2 - S_2$. Let $S := S_1 = S_2$. Since every vw-path in G contains the vertex u, it follows that dist_{G-S}(v, w) = dist_{G-S}(v, u) + dist_{G-S}(u, w) = x + x - x = x. It follows that

$$\min_{x \in \{0,...,x\}} (C[_1, x] + C[_2, x - x]) = |S_1| + |S_2| = |S| \quad C[_, x]$$

Case 3: If V(T) is an inner node of T with () = P, and children ₁ and ₂, then

$$C[, x] = C[_1, x] + C[_2, x].$$

Correctness: Let G , G , and G , be the graphs corresponding to nodes , 1, and 2, respectively. Let v, w denote the terminals of G , and let v, w and v, w be the terminals of G , and G , respectively. Recall that G is the parallel composition of G , and G , thus G is obtained by identifying v with v as v and w with w as w.

Let S E(G) be a set of C[, x] edges such that there is no vw-path of length smaller than x in G - S. Since G is the parallel composition, it holds that S = S₁ S₂ with S₁ $E(G_1)$ and S₂ $E(G_2)$. Observe that there is a vw-path of length smaller than x in G - S if and only if there is a v w -path or a v w -path of length smaller than x in G₁ - S₁ or in G₂ - S₂. The observation follows immediately from the definition of parallel compositions and the fact that v is identified with v as v and w is identified with w as w. It follows that

$$C[, x] = |S| = |S_1| + |S_2| C[_1, x] + C[_2, x].$$

Conversely, let $S_1 = E(G_1)$ and $S_2 = E(G_2)$ with $|S_1| = C[1, x]$ and $|S_2| = C[2, x]$ such that there is no v w -path of length smaller than x in $G_1 - S_1$ and no v w -path of length smaller than x in $G_2 - S_2$. Let $S := S_1 - S_2$. Following the preceding observation, we obtain

$$C[_1, x] + C[_2, x] = |S_1| + |S_2| = |S| - C[_, x].$$

We consider C as a table in the remainder. We fill C in post-order on T, that is, whenever the entries for an inner node are to be filled, the entries of the child nodes are filled before. By

the correctness of the cases above, if V(T) denotes the root of T, then C[,] denotes the minimum number of edge deletions such that there is no st-path in G of length smaller than .

Since every edge in G one-to-one corresponds to a leaf in T, there are O(m) nodes in T. Hence, the table C has O(m \cdot) entries. In Case 2, we have to find a minimum in O() time. Altogether, the algorithm takes O(m \cdot ²) time.

Remark 1. With a similar dynamic programming approach one can show an algorithm solving Max-Length SP-MVE in $O(m \cdot k^2)$ time, see Stahlberg [34, Theorem 8.4] for details. Furthermore, both the $O(m \cdot ^2)$ -time algorithm above and the $O(m \cdot k^2)$ -time algorithm extend to the case where the edges have integral edge-deletion costs. This problem variant with both edge-deletion costs and edge lengths was shown to be (weakly) NP-hard on series-parallel graphs with s and t being the two terminals by Baier et al. [2]. The two algorithms above complement this with fixed-parameter tractability with respect to each k and .

In Theorem 4 we showed that SP-MVE with unit-length edges on split graphs remains NPhard. Since split graphs are of diameter at most three, SP-MVE with unit-length edges remains NP-hard on graphs of diameter at least three. The last result of this section shows that this bound on the diameter is strict.

Proposition 1. SP-MVE with unit-length edges is linear-time solvable on graphs of diameter at most two.

Proof. Itai et al. [22] proved that for 4, SP-MVE with unit-length edges is solvable in polynomial time. Hence, it remains to consider the case where 5.

Payne et al. [32] showed that in any graph H of diameter two, for each pair of distinct vertices v, w = V(H), there are min{deg(v), deg(w)} many edge-disjoint paths of length at most four. Hence, to achieve a distance of five or more between s and t we have to delete min{deg(s), deg(t)} edges, which is su cient to cut s from t. Thus, any instance (G, s, t, k,) with 5 and G being a graph of diameter two is a yes-instance if and only if k min{deg(s), deg(t)}. This can be decided in linear time.

Observe that each connected component of a *cograph* (a graph without an induced P_4) has diameter two. Note that threshold graphs are cographs. Thus, the preceding result also shows that SP-MVE with unit-length edges is linear-time solvable on cographs and threshold graphs.

5 Algorithms for some NP-hard cases

In this section, we present fixed-parameter and approximation algorithms. First, we consider bounded-degree graphs. Here, the basic observation is that the maximum vertex degree of a graph upper-bounds the number of deleted edges for SP-MVE: a budget of would allow to disconnect s from t by deleting all edges incident to s.

Proposition 2. SP-MVE can be solved in $O(m^{-1}(m + n \log n))$ time.

Proof. Recall that we assume k to be smaller than the maximum degree as otherwise we could simply delete all edges incident to s. The straightforward algorithm branching into all $O(m^k)$ cases to delete at most k edges and checking with Dijkstra's shortest path algorithm whether the distance between s and t is high enough runs in $O(m^k(m + n \log n)) = O(m^{-1}(m + n \log n))$ time.

The question whether one can replace m⁻¹ by $f() \cdot m^{O(1)}$ for some function f, that is, whether SP-MVE is not only in XP (as shown by Proposition 2) but also fixed-parameter tractable with respect to , remains open.

Golovach and Thilikos [17] used a search tree algorithm to show that SP-MVE is fixedparameter tractable when combining the parameters number k of removed edges and minimum st-path length to be achieved. We next state the result and describe the search tree since we will adapt it in the following. **Proposition 3** (Golovach and Thilikos [17]). SP-MVE can be solved in $O((-1)^k \cdot (n \log n + m))$ time.

Proof. We employ a simple depth-bounded search tree: the basic idea is to search for a shortest st-path and to "destroy" it by deleting one of the edges (trying all possibilities). This is repeated until every shortest st-path has length at least . For each such shortest path, we branch into at most -1 possibilities to delete one of its edges, and the depth of the corresponding search tree is at most k (our "deletion budget") since otherwise we cannot find a solution with at most k edge deletions. The correctness is obvious. Hence, we arrive at a search tree of size at most (-1)^k where in each step we need to compute a shortest path. Using Dijkstra's shortest algorithm, this can be done in O(n log n + m) time. The overall running time is thus O((-1)^k (n log n + m)).

Using the search tree described in the proof of Proposition 3 to destroy all paths of length at most $2^{O(\log n)}$ yields the following.

Corollary 1. For any constant c, Max-Length SP-MVE with unit-length edges can be approximated within a factor of $n/2^{c}$. $\log n$ in $O(2^{k^2}k(n\log n + m) + n^{c^2+3})$ time.

Proof. We employ the search tree algorithm behind Proposition 3; it has size $O((-1)^k)$. The idea now is to either compute an optimal solution in fpt-time or to derive the stated approximation in polynomial time.

Our parameterized approximation algorithm works as follows. Trying = 1, 2, ..., g(n) (where g(n) is determined below) we employ the search tree to detect whether there is an optimal solution of length smaller than g(n). Namely, if the search tree for some -value says no, then we know that we found an optimal solution with the previous search tree and output this. Otherwise, we reach = g(n) and thus, since the optimal value is at most n - 1, this means that we have a factor-n/g(n)-approximation.

Overall, this procedure has at most g(n) iterations and each has a running time of $O(g(n)^k \cdot (n \log n + m))$. It remains to determine for which (maximum) function g(n) this still yields fpt running time for parameter k. First, if $k > \log(g(n))$, then $g(n)^k = 2^{k \cdot \log g(n)}$ can be upper-bounded by 2^{k^2} and we are done. Second, if $k = \log(g(n))$, then we have that $g(n)^k = g(n)^{\log(g(n))} = 2^{(\log(g(n))^2)}$. The latter term is polynomial if and only if $g(n) = 2^{O(-\log n)}$. More precisely, if for any constant c we have $g(n) = 2^{c \cdot -\log n}$, then we get the bound $2^{(\log(g(n))^2)} = n^{c^2}$. In total the running time in this second case is bounded by $O(n^{c^2+3})$.

By deleting every edge on too short st-paths, we obtain an -approximation.

Proposition 4. Min-Cost SP-MVE can be approximated within a factor of $in O(n^2 \log n + nm)$ time.

Proof. Let I = (G = (V, E), , s, t,) be an instance of Min-Cost SP-MVE. We repeat the following algorithm until the shortest st-path has length at least . Set G := G and let P be a shortest st-path in G. If the length (P) of P is less than , then set G := G - E(P) and proceed. Denote by i the number of iterations the algorithm realizes. Let E be the set of all edges of the i shortest paths removed from G. The size of E is |E| i since at each step at most edges are deleted. Moreover, opt(I) i since an optimal solution contains at least one edge of each of these i paths. The number of iteration is at most n and each iteration can be done in O(n log n + m) time.

Baier et al. [2, Corollary 3.14] provided a b-approximation algorithm for Min-Cost SP-MVE running in $O(b \cdot n \cdot m)$ time. Observe that our approximation algorithm in Proposition 4 provides a weaker approximation factor but a faster running time.

Combining the previous approximation algorithm with a tradeo between running time and approximation factor [6, Lemma 2], we obtain the following.

Corollary 2. For every increasing function r, Min-Cost SP-MVE is parameterized r(n)-approximable with respect to the parameter .

Parameter feedback edge set number. We next provide a linear-size problem kernel for SP-MVE parameterized by the feedback edge set number. An edge set F = E is called *feedback edge set* for a graph G = (V, E) if G - F is a tree or a forest. The feedback edge set number of G is the size of a minimum feedback edge set. Note that if G is connected, then the feedback edge set number equals m - n + 1. Computing a spanning tree, one can determine a minimum feedback edge set F with |F| = f for our input instance (G = (V, E), k, s, t, c). We start with two simple data reduction rules dealing with degree-one and degree-two vertices.

Rule 1. Let (G = (V, E), k, s, t,) be an SP-MVE instance and let $v \in V \setminus \{s, t\}$ be a vertex of degree one. Then, delete v.

The correctness of **Rule 1** is obvious as no shortest path uses a degree-one vertex. We deal with degree-two vertices as follows.

Rule 2. Let (G = (V, E), k, s, t,) be an SP-MVE instance and let $v = V \setminus \{s, t\}$ be a vertex of degree two with $N_G(v) = \{u, w\}$ and $\{u, w\} / E$. Then add the edge $\{u, w\}$ with the length $(\{u, w\}) := (\{u, v\}) + (\{v, w\})$ and delete v.

The correctness of **Rule 2** follows from the fact that on an induced path at most one edge will be deleted and it does not matter which one will get deleted. Applying both rules exhaustively can be done in linear time and leads to the following problem kernel.

Theorem 7. SP-MVE admits a linear-time computable problem kernel with 5f + 2 vertices and 6f + 2 edges.

Proof. Let (G = (V, E), k, s, t,) be the input instance of SP-MVE. First, we exhaustively apply Rules 1 and 2. It remains to upper-bound the size of the reduced graph G. To this end, first observe that G contains at most f degree-two vertices as every degree-two vertex that is not deleted by Rule 2 has two neighbors that are adjacent to each other and thus induces together with its neighbors a cycle. It remains to upper-bound the number of vertices with degree at least three. To this end, let r denote the number of leaves in the tree G - F. Thus, G - F contains at most r - 2 vertices of degree at least three. Due to Rule 1, G contains at most two degree-one vertices (s and t) and, hence, r - 2f + 2. Furthermore, there are at most 2f degree-three vertices in G that are incident to an edge in F. Hence, G contains at most 4f + 2 vertices of degree at least three. In total, G contains at most 5f + 2 vertices and, thus, 6f + 2 edges.

We now discuss the running time. To apply the rules, start with sorting the vertices by degree in non-decreasing order. Since all degrees are smaller than n, the sorting can be done in O(n) time using e.g. Bucket sort. Then, deleting all degree-one vertices and updating their neighbors' degrees can be done in linear time. Similarly, once Rule 1 is no more applicable, the degree-two vertices can be dealt with in similar fashion. Note that applying Rule 2 does not change the degrees of the neighbors of the degree-two vertex. Thus, for each degree-two vertex removing it and adding the extra edge can be done in constant time. Hence, the overall time to apply both rules is linear.

By simply trying all possibilities to delete edges in the problem kernel and checking with Dijkstra's algorithm the distance between s and t, we obtain the following.

Corollary 3. SP-MVE can be solved in $O(2^{6f} (n \log n + m))$ time where f is the feedback edge set number.

Parameter cluster vertex deletion number. We now prove that SP-MVE restricted to unitlength edges is fixed-parameter tractable with respect to the parameter cluster vertex deletion number x. A graph G is a *cluster graph* if it is a disjoint union of cliques. A vertex set X V is called *cluster vertex deletion set* if G[V \ X] is a cluster graph [20]. The cluster vertex deletion number is the size of a minimum cluster vertex deletion set. Recall that SP-MVE with arbitrary edge lengths is NP-complete on complete graphs (see Theorem 5). Thus, the algorithm presented below for the unit-length case cannot be extended to the more general case with arbitrary edge lengths since a clique has cluster vertex deletion number zero.

We assume in the following that for the input instance (G = (V, E), k, s, t) we are given a cluster vertex deletion set X of size x. If X is not already given, then we can compute X in $O(1.92^{x} \cdot (n + m))$ time [8]. Our algorithm is based on the observation that twins can be handled equally in a solution. This follows from a more general statement provided in the following lemma. It shows that for any set T V \ {s, t} of vertices that have the same neighborhood in V \ T, we can assume that we do not delete edges in G[T] and that the vertices in T behave the same, that is, one deletes either all edges or no edge between a vertex v V \ T and the vertices in T.

Lemma 1. Let G = (V, E) be an undirected graph with unit-length edges, let s, t = V be two vertices, and let $T = \{v_1, \ldots, v_t\} = V \setminus \{s, t\}$ be a set of vertices such that $N_G(v_1) \setminus T = N_G(v_2) \setminus T = \ldots = N_G(v_t) \setminus T$. Then, for every edge subset S = E, there exists an edge subset S = E such that dist_{G-S} (s, t) = dist_{G-S} (s, t), |S| = |S|, and $N_{G[S]}(v_1) = N_{G[S]}(v_2) = \ldots = N_{G[S]}(v_t)$.

Proof. Starting from the edge subset S E, we construct S having the desired properties. To this end, we abbreviate $:= dist_{G-S}(s,t)$. Let T V \{s,t} be a set of vertices such that N(u) \T = N(v) \T for each pair u, v T. Assume that the vertices in T do not have the same neighborhood in G[S]; otherwise, we simply set S := S. Let u T be a vertex that has in the graph (V, S) the smallest degree of all vertices in T, that is, the vertex in T that is incident to the least number of edges in S. Now, construct S as follows. First, initialize S as a copy of S. Second, remove all edges of S that have both endpoints in T. Third, for each v T \ {u} remove all edges incident to v from S and add for each edge {u, w} S the edge {v, w}. Summarizing, S is composed as follows:

$$\begin{split} S &:= (S \setminus \{\{v,w\} \mid v \quad T \setminus \{u\} \quad w \quad V \}) \\ & \quad \{\{v,w\} \mid v \quad T \setminus \{u\} \quad w \quad V \setminus T \quad \{u,w\} \quad S\}. \end{split}$$

By construction of S we have |S| = |S|. Furthermore, we have $N_{G[S]}(v) = N_{G[S]}(u) \setminus T$ for all v T and thus $N_{G[S]}(v) = N_{G[S]}(v)$ for each pair v, v T. It remains to show that in G - S the distance between s and t is at least . To this end, assume by contradiction that G - S contains an st-path P of length less than . Since, by construction of S, each edge in S \ S has at least one endpoint in T, it follows that P contains at least one vertex of T. Let v and v be the first respectively last vertex of T on P (possibly v = v) and let w, w be the vertices before v respectively after v on P, that is

$$\mathsf{P} = \mathsf{s} \cdot \ldots \cdot \mathsf{w} \cdot \mathsf{v} \cdot \ldots \cdot \mathsf{v} \cdot \mathsf{w} \cdot \ldots \cdot \mathsf{t}.$$

Since w, w / T, $N_G(v) \setminus T = N_G(v) \setminus T$, and $N_{G[S]}(v) = N_{G[S]}(v)$, it follows that P w-v-w P is also an st-path with length less than in G - S. Similarly, it follows that P := Pw-u-w P is also an st-path with length less than in G - S (where u is the vertex used in the construction of S). Since $N_{G[S]}(u) = N_{G[S]}(u) \setminus T$ it follows that $\{u, w\}, \{u, w\} / S$, implying that P is an st-path of length less than in G - S; a contradiction to the assumption that dist_{G-S}(s, t) = .

Using Lemma 1 we can show that SP-MVE with unit-length edges is linear-time fixed-parameter tractable with respect to the parameter cluster vertex deletion number.

Theorem 8. SP-MVE with unit-length edges can be solved in $2^{2^{O(x)}}(n + m)$ time where x is the cluster vertex deletion number.

Proof. Let (G = (V, E), k, s, t) be the input instance of SP-MVE and let X V be a cluster vertex deletion set of size x. Hence, G - X is a cluster graph and the vertex sets C_1, \ldots, C_r form the cliques (clusters) for some r \mathbb{N} . We set $C := \{C_1, \ldots, C_r\}$. Assume that there is an SP-MVE solution S E of size at most k; otherwise the algorithm will output 'no' as it finds no solution. We describe an algorithm that finds S.

Our algorithm is based on the following observation. Let P be an arbitrary shortest st-path that goes through a clique C C in G – S. Then, P contains at most 2^x vertices from C: By Lemma 1, we can assume that the twins in G are still twins in G – S. Since P is a shortest path, P does not contain two vertices that are twins. As the vertices in C form a clique, they only di er in how they are connected to vertices in X. Thus, C contains at most 2^x "di erent" vertices, that is, vertices with pairwise di erent neighborhoods.

Now, consider two non-adjacent vertices u, v = X. From the above considerations it follows that in G - S a uv-path avoiding the vertices in X has length between one and $2^x + 1$ as it can pass through at most one clique. Our algorithm tries for each vertex pair from X all possibilities for the distance it has in G - S and then tries to realize the current possibility. After the current possibility is realized, the cliques in C are obsolete and thus the instance size can be upper-bounded in a function of x. More precisely, our algorithm works as follows:

- 1. Branch into all possibilities to delete edges contained in G[X]. Decrease the budget k accordingly.
- Branch into all possibilities to add for each pair u, v of non-adjacent vertices in X an edge with a length lying in {2, 3, ..., 2^x, 2^x + 1, } and indicating the length of a shortest path between u and v that does not contain any vertex in X.
- Delete for each clique containing neither s nor t the *minimum number* of edges to ensure that a shortest path between each pair of vertices in X is completely contained in G[X]. Decrease the budget k accordingly.
- 4. Remove all cliques except the ones that contain s or t. Do not change the budget k.
- 5. Solve the problem on the remaining graph with the remaining budget (that was not spent in Steps 1 and 3).

Note that Step 2 is performed for each possibility in Step 1. Hence, in Steps 1 and 2 at most $2^{x^2} \cdot (2^x + 1)^{x^2}$ possibilities are considered and for each of these possibilities Step 3 is invoked.

In Step 3, the algorithm tries to realize the prediction made in Step 2. To this end, let C C be a clique containing neither s nor t. The algorithm branches into all possibilities to delete edges in G[C] or edges with one endpoint in C and the other endpoint in X. Since G[C] contains at most 2^x di erent vertices, it follows from Lemma 1 that at most $2^{(2^x)^2+2^x \cdot x} = 2^{(4^x)+2^x \cdot x}$ possibilities need to be considered to delete edges. For each possibility, the algorithm checks in $x^{O(1)}$ time whether all shortest paths between a pair of vertices of X go through C. If yes, then the algorithm discards the currently considered branch; if no, then the current branch is called valid. From all valid branches for C, the algorithm picks the one that deletes the minimum amount of edges and proceeds with the next clique. Observe that since X is a vertex separator for all cliques in C, the algorithm can solve Step 3 for each clique independently of the outcome in the other cliques. Hence, the overall running time for Step 3 is $2^{2^{O(x)}} \cdot n$ as |C| = n.

As discussed above, the cliques in C containing neither s nor t are now obsolete as there is always a shortest path avoiding these cliques. Hence, the algorithm removes these cliques (Step 4). This can be done in linear time. The remaining instance consists of the vertices in X and the at most two cliques containing s and t. As the algorithm deleted the edges within G[X] in Step 1, it remains to consider deleting edges within the two cliques or between the two cliques and the vertices in X. Again, by Lemma 1, the algorithm only needs to branch into $2^{2 \cdot (4^x + x \cdot 2^x)}$ possibilities to delete edges and check for each branch whether s and t have distance at least and the overall budget k is not exceeded. If one branch succeeds, then the algorithm found a solution and returns it. If no branch succeeds, then there exists no solution of size k since the algorithm performed an exhaustive search. Overall, the running time is $2^{2^{O(x)}} \cdot (n + m)$.

Obviously, it would be interesting to improve the above algorithm by obtaining linear-time fixed-parameter tractability with a single-exponential-time algorithm.

6 Conclusion

The Shortest Path Most Vital Edges (SP-MVE) problem is a natural edge deletion problem that is amenable to a rich body of fine-grained (multivariate) computational complexity analysis. Such a study has been initiated here, identifying numerous challenges for future work. Figure 1 in the introductory section depicts a wide range of graph parameters for which the parameterized complexity status of SP-MVE is unknown. Also concerning the approximation point of view not much is known. There is a huge gap between the known lower and upper bounds of the approximation factor achievable in polynomial time. Further, from a practical point of view it would make sense to extend our studies by restricting the input to planar graphs [12, 31]—here one might hope for further fixed-parameter tractability results. Moreover, the complexity of SP-MVE remains open even for highly structured graphs such as interval or proper interval graphs; we conjecture that SP-MVE is polynomial-time solvable on proper interval graphs [34]. Finally, also in terms of parameterized approximability [28] Shortest Path Most Vital Edges o ers a number of interesting challenges for future work.

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