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Article

Parameterised Enumeration for Modification Problems

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Abstract: Recently, Creignou et al. (Theory Comput. Syst. 2017) have introduced the class DelayFPT into parameterised complexity theory in order to capture the notion of efficiently solvable parameterised enumeration problems. In this paper, we propose a framework for parameterised *ordered* enumeration and will show how to obtain enumeration algorithms running with an FPT delay in the context of general modification problems. We study these problems considering two different orders of solutions, namely, lexicographic order and order by size. Furthermore, we present two generic algorithmic strategies. The first one is based on the well-known principle of self-reducibility and is used in the context of lexicographic order. The second one shows that the existence of a neighbourhood structure among the solutions implies the existence of an algorithm running with FPT delay which outputs all solutions ordered non-decreasingly by their size.

Keywords: Parameterised complexity; Enumeration; Bounded Search Tree; Parameterised Enumeration; Ordering

1. Introduction

The study of enumeration problems, that is, the task of generating all solutions of a given computational problem, finds a wealth of applications, e.g., in query answering in databases [1] and web search engines [2], bioinformatics [3] and computational linguistics [4]. From a complexity-theoretic viewpoint, the notion of DelayP, the class of problems whose instance solutions can be output in such a way that the delay between two outputs is bounded by a polynomial, is of high importance [5].

For many enumeration problems often it is central that the output solutions obey a given ordering: in many applications it is interesting to get the solutions with the smallest “cost” at the beginning. Enumerating all solutions in non-decreasing order allows to determine not only the smallest solution, but also the *k*th-smallest one. Also with such a generating algorithm, it is possible to find the smallest solution satisfying additional constraints in checking at each generation step whether these constraints are satisfied. The disadvantage of this method is that it cannot guarantee fast results because a long prefix of candidates may not satisfy them. However, this technique has the advantage to be applicable to any additional decidable constraint (see, e.g., [6]). Let us illustrate this with some examples.

The question for which classes of propositional CNF formulas enumerating all satisfying solutions is possible in DelayP, as defined above, was studied by Creignou and Hébrard [7]. In terms of the

30 well-known Schaefer framework for classification of Boolean constraint satisfaction problems, it was
31 shown that for the classes of Horn, anti-Horn, affine or bijunctive formulas, such an algorithm exists.
32 For any other classes of formulas, the existence of a DelayP-algorithm implies $P = NP$. It is interesting
33 to note that the result hinges on the self-reducibility of the propositional satisfiability problem. Since
34 variables are tried systematically first with an assignment 0 and then 1, it can be observed that the
35 given enumeration algorithms output all satisfying assignments in lexicographic order.

36 Creignou et al. [8] studied the enumeration of satisfying assignments for propositional formulas
37 under a different order, namely in non-decreasing weight, and it was shown that under this new
38 requirement, enumerating with polynomial delay is only possible for Horn formulas and width-2
39 affine formulas (i.e., affine formulas with at most 2 literals per clause). One of the main ingredients of
40 these algorithms is the use of a priority queue to ensure enumeration in order (as is observed already
41 by Johnson et al. [5]).

42 While parameterised enumeration has already been considered before (see, e.g., the works
43 of Fernau, Damaschke and Fomin et al. [9–11]), the notion of fixed-parameter tractable delay was
44 introduced only recently in this context, leading to the definition of the complexity class DelayFPT [12].
45 The “polynomial time” in the definition of DelayP here is simply replaced by a time-bound of the form
46 $p(n) \cdot f(k)$, where n denotes the input length, k is the input parameter, p is an arbitrary polynomial, and
47 f is a computable function. By this, the notion of efficiency in the context of the parameterised world,
48 i.e., fixed-parameter tractability (FPT), has been combined with the enumeration framework. A number
49 of problems from propositional logic were studied by Creignou et al. [12] and enumeration algorithms
50 based on self-reducibility and on the technique of kernelisation were developed. In particular, it was
51 shown that membership of an enumeration problem in DelayFPT can be characterised by a certain
52 tailored form of kernelisability, very much as in the context of usual decision problems.

53 As this area of parameterised enumeration is rather young and has received less attention, we
54 want to further push the topic with this paper. Here, we study *ordered enumeration* in the context of
55 *parameterised complexity*. First, we develop a novel formal framework for enumeration with *arbitrary*
56 orders. Then we consider the special context of graph modification problems where we are interested in
57 ordered enumeration for the two mostly studied orders, namely by *lexicographic* and by non-decreasing
58 *size* (where the size is the number of modifications that have to be made). We use two algorithmic
59 strategies, depending on the respective order as follows. Based on the principle of self-reducibility we
60 obtain DelayFPT (and polynomial-space) enumeration algorithms for lexicographic order, as soon as
61 the decision problem is efficiently solvable. Secondly, we present a DelayFPT enumeration algorithm
62 for order by size as soon as a certain FPT-computable neighbourhood function on the solutions
63 set exists (see Theorem 1). Notice that, the presented enumeration algorithms do not start from a
64 minimal solution but solutions of bounded size. Extending to such solutions from minimal ones in the
65 enumeration process is not generally trivial. To cope with the order, we use a priority queue that may
66 require exponential space in the input length (as there exist potentially that many solutions).

67 Eventually, we show that the observed principles and algorithmic strategies can be applied to
68 general modification problems as well. It is a rather rare situation that a general algorithmic scheme
69 is developed. Usually algorithms are devised on a very individual basis. We prove a wide scope
70 of applicability of our method by presenting new FPT-delay ordered enumeration algorithms for
71 a large variety of problems, such as cluster editing [13], triangulation [14], triangle deletion [15],
72 closest-string [16], and backdoor sets [17]. Furthermore, there already exists work which adopts the
73 introduced framework of Creignou et al. [12] in the area of conjunctive query enumeration [18], triangle
74 enumeration [19], combinatorial optimisation [20], abstract argumentation [21], and global constraints
75 [22].

76 2. Preliminaries

77 We start by defining parameterised enumeration problems with a specific ordering and their
78 corresponding enumeration algorithms. Most definitions in this section transfer those of Johnson et al.

79 and Schmidt [5,23] from the context of enumeration and those of Creignou et al. [12] from the context
80 of parameterised enumeration to the context of parameterised *ordered* enumeration.

81 The studied orderings of enumeration problems in this paper are quasi-orders which will be
82 defined in the following.

83 **Definition 1** (Quasi-order). *Let R be a set and \preceq a binary relation on R . Then \preceq is a preorder (or quasi-order)*
84 *if we have for all elements $a, b, c \in R$:*

- 85 • $a \preceq a$, and
- 86 • if $a \preceq b$ and $b \preceq c$ then $a \preceq c$.

87 We will write $z \not\preceq y$ whenever $z \preceq y$ is not true.

88 Now, we proceed by introducing parameterised enumeration problems with ordering. Intuitively,
89 the corresponding enumeration algorithm for such problems has to obey the given ordering, that is, it
90 has to produce solutions without violating that ordering.

91 **Definition 2.** *A parameterised enumeration problem with ordering is a quadruple $E = (I, \kappa, \text{Sol}, \preceq)$*
92 *such that the following holds:*

- 93 • I is the set of instances.
- 94 • $\kappa: I \rightarrow \mathbb{N}$ is the parameterisation function; κ is required to be polynomial-time computable.
- 95 • Sol is a function such that for all $x \in I$, $\text{Sol}(x)$ is a finite set, the set of solutions of x . Further we write
96 $\mathcal{S} = \bigcup_{x \in I} \text{Sol}(x)$.
- 97 • \preceq is a quasi-order on \mathcal{S} .

98 Notice that this order on all solutions is only a lazy way of simultaneously giving an order for
99 each instance. Furthermore, we will write an index E letter, e.g., I_E, κ_E , to denote that we talk about
100 instance set, parameterisation function, etc. of a given enumeration problem E . In the next step, we fix
101 the notion of enumeration algorithms in our setting.

102 **Definition 3** (Enumeration Algorithm). *Let $E = (I, \kappa, \text{Sol}, \preceq)$ be a parameterised enumeration problem*
103 *with ordering. Then an algorithm \mathcal{A} is an enumeration algorithm for E if the following holds:*

- 104 • For every $x \in I$, $\mathcal{A}(x)$ terminates after a finite number of steps.
- 105 • For every $x \in I$, $\mathcal{A}(x)$ outputs exactly the elements of $\text{Sol}(x)$ without duplicates.
- 106 • For every $x \in I$ and $y, z \in \text{Sol}(x)$, if $y \preceq z$ and $z \not\preceq y$ then $\mathcal{A}(x)$ outputs solution y before solution z .

107 Before we define complexity classes for parameterised enumeration, we need the notion of delay
108 for enumeration algorithms.

109 **Definition 4** (Delay). *Let $E = (I, \kappa, \text{Sol}, \preceq)$ be a parameterised enumeration problem with ordering and \mathcal{A} be*
110 *an enumeration algorithm for E . Let $x \in I$ be an instance. The i -th delay of \mathcal{A} is the elapsed runtime with respect*
111 *to $|x|$ of \mathcal{A} between outputting the i -th and $(i + 1)$ -st solution in $\text{Sol}(x)$. The 0-th delay is the precomputation*
112 *time which is the elapsed runtime with respect to $|x|$ of \mathcal{A} from the start of the computation to the first output*
113 *statement. Analogously, the n -th delay, for $n = |\text{Sol}(x)|$, is the postcomputation time which is the elapsed*
114 *runtime with respect to $|x|$ of \mathcal{A} after the last output statement until \mathcal{A} terminates. Then, the delay of \mathcal{A} is the*
115 *maximum over all $0 \leq i \leq n$ of the i -th delay of \mathcal{A} .*

116 Now we are able to define two different complexity classes for parameterised enumeration
117 following the notion of Creignou et al. [12].

118 **Definition 5.** *Let $E = (I, \kappa, \text{Sol}, \preceq)$ be a parameterised enumeration problem. We say E is FPT-enumerable if*
119 *there exists an enumeration algorithm \mathcal{A} , a computable function $f: \mathbb{N} \rightarrow \mathbb{N}$, and a polynomial p such that for*
120 *every $x \in I$, \mathcal{A} outputs all solutions of $\text{Sol}(x)$ in time $f(\kappa(x)) \cdot p(|x|)$.*

121 An enumeration algorithm \mathcal{A} is a DelayFPT-algorithm if there exists a computable function $f: \mathbb{N} \rightarrow \mathbb{N}$,
 122 and a polynomial p such that for every $x \in I$, \mathcal{A} outputs all solutions of $\text{Sol}(x)$ with delay of at most
 123 $f(\kappa(x)) \cdot p(|x|)$.

124 The class DelayFPT consists of all parameterised enumeration problems that admit a
 125 DelayFPT-enumeration algorithm.

126 Some of our enumeration algorithms will make use of the concept of *priority queues* to
 127 enumerate all solutions in the correct order and to avoid duplicates. We will follow the approach
 128 of Johnson et al. [5]. For an instance x of a parameterised enumeration problem whose sizes of
 129 solutions are polynomially bounded in $|x|$, we use a priority queue Q to store a subset of $\text{Sol}(x)$, of
 130 cardinality potentially exponential in $|x|$. The *insert operation* of Q requires $O(|x| \cdot \log |\text{Sol}(x)|)$ time.
 131 The *extract minimum operation* requires $O(|x| \cdot \log |\text{Sol}(x)|)$ time, too. It is important, however, that the
 132 computation of the order between two elements takes at most $O(|x|)$ time. As pointed out by Johnson
 133 et al. the required queue can be implemented with the help of standard balanced tree schemes [24].

134 2.1. Graph Modification Problems

135 Graph modifications problems have been studied for a long time in computational complexity
 136 theory [25]. Already in the monograph by Garey and Johnson [26], among the graph-theoretic problems
 137 considered, many fall into this problem class. To the best of our knowledge, graph modification
 138 problems were studied in the context of parameterised complexity for the first time in [27].

139 In this paper, we consider only undirected graphs. Let \mathcal{G} denote the set of all undirected graphs.
 140 A graph property $\mathcal{P} \subseteq \mathcal{G}$ is a set of graphs.

141 **Definition 6** (Graph Operations). Given a graph property \mathcal{P} and an undirected graph G , we write $G \models \mathcal{P}$ if
 142 the graph G obeys the property \mathcal{P} , that is, $G \in \mathcal{P}$. A (graph) operation for G is either of the following:

- 143 • removing a vertex: a function $\text{rem}_v: \mathcal{G} \rightarrow \mathcal{G}$ such that $\text{rem}_v(G)$ is the graph obtained by removing the
 144 vertex v from G (if v is present; otherwise rem_v is the identity) and deleting all incident edges to v ,
- 145 • adding/removing an edge: a function $\text{add}_{\{u,v\}}, \text{rem}_{\{u,v\}}: \mathcal{G} \rightarrow \mathcal{G}$ such that $\text{add}_{\{u,v\}}(G), \text{rem}_{\{u,v\}}(G)$
 146 is the graph obtained by adding/removing the edge $\{u,v\}$ to G if u and v are present in G ; otherwise both
 147 functions are the identity

148 Two operations o, o' are dependent if

- 149 • $o = \text{rem}_v$ and $o' = \text{rem}_{\{u,v\}}$ (one removes a vertex v and the other removes or adds an edge incident to
 150 v), or
- 151 • $o = \text{rem}_{\{u,v\}}$ and $o' = \text{add}_{\{u,v\}}$ (one removes an edge $\{u,v\}$ and the other adds the same edge $\{u,v\}$
 152 again).

153 A set of operations is consistent if it does not contain two dependent operations. Given such a consistent set of
 154 operations S , the graph obtained from G by applying the operations in S on G is denoted by $S(G)$.

155 Now, we turn towards the definition of solutions and will define minimality in terms of being
 156 inclusion-minimal.

157 **Definition 7** (Solutions). Given a graph property \mathcal{P} , a graph G , $k \in \mathbb{N}$, and a set of operations O , we say that
 158 S is a solution for (G, k, O) with respect to \mathcal{P} if the following three properties hold:

- 159 1. $S \subseteq O$ is a consistent set of operations,
- 160 2. $|S| \leq k$, and
- 161 3. $S(G) \models \mathcal{P}$.

162 A solution S is minimal if there is no solution S' such that $S' \subsetneq S$.

163 Cai [27] was interested in the following parameterised graph modification decision problem with
 164 respect to a given graph property \mathcal{P} :

Problem:	$\mathcal{M}_{\mathcal{P}}$
Input:	(G, k, O) , G undirected graph, $k \in \mathbb{N}$, O set of operations on G .
Parameter:	The integer k .
Question:	Does there exist a solution for (G, k, O) with respect to \mathcal{P} ?

166 Some of the most important examples of graph modification problems are presented now. A *chord*
 167 in a graph $G = (V, E)$ is an edge between two vertices of a cycle C in G which is not part of C . A given
 168 graph $G = (V, E)$ is *triangular* (or *chordal*) if each of its induced cycles of 4 or more nodes has a chord.
 169 The problem TRIANGULATION then asks, given an undirected graph G and $k \in \mathbb{N}$, whether there
 170 exists a set of at most k edges such that adding this set of edges to G makes it triangular. Yannakakis
 171 showed that this problem is NP-complete [14]. Kaplan et al. [28], and independently Cai [27] have
 172 shown that the parameterised problem is in FPT. For this problem, a solution is a set of edges which
 173 have to be added to the graph to make the graph triangular. Observe that, in this special case of the
 174 modification problem, the underlying property \mathcal{P} , “to be triangular”, does not have a finite forbidden
 175 set characterisation (since cycles of any length are problematic). Nevertheless, we will see later, that
 176 one can efficiently enumerate all minimal solutions as well.

177 A *cluster* is a graph such that all its connected components are cliques. In order to transform
 178 (or modify) a graph G we allow here only two kinds of operations: adding or removing an edge.
 179 CLUSTER-EDITING asks, given a graph G and a parameter k , whether there exists a consistent set of
 180 operations of cardinality at most k such that $S(G)$ is cluster. It was shown by Shamir et al. that the
 181 problem is NP-complete [13].

182 The problem TRIANGLE-DELETION asks whether a given graph can be transformed into a
 183 triangle-free graph by deletion of at most k vertices. Yannakakis has shown that the problem is
 184 NP-complete [15].

185 Analogous problems can be defined for many other classes of graphs, e.g., line graphs, claw-free
 186 graphs, Helly circular-arc graphs, etc., see [29].

187 Now, we turn towards the main focus of the paper. Here, we are interested in corresponding
 188 enumeration problems with ordering. In particular, we will focus on two well-known preorders,
 189 lexicographic ordering and ordering by size. Since our solutions are subsets of an ordered set of
 190 operations, they can be encoded as binary strings in which the i th bit from right indicates whether the
 191 i th operation is in the subset. We define the lexicographic ordering of solutions as the lexicographic
 192 ordering of these strings. Then, the size of a solution simply is its cardinality.

Problem:	ENUM- $\mathcal{M}_{\mathcal{P}}^{\text{LEX}}$
Input:	(G, k, O) , G undirected graph, $k \in \mathbb{N}$, O ordered set of operations on G .
Parameter:	The integer k .
Output:	All solutions of (G, k, O) with respect to \mathcal{P} in lexicographic order.
Problem:	ENUM- $\mathcal{M}_{\mathcal{P}}^{\text{SIZE}}$
Input:	(G, k, O) , G undirected graph, $k \in \mathbb{N}$, O set of operations on G .
Parameter:	The integer k .
Output:	All solutions of (G, k, O) with respect to \mathcal{P} in non-decreasing size.

195 If the context is clear, we omit the subscript \mathcal{P} for the graph modification problem and simply
 196 write \mathcal{M} . Furthermore, we write $\text{Sol}_{\mathcal{M}}(x)$ for the function associating solutions to a given instance,
 197 and also $\mathcal{S}_{\mathcal{M}}$ for the set of all solutions of \mathcal{M} .

198 3. Enumeration of Graph Modification Problems with Ordering

199 In this section, we study the two previously introduced parameterised enumeration problems
200 with ordering (lexicographic and size ordering).

201 3.1. Lexicographic Ordering

202 We first prove that, for any graph property \mathcal{P} , if the decision problem $\mathcal{M}_{\mathcal{P}}$ is in FPT then there is
203 an efficient enumeration algorithm for $\text{ENUM-}\mathcal{M}_{\mathcal{P}}^{\text{LEX}}$.

204 **Lemma 1.** *Let $\mathcal{M}_{\mathcal{P}}$ be a graph modification problem. If $\mathcal{M}_{\mathcal{P}}$ is in FPT then $\text{ENUM-}\mathcal{M}_{\mathcal{P}}^{\text{LEX}} \in \text{DelayFPT}$ with
205 polynomial space.*

206 **Proof.** Algorithm 1 enumerates all solutions of an instance of a given modification problem $\mathcal{M}_{\mathcal{P}}$ by
207 the method of self-reducibility (it is an extension of the flash light search of Creignou and Hébrard [7]).
208 The algorithm uses a function $\text{ExistsSol}(G, k, O)$ that tests if the instance (G, k, O) of the modification
209 problem $\mathcal{M}_{\mathcal{P}}$ has a solution. By assumption of the lemma, $\mathcal{M}_{\mathcal{P}} \in \text{FPT}$ so this function runs in fpt-time.
210 We use calls to this function to avoid exploration of branches of the recursion tree that do not lead to
211 any output. Also, we ensure that the solutions using o_p have to be consistent. This consistency check
212 runs in polynomial time for graph operations. The rest yields a search tree of depth at most k . From
213 this it follows that, for any instance of length n , the time between the output of any two solutions is
214 bounded by $f(k) \cdot p(n)$ for some polynomial p and a computable function f . \square

Algorithm 1: Enumerate all solutions of $\mathcal{M}_{\mathcal{P}}$ in lexicographic order

Input: (G, k, O) : a graph G , $k \in \mathbb{N}$, an ordered set of operations $O = \{o_1, \dots, o_n\}$
Output: all consistent sets $S \subseteq O$ s.t. $|S| \leq k$, $S(G) \models \mathcal{P}$ in lexicographic order
1 **if** $\text{ExistsSol}(G, k, O)$ **then** $\text{Generate}(G, k, O, \emptyset)$;

Procedure $\text{Generate}(G, k, O, S)$:
1 **if** $O = \emptyset$ or $k = 0$ **then return** S ;
2 **else**
3 let o_p be the lexicographically last operation in O , $O := O \setminus \{o_p\}$;
4 **if** $\text{ExistsSol}(S(G), k, O)$ **then** $\text{Generate}(S(G), k, O, S)$;
5 **if** $S \cup \{o_p\}$ is consistent and $\text{ExistsSol}((S \cup \{o_p\})(G), k - 1, O)$ **then**
6 $\text{Generate}((S \cup \{o_p\})(G), k - 1, O, S \cup \{o_p\})$.

215 **Corollary 1.** $\text{ENUM-TRIANGULATION}^{\text{LEX}} \in \text{DelayFPT}$ with polynomial space.

216 **Proof.** Kaplan et al. [28] and Cai [27] showed that $\text{TRIANGULATION} \in \text{FPT}$. Now, by applying
217 Lemma 1, we get the result. \square

218 Cai [27] identified a class of graph properties whose associated modification problems belong to
219 FPT. Let us introduce some terminology.

220 **Definition 8.** *Given two graphs $G = (V, E)$ and $H = (V', E')$, we write $H \trianglelefteq G$ if H is an induced subgraph
221 of G , i.e., $V' \subseteq V$ and $E' = E \cap (V' \times V')$. Let \mathcal{F} be a set of graphs and \mathcal{P} be a graph property. We say that \mathcal{F}
222 is a forbidden set characterisation of \mathcal{P} if for any graph G it holds that: $G \models \mathcal{P}$ iff for all $H \in \mathcal{F}$, $H \not\trianglelefteq G$.*

223 Among the problems presented in the previous section (see page 5), TRIANGLE-DELETION and
224 CLUSTER-EDITING have a finite forbidden set characterisation, namely by triangles and paths of
225 length two. In contrast to that, TRIANGULATION has a forbidden set characterisation which is infinite,

226 since cycles of arbitrary length are problematic. Actually, for properties having a finite forbidden set
 227 characterisation, the corresponding modification problem is fixed-parameter tractable. Together with
 228 Lemma 1, this provides a positive result in terms of enumeration.

229 **Proposition 1** ([27]). *If a property \mathcal{P} has a finite forbidden set characterisation then $\mathcal{M}_{\mathcal{P}}$ is in FPT.*

230 **Corollary 2.** *For any graph modification problem, if \mathcal{P} has a finite forbidden set characterisation then*
 231 $\text{ENUM-}\mathcal{M}_{\mathcal{P}}^{\text{LEX}} \in \text{DelayFPT}$ *with polynomial space.*

232 **Proof.** This result follows by combining Proposition 1 with Lemma 1. \square

233 3.2. Size Ordering

234 A common strategy in the enumeration context consists of defining a notion of a neighbourhood
 235 that allows to compute a new solution from a previous one with small amounts of computation
 236 time (see, e.g., the work of Avis and Fukuda [30]). We introduce the notion of a neighbourhood
 237 function, which, roughly speaking, generates some initial solutions from which all solutions can
 238 be produced. A priority queue then takes care of the ordering and avoids duplicates, which may
 239 require exponential space. For the graph modification problems of interest, we show that if the
 240 inclusion-minimal solutions can be generated in FPT, then such a neighbourhood function exists,
 241 accordingly providing a DelayFPT-enumeration algorithm. In the following, \circledast (the “seed”) is a
 242 technical symbol that will be used to generate the initial solutions.

243 **Definition 9.** *Let \mathcal{M} be a graph modification problem. A neighbourhood function for \mathcal{M} is a (partial)*
 244 *function $\mathcal{N}_{\mathcal{M}}: I_{\mathcal{M}} \times (\mathcal{S}_{\mathcal{M}} \cup \{\circledast\}) \rightarrow 2^{\mathcal{S}_{\mathcal{M}}}$ such that the following holds:*

- 245 1. *For all $x = (G, k, O) \in I_{\mathcal{M}}$ and $S \in \text{Sol}_{\mathcal{M}}(x) \cup \{\circledast\}$, $\mathcal{N}_{\mathcal{M}}(x, S)$ is defined.*
- 246 2. *For all $x \in I_{\mathcal{M}}$, $\mathcal{N}_{\mathcal{M}}(x, \circledast) = \emptyset$ if $\text{Sol}_{\mathcal{M}}(x) = \emptyset$, and $\mathcal{N}_{\mathcal{M}}(x, \circledast)$ is an arbitrary set of solutions*
 247 *otherwise.*
- 248 3. *For all $x \in I_{\mathcal{M}}$ and $S \in \text{Sol}_{\mathcal{M}}(x)$, if $S' \in \mathcal{N}_{\mathcal{M}}(x, S)$ then $|S| < |S'|$.*
- 249 4. *For all $x \in I_{\mathcal{M}}$ and all $S \in \text{Sol}_{\mathcal{M}}(x)$, there exists $p > 0$ and $S_1, \dots, S_p \in \text{Sol}_{\mathcal{M}}(x)$ such that (i)*
 250 *$S_1 \in \mathcal{N}_{\mathcal{M}}(x, \circledast)$, (ii) $S_{i+1} \in \mathcal{N}_{\mathcal{M}}(x, S_i)$ for $1 \leq i < p$, and (iii) $S_p = S$.*

251 *Furthermore, we say that $\mathcal{N}_{\mathcal{M}}$ is FPT-computable, when $\mathcal{N}_{\mathcal{M}}(x, S)$ is computable in time $f(\kappa(x)) \cdot \text{poly}(|x|)$*
 252 *for any $x \in I_{\mathcal{M}}$ and $S \in \text{Sol}_{\mathcal{M}}(x)$.*

253 As a result, a neighbourhood function for a problem \mathcal{M} is a function that in a first phase computes
 254 from scratch an initial set of solutions (see Definition 9(2)). In many of our applications below,
 255 $\mathcal{N}_{\mathcal{M}}(x, \circledast)$ will be the set of all minimal solutions for x . In a second phase these solutions are iteratively
 256 extended (see condition (3)), where condition (4) guarantees that we do not miss any solution, as we
 257 will see in the next theorem.

258 **Theorem 1.** *Let \mathcal{M} be a graph modification problem. If \mathcal{M} admits a neighbourhood function $\mathcal{N}_{\mathcal{M}}$ that is*
 259 *FPT-computable, then $\text{ENUM-}\mathcal{M}^{\text{SIZE}} \in \text{DelayFPT}$.*

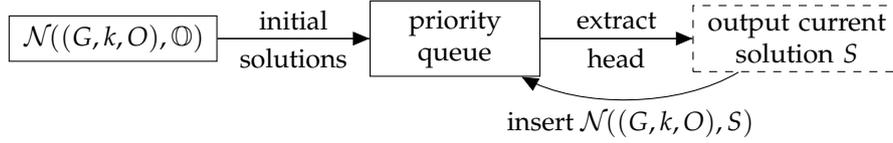
260 **Proof.** Algorithm 2 outputs all solutions in DelayFPT-time. By the definition of the priority queue
 261 (recall in particular that insertion of an element is done only if the element is not yet present in the
 262 queue) and by the fact that all elements of $\mathcal{N}_{\mathcal{M}}((G, k, O), S)$ are of bigger size than S by Definition 9(3),
 263 it is easily seen that *the solutions are output in the right order and that no solution is output twice.*

264 Besides, *no solution is omitted.* Indeed, given $S \in \text{Sol}_{\mathcal{M}}(G, k, O)$ and S_1, \dots, S_p associated with S
 265 by Definition 9(4), we prove by induction that each S_i is inserted in Q during the run of the algorithm:

266 $i = 1$: This proceeds from line 2 of the algorithm.

Algorithm 2: DelayFPT algorithm for ENUM- \mathcal{M}

Input : (G, k, O) : G is an undirected graph, $k \in \mathbb{N}$, and O is a set of operations.
1 compute $\mathcal{N}_{\mathcal{M}}((G, k, O), \emptyset)$;
2 insert all elements of $\mathcal{N}_{\mathcal{M}}((G, k, O), \emptyset)$ into priority queue Q (ordered by size);
3 **while** Q is not empty **do**
4 **extract** the minimum solution S of Q and output it;
5 **insert** all elements of $\mathcal{N}_{\mathcal{M}}((G, k, O), S)$ into Q ;

**Figure 1.** Structure of Algorithm 2.

267 $i > 1$: The solution S_{i-1} is inserted in Q by induction hypothesis and hence all elements of
268 $\mathcal{N}_{\mathcal{M}}((G, k, O), S_{i-1})$, including S_i , are inserted in Q (line 5 of Algorithm 2). Consequently,
269 each S_i is inserted in Q and then output during the run. In particular, this is true for $S = S_p$.

270 Finally, we claim that Algorithm 2 runs in DelayFPT-time. Indeed, the delay between the output
271 of two consecutive solutions is bounded by the time required to compute a neighbourhood of the form
272 $\mathcal{N}_{\mathcal{M}}((G, k, O), \emptyset)$ or $\mathcal{N}_{\mathcal{M}}((G, k, O), S)$ and to insert all its elements in the priority queue. This is in
273 FPT due to the assumption on $\mathcal{N}_{\mathcal{M}}$ being FPT-computable and as there is only a single extraction and
274 FPT-many insertion operations on the queue. \square

275 A natural way to provide a neighbourhood function for a graph modification problem \mathcal{M} is to
276 consider the *inclusion minimal* solutions of \mathcal{M} . Let us denote by MIN- \mathcal{M} the problem of enumerating
277 all inclusion minimal solutions of \mathcal{M} .

278 **Theorem 2.** Let \mathcal{M} be a graph modification problem. If MIN- \mathcal{M} is FPT-enumerable then ENUM- $\mathcal{M}^{\text{SIZE}} \in$
279 DelayFPT.

280 **Proof.** Let \mathcal{A} be an FPT-algorithm for MIN- \mathcal{M} . Because of Theorem 1, it is sufficient to build an
281 FPT-neighbourhood function for \mathcal{M} . For an instance (G, k, O) of \mathcal{M} and for $S \in \text{Sol}_{\mathcal{M}}(G, k, O) \cup \{\emptyset\}$,
282 we define $\mathcal{N}_{\mathcal{M}}((G, k, O), S)$ as the result of Algorithm 3.

283 Accordingly, the function $\mathcal{N}_{\mathcal{M}}$ clearly fulfils Conditions 2 and 3 of Definition 9. We prove by
284 induction that it also satisfies Condition 4 (that is, each solution T of size k comes with a sequence
285 $T_1, \dots, T_p = T$ such that $T_1 \in \mathcal{N}_{\mathcal{M}}((G, k, O), \emptyset)$ and $T_{i+1} \in \mathcal{N}_{\mathcal{M}}((G, k, O), T_i)$ for each i). If T is a
286 minimal solution for (G, k, O) , then $T \in \mathcal{N}_{\mathcal{M}}((G, k, O), \emptyset)$ and the expected sequence $(T_i)_{1 \leq i \leq p}$
287 reduces to $T_1 = T$. Otherwise, there exists an $S \in \text{Sol}_{\mathcal{M}}(G, k, O)$ and a non-empty set of transformations, say
288 $S' \cup \{t\}$, such that $T = S \cup S' \cup \{t\}$ and there is no solution for G between S and $S \cup S' \cup \{t\}$. This entails
289 that S' is a minimal solution for $((S \cup \{t\})(G), k - |S| - 1)$ and, as a consequence, $T \in \mathcal{N}_{\mathcal{M}}((G, k, O), S)$
290 (see lines 4–5 of Algorithm 3). The conclusion follows from the induction hypothesis that guarantees
291 the existence of solutions S_1, \dots, S_q such that $S_1 \in \mathcal{N}_{\mathcal{M}}((G, k, O), \emptyset)$, $S_{i+1} \in \mathcal{N}_{\mathcal{M}}((G, k, O), S_i)$ and
292 $S_q = S$. The expected sequence T_1, \dots, T_p for T is nothing but S_1, \dots, S_q, T . To conclude, it remains to
293 show that Algorithm 3 is FPT. This follows from the fact that \mathcal{A} is an FPT-algorithm (Lines 1 and 4 of
294 Algorithm 3). \square

295 **Corollary 3.** ENUM-TRIANGULATION^{SIZE} \in DelayFPT.

Algorithm 3: Procedure for computing $\mathcal{N}_{\mathcal{M}}((G, k, O), S)$

Input : $(G, k, O), S$: G is an undirected graph, $k \in \mathbb{N}$, O and S are sets of operations.

- 1 **if** $S = \emptyset$ **then return** $\mathcal{A}(G, k, O)$;
- 2 $\text{res} := \emptyset$;
- 3 **forall the** $t \in O$ **do**
- 4 **forall the** $S' \in \mathcal{A}((S \cup \{t\})(G), k - |S| - 1, O \setminus \{t\})$ **do**
- 5 **if** $S \cup S' \cup \{t\}$ **is consistent** **then** $\text{res} := \text{res} \cup \{S \cup S' \cup \{t\}\}$;
- 6 **return** res ;

296 **Proof.** All size minimal k -triangulations can be output in time $O(2^{4k} \cdot |E|)$ for a given graph G and
 297 $k \in \mathbb{N}$ as shown by Kaplan et al. [28, Thm. 2.4]. This immediately yields the expected result via
 298 Theorem 2. \square

299 **Corollary 4.** For any property \mathcal{P} that has a finite forbidden set characterisation, the problem $\text{ENUM-}\mathcal{M}_{\mathcal{P}}^{\text{SIZE}}$ is
 300 in DelayFPT.

301 **Proof.** The algorithm developed by Cai [27] for the decision problem is based on a bounded search
 302 tree, whose exhaustive examination provides all size minimal solutions in FPT. Theorem 2 yields the
 303 conclusion. \square

304 **Corollary 5.** $\text{ENUM-CLUSTER-EDITING}^{\text{SIZE}}$ and $\text{ENUM-TRIANGLE-DELETION}^{\text{SIZE}}$ are in DelayFPT.

305 **Proof.** Both problems have a finite forbidden set characterisation. For the cluster editing problem,
 306 paths of length two are the forbidden pattern, and, Regarding $\text{ENUM-TRIANGLE-DELETION}^{\text{SIZE}}$, the
 307 forbidden patterns are obviously just triangles. Finally, just apply Corollary 4. \square

308 4. Generalisation to Modification Problems

309 In this section, we will show how the algorithmic strategy that has been defined and formalised
 310 in the context of graph modification can be of use for many other problems, coming from various
 311 combinatorial frameworks.

312 **Definition 10** (General Operations). Let $Q \subseteq \Sigma^*$ be a language defined over an alphabet, and $x \in \Sigma^*$ be an
 313 input. A set of operations $\Omega(Q) = \{\omega_n: \Sigma^* \rightarrow \Sigma^* \mid n \in \mathbb{N}\}$ is an infinite set of operations on instances of
 314 Q . We say an operation ω is valid with respect to an instance $x \in Q$, if $\omega(x) \in Q$. We write Ω/x as the set of
 315 possible (valid) operations on an instance x .

316 Two operations ω, ω' are dependent with respect to an instance $x \in Q$ if

- 317 • $\omega(\omega'(x)) = x$, or
 318 • $\omega(\omega'(x)) = \omega'(x)$ or $\omega(\omega'(x)) = \omega(x)$

319 A set of operations $O \subseteq \Omega/x$ is consistent with respect to x if it does not contain two dependent operations.

320 For instance, the set Ω could contain operations that add edges or, in another case, flip bits. It
 321 highly is the subject to the repective language Q .

Example 1. Let $\mathcal{G} \subseteq \{0, 1\}^*$ be the language of all undirected graphs encoded by adjacency matrices. Then $\Omega(\mathcal{G})$ is the set of all graph operations in the sense of Definition 6: removing vertices or edges, adding edges. Note that $\Omega(\mathcal{G})$ contains all operations of the kind

$$\text{rem}_j: \mathcal{G} \rightarrow \mathcal{G}, \quad \text{rem}_{\{i,j\}}: \mathcal{G} \rightarrow \mathcal{G}, \quad \text{add}_{\{i,j\}}: \mathcal{G} \rightarrow \mathcal{G}$$

322 for all $i, j \in \mathbb{N}$. Furthermore, let $G = (V, E) \in \{0, 1\}^*$ be a concrete input graph. As a result, Ω/G then is the
 323 restriction of Ω to those $i, j \in \mathbb{N}$ such that $v_i, v_j \in V$ encode vertices in G .

324 Similarly as defined in Subsection 2.1, a property is just a set. In the following context, it is a
 325 subset of a considered language Q . Intuitively, you may think, in the view of graph modification
 326 problems, of Q as \mathcal{G} . Then a graph property \mathcal{P} was just a subset of \mathcal{G} .

Definition 11 (General Solutions). Let $Q \subseteq \Sigma^*$ be a language defined over an alphabet, $S \subseteq \Omega/x$ be a finite set of operations on $x \in Q$ and $\mathcal{P} \subseteq Q$ be a property. We say S is a solution (of x) if there exists an ordering of $S = \{\omega_{i_1}, \dots, \omega_{i_k}\}$ such that $\omega_{i_1}(\omega_{i_2}(\dots \omega_{i_k}(x) \dots)) \in \mathcal{P}$ for $i_\mu \in \mathbb{N}$ and $1 \leq \mu \leq k$. In such a case, we also just write $\omega_{i_1}(\omega_{i_2}(\dots \omega_{i_k}(x) \dots)) \models \mathcal{P}$. If for every pair of permutations on k elements α, β we have that

$$\omega_{\alpha(1)}(\omega_{\alpha(2)}(\dots \omega_{\alpha(k)}(x) \dots)) = \omega_{\beta(1)}(\omega_{\beta(2)}(\dots \omega_{\beta(k)}(x) \dots)),$$

327 then we say S is consistent.

328 If S is a consistent set of operations then we write $S(x)$ for the application of the operations in S to x . In short,
 329 whenever S is a consistent solution we just write $S(x) \models \mathcal{P}$. Similarly, we say an operation ω is consistent with
 330 a set S if and only if $S \cup \{\omega\}$ is consistent. Furthermore, we denote by $\mathcal{S}_Q := \bigcup_{x \in Q} \{S \mid S \text{ is a solution of } x\}$
 331 the set of all solutions for every instance $x \in Q$. Also $\text{Sol}(x)$ is the set of solutions for every instance $x \in Q$.

332 **Example 2.** Continuing the previous example, if the property \mathcal{P} is “to be a cluster” then a consistent solution
 333 S to a given graph just then is a sequence of removing vertices, adding and deleting of edges where

- 334 • there is no edge (i, j) added or deleted such that vertex i or j is removed,
- 335 • there is no edge (i, j) added and removed, and
- 336 • $S(G) \models \mathcal{P}$.

337 Similarly, adding edge (i, j) together with removing vertex i or j or removing edge (i, j) is an inconsistent set of
 338 operations.

339 Now we want to define the corresponding decision and enumeration tasks. On that account, let
 340 \mathcal{P} be a property, $\Pi = (Q, \kappa)$ be a parametrised problem with $Q \subseteq \Sigma^*$, and Ω be a set of operations.

341	Problem: $\Pi_{\mathcal{P}}$ — parameterised modification problem Π regarding property \mathcal{P} over Σ
	Input: $x \in \Sigma^*, k \in \mathbb{N}, \Omega/x$ set of operations.
	Parameter: The integer k .
	Question: Is there a consistent solution $S \subseteq \Omega/x$ such that $S(x) \models \mathcal{P}$ and $ S \leq k$?
342	Problem: ENUM-MIN- $\Pi_{\mathcal{P}}$ — parameterised minimum enumeration modification problem regarding property \mathcal{P} over Σ
	Input: $x \in \Sigma^*, k \in \mathbb{N}, \Omega/x$ set of operations.
	Parameter: The integer k .
	Output: All minimal (w.r.t. some order) consistent solutions $S \subseteq \Omega/x$ with $ S \leq k$ such that $S(x) \models \mathcal{P}$.

343 The enumeration modification problem where we want to output all possible sets of
 344 transformations on a given instance x (and not only the minimum ones) then is ENUM- $\Pi_{\mathcal{P}}$.

345 In the following, we show how the notion of neighbourhood functions can be generalised as well.
 346 This will in turn yield generalisations of the results for graph modification problems afterwards.

347 **Definition 12.** Let Σ be an alphabet, $\mathcal{P} \subseteq \Sigma^*$ be a property and $\Pi_{\mathcal{P}}$ be a parameterised modification problem
 348 over Σ . A neighbourhood function for $\Pi_{\mathcal{P}}$ is a (partial) function $\mathcal{N}_{\Pi_{\mathcal{P}}}: \Sigma^* \times (\mathcal{S}_{\Pi_{\mathcal{P}}} \cup \{\emptyset\}) \rightarrow 2^{\mathcal{S}_{\Pi_{\mathcal{P}}}}$ such
 349 that the following holds:

- 350 1. For all $x \in \Sigma^*$ and $S \in \text{Sol}_{\Pi_p}(x) \cup \{\emptyset\}$, $\mathcal{N}_{\Pi_p}(x, S)$ is defined.
 351 2. For all $x \in \Sigma^*$, $\mathcal{N}_{\Pi_p}(x, \emptyset) = \emptyset$ if $\text{Sol}_{\Pi_p}(x) = \emptyset$, and $\mathcal{N}_{\Pi_p}(x, \emptyset)$ is an arbitrary set of solutions
 352 otherwise.
 353 3. For all $x \in \Sigma^*$ and $S \in \text{Sol}_{\Pi_p}(x)$, if $S' \in \mathcal{N}_{\Pi_p}(x, S)$ then $|S| < |S'|$.
 354 4. For all $x \in \Sigma^*$ and all $S \in \text{Sol}_{\Pi_p}(x)$, there exists $p > 0$ and $S_1, \dots, S_p \in \text{Sol}_{\Pi_p}(x)$ such that (i)
 355 $S_1 \in \mathcal{N}_{\Pi_p}(x, \emptyset)$, (ii) $S_{i+1} \in \mathcal{N}_{\Pi_p}(x, S_i)$ for $1 \leq i < p$, and (iii) $S_p = S$.

356 Furthermore, we say that \mathcal{N}_{Π_p} is FPT-computable, when $\mathcal{N}_{\Pi_p}(x, S)$ is computable in time $f(k) \cdot \text{poly}(|x|)$ for
 357 any $x \in \Sigma^*$ and $S \in \text{Sol}_{\Pi_p}(x)$.

358 As already announced before, we are able to state generalised versions of Theorems 1 and 2 which
 359 can be proven in a similar way. However, one has to replace the graph modification problems by
 360 general modification problems.

361 **Corollary 6.** Let \mathcal{P} be a property, $\Pi \subseteq \Sigma^* \times \mathbb{N}$ be a parameterised modification problem, and Ω be a set of
 362 operations such that Ω/x is finite for all $x \in \Sigma^*$. If $\Pi_{\mathcal{P}}$ admits a neighbourhood function that is FPT-computable
 363 then $\text{ENUM-}\Pi_{\mathcal{P}} \in \text{DelayFPT}$ and

- 364 • polynomial space for lexicographic order, and
- 365 • exponential space for size order.

366 **Corollary 7.** Let \mathcal{P} be a property, $\Pi \subseteq \Sigma^* \times \mathbb{N}$ be a parameterised modification problem, and Ω be a set of
 367 operations such that Ω/x is finite for all $x \in \Sigma^*$. If $\text{ENUM-MIN-}\Pi_{\mathcal{P}}$ is FPT-enumerable and consistency of
 368 solutions can be checked in FPT then $\text{ENUM-}\Pi_{\mathcal{P}} \in \text{DelayFPT}$ and

- 369 • polynomial space for lexicographic order, and
- 370 • exponential space for size order.

371 4.1. Closest String

372 In the following, we consider a central NP-complete problem in coding theory [31]. Given a set of
 373 binary strings I , we want to find a string s whose maximum Hamming distance $\max\{d_H(s, s') \mid s' \in$
 374 $I\} \leq d$ for a $d \in \mathbb{N}$, where $d_H(s, s')$ is the Hamming distance between two strings.

375 **Definition 13** (Bit-flip operation). Given a string $w = w_1 \cdots w_n$ with $w_i \in \{0, 1\}$, $n \in \mathbb{N}$, and a set
 376 $S \subseteq \{1, \dots, n\}$, $S(w)$ denotes the string obtained from w in flipping the bits indicated by S , more formally
 377 $S(w) := S(w_1) \cdots S(w_n)$, where $S(w_i) = 1 - w_i$ if $i \in S$ and $S(w_i) = w_i$ otherwise.

378 The corresponding parametrised version is the following.

Problem:	CLOSEST-STRING
Input:	A sequence (s_1, s_2, \dots, s_k) of k strings over $\{0, 1\}$ each of given length $n \in \mathbb{N}$ and 379 an integer $d \in \mathbb{N}$.
Parameter:	The integer d .
Question:	Does there exist $S \subseteq \{1, \dots, n\}$ such that $d_H(S(s_1), s_i) \leq d$ for all $1 \leq i \leq k$?

380 **Proposition 2** ([16]). CLOSEST-STRING is in FPT.

381 Moreover, an exhaustive examination of a bounded search tree constructed from the idea of
 382 Gramm et al. [16, Fig. 1] allows to produce all minimal solutions of this problem in FPT. Accordingly,
 383 we get the following result for the corresponding enumeration problems.

384 **Theorem 3.**

- 385 • $\text{ENUM-CLOSEST-STRING}^{\text{LEX}} \in \text{DelayFPT}$ with polynomial space.

386 • ENUM-CLOSEST-STRING^{SIZE} \in DelayFPT with exponential space.

387 **Proof.** Ω is just the set of operations which flip the i -th bit of a string for every $i \in \mathbb{N}$. Then use
388 Proposition 2 and Corollary 7. \square

389 4.2. Backdoors

390 In this section, we will consider the concept of backdoors. Let \mathcal{C} be a class of propositional
391 formulas. Intuitively, a \mathcal{C} -backdoor is a set of variables of a given propositional formula with the
392 following property. Applying assignments over these variables to the formula always yields a formula
393 in the class \mathcal{C} . Of course, one aims for formula classes for which satisfiability can be decided efficiently.
394 Informally speaking, with the parameter backdoor size of a formula one tries to describe a distance to
395 tractability. This definition was first introduced by Golmes, Williams and Selman [17] to model short
396 distances to efficient subclasses. Until today, backdoors gained copious attention in many different
397 areas: abduction [32], answer set programming [33,34], argumentation [35], default logic [36], temporal
398 logic [37], planning [38], and constraint satisfaction [39,40].

399 Consider a formula ϕ in conjunctive normal form. Denote by $\phi[\tau]$ for a partial truth assignment τ
400 the result of removing all clauses from ϕ which contain a literal ℓ with $\tau(\ell) = 1$ and removing literals
401 ℓ with $\tau(\ell) = 0$ from the remaining clauses.

402 **Definition 14.** Let \mathcal{C} be a class of CNF-formulas and ϕ be a CNF-formula. A set $V \subseteq \text{Vars}(\phi)$ of variables of
403 ϕ is a strong \mathcal{C} -backdoor set of ϕ if for all truth assignments $\tau: V \rightarrow \{0, 1\}$ we have that $\phi[\tau] \in \mathcal{C}$.

404 **Definition 15** ([41,42]). Let \mathcal{C} be a class of CNF-formulas and ϕ be a CNF-formula. A set $V \subseteq \text{Vars}(\phi)$ of
405 variables of ϕ is a \mathcal{C} -deletion backdoor set of ϕ if $\phi[V]$ is in \mathcal{C} , where $\phi[V]$ denotes the formula obtained from
406 ϕ in deleting in ϕ all occurrences of variables from V .

407 **Definition 16** (Weak Backdoor Sets). Let \mathcal{C} be a class of CNF-formulas, and ϕ be a propositional CNF
408 formula. A set $V \subseteq \text{Vars}(\phi)$ of variables from ϕ is a weak \mathcal{C} -backdoor set of ϕ if there exists an assignment
409 $\theta \in \Theta(V)$ such that $\phi[\theta] \in \mathcal{C}$ and $\phi[\theta]$ is satisfiable.

410 Now let us consider the following enumeration problem.

Problem:	ENUM-WEAK-BACKDOORSET(\mathcal{C})
Input:	A formula ϕ in CNF, $k \in \mathbb{N}$.
Parameter:	The integer k .
Output:	The set of all weak \mathcal{C} -backdoor sets of ϕ of size at most k .

412 Similarly, define ENUM-STRONG-BACKDOORSET(\mathcal{C}) as the set of all strong \mathcal{C} -backdoor sets of ϕ
413 of size at most k . Observe that the backdoor set problems can be seen as modification problems where
414 solutions are sequences of variable assignments. The target property then simply is the class of CNF
415 formulas \mathcal{C} .

416 Notice that Creignou et al. [12, Thm. 4] have studied enumeration for exact strong
417 HORN-backdoor sets and provided an algorithm running in DelayFPT, where HORN denotes the
418 set of all Horn-formulas, that is, CNF-formulas whose clauses contain at most one positive literal.

419 **Definition 17** (Base Class, [43]). The class \mathcal{C} is a base class if it can be recognised in P (that is, $\mathcal{C} \in \text{P}$),
420 satisfiability of its formulas is in P, and the class is closed under isomorphisms w.r.t. variable names. We say \mathcal{C} is
421 clause-defined if for every CNF-formula ϕ we have: $\phi \in \mathcal{C}$ if and only if $\{C\} \in \mathcal{C}$ for all clauses C from ϕ .

422 **Proposition 3** ([43, Prop. 2]). For every clause-defined base class \mathcal{C} , detection of weak \mathcal{C} -backdoor sets is in
423 FPT for input formulas in 3CNF.

424 In their proof, Gaspers and Szeider [43] describe how utilising a bounded search tree allows one
 425 to solve the detection of weak \mathcal{C} -backdoors in FPT time. Interestingly to note, this technique results in
 426 obtaining *all minimal solutions* in FPT-time. This observation results in the following theorem.

427 **Theorem 4.** *For every clause-defined base class \mathcal{C} and input formulas in 3-CNF*

- 428 • ENUM-WEAK- \mathcal{C} -BACKDOORS^{LEX} \in DelayFPT *with polynomial space, and*
- 429 • ENUM-WEAK- \mathcal{C} -BACKDOORS^{SIZE} \in DelayFPT *with exponential space.*

430 **Proof.** The set of operations Ω then contains functions that replace a specific variable $i \in \mathbb{N}$
 431 by a truth value $t \in \{0,1\}$. A solution then encodes the chosen backdoor sets together with
 432 the required assignment. Then, Proposition 3 yields ENUM-MIN-WEAK- \mathcal{C} -BACKDOORS^{LEX}, resp.,
 433 ENUM-MIN-WEAK- \mathcal{C} -BACKDOORS^{SIZE} being FPT-enumerable. As the consistency check for solutions
 434 is in polynomial time, applying Corollary 7 completes the proof. \square

435 In the following result, we will examine the parametrised enumeration complexity of the task to
 436 enumerate all strong \mathcal{C} -backdoor sets of a given 3CNF formula for some clause-defined base class \mathcal{C} .
 437 Crucially, every strong backdoor set has to contain at least one variable from a clause that is not in \mathcal{C}
 438 which relates to ‘hitting all bad clauses’ like in the definition of deletion backdoors (see Def. 15).

439 **Theorem 5.** *For every clause-defined base class \mathcal{C} and input formulas in 3-CNF:*

- 440 • ENUM-STRONG- \mathcal{C} -BACKDOORS^{LEX} \in DelayFPT *with polynomial space, and*
- 441 • ENUM-STRONG- \mathcal{C} -BACKDOORS^{SIZE} \in DelayFPT *with exponential space.*

442 **Proof.** We show that for every clause-defined base class \mathcal{C} and input formulas in 3-CNF, the problem
 443 MIN-STRONG- \mathcal{C} -BACKDOORS is FPT-enumerable. Indeed, we only need to branch on the variables
 444 from a clause $C \notin \mathcal{C}$ and remove the corresponding literals over the considered variable from ϕ . The
 445 size of the branching tree is at most 3^k . As for base classes the satisfiability test is in P, this yields
 446 an FPT-algorithm. The neighbourhood function $\mathcal{N}(x, S)$ for $x = (\phi, k)$ is defined to be the set of the
 447 pairwise unions of all minimal strong \mathcal{C} -backdoors of $(\phi[(S \cup \{x_i\})], k - |S| - 1)$ together with $S \cup \{x_i\}$
 448 for all variables $x_i \notin S$. If $\text{Vars}(\phi) = \{x_1, \dots, x_n\}$, then the operations are $\omega_i: \phi \mapsto \phi(0/x_i) \wedge \phi(1/x_i)$.
 449 As applying the functions ω_i happens only with respect to the backdoor size k , which is the parameter,
 450 the formula size increases by an exponential factor in the parameter only. This yields the preconditions
 451 for Corollary 7 constituting the proof. \square

452 4.3. Weighted Satisfiability Problems

453 Finally, we consider satisfiability questions for formulas in the Schaefer framework [44]. A
 454 *constraint language* Γ is a finite set of relations. A Γ -*formula* ϕ , is a conjunction of constraints using only
 455 relations from Γ and, consequently, is a quantifier-free first order formula.

456 As opposed to the approach of Creignou et al. [12], who examined maximum satisfiability, we
 457 now focus on the problem MINONES-SAT(Γ) defined below.

458 **Definition 18 (Minimality).** *Given a propositional formula ϕ and an assignment θ over the variables in ϕ*
 459 *with $\theta \models \phi$, we say θ is minimal if there does not exist an assignment $\theta' \subset \theta$ and $\theta' \models \phi$. The size $|\theta|$ of θ is*
 460 *the number of variables it sets to true.*

461 Formally, the problem of interest is defined with respect to any fixed constraint language Γ :

Problem:	MIN-MINONES-SAT ^{SIZE} (Γ)
Input:	(ϕ, k) , a Γ -formula ϕ , $k \in \mathbb{N}$.
Parameter:	The integer k .
Output:	Generate all inclusion-minimal satisfying assignments θ of ϕ with $ \theta \leq k$ by non-decreasing size.

Similarly, the problem ENUM-MINONES-SAT(Γ) asks for all satisfying assignments θ of ϕ with $|\theta| \leq k$. In this context, the operations in Ω are functions that replace the variable with index $i \in \mathbb{N}$ by true.

Theorem 6. *For all constraint languages Γ , we have: MIN-MINONES-SAT^{SIZE}(Γ) is FPT-enumerable and ENUM-MINONES-SAT^{SIZE}(Γ) \in DelayFPT with exponential space.*

Proof. For the first claim we can simply compute the minimal assignments by a straight forward branching algorithm: initially, begin with the all 0-assignment, then consider all unsatisfied clauses in turn and flip one of the occurring variables to true. The second claim follows by a direct application of Corollary 7. \square

5. Conclusion

We presented FPT-delay ordered enumeration algorithms for a large variety of problems, such as cluster editing, chordal completion, closest-string, and weak and strong backdoors. An important point of our paper is that we propose a general strategy for efficient enumeration. This is rather rare in the literature, where algorithms are usually devised individually for specific problems. In particular, our scheme yields DelayFPT algorithms for all graph modification problems that are characterised by a finite set of forbidden patterns.

Initially, we focussed on graph-theoretic problems. Afterwards, the generic approach we presented, covers problems which are not only of a graph-theoretic nature. Here, we defined general modification problems, detached from graphs and constructed generic enumeration algorithms for arising problems in the world of strings, numbers, formulas, constraints, etc.

As an observation we would like to mention that the DelayFPT algorithms presented in this paper require exponential space due to the inherent use of the priority queues to achieve *ordered* enumeration. An interesting question, continuing research of Meier [45], is whether there is a method which requires less space but uses a comparable delay between the output of solutions and still obeys the underlying order on solutions.

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