

The Good, the Bad, and the Odd: Cycles in Answer-Set Programs

Johannes Klaus Fichte
Institute of Information Systems
Vienna University of Technology, Vienna, Austria
fichte@kr.tuwien.ac.at

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Abstract

Backdoors of answer-set programs are sets of atoms that represent “clever reasoning shortcuts” through the search space. Assignments to backdoor atoms reduce the given program to several programs that belong to a tractable target class. Previous research has considered target classes based on notions of acyclicity where various types of cycles (good and bad cycles) are excluded from graph representations of programs. We generalize the target classes by taking the parity of the number of negative edges on bad cycles into account and consider backdoors for such classes. We establish new hardness results and non-uniform polynomial-time tractability relative to directed or undirected cycles.

1 Introduction

Answer-set programming (ASP) is a popular framework to describe concisely search and combinatorial problems [14, 16]. It has been successfully applied in crypto-analysis, code optimization, the semantic web, and several other fields [18]. Problems are encoded by rules and constraints into disjunctive logic programs whose solutions are answer-sets (stable models). The complexity of finding an answer-set for a disjunctive logic program is Σ_2^P -complete [4]. However this hardness result does not exclude quick solutions for large instances if we can exploit structural properties that might be present in real-world instances.

Recently, Fichte and Szeider [5] have established a new approach to ASP based on the idea of backdoors, a concept that originates from the area of satisfiability [20]. Backdoors exploit the structure of instances by identifying sets of atoms that are important for reasoning. A *backdoor* of a disjunctive logic program is a set of variables such that any instantiation of the variables yields a simplified logic program that lies in a class of programs where the decision problem we are interested in is tractable. By means of a backdoor of size k for a disjunctive logic program we can solve the program by solving all the

2^k tractable programs that correspond to the truth assignments of the atoms in the backdoor. For each answer set of each of the 2^k tractable programs we need to check whether it gives rise to an answer set of the given program. In order to do this efficiently we consider tractable programs that have a small number of answer sets (e.g., stratified programs [9]).

We consider target classes based on various notions of acyclicity on the *directed/undirected dependency graph* of the disjunctive logic program. A cycle is *bad* if it contains an edge that represents an atom from a negative body of a rule. Since larger target classes facilitate smaller backdoors, we are interested in large target classes that allow small backdoors and efficient algorithms for finding the backdoors.

Contribution

In this paper, we extend the backdoor approach of [5] using ideas from Zhao [23]. We enlarge the target classes by taking the parity of the number of negative edges or vertices on bad cycles into account and consider backdoors with respect to such classes. This allows us to consider larger classes that also contain non-stratified programs. Our main results are as follows:

1. For target classes based on directed bad even cycles, the detection of backdoors of bounded size is intractable (Theorem 1).
2. For target classes based on undirected bad even cycles, the detection of backdoors is polynomial-time tractable (Theorem 3).

The result (2) is a *non-uniform* polynomial-time result since the order of the polynomial depends on the backdoor size. An algorithm is *uniform* polynomial-time tractable if it runs in time $\mathcal{O}(f(k) \cdot n^c)$ where f is an arbitrary function and c is a constant independent from k . Uniform polynomial-time tractable problems are also known as fixed-parameter tractable problems [3]. We provide strong theoretical evidence that result (2) cannot be extended to uniform polynomial-time tractability. Further, we establish that result (2) generalizes a result of Lin and Zhao [13].

2 Formal Background

We consider a universe U of propositional *atoms*. A *literal* is an atom $a \in U$ or its negation $\neg a$. A *disjunctive logic program* (or simply a *program*) P is a set of *rules* of the following form

$$x_1 \vee \dots \vee x_l \quad \leftarrow \quad y_1, \dots, y_n, \neg z_1, \dots, \neg z_m. \quad (1)$$

where $x_1, \dots, x_l, y_1, \dots, y_n, z_1, \dots, z_m$ are atoms and l, n, m are non-negative integers. Let r be a rule. We write $\{x_1, \dots, x_l\} = H(r)$ (the *head* of r) and $\{y_1, \dots, y_n, z_1, \dots, z_m\} = B(r)$ (the *body* of r). We abbreviate the positive literals of the body by $B^+(r) = \{y_1, \dots, y_n\}$ and the negative literals by

$B^-(r) = \{z_1, \dots, z_m\}$. We denote the sets of atoms occurring in a rule r or in a program P by $\text{at}(r) = H(r) \cup B(r)$ and $\text{at}(P) = \bigcup_{r \in P} \text{at}(r)$, respectively. A rule r is *normal* if $|H(r)| = 1$. A rule is *Horn* if normal and $|B^-(r)| = 0$. We say that a program has a certain property if all its rules have the property. **Horn** refers to the class of all Horn programs.

A set M of atoms *satisfies* a rule r if $(H(r) \cup B^-(r)) \cap M \neq \emptyset$ or $B^+(r) \setminus M \neq \emptyset$. M is a *model* of P if it satisfies all rules of P . The *Gelfond-Lifschitz (GL) reduct* of a program P under a set M of atoms is the program P^M obtained from P by first removing all rules r with $B^-(r) \cap M \neq \emptyset$ and second removing all $\neg z$ where $z \in B^-(r)$ from the remaining rules r [10]. M is an *answer-set* (or *stable set*) of a program P if M is a minimal model of P^M . We denote by $\text{AS}(P)$ the set of all answer-sets of P . The main computational problems in ASP are:

- CONSISTENCY: given a program P , does P have an answer-set?
- CREDULOUS/SKEPTICAL REASONING: given a program P and an atom $a \in \text{at}(P)$, is a contained in some/all answer-set(s) of P ?
- AS COUNTING: how many answer-sets does P have?
- AS ENUMERATION: list all answer-sets of P .

A *truth assignment* is a mapping $\tau : X \rightarrow \{0, 1\}$ defined for a set $X \subseteq U$ of atoms. For $x \in X$ we put $\tau(\neg x) = 1 - \tau(x)$. By $\text{ta}(X)$ we denote the set of all truth assignments $\tau : X \rightarrow \{0, 1\}$. Let $\tau \in \text{ta}(X)$ and P be a program.

2.1 Strong Backdoors

Backdoors are small sets of atoms which can be used to simplify the considered computational problems in ASP. They have originally been introduced by Williams, Gomes, and Selman [20, 21] as a concept to the analysis of decision heuristics in propositional satisfiability [6]. Fichte and Szeider [5] have recently adapted backdoors to the field of ASP. First, we define a reduct of a program with respect to a given set of atoms. Subsequently, we give the notion of strong backdoors. In the following we refer to \mathcal{C} as the *target class* of the backdoor.

Definition 1. Let P be a program, X a set of atoms, and $\tau \in \text{ta}(X)$. The truth assignment reduct of P under τ is the logic program P_τ obtained by

1. removing all rules r with $H(r) \cap \tau^{-1}(1) \neq \emptyset$ or $H(r) \subseteq X$;
2. removing all rules r with $B^+(r) \cap \tau^{-1}(0) \neq \emptyset$;
3. removing all rules r with $B^-(r) \cap \tau^{-1}(1) \neq \emptyset$;
4. removing from the heads and bodies of the remaining rules all literals $v, \neg v$ with $v \in X$.

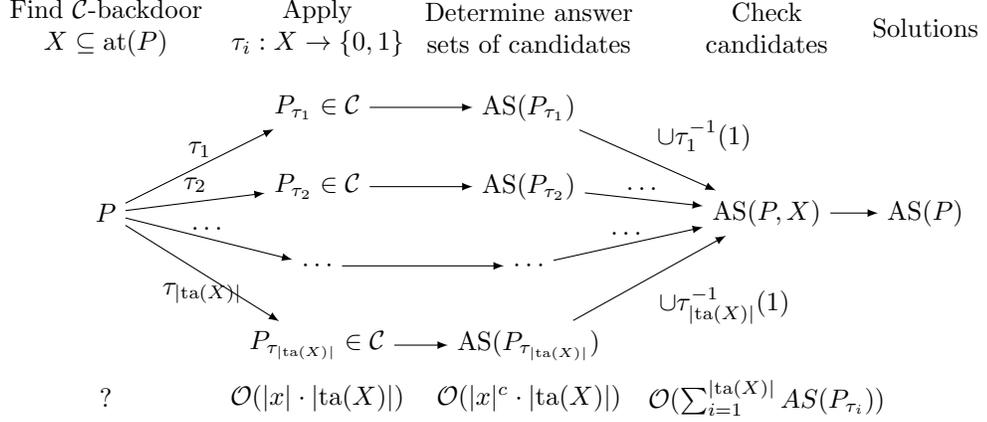


Figure 1: Exploit pattern of ASP backdoors if the target class \mathcal{C} is enumerable.

Definition 2. A set X of atoms is a strong \mathcal{C} -backdoor of a program P if $P_\tau \in \mathcal{C}$ for all truth assignments $\tau \in \text{ta}(X)$. We define the problem of finding strong backdoors as follows: **k -STRONG \mathcal{C} -BACKDOOR DETECTION**: given a program P , find a strong \mathcal{C} -backdoor X of P of size at most k , or report that such X does not exist.

Example 1. Consider the program $P = \{b \leftarrow a; d \leftarrow a; b \leftarrow \neg c; a \leftarrow d, \neg c; a \vee c \leftarrow d, \neg b; d\}$. The set $X = \{b, c\}$ is a strong **Horn**-backdoor since the truth assignment reducts $P_{b=0, c=0} = P_{00} = \{\leftarrow a; d \leftarrow a; a \leftarrow d; d\}$, $P_{01} = \{\leftarrow a; d \leftarrow a; d\}$, $P_{10} = \{d \leftarrow a; a \leftarrow d; d\}$, and $P_{11} = \{d \leftarrow a; d\}$ are in the target class **Horn**.

Definition 3. Let P be a program and X a set of atoms. We define

$$\text{AS}(P, X) = \{M \cup \tau^{-1}(1) : \tau \in \text{ta}(X \cap \text{at}(P)), M \in \text{AS}(P_\tau)\}.$$

Lemma 1 ([5]). $\text{AS}(P) \subseteq \text{AS}(P, X)$ holds for every program P and every set X of atoms.

Figure 1 illustrates how we can exploit backdoors to find answer sets of a program. Once we have found a strong \mathcal{C} -backdoor X , we can simplify the program P to programs which belong to the target class \mathcal{C} . Then we consider all $|\text{ta}(X)|$ truth assignments to the atoms in the backdoor X . We compute the answer sets $\text{AS}(P_\tau)$ for all $\tau \in \text{ta}(X)$. Finally, we obtain the answer set $\text{AS}(P)$ by checking for each $M \in \text{AS}(P_\tau)$ whether it gives rise to an answer-set of P .

Example 2. We consider the program of Example 1. The answer-sets of P_τ are $\text{AS}(P_{00}) = \{\{a, d\}\}$, $\text{AS}(P_{01}) = \{\{d\}\}$, $\text{AS}(P_{10}) = \{\{a, d\}\}$, and $\text{AS}(P_{11}) = \{\{d\}\}$. Thus $\text{AS}(P, X) = \{\{a, d\}, \{c, d\}, \{a, b, d\}, \{b, c, d\}\}$, and since $\{c, d\}$ and $\{a, b, d\}$ are answer-sets of P , we obtain $\text{AS}(P) = \{\{a, b, d\}, \{c, d\}\}$.

Definition 4. A class \mathcal{C} of programs is enumerable if for each $P \in \mathcal{C}$ we can compute $\text{AS}(P)$ in polynomial time.

2.2 Deletion Backdoors

For a program P and a set X of atoms we define $P - X$ as the program obtained from P by deleting all atoms contained in X from the heads and bodies of all the rules of P and their negations. The definition gives rise to deletion backdoors and the problem of finding deletion backdoors, which is in some cases easier to solve than the problem of finding strong backdoors.

Definition 5 (Deletion \mathcal{C} -backdoor). *Let \mathcal{C} be a class of programs. A set X of atoms is a deletion \mathcal{C} -backdoor of a program P if $P - X \in \mathcal{C}$. We define the problem k -DELETION \mathcal{C} -BACKDOOR DETECTION as follows: given a program P , find a deletion \mathcal{C} -backdoor X of P of size at most k , or report that such X does not exist.*

2.3 Target Classes

As explained above, we need to consider target classes of programs that only have a small number of answer sets. There are two causes for a program to have a large number of answer sets: (i) disjunctions in the heads of rules, and (ii) certain cyclic dependencies between rules. Disallowing both causes yields so-called *stratified* programs [9]. In the following we require normality and consider various types of acyclicity to describe target classes. In order to define acyclicity we associate with each normal program P its *directed dependency graph* D_P [1], and its *undirected dependency graph* U_P [11]. D_P has as vertices the atoms of P and a directed edge (x, y) between any two atoms x, y for which there is a rule $r \in P$ with $x \in H(r)$ and $y \in B(r)$; if there is a rule $r \in P$ with $x \in H(r)$ and $y \in B^-(r)$, then the edge (x, y) is called a *negative edge*. U_P is obtained from D_P by replacing each negative edge $e = (x, y)$ with two undirected edges $\{x, v_e\}, \{v_e, y\}$ where v_e is a new *negative vertex*, and by replacing each remaining directed edge (u, v) with an undirected edge $\{u, v\}$. By an *(un)directed cycle* of P we mean an (un)directed cycle in D_P (U_P). An (un)directed cycle is *bad* if it contains a negative edge (a negative vertex), otherwise it is *good*.

In recent research, Fichte and Szeider [5] have considered target classes that consist of normal programs without directed bad cycles (**no-DBC**), without undirected bad cycles (**no-BC**), without directed cycles (**no-DC**), and without undirected cycles (**no-C**). **no-DBC** is exactly the class that contains all stratified programs [1]. Fichte and Szeider have examined the problems k -STRONG \mathcal{C} -BACKDOOR DETECTION and k -DELETION \mathcal{C} -BACKDOOR DETECTION on the target classes $\mathcal{C} \in \{\mathbf{no-C}, \mathbf{no-BC}, \mathbf{no-DC}, \mathbf{no-DBC}\}$.

Example 3. *The set $X = \{a, b\}$ is a deletion **no-DBEC**-backdoor of the program P of Example 1, since the simplification $P - X = \{d; \leftarrow \neg c; \leftarrow d, \neg c\}$ is in the target class **no-DBEC**. We observe easily that there exists no deletion **no-DBEC**-backdoor of size 1.*

3 Parity Cycles

In this section, we generalize the acyclicity based target classes by taking the parity of the number of negative edges (vertices) into account and consider backdoors for such classes. We say that an (un)directed cycle in a given program P is *even* if the cycle has an even number of negative edges (vertices). The definition gives rise to the new target classes of all *normal* programs without directed bad even cycles (**no-DBEC**), without undirected bad even cycles (**no-BEC**), without directed even cycles (**no-DEC**), and without even cycles (**no-EC**).

Example 4. For instance in the program P of Example 1 the sequence (a, b, c, a) is a directed bad even cycle, $(a, b, v_{(b,c)}, c, v_{(c,a)}, a)$ is an undirected bad even cycle, (a, d, a) is a directed even cycle, and $(a, b, v_{(b,c)}, c, v_{(c,a)}, a)$ is an undirected even cycle (see Figure 2). The set $X = \{c\}$ is a strong **no-DBEC**-backdoor since the truth assignment reduces $P_{c=0} = P_0 = \{b \leftarrow a; d \leftarrow a; b; a \leftarrow d; a \leftarrow d, \neg b; d\}$ and $P_1 = \{b \leftarrow a; d \leftarrow a; d\}$ are in the target class **no-DBEC**. The answer-sets of P_τ are $AS(P_0) = \{\{a, b, d\}\}$ and $AS(P_1) = \{\{d\}\}$. Thus $AS(P, X) = \{\{a, b, d\}, \{c, d\}\}$, and since $\{a, b, d\}$ and $\{c, d\}$ are answer-sets of P , we obtain $AS(P) = \{\{a, b, d\}, \{c, d\}\}$.

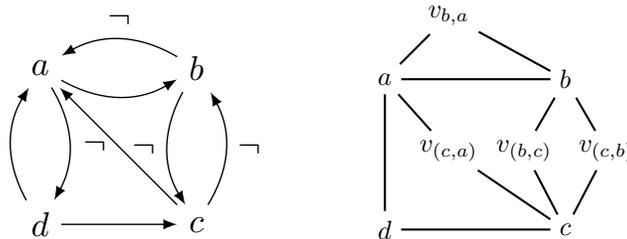


Figure 2: Directed dependency graph D_P (left) and undirected dependency graph U_P (right) of the program P of Example 1.

3.1 Computing Answer-Sets

First, we discuss the connection between the problem of finding bad even cycles in signed graphs and even cycles in graphs. A *signed (directed) graph* is a graph whose edges are either positive (unlabeled) or negative. We construct the *unlabeled directed graph* G' of a signed directed graph $G = (V, E)$ as follows: we replace in G each positive edge $e = (u, v) \in E$ by two edges $(u, v_e), (v_e, v)$ where v_e is a new vertex. Then we remove the labels from the negative edges. Analogously, we construct the *unlabeled undirected graph* where we ignore the direction of the edges. The following connection was already observed by Aracena, Gajardo, and Montalva [15].

Lemma 2 ([15]). *A signed (un)directed graph G has an even cycle if and only if its unlabeled (un)directed graph G' has a cycle of even length.*

Proof. Let $G = (V, E)$ be the signed directed graph and $G' = (V', E')$ its unlabeled directed graph. Since every positive edge $e \in E$ corresponds to two edges $e_1, e_2 \in E'$ and every negative edge $e \in E$ corresponds to one edge $e \in E'$, a cycle in G with an even number of negative edges gives a cycle of even length in G' . Conversely, let $G' = (V', E')$ be an unlabeled directed graph that contains a cycle of even length. Then G contains an even cycle since every two edges $e_1, e_2 \in E'$ correspond either to two negative edges or no negative edge. The proof works analogously for undirected graphs. \square

The well-founded reduct of a program P under an interpretation τ is the logic program P_τ^{WF} we obtain by removing all rules $r \in P$ where some $\text{at}(r) \in \tau^{-1}(0)$, and removing from all rules $r \in P$ all literals $x \in \text{at}(r)$ where $x \in \tau^{-1}(1)$. We obtain the program P^+ (P^-) by removing all rules from P where $B^+(r) \neq \emptyset$ ($B^-(r) \neq \emptyset$ respectively). The *well-founded model* $\text{WFM}(P)$ of a program P is the least fixed point of the sequence of interpretations where $\tau_0 := \emptyset$, $\tau_{k+1}^{-1}(1)$ consists of the least model of $P^+(P_\tau^{\text{WF}})$ and $\tau_{k+1}^{-1}(0)$ consists of the atoms of $P^-(P_\tau^{\text{WF}})$ that are not in the least model.

Lemma 3. *The target classes **no-DBEC**, **no-BEC**, **no-DEC**, **no-EC** are enumerable.*

Proof. Zhao [23] has shown that a program without a bad even cycle has either no answer-set or the well-founded model is its answer-set. Since in the definition of the well-founded model the sequence of τ_0, τ_1, \dots is monotone for a normal program, there is a least fixed point and it can be computed in polynomial time [7, 8]. Thus the answer sets can be computed in polynomial time. Let P be a program and D_P (U_P) its (un)directed dependency graph. Since every bad even cycle in D_P is also a bad even cycle in U_P , this holds for the undirected case. Considering the fact that every bad even cycle in D_P is also an even cycle in D_P , the lemma sustains for the target class **no-DEC**. Since every bad even cycle in D_P is also an even cycle U_P , it prevails for the remaining target class **no-EC**. \square

Proposition 1. *The problems CONSISTENCY, CREDULOUS and SKEPTICAL REASONING, AS COUNTING and AS ENUMERATION are all polynomial-time solvable for programs with strong \mathcal{C} -backdoor of bounded size, $\mathcal{C} \in \{\mathbf{no-DBEC}, \mathbf{no-BEC}, \mathbf{no-DEC}, \mathbf{no-EC}\}$, assuming that the backdoor is given as an input.*

Proof. Let X be the given backdoor. By Lemma 3 each target class \mathcal{C} is enumerable. Since we have $|\text{AS}(P, X)| \leq 2^{|\dot{X}|}$, we can solve each listed problem by making at most $2^{|\dot{X}|}$ polynomial checks. \square

If the problem of determining backdoors is also polynomial-time solvable with respect to the fixed size of a smallest strong \mathcal{C} -backdoor, then the ASP problems are polynomial-time solvable.

Lemma 4. *For all target classes $\mathcal{C} \in \{\mathbf{no-DBEC}, \mathbf{no-BEC}, \mathbf{no-DEC}, \mathbf{no-EC}\}$ every deletion \mathcal{C} -backdoor is also a strong \mathcal{C} -backdoor.*

Proof. We show the statement by proving that $P_\tau \subseteq P - X$ for every $\tau \in \text{ta}(X)$ and for every program $P \in \mathcal{C}$. Let P be a program and X a set of atoms of P . We choose arbitrarily a truth assignment $\tau \in \text{ta}(X)$. For a rule $r \in P$ if $H(r) \cap \tau^{-1}(1) \neq \emptyset$ or $H(r) \subseteq X$ or $B^+(r) \cap \tau^{-1}(0) \neq \emptyset$ or $B^-(r) \cap \tau^{-1}(1) \neq \emptyset$, then r is removed from P by the truth assignment reduct of P under τ . However removing all literals $x, \neg x$ with $x \in X$ from the head $H(r)$ and the body $B(r)$ yields a new rule $r' \in P - X$. Thus $r' \notin P_\tau$ and $r' \in P - X$ where $r' \subseteq r$. If the conditions (1), (2), and (3) of Definition 1 above do not apply, then all literals $v, \neg v$ with $v \in X$ are removed from the heads $H(r)$ and bodies $B(r)$ by the truth assignment reduct of P under τ . This is also done by $P - X$. Hence $P_\tau \subseteq P - X$. □

3.2 Backdoor Detection for Directed Target Classes

In order to apply backdoors we need to find them first. In this section we consider the problems k -STRONG \mathcal{C} -BACKDOOR DETECTION and k -DELETION \mathcal{C} -BACKDOOR DETECTION for the target classes $\mathcal{C} \in \{\mathbf{no-DEC}, \mathbf{no-DBEC}\}$.

For an unlabeled directed graph $G = (V, E)$ and fixed vertices $s, m, t \in V$ we define the *program* $P_{s,m,t}(G)$ as follows: For each edge $e = (v, w) \in E$ where $v, w \in V$ and $w \neq m$ we construct a rule $r_e: v \leftarrow w$. For the edges $e' = (v', m)$ where $v' \in V$ we construct a rule $r_{e'}: v' \leftarrow \neg m$. Then we add the rule $r_{s,t}: t \leftarrow \neg s$.

Lemma 5. *Let $G = (V, E)$ be a directed graph and s, m, t three distinct vertices of G . Then G has a simple path from s to t via m if and only if $P_{s,m,t}(G) \notin \mathbf{no-DBEC}$.*

Proof. Let G be a graph and and $p = (s, s_1, \dots, s_k, m, t_1, \dots, t_l, t)$ a path in G where $s \neq m, m \neq t, s \neq t$. The construction $P_{s,m,t}$ gives rules $\{s \leftarrow s_1; s_1 \leftarrow s_2; \dots s_k \leftarrow \neg m; m \leftarrow t_1; t_1 \leftarrow t_2; \dots t \leftarrow t; t \leftarrow \neg s\} \in P_{s,m,t}(G)$. Since D_P contains the cycle $c = (s, s_1, \dots, s_k, m, t_1, \dots, t, s)$ and c contains an even number of negative edges, the program $P_{s,m,t}(G) \notin \mathbf{no-DBEC}$.

Conversely, let $P_{s,m,t}(G) \in \mathbf{no-DBEC}$, then $P_{s,m,t}(G)$ contains a bad even cycle c . Since the construction of $P_{s,m,t}(G)$ gives only negative edges $(t, s) \in D_{P_{s,m,t}(G)}$ and $(v, m) \in D_{P_{s,m,t}(G)}$ where $v \in \text{at}(P_{s,m,t}(G))$, the cycle c must have the vertices s, m , and t . Further every rule $r_e \in P_{s,m,t}(G)$ corresponds to an edge $e \in E$. It follows that there is a simple path s, \dots, m, \dots, t . □

Theorem 1. *The problems k -STRONG $\mathbf{no-DBEC}$ -BACKDOOR DETECTION and k -DELETION $\mathbf{no-DBEC}$ -BACKDOOR DETECTION are co-NP-hard for every constant $k \geq 0$.*

Proof. Let $k \geq 0$. Let G be a given directed graph and t, m, s vertices of G . It was shown by Lapaugh and Papadimitriou [12] that deciding whether G contains a simple path from s to t via m is NP-complete. By Lemma 5, such a path exists if and only if $P_{s,m,t}(G) \notin \mathbf{no-DBEC}$, hence recognizing $\mathbf{no-DBEC}$ is co-NP-hard. Let G_k denote the graph obtained from G by adding k disjoint bad even cycles. Clearly G_k has a deletion $\mathbf{no-DBEC}$ -Backdoor of size $\leq k$ if and only if $P_{s,m,t}(G_k) \in \mathbf{no-DBEC}$, hence k -DELETION $\mathbf{no-DBEC}$ -BACKDOOR DETECTION is co-NP-hard. Similarly, G_k has a strong $\mathbf{no-DBEC}$ -Backdoor of size $\leq k$ if and only if $P_{s,m,t}(G_k) \in \mathbf{no-DBEC}$, and so k -STRONG $\mathbf{no-DBEC}$ -BACKDOOR DETECTION is co-NP-hard as well. \square

Theorem 2. *Let $k > 0$ be a constant. The problems k -DELETION $\mathbf{no-DEC}$ -BACKDOOR DETECTION and k -STRONG $\mathbf{no-DEC}$ -BACKDOOR DETECTION are polynomial-time tractable.*

Proof. By Lemma 2, we can reduce to the problem of finding a cycle of even length in the unlabeled dependency graph. Vazirani and Yannakakis [19] have shown that finding a cycle of even length in a directed graph is equivalent to finding a Pfaffian orientation of a graph. Since Robertson, Seymour, and Thomas [17] have shown that a Pfaffian orientation can be found in polynomial time. For each possible backdoor of size k we need to test $\binom{n}{k} \leq n^k$ subsets $S \subseteq V$ of size k whether $D_P - S$ contains a cycle of even length, respectively D_{P_τ} for $\tau \in \text{ta}(S)$. Since we can do this in polynomial time for each fixed k , the theorem follows. \square

In Theorem 2 we consider k as a constant. In the following proposition we show that if k is considered as part of the input, then the problem k -STRONG $\mathbf{no-DEC}$ -BACKDOOR DETECTION is polynomial-time equivalent to the problem HITTING SET and k is preserved. An instance of this problem is a pair (\mathbf{S}, k) where $\mathbf{S} = \{S_1, \dots, S_m\}$ is a family of sets and k is an integer. The question is whether there exists a set H of size at most k which intersects with all the S_i ; such H is a hitting set. Note that there is strong theoretical evidence that the problem HITTING SET does not admit uniform polynomial-time tractability [3].

Proposition 2. *The problem k -STRONG $\mathbf{no-DEC}$ -BACKDOOR DETECTION is polynomial-time equivalent to the problem HITTING SET.*

Proof. The proof is very similar to the proof for target classes without respecting the parity by Fichte and Szeider [5]. We construct a program P as follows. As atoms we take the elements of $\mathcal{S} = \bigcup_{i=1}^m S_i$ and new atoms a_i^j and b_i^j for $1 \leq i \leq m$, $1 \leq j \leq k+1$. For each $1 \leq i \leq m$ and $1 \leq j \leq k+1$ we take two rules r_i^j, s_i^j where $H(r_i^j) = \{a_i^j\}$, $B^-(r_i^j) = S_i \cup \{b_i^j\}$, $B^+(r_i^j) = \emptyset$; $H(s_i^j) = \{b_i^j\}$, $B^-(s_i^j) = \{a_i^j\}$, $B^+(s_i^j) = \mathcal{S}$.

We show that \mathbf{S} has a hitting set of size at most k if and only if P has a strong $\mathbf{no-DEC}$ -backdoor of size at most k . Let \mathbf{S} be a family of sets and H an hitting set of \mathbf{S} of size at most k . Choose arbitrarily an atom $s_i \in \text{at}(P) \cap \mathcal{S}$ and a truth assignment $\tau \in \text{ta}(H)$. If $s_i \in \tau^{-1}(0)$, then $B^+(s_i^j) \cap \tau^{-1}(0) \neq \emptyset$

for $1 \leq j \leq k + 1$. Thus $s_i^j \notin P_\tau$. If $s_i \in \tau^{-1}(1)$, then $B^-(r_i^j) \cap \tau^{-1}(1) \neq \emptyset$ for $1 \leq j \leq k + 1$. Thus $r_i^j \notin P_\tau$. Since H contains at least one element $e \in S$ from each set $S \in \mathcal{S}$, the truth assignment reduct $P_\tau \in \mathbf{no-DEC}$. We conclude that H is a strong **no-DEC**-backdoor of P of size at most k .

Conversely, let X be a strong **no-DEC**-backdoor of P of size at most k . Since the directed dependency graph D_P contains $k + 1$ directed even cycles (a_i^j, b_i^j, a_i^j) and a_i^j (respectively b_i^j) is contained in exactly one rule r_i^j (respectively s_i^j), $|\bigcup a_i^j| > k$ and $|\bigcup b_i^j| > k$. Hence we have to select atoms from S_i . Since $S_i \subseteq B^-(r_i^j)$ for $1 \leq i \leq m$ and $1 \leq j \leq k + 1$, we have to select at least one element from each S_i into the backdoor X . Thus we have established that X is a hitting set of S , and so the theorem follows. \square

3.3 Backdoor Detection for Undirected Target Classes

The results of Theorem 1 suggest to consider the backdoor detection on the weaker target classes based on undirected even acyclicity.

Lemma 6. *Let P be a program, $P \in \mathbf{no-EC}$ can be decided in polynomial time.*

Proof. Let P be a program and G its dependency graph U_P . Lemma 2 allows to consider the problem of finding an even cycle in the unlabeled version of U_P . Since Yuster and Zwick [22] have shown that finding an even cycle in an undirected graph is polynomial-time solvable, the lemma holds. \square

Lemma 7. *Let P be a program. The problem of deciding whether $P \in \mathbf{no-BEC}$ can be solved in polynomial time.*

Proof. Let P be a program and G its dependency graph U_P . For a negative edge e of G we define G_e to be the unlabeled graph of $G - e$. Now G contains a bad even cycle if and only if G has an edge $e = \{s, t\}$ such that G_e contains an odd path from s to t . Since Arikati and Peled [2] have shown that finding an odd path in an undirected graph is polynomial-time solvable, the lemma follows. \square

Theorem 3. *Let $k > 0$ be a constant. For the target classes $\mathcal{C} \in \{\mathbf{no-EC}, \mathbf{no-BEC}\}$ the problems k -DELETION \mathcal{C} -BACKDOOR DETECTION and k -STRONG \mathcal{C} -BACKDOOR DETECTION are non-uniform polynomial-time tractable.*

Proof. Let P be a program and $U_P = (V, E)$ its undirected dependency graph. Let n be the size of V . For each possible backdoor of size k we need to test $\binom{n}{k} \leq n^k$ subsets $S \subseteq V$ of size k whether $U_P - S$ contains a (bad) cycle of even length, respectively U_{P_τ} for $\tau \in \text{ta}(S)$. Since we can do this in polynomial time for each fixed k , the problems k -DELETION \mathcal{C} -BACKDOOR and k -STRONG \mathcal{C} -BACKDOOR DETECTION are non-uniform polynomial-time tractable. \square

In Theorem 3 we consider k as a constant. If k is considered as part of the input we can show that for each class $\mathcal{C} \in \{\mathbf{no-EC}, \mathbf{no-BEC}\}$ the problem k -STRONG \mathcal{C} -BACKDOOR DETECTION is polynomial-time equivalent to HITTING SET [5]. As mentioned before for **no-DEC** there is strong theoretical

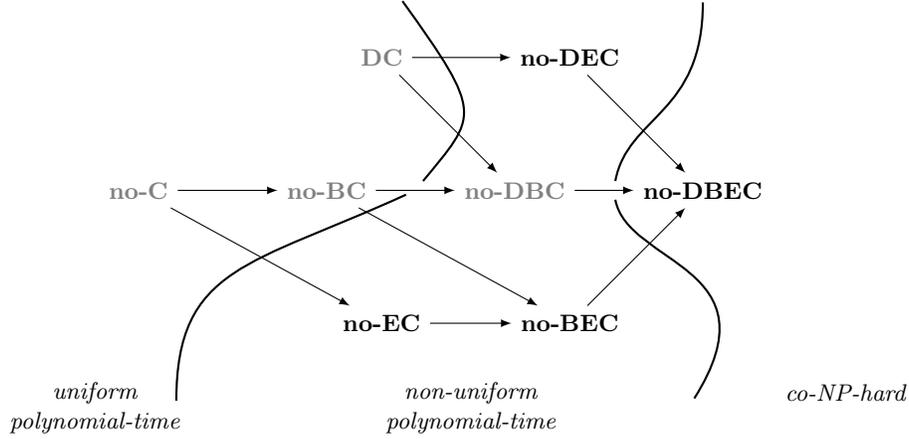


Figure 3: Relationship between classes of programs and state of the knowledge regarding the complexity of the problem DELETION \mathcal{C} -BACKDOOR. The new results are colored in black.

evidence that k -STRONG \mathcal{C} -BACKDOOR DETECTION does not admit a uniform polynomial-time tractability result.

Proposition 3. *The problem k -STRONG \mathcal{C} -BACKDOOR DETECTION is polynomial-time equivalent to the problem HITTING SET for each class $\mathcal{C} \in \{\mathbf{no-EC}, \mathbf{no-BEC}\}$.*

Proof. We modify the above reduction from HITTING SET by redefining the rules r_i^j, s_i^j . We put $H(r_i^j) = \{a_i^j\}$, $B^-(r_i^j) = S_i \cup \{b_i^j\}$, $B^+(r_i^j) = S_i$; $H(s_i^j) = \{b_i^j\}$, $B^-(s_i^j) = \{a_i^j\}$, $B^+(s_i^j) = \emptyset$. □

4 Relationship between Target Classes

In this section, we compare ASP parameters in terms of their *generality*. We have already observed that every deletion \mathcal{C} -backdoor is a strong \mathcal{C} -backdoor for a target class $\mathcal{C} \in \{\mathbf{no-EC}, \mathbf{no-DEC}, \mathbf{no-BEC}, \mathbf{no-DBEC}\}$. For the considered target classes it is easy to see that if $\mathcal{C} \subseteq \mathcal{C}'$, then every \mathcal{C}' backdoor of a program P is also a \mathcal{C} -backdoor, but there might exist smaller \mathcal{C}' -backdoors. Thus we compare the target classes among each other instead of the backdoors. By definition we have $\mathbf{no-DBC} \subsetneq \mathbf{no-DBEC}$, $\mathbf{no-DEC} \subsetneq \mathbf{no-DBEC}$, $\mathbf{no-EC} \subsetneq \mathbf{no-BEC}$, $\mathbf{no-C} \subsetneq \mathbf{no-EC}$, and $\mathbf{no-DC} \subsetneq \mathbf{no-DEC}$. The diagram in Fig. 3 shows the relationship between the various classes, an arrow from \mathcal{C} to \mathcal{C}' indicates that \mathcal{C} is a proper subset of \mathcal{C}' . If there is no arrow between two classes (or the arrow does not follow by transitivity of set inclusion), then the two classes are incomparable.

Lin and Zhao [13] have studied even cycles as a parameter to ASP. They have proved that for fixed k the main reasoning problems are polynomial-time solvable if the number of the shortest even cycles is bounded. The following proposition states that size of **no-DBEC**-backdoors is a more general parameter than the number of even cycles.

Proposition 4. *There is a function f such that $k \leq f(l)$ and no function g such that $l < g(k)$ for all programs P where k is the size of the smallest deletion-**no-DBEC**-backdoor of P and l is the number of even cycles in D_P .*

Proof. Let P be some program. If P has at most k bad even cycles, we can construct a **no-DBEC**-backdoor X for P by taking one element from each bad even cycle into X . Thus there is a function f such that $k \leq f(l)$. If a program P has a **no-DBEC**-backdoor of size 1, it can have arbitrary many even cycles that run through the atom in the backdoor. It follows that there is no function g such that $l < g(k)$ and the proposition holds. \square

5 Conclusion

We have extended the backdoor approach of [5] by taking the parity of the number of negative edges on bad cycles into account. In particular, this allowed us to consider target classes that contain non-stratified programs. We have established new hardness results and non-uniform polynomial-time tractability depending on whether we consider directed or undirected even cycles. We have shown that the backdoor approach with parity target classes generalize a result of Lin and Zhao [13]. Since Theorem 1 states that target classes based on directed even cycles are intractable, we think these target classes are of limited practical interest. The results of this paper give rise to research questions that are of theoretical interest. For instance, it would be stimulating to find out whether the problem k -STRONG \mathcal{C} -BACKDOOR DETECTION is uniform polynomial-time solvable (fixed-parameter tractable) for the classes **no-BC** and **no-BEC**, which is related to the problems parity feedback vertex set and parity subset feedback vertex set.

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