

Generating all vertices of a polyhedron is hard*

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Abstract

We show that generating all negative cycles of a weighted graph is a hard enumeration problem, in both the directed and undirected cases. More precisely, given a family of negative (directed) cycles, it is an NP-complete problem to decide whether this family can be extended or there are no other negative (directed) cycles in the graph, implying that (directed) negative cycles cannot be generated in polynomial output time, unless $P=NP$. As a corollary, we solve in the negative two well-known generating problems from linear programming: (i) Given an infeasible system of linear inequalities, generating all minimal infeasible subsystems is hard. Yet, for generating maximal feasible subsystems the complexity remains open. (ii) Given a feasible system of linear inequalities, generating all vertices of the corresponding polyhedron is hard. Yet, in the case of bounded polyhedra the complexity remains open. Equivalently the complexity of generating vertices and extreme rays of polyhedra remains open.

keywords polytope, polyhedron, polytope-polyhedron problem, vertex, face, facet, enumeration problem, vertex enumeration, facet enumeration, graph, cycle, negative cycle, linear inequalities, feasible system

1 Introduction and Main Results

Let $G = (V, E)$ be a directed graph (digraph) and $w : E \rightarrow \mathbb{R}$ be a real-valued weight function defined on its arcs. We call such a pair a *weighted digraph* and denote it by (G, w) . For every subset of arcs $F \subseteq E$ its weight is defined as the total weight of all its arcs, $w(F) = \sum_{e \in F} w(e)$. Let us call a simple directed cycle a *circuit*. A circuit is called *negative* if its weight is negative. Finally,

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let us denote by $\mathcal{C}^- = \mathcal{C}^-(G, w)$ the family of negative circuits of (G, w) , i.e., $\mathcal{C}^- = \{C \subseteq E \mid C \text{ is a circuit with } w(C) < 0\}$.

First we consider the problem of generating exhaustively all negative circuits of a given weighted directed graph (G, w) , in other words the problem of enumerating the family $\mathcal{C}^-(G, w)$. Since the number of negative circuits may be exponential in the size of the input description, i.e., the size of G and w , the efficiency of such enumeration algorithms is measured customarily in both the input and output sizes (see e.g., [39, 29, 26]). More precisely, such an enumeration problem is said to be solvable in *polynomial total time* if the output can be generated in time polynomial in the input and output sizes. It is easy to see that for self-reducible problems a family \mathcal{C} is enumerable in polynomial total time if and only if for each subfamily $\mathcal{X} \subseteq \mathcal{C}$, the problem of deciding $\mathcal{X} \neq \mathcal{C}$, and if yes, finding $C \in \mathcal{C} \setminus \mathcal{X}$, is solvable in time polynomial in $size(G, w)$ and $|\mathcal{X}|$. On the other hand, when this decision problem is NP-hard, the enumeration problem is called NP-hard, too (see [29]). Thus, NP-hard enumeration problems are unlikely to have total polynomial time enumeration algorithms, unless $P=NP$.

Our main result claims that enumerating negative circuits of a weighted directed graph is a hard enumeration problem.

Theorem 1 *Given a weighted digraph $G = (V, E)$, $w : E \rightarrow \mathbb{R}$ and a family $\mathcal{X} \subseteq \mathcal{C}^-$ of its negative circuits, it is an NP-complete problem to decide whether $\mathcal{X} \neq \mathcal{C}^-$, even if w takes only two different values.*

Let us add that the analogous hardness result can be shown for undirected graphs, as well. In this case we also call a simple cycle a circuit and we denote by $\mathcal{C}^- = \mathcal{C}^-(G, w)$ the family of all negative circuits of an undirected graph $G = (V, E)$.

Theorem 2 *Given a weighted undirected graph $G = (V, E)$, $w : E \rightarrow \mathbb{R}$, and a family $\mathcal{X} \subseteq \mathcal{C}^-(G, w)$ of its negative circuits, it is an NP-complete problem to decide whether $\mathcal{X} \neq \mathcal{C}^-$, even if w takes only two different values.*

Let us remark that all circuits of a directed or undirected graph can be enumerated efficiently, e.g., by a simple backtracking algorithm [35].

Let us note that if w takes the same value for all edges (arcs), then negative circuits either do not exist, or all circuits are negative. Thus, the enumeration problems for both directed and undirected graphs can efficiently be solved, as we noted above. Furthermore, when w takes only two different values, those can be assumed to be integers, and hence by edge (arc) splitting the input can be transformed to one in which all edges (arcs) have weight ± 1 . Though this transformation may increase the size of the input in a non polynomial way, in case of the specific constructions we provide in the proofs of the above two theorems, it is a polynomial transformation, implying that generating all negative circuits is NP-hard even if all edges (arcs) have weights ± 1 .

We shall derive several consequences of the above results, including the hardness of generating all vertices of a (possibly unbounded) polyhedron, generating all minimal infeasible subsystems of a system of linear inequalities, etc. We prove Theorems 1 and 2 in sections 2 and 3, respectively.

1.1 Negative Circuits and Minimal Infeasible Subsystems

Let us first note that deciding the existence and finding a negative circuit in a weighted directed graph are polynomially solvable tasks. Gallai [24] proved that (G, w) has no negative circuit if and only if by a potential transformation all edge weights can be changed to nonnegative values. Furthermore, a negative circuit can be found in $O(|V|^3)$ time, if the graph has negative circuits at all [21, 40]. We shall utilize Gallai's approach to reformulate the problem and derive some interesting consequences.

To a weighted digraph (G, w) , where $G = (V, E)$ and $w : E \rightarrow \mathbb{R}$, let us associate a polyhedron $P(E, w)$ defined by

$$P(E, w) = \{x \in \mathbb{R}^V \mid x_u - x_v \geq -w(u, v) \text{ for all arcs } (u, v) \in E\} \quad (1)$$

Let us note that every vector $x \in P(E, w)$ is a potential in the sense Gallai [24] defined it, proving that G is negative circuit free. Namely, defining $w'(u, v) = w(u, v) + x_u - x_v$ for all arcs $(u, v) \in E$ we get another weighting of the arcs of G , such that $w'(C) = w(C)$ for all directed circuits $C \subseteq E$, and for which $w'(u, v) \geq 0$ for all arcs $(u, v) \in E$, according to the definition of $P(E, w)$. This latter shows that G is indeed negative cycle free.

Applying thus Gallai's result to subgraphs of G we obtain that $P(E', w) = \emptyset$ for some $E' \subseteq E$ if and only if the subgraph $G' = (V, E')$ contains a negative cycle with respect to the weight function w . Therefore, the minimal infeasible subsystems of the system of linear inequalities (1) correspond in a one-to-one way to the negative circuits of (G, w) . Hence, Theorem 1 implies the following result.

Corollary 1 *Enumerating all minimal infeasible subsystems of a system of linear inequalities is an NP-hard enumeration problem, even if we restrict the input to linear systems involving at most two variables in each inequality.* \square

The problems of finding minimal infeasible subsystems of a system of linear inequalities, sometimes called *IIS (Irreducible Inconsistent Subsystems)* or *Helly systems*, and its natural dual of finding maximal feasible subsystems received ample attention in the literature, see e.g., [36, 33, 2]. The optimization versions of these problems, i.e., finding a maximum cardinality feasible subsystem, and finding a minimum cardinality infeasible subsystem are known to be NP-hard, see e.g., [27, 11, 33].

1.2 Negative Circuits and Vertex Enumeration

Let us recall that the infeasibility of a system of linear inequalities is well characterized by Farkas' lemma: either the system $Ax \geq b$ has a solution, or there exists a nonnegative vector $y \geq 0$ such that $y^T A = 0$ and $y^T b > 0$, but not both (see [20]). Using this claim, [25] associated to a system of linear inequalities $Ax \geq b$, $A \in \mathbb{R}^{m \times n}$ and $b \in \mathbb{R}^m$, a so called *alternative polyhedron* defined as $Q = \{y \in \mathbb{R}_+^m \mid y^T A = 0, y^T b = 1\}$, and observed that minimal infeasible subsystems of $Ax \geq b$ are in a one-to-one correspondence with vertices of Q . Indeed, for every vector $y \in Q$ let us consider the subsystem of $Ax \geq b$ corresponding to the support set $S(y) = \{i \mid y_i \neq 0\}$. By Farkas' lemma, we have that these corresponding subsystems are indeed infeasible. Conversely, if S is the index set of an infeasible subsystem of $Ax \geq b$, then again by Farkas' lemma we have a vector $y \in Q$ for which $S(y) \subseteq S$. Thus, minimal infeasible subsystems correspond to vectors $y \in Q$ with minimal support sets, and hence those are indeed vertices of Q .

This observation, coupled with Corollary 1 implies the hardness of enumerating the vertices of polyhedra.

Corollary 2 *Enumerating all vertices of a rational polyhedron, given as the intersection of finitely many closed half-spaces, is an NP-hard enumeration problem.*

Proof: Let us consider an infeasible system of rational linear inequalities $Ax \geq b$, and its alternative polyhedron Q . We can write Q equivalently as $Q = \{y \in \mathbb{R}^m \mid y \geq 0, A^T y \geq 0, -A^T y \geq 0, b^T y \geq 1, -b^T y \geq -1\}$, i.e., as the intersection of $m + 2n + 2$ closed half-spaces. Thus, by the above observation, enumerating the vertices of this rational polyhedron would also enumerate all minimal infeasible subsystems of $Ax \geq b$, which is an NP-hard enumeration problem according to Corollary 1. \square

Vertex enumeration is a fundamental problem in computational geometry and polyhedral combinatorics (see e.g., [18] for a list of applications), and has many equivalent formulations. Most notably for bounded polyhedra vertex enumeration is equivalent with *facet generation*, i.e., enumerating the facets of a polytope given by an explicit list of its vertices (see e.g., the so called polytope-polyhedron problem in [30]).

Let us emphasize that whenever the system of equations $A^T y = 0$, $b^T y = 0$ has a nontrivial solution for which $y \geq 0$, then Q in the above Corollary 2 is an unbounded polyhedron. Thus, our reduction through Theorem 1 yields in general, unbounded polyhedra, and hence does not imply the hardness of vertex generation for bounded polyhedra, which still remains an open problem. Furthermore and equivalently, the complexity of enumerating together vertices and extreme rays of polyhedra is also an open problem.

Numerous algorithmic ideas have been introduced in the literature (either for vertex or for facet enumeration, see e.g., [13, 32, 6, 12, 31, 18, 15, 17, 38, 37, 4, 34, 5, 9, 10, 1]). Efficient algorithms (typically linear in the number of vertices) were proposed for several special cases, including simple polyhedra, i.e. in which every vertex is incident with exactly n facets, [4], network polytopes [34], polytopes with zero-one vertices [10], and polyhedra in which every facet defining inequality involves at most two nonzero coefficients [1]. Furthermore, for fixed dimension both vertices and rays of a polyhedron can be enumerated efficiently [14]. However, no method proved to be efficient (yet) for the general case. In fact several publications [3, 23, 8] analyzed the proposed general purpose methods for vertex/facet enumeration, and showed that all of the known algorithms may require in the worst case superpolynomial time in the output size. Along the same lines, Corollary 2 shows that vertex enumeration is indeed a hard enumeration problem for unbounded polyhedra (unless of course $P=NP$).

In analyzing the reasons why backtracking methods are not efficient for vertex enumeration, in general, [23] noted that such methods require solving repeatedly decision problems, which turn out to be NP-hard. In particular, they showed that for a given rational polyhedron P and an open rational half-space $H = \{x \in \mathbb{R}^n \mid \alpha^T x > \beta\}$, it is NP-hard to decide if P has a vertex in H . Let us note that the same decision problem for bounded polyhedra is much easier, since it can be decided by maximizing $\alpha^T x$ over P , which is a linear programming problem, known to be polynomially solvable by [28]. We can show, as a next corollary of Theorem 1 that the enumerative version of this decision problem is hard, already for bounded polyhedra.

To arrive to this claim, let us recall that the nonzero vertices of the circulation polytope

$$P(G) = \left\{ y \in \mathbb{R}^E \mid \begin{array}{l} \sum_{v:(u,v) \in E} y_{uv} - \sum_{w:(w,u) \in E} y_{wu} = 0 \quad \forall u \in V \\ \sum_{(u,v) \in E} y_{uv} = 1 \\ 0 \leq y_{uv} \quad \forall (u,v) \in E \end{array} \right\}$$

of a directed graph $G = (V, E)$ correspond to circuits of G , namely for every vertex y of $P(G)$ its support set $S(y) = \{(u, v) \in E \mid y_{uv} \neq 0\}$ is a circuit in G . Let us remark that $P(G)$ is frequently occurring in the optimization literature, the vertices and facial structure of which are well studied and understood. In particular, the vertices of $P(G)$ can be generated in linear (output) time by cycle enumeration [35] (or by the method proposed in [10]).

Associating further to a rational weight function $w : E \rightarrow \mathbb{R}$ an open rational half-space defined by

$$H = \left\{ y \in \mathbb{R}^E \mid \sum_{(u,v) \in E} w(u,v)y_{uv} < 0 \right\}$$

we get that the support sets of vertices of $P(G)$ belonging to H are exactly the negative circuits of the weighted directed graph (G, w) . Thus, Theorem 1 readily implies the following claim.

Corollary 3 *Given a rational polyhedron P and an open rational half-space H , it is NP-hard to enumerate all vertices of P which belong to H , even if P is bounded.* \square

Many applications (see e.g. [18]) call for the enumeration of all those basic feasible solutions to a linear programming problem (i.e., vertices of the corresponding polyhedron), the corresponding objective function value of which is above a given threshold. Corollary 3 indicates that unfortunately such enumeration problems are difficult in general, unless $P=NP$.

A further consequence of Theorem 1 is that enumerating all vertices of a bounded polyhedron P which do not belong to a given face of P is also hard, in general.

Corollary 4 *Given a bounded polyhedron P and a proper face F of it, it is NP-hard to enumerate the vertices of P which do not belong to F .*

Proof: Note that $P' = P(G) \cap H$, as defined above, is a bounded polyhedron, for which H is facet defining, and the vertices of which outside H correspond in a one-to-one way to the negative circuits of the weighted graph (G, w) to which we associated H and $P(G)$. Thus the claim follows by Theorem 1. \square

By Corollary 2 there exists no algorithm to output in incremental (or total) polynomial time, the vertices and then the extreme directions of a polyhedron, in that order. In contrast we have the following statement.

Proposition 1 *If there exists an algorithm which enumerates all vertices of a bounded polyhedron in incremental polynomial time then we can enumerate all extreme rays and then all vertices (in this order) of a polyhedron in incremental polynomial time.*

Proof: Let $P = \{x \in \mathbb{R}^n : a_i^T x \leq b_i, i = 1, \dots, m\}$ be an unbounded polyhedron and let V and R denote the set of vertices and the set of extreme rays of P , respectively. We can assume that V contains at least one vertex v , since checking if $V = \emptyset$ or finding a vertex of P can be done in polynomial time. Let $a = \sum_{i: a_i^T v = b_i} a_i$. Then $P' = \{x : a_i^T x \leq b_i, i = 1, \dots, m, a^T x = -M\}$ is a bounded polyhedron whose vertices correspond to R . Furthermore $P'' = \{x : a_i^T x \leq b_i, i = 1, \dots, m, a^T x \geq -M\}$ is a bounded polyhedron whose vertices correspond to $V \cup R$.

We assume that we have an incremental polynomial time algorithm **A** to enumerate all vertices of a bounded polyhedron. This implies that given a

subset of vertices of a bounded polyhedron \mathbf{A} can find another vertex of this polyhedron or determine that there is none in time polynomial in the combined size of the input and the given subset of vertices.

By applying \mathbf{A} to P'' we can enumerate $V \cup R$ efficiently. Note that enumerating V is NP-hard by Corollary 4. To enumerate R and then V we apply \mathbf{A} to P' and then starting with R we apply \mathbf{A} to P'' to find the remaining vertices of P'' . \square

1.3 Four Geometric Enumeration Problems

Let us finally recall four strongly related geometric enumeration problems. Let $\mathcal{A} \subseteq \mathbb{R}^n$ be a given subset of vectors in \mathbb{R}^n , fix a point $z \in \mathbb{R}^n$ called the *center*, and consider the following four definitions:

- A *simplex* is a minimal subset $X \subseteq \mathcal{A}$ containing the center in its convex hull, i.e., $z \in \text{conv}(X)$.
- An *anti-simplex* is a maximal subset $X \subseteq \mathcal{A}$ not containing the center in its convex hull, i.e., $z \notin \text{conv}(X)$.
- A *body* is a minimal (full-dimensional) subset $X \subseteq \mathcal{A}$ containing the center in the interior of its convex hull, i.e., $z \in \text{int}(\text{conv}(X))$.
- An *anti-body* is a maximal subset $X \subseteq \mathcal{A}$ not containing the center in the interior of its convex hull, i.e., $z \notin \text{int}(\text{conv}(X))$.

Equivalently, a simplex (body) is a minimal collection of the given vectors not contained in an *open* (*closed*) half-space through the center, while an anti-simplex (anti-body) is a maximal collection of vectors contained in an open (*closed*) half space through the center. It can be seen easily that $|X| \leq n + 1$ for a simplex, and that $n + 1 \leq |X| \leq 2n$ for a body.

In the sequel we assume that the center is at the origin, i.e., $z = 0$. For a given point set $\mathcal{A} \subseteq \mathbb{R}^n$ let us denote respectively by \mathcal{S} and \mathcal{B} the hypergraphs on the base set \mathcal{A} , consisting of all simplices, and respectively all bodies of \mathcal{A} . The corresponding families of maximal independent sets of these two hypergraphs are respectively all anti-simplices and anti-bodies of \mathcal{A} , denoted respectively by \mathcal{S}^* and \mathcal{B}^* , i.e.,

$$\begin{aligned} \mathcal{S}^* &= \{ X \subseteq \mathcal{A} \mid X \text{ is maximal such that } X \not\subseteq S \ \forall S \in \mathcal{S} \}, \\ \mathcal{B}^* &= \{ Y \subseteq \mathcal{A} \mid Y \text{ is maximal such that } Y \not\subseteq B \ \forall B \in \mathcal{B} \}. \end{aligned}$$

Simplices, anti-simplices, bodies and anti-bodies can naturally be related to minimal infeasible or maximal feasible subsystems of certain linear systems of inequalities. Namely, let us denote by $A \in \mathbb{R}^{m \times n}$, where $m = |\mathcal{A}|$, the matrix whose row vectors are the vectors of \mathcal{A} , and let $e \in \mathbb{R}^m$ denote the m -dimensional vector of all ones.

It follows from the above definitions that simplices and anti-simplices are in a one-to-one correspondence respectively with the minimal infeasible and maximal feasible subsystems of the linear system of inequalities:

$$Ax \geq \mathbf{e}, \quad x \in \mathbb{R}^n. \quad (2)$$

Similarly, it follows that bodies and anti-bodies correspond, in a one-to-one way, respectively to the minimal infeasible and maximal feasible subsystems of the system:

$$Ax \geq 0, \quad x \neq 0. \quad (3)$$

As of the complexity of these enumeration problems, it is known that the generation of anti-bodies is a hard problem:

Proposition 2 ([7]) *Given a set of vectors $\mathcal{A} \subseteq \mathbb{R}^n$, and a partial list $\mathcal{X} \subseteq \mathcal{B}^*$ of the anti-bodies of S , it is NP-hard to determine if the given list is incomplete, i.e. $\mathcal{X} \neq \mathcal{B}^*$, or not. Equivalently, given an infeasible system (3), and a partial list of its maximal feasible subsystems, it is NP-hard to determine if the given partial list is incomplete, or not. \square*

Enumeration of bodies turns out to be at least as hard as the well-known *hypergraph transversal problem* [19], which is not known to be solvable in incremental polynomial time.

Proposition 3 ([7]) *The problem of incrementally enumerating bodies, for a given set of $m + n$ points $\mathcal{A} \subseteq \mathbb{R}^n$, includes as a special case the problem of enumerating all minimal transversals for a given hypergraph \mathcal{H} with n hyperedges on m vertices. Equivalently, generating minimal infeasible subsystems of (3) is at least as hard as hypergraph transversal generation. \square*

It should be added that the best currently known algorithm for the hypergraph transversal problem runs in incremental *quasi-polynomial* time [22].

The problem of generating simplices turns out to be equivalent, in general, with the problem of enumerating the vertices of bounded polyhedra, or enumerating the vertices and extreme rays of possibly unbounded polyhedra. To see this, let us consider a vector set $\mathcal{A} = \{a_1, \dots, a_n, b\} \subseteq \mathbb{R}^d$ and associate to it a polyhedron $P = \{x \in \mathbb{R}^n \mid a_i^T x = -b_i, i = 1, \dots, n, x \geq 0\}$.

Recall that for a vector $y \in \mathbb{R}^d$ we called the set $S(y) = \{i \mid y_i \neq 0\}$ its support set.

Proposition 4 *If $y \in P$ is a vertex of P then the set $\{a_i \mid i \in S(y)\} \cup \{b\}$ is a simplex of \mathcal{A} , while if $y \in P$ is an extreme ray of P then the set $\{a_i \mid i \in S(y)\}$ is a simplex of \mathcal{A} . Furthermore, every simplex of \mathcal{A} corresponds in this way either to a vertex or to an extreme ray of P .*

Proof: The first two claims are easy to see by the definitions. For the last claim, let $S \subseteq \mathcal{A}$ be a simplex, i.e., a minimal subset for which $0 \in \text{conv}(S)$. If $b \in S$, then we have for some $\lambda_a \geq 0$, $a \in S \setminus \{b\}$ and $\lambda_b \geq 0$, with $\lambda_b + \sum_{a \in S \setminus \{b\}} \lambda_a = 1$ that

$$-\lambda_b b = \sum_{a \in S \setminus \{b\}} \lambda_a a.$$

Since S is minimal, we must have all these coefficients positive, and thus

$$-b = \sum_{a \in S \setminus \{b\}} \frac{\lambda_a}{\lambda_b} a.$$

Thus the vector $x \in \mathbb{R}^n$, defined by

$$x_i = \begin{cases} \frac{\lambda_{a_i}}{\lambda_b} & \text{if } a_i \in S \setminus \{b\}, \\ 0 & \text{otherwise} \end{cases}$$

for $i = 1, \dots, n$, is a vertex of P , again by the minimality of S . While if $b \notin S$, then we have

$$0 = \sum_{a \in S} \lambda_a a$$

for some positive coefficients $\lambda_a > 0$, $a \in S$ for which $\sum_{a \in S} \lambda_a = 1$, and thus the vector $x \in \mathbb{R}^n$, defined by

$$x_i = \begin{cases} \lambda_{a_i} & \text{if } a_i \in S, \\ 0 & \text{otherwise} \end{cases}$$

for $i = 1, \dots, n$, is an extreme ray of P , once more by the minimality of S . \square

In particular, if $P = \{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$ is a bounded polyhedron, i.e., if $Ax = 0$ has no nontrivial nonnegative solutions, then the vertices of P correspond in a one-to-one way to the simplices of the set \mathcal{A} formed by the column vectors of A and b .

For the special case of vectors $\mathcal{A} \subseteq \mathbb{R}^n$ in general position, we have $\mathcal{B} = \mathcal{S}$, and consequently the problem of enumerating bodies of \mathcal{A} turns into the problem of enumerating vertices of the bounded polyhedron $\{x \in \mathbb{R}^n \mid Ax = 0, \mathbf{e}^T x = 1, x \geq 0\}$, each vertex of which is non-degenerate and has exactly $n + 1$ positive components. For such kinds of *simple* bounded polyhedra there exist algorithms that generate all vertices with polynomial delay (see e.g., [15, 4]).

Let us finally mention that, although the status of the problem of enumerating all maximal feasible subsystems of (2) is not known in general, the situation changes if we fix a consistent subfamily of inequalities, and ask for enumerating all its extensions to a maximal feasible subsystem. In fact, such a problem turns out to be NP-hard, even if we fix only non-negativity constraints.

Proposition 5 ([7]) *Let $A \in \mathbb{R}^{m \times n}$ be an $m \times n$ matrix, $b \in \mathbb{R}^m$ be an m -dimensional vector, and assume that the system*

$$Ax \geq b, \quad x \in \mathbb{R}^n \tag{4}$$

has no solution $x \geq 0$. Let \mathcal{F} be the family of all maximal subsystems of (4) which can be satisfied by a non-negative solution x . Then, given a partial list $\mathcal{X} \subseteq \mathcal{F}$, it is an NP-complete problem to determine if the list is incomplete, i.e., if $\mathcal{X} \neq \mathcal{F}$, even if b is a unit vector, and entries in A are either, $-1, 1$, or 0 . \square

We conclude with the observation that the problem of finding, for an infeasible system

$$A'x \geq b', \quad A''x \geq b'', \tag{5}$$

all maximal feasible subsystems extending the feasible subsystem $A''x \geq b''$, naturally includes both the problems of generating anti-simplices and simplices. Clearly, the former problem can be written in the form (5) by considering (2) and all maximal extensions of an empty subsystem. For the latter problem, note that the vertices of the bounded polyhedron $\{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$, where $b \neq 0$, are in one-to-one correspondence with the maximal feasible extensions of the subsystem $Ax = b, x \geq 0$ in the infeasible system $Ax = b, x \geq 0, x \leq 0$. Although the general problem of generating maximal feasible extensions is NP-hard as stated above, the special cases of generating simplices and anti-simplices remain open.

2 Proof of Theorem 1

In this section we prove Theorem 1 by a reduction from satisfiability, a well-known NP-complete problem (see [16]).

Let us consider n propositional Boolean variables $X_j, j = 1, \dots, n$, denote by $\overline{X} = 1 - X$ the negation of X , call variables and their negations *literals*, and elementary disjunctions of literals *clauses*. Let us next consider an arbitrary conjunctive normal form (CNF) $\phi = C_1 \wedge C_2 \dots \wedge C_m$, i.e., where $C_i, i = 1, \dots, m$ are clauses. A truth assignment to the variables is called *satisfying* for the CNF ϕ , if ϕ evaluates to true, i.e., if at least one literal evaluates to true in each of the clauses of ϕ .

In what follows, we shall associate to ϕ a weighted directed graph (G, w) and a set \mathcal{X} of negative circuits of G such that (G, w) has a negative circuit not belonging to \mathcal{X} if and only if ϕ has a satisfying assignment. Because (G, w) and \mathcal{X} are constructed from ϕ in $O(mn)$ time, and the weight function w uses only two different values (1 and -1), Theorem 1 follows readily from this construction, since the decision problem "Is there a negative circuit in (G, w) which does not belong to \mathcal{X} ?" is in NP. To complete the proof of Theorem 1, we provide

below a construction with these properties, such that there is a one-to-one correspondence between satisfying assignments to ϕ and negative circuits of (G, w) which do not belong to \mathcal{X} .

To describe our construction, let us denote for $j = 1, \dots, n$ respectively by o_j and \bar{o}_j the number of occurrences of literal X_j and its negation \bar{X}_j , denote by x_j^k the k th occurrence of X_j , $k = 1, \dots, o_j$, and by \bar{x}_j^k the k th occurrence of \bar{X}_j , $k = 1, \dots, \bar{o}_j$, and let L denote the set of all literal occurrences, i.e.,

$$|L| = \sum_{i=1}^m |C_i| = \sum_{j=1}^n (o_j + \bar{o}_j).$$

Since monotone variables, i.e., ones for which $o_j = 0$ or $\bar{o}_j = 0$, can be easily eliminated from a satisfiability problem, we can assume without any loss of generality that $o_j > 0$ and $\bar{o}_j > 0$ hold for all variables $j = 1, \dots, n$.

For instance, if $n = 3$ and

$$\phi = (X_1 \vee X_2 \vee \bar{X}_3) \wedge (X_1 \vee \bar{X}_2 \vee X_3) \wedge (\bar{X}_1 \vee X_2 \vee \bar{X}_3), \quad (6)$$

then we have $o_1 = 2$, $\bar{o}_1 = 1$, $o_2 = 2$, $\bar{o}_2 = 1$, $o_3 = 1$, $\bar{o}_3 = 2$,

$$L = \{x_1^1, x_2^1, \bar{x}_3^1, x_1^2, \bar{x}_2^1, x_3^1, \bar{x}_1^1, x_2^2, \bar{x}_3^2\}.$$

We define the vertex set of the graph $G = (V, E)$ associated to ϕ as

$$V = U \cup Q \cup \bigcup_{j=1}^n (Y_j \cup Z_j),$$

where U , Q , and Y_j and Z_j for $j = 1, \dots, n$ are pairwise disjoint, defined as

$$\begin{aligned} U &= \{u_k \mid k = 0, 1, \dots, m+n\}, \\ Q &= \{a(\ell), b(\ell) \mid \ell \in L\}, \\ Y_j &= \{y_{jk} \mid k = 1, \dots, o_j - 1\} \text{ for } j = 1, \dots, n, \text{ and} \\ Z_j &= \{z_{jk} \mid k = 1, \dots, \bar{o}_j - 1\} \text{ for } j = 1, \dots, n. \end{aligned}$$

The graph itself has a ring structure, the skeleton of which is the set U . For every variable X_j of ϕ we have two parallel directed paths from u_{j-1} to u_j . The first path corresponding to X_j contains vertices Y_j (and some other vertices), while the second path, corresponding to \bar{X}_j passes through vertices of Z_j ($j = 1, \dots, n$). For convenience, we also introduce the notations

$$y_{j0} = z_{j0} = u_{j-1} \quad \text{and} \quad y_{j,o_j} = z_{j,\bar{o}_j} = u_j \quad (7)$$

for $j = 1, \dots, n$. To every clause C_i of ϕ we associate $|C_i|$ parallel directed paths from u_{n+i-1} to u_{n+i} , one for each of the literals in C_i ($i = 1, \dots, m$). Finally vertices $a(\ell)$ and $b(\ell)$ correspond exclusively to literal occurrence $\ell \in L$.

Let us consider next the weighted graph $H(a, b, p, q, r, s)$ (see Figure 1) on six nodes a, b, p, q, r and s , having six arcs, the weights of which are as follows:

$$\begin{aligned} w(a, b) &= w(b, a) = -2 \quad \text{and} \\ w(p, a) &= w(b, q) = w(r, b) = w(a, s) = 1. \end{aligned} \tag{8}$$

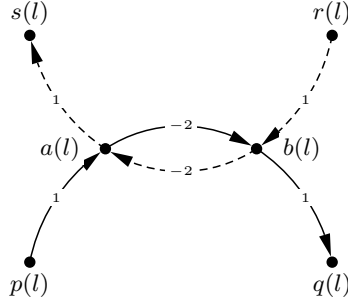


Figure 1: The directed graph $H(a, b, p, q, r, s)$ associated with literal occurrences.

To every literal occurrence $\ell \in L$ we associate a disjoint copy of $H(a, b, p, q, r, s)$, and denote by $a(\ell), b(\ell)$, etc., its nodes, and by E_ℓ its arc set. Note that each of these small subgraphs can be decomposed into two directed paths, each consisting of three arcs, $E_\ell = E_\ell^v \cup E_\ell^c$, where

$$\begin{aligned} E_\ell^v &= \{ (p(\ell), a(\ell)), (a(\ell), b(\ell)), (b(\ell), q(\ell)) \}, \text{ and} \\ E_\ell^c &= \{ (r(\ell), b(\ell)), (b(\ell), a(\ell)), (a(\ell), s(\ell)) \}. \end{aligned}$$

Finally we set

$$E = E_0 \cup \bigcup_{\ell \in L} E_\ell$$

where $E_0 = \{(u_{m+n}, u_0)\}$ with weight $w(u_{m+n}, u_0) = -1$.

In each of the subgraphs corresponding to the literal occurrences $\ell \in L$, we have the nodes $a(\ell)$ and $b(\ell)$ already introduced in $Q \subseteq V$, while the nodes $p(\ell), q(\ell), r(\ell)$ and $s(\ell)$ for $\ell \in L$ are corresponding to some other vertices of G , according to the following definitions:

$$\begin{aligned} p(\ell) &= y_{j, k-1} \quad \text{and} \quad q(\ell) = y_{jk} && \text{if } \ell = x_j^k, \\ p(\ell) &= z_{j, k-1} \quad \text{and} \quad q(\ell) = z_{jk} && \text{if } \ell = \bar{x}_j^k, \text{ and} \\ r(\ell) &= u_{n+i-1} \quad \text{and} \quad s(\ell) = u_{n+i} && \text{if } \ell \in C_i. \end{aligned}$$

In other words, for every literal occurrence ℓ of clause C_i the set E_ℓ^c forms a 3-arc directed path from u_{n+i-1} to u_{n+i} . Furthermore by (7) and by the above definitions, the sets E_ℓ^v for $\ell = x_j^1, x_j^2, \dots, x_j^{o_j}$ form a directed path from u_{j-1}

to u_j through the vertices of Y_j , consisting of $3o_j$ arcs, for every variable X_j . Similarly, the sets E_ℓ^v for $\ell = \bar{x}_j^1, \bar{x}_j^2, \dots, \bar{x}_j^{o_j}$ form another directed path from u_{j-1} to u_j through the vertices of Z_j , consisting of $3\bar{o}_j$ arcs.

In summary, $G = (V, E)$ consists of $|V| = 3|L| + m - n + 1$ vertices and $|E| = 6|L| + 1$ arcs, and the weight function w takes only values in $\{-2, -1, 1\}$. Note that we can split arcs of weight -2 to obtain a graph whose all arcs have weight ± 1 .

Returning to the example CNF ϕ given in (6), the corresponding graph $G = (V, E)$ is shown in Figure 2. To make the drawing of such a graph visually more clear, for every literal occurrence ℓ nodes $a(\ell)$ and $b(\ell)$ of G are represented by two separate points of the picture each, labeled as $a(\ell)$ and $a'(\ell)$, and as $b(\ell)$ and $b'(\ell)$, respectively. Similarly, node u_n is represented by two points in the figure, labeled by u_n and u'_n . Arcs in the sets E_ℓ^c for $\ell \in L$ are drawn as dashed lines, while those belonging to E_ℓ^v for $\ell \in L$ are drawn as solid lines.

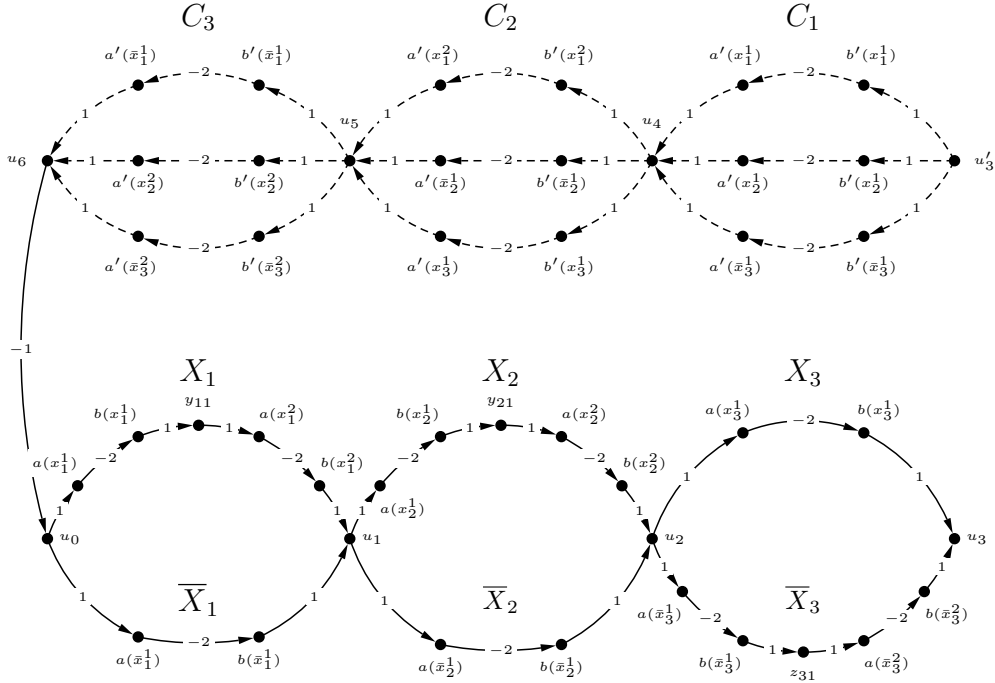


Figure 2: G is obtained by identifying vertices $a(l)$, $a'(l)$, and $b(l)$, $b'(l)$, for each literal occurrence l , and u_3 , u'_3 in the graph above. The lower part of the graph corresponds to the literals and the upper part corresponds to the clauses.

Let us observe first that the arcs $(a(\ell), b(\ell))$ and $(b(\ell), a(\ell))$ form a circuit of total weight -4 for every literal occurrence $\ell \in L$. Let us denote by \mathcal{X} the set

of these circuits, i.e., $|\mathcal{X}| = |L|$, and let us denote by \mathcal{F} the set of all directed negative circuits of G .

We claim that from every satisfying assignment X of ϕ we can construct a directed negative circuit $D^X \in \mathcal{F} \setminus \mathcal{X}$, and conversely, from every directed negative circuit $D \in \mathcal{F} \setminus \mathcal{X}$ we can construct a satisfying assignment X^D of ϕ . As we noted at the beginning of this section, this claim implies Theorem 1.

To see this claim, let us first consider a satisfying assignment $X = (X_1, \dots, X_n) \in \{0, 1\}^n$ of ϕ . Since X satisfies ϕ , we have a literal occurrence ℓ_i in every clause C_i , $i = 1, \dots, m$, such that ℓ_i evaluates to true at X (i.e., $\ell_i(X) = 1$). Let us also denote by W the set of all those literal occurrences which evaluate to false at X , i.e., $W = \{\ell \in L \mid \ell(X) = 0\}$. Clearly, $\ell_i \notin W$ for $i = 1, \dots, m$ by the above definitions. Then, the set of arcs

$$D^X = \left(\bigcup_{i=1}^m E_{\ell_i}^c \right) \cup \left(\bigcup_{\ell \in W} E_{\ell}^v \right) \cup \{(u_{m+n}, u_0)\}.$$

forms a circuit in G not belonging to \mathcal{X} . Since we have $w(E_{\ell}^c) = w(E_{\ell}^v) = 0$ for all literal occurrences $\ell \in L$, it follows by the above definitions that $w(D^X) = w(u_{m+n}, u_0) = -1$, i.e., $D^X \in \mathcal{F} \setminus \mathcal{X}$ as claimed.

We again return to the CNF ϕ given in (6). Let us consider the satisfying assignment $X = (1, 0, 0)$ of ϕ . We choose literal occurrences $\bar{x}_3^1 \in C_1$, $\bar{x}_1^2 \in C_2$ and $\bar{x}_3^2 \in C_3$ that evaluate to true at X . Figure 3 depicts the negative circuit $D^X = E_{\bar{x}_3^1}^c \cup E_{\bar{x}_1^2}^c \cup E_{\bar{x}_3^2}^c \cup E_{\bar{x}_1^1}^v \cup E_{\bar{x}_2^1}^v \cup E_{\bar{x}_2^2}^v \cup E_{\bar{x}_3^3}^v \cup (u_6, u_0)$.

Before proving the reverse direction of our main claim, let us first observe some simple properties of our construction. To simplify notation, recall that $E_{\ell} = E_{\ell}^c \cup E_{\ell}^v$ for $\ell \in L$, and that the 6-vertex subgraphs induced by the arc set E_{ℓ} have the same structure and weights, as in Figure 1, for all $\ell \in L$. The following property of these subgraphs will be instrumental in our proof.

Lemma 1 *Given a circuit $D \subseteq E$ of G , not belonging to \mathcal{X} , and given a literal occurrence $\ell \in L$, we have*

$$w(D \cap E_{\ell}) \in \{0, 2, 4\}.$$

Moreover, $w(D \cap E_{\ell}) = 0$ only if the set $D \cap E_{\ell}$ is one of the following three subsets of E_{ℓ} : E_{ℓ}^c , E_{ℓ}^v , or \emptyset .

Proof: Since D is a circuit not belonging to \mathcal{X} , D cannot contain both arcs $(a(\ell), b(\ell))$ and $(b(\ell), a(\ell))$. Thus, denoting by $A_{\ell} = \{(p(\ell), a(\ell)), (a(\ell), s(\ell))\}$ and $B_{\ell} = \{(r(\ell), b(\ell)), (b(\ell), q(\ell))\}$ we have that $D \cap E_{\ell}$ is one of the following six sets: \emptyset , A_{ℓ} , B_{ℓ} , $A_{\ell} \cup B_{\ell}$, E_{ℓ}^c , and E_{ℓ}^v . Since we have $w(\emptyset) = w(E_{\ell}^c) = w(E_{\ell}^v) = 0$, $w(A_{\ell}) = w(B_{\ell}) = 2$ and hence $w(A_{\ell} \cup B_{\ell}) = 4$, the statement follows. \square

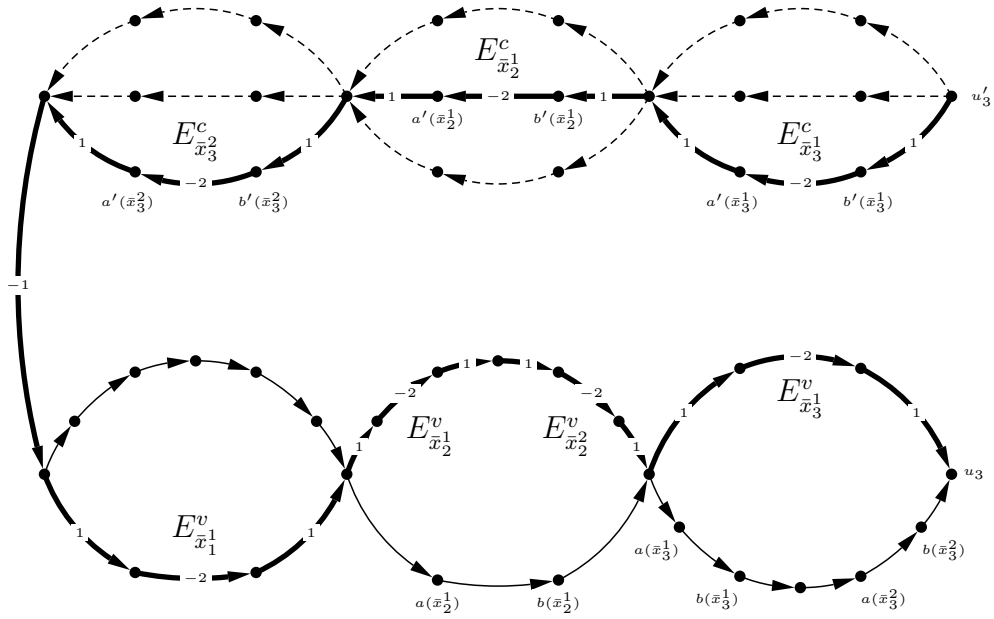


Figure 3: Thick lines are edges of the negative circuit D^X corresponding to the satisfying assignment $X = (1, 0, 0)$. The vertices u_3, u'_3 are identified. Since D^X contains arcs $(b'(\bar{x}_3^1)a'(\bar{x}_3^1))$, $(b'(\bar{x}_2^1)a'(\bar{x}_2^1))$ and $(b'(\bar{x}_3^2)a'(\bar{x}_3^2))$ but it does not contain arcs $(a(\bar{x}_3^1)b(\bar{x}_3^1))$, $(a(\bar{x}_2^1)b(\bar{x}_2^1))$ and $(a(\bar{x}_3^2)b(\bar{x}_3^2))$, no circuit of \mathcal{X} is contained in D^X .

Returning to the reverse direction of our main claim, let us consider a negative circuit $D \in \mathcal{F} \setminus \mathcal{X}$ of G . Since

$$w(D) = \sum_{\ell \in L} w(D \cap E_\ell) + w(D \cap \{(u_{m+n}, u_0)\})$$

we must have by Lemma 1 that $(u_{m+n}, u_0) \in D$ and

$$w(D \cap E_\ell) = 0 \quad \text{for all } \ell \in L. \quad (9)$$

We show first that D passes through all vertices in U , includes exactly one of the two parallel paths between u_{j-1} and u_j for $j = 1, \dots, n$, and exactly one of the parallel paths between u_{n+i-1} and u_{n+i} for all $i = 1, \dots, m$.

As we observed above, we have u_0 as a vertex of D . Thus D must contain an arc leaving u_0 , say it contains $(u_0, a_{x_1^1})$. Then, by (9) and by Lemma 1 we must have $E_{x_1^1}^v \subseteq D$, i.e., D must pass through vertex y_{11} . Since only $(y_{11}, a(x_1^2))$ is leaving y_{11} , by repeating the above argument we can conclude that we must also have $E_{x_1^2}^v \subseteq D$, etc., finally arriving at $E_{x_{o_1}^v} \subseteq D$, i.e., that D includes u_1 as a vertex. Repeating the same argument, we can prove by induction that for all indices $j = 1, \dots, n$, if $E_{x_j^1}^v \subseteq D$, then we must have $E_{x_j^k}^v \subseteq D$ for all $k = 1, \dots, o_j$, and that if $E_{x_j^1}^v \subseteq D$, then we must also have $E_{x_j^k}^v \subseteq D$ for all $k = 1, \dots, \bar{o}_j$. Let us then define a truth assignment X^D by

$$X_j^D = \begin{cases} 1 & \text{if } E_{x_j^1}^v \subseteq D, \\ 0 & \text{if } E_{x_j^1}^v \not\subseteq D. \end{cases}$$

Furthermore, repeating a similar argument for vertices $u_n, u_{n+1}, \dots, u_{n+m-1}, u_{n+m}$ we can also conclude that D must contain the set $E_{\ell_i}^c$ for exactly one of the literals $\ell_i \in C_i$, for each clause C_i of ϕ . Since D is a circuit in which no vertex $a(\ell)$ or $b(\ell)$ is repeated, we must have that $\ell_i(X^D) = 1$ for all $i = 1, \dots, m$, i.e., that X^D is indeed a satisfying assignment of ϕ .

These observations prove the reverse direction of our main claim, and hence conclude the proof of Theorem 1. \square

3 Proof of Theorem 2

We can repeat essentially the same proof as for the directed case, with the exception that we associate with every literal occurrence $\ell \in L$ a different subgraph denoted by E_ℓ : Let us associate now with $\ell \in L$ six nodes, $a = a(\ell), b = b(\ell), c = c(\ell), d = d(\ell), e = e(\ell)$, and $f = f(\ell)$, and the following 10 edges

$$E_\ell = \{(a, b), (b, c), (c, d), (d, e), (e, f), (a, f), (a, p), (b, q), (d, r), (e, s)\},$$

where nodes $p = p(\ell)$, $q = q(\ell)$, $r = r(\ell)$ and $s = s(\ell)$ are identified with the other nodes of G , in the same way as in the previous proof. To simplify notation, we shall omit the reference to ℓ , whenever it is clear from the context which literal occurrence we talk about. The weights of the edges of E_ℓ are defined as

$$w(a, p) = w(b, q) = w(d, r) = w(e, s) = \frac{5}{2}, \text{ and}$$

$$w(a, b) = w(b, c) = w(c, d) = w(d, e) = w(e, f) = w(a, f) = -1.$$

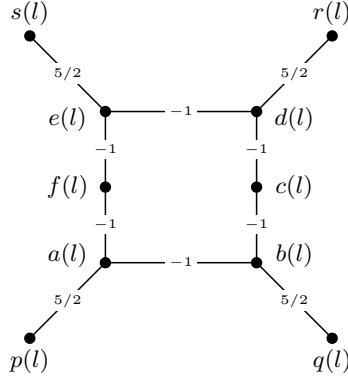


Figure 4: The undirected graph associated with literal occurrences.

Let us note that in each of these subgraphs there is a negative circuit (see Figure 4), formed by the six edges $D_\ell = \{(a, b), (b, c), (c, d), (d, e), (e, f), (a, f)\}$. Let us denote by $\mathcal{X} = \{D_\ell \mid \ell \in L\}$ the collection of these negative circuits, and let \mathcal{F} denote the family of all negative circuits in G .

By an analogous proof as in the previous section, we can show that there exists a negative circuit belonging to $\mathcal{F} \setminus \mathcal{X}$ if and only if ϕ has a satisfying assignment. The key observation in this case, the analogue of Lemma 1, is the following claim, which can easily be verified e.g., by looking at Figure 4.

Lemma 2 *For a circuit D of G not belonging to \mathcal{X} and literal occurrence $\ell \in L$ we have*

$$w(D \cap E_\ell) \in \{0, 1, 2, 4\}$$

and it is equal to 0 only if $D \cap E_\ell$ is one of the following three sets: \emptyset ,

$$E_\ell^v = \{(b, c), (c, d), (d, e), (e, f), (a, f), (a, p), (b, q)\}, \text{ or}$$

$$E_\ell^c = \{(a, b), (b, c), (c, d), (e, f), (a, f), (d, r), (e, s)\}.$$

□

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