

# On Generating All Minimal Integer Solutions for a Monotone System of Linear Inequalities<sup>\*</sup>

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**Abstract.** We consider the problem of enumerating all minimal integer solutions of a monotone system of linear inequalities. We first show that for any monotone system of  $r$  linear inequalities in  $n$  variables, the number of maximal infeasible integer vectors is at most  $rn$  times the number of minimal integer solutions to the system. This bound is accurate up to a  $\text{polylog}(r)$  factor and leads to a polynomial-time reduction of the enumeration problem to a natural generalization of the well-known dualization problem for hypergraphs, in which dual pairs of hypergraphs are replaced by dual collections of integer vectors in a box. We provide a quasi-polynomial algorithm for the latter dualization problem. These results imply, in particular, that the problem of incrementally generating minimal integer solutions of a monotone system of linear inequalities can be done in quasi-polynomial time.

**Keywords:** Integer programming, complexity of incremental algorithms, dualization, quasi-polynomial time, monotone discrete binary functions, monotone inequalities, regular discrete functions.

## 1 Introduction

Consider a system of  $r$  linear inequalities in  $n$  integer variables

$$Ax \geq b, \quad x \in \mathcal{C} = \{x \in \mathbb{Z}^n \mid 0 \leq x \leq c\}, \quad (1)$$

where  $A$  is a rational  $r \times n$ -matrix,  $b$  is a rational  $r$ -vector, and  $c$  is a non-negative integral  $n$ -vector some or all of whose components may be infinite. We assume that (1) is a monotone system of inequalities: if  $x \in \mathcal{C}$  satisfies (1) then any vector  $y \in \mathcal{C}$  such that  $y \geq x$  is also feasible. For instance, (1) is monotone if the

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matrix  $A$  is non-negative. Let us denote by  $\mathcal{F} = \mathcal{F}_{A,b,c}$  the set of all minimal feasible integral vectors for (1), i.e.  $y \in \mathcal{F}$  if there is no solution  $x$  of (1) such that  $x \leq y$ ,  $x \neq y$ . In particular, we have  $\{x \in \mathcal{C} \mid Ax \geq b\} = \bigcup_{y \in \mathcal{F}} \{x \in \mathcal{C} \mid x \geq y\}$ . In this paper, we are concerned with the problem of incrementally generating  $\mathcal{F}$ :

*GEN*( $\mathcal{F}_{A,b,c}, \mathcal{X}$ ): Given a monotone system (1) and a set  $\mathcal{X} \subseteq \mathcal{F}_{A,b,c}$  of minimal feasible vectors for (1), either find a new minimal integral vector satisfying (1), or show that  $\mathcal{X} = \mathcal{F}_{A,b,c}$ .

The entire set  $\mathcal{F} = \mathcal{F}_{A,b,c}$  can be constructed by initializing  $\mathcal{X} = \emptyset$  and iteratively solving the above problem  $|\mathcal{F}| + 1$  times.

If  $A$  is a binary matrix, and  $b, c$  are vectors of all ones, then  $\mathcal{F}$  is the set of (characteristic vectors of) all minimal transversals to the hypergraph defined by the rows of  $A$ . In this case, problem *GEN*( $\mathcal{F}_{A,b,c}, \mathcal{X}$ ) turns into the well-known *hypergraph dualization problem*: incrementally enumerate all the minimal transversals (equivalently, all the maximal independent sets) for a given hypergraph (see, e.g., [3,10]). Some applications of the hypergraph dualization problem are discussed in [1,7,9]. The case where  $A$  is binary,  $c$  is the vector of all ones and  $b$  is arbitrary, is equivalent with the generation of so-called *multiple transversals* [5]. If  $A$  is integral and  $c = +\infty$ , the generation of  $\mathcal{F}$  can also be regarded as the computation of the Hilbert basis for the ideal  $\{x \in \mathbb{Z}^n \mid Ax \geq b, x \geq 0\}$ . One more application of problem *GEN*( $\mathcal{F}_{A,b,c}, \mathcal{X}$ ) is related to stochastic programming, more precisely to the generation of minimal *p-efficient* points for a given probability distribution of a discrete random variable  $\xi \in \mathbb{Z}^n$ . An integer vector  $y \in \mathbb{Z}^n$  is called *p-efficient*, if  $Prob(\xi \leq y) \geq p$ . It is known that for every probability distribution and every  $p > 0$  there are finitely many minimal *p-efficient* points, furthermore that for *r-concave* probability distributions these points are exactly the minimal integral points of a corresponding convex monotone system (see, e.g., [14]).

Let  $J^* = \{j \mid c_j = \infty\}$  and  $J_* = \{1, \dots, n\} \setminus J^*$  be, respectively, the sets of unbounded and bounded integer variables in (1). Consider an arbitrary vector  $x = (x_1, \dots, x_n) \in \mathcal{F}_{A,b,c}$  such that  $x_j > 0$  for some  $j \in J^*$ . Then it is easy to see that

$$x_j \leq \max_{i: a_{ij} > 0} \left\lceil \frac{b_i - \sum_{k \in J_*} \min\{0, a_{ik}\} c_k}{a_{ij}} \right\rceil < +\infty. \tag{2}$$

Since the bounds of (2) are easy to compute, and since appending these bounds to (1) does not change the set  $\mathcal{F}_{A,b,c}$ , we shall assume in the sequel that all components of the non-negative vector  $c$  are finite, even though this may not be the case for the original system. This assumption does not entail any loss of generality and allows us to consider  $\mathcal{F}_{A,b,c}$  as a system of integral vectors in a finite box. We shall also assume that the input monotone system (1) is feasible, i.e.,  $\mathcal{F}_{A,b,c} \neq \emptyset$ . For a finite and non-negative  $c$  this is equivalent to  $Ac \geq b$ .

Let  $\mathcal{A}$  be a collection of integral vectors in  $\mathcal{C}$  and let  $\mathcal{A}^+ = \{x \in \mathcal{C} \mid x \geq a \text{ for some } a \in \mathcal{A}\}$  and  $\mathcal{A}^- = \{x \in \mathcal{C} \mid x \leq a \text{ for some } a \in \mathcal{A}\}$  denote the ideal and filter generated by  $\mathcal{A}$ . Any element in  $\mathcal{C} \setminus \mathcal{A}^+$  is called *independent of  $\mathcal{A}$* . Let  $\mathcal{I}(\mathcal{A})$  be the set of all maximal independent elements for  $\mathcal{A}$ , then for any finite box  $\mathcal{C}$  we have the decomposition:

$$\mathcal{A}^+ \cap \mathcal{I}(\mathcal{A})^- = \emptyset, \quad \mathcal{A}^+ \cup \mathcal{I}(\mathcal{A})^- = \mathcal{C}. \tag{3}$$

In particular, if  $\mathcal{A}$  is the set  $\mathcal{F} = \mathcal{F}_{A,b,c}$  of all minimal feasible integral vectors for (1), then the ideal  $\mathcal{F}^+$  is the solution set of (1), while the filter  $\mathcal{C} \setminus \mathcal{F}^+$  is generated by the set  $\mathcal{I}(\mathcal{F})$  of all maximal infeasible integral vectors for (1):

$$\{x \in \mathcal{C} \mid Ax \not\geq b\} = \bigcup_{y \in \mathcal{I}(\mathcal{F})} \{y\}^-.$$

It is known that the problem of incrementally generating all maximal infeasible vectors for (1) is NP-hard even if  $c$  is the vector of all ones and the matrix  $A$  is binary:

**Proposition 1 (c.f. [12]).** *Given a binary matrix  $A$  and a set  $\mathcal{X} \subseteq \mathcal{I}(\mathcal{F}_{A,b,c})$  of maximal infeasible Boolean vectors for  $Ax \geq b$ ,  $x \in \{0, 1\}^n$ , it is NP-complete to decide if the set  $\mathcal{X}$  can be extended, that is if  $\mathcal{I}(\mathcal{F}_{A,b,c}) \setminus \mathcal{X} \neq \emptyset$ .*

In contrast to that, we show in this paper that the problem of incrementally generating all minimal feasible vectors for (1) is unlikely to be NP-hard.

**Theorem 1.** *Problem  $GEN(\mathcal{F}_{A,b,c}, \mathcal{X})$  can be solved in quasi-polynomial time  $poly(|input|) + t^{o(\log t)}$ , where  $t = \max\{n, r, |\mathcal{X}|\}$ .*

It was conjectured in [11] that problem  $GEN(\mathcal{F}_{A,b,c}, \mathcal{X})$  cannot be solved in polynomial time unless  $P=NP$ .

To prove Theorem 1, we first bound the number of maximal infeasible vectors for (1) in terms of the dimension of the system and the number of minimal feasible vectors.

**Theorem 2.** *Suppose that the monotone system (1) is feasible, i.e.,  $Ac \geq b$ . Then for any non-empty set  $\mathcal{X} \subseteq \mathcal{F}_{A,b,c}$  we have*

$$|\mathcal{I}(\mathcal{X}) \cap \mathcal{I}(\mathcal{F}_{A,b,c})| \leq r \sum_{x \in \mathcal{X}} p(x), \tag{4}$$

where  $p(x)$  is the number of positive components of  $x$ . In particular,

$$|\mathcal{I}(\mathcal{X}) \cap \mathcal{I}(\mathcal{F}_{A,b,c})| \leq rn|\mathcal{X}|,$$

which for  $\mathcal{X} = \mathcal{F}_{A,b,c}$  leads to the inequality  $|\mathcal{I}(\mathcal{F}_{A,b,c})| \leq rn|\mathcal{F}_{A,b,c}|$ .

It should be mentioned that the bounds of Theorem 2 are sharp for  $r = 1$ , e.g., for the inequality  $x_1 + \dots + x_n \geq n$ . For large  $r$ , these bounds are accurate up to a factor poly-logarithmic in  $r$ . To see this, let  $n = 2k$  and consider the monotone system of  $r = 2^k$  inequalities of the form

$$x_{i_1} + x_{i_2} + \dots + x_{i_k} \geq 1, \quad i_1 \in \{1, 2\}, \quad i_2 \in \{3, 4\}, \dots, \quad i_k \in \{2k - 1, 2k\},$$

where  $x = (x_1, \dots, x_n) \in \mathcal{C} = \{x \in \mathbb{Z}^n \mid 0 \leq x \leq c\}$ . For any positive integral vector  $c$ , this system has  $2^k$  maximal infeasible integral vectors and only  $k$  minimal feasible integral vectors, i.e.,

$$|\mathcal{I}(\mathcal{F}_{A,b,c})| = \frac{rn}{2(\log r)^2} |\mathcal{F}_{A,b,c}|.$$

Needless to say that in general,  $|\mathcal{F}_{A,b,c}|$  cannot be bounded by a polynomial in  $r$ ,  $n$ , and  $|\mathcal{I}(\mathcal{F}_{A,b,c})|$ . For instance, for  $n = 2k$  the system of  $k$  inequalities  $x_1 + x_2 \geq 1, \quad x_3 + x_4 \geq 1, \dots, \quad x_{2k-1} + x_{2k} \geq 1$  has  $2^k$  minimal feasible binary vectors and only  $k$  maximal infeasible binary vectors.

Let us add finally that if the number of inequalities in (1) is fixed, then  $|\mathcal{F}_{A,b,c}|$  can also be polynomially bounded by  $|\mathcal{I}(\mathcal{F}_{A,b,c})|$ , and accordingly, the set of all maximal infeasible integer vectors for (1) can be generated in quasi-polynomial time. In other words, Proposition 1 cannot hold for  $r = \text{const}$  unless any problem in  $NP$  can be solved in quasi-polynomial time. Furthermore, for systems with fixed number of non-zero coefficients per inequality and bounded box size, problem  $GEN(\mathcal{F}_{A,b,c}, \mathcal{X})$  can be efficiently solved in parallel (see [4]).

We prove Theorem 2 in Section 2, and then use this theorem in the next section to reduce problem  $GEN(\mathcal{F}_{A,b,c}, \mathcal{X})$  to a natural generalization of the hypergraph dualization problem. Our generalized dualization problem replaces hypergraphs by collections of integer vectors in a box.

**Theorem 3.**  *$GEN(\mathcal{F}_{A,b,c}, \mathcal{X})$  is polynomial-time reducible to the following problem:*

*$DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$ : Given an integral box  $\mathcal{C}$ , a family of vectors  $\mathcal{A} \subseteq \mathcal{C}$ , and a collection of maximal independent elements  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{A})$ , either find a new maximal independent element  $x \in \mathcal{I}(\mathcal{A}) \setminus \mathcal{B}$ , or prove that  $\mathcal{B} = \mathcal{I}(\mathcal{A})$ .*

Note that for  $\mathcal{C} = \{0, 1\}^n$ , problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$  turns into the hypergraph dualization problem. Other applications of the dualization problem on boxes can be found in [2,6,13]. In Section 4 we extend the hypergraph dualization algorithm of [8] to problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$  and show that the latter problem can be solved in quasi-polynomial time:

**Theorem 4.** *Given two sets  $\mathcal{A}$ , and  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{A})$  in an integral box  $\mathcal{C} = \{x \in \mathbb{Z}^n \mid 0 \leq x \leq c\}$ , problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$  can be solved in  $\text{poly}(n, m) + m^{o(\log m)}$  time, where  $m = |\mathcal{A}| + |\mathcal{B}|$ .*

Clearly, Theorem 1 follows from Theorems 3 and 4. The special cases of Theorems 2 and 3 for Boolean systems  $x \in \{0, 1\}^n$  can be found in [5].

The remainder of the paper consists of the proofs of Theorems 2, 3 and 4 in Sections 2, 3, 4 respectively.

## 2 Bounding the Number of Maximal Infeasible Vectors

In this section we prove Theorem 2. We first need some notations and definitions.

Let  $\mathcal{C} = \{x \in \mathbb{Z}^n \mid 0 \leq x \leq c\}$  be a box and let  $f : \mathcal{C} \rightarrow \{0, 1\}$  be a discrete binary function. The function  $f$  is called *monotone* if  $f(x) \geq f(y)$  whenever  $x \geq y$  and  $x, y \in \mathcal{C}$ . We denote by  $T(f)$  and  $F(f)$  the sets of all true and all false vectors of  $f$ , i.e.,

$$T(f) = \{x \in \mathcal{C} \mid f(x) = 1\} = (\min[f])^+, \quad F(f) = \{x \in \mathcal{C} \mid f(x) = 0\} = (\max[f])^-,$$

where  $\min[f]$  and  $\max[f]$  are the sets of all minimal true and all maximal false vectors of  $f$ , respectively.

Let  $\sigma \in \mathbb{S}_n$  be a permutation of the coordinates and let  $x, y$  be two  $n$ -vectors. We say that  $y$  is a *left-shift* of  $x$  and write  $y \succeq_\sigma x$  if the inequalities

$$\sum_{j=1}^k y_{\sigma_j} \geq \sum_{j=1}^k x_{\sigma_j}$$

hold for all  $k = 1, \dots, n$ . A discrete binary function  $f : \mathcal{C} \rightarrow \{0, 1\}$  is called *2-monotonic with respect to  $\sigma$*  if  $f(y) \geq f(x)$  whenever  $y \succeq_\sigma x$  and  $x, y \in \mathcal{C}$ . Clearly,  $y \geq x$  implies  $y \succeq_\sigma x$  for any  $\sigma \in \mathbb{S}_n$ , so that any 2-monotonic function is monotone.

The function  $f$  will be called *regular* if it is 2-monotonic with respect to the identity permutation  $\sigma = (1, 2, \dots, n)$ . Any 2-monotonic function can be transformed into a regular one by appropriately re-indexing its variables. To simplify notations, we shall state Lemma 1 below for regular functions, i.e., we fix  $\sigma = (1, 2, \dots, n)$  in this lemma.

For a given subset  $\mathcal{A} \subseteq \mathcal{C}$  let us denote by  $\mathcal{A}^*$  all the vectors which are left-shifts of some vectors of  $\mathcal{A}$ , i.e.,  $\mathcal{A}^* = \{y \in \mathcal{C} \mid y \succeq x \text{ for some } x \in \mathcal{A}\}$ . Clearly,  $T(f) = (\min[f])^*$  for a regular function  $f$  (in fact, the subfamily of *right-most* vectors of  $\min[f]$  would be enough to use here.)

Given monotone discrete functions  $f$  and  $g$ , we call  $g$  a *regular majorant* of  $f$ , if  $g(x) \geq f(x)$  for all  $x \in \mathcal{C}$ , and  $g$  is regular. Clearly,  $T(g) \supseteq (\min[f])^*$  must hold in this case, and the discrete function  $h$  defined by  $T(h) = (\min[f])^*$  is the unique minimal regular majorant of  $f$ .

For a vector  $x \in \mathcal{C}$ , and for an index  $1 \leq k \leq n$ , let the vectors  $x^{(k)}$  and  $x^{[k]}$  be defined by

$$x_j^{(k)} = \begin{cases} x_j & \text{for } j \leq k, \\ 0 & \text{otherwise,} \end{cases}$$

and

$$x_j^{[k]} = \begin{cases} x_j & \text{for } j \geq k, \\ 0 & \text{otherwise.} \end{cases}$$

Let us denote by  $e$  the  $n$ -vector of all 1's, let  $e_j$  denote the  $j^{\text{th}}$  unit vector,  $j = 1, \dots, n$ , and let  $p(x)$  denote the number of positive components of the vector  $x \in \mathcal{C}$ .

**Lemma 1.** *Given a monotone discrete binary function  $f : \mathcal{C} \rightarrow \{0, 1\}$  such that  $f \not\equiv 0$ , and a regular majorant  $g \geq f$ , we have the inequality*

$$|F(g) \cap \max[f]| \leq \sum_{x \in \min[f]} p(x). \tag{5}$$

*Proof.* Let us denote by  $h$  the unique minimal regular majorant of  $f$ . Then we have  $F(g) \cap \max[f] \subseteq F(h) \cap \max[f]$ , and hence it is enough to show the statement for  $g = h$ , i.e. when  $T(g) = (\min[f])^*$ .

For a vector  $y \in \mathcal{C} \setminus \{c\}$  let us denote by  $l = l_y$  the index of the last component which is less than  $c_l$ , i.e.,  $l = \max\{j \mid y_j < c_j\} \in \{1, \dots, n\}$ . We claim that for every  $y \in F(h) \cap \max[f]$  there exists an  $x \in \min[f]$  such that

$$y = x^{(l-1)} + (x_l - 1)e_l + c^{[l+1]}, \tag{6}$$

where  $l = l_y$ . To see this claim, first observe that  $y \neq c$  because  $y \in F(f)$  and  $f \not\equiv 0$ . Second, for any  $j$  with  $y_j < c_j$  we know that  $y + e_j \in T(f)$ , by the definition of a maximal false point. Hence there exists a minimal true-vector  $x \in \min[f]$  such that  $x \leq y + e_l$  for  $l = l_y$ . We must have  $x^{(l-1)} = y^{(l-1)}$ , since if  $x_i < y_i$  for some  $i < l$ , then  $y \geq x + e_i - e_l \succeq x$  would hold, i.e.  $y \succeq x$  would follow, implying  $y \in (\min[f])^*$  and yielding a contradiction with  $y \in F(h) = \mathcal{C} \setminus (\min[f])^*$ . Finally, the definition of  $l = l_y$  implies that  $y^{[l+1]} = c^{[l+1]}$ . Hence, our claim and the equality (6) follow.

The above claim implies that

$$F(h) \cap \max[f] \subseteq \{x^{(l-1)} + (x_l - 1)e_l + c^{[l+1]} \mid x \in \min[f], x_l > 0\},$$

and hence (5) and thus the lemma follow. □

**Lemma 2.** *Let  $f : \mathcal{C} \rightarrow \{0, 1\}$  be a monotone discrete binary function such that  $f \not\equiv 0$  and*

$$x \in T(f) \Rightarrow \alpha x \stackrel{\text{def}}{=} \alpha_1 x_1 + \dots + \alpha_n x_n \geq \beta, \tag{7}$$

where  $\alpha = (\alpha_1, \dots, \alpha_n)$  is a given real vector and  $\beta$  is a real threshold. Then

$$|\{x \in \mathcal{C} \mid \alpha x < \beta\} \cap \max[f]| \leq \sum_{x \in \min[f]} p(x).$$

*Proof.* Suppose that some of the weights  $\alpha_1, \dots, \alpha_n$  are negative, say  $\alpha_1 < 0, \dots, \alpha_k < 0$  and  $\alpha^{[k+1]} \geq 0$ . Since  $\alpha x \geq \beta$  for any  $x \in T(f)$  and since  $f$  is monotone, we have  $x \in T(f) \Rightarrow \alpha^{[k+1]}x \geq \beta - \alpha^{[k]}c^{[k]}$ . For any  $x \in \mathcal{C}$  we also have  $\{x \mid \alpha x < \beta\} \subseteq \{x \mid \alpha^{[k+1]}x < \beta - \alpha^{[k]}c^{[k]}\}$ . Hence it suffices to prove the lemma for the non-negative weight vector  $\alpha^{[k+1]}$  and the threshold  $\beta - \alpha^{[k]}c^{[k]}$ . In other words, we can assume without loss of generality that the original weight vector  $\alpha$  is non-negative.

Let  $\sigma \in \mathbb{S}^n$  be a permutation such that  $\alpha_{\sigma_1} \geq \alpha_{\sigma_2} \geq \dots \geq \alpha_{\sigma_n} \geq 0$ . Then the threshold function

$$g(x) = \begin{cases} 1 & \text{if } \alpha x \geq \beta, \\ 0 & \text{otherwise.} \end{cases}$$

is 2-monotonic with respect to  $\sigma$ . By (7), we have  $g \geq f$  for all  $x \in \mathcal{C}$ , i.e.,  $g$  majorates  $f$ . In addition,  $F(g) = \{x \in \mathcal{C} \mid \alpha x < \beta\}$ , and hence Lemma 2 follows from Lemma 1.  $\square$

We are now ready to show inequality (4) and finish the proof of Theorem 2. Given a non-empty set  $\mathcal{X} \subseteq \mathcal{F}_{A,b,c}$ , consider the monotone discrete function  $f : \mathcal{C} \rightarrow \{0, 1\}$  defined by the condition  $\min[f] = \mathcal{X}$ . Since (1) is monotone, any true vector of  $f$  also satisfies (1):

$$x \in T(f) \Rightarrow a_{k1}x_1 + \dots + a_{kn}x_n \geq b_k$$

for all  $k = 1, \dots, r$ . In addition,  $f \neq 0$  because  $\mathcal{X} \neq \emptyset$ . Thus, by Lemma 2 we have the inequalities

$$|\{x \mid a_{k1}x_1 + \dots + a_{kn}x_n < b_k\} \cap \max[f]| \leq \sum_{x \in \mathcal{X}} p(x) \tag{8}$$

for each  $k = 1, \dots, r$ . Now, from  $\max[f] = \mathcal{I}(\mathcal{X})$  we deduce that

$$\mathcal{I}(\mathcal{F}_{A,b,c}) \cap \mathcal{I}(\mathcal{X}) \subseteq \bigcup_{k=1}^r \{x \mid a_{k1}x_1 + \dots + a_{kn}x_n < b_k\} \cap \max[f],$$

and thus (4) and the theorem follow by (8).

### 3 Generating Minimal Integer Solutions via Integral Dualization

The proof of Theorem 3 has two ingredients. First, we show that given a monotone system (1), the sets  $\mathcal{I}(\mathcal{F}_{A,b,c})$  and  $\mathcal{F}_{A,b,c}$  can be *jointly* enumerated by iteratively solving the dualization problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$  introduced in Theorem 3. Second, we invoke Theorem 2 and argue that since the number of maximal infeasible vectors is relatively small, the generation of  $\mathcal{F}_{A,b,c}$  polynomially reduces to the joint generation of  $\mathcal{I}(\mathcal{F}_{A,b,c})$  and  $\mathcal{F}_{A,b,c}$ .

#### 3.1 Joint Generation of Dual Subsets in an Integral Box

Let  $\mathcal{F} = \mathcal{F}_{A,b,c}$  be the set of minimal integral vectors for (1), and consider the following problem of jointly generating all points of  $\mathcal{F}$  and  $\mathcal{I}(\mathcal{F})$ :

*GEN*( $\mathcal{F}, \mathcal{I}(\mathcal{F}), \mathcal{A}, \mathcal{B}$ ): Given two explicitly listed collections  $\mathcal{A} \subseteq \mathcal{F}$  and  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{F})$ , either find a new point in  $(\mathcal{F} \setminus \mathcal{A}) \cup (\mathcal{I}(\mathcal{F}) \setminus \mathcal{B})$ , or prove that these collections are complete:  $(\mathcal{A}, \mathcal{B}) = (\mathcal{F}, \mathcal{I}(\mathcal{F}))$ .

**Proposition 2.** *Problem  $GEN(\mathcal{F}, \mathcal{I}(\mathcal{F}), \mathcal{A}, \mathcal{B})$  can be solved in time  $poly(n, |\mathcal{A}|, |\mathcal{B}|, \log \|c\|_\infty) + T_{dual}$ , where  $T_{dual}$  denotes the time required to solve problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$ .*

*Proof.* The reduction is via the following Algorithm  $\mathcal{J}$ :

*Step 1.* Check whether  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{A})$ . If there is an  $x \in \mathcal{B} \setminus \mathcal{I}(\mathcal{A})$ , then  $x \notin \mathcal{F}^+$  because  $x \in \mathcal{B} \subseteq \mathcal{I}(\mathcal{F})$ . This and the inclusion  $\mathcal{A} \subseteq \mathcal{F}$  imply that  $x \notin \mathcal{A}^+$ . Since  $x \notin \mathcal{I}(\mathcal{A})$ , we can find a coordinate  $j \in \{1, \dots, n\}$  for which  $y = x + e_j \notin \mathcal{A}^+$ . By the maximality of  $x$  in  $\mathcal{C} \setminus \mathcal{F}^+$ ,  $y$  belongs to  $\mathcal{F}^+$  and therefore, there must exist a  $z \in \mathcal{F}$  such that  $z \leq y$ . Since  $z \notin \mathcal{A}^+$ , we have  $z \in \mathcal{F} \setminus \mathcal{A}$ , i.e.,  $z$  is a *new* minimal integral vector in  $\mathcal{F}$  which can be found in  $poly(n, |\mathcal{A}|, |\mathcal{B}|, \log \|c\|_\infty)$  time by performing coordinate binary searches on the box  $\{z \in \mathbb{Z}^n \mid 0 \leq z \leq y\}$ .

*Step 2* is similar to the previous step: we check whether  $\mathcal{A} \subseteq \mathcal{I}^{-1}(\mathcal{B})$ , where  $\mathcal{I}^{-1}(\mathcal{B})$  is the set of integral vectors minimal in  $\mathcal{C} \setminus \mathcal{B}^-$ . If  $\mathcal{A}$  contains an element that is not minimal in  $\mathcal{C} \setminus \mathcal{B}^-$ , we can find a new point in  $\mathcal{I}(\mathcal{F}) \setminus \mathcal{B}$  and halt.

*Step 3.* Suppose that  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{A})$  and  $\mathcal{A} \subseteq \mathcal{I}^{-1}(\mathcal{B})$ . Then  $(\mathcal{A}, \mathcal{B}) = (\mathcal{F}, \mathcal{I}(\mathcal{F})) \Leftrightarrow \mathcal{B} = \mathcal{I}(\mathcal{A})$ . (To see this, assume that  $\mathcal{B} = \mathcal{I}(\mathcal{A})$ , and suppose on the contrary that there is an  $x \in \mathcal{F} \setminus \mathcal{A}$ . Since  $x \notin \mathcal{A} = \mathcal{I}^{-1}(\mathcal{B})$  and  $x \notin \mathcal{B}^- \subseteq \mathcal{I}(\mathcal{F})^-$ , there must exist a  $y \in \mathcal{I}^{-1}(\mathcal{B}) = \mathcal{A} \subseteq \mathcal{F}$  such that  $y \leq x$ . Hence we get two distinct elements  $x, y \in \mathcal{F}$  such that  $y \leq x$ , which contradicts the definition of  $\mathcal{F}$ . The existence of an  $x \in \mathcal{I}(\mathcal{F}) \setminus \mathcal{B}$  leads to a similar contradiction.) To check the condition  $\mathcal{B} = \mathcal{I}(\mathcal{A})$ , we solve problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$ . If  $\mathcal{B} \neq \mathcal{I}(\mathcal{A})$ , we obtain a new point  $x \in \mathcal{I}(\mathcal{A}) \setminus \mathcal{B}$ . By (3), either  $x \in \mathcal{F}^+$ , or  $x \in \mathcal{I}(\mathcal{F})^-$  and we can decide which of these two cases holds by checking the feasibility of  $x$  for (1). In the first case, we obtain a new point  $y \in \{x\}^- \cap (\mathcal{F} \setminus \mathcal{A})$  by performing binary searches on the coordinates of the box  $\{y \in \mathbb{Z}^n \mid 0 \leq y \leq x\}$ . In the second case, a new point in  $\{x\}^+ \cap (\mathcal{I}(\mathcal{F}) \setminus \mathcal{B})$  can be obtained by searching the box  $\{y \in \mathbb{Z}^n \mid x \leq y \leq c\}$ .  $\square$

Let  $\mathcal{F} \subseteq \mathcal{C}$  be an arbitrary antichain, i.e., a system of integral vectors such that  $x \not\leq y$  for any two distinct elements  $x, y \in \mathcal{F}$ . It is easy to see that Algorithm  $\mathcal{J}$  and Proposition 2 can be used for any class of antichains  $\mathcal{F}$  defined by a polynomial-time membership oracle for  $\mathcal{F}^+$ .

### 3.2 Uniformly Dual-Bounded Antichains

Extending the definition of dual-bounded hypergraphs in [5], we say that (a class of antichains)  $\mathcal{F} \subseteq \mathcal{C}$  is *uniformly dual-bounded* if there exists a polynomial  $p$  such that, for any nonempty subset  $\mathcal{X} \subseteq \mathcal{F}$ , we have

$$|\mathcal{I}(\mathcal{F}) \cap \mathcal{I}(\mathcal{X})| \leq p(|\mathcal{X}|).$$

**Proposition 3.** *Suppose that  $\mathcal{F}$  is uniformly dual-bounded and there exists a polynomial-time membership oracle for  $\mathcal{F}^+$ . Then problem  $GEN(\mathcal{F})$  is polynomial-time reducible to problem  $DUAL(\mathcal{C}, \mathcal{A}, \mathcal{B})$ .*



*Proof.* Given a set  $\mathcal{X}$  in  $\mathcal{F}$ , we repeatedly run Algorithm  $\mathcal{J}$  until it either produces a new element in  $\mathcal{F} \setminus \mathcal{X}$  or proves that  $\mathcal{X} = \mathcal{F}$  by generating the entire family  $\mathcal{I}(\mathcal{F})$ . By Step 1, as long as Algorithm  $\mathcal{J}$  outputs elements of  $\mathcal{I}(\mathcal{F})$ , these elements also belong to  $\mathcal{I}(\mathcal{X})$ , and hence the total number of such elements does not exceed  $p(|\mathcal{X}|)$ .  $\square$

By Theorem 2, the set of minimal integral solutions to any monotone system of linear inequalities is uniformly-dual bounded, and hence Theorem 3 is a corollary of Proposition 3.

## 4 Dualization in Products of Chains

Let  $\mathcal{C} \stackrel{\text{def}}{=} \mathcal{C}_1 \times \dots \times \mathcal{C}_n$  be an integer box defined by the product of  $n$  chains  $\mathcal{C}_i = [l_i : u_i]$  where  $l_i, u_i \in \mathbb{Z}$  are, respectively, the lower and upper bounds of chain  $\mathcal{C}_i$ . Given an antichain  $\mathcal{A} \subseteq \mathcal{C}$ , and an antichain  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{A})$ , we say that  $\mathcal{B}$  is *dual to*  $\mathcal{A}$  if  $\mathcal{B} = \mathcal{I}(\mathcal{A})$ , i.e.,  $\mathcal{B}$  contains all the maximal elements of  $\mathcal{C} \setminus \mathcal{A}^+$ . If  $\mathcal{C}$  is the unit cube, we obtain the familiar notion of dual hypergraphs, where  $\mathcal{I}(\mathcal{A})$  becomes the complementary set of the transversal hypergraph of  $\mathcal{A}$ . In this section, we show how to extend the hypergraph dualization algorithm of [8] to arbitrary systems  $\mathcal{A}$  of integral vectors in a box  $\mathcal{C}$ .

As in [8], we shall analyze the running time of the algorithm in terms of the “volume”  $v = v(\mathcal{A}, \mathcal{B}) \stackrel{\text{def}}{=} |\mathcal{A}||\mathcal{B}|$  of the input problem. In general, a given problem will be decomposed into a number of subproblems which we solve recursively. Since we have assumed that  $\mathcal{B} \subseteq \mathcal{I}(\mathcal{A})$ , (3) implies that the following condition holds for the original problem and all subsequent subproblems:

$$a \not\leq b, \text{ for all } a \in \mathcal{A}, b \in \mathcal{B}. \tag{9}$$

Let  $R(v) = R(v(\mathcal{A}, \mathcal{B}))$  denote the number of subproblems that have to be solved in order to solve the original problem, and let  $m$  denote  $|\mathcal{A}| + |\mathcal{B}|$ , and  $[n] \stackrel{\text{def}}{=} \{1, \dots, n\}$ . We start with the following proposition that provides the base case for recursion.

**Proposition 4.** *Suppose  $\min\{|\mathcal{A}|, |\mathcal{B}|\} \leq \text{const}$ , then problem  $\text{DUAL}(\mathcal{C}, \mathcal{A}, \mathcal{B})$  is solvable in polynomial time.*

*Proof.* Let us assume without loss of generality that  $\mathcal{B} = \{b^1, \dots, b^k\}$ , for some constant  $k$ . For  $t \in [n]^k$  and  $i \in [n]$ , let  $I_i^t = \{j \in [k] \mid t_j = i\}$ . Then  $\mathcal{C} = \mathcal{A}^+ \cup \mathcal{B}^-$  if and only if for every  $t \in [n]^k$  for which

$$b_i^j \neq u_i, \text{ for all } i \in [n], j \in I_i^t, \tag{10}$$

there exists an  $a \in \mathcal{A}$  such that

$$a_i \leq \max\{b_i^j + 1 \mid j \in I_i^t\} \text{ if } I_i^t \neq \emptyset, \text{ and } a_i = l_i \text{ otherwise.} \tag{11}$$

To see this, assume first that  $\mathcal{C} = \mathcal{A}^+ \cup \mathcal{B}^-$  and consider any  $t \in [n]^k$  such that (10) holds. Let  $x \in \mathcal{C}$  be defined by taking  $x_i = \max\{b_i^j + 1 \mid j \in I_i^t\}$  if  $I_i^t \neq \emptyset$ ,

and  $x_i = l_i$  otherwise. Then  $x \in \mathcal{C} \setminus \mathcal{B}^-$  and hence  $x \in \mathcal{A}^+$ , implying that there is an  $a \in \mathcal{A}$  satisfying (11). On the other hand, let us assume that for every  $t \in [n]^k$  satisfying (10), there is an  $a \in \mathcal{A}$  for which (11) holds. Consider an  $x \in \mathcal{C} \setminus \mathcal{B}^-$ , then there must exist, for every  $j \in [k]$ , a  $t_j \in [n]$ , such that  $x_{t_j} \geq b_{t_j}^j + 1$ . Clearly  $t = (t_1, \dots, t_k) \in [n]^k$  satisfies (10), and therefore, there is an  $a \in \mathcal{A}$  such that  $a_i \leq \max\{b_i^j + 1 \mid j \in I_i^t\} \leq x_i$  if  $I_i^t \neq \emptyset$ , and  $a_i = l_i$  otherwise. This gives  $x \in \mathcal{A}^+$ .  $\square$

*Remark.* Having found an  $x \in \mathcal{C} \setminus (\mathcal{A}^+ \cup \mathcal{B}^-)$ , it is always possible to extend it to a maximal point with the same property in  $O(nm \log m)$  time as follows. Let  $\mathcal{Q}_i = \{a_i - 1 \mid a \in \mathcal{A}\} \cup \{x_i, u_i\}$ ,  $i = 1, \dots, n$ , and assume that this list is kept in sorted order for each  $i$ . For  $i = 1, \dots, n$ , we iterate  $x_i \leftarrow \max\{z \in \mathcal{Q}_i \mid (x_1, \dots, x_{i-1}, z, x_{i+1}, \dots, x_n) \notin \mathcal{A}^+\}$ . Then the resulting point  $x$  is maximal in  $\mathcal{C} \setminus (\mathcal{A}^+ \cup \mathcal{B}^-)$ .

Now given two integral antichains  $\mathcal{A}, \mathcal{B}$  that satisfy the necessary duality condition (9), we proceed as follows:

*Step 1.* If  $\min\{|\mathcal{A}|, |\mathcal{B}|\} \leq 2$ , the duality of  $\mathcal{A}$  and  $\mathcal{B}$  can be tested in  $O(n^3m)$  time using Proposition 4.

*Step 2.* For each  $k \in [n]$ :

1. if  $a_k > u_k$  for some  $a \in \mathcal{A}$  ( $b_k < l_k$  for some  $b \in \mathcal{B}$ ), then  $a$  (respectively,  $b$ ) can be clearly discarded from further consideration;
2. if  $a_k < l_k$  for some  $a \in \mathcal{A}$  ( $b_k > u_k$  for some  $b \in \mathcal{B}$ ), we set  $a_k \leftarrow l_k$  (respectively,  $b_k \leftarrow u_k$ ). Note that the duality condition (9) continues to hold after such replacements.

Thus we may assume for next steps that  $\mathcal{A}, \mathcal{B} \subseteq \mathcal{C}$ .

*Step 3.* Let  $a^o \in \mathcal{A}$ ,  $b^o \in \mathcal{B}$ . By (9), there exists an  $i \in [n]$ , such that  $a_i^o > b_i^o$ . Assume, with no loss of generality, that  $i = 1$  and set  $\mathcal{C}'_1 \leftarrow [a_1^o : u_1]$ ,  $\mathcal{C}''_1 \leftarrow [l_1 : a_1^o - 1]$ . (Alternatively, we may set  $\mathcal{C}''_1 \leftarrow [l_1 : b_1^o]$  and  $\mathcal{C}'_1 \leftarrow [b_1^o + 1 : u_1]$ .) Define

$$\begin{aligned} \mathcal{A}'' &= \{a \in \mathcal{A} \mid a_1 < a_1^o\}, & \mathcal{A}' &= \mathcal{A} \setminus \mathcal{A}'', & \epsilon_1^{\mathcal{A}} &= \frac{|\mathcal{A}'|}{|\mathcal{A}|}, \\ \mathcal{B}' &= \{b \in \mathcal{B} \mid b_1 \geq a_1^o\}, & \mathcal{B}'' &= \mathcal{B} \setminus \mathcal{B}', & \epsilon_1^{\mathcal{B}} &= \frac{|\mathcal{B}''|}{|\mathcal{B}|}. \end{aligned}$$

Observe that  $\epsilon_1^{\mathcal{A}} > 0$  and  $\epsilon_1^{\mathcal{B}} > 0$  since  $a^o \in \mathcal{A}'$  and  $b^o \in \mathcal{B}''$ .

Denoting by  $\mathcal{C}' = \mathcal{C}'_1 \times \mathcal{C}_2 \times \dots \times \mathcal{C}_n$ , and  $\mathcal{C}'' = \mathcal{C}''_1 \times \mathcal{C}_2 \times \dots \times \mathcal{C}_n$  the two half-boxes of  $\mathcal{C}$  induced by the above partitioning, it is then easy to see that  $\mathcal{A}$  and  $\mathcal{B}$  are dual in  $\mathcal{C}$  if and only if

$$\mathcal{A}, \mathcal{B}' \text{ are dual in } \mathcal{C}', \text{ and} \tag{12}$$

$$\mathcal{A}'', \mathcal{B} \text{ are dual in } \mathcal{C}''. \tag{13}$$

*Step 4.* Define  $\epsilon(v) = 1/\chi(v)$ , where  $\chi(v)^{\chi(v)} = v = v(\mathcal{A}, \mathcal{B})$ . If  $\min\{\epsilon_1^{\mathcal{A}}, \epsilon_1^{\mathcal{B}}\} > \epsilon(v)$ , we use the decomposition rule given above, which amounts to solving recursively two subproblems (12), (13) of respective volumes:

$$\begin{aligned} v(\mathcal{A}, \mathcal{B}') &= |\mathcal{A}||\mathcal{B}'| = |\mathcal{A}|(1 - \epsilon_1^{\mathcal{B}})|\mathcal{B}| = (1 - \epsilon_1^{\mathcal{B}})v(\mathcal{A}, \mathcal{B}) \leq (1 - \epsilon(v))v, \\ v(\mathcal{A}'', \mathcal{B}) &= |\mathcal{A}''||\mathcal{B}| = (1 - \epsilon_1^{\mathcal{A}})|\mathcal{A}||\mathcal{B}| = (1 - \epsilon_1^{\mathcal{A}})v(\mathcal{A}, \mathcal{B}) \leq (1 - \epsilon(v))v. \end{aligned}$$

This gives rise to the recurrence

$$R(v) \leq 1 + R((1 - \epsilon_1^{\mathcal{B}})v) + R((1 - \epsilon_1^{\mathcal{A}})v) \leq 1 + 2R((1 - \epsilon(v))v).$$

*Step 5.* Let us now suppose that  $\epsilon_1^{\mathcal{B}} \leq \epsilon(v)$ . In this case, we begin by solving subproblem (12). If  $\mathcal{A}, \mathcal{B}'$  are not dual in  $\mathcal{C}'$ , we get a point  $x$  maximal in  $\mathcal{C}' \setminus [\mathcal{A}^+ \cup (\mathcal{B}')^-]$ , and we are done. Otherwise we claim that

$$\mathcal{A}'', \mathcal{B} \text{ are dual in } \mathcal{C}'' \iff \forall a \in \tilde{\mathcal{A}} : \mathcal{A}'', \mathcal{B}'' \text{ are dual in } \mathcal{C}''(a), \quad (14)$$

where  $\tilde{\mathcal{A}} = \{a \in \mathcal{A} \mid a_1 \leq a_1^o\}$ , and  $\mathcal{C}''(a) = \mathcal{C}_1'' \times [a_2 : u_2] \times \dots \times [a_n : u_n]$ .

*Proof of (14).* The forward direction does not use (12). Suppose that there is an  $x \in \mathcal{C}''(a) \setminus [(\mathcal{A}'')^+ \cup (\mathcal{B}'')^-]$  for some  $a \in \tilde{\mathcal{A}}$ , then  $x_i \geq a_i$ , for  $i = 2, \dots, n$ . If  $x \in (\mathcal{B}')^-$ , i.e.,  $x \leq b$  for some  $b \in \mathcal{B}'$ , then by the definition of  $\mathcal{B}'$ ,  $b_1 \geq a_1^o$ . On the other hand,  $a \in \tilde{\mathcal{A}}$  implies that  $a_1 \leq a_1^o$ . But then,

$$(a_1, a_2, \dots, a_n) \leq (a_1^o, x_2, \dots, x_n) \leq (b_1, b_2, \dots, b_n),$$

which contradicts the assumed duality condition (9). This shows that  $x \in \mathcal{C}'' \setminus [(\mathcal{A}'')^+ \cup (\mathcal{B}' \cup \mathcal{B}'')^-]$ .

For the other direction, let  $x \in \mathcal{C}'' \setminus [(\mathcal{A}'')^+ \cup \mathcal{B}^-]$ . Since  $x \notin (\mathcal{B}')^-$  and  $x = (x_1, x_2, \dots, x_n) < y \stackrel{\text{def}}{=} (a_1^o, x_2, \dots, x_n)$ , the vector  $y$  is also not covered by  $\mathcal{B}'$ . Thus  $y \in \mathcal{C}' \setminus (\mathcal{B}')^-$ . We conclude therefore, assuming (12), that  $y \in \mathcal{A}^+$ , i.e., there is an  $a \in \mathcal{A}$  such that  $a \leq y$ . But this implies that  $a \in \tilde{\mathcal{A}}$  and hence that  $x \in \mathcal{C}''(a) \setminus [(\mathcal{A}'')^+ \cup (\mathcal{B}'')^-]$  for some  $a \in \tilde{\mathcal{A}}$ .  $\square$

It follows by (14) that, once we discover that (12) holds, we can reduce the solution of subproblem (13) to solving  $|\tilde{\mathcal{A}}|$  subproblems, each of which has a volume of  $v(|\mathcal{A}''|, |\mathcal{B}''|) \leq \epsilon_1^{\mathcal{B}} v(\mathcal{A}, \mathcal{B})$ . Thus we obtain the recurrence

$$R(v) \leq 1 + R((1 - \epsilon_1^{\mathcal{B}})v) + |\mathcal{A}|R(\epsilon_1^{\mathcal{B}}v) \leq R((1 - \epsilon_1^{\mathcal{B}})v) + \frac{v}{2}R(\epsilon_1^{\mathcal{B}}v),$$

where the last inequality follows from  $|\mathcal{A}| \leq v/3$  and  $v \geq 9$ .

*Step 6.* Finally, if  $\epsilon_1^{\mathcal{A}} \leq \epsilon(v) < \epsilon_1^{\mathcal{B}}$ , we solve subproblem (13), and if we discover that  $\mathcal{A}'', \mathcal{B}$  are dual in  $\mathcal{C}''$ , we obtain the following rule, symmetric to (14):

$$\mathcal{A}, \mathcal{B}' \text{ are dual in } \mathcal{C}' \iff \forall b \in \tilde{\mathcal{B}} : \mathcal{A}', \mathcal{B}' \text{ are dual in } \mathcal{C}'(b),$$

where  $\tilde{\mathcal{B}} = \{b \in \mathcal{B} \mid b_1 \geq a_1^o - 1\}$ , and  $\mathcal{C}'(b) = \mathcal{C}_1' \times [l_2 : b_2] \times \dots \times [l_n : b_n]$ . This reduces our original problem into one subproblem of volume  $\leq (1 - \epsilon_1^{\mathcal{A}})v$ , plus  $|\tilde{\mathcal{B}}|$  subproblems, each of volume at most  $\epsilon_1^{\mathcal{A}}v$ , thus giving the recurrence

$$R(v) \leq 1 + R((1 - \epsilon_1^{\mathcal{A}})v) + |\mathcal{B}|R(\epsilon_1^{\mathcal{A}}v) \leq R((1 - \epsilon_1^{\mathcal{A}})v) + \frac{v}{2}R(\epsilon_1^{\mathcal{A}}v).$$

Using induction on  $v \geq 9$ , it can be shown that the above recurrences imply that  $R(v) \leq v^{\chi(v)}$  (see [8]). As  $\chi(m^2) < 2\chi(m)$  and  $v(\mathcal{A}, \mathcal{B}) < m^2$ , we get  $\chi(v) < \chi(m^2) < 2\chi(m) \sim 2 \log m / \log \log m$ . Let us also note that every step above can be implemented in at most  $O(n^3 m)$  time, independent of the chains sizes  $|\mathcal{C}_i|$ . This establishes the bound stated in Theorem 4.

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