

# An inequality for polymatroid functions and its applications <sup>★</sup>

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## Abstract

An integral-valued set function  $f : 2^V \mapsto \mathbb{Z}$  is called polymatroid if it is submodular, non-decreasing, and  $f(\emptyset) = 0$ . Given a polymatroid function  $f$  and an integer threshold  $t \geq 1$ , let  $\alpha = \alpha(f, t)$  denote the number of maximal sets  $X \subseteq V$  satisfying  $f(X) < t$ , let  $\beta = \beta(f, t)$  be the number of minimal sets  $X \subseteq V$  for which  $f(X) \geq t$ , and let  $n = |V|$ . We show that if  $\beta \geq 2$  then  $\alpha \leq \beta^{(\log t)/c}$ , where  $c = c(n, \beta)$  is the unique positive root of the equation  $1 = 2^c(n^{c/\log \beta} - 1)$ . In particular, our bound implies that  $\alpha \leq (n\beta)^{\log t}$  for all  $\beta \geq 1$ . We also give examples of polymatroid functions with arbitrarily large  $t, n, \alpha$  and  $\beta$  for which  $\alpha \geq \beta^{(.551 \log t)/c}$ .

More generally, given a polymatroid function  $f : 2^V \mapsto \mathbb{Z}$  and an integral threshold  $t \geq 1$ , consider an arbitrary hypergraph  $\mathcal{H}$  such that  $|\mathcal{H}| \geq 2$  and  $f(H) \geq t$  for all  $H \in \mathcal{H}$ . Let  $\mathcal{S}$  be the family of all maximal independent sets  $X$  of  $\mathcal{H}$  for which  $f(X) < t$ . Then  $|\mathcal{S}| \leq |\mathcal{H}|^{(\log t)/c(n, |\mathcal{H}|)}$ . As an application, we show that given a system of polymatroid inequalities  $f_1(X) \geq t_1, \dots, f_m(X) \geq t_m$  with quasi-polynomially bounded right hand sides  $t_1, \dots, t_m$ , all minimal feasible solutions to this system can be generated in incremental quasi-polynomial time. In contrast to this result, the generation of all maximal infeasible sets is an NP-hard problem for many polymatroid inequalities of small range.

*Key words:* dualization, incremental algorithm, independent set, matroid, submodular function, polymatroid function, system of polymatroid inequalities, transversal hypergraph.

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# 1 Introduction

## 1.1 Main results

Let  $V$  be a finite set of cardinality  $|V| = n$ , let  $f : 2^V \mapsto \mathbb{Z}_+$  be a set-function taking non-negative integral values, and let  $r = r(f)$  denote the *range* of  $f$ , i.e.,  $r(f) = \max\{f(X) \mid X \subseteq V\}$ . The set-function  $f$  is called *Boolean* if  $r = 1$ , *monotone* if  $f(X) \leq f(Y)$  whenever  $X \subseteq Y$ , and *submodular* if

$$f(X \cup Y) + f(X \cap Y) \leq f(X) + f(Y) \quad (1)$$

holds for all subsets  $X, Y \subseteq V$ . Finally,  $f$  is called a *polymatroid function* if it is monotone, submodular and  $f(\emptyset) = 0$ . Let us remark that in general, determining the range of a submodular set function may be an NP-hard problem, while  $r(f) = f(V)$  holds for polymatroid functions.

Given a monotone function  $f : 2^V \mapsto \{0, 1, \dots, r\}$ , and an integral threshold  $t \in \{1, \dots, r\}$  let us denote by  $\mathcal{B}_t = \mathcal{B}_t(f)$  the family of all minimal subsets  $X \subseteq V$  for which  $f(X) \geq t$ , and analogously, let us denote by  $\mathcal{A}_t = \mathcal{A}_t(f)$  the family of all maximal subsets  $X \subseteq V$  for which  $f(X) < t$ . It is easy to see that  $\mathcal{A}_t = \mathcal{I}(\mathcal{B}_t)$ , where  $\mathcal{I}(\cdot)$  denotes the family of all maximal independent sets for the hypergraph  $(\cdot)$ . Throughout the paper we shall use the notation  $\alpha = |\mathcal{A}_t(f)|$  and  $\beta = |\mathcal{B}_t(f)|$ .

**Theorem 1** *For every polymatroid function  $f$  and threshold  $t \in \{1, \dots, r(f)\}$  such that  $\beta \geq 2$  we have the inequality*

$$\alpha \leq \beta^{(\log t)/c(n, \beta)}, \quad (2)$$

where  $c(n, \beta)$  is the unique positive root of the equation

$$2^c(n^{c/\log \beta} - 1) = 1. \quad (3)$$

In addition,  $\alpha \leq n$  holds if  $\beta = 1$ .

Let us first remark that by (3),  $1 = n^{-c/\log \beta} + (n\beta)^{-c/\log \beta} \geq 2(n\beta)^{-c/\log \beta}$ , and hence  $\beta^{1/c(n, \beta)} \leq n\beta$ . Consequently, for  $\beta \geq 2$  (in which case  $n \geq 2$  is implied, too) we can replace (2) by the simpler but weaker inequality

$$\alpha \leq (n\beta)^{\log t}. \quad (4)$$

In fact, (4) holds even in case of  $\beta = 1$ , because if the hypergraph  $\mathcal{B}_t$  consists only of a single hyperedge  $X \subseteq V$ , then  $|\mathcal{A}_t| \leq |X| \leq n$  follows immediately by the relation  $\mathcal{A}_t = \mathcal{I}(\mathcal{B}_t)$ . On the other hand, for large  $\beta$  the bound of

Theorem 1 becomes increasingly stronger than (4). For instance,  $c(n, n) = \log(1 + \sqrt{5}) - 1 > .694$ ,  $c(n, n^2) > 1.102$ , and  $c(n, n^\sigma) \sim \log \sigma$  for large  $\sigma$ .

Let us remark next that the bound of Theorem 1 is reasonably sharp. As we shall show in Section 3, for any positive integers  $k$  and  $l$  there exists a polymatroid function  $f$  of range  $r = 2^k$  for which  $n = kl$ ,  $|\mathcal{A}_r| = l^k$ , and  $|\mathcal{B}_r| = kl(l-1)/2$ . Thus, letting  $t = r$  and  $l = 2^k$ , we obtain an infinite family of polymatroid functions for which

$$\alpha \geq \beta^{(.551 \log t)/c(n, \beta)} \quad \text{and} \quad \alpha \geq (n\beta)^{(\frac{1}{3} + o(1)) \log t}, \quad (5)$$

as  $t = r \rightarrow \infty$ , see Section 3 for more detail. In Section 3 we also show that our lower bounds (5) can be achieved within the subclass of rank functions defined on the subsets of some linear space. Namely, we can construct  $kl$  subspaces  $V_{ij} \subseteq \mathbb{R}^{2^k}$ ,  $i = 1, \dots, k$ ,  $j = 0, \dots, l-1$  of dimension  $2^{k-1}$  each, such that for any  $i$  and  $j \neq j'$  we have  $\dim(V_{ij} \cup V_{ij'}) = 2^k$ , while for every  $(j_1, j_2, \dots, j_k) \in \{0, 1, \dots, l-1\}^k$  the inequality  $\dim(\bigcup_{i=1}^k V_{ij_i}) < 2^k$  holds.

Let us finally note that for many classes of polymatroid functions,  $\beta$  cannot be bounded by a quasi-polynomial estimate of the form  $(n\alpha)^{\text{poly} \log r}$ . Let us consider for instance, a graph  $G = t \times K_2$  consisting of  $t$  disjoint edges, and let  $f(X)$  be the number of edges  $X$  intersects, for  $X \subseteq V(G)$ . Then  $f$  is a polymatroid function of range  $r = t$ , and we have  $n = 2t$ ,  $\alpha = |\mathcal{A}_t| = t$  and  $\beta = |\mathcal{B}_t| = 2^t$ .

We will strengthen Theorem 1 as follows. Given a non-empty hypergraph  $\mathcal{H}$  on the vertex set  $V$ , a polymatroid function  $f : 2^V \mapsto \mathbb{Z}_+$ , and a integral positive threshold  $t$ , the pair  $(f, t)$  is called a *polymatroid separator* for  $\mathcal{H}$  if  $f(H) \geq t$  for all  $H \in \mathcal{H}$ .

**Theorem 2** *Let  $(f, t)$  be a polymatroid separator for a hypergraph  $\mathcal{H}$  of cardinality  $|\mathcal{H}| \geq 2$ . Then*

$$|\mathcal{A}_t(f) \cap \mathcal{I}(\mathcal{H})| \leq |\mathcal{H}|^{(\log t)/c(n, |\mathcal{H}|)}, \quad (6)$$

where  $\mathcal{I}(\mathcal{H})$  is the family of all maximal independent sets for  $\mathcal{H}$ .

In particular, if  $(f, t)$  is a polymatroid separator for a non-empty hypergraph  $\mathcal{H}$ , then  $|\mathcal{A}_t(f) \cap \mathcal{I}(\mathcal{H})| \leq (n|\mathcal{H}|)^{\log t}$ .

Clearly, Theorem 1 is a special case of Theorem 2 for  $\mathcal{H} = \mathcal{B}_t(f)$ . Since the right-hand side of (6) monotonically increases with  $|\mathcal{H}|$ , we will only need to prove Theorem 2 for *Sperner* hypergraphs  $\mathcal{H}$ , i.e., under the assumption that that none of the hyperedges of  $\mathcal{H}$  contains another hyperedge of  $\mathcal{H}$ .

Let  $\mathcal{H}$  be a non-empty Sperner hypergraph on the vertex set  $V$ . A polymatroid separator  $(f, t)$  is called *exact* for  $\mathcal{H}$  if  $\mathcal{H} = \mathcal{B}_t(f)$  and consequently,  $\mathcal{I}(\mathcal{H}) = \mathcal{A}_t(f)$ . It is immediate to see that for exact separators  $(f, t)$ , Theorem 2 implies  $r(f) \geq t \geq |\mathcal{I}(\mathcal{H})|^{c(n, |\mathcal{H}|)/\log |\mathcal{H}|}$  whenever  $|\mathcal{H}| \geq 2$ . As we shall see in Section 2.1, any Sperner hypergraph  $\mathcal{H}$  has an exact polymatroid separator  $(f, t)$  such that  $r(f) = t = |\mathcal{I}(\mathcal{H})|$  holds.

A polymatroid separator  $(f, t)$  is called *linear* if the set-function  $f$  is modular, i.e., if equality holds in (1) for all  $X, Y \subseteq V$ . As shown in [9,12], if  $(f, t)$  is a linear separator for a non-empty hypergraph  $\mathcal{H}$ , then  $|\mathcal{A}_t(f) \cap \mathcal{I}(\mathcal{H})| \leq n|\mathcal{H}|$ . Unlike (6), this bound does not depend on  $t$  and holds even for real-valued linear separators.

## 1.2 Generating minimal feasible sets for systems of polymatroid inequalities

Before proceeding further we shall discuss some algorithmic implications of Theorem 2 related to the complexity of enumerating all minimal solutions to a system of polymatroid inequalities

$$f_i(X) \geq t_i, \quad i = 1, \dots, m, \quad (7)$$

over the subsets  $X \subseteq V$ . Specifically, letting  $\mathcal{B}$  denote the family of all minimal feasible sets for (7), we consider the following problem:

*GEN*( $\mathcal{B}, \mathcal{H}$ ): Given a system of polymatroid inequalities (7) and a collection  $\mathcal{H} \subseteq \mathcal{B}$  of minimal feasible sets for (7), either find a new minimal feasible set  $H \in \mathcal{B} \setminus \mathcal{H}$  for (7), or show that  $\mathcal{H} = \mathcal{B}$ .

In what follows, we assume that each of the polymatroid functions  $f_i : 2^V \mapsto \mathbb{Z}$  is defined via a (quasi) polynomial-time evaluation oracle, and that  $t_1, \dots, t_m$  are given positive integral thresholds. Let us also note that since  $\min\{f, t\}$  is a polymatroid function whenever  $f$  is polymatroid, we could also assume without loss of generality that  $t_i = \text{range}(f_i)$ .

Clearly, we can incrementally generate all sets in  $\mathcal{B}$  by initializing  $\mathcal{H} = \emptyset$  and then iteratively solving the above generation problem  $|\mathcal{B}| + 1$  times. It is easy to see that the first minimal feasible set  $H \in \mathcal{B}$  can be found (or  $\mathcal{B} = \emptyset$  can be recognized) by evaluating (7)  $n + 1$ -times. Furthermore, since  $\mathcal{I}(\{H\}) = \{V \setminus \{x\} \mid x \in H\}$ , the second minimal feasible set can also be identified (or  $\mathcal{B} = \{H\}$  can be recognized) in another  $n + |H|$  evaluations of (7). Thus, in what follows we can assume  $|\mathcal{H}| \geq 2$  without any loss of generality.

By definition, each pair  $(f_i, t_i)$  is a polymatroid separator for  $\mathcal{H}$ , and therefore Theorem 2 implies the inequalities

$$|\mathcal{A}_{t_i}(f_i) \cap \mathcal{I}(\mathcal{H})| \leq |\mathcal{H}|^{(\log t_i)/c(n, |\mathcal{H}|)}, \quad i = 1, \dots, m.$$

Let  $\mathcal{A} = \mathcal{I}(\mathcal{B})$  be the hypergraph of all maximal infeasible sets for (7), then  $\mathcal{A} \subseteq \bigcup_{i=1}^m \mathcal{A}_{t_i}(f_i)$ . Hence we arrive at the following bound.

**Corollary 1** *Let  $\mathcal{B}$  be the family of all minimal feasible sets for the system of polymatroid inequalities (7) and let  $\mathcal{H} \subseteq \mathcal{B}$  be an arbitrary subfamily of  $\mathcal{B}$  of size  $|\mathcal{H}| \geq 2$ . Then*

$$|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})| \leq m |\mathcal{H}|^{(\log t)/c(n, |\mathcal{H}|)} \leq m(n|\mathcal{H}|)^{\log t}, \quad (8)$$

where  $t = \max\{t_1, \dots, t_m\}$ . In particular,  $|\mathcal{I}(\mathcal{B})| \leq m|\mathcal{B}|^{(\log t)/c(n, |\mathcal{B}|)}$ .

As mentioned earlier, if the functions  $f_i$  in (7) are linear, then (8) can be improved to show that  $|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})| \leq mn|\mathcal{H}|$  (see [9,12]).

Corollary 1 shows that if the right hand sides of (7) are quasi-polynomially bounded, i.e.,  $t = \max\{t_1, \dots, t_m\} \leq 2^{\text{poly} \log nm}$ , then for any non-empty hypergraph  $\mathcal{H} \subseteq \mathcal{B}$  we have

$$|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})| \leq \text{quasi-poly}(n, m, |\mathcal{H}|). \quad (9)$$

By definition, the family  $\mathcal{B} \subseteq 2^V$  of all minimal feasible sets for (7) is a Sperner hypergraph. Furthermore, the hypergraph  $\mathcal{B}$  has a (quasi) polynomial-time superset oracle, i.e., given a set  $X \subseteq V$ , we can determine in (quasi) polynomial time whether or not  $X$  contains some set  $H \in \mathcal{B}$  (this is equivalent to checking the feasibility of  $X$  for (7)). As shown in [6,9,12,17], for any Sperner hypergraph  $\mathcal{B}$  defined via a (quasi) polynomial-time superset oracle, problem  $GEN(\mathcal{B}, \mathcal{H})$  reduces in (quasi) polynomial time to  $|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})|$  instances of the *hypergraph dualization problem*:

*Given two explicitly listed Sperner families  $\mathcal{H} \subseteq 2^V$  and  $\mathcal{G} \subseteq \mathcal{I}(\mathcal{H})$ , either find a new maximal independent set  $X \in \mathcal{I}(\mathcal{H}) \setminus \mathcal{G}$  or show that  $\mathcal{G} = \mathcal{I}(\mathcal{H})$ .*

To see this reduction in our case, consider a hypergraph  $\mathcal{H} \subseteq \mathcal{B}$  of minimal feasible solutions to (7). Start generating maximal independent sets for  $\mathcal{H}$  checking, for each generated set  $X \in \mathcal{I}(\mathcal{H})$ , whether or not  $X$  is feasible for (7). If  $X$  is feasible for (7) then  $X$  contains a new minimal solution to (7) which can be found by querying the feasibility oracle at most  $|X| + 1$  times. If  $X \in \mathcal{I}(\mathcal{H})$  is infeasible for (7), then it is easy to see that  $X \in \mathcal{I}(\mathcal{B})$ , and hence the number of such infeasible sets  $X$  is bounded by  $|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})|$ .

The hypergraph dualization problem can be solved in quasi-polynomial time  $\text{poly}(n) + (|\mathcal{H}| + |\mathcal{G}|)^{o(\log(|\mathcal{H}| + |\mathcal{G}|))}$ , see [16]. (Moreover, for the hypergraphs of bounded edge-sizes the dualization problem can be efficiently solved in parallel, see [7].) Hence from the above reduction we readily obtain the following result.

**Theorem 3** *Consider a system of polymatroid inequalities (7) in which the right hand sides are bounded by a quasi-polynomial in the dimension of the system. Suppose further that (7) has a quasi-polynomial-time feasibility oracle. Then problem  $\text{GEN}(\mathcal{B}, \mathcal{H})$  can be solved in quasi-polynomial time, and hence all minimal feasible sets for (7) can be enumerated in incremental quasi-polynomial time.*

We discuss some special cases and applications of Theorem 3 in Section 2. It is worth mentioning that in most of our examples, generating all maximal infeasible sets for (7) turns out to be NP-hard.

### 1.3 Proper mappings of independent sets into binary trees

Our proof of Theorem 2 makes use of a combinatorial construction which may be of independent interest. Theorem 2 states that for any polymatroid separator  $(f, t)$  of a hypergraph  $\mathcal{H}$  we have

$$r(f) \geq t \geq |\mathcal{S}|^{c(n, |\mathcal{H}|) / \log(|\mathcal{H}|)},$$

where  $\mathcal{S} = \mathcal{I}(\mathcal{H}) \cap \{X \mid f(X) < t\}$ , i.e., the range of  $f$  must increase with the size of  $\mathcal{S} \subseteq \mathcal{I}(\mathcal{H})$ . Thus, to prove the theorem we must first find ways to provide lower bounds on the range of a polymatroid function. To this end we shall show that the number of independent sets which can be organized in a special way into a binary tree structure provides such a lower bound.

Let  $\mathbf{T}$  denote a binary tree,  $V(\mathbf{T})$  denote its node set, and let  $L(\mathbf{T})$  denote the set of its leaves. For every node  $v \in V(\mathbf{T})$  let  $\mathbf{T}(v)$  be the binary sub-tree rooted at  $v$ . Obviously, for every two nodes  $u, v$  of  $\mathbf{T}$  either the sub-trees  $\mathbf{T}(u)$  and  $\mathbf{T}(v)$  are disjoint, or one of them is a sub-tree of the other one. The nodes  $u$  and  $v$  are called *incomparable* in the first case, and *comparable* in the second case.

Given a Sperner hypergraph  $\mathcal{H}$  and a binary tree  $\mathbf{T}$ , let us consider mappings  $\phi : L(\mathbf{T}) \mapsto \mathcal{I}(\mathcal{H})$  assigning maximal independent sets  $I_l \in \mathcal{I}(\mathcal{H})$  to the leaves  $l \in L(\mathbf{T})$ . Let us associate furthermore to every node  $v \in V(\mathbf{T})$  the intersection  $S_v = \bigcap_{l \in L(\mathbf{T}(v))} I_l$ . Let us call finally the mapping  $\phi$  *proper* if it is injective, i.e., assigns different independent sets to different leaves, and if the sets  $S_u \cup S_v$  are not independent whenever  $u$  and  $v$  are incomparable nodes of

**T.** Let us point out that the latter condition means that the set  $S_u \cup S_v$ , for incomparable nodes  $u$  and  $v$ , must contain a hyperedge  $H \in \mathcal{H}$ , as a subset. Since the intersection of independent sets is always independent, it follows, in particular that both  $S_v$  and  $S_u$  are non-empty independent sets (otherwise their union could not be non-independent.) Finally, since all non-root nodes  $u \in V(\mathbf{T})$  have at least one incomparable node  $v \in V(\mathbf{T})$ , we get that the sets  $S_u$ , for  $u \in V(\mathbf{T}) \setminus \{s\}$  are all non-empty and independent.

**Lemma 1** *Let us consider a Sperner hypergraph  $\mathcal{H}$  and a polymatroid separator  $(f, t)$  of it, and let us denote by  $\mathcal{S}$  the subfamily of maximal independent sets, separated by  $(f, t)$  from  $\mathcal{H}$ , as before. Let us assume further that  $\mathbf{T}$  is a binary tree for which there exists a proper mapping  $\phi : L(\mathbf{T}) \mapsto \mathcal{S}$ . Then, we have*

$$r(f) \geq t \geq |L(\mathbf{T})|. \quad (10)$$

Let us note that if a proper mapping exists for a binary tree  $\mathbf{T}$ , then we can associate a hyperedge  $H_u \in \mathcal{H}$  to every node  $u \in V(\mathbf{T}) \setminus L(\mathbf{T})$  in the following way: Let  $v$  and  $w$  be the two successors of  $u$ . Since  $v$  and  $w$  are incomparable, the union  $S_v \cup S_w$  must contain a hyperedge from  $\mathcal{H}$ . Let us choose such a hyperedge, and denote it by  $H_u$ . Let us observe next that if  $l \in L(\mathbf{T}(v))$  and  $l' \in L(\mathbf{T}(w))$ , then  $S_v \subseteq I_l$  and  $S_w \subseteq I_{l'}$ , and thus  $H_u \subseteq I_l \cup I_{l'}$ . In other words, to construct a large binary tree for which there exists a proper mapping, we have to find a way of splitting the family of independent sets, repeatedly, such that the union of any two independent sets, belonging to different parts of the split must contain a hyperedge of  $\mathcal{H}$ . We shall show next that indeed, such a construction is possible.

**Lemma 2** *For every Sperner hypergraph  $\mathcal{H} \subseteq 2^V$ ,  $|\mathcal{H}| \geq 2$ , and for every subfamily  $\mathcal{S} \subseteq \mathcal{I}(\mathcal{H})$  of its maximal independent sets there exists a binary tree  $\mathbf{T}$  and a proper mapping  $\phi : L(\mathbf{T}) \mapsto \mathcal{S}$ , such that*

$$|L(\mathbf{T})| \geq |\mathcal{S}|^{c(n, |\mathcal{H}|) / \log |\mathcal{H}|}, \quad (11)$$

where  $n = |V|$ .

Clearly, Lemmas 1 and 2 imply Theorem 2, which in turn implies Theorem 1. Lemmas 1 and 2 will be proved in Section 4.

#### 1.4 Related work

For the special case when the Sperner hypergraph  $\mathcal{H} = E(G)$  is the edge set of a connected graph  $G$  and  $\mathcal{S} = \mathcal{I}(\mathcal{H})$ , Lemma 2 follows from the results by

Balas and Yu (1989) (see also [2,29]): Theorem 4 of [5] claims that

$$2^p \leq |\mathcal{I}(\mathcal{H})| \leq \delta^p + 1, \quad (12)$$

where  $\delta$  is the number of pairs of vertices in  $V$  at distance 2 (in particular,  $\delta < n^2/2$ ), and  $p$  is the cardinality of a maximum induced matching in  $G$ . Any such matching can be used to construct a proper mapping  $\phi : L(\mathbf{T}) \mapsto \mathcal{I}(\mathcal{H})$  for a uniform binary tree  $\mathbf{T}$  of depth  $p$ , i.e., for which  $|L(\mathbf{T})| = 2^p$ . Namely, let  $e_i = (v_0^i, v_1^i)$  for  $i = 1, \dots, p$  denote the edges of the induced matching. For each binary vector  $x = (x_1, x_2, \dots, x_p) \in \{0, 1\}^p$  let us associate a subset  $\tilde{I}_x$  defined by  $\tilde{I}_x = \{v_{x_i}^i \mid i = 1, \dots, p\}$ . Finally, let  $I_x \in \mathcal{I}(\mathcal{H})$  be a maximal independent set of  $G$  containing  $\tilde{I}_x$ , for  $x \in \{0, 1\}^p$ . (Since the edges  $e_1, \dots, e_p$  form an induced matching, all sets  $\tilde{I}_x$ ,  $x \in \{0, 1\}^p$  are independent.) Thus, the sets  $I_x$ ,  $x \in \{0, 1\}^p$  are pairwise distinct maximal independent sets of  $G$  by the above construction, and  $e_i \subseteq I_x \cup I_y$  whenever  $x_i \neq y_i$ . Naturally, the leaves of the binary tree  $\mathbf{T}$  of depth  $p$  are also encoded by the binary sequences of length  $p$ . Thus a mapping  $\phi$  is defined by assigning  $I_x$  to the leaf coded with  $x$ , and it is not difficult to verify that this mapping is proper (and that the edge  $e_i$  will be the associated hyperedge at each node of  $\mathbf{T}$  at depth  $i - 1$ , for  $i = 1, \dots, p$ ).

By these definitions,  $\log |L(\mathbf{T})| = p$  and  $\log |\mathcal{I}(\mathcal{H})| \leq p \log \delta$ , according to (12), and hence

$$|L(\mathbf{T})| \geq |\mathcal{I}(\mathcal{H})|^{1/\log \delta} \geq |\mathcal{I}(\mathcal{H})|^{1/(2 \log n)},$$

whereas the bound of Lemma 2 gives:

$$|L(\mathbf{T})| \geq |\mathcal{I}(\mathcal{H})|^{c(n, |\mathcal{H}|)/\log |\mathcal{H}|}.$$

For this reason, Lemma 2 can be viewed as an extension of Theorem 4 of [5] from graphs to hypergraphs (even though we could not generalize directly the notion of an induced matching).

Let us remark here that according to the above results, the existence of a “large” induced matching in a graph, or more generally, the existence of a “large” binary tree with a proper mapping in a hypergraph can be viewed as the reason for the existence of “many” maximal independent sets. These reasons, however, may not be easy to exhibit for a given graph or hypergraph. The problem of finding the maximum size induced matching in a graph is known to be NP-hard (see e.g. [31]) and it is even hard to approximate it well (see [15]). The complexities of the corresponding problems of finding or approximating well the largest binary tree with a proper mapping are open.

The problem of generating all maximal independent sets for a general independence system was shown to be NP-hard in [23]. Some special classes of independence systems were also discussed in the same paper. In particular, it was conjectured that, for linear systems (7), problem  $GEN(\mathcal{B}, \mathcal{H})$  cannot

be solved in polynomial time unless  $P=NP$ . Furthermore, an algorithm was given, to generate all maximal independent sets in the intersection of  $m$  matroids, whose running time was exponential in  $m$ . In contrast to these results, Corollary 1 implies that these two problems can be solved in quasi-polynomial time.

Other algorithms for generating combinatorial structures, described by polymatroid inequalities (e.g., spanning trees, maximal cliques and cycles in a graph) can be found in [20,30,32], see also Section 2 below for other examples. It should be mentioned also that, even though generating all maximal infeasible sets for (7) is NP-hard, there is a polynomial randomized scheme for nearly uniform sampling from the set of *all* infeasible sets for linear systems (7) ([21,28]). On the other hand, by using the amplification technique of [19] it is easy to show that a similar randomized scheme for nearly uniform sampling from within the set of all minimal feasible sets for a given system (7) would imply that any NP-complete problem can be solved in polynomial time by a randomized algorithm with arbitrary small one-sided failure probability (see [17] for more detail). The complexity of counting the number of minimal solutions for linear systems (7) is studied in [14].

Finally, we remark that similar inequalities for some special cases of Theorem 1 appear in [4,8,9,11,13], see also Section 2; many other intersection inequalities for hypergraphs can be found in [3], chapter 4.

## 2 Some polymatroid separators

In this Section we discuss a number of polymatroid separators as well as some stronger forms of Theorems 1 and 2 for specific cases.

### 2.1 *Partial unions and transversals of a hypergraph. Fairly independent sets.*

Let  $\mathcal{H}$  be an explicitly given hypergraph on a finite vertex set  $V$ . For a set  $X \subseteq V$ , define

$$f(X) = |\{H \in \mathcal{H} \mid X \cap H \neq \emptyset\}| \quad (13)$$

to be the number of hyperedges of  $\mathcal{H}$  which have a non-empty intersection with  $X$ . The function  $f$  is polymatroid and  $r(f) = |\mathcal{H}|$ . For a given threshold  $t \in \{1, \dots, |\mathcal{H}|\}$ , the hypergraph  $\mathcal{B}_t$  consists of all *partial  $t$ -transversals*, i.e., all minimal vertex sets which intersect at least  $t$  hyperedges of  $\mathcal{H}$ . (Equivalently,

the complement to each partial  $t$ -transversal is a *fairly independent set*, i.e., a maximal vertex set containing at most  $|\mathcal{H}| - t$  hyperedges of  $\mathcal{H}$ .)

For  $t = |\mathcal{H}|$  the set family  $\mathcal{B}_t$  is identical to the transversal hypergraph  $\mathcal{H}^d$  of  $\mathcal{H}$ . This readily implies the following claim.

**Proposition 1** *Every non-empty Sperner hypergraph  $\mathcal{H}$  has an exact polymatroid separator of the form (13).*

**Proof.** Since  $\mathcal{H}^{dd} = \mathcal{H}$  holds for any Sperner hypergraph  $\mathcal{H}$ , applying definition (13) to  $\mathcal{H}^d$  we obtain a polymatroid function  $f'$  of range  $r(f') = |\mathcal{H}^d|$  such that the family of all minimal feasible sets for  $f'(X) \geq r$  is exactly  $\mathcal{H}$ .  $\square$

We mention in passing that each hypergraph  $\mathcal{H}$  also admits a monotone *supermodular* separator  $h(X) = |\{H \in \mathcal{H} \mid H \subseteq X\}|$  of range  $|\mathcal{H}|$ .

Returning to the polymatroid function  $f$  defined by (13), we can identify the hypergraph  $\mathcal{A}_t = \mathcal{A}_t(f)$  with the family of all maximal subsets of  $V$  that avoid at least  $|\mathcal{H}| - t + 1$  hyperedges of  $\mathcal{H}$ . Equivalently, the complement to each set  $X \in \mathcal{A}_t$  is a *partial  $(|\mathcal{H}| - t + 1)$ -union*, that is a minimal vertex set containing at least  $|\mathcal{H}| - t + 1$  hyperedges of  $\mathcal{H}$ .

As shown in [10,11], the inequality of Theorem 1 can be strengthened for the class of polymatroid functions (13) as follows:

$$|\mathcal{A}_t| \leq t|\mathcal{B}_t| \quad \text{for any } t \in \{1, \dots, r\}. \quad (14)$$

Moreover, the above bounds may be sharp for arbitrarily large hypergraphs  $\mathcal{H}$  and  $t \in \{1, \dots, |\mathcal{H}|\}$ . (Theorem 2 can also be strengthened to a similar sharp linear bound, see [10,11].)

Given a list of hyperedges of  $\mathcal{H}$  and a threshold  $t \in \{1, \dots, |\mathcal{H}|\}$ , we can easily check whether or not a given vertex set  $X$  contains a partial  $t$ -transversal (we only need to check if  $X$  intersects at least  $t$  hyperedges of  $\mathcal{H}$ ). This gives a polynomial-time feasibility oracle for  $f(X) \geq t$ . From Theorem 3 we thus conclude that all partial  $t$ -transversals  $X \in \mathcal{B}_t$  for a given hypergraph  $\mathcal{H}$  can be generated in incremental quasi-polynomial time.

Since problem  $GEN(\mathcal{B}_t, \mathcal{X})$  can be solved in quasi-polynomial time, it is unlikely to be NP-hard. In contrast to this result, problem  $GEN(\mathcal{A}_t, \mathcal{X})$  is known to be NP-hard. Specifically, it is NP-complete to decide if  $\mathcal{X} \neq \mathcal{A}_t$  for an explicitly given set family  $\mathcal{X} \subseteq \mathcal{A}_t$  (see [26] and also [11] for more detail).

## 2.2 Maximal frequent and minimal infrequent sets for binary matrices

The notion of frequent sets in data-mining can be related naturally to the polymatroid separator (13) considered above. Let  $\mathcal{D} : \mathcal{R} \times \mathcal{C} \mapsto \{0, 1\}$  be a given  $m \times n$  binary matrix. To each subset of columns  $X \subseteq \mathcal{C}$ , we associate the subset  $R(X) \subseteq \mathcal{R}$  of all those rows  $r \in \mathcal{R}$  for which  $\mathcal{D}(r, c) = 1$  in all columns  $c \in X$ . The cardinality of  $R(X)$  is called the *support* of  $X$  and is easily seen to be an anti-monotone supermodular function of  $X$ . Hence

$$f(X) = m - |R(X)| \quad (15)$$

is a polymatroid function of range  $m$ . In fact, the above definition is identical to that given in (13) if we let  $\mathcal{H}$  to be the hypergraph defined by the anti-incidence matrix  $\mathcal{D}$ .

A column set  $X$  is called *t-frequent* if  $|R(X)| \geq t$  and *maximal t-frequent* if no superset of  $X$  is *t-frequent*. The generation of (maximal) frequent sets of a given binary matrix is an important task of knowledge discovery and data mining (see, for instance, [1]). Let  $\mathcal{F}_t(\mathcal{D})$  be the family of all maximal frequent sets for  $\mathcal{D}$ , then  $\mathcal{F}_t(\mathcal{D})$  is exactly the hypergraph  $\mathcal{A}_{m-t+1}(f)$  for the submodular function  $f$  defined by (15). Denoting by  $\mathcal{I}_t(\mathcal{D}) = \mathcal{B}_{m-t+1}(f)$  the family of all minimal columns sets  $X$  for which  $|R(X)| < t$ , i.e., all minimal *t-infrequent sets*, we can conclude that

$$|\mathcal{F}_t(\mathcal{D})| \leq (m - t + 1)|\mathcal{I}_t(\mathcal{D})|, \quad t \in \{1, \dots, m\}. \quad (16)$$

As for (14), these inequalities are best possible. For instance, they are sharp when  $\mathcal{D}$  is an  $m \times (m - t + 1)$  matrix, in which every entry is 1, except the diagonal entries in the first  $m - t + 1$  rows, which are 0. In addition, (16) stays accurate, up to a factor of  $\log m$ , even when  $m \gg n$  and  $|\mathcal{F}_t|$  and  $|\mathcal{I}_t|$  are arbitrarily large, see [11] for more detail. It is also worth mentioning that  $|\mathcal{F}_t|$  cannot be bounded by a quasi-polynomial in  $|\mathcal{I}_t|$  and  $n$ , the number of columns of  $\mathcal{D}$ .

## 2.3 Weighted transversals

Extending the notion of partial *t*-transversals to weighted hypergraphs we arrive at weighted transversals [12]. Given a hypergraph  $\mathcal{H} \subseteq 2^V$ , we assign an  $m$ -dimensional non-negative integral weight vector  $w = w(H) \in \mathbb{Z}_+^m$  to each hyperedge  $H \in \mathcal{H}$  and consider the system of  $m$  polymatroid inequalities

$$f(X) = \sum \{w(H) \mid X \cap H \neq \emptyset, H \in \mathcal{H}\} \geq t \quad (17)$$

where  $t \in \mathbb{Z}_+^m$  is a given threshold vector. The minimal solutions  $X \subseteq V$  to the above system are called *weighted transversals* for  $(\mathcal{H}, w)$ . Let  $\mathcal{B}$  be the set of all weighted transversals for (17) and let  $\mathcal{A} = \mathcal{I}(\mathcal{B})$  denote the hypergraph of all maximal infeasible sets for (17). Considering that each weighted hypergraph  $\mathcal{H}$  can be regarded as a collection of  $m$  multi-hypergraphs  $(\mathcal{H}_1, \dots, \mathcal{H}_m)$ , where multi-hypergraph  $\mathcal{H}_i$  contains  $w_i(H)$  copies of edge  $H \in \mathcal{H}$ , we conclude from (14) that  $|\mathcal{A}| \leq (t_1 + \dots + t_m)|\mathcal{B}|$ . This bound depends on the threshold vector and can be arbitrarily large with respect to  $n$  and  $m$  when the range of (17) becomes high. However, it can be shown [12] that regardless of the weights used in the definition of the hypergraph of weighted transversals, for any non-empty hypergraph  $\mathcal{X} \subseteq \mathcal{B}$  we have the inequality

$$|\mathcal{I}(\mathcal{X}) \cap \mathcal{I}(\mathcal{B})| \leq m \sum_{X \in \mathcal{X}} |\{H \in \mathcal{H} \mid H \cap X \neq \emptyset\}| \quad (18)$$

where as before  $\mathcal{I}(\mathcal{X})$  denotes the family of all maximal independent sets for  $\mathcal{X}$ . In particular, it follows that  $|\mathcal{I}(\mathcal{X}) \cap \mathcal{I}(\mathcal{B})| \leq m|\mathcal{H}||\mathcal{X}|$ , which for  $\mathcal{X} = \mathcal{B}$  gives

$$|\mathcal{I}(\mathcal{B})| \leq m|\mathcal{H}||\mathcal{B}|. \quad (19)$$

As earlier, Theorem 3 implies an incremental quasi-polynomial algorithm for generating all minimal weighted transversals for a given weighted hypergraph  $(\mathcal{H}, w)$  and threshold vector  $t$ . Generating all maximal infeasible vectors to (17) is NP-hard already for scalar unit weights  $w(H) \equiv 1$ .

#### 2.4 Monotone systems of Boolean and integer inequalities

This example can be viewed as a special case of weighted transversals. Consider a system of  $m$  linear inequalities in  $n$  binary variables

$$Ax \geq b, \quad x \in \{0, 1\}^n \quad (20)$$

where  $A$  is a given non-negative  $m \times n$ -matrix and  $b$  is a given  $m$ -vector. Let  $\mathcal{C}$  be the set of columns of  $A$ , and let  $\mathcal{H}$  be the hypergraph on the vertex set  $\mathcal{C}$  whose hyperedges are the  $n$  singletons (columns). We have  $|\mathcal{H}| = n$  and each column of the  $m \times n$  matrix  $A$  can now be interpreted as a non-negative  $m$ -dimensional weight vector assigned to the corresponding hyperedge of  $\mathcal{H}$ . Under this interpretation, (20) is a special case of (17) and we can identify the hypergraph  $\mathcal{B}$  of weighted transversals with the set of all minimal Boolean solutions to (20). Accordingly, the hypergraph  $\mathcal{A} = \mathcal{I}(\mathcal{B})$  can be viewed as the set of all maximal infeasible vectors for (20). From (18) we now conclude that for any non-empty set  $\mathcal{X} \subseteq \mathcal{B}$  of minimal feasible solutions to (20) we have

the inequalities

$$|\mathcal{I}(\mathcal{X}) \cap \mathcal{I}(\mathcal{B})| \leq m \sum_{x \in \mathcal{X}} p(x) \leq nm|\mathcal{X}|, \quad (21)$$

where  $p(x)$  is number of positive components of  $x$ . In particular, for any feasible system (20) we obtain

$$|\mathcal{I}(\mathcal{B})| \leq nm|\mathcal{B}|. \quad (22)$$

The above bounds are sharp when  $m = 1$ , for instance, for the inequality  $x_1 + \dots + x_n \geq n$ . For large  $m$ , these bounds are accurate up to a factor polylogarithmic in  $m$ . For instance, for any positive integer  $k$ , the system of  $m = 2^k$  inequalities in  $n = 2k$  binary variables

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

has  $2^k$  maximal infeasible binary vectors and only  $k$  minimal feasible binary vectors, i.e.,

$$|\mathcal{I}(\mathcal{B})| = \frac{nm}{2(\log m)^2} |\mathcal{B}|.$$

As shown in [9,10], inequalities (21) and (22) actually hold for any *monotone* system of inequalities (20) in binary variables, i.e., under the assumption that for any feasible vector  $x \in \{0, 1\}^n$ , any binary vectors  $y \geq x$  is also feasible for (20). (For instance, (20) is monotone if the matrix  $A$  is non-negative.) In fact, inequalities (21) and (22) also hold for any monotone system of  $m$  linear inequalities in  $n$  *integer* variables

$$Ax \geq b, \quad x \in \mathcal{C} \stackrel{\text{def}}{=} \{x \in \mathbb{Z}^n \mid 0 \leq x \leq c\}, \quad (23)$$

where  $c$  is a given  $n$  vector some or all components of which may be infinite. All minimal feasible integer solutions to a given monotone system of integer inequalities (23) can be generated in incremental quasi-polynomial time [8,12], which should be contrasted with the conjecture of E. Lawler, Lenstra and Rinnooy Kan [23] that for non-negative  $A$  and  $d = (1, \dots, 1)$  this problem cannot be solved in incremental polynomial time unless  $P=NP$ . On the other hand, the problem of generating all maximal infeasible binary vectors for (20) is NP-hard already for binary matrices  $A$  and  $b = (2, 2, \dots, 2)$ , see [25] and also [9] for more detail.

### 2.5 Spanning a linear space by linear subspaces.

The transversal hypergraph problem is equivalent to the following set covering problem: Given an  $r$ -element ground set  $\mathcal{R}$  and a family  $\mathcal{V}$  of  $n$  subsets of  $\mathcal{R}$ , enumerate all minimal subfamilies of  $\mathcal{V}$  which cover the entire set  $\mathcal{R}$ . Replacing

$\mathcal{R}$  by the vector space  $\mathbf{F}^r$  over some field  $\mathbf{F}$ , and replacing each given subset of  $\mathcal{V}$  by a linear subspace of  $\mathbf{F}^r$ , we arrive at the following *space covering* problem:

*Given a collection  $\mathcal{V} = \{V_1, \dots, V_n\}$  of  $n$  linear subspaces of  $\mathbf{F}^r$ , enumerate all minimal subsets  $X$  of  $V = \{1, \dots, n\}$  such that  $\text{Span}\langle \bigcup_{i \in X} V_i \rangle = \mathbf{F}^r$ .*

Generalizing further, consider the polymatroid inequality

$$f(X) = \dim\left(\bigcup_{i \in X} V_i\right) \geq t, \quad (24)$$

where  $t \in \{1, \dots, r\}$  is a given threshold. Note that when each subspace  $V_i$  is spanned by a subset  $R_i$  of vectors from some fixed basis of  $\mathbf{F}^r$ , the value of  $f(X)$  is just the size of  $\bigcup_{i \in X} R_i$ , and hence (24) includes the class of polymatroid inequalities discussed in Sections 2.1 and 2.2. This shows that generating all maximal infeasible sets  $\mathcal{A}_t = \mathcal{A}_t(f)$  for (24) is NP-hard.

Let  $\mathcal{B}_t = \mathcal{B}_t(f)$  be the collection of all minimal solutions  $X \subseteq V$  to (24). When  $t = r$  and the given subspaces  $V_i$  are all lines, i.e.,

$$V_i = \text{Span}\langle b_i \rangle \text{ for given vectors } b_i \in \mathbf{F}^r, \quad i = 1, \dots, n,$$

the set  $\mathcal{B}_t$  can be identified with the collection of all column bases of the  $r \times n$  matrix  $[b_1, \dots, b_n]$ . Accordingly, we then have  $|\mathcal{A}_r| \leq r|\mathcal{B}_r|$ , where  $\mathcal{A}_r = \mathcal{I}(\mathcal{B}_r)$  is the collection of all hyperplanes of the vectorial matroid  $M = \{b_1, \dots, b_n\}$ . It is an open problem whether there exists an efficient randomized algorithm for nearly uniform sampling from within the set of bases of a given matrix. However, for any matroid  $M$  defined by a polynomial-time independence oracle, all bases of  $M$  can be enumerated in incremental polynomial time by using depth-first search on  $\mathcal{B}_r$  with the obvious adjacency relation: two bases  $X, X' \in \mathcal{B}_r$  are adjacent if  $|X \cap X'| \geq r - 1$ , where  $r$  is the rank of  $M$ . (All hyperplanes of  $M$  can also be enumerated in incremental polynomial time by using the characteristic property of hyperplanes:

(\*) If  $H$  and  $H'$  are two distinct hyperplanes in  $M$  and  $b \notin H \cup H'$  then there exists a hyperplane  $H''$  such that  $b \cup (H \cap H') \subseteq H''$ .

If  $\mathcal{H} = \{H_1, \dots, H_k\}$  is an incomplete collection of hyperplanes closed with respect to (\*) then  $\mathcal{H}$  defines a matroid  $M'$  of rank  $r' < r$ , which can be checked by constructing a base for  $M'$ , i.e., a minimal transversal to the collection of co-circuits  $M \setminus H_1, \dots, M \setminus H_k$ .)

It is easy to show that if the dimension of each input subspace  $V_i$ ,  $i = 1, \dots, n$ , is bounded by some constant  $d$ , then

$$|\mathcal{A}_r| \leq rn^{d-1}|\mathcal{B}_r|, \quad (25)$$

i.e., the size of  $\mathcal{A}_r$  can still be bounded by a polynomial in  $n$  and the size of  $\mathcal{B}_r$ . (The bound of (25) follows from the fact that each element  $X \in \mathcal{A}_r$  can be completely defined by a minimal subset  $Y \subseteq X$  that spans the same subspace as  $X$ , and that  $Y$  can be transformed into an element of  $\mathcal{B}_r$  by adding one and then deleting at most  $d - 1$  of the input subspaces  $V_i$ .) Theorems 1 and 3 state that for all  $t \in \{1, \dots, r\}$ , the size of  $\mathcal{A}_t$  can be bounded by a  $\log t$ -degree polynomial in  $n$  and  $|\mathcal{B}_t|$ , and that all sets in  $\mathcal{B}_t$  can be enumerated in incremental quasi-polynomial time, regardless of the dimensions of the input subspaces  $V_i$ .

It is interesting to mention that even though the space covering problem can be solved in incremental quasi-polynomial time for any input subspaces  $V_1, \dots, V_n$ , the following close modification of the problem is NP-hard: Enumerate all minimal subsets  $X \subseteq V$  such that  $\text{Span}\langle \bigcup_{i \in X} V_i \rangle$  contains a given linear subspace  $V_0$ . In fact, the above *subspace* covering problem is NP-hard even when  $V_0$  is a line and  $\dim(V_i) = 2$  for all  $i = 1, \dots, n$ . If all the input subspaces are lines, i.e.,  $V_i = \text{Span}\langle b_i \rangle$ ,  $i = 0, 1, \dots, n$ , the problem calls for enumerating all minimal dependent sets ( $\equiv$  circuits) containing  $b_0$  in the vectorial matroid  $M = \{b_0, b_1, \dots, b_n\}$ . The latter problem can be done in incremental polynomial time for any matroid  $M$  defined by a polynomial-time independence oracle. (First, we can enumerate all circuits of  $M$  in incremental polynomial time because any circuit of  $M$  is a co-circuit of the dual matroid  $M^*$  and vice versa. Second, assuming without loss of generality that  $M$  is connected, by Lehman's theorem ([33], p.74) any circuit in  $M$  can be expressed via two circuits containing  $b_0$ , and hence  $(\# \text{ circuits of } M) \leq (\# \text{ circuits through } b_0)^2$ . More efficient cycle generation algorithm for the special case of graphs can be found in [30].)

It is also worth mentioning the NP-hardness of the *conic* variant of the space covering problem: Given a collection of  $r$ -dimensional polyhedral cones  $K_1, \dots, K_n$  defined by their rational generators, enumerate all minimal sets  $X \subseteq \{1, \dots, n\}$  such that  $\text{Cone}\langle \bigcup_{i \in X} K_i \rangle$  spans the entire space. The following problem is also NP-hard [22]: Given a rational  $r$ -vector  $b$  and a collection of  $n$  dihedral cones  $K_i = \text{Cone}\langle a_i, a'_i \rangle$ ,  $i = 1, \dots, n$ , enumerate all minimal sets  $X \subseteq \{1, \dots, n\}$  for which  $b \in \text{Cone}\langle \bigcup_{i \in X} K_i \rangle$ . Replacing the input dihedral cones by rays we obtain the *vertex enumeration* problem: Enumerate all vertices of a given polyhedron  $P = \{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$ . The incremental complexity of the vertex enumeration problem is not known. In particular, it is not known whether there exists a quasi-polynomial-time algorithm which, given a rational polytope  $P = \{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$  and a collection  $Q$  of the vertices of  $P$ , can determine if  $P = \text{Conv.hull}\langle Q \rangle$ .

## 2.6 Spanning collections of graphs.

Let  $R$  be a finite set of  $r$  vertices and let  $E_1, \dots, E_n \subseteq R \times R$  be a collection of  $n$  graphs on  $R$ . Given a set  $X \subseteq \{1, \dots, n\}$  define  $k(X)$  to be the number of connected components in the graph  $(R, \bigcup_{i \in X} E_i)$ . Then  $k(X)$  is an anti-monotone supermodular function and hence for any integral threshold  $t$ , the inequality

$$f(X) = r - k(X) \geq t$$

is polymatroid. In particular,  $\mathcal{B}_{r-t}$  is the family of all minimal collections of the input graphs  $E_1, \dots, E_n$  which interconnect all vertices in  $R$ . (If the  $n$  input graphs are just  $n$  disjoint edges, then  $\mathcal{B}_{r-t}$  is the set of all spanning trees in the graph  $E_1 \cup \dots \cup E_n$ .) Since  $k(X)$  can be evaluated at any set  $X$  in polynomial time, Theorem 3 implies that for each  $t \in \{1, \dots, r\}$ , all elements of  $\mathcal{B}_t$  can be enumerated in incremental quasi-polynomial time. In particular, given a collection of  $n$  *equivalence relations* (partitions) on  $R$ , we can enumerate in incremental quasi-polynomial time all minimal subsets of the given relations whose transitive closure puts all elements of  $R$  in one equivalence class (or produces at most  $r - t$  equivalence classes).

Interestingly, enumerating all minimal collections of  $E_1, \dots, E_n$  connecting two distinguished vertices  $s, t \in R$  turns out to be NP-hard even if the input sets  $E_1, \dots, E_n$  are disjoint and contain at most 2 edges each, see [17]. Needless to say that as before, generating all maximal collections of  $E_1, \dots, E_n$  for which the number of connected components of  $(R, \bigcup_{i \in X} E_i)$  exceeds a given threshold remains NP-hard.

## 2.7 Matroid Intersections

Let  $M_1, \dots, M_m$  be  $m$  matroids defined on a common ground set  $V$  by  $m$  polynomial-time independence oracles. Lawler, Lenstra and Rinnooy Kan [23] considered the following *matroid intersection problem*: incrementally generate all maximal subsets  $Y \subseteq V$  independent for all matroids  $M_1, \dots, M_m$ . Let  $\rho_1, \dots, \rho_m : 2^V \mapsto \{0, 1, \dots, |V|\}$  be the rank functions of  $M_1, \dots, M_m$ , then the matroid intersection problem calls for the enumeration of all maximal feasible solutions to the system of  $m$  inequalities

$$\rho_i(Y) \geq |Y|, \quad i = 1, \dots, m, \quad Y \subseteq V. \quad (26)$$

Letting  $X = V \setminus Y$ , (26) can equivalently be stated as follows: enumerate all minimal solutions  $X$  to the system of polymatroid inequalities

$$f_i(X) \stackrel{\text{def}}{=} |X| + \rho_i(V \setminus X) - \rho_i(V) \geq |V| - \rho_i(V), \quad i = 1, \dots, m, \quad Y \subseteq V. \quad (27)$$

Since the polymatroid functions  $f_i(X)$  can be evaluated at any set  $X$  in polynomial time, Theorem 3 implies that the matroid intersection problem can be solved in incremental quasi-polynomial time. This substantially improves the algorithm suggested in [23], whose running time is exponential in  $m$ .

Several other examples of polymatroid functions can be found, for instance, in Lovász [24] and Welsh [33]. Let  $M$  be a matroid on a ground set  $U$  with the rank function  $\rho : 2^U \mapsto \{0, 1, \dots\}$ , and let  $U_1, \dots, U_n$  be some subsets of  $U$ . For each  $X \subseteq V \stackrel{\text{def}}{=} \{1, \dots, n\}$ , let  $f(X) = \rho(\cup_{i \in X} U_i)$ . Then  $f$  is a polymatroid function. In fact, every polymatroid function arises by this construction from some matroid, see [18,27], and also [24].

### 3 A lower bound for $|\mathcal{A}_t|$

In this section we demonstrate that inequality (2) of Theorem 1 is reasonably tight.

#### 3.1 A hypergraph example

In our first example, let  $\mathcal{H}$  be the edge set of the graph  $G = k \times K_l$  consisting of  $k$  pairwise disjoint copies of a clique on  $l$  vertices. In this (hyper)graph, the number of vertices is  $n = |V| = kl$ , the number of (hyper)edges is  $|\mathcal{H}| = k \binom{l}{2}$ , the number of maximal independent sets is  $|\mathcal{I}(\mathcal{H})| = l^k$ , and we can prove the following statement.

**Lemma 3** *For the hypergraph  $\mathcal{H}$  defined above, there exists an exact polymatroid separator  $(f, t)$ , such that  $t = r(f) = 2^k$ .*

**Proof.** For  $X \subseteq V$ , define  $f(X)$  by

$$f(X) = \begin{cases} 2^k, & \text{if } X \text{ contains an edge of } \mathcal{H} \\ 2^k - 2^{\gamma(X)}, & \text{otherwise,} \end{cases}$$

where  $\gamma(X)$  denotes the number of  $l$ -cliques of  $G$  disjoint from  $X$ . In particular,  $f(X) = 2^k - 1$  if (and only if)  $X$  is a maximal independent set of  $\mathcal{H}$ , and  $f(X) = 2^k$  if  $X$  contains an edge of  $\mathcal{H}$ . Let us also note that  $f$  is obviously monotone, by the above definition. Thus, with  $t = 2^k$  the pair  $(f, t)$  is indeed an exact separator of  $\mathcal{H}$ .

It remains to show that  $f$  is submodular. For this, let  $X$  and  $Y$  be two arbitrary subsets of the vertex set  $V$ . If both  $X$  and  $Y$  contain an edge of  $\mathcal{H}$ , then (1) holds trivially, since we have  $2^{k+1}$  on the right hand side, and we have  $f(Z) \leq 2^k$  for all subsets  $Z \subseteq V$  by definition. Furthermore, if one of these sets contains an edge of  $\mathcal{H}$ , say  $X$ , then of course  $X \cup Y$  does too, and hence (1) reduces to  $f(X \cap Y) \leq f(Y)$  which holds again trivially by the monotonicity of  $f$ . Let us assume next that neither  $X$  nor  $Y$  contain an edge of  $\mathcal{H}$ , but  $X \cup Y$  does. In this case  $\gamma(X \cap Y) \geq 1 + \max(\gamma(X), \gamma(Y))$  holds, implying  $2^{\gamma(X \cap Y)} \geq 2^{\gamma(X)} + 2^{\gamma(Y)}$ , from which (1) follows. Let us assume finally that  $X \cup Y$  does not contain an edge of  $\mathcal{H}$ . In this case  $\gamma(X \cap Y) \geq \max(\gamma(X), \gamma(Y))$  and  $\gamma(X \cap Y) + \gamma(X \cup Y) \geq \gamma(X) + \gamma(Y)$  both hold, implying  $2^{\gamma(X \cap Y)} + 2^{\gamma(X \cup Y)} \geq 2^{\gamma(X)} + 2^{\gamma(Y)}$ , from which (1) follows again.  $\square$

The above lemma now implies that with  $t = r(f) = 2^k$  we have  $\beta = |\mathcal{H}| = k \binom{l}{2}$ ,  $\alpha = l^k$  and  $n = |V| = kl$ . For  $l = 2^k$  and  $k \rightarrow \infty$ , we thus obtain  $\log n / \log \beta \rightarrow 1/2$  and hence  $c(n, \beta)$  converges to the root of the equation  $2^c(2^{c/2} - 1) = 1$ . This gives  $c = 1.102\dots$ , and consequently

$$\alpha > \beta^{(.551 \log t)/c(n, \beta)}$$

for  $k$  sufficiently large.

### 3.2 A rank function example

Let us next show that the polymatroid function  $f$  defined above can be realized as the rank function of some linear subspaces of the vector space  $\mathbf{F}^r$ ,  $r \in \mathbb{Z}_+$  over a (possibly large) field  $\mathbf{F}$ .

For a positive integer  $l$  let  $\mathbf{F}$  be a field with  $l \leq |\mathbf{F}|$  (we shall use, as customary,  $+$  and  $\times$  to denote the two field operations, and we write  $0$  and  $1$  for the unit elements of these operations, respectively). Furthermore, let  $n = kl$ , let  $G = k \times K_l$  be the graph, as above, and let  $\mathcal{H}$  be again the edge set of  $G$ . Let us introduce the notations  $K = \{1, 2, \dots, k\}$  and  $L = \{0, 1, \dots, l-1\}$ , and let us denote the vertex set of  $G$  by  $V = K \times L$ . We shall associate to each vertex  $(i, j) \in V$  a linear subspace  $V_{ij}$  of  $\mathbf{F}^r$ , where  $r = 2^k$ . These subspaces will be chosen in such a way that every two subspaces corresponding to the same clique of  $G$  intersect only in the origin (and hence generate the whole space  $\mathbf{F}^r$ ), while the intersection of arbitrary  $s$  subspaces ( $1 \leq s \leq k$ ), each corresponding to distinct cliques of  $G$ , is of dimension  $2^{k-s}$ .

Let  $\{b_x \mid x \in \{0, 1\}^k\}$  be an arbitrary basis in  $\mathbf{F}^r$ , indexed by the  $r = 2^k$  elements of the binary cube of dimension  $k$ , and let  $\lambda_0 = 0, \lambda_1 = 1, \lambda_2 \dots, \lambda_{l-1}$

be distinct elements of  $\mathbf{F}$  (hence the requirement  $l \leq |\mathbf{F}|$ ). For every  $z \in L^k$ , and every index vector  $x \in \{0, 1\}^k$ , define the product

$$\Lambda_z(x) = \prod_{i:x_i=1} \lambda_{z_i} \prod_{i:x_i=0} (1 - \lambda_{z_i}).$$

It is easily verified that  $\sum_{x \in \{0,1\}^k} \Lambda_z(x) = 1$  for all  $z \in L^k$ , and that for any two binary vectors  $x, y \in \{0, 1\}^k$ , we have  $\Lambda_x(y) = 1$  if  $x = y$ , and  $\Lambda_x(y) = 0$  otherwise. Let us now associate a (unique) vector

$$b_z = \sum_{x \in \{0,1\}^k} \Lambda_z(x) b_x \quad (28)$$

of  $\mathbf{F}^r$  to every  $z \in L^k$ . Observe that for  $z = x \in \{0, 1\}^k$ , we get a basis element  $b_z = b_x$  by our selection of  $\lambda_0 = 0, \lambda_1 = 1$ .

Let us next define the linear subspace  $V_{ij}$ , for  $(i, j) \in V$ , to be the subspace generated by the vectors  $b_z \in \mathbf{F}^r$  whose index vector  $z$  has value  $j$  in its  $i$ -th coordinate:

$$V_{ij} = \langle b_z \mid z_i = j \rangle.$$

We will show below that this construction has the announced properties. To simplify notation, we shall need a few more definitions.

For index vectors  $x, y \in L^k$ , and a subset  $S \subseteq K$ , we denote by  $x[S]$  the restriction of  $x$  to  $S$ , by  $y[\overline{S}]$  the restriction of  $y$  to the complementary set  $\overline{S} = K \setminus S$ , and by  $z = x[S], y[\overline{S}]$  the vector defined by

$$z_j = \begin{cases} x_j & \text{if } j \in S, \\ y_j & \text{if } j \notin S. \end{cases}$$

Let us note that  $\Lambda_{a[S], b[\overline{S}]}(x[S], y[\overline{S}]) = \Lambda_{a[S]}(x[S]) \Lambda_{b[\overline{S}]}(y[\overline{S}])$  holds by the above definitions, for all  $a, b \in L^k$ ,  $x, y \in \{0, 1\}^k$  and  $S \subseteq K$ .

**Lemma 4** *For any  $S \subseteq K$  and  $w = (j_i \mid i \in S) \in L^S$ , the set of vectors*

$$\left\{ \sum_{y \in \{0,1\}^S} \Lambda_w(y) b_{y,x} \mid x \in \{0, 1\}^{\overline{S}} \right\} \quad (29)$$

*forms a basis for the vector space  $V_{S,w} \stackrel{\text{def}}{=} \langle b_z \mid z[S] = w \rangle$ . In particular,  $\dim(V_{S,w}) = 2^{k-|S|}$ .*

**Proof.** First, let us observe that for each  $x \in \{0, 1\}^{\overline{S}}$ , the vector  $a = \sum_{y \in \{0,1\}^S} \Lambda_w(y) b_{y,x}$  lies in the space  $V_{S,w}$  since  $a = b_z$  with  $z[S] = w$  and  $z[\overline{S}] = x$ . Let

us observe next that these vectors are linearly independent: suppose, on the contrary, that there exist scalars  $\mu_x \in \mathbf{F}$ ,  $x \in \{0, 1\}^{\bar{S}}$ , not all zero, such that

$$0 = \sum_{x \in \{0, 1\}^{\bar{S}}} \mu_x \left( \sum_{y \in \{0, 1\}^S} \Lambda_w(y) b_{y,x} \right).$$

Then, by the linear independence of the basis  $\{b_x \mid x \in \{0, 1\}^k\}$  of  $\mathbf{F}^r$ , we obtain that

$$\mu_x \Lambda_w(y) = 0, \quad \text{for all } x \in \{0, 1\}^{\bar{S}} \text{ and } y \in \{0, 1\}^S. \quad (30)$$

But summing up equations (30) for a particular  $x \in \{0, 1\}^{\bar{S}}$  over all  $y \in \{0, 1\}^S$ , and using  $\sum_{y \in \{0, 1\}^S} \Lambda_w(y) = 1$ , we get  $\mu_x = 0$ , for all  $x \in \{0, 1\}^{\bar{S}}$ , a contradiction, proving that (29) is indeed a family of linearly independent vectors. Let us note finally that these vectors span the entire subspace  $V_{S,w}$ , since any vector  $b_z$  with  $z[S] = w$  in this subspace can be written as:

$$\begin{aligned} b_z &= \sum_{u \in \{0, 1\}^k} \Lambda_z(u) b_u \\ &= \sum_{u \in \{0, 1\}^k} \Lambda_{z[\bar{S}]}(u[\bar{S}]) \Lambda_{z[S]}(u[S]) b_u \\ &= \sum_{x \in \{0, 1\}^{\bar{S}}} \Lambda_{z[\bar{S}]}(x) \left( \sum_{y \in \{0, 1\}^S} \Lambda_w(y) b_{y,x} \right). \end{aligned}$$

The lemma follows from the above observations.  $\square$

For  $z', z'' \in L^k$ , let us denote by  $[z', z'']$  the set of all those vectors  $z \in L^k$  for which  $z_i \in \{z'_i, z''_i\}$  for  $i = 1, \dots, k$ .

**Lemma 5** *Let  $z', z'' \in L^k$  be such that  $z'_i \neq z''_i$  for all  $i \in K$ . Then the set  $B_{z', z''} \stackrel{\text{def}}{=} \{b_z \mid z \in [z', z'']\}$  forms a basis for  $\mathbf{F}^r$ .*

**Proof.** Let  $M_{z', z''} \stackrel{\text{def}}{=} (\Lambda_z(x))_{x,z}$  be the  $2^k \times 2^k$ -matrix whose rows are indexed by the vectors  $x \in \{0, 1\}^k$ , and whose columns are indexed by the vectors  $z \in [z', z'']$ . To prove that the set  $B_{z', z''}$  is linearly independent, it is enough by (28) to show that the matrix  $M_{z', z''}$  is non-singular. Indeed, we claim that

$$|\det(M_{z', z''})| = \prod_{i=1}^k (\lambda_{z'_i} - \lambda_{z''_i})^{2^{k-1}}, \quad (31)$$

from which the lemma will follow by the distinctness of  $\lambda_0, \lambda_1, \dots, \lambda_{l-1}$ . To prove (31), we first observe that the left hand side is a polynomial in  $\mathbf{F}[\lambda_{z'_1}, \dots, \lambda_{z'_k}, \lambda_{z''_1}, \dots, \lambda_{z''_k}]$ , of degree  $2^{k-1}$  in each variable  $\lambda_{z_i}$ . Let  $i \in K$  and let  $u, v \in L^k$  be such that  $u_i = z'_i$ ,  $v_i = z''_i$ , and  $u[K \setminus \{i\}] = v[K \setminus \{i\}] = w \in L^{K \setminus \{i\}}$ . Then for any  $x \in \{0, 1\}^k$ , it is easy to see that  $\Lambda_u(x) - \Lambda_v(x) = (-1)^{x_i} (\lambda_{z'_i} - \lambda_{z''_i}) \Lambda_w(x[K \setminus \{i\}])$ . In particular, if we subtract the two columns of  $M_{z', z''}$

indexed by  $u, v$ , we obtain  $\lambda_{z'_i} - \lambda_{z''_i}$  as a factor for the determinant expression in (31). Repeating this argument for every  $i \in K$  and every  $w \in L^K \setminus \{i\}$ , we conclude that the right hand side of (31) is a divisor of the left hand side. Since both polynomials are of the same degree in all variables by our earlier observation, and since they attain the same value at, say,  $z' = (0, \dots, 0)$ ,  $z'' = (1, \dots, 1)$ , (31) follows.  $\square$

**Lemma 6** *For all  $i \in K$ , and for all  $j, j' \in L$ ,  $j \neq j'$ , the subspaces  $V_{ij}$  and  $V_{ij'}$  span the entire space  $\mathbf{F}^r$ , i.e.,  $\dim(V_{ij} \cup V_{ij'}) = 2^k$ .*

**Proof.** Let  $z', z'' \in L^k$  be such that  $z'_i \neq z''_i$  for all  $i \in K$ ,  $z'_i = j$ , and  $z''_i = j'$ . Since the basis set  $B_{z', z''}$  is contained in  $V_{ij} \cup V_{ij'}$ , the lemma follows.  $\square$

**Lemma 7** *Let  $z', z'' \in L^k$  be such that  $z'_i \neq z''_i$  for all  $i \in K$ . Then for any  $i \in K$ , we have  $V_{i, z'_i} = \langle b_z \mid z \in [z', z''], z_i = z'_i \rangle$ .*

**Proof.** From Lemma 4, we have  $\dim(V_{i, z'_i}) = 2^{k-1}$ , and from Lemma 5, the set  $\{b_z \mid z \in [z', z''], z_i = z'_i\}$  is linearly independent. Since this set is contained in  $V_{i, z'_i}$  by definition, the lemma follows.  $\square$

**Lemma 8** *For  $S \subseteq K$  and  $w = (j_i \mid i \in S) \in L^S$ , we have*

$$\dim\left(\bigcup_{i \in S} V_{i, j_i}\right) = 2^k - 2^{k-|S|}. \quad (32)$$

**Proof.** Fix  $z', z'' \in L^k$  such that  $z'_i \neq z''_i$  for all  $i \in K$ , and  $z'[S] = w$ , and let  $B = B_{z', z''}$  be the basis set defined by these two vectors. For  $i \in S$  let  $B^i \stackrel{\text{def}}{=} B \cap V_{i, j_i}$ , and let  $B^{S, w} = \bigcup_{i \in S} B^i$ . It is then immediate from the definitions and Lemma 7 that  $\bigcup_{i \in S} V_{i, j_i} = \langle B^{S, w} \rangle$ , and thus it is enough, by Lemma 5, to count the number of elements in the set  $B^{S, w}$ . Using the inclusion-exclusion formula, we obtain

$$\dim\left(\bigcup_{i \in S} V_{i, j_i}\right) = |B^{S, w}| = \sum_{\substack{Q \subseteq S \\ Q \neq \emptyset}} (-1)^{|Q|-1} \left| \bigcap_{i \in Q} B^i \right|.$$

Therefore, since  $\left| \bigcap_{i \in Q} B^i \right| = |\{b_z \in B_{z', z''} : z[Q] = w[Q]\}| = 2^{k-|Q|}$ , we get

$$\dim\left(\bigcup_{i \in S} V_{i, j_i}\right) = \sum_{\substack{Q \subseteq S \\ Q \neq \emptyset}} (-1)^{|Q|-1} 2^{k-|Q|} = \sum_{m=1}^{|S|} (-1)^{m-1} \binom{|S|}{m} 2^{k-m} = 2^k - 2^{k-|S|},$$

implying the Lemma.  $\square$

We are now ready to verify that our construction indeed has the desired prop-

erties. For a subset  $X \subseteq V = K \times L$  let us define

$$g(X) = \dim\left(\bigcup_{(i,j) \in X} V_{ij}\right),$$

and let us set  $t = 2^k$ . It follows by Lemma 6 that if  $X$  contains an edge of the graph  $G$ , then  $g(X) = 2^k$ , i.e., that  $\mathcal{B}_t(g) = \mathcal{H}$ . It also follows by Lemma 8 that  $g(X) = 2^k - 2^{k-|X|} \leq 2^k - 1$  for any independent set  $X \subset V$ . i.e.,  $\mathcal{A}_t(g) = \mathcal{I}(\mathcal{H})$ . In other words,  $g$  is the same set-function as the function  $f$  described in the previous subsection.

## 4 Proofs

In this section we prove Lemmas 1 and 2, which are the key statements needed to prove our main results.

**Proof of Lemma 1.** Let us recall that  $(f, t)$  is a polymatroid separator of the hypergraph  $\mathcal{H}$ , separating the maximal independent sets  $\mathcal{S} = \mathcal{S}(\mathcal{H}, f, t)$  from  $\mathcal{H}$ , and that to every node  $v$  of  $\mathbf{T}$  we have associated an independent set  $S_v = \bigcap_{l \in L(\mathbf{T}(v))} I_l$ , where  $I_l \in \mathcal{S}$  denotes the maximal independent set assigned to the leaf  $l \in L(\mathbf{T})$  by the proper assignment  $\phi$ .

To prove the statement of the lemma, we shall show by induction that

$$f(S_w) \leq t - |L(\mathbf{T}(w))| \tag{33}$$

holds for every node  $w$  of the binary tree  $\mathbf{T}$ . Since  $f$  is non-negative, it follows that

$$|L(\mathbf{T}(w))| \leq t \leq r(f)$$

which, if applied to the root of  $\mathbf{T}$ , proves the lemma. To see (33), let us apply induction by the size of  $L(\mathbf{T}(w))$ . Clearly, if  $w = l$  is a leaf of  $\mathbf{T}$ , then  $|L(\mathbf{T}(l))| = 1$ ,  $S_w = I_l \in \mathcal{S}$ , and (33) follows by the assumption that  $(f, t)$  is separating  $\mathcal{H}$  from  $\mathcal{S}$ . Let us assume now that  $w$  is a node of  $\mathbf{T}$  with  $u$  and  $v$  as its immediate successors. Then  $|L(\mathbf{T}(w))| = |L(\mathbf{T}(u))| + |L(\mathbf{T}(v))|$ , and  $S_w = S_u \cap S_v$ . By our inductive hypothesis, and since  $f$  is submodular, we have the inequalities

$$\begin{aligned} f(S_u \cup S_v) + f(S_w) &\leq f(S_u) + f(S_v) \\ &\leq t - |L(\mathbf{T}(u))| + t - |L(\mathbf{T}(v))| \\ &= 2t - |L(\mathbf{T}(w))|. \end{aligned}$$

Since  $\phi$  is a proper mapping, the set  $S_u \cup S_v$  contains a hyperedge  $H \in \mathcal{H}$ , and thus  $f(S_u \cup S_v) \geq f(H) \geq t$  by the monotonicity of  $f$ , and by our assumption that  $(f, t)$  is a separator for  $\mathcal{H}$ . Thus, from the above inequality we get  $t + f(S_w) \leq f(S_u \cup S_v) + f(S_w) \leq 2t - |L(\mathbf{T}(w))|$ , from which (33) follows.  $\square$

For a hypergraph  $\mathcal{H}$  and a vertex  $v \in V = V(\mathcal{H})$  let us denote by  $d_{\mathcal{H}}(v)$  the degree of vertex  $v$  in  $\mathcal{H}$ , i.e.,  $d_{\mathcal{H}}(v)$  is the number of hyperedges of  $\mathcal{H}$  containing  $v$ .

**Lemma 9** *For every Sperner hypergraph  $\mathcal{H} \subseteq 2^V$  on  $n = |V| > 1$  vertices, with  $m = |\mathcal{H}| \geq n$  hyperedges, there exists a vertex  $v \in V$  for which*

$$m \frac{1}{n} \leq d_{\mathcal{H}}(v) \leq m \left(1 - \frac{1}{n}\right).$$

**Proof.** Let us define

$$X = \{v \in V \mid d_{\mathcal{H}}(v) < m \frac{1}{n}\}$$

and

$$Y = \{v \in V \mid d_{\mathcal{H}}(v) > m(1 - \frac{1}{n})\},$$

and let us assume indirectly that  $X \cup Y = V$  forms a partition of the vertex set.

Let us observe first that  $|X| < n$  must hold, since otherwise a contradiction

$$m \leq \sum_{H \in \mathcal{H}} |H| = \sum_{v \in X} d_{\mathcal{H}}(v) < n \frac{m}{n} = m,$$

would follow.

Let us observe next that  $|X| > 0$  must hold, since otherwise

$$\sum_{H \in \mathcal{H}} |H| = \sum_{v \in V} d_{\mathcal{H}}(v) = \sum_{v \in Y} d_{\mathcal{H}}(v) > n \times m(1 - \frac{1}{n}) = m(n - 1)$$

follows, implying the existence of a hyperedge  $H \in \mathcal{H}$  of size  $|H| = n$ , i.e.,  $V \in \mathcal{H}$ . Since  $\mathcal{H}$  is Sperner,  $1 = m < n$  would follow, contradicting our assumptions.

Let us observe finally that the number of those hyperedges which avoid some points of  $Y$  cannot be more than  $|Y|m/n$ , and since  $|Y| < n$  by our previous observation, there must exist a hyperedge  $H \in \mathcal{H}$  containing  $Y$ . Thus, all

other hyperedges must intersect  $X$ , and hence we have

$$m - 1 \leq \sum_{H \in \mathcal{H}} |H \cap X| = \sum_{v \in X} d_{\mathcal{H}}(v) < |X| \frac{m}{n} \leq m \frac{n-1}{n}$$

by our first observation. From this  $m < n$  would follow, contradicting again our assumption that  $m \geq n$ . This last contradiction hence proves  $X$  and  $Y$  cannot cover  $V$ , and thus follows the lemma.  $\square$

For a subset  $X \subseteq V$  let  $\mathcal{H}^X \stackrel{\text{def}}{=} \{H \in \mathcal{H} \mid H \supseteq X\}$ , and let us simply write  $\mathcal{H}^v$  if  $X = \{v\}$ .

**Lemma 10** *Given a hypergraph  $\mathcal{H}$  and a subfamily  $\mathcal{S} \subseteq \mathcal{I}(\mathcal{H})$  of its maximal independent sets,  $|\mathcal{S}| \geq 2$ , there exists a hyperedge  $H \in \mathcal{H}$  and a vertex  $v \in H$  such that*

$$|\mathcal{S}^v| \geq \frac{|\mathcal{S}|}{n} \text{ and } |\mathcal{S}^{H \setminus v}| \geq \frac{|\mathcal{S}|}{n|\mathcal{H}|}.$$

**Proof.** Let us note first that if  $2 \leq |\mathcal{S}| < n$ , then the statement is almost trivially true. To see this, let us choose two distinct maximal independent sets  $S_1$  and  $S_2$  from  $\mathcal{S}$ , and a vertex  $v \in S_2 \setminus S_1$ . Since  $S_1 \cup \{v\}$  is not independent, there exists a hyperedge  $H \in \mathcal{H}$  for which  $v \in H \cap S_2$  and  $H \setminus \{v\} \subseteq S_1$ , implying thus that both  $|\mathcal{S}^v|$  and  $|\mathcal{S}^{H \setminus v}|$  are at least 1, and the right hand sides in the claimed inequalities are not more than 1.

Thus, we can assume in the sequel that  $|\mathcal{S}| \geq n$ . Let us then apply Lemma 9 for the Sperner hypergraph  $\mathcal{S}^c \stackrel{\text{def}}{=} \{V \setminus I \mid I \in \mathcal{S}\}$ , and obtain that

$$\frac{|\mathcal{S}|}{n} \leq d_{\mathcal{S}^c}(v) \leq |\mathcal{S}| \left(1 - \frac{1}{n}\right)$$

holds for some  $v \in V$ , since  $|\mathcal{S}| = |\mathcal{S}^c|$  obviously. Thus, from the second inequality we obtain

$$|\mathcal{S}^v| \geq \frac{|\mathcal{S}|}{n}.$$

To see the second inequality of Lemma 10, let us note that members of  $\mathcal{S}^c$  are minimal transversals of  $\mathcal{H}$ , and thus for every  $T \in \mathcal{S}^c$ ,  $T \ni v$  there exists a hyperedge  $H \in \mathcal{H}$  for which  $H \cap T = \{v\}$ , by the definition of minimal transversals. Thus,

$$\bigcup_{H \in \mathcal{H}: H \ni v} \{T \in \mathcal{S}^c \mid T \cap H = \{v\}\} \supseteq \{T \in \mathcal{S}^c \mid T \ni v\}$$

holds, from which

$$\sum_{H \in \mathcal{H}: H \ni v} |\mathcal{S}^{H \setminus v}| \geq d_{\mathcal{S}^c}(v) \geq \frac{|\mathcal{S}|}{n}$$

follows. Therefore, since  $|\{H \in \mathcal{H} \mid H \ni v\}| = d_{\mathcal{H}}(v) \leq |\mathcal{H}|$  holds obviously, there must exist a hyperedge  $H \in \mathcal{H}$ ,  $H \ni v$ , for which

$$|\mathcal{S}^{H \setminus v}| \geq \frac{|\mathcal{S}|}{n|\mathcal{H}|}$$

holds, implying thus the lemma.  $\square$

**Proof of Lemma 2.** Let us denote by  $L(\alpha)$  the maximum number of leaves of a binary tree  $\mathbf{T}$  with a proper mapping  $\phi : V(\mathbf{T}) \rightarrow \mathcal{S}$ , where  $\mathcal{S} \subseteq \mathcal{I}(\mathcal{H})$  is an arbitrary subfamily of maximal independent sets of  $\mathcal{H}$ . To simplify notation, let us write  $\alpha = |\mathcal{S}|$  and  $\beta = |\mathcal{H}|$ . To prove the statement, we need to show that

$$L(\alpha) \geq \alpha^{c/\log \beta} \tag{34}$$

where  $c = c(n, \beta)$  is as defined in (3).

Let us prove this inequality by induction on  $\alpha$ . Clearly, if  $\alpha = 1$ , then  $L(1) = 1$  holds, and we have equality in (34).

Let us assume next that we already have verified the claim for all subfamilies of size smaller than  $\alpha$ , and let us consider a subfamily  $\mathcal{S} \subseteq \mathcal{I}(\mathcal{H})$  of size  $\alpha = |\mathcal{S}|$ . According to Lemma 10, we can choose two disjoint subfamilies  $\mathcal{S}', \mathcal{S}'' \subseteq \mathcal{S}$  such that  $|\mathcal{S}'| \geq \frac{\alpha}{n}$  and  $|\mathcal{S}''| \geq \frac{\alpha}{n\beta}$ , and such that for any pair of sets  $S' \in \mathcal{S}'$  and  $S'' \in \mathcal{S}''$  the union  $S' \cup S''$  contains a member of  $\mathcal{H}$ . Thus, building binary trees with proper mappings separately for  $\mathcal{S}'$  and  $\mathcal{S}''$ , and joining them as two siblings of a common root, we obtain a binary tree with a proper mapping for  $\mathcal{S}$ . Since the right hand side of our claim is a monotone function of  $\alpha$ , we can conclude for the number of leaves in the obtained binary tree that

$$L(\alpha) \geq L\left(\frac{\alpha}{n}\right) + L\left(\frac{\alpha}{n\beta}\right). \tag{35}$$

Applying now our inductive hypothesis, we get

$$\begin{aligned} L(\alpha) &\geq \left(\frac{\alpha}{n}\right)^{c/\log \beta} + \left(\frac{\alpha}{n\beta}\right)^{c/\log \beta} \\ &= \alpha^{c/\log \beta} \left[ n^{-c/\log \beta} + (n\beta)^{-c/\log \beta} \right] \\ &= \alpha^{c/\log \beta}, \end{aligned}$$

where the last equality holds by (3). This proves (34), and hence the lemma follows.  $\square$

Note that the right hand side of (34) is the least possible solution of the recursion (35).

## 5 Generating minimal integer solutions of a system of polymatroid inequalities in an integral box

In this section, we discuss a generalization of the previous results which replaces polymatroid set-functions by polymatroid functions defined on integral boxes. Let  $\mathcal{C} \stackrel{\text{def}}{=} \{x \in \mathbb{Z}^n \mid 0 \leq x \leq c\}$  be an integral box, where  $c \in \mathbb{Z}_+^n$  is a given finite  $n$ -vector. A function  $f : \mathcal{C} \mapsto \{0, 1, \dots, r\}$ , where  $r \in \mathbb{Z}_+$ , is said to be submodular if

$$f(x \vee y) + f(x \wedge y) \leq f(x) + f(y)$$

holds for all  $x, y \in \mathcal{C}$ , where  $\vee$  and  $\wedge$  denote, as usual, the join and meet operators over  $\mathcal{C}$ :

$$\begin{aligned} x \vee y &= (\max\{x_1, y_1\}, \dots, \max\{x_n, y_n\}), \\ x \wedge y &= (\min\{x_1, y_1\}, \dots, \min\{x_n, y_n\}). \end{aligned}$$

As before,  $f$  is said to be monotone if  $f(x) \leq f(y)$  whenever  $x \leq y$ , and is called polymatroid if it is monotone, submodular, and  $f(0) = 0$ .

Given a polymatroid function  $f$  with range  $r$  and an integral threshold  $t \in \{1, \dots, r\}$  let us denote by  $\mathcal{B}_t = \mathcal{B}_t(f)$  the family of all minimal vectors  $x \in \mathcal{C}$  for which  $f(x) \geq t$ , and by  $\mathcal{A}_t = \mathcal{A}_t(f)$  the family of all maximal vectors  $x \in \mathcal{C}$  for which  $f(x) < t$ . It follows that  $\mathcal{A}_t = \mathcal{I}(\mathcal{B}_t)$ , where  $\mathcal{I}(\mathcal{B}) \stackrel{\text{def}}{=} \{\text{maximal } x \in \mathcal{C} \mid x \not\geq b, \text{ for all } b \in \mathcal{B}\}$  is the family of all maximal independent vectors for  $\mathcal{B} \subseteq \mathcal{C}$ . As in the Boolean case  $\mathcal{C} = \{0, 1\}^n$ , we shall use the notation  $\alpha = |\mathcal{A}_t|$  and  $\beta = |\mathcal{B}_t|$ .

Theorems 1 and 2 admit the following generalizations:

**Theorem 4** *For every polymatroid function  $f : \mathcal{C} \mapsto \{0, 1, \dots, r\}$  and threshold  $t \in \{1, \dots, r\}$  such that  $\beta \geq 2$  we have the inequality*

$$\alpha \leq \beta^{(\log t)/c(2n, \beta)}. \tag{36}$$

Let  $\mathcal{H} \subseteq \mathcal{C}$  be a set of integral  $n$ -vectors,  $f : \mathcal{C} \mapsto \{0, 1, \dots, r\}$  a polymatroid function and  $t \in \{1, \dots, r\}$  a threshold. As before, we say that  $(f, t)$  is a separator for  $\mathcal{H}$  if  $f(x) \geq t$  for all  $x \in \mathcal{H}$ .

**Theorem 5** *Let  $(f, t)$  be a polymatroid separator for a set  $\mathcal{H}$  of at least two points in  $\mathcal{C}$ . Then*

$$|\mathcal{A}_t(f) \cap \mathcal{I}(\mathcal{H})| \leq |\mathcal{H}|^{(\log t)/c(2n, |\mathcal{H}|)}. \quad (37)$$

Using the same argument as that preceding Theorem 3, and the fact that the dualization problem on boxes can still be solved in quasi-polynomial time, see [8], we readily arrive at the following generalization of Theorem 3.

**Theorem 6** *Consider a system of polymatroid inequalities*

$$f_i(x) \geq t_i, \quad i = 1, \dots, m, \quad x \in \mathcal{C},$$

*in which the right hand sides are bounded by a quasi-polynomial function in the size of the system. Suppose further that this system has a quasi-polynomial time feasibility oracle. Then all minimal feasible solutions for the system can be enumerated in incremental quasi-polynomial time.*

Our proof of Theorem 5 makes use of a generalization of Lemma 2. To state this generalization, we need first to extend the notion of proper mappings of maximal independent sets to binary trees. Call a family  $\mathcal{A} \subseteq \mathcal{C}$  an antichain if no two elements are comparable in  $\mathcal{A}$ . Given a binary tree  $\mathbf{T}$ , an antichain  $\mathcal{B} \subseteq \mathcal{C}$ , and a collection  $\mathcal{A} \subseteq \mathcal{I}(\mathcal{B})$  of maximal independent elements of  $\mathcal{B}$ , let us consider again mappings  $\phi : L(\mathbf{T}) \mapsto \mathcal{I}(\mathcal{B})$  that assign a maximal independent element  $a^l \in \mathcal{A}$  to each leaf  $l$  of  $\mathbf{T}$ . To each node  $v$  of the tree  $\mathbf{T}$ , we associate the element  $x^v = \bigwedge_{l \in L(\mathbf{T}(v))} a^l$ . The mapping  $\phi$  will be called proper if it assigns different independent elements to different leaves, and if the element  $x^u \vee x^v$  is not independent whenever  $u$  and  $v$  are incomparable nodes of  $\mathbf{T}$ . The latter condition implies that for every pair of incomparable nodes  $u, v \in V(\mathbf{T})$ , there exists an element  $a \in \mathcal{A}$  for which  $a \leq x^u \vee x^v$ .

**Lemma 11** *Let  $\mathcal{B} \subseteq \mathcal{C}$  be an antichain of size  $|\mathcal{B}| \geq 2$  in an  $n$ -dimensional integral box  $\mathcal{C}$  and let  $\mathcal{A} \subseteq \mathcal{I}(\mathcal{B})$ . Then there exists a binary tree  $\mathbf{T}$  and a proper mapping  $\phi : L(\mathbf{T}) \mapsto \mathcal{A}$ , such that*

$$|L(\mathbf{T})| \geq |\mathcal{A}|^{c(2n, |\mathcal{B}|)/\log |\mathcal{B}|}. \quad (38)$$

**Proof.** By induction on  $|\mathcal{A}|$ . For  $b \in \mathcal{B}$  and  $i \in V \stackrel{\text{def}}{=} \{1, \dots, n\}$ , let

$$\begin{aligned} \mathcal{A}_b^i &\stackrel{\text{def}}{=} \{a \in \mathcal{A} \mid a_i \geq b_i\}, \text{ and} \\ \mathcal{A}_b^{V \setminus i} &\stackrel{\text{def}}{=} \{a \in \mathcal{A} \mid a_j \geq b_j \text{ for all } j \in V \setminus i\}. \end{aligned}$$

We shall make use of the following lemma.

**Lemma 12** *For every antichain  $\mathcal{B} \subseteq \mathcal{C}$  and every  $\mathcal{A} \subseteq \mathcal{I}(\mathcal{B})$ , there exist  $b \in \mathcal{B}$  and an  $i \in V$  such that*

$$|\mathcal{A}_b^i| \geq \frac{|\mathcal{A}|}{2n} \text{ and } |\mathcal{A}_b^{V \setminus i}| \geq \frac{|\mathcal{A}|}{2n|\mathcal{B}|}.$$

Then, to build a binary tree  $\mathbf{T}$  on  $\mathcal{A}$ , let us associate to its root the pair  $i \in V$ ,  $b \in \mathcal{B}$  which satisfies the conditions of Lemma 12, and let us use members of  $\mathcal{A}_a^i$  to label all left-leaves of  $\mathbf{T}$ , and use  $\mathcal{A}_a^{V \setminus i}$  to label all right-leaves. Clearly,  $0 < |\mathcal{A}_a^i| < |\mathcal{A}|$ ,  $0 < |\mathcal{A}_a^{V \setminus i}| < |\mathcal{A}|$ , and both are collections of maximal independent elements of  $\mathcal{B}$ . Therefore, we can conclude by induction that there exist binary trees  $\mathbf{T}_1$  and  $\mathbf{T}_2$  of sufficiently large number of leaves, and proper mappings  $\phi_1 : L(\mathbf{T}_1) \mapsto \mathcal{A}_b^i$  and  $\phi_2 : L(\mathbf{T}_2) \mapsto \mathcal{A}_b^{V \setminus i}$  such that

$$\bigwedge_{l \in L(\mathbf{T}_1)} a_l^i \geq b_i \quad \text{and} \quad \bigwedge_{l \in L(\mathbf{T}_2)} a_l^{V \setminus i} \geq b[V \setminus i],$$

and thus the join of these two elements is at least  $b$ . The lemma then follows by the argument used in the proof of Lemma 2.  $\square$

For  $\mathcal{A} \subseteq \mathcal{C}$  and  $x \in \mathbb{Z}$ , let us define

$$d_{\mathcal{A}}(x) \stackrel{\text{def}}{=} |\{a \in \mathcal{A} \mid a_i \geq x\}|.$$

**Lemma 13** *Let  $\mathcal{A} \subseteq \mathcal{C}$  be an antichain in an  $n$ -dimensional integral box  $\mathcal{C}$ , and let  $\epsilon > 0$  be a given constant. If  $n > 1$  and  $m = |\mathcal{A}| \geq 1 + 1/\epsilon$ , then there exists an  $x \in \mathbb{Z}$  such that  $0 \leq x \leq \max\{c_1, \dots, c_n\}$  and*

$$m \frac{1}{(1 + \epsilon)n} \leq d_{\mathcal{A}}(x) \leq m \left(1 - \frac{1}{(1 + \epsilon)n}\right).$$

**Proof.** Let us define the vector  $y \in \mathbb{C}$  by setting

$$y_i = \max\{x \in \mathbb{Z} \mid 0 \leq x \leq c_i, d_{\mathcal{A}}(x) > m(1 - \frac{1}{(1 + \epsilon)n})\},$$

for  $i = 1, \dots, n$ , and let  $Y = \{i \in V \mid y_i = c_i\}$ . Assume indirectly that for every  $i \in V$ , and every  $x \in \mathbb{Z}$  such that  $y_i < x \leq c_i$ , we have  $d_{\mathcal{A}}(x) < m/(1 + \epsilon)n$ . Let us observe first that  $Y \neq V$  must hold, since otherwise

$$\begin{aligned} \left| \bigcup_{i=1}^n \{a \in \mathcal{A} \mid a_i < y_i\} \right| &\leq \sum_{i=1}^n |\{a \in \mathcal{A} \mid a_i < y_i\}| \\ &= \sum_{i=1}^n (m - d_{\mathcal{A}}(y_i)) < n \frac{m}{(1 + \epsilon)n} < m, \end{aligned}$$

implying that there exists an  $a^o \in \mathcal{A}$  such that  $a^o \geq y = c$ . But this would imply that  $m = 1$ , since  $\mathcal{A}$  is an antichain, contradicting our assumptions. Now since there is an  $a^o \in \mathcal{A}$  such that  $a^o \geq y$  and  $\mathcal{A}$  is an antichain, it follows that, for all other elements  $a \in \mathcal{A} \setminus \{a^o\}$ , there must exist an  $i \in V$  for which  $a_i > y_i$ . Consequently,

$$m - 1 \leq \sum_{a \in \mathcal{A}} |\{i \in V \mid a_i > y_i\}| = \sum_{i \notin Y} d_{\mathcal{A}}(y_i + 1) < n \frac{m}{(1 + \epsilon)n},$$

and therefore we get  $m < 1 + 1/\epsilon$ , again a contradiction.  $\square$

**Proof of Lemma 12.** Assume without loss of generality that  $|\mathcal{A}| \geq 2$ . Let us apply Lemma 13 with  $\epsilon = 1$  for the antichain  $\mathcal{A} \subseteq \mathcal{I}(\mathcal{B})$ , and obtain an  $i \in V$  and an  $0 \leq x \leq c_i$  such that

$$|\mathcal{A}| \frac{1}{2n} \leq d_{\mathcal{A}}(x) \leq |\mathcal{A}| \left(1 - \frac{1}{2n}\right).$$

Let us next note that for every  $a \in \mathcal{A}$  for which  $a_i < x$ , there exists an element  $b \in \mathcal{B}$  such that  $b_i \leq x$ , and  $b_j \leq a_j$  for all  $j \neq i$ , by the maximality of the independent element  $a$ . Thus,

$$\frac{|\mathcal{A}|}{2n} \leq |\mathcal{A}| - d_{\mathcal{A}}(x) \leq \left| \bigcup_{b \in \mathcal{B}} \{a \in \mathcal{A} \mid a_i < x, a_j \geq b_j \text{ for all } j \neq i\} \right|$$

holds, from which we conclude that there must exist a  $b \in \mathcal{B}$  such that

$$|\{a \in \mathcal{A} \mid a_i < x, a_j \geq b_j \text{ for all } j \neq i\}| \geq \frac{|\mathcal{A}|}{2n|\mathcal{B}|}.$$

This gives immediately  $|\mathcal{A}_b^{V \setminus i}| \geq |\mathcal{A}|/(2n|\mathcal{B}|)$ . On the other hand, since  $b_i \leq x$  and  $d_{\mathcal{A}}(x) \geq |\mathcal{A}|/(2n)$ , we obtain  $|\mathcal{A}_b^i| = d_{\mathcal{A}}(b_i) \geq |\mathcal{A}|/(2n)$ , and the first inequality of the lemma follows.  $\square$

We mention in closing that Theorem 2 can also be generalized for polymatroid functions  $f : \mathcal{L}_1 \times \cdots \times \mathcal{L}_n \mapsto \{0, 1, \dots\}$ , where  $\mathcal{L}_1, \dots, \mathcal{L}_n$  are arbitrary lattices (i.e. partially ordered sets with meet  $\wedge$  and join  $\vee$  operations). Let  $(f, t)$  be a polymatroid separator for a set  $\mathcal{X} \subseteq \mathcal{L}_1 \times \cdots \times \mathcal{L}_n$  of size  $|\mathcal{X}| \geq 2$ , then it can be shown that

$$|\mathcal{A}_t(f) \cap \mathcal{I}(\mathcal{X})| \leq |\mathcal{X}|^{(\log t)/c(2Q, |\mathcal{X}|)}, \quad (39)$$

where  $Q = \sum_{i=1}^n |\mathcal{L}_i|$ . In particular,  $|\mathcal{A}_t(f)| \leq \max(Q, \mathcal{B}_t(f)^{(\log t)/c(2Q, \mathcal{B}_t(f))})$ . Note, however, that for the case where each lattice  $\mathcal{L}_i$  is a chain, (39) is weaker than (37).

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