

# Matroid Intersections, Polymatroid Inequalities, and Related Problems\*

Endre Boros<sup>1</sup>, Khaled Elbassioni<sup>2</sup>, Vladimir Gurvich<sup>1</sup>, and Leonid Khachiyan<sup>2</sup>

<sup>1</sup> RUTCOR, Rutgers University,  
640 Bartholomew Road, Piscataway NJ 08854-8003,  
{boros,gurvich}@rutcor.rutgers.edu

<sup>2</sup> Department of Computer Science, Rutgers University,  
110 Frelinghuysen Road, Piscataway NJ 08854-8003,  
elbassio@paul.rutgers.edu,leonid@cs.rutgers.edu

**Abstract.** Given  $m$  matroids  $M_1, \dots, M_m$  on the common ground set  $V$ , it is shown that all maximal subsets of  $V$ , independent in the  $m$  matroids, can be generated in quasi-polynomial time. More generally, 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  $X \subseteq V$  to the system can be generated in incremental quasi-polynomial time. Our proof of these results is based on a combinatorial inequality for polymatroid functions which may be of independent interest. Precisely, for 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  $\alpha \leq \max\{n, \beta^{(\log t)^c}\}$ , where  $c = c(n, \beta)$  is the unique positive root of the equation  $2^c(n^{c/\log \beta} - 1) = 1$ . In particular, our bound implies that  $\alpha \leq (n\beta)^{\log t}$ . We also give examples of polymatroid functions with arbitrarily large  $t, n, \alpha$  and  $\beta$  for which  $\alpha = \beta^{(1-o(1)) \log t/c}$ .

## 1 Introduction

Given  $m$  matroids  $M_1, \dots, M_m$  on the common ground set  $V$  of cardinality  $n$ , Lawler, Lenstra and Rinnooy Kan [14] in 1980 asked the question of the complexity of generating all maximal sets independent in all the matroids, and gave an exponential-time algorithm whose running time is  $O(n^{m+2})$  per each generated maximal independent set. This matroid intersection problem has interesting applications in a variety of fields including combinatorial optimization [13,19] and symbolic analysis of electrical circuits [10]. In this paper, we show that all

---

\* This research was supported by the National Science Foundation (Grant IIS-0118635), and by the Office of Naval Research (Grant N00014-92-J-1375). The second and third authors are also grateful for the partial support by DIMACS, the National Science Foundation's Center for Discrete Mathematics and Theoretical Computer Science.

maximal sets independent in  $m$  matroids can be generated in incremental quasi-polynomial time. More precisely, assume that each matroid  $M_i$  is described by an independence oracle, i.e., an algorithm that, given a set  $X \subseteq V$ , determines whether or not  $X$  is independent in  $M_i$ .

**Theorem 1.** *Let  $M_1, \dots, M_m$  be  $m$  matroids on the common ground set  $V$ ,  $|V| = n$ , and let  $\mathcal{F} \subseteq 2^V$  be the family of all maximal sets independent in all the matroids. Given a partial list  $\mathcal{H} \subseteq \mathcal{F}$ , either a new element in  $\mathcal{F} \setminus \mathcal{H}$  can be computed, or  $\mathcal{F} = \mathcal{H}$  can be recognized, in  $k^{o(\log k)}$  time and  $\text{poly}(k)$  calls to the independence oracles, where  $k \stackrel{\text{def}}{=} \max\{m, n, |\mathcal{H}|\}$ .*

In fact, we shall consider a wider class of problems of which matroid intersection is a special case. 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 *monotone* if  $f(X) \leq f(Y)$  whenever  $X \subseteq Y$ , and *submodular* if

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

holds for all subsets  $X, Y \subseteq V$ . Finally,  $f$  is called a *polymatroid function* if it is monotone, submodular and  $f(\emptyset) = 0$ . Given a system of *polymatroid inequalities*:

$$f_i(X) \geq t_i, \quad i = 1, \dots, m, \tag{1}$$

where each of the polymatroid functions  $f_i : 2^V \mapsto \mathbb{Z}_+$  is defined via an *evaluation oracle*, and  $t_1, \dots, t_m$  are given positive integral thresholds, let  $\mathcal{A}$  and  $\mathcal{B}$ , respectively, denote the family of all maximal infeasible and minimal feasible sets for (1). It is easy to see that  $\mathcal{A} = \mathcal{I}(\mathcal{B})$ , where  $\mathcal{I}(\cdot)$  denotes the family of all maximal independent sets for the hypergraph  $(\cdot)$ . Consider the following problem:

*GEN*( $\mathcal{B}, \mathcal{H}$ ): *Given a system of polymatroid inequalities (1) and a collection  $\mathcal{H} \subseteq \mathcal{B}$  of minimal feasible sets for (1), either find a new minimal feasible set  $H \in \mathcal{B} \setminus \mathcal{H}$  for (1), or show that  $\mathcal{H} = \mathcal{B}$ .*

Clearly, the matroid intersection problem can be described as a system of polymatroid inequalities (1). [Indeed, let  $\rho_i : 2^V \mapsto \{0, 1, \dots, n\}$  be the rank function of  $M_i$ . Then the rank function of the dual matroid

$$f_i(X) = \rho_i(V \setminus X) + |X| - \rho_i(V) : 2^V \mapsto \{0, 1, \dots, n\}$$

is a polymatroid function. Furthermore, a set  $X \subseteq V$  is independent in  $M_i$  if and only if  $f_i(V \setminus X) \geq n - \rho_i(V)$ . Letting  $\mathcal{B} \stackrel{\text{def}}{=} \{V \setminus X \mid X \in \mathcal{F}\}$ , we conclude, therefore, that  $\mathcal{B}$  is the family of minimal solutions for the system of polymatroid inequalities  $f_i(X) \geq n - \rho_i(V)$ ,  $i = 1, \dots, m$ .]

The main result of this paper, Theorem 2 below, generalizes Theorem 1 to systems of polymatroid inequalities (1). Let, as before,  $\mathcal{B}$  denote the family of all minimal solutions to (1). A generation algorithm for  $\mathcal{B}$  is said to

run in *incremental quasi-polynomial* time if it can solve problem  $GEN(\mathcal{B}, \mathcal{H})$  in  $2^{\text{polylog}^k}$  operations and calls to the evaluation oracles for  $f_1, \dots, f_m$ , where  $k = \max\{m, n, |\mathcal{H}|\}$ .

**Theorem 2.** *Consider a system of polymatroid inequalities (1) in which the right-hand sides are bounded by a quasi-polynomial in the dimension of the system:*

$$\max\{t_1, \dots, t_m\} \leq 2^{\text{polylog}(nm)}.$$

*Then all minimal solutions to (1) can be generated in incremental quasi-polynomial time.*

Theorem 2 can be complemented with the following negative result.

**Proposition 1.** *There exist polymatroid inequalities  $f(X) \geq t$ , with polynomial-time computable left-hand side, for which problem  $GEN(\mathcal{B}, \mathcal{H})$  is NP-hard for exponentially large  $t$ .*

The paper is organized as follows. In section 2, we present a combinatorial inequality bounding the number of maximal infeasible sets  $\mathcal{A}$  by a quasi-polynomial in  $n$  and the number of minimal feasible sets  $\mathcal{B}$ . We then use this inequality in Section 4 to reduce problem  $GEN(\mathcal{B}, \mathcal{H})$ , in quasi-polynomial time, to the well-known hypergraph dualization problem, that is, the generation of all maximal independent sets of an explicitly given hypergraph. Since the hypergraph dualization problem can be solved in incremental quasi-polynomial time [9], this will prove Theorem 2 and allow for the efficiently incremental solution of a number of applications, in addition to matroid intersections. Some of these applications are briefly discussed in Section 3, including the generation of minimal feasible solutions to a system of non-negative linear inequalities in Boolean variables (integer programming), minimal infrequent sets of a database (data mining), minimal connectivity ensuring collections of subgraphs from a given list (reliability theory), and minimal spanning collections of subspaces from a given list (linear algebra). The proof of the polymatroid inequality will be given in Sections 5 and 6.

## 2 An Inequality for Polymatroid Functions

Given a polymatroid 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  $f(X) < t$ . Throughout the paper we shall use the notation  $\alpha = |\mathcal{A}_t(f)|$  and  $\beta = |\mathcal{B}_t(f)|$ .

**Theorem 3.** *For every polymatroid function  $f$  and threshold  $t \in \{1, \dots, r(f)\}$  such that  $\beta \geq 2$  we have the inequality*

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

where  $c(n, \beta)$  is the unique positive root of the equation<sup>1</sup>

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

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

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

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 3 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 3 is reasonably sharp. For instance, given positive integers  $k, l$ , and  $d$ , let  $V = V_1 \cup \dots \cup V_k$  be the disjoint union of  $k$  sets of  $l$  vertices each, and for  $X \subseteq V$ , define  $f(X) = d^k$  if  $|X \cap V_i| \geq d$  for some  $i \in \{1, \dots, k\}$ , and  $f(X) = d^k - \prod_{i=1}^k (d - |X \cap V_i|)$  otherwise. Then  $f$  is a polymatroid function of range  $r = d^k$  for which  $n = kl$ ,  $|\mathcal{A}_r| = \binom{l}{d-1}^k$ , and  $|\mathcal{B}_r| = k \binom{l}{d}$ . Thus, letting  $t = r$ ,  $d = k$ , and  $l = 2^k$ , we obtain an infinite family of polymatroid functions for which  $c(n, \beta) = (1 - o(1)) \log k$  and

$$\alpha = \beta^{(1-o(1)) \log t / c(n,\beta)},$$

as  $k \rightarrow \infty$ .

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  $\beta = |\mathcal{B}_t| = 2^t$ .

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  $H \in \mathcal{H}$ . We can further strengthen Theorem 3 as follows.

**Theorem 4.** *Let  $(f, t)$  be a polymatroid separator for a hypergraph  $\mathcal{H}$  of cardinality  $|\mathcal{H}| \geq 2$ . Then*

$$|\mathcal{A}_t(f) \cap \mathcal{I}(\mathcal{H})| \leq |\mathcal{H}|^{(\log t)/c(n, |\mathcal{H}|)}, \tag{5}$$

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

Clearly, Theorem 3 is a special case of Theorem 4 for  $\mathcal{H} = \mathcal{B}_t(f)$ . Since the right-hand side of (5) monotonically increases with  $|\mathcal{H}|$ , we can assume without loss of generality that  $\mathcal{H}$  is Sperner, i.e., none of the hyperedges of  $\mathcal{H}$  contains another hyperedge of  $\mathcal{H}$ .

<sup>1</sup> All logarithms in this paper are assumed to have base 2

### 3 Applications

Before proving Theorems 2 and 4, let us consider first some applications.

*Monotone Systems of Linear Inequalities in Binary and Integer Variables:* Consider a system  $Ax \geq b$  of  $m$  linear inequalities in  $n$  Boolean variables, where  $A$  is a given non-negative integer  $m \times n$ -matrix and  $b$  is given integer  $m$ -vector. Since a linear inequality with non-negative integer coefficients is clearly polymatroid, all minimal Boolean vectors  $x$  feasible for the system can be generated in quasi-polynomial time by Theorem 2, provided that the right-hand side  $b$  is bounded by a quasi-polynomial in  $n$  and  $m$ . In fact, for linear systems the latter condition can be dropped and the bound of Theorem 3 can be strengthened to a linear bound valid even for real  $A$  and  $b$  and integer  $x$  in an arbitrary box  $0 \leq x \leq c$ . This gives an incremental quasi-polynomial algorithm for enumerating minimal solutions to an arbitrary nonnegative system of linear inequalities in Boolean or integer variables (see [4,6] for more details). Thus knapsack, generalized knapsack, and set covering problems are all included as special cases. The quasi-polynomial generation of all maximal feasible solutions to a generalized knapsack problem improves on known results, since for instance Lawler, Lenstra and Rinnooy Kan [14] conjectured that the generation of the maximal binary feasible solutions of a generalized knapsack problem cannot be done in incremental polynomial time, unless  $P = NP$ .

*Minimal Infrequent Sets for a Database:* Given a hypergraph  $\mathcal{H} \subseteq 2^V$  (or equivalently, a database with binary attributes), and an integer threshold  $t$ , a set  $X \subseteq V$  is called  $t$ -frequent if it is contained in at least  $t$  hyperedges of  $\mathcal{H}$ , and is called  $t$ -infrequent otherwise. The generation of maximal frequent and minimal infrequent sets for are important tasks in knowledge discovery and data mining applications (see, for instance, [1,2,18]). Since the function  $f(X) \stackrel{\text{def}}{=} |\{H \in \mathcal{H} \mid H \not\supseteq X\}|$  is polymatroid of range  $|\mathcal{H}|$ , Theorems 3 and 2 imply respectively that the number of maximal frequent sets can be bounded by a quasi-polynomial in the number of minimal infrequent sets and the sizes of  $V, \mathcal{H}$ , and that the minimal infrequent sets can be generated in quasi-polynomial time. In fact, the bound of Theorem 4 can be strengthened to a sharp linear bound in this case, see [7].

*Connectivity Ensuring Collections of Subgraphs:* 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,  $\mathcal{B}_{r-1}(f)$  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  $\mathcal{B}_{r-1}$  is the set of all spanning trees in the graph  $E_1 \cup \dots \cup E_n$ , see [17].) Since  $k(X)$  can be evaluated at any set  $X$  in polynomial time, Theorem 2 implies that for each  $t \in \{1, \dots, r\}$ , all elements of

$\mathcal{B}_t$  can be enumerated in incremental quasi-polynomial time. This problem has applications in reliability theory [8,16].

*Spanning a Linear Space by Linear Subspaces:* Given a collection  $\mathcal{V} = \{\mathcal{V}_1, \dots, \mathcal{V}_n\}$  of  $n$  linear subspaces of  $\mathbf{F}^r$ , for some field  $\mathbf{F}$ , consider the problem of enumerating all minimal sub-collections  $X$  of  $V = \{1, \dots, n\}$  such that  $\text{Span}(\bigcup_{i \in X} \mathcal{V}_i) = \mathbf{F}^r$ . More generally, consider the polymatroid inequality

$$f(X) = \dim\left(\bigcup_{i \in X} \mathcal{V}_i\right) \geq t, \tag{6}$$

where  $t \in \{1, \dots, r\}$  is a given threshold. Then the set  $\mathcal{B}_t(f)$  of minimal solutions to (6) is the collection of all minimal subsets of  $\mathcal{V}$  the dimension of whose union is at least  $t$ . Theorem 4 then states that for all  $t \in \{1, \dots, r\}$ , the size of  $\mathcal{A}_t(f)$  can be bounded by a log  $t$ -degree polynomial in  $n$  and  $|\mathcal{B}_t(f)|$ , and thus all sets in  $\mathcal{B}_t(f)$  can be enumerated in incremental quasi-polynomial time.

It is worth mentioning that in all of the above examples, generating all maximal infeasible sets for (1) turns out to be NP-hard, see [7,11,15].

## 4 Proof of Theorem 2

In this Section we show that Theorem 2 follows from Theorem 4.

Let  $\mathcal{B}$  be the set of minimal feasible sets for (1). Clearly, we can incrementally generate all sets in  $\mathcal{B}$  by initializing  $\mathcal{H} = \emptyset$  and then iteratively solving problem  $GEN(\mathcal{B}, \mathcal{H})$  a number of  $|\mathcal{B}| + 1$  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 (1)  $n + 1$ -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 (1). Thus, in what follows we can assume without loss of generality that the current set  $\mathcal{H} \subseteq \mathcal{B}$  of minimal solutions to (1) has cardinality of at least 2.

By definition, each pair  $(f_i, t_i)$  is a polymatroid separator for  $\mathcal{H}$ , and therefore Theorem 4 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 (1), then  $\mathcal{A} \subseteq \bigcup_{i=1}^m \mathcal{A}_{t_i}(f_i)$ . Hence we arrive at the following bound:

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

where  $t = \max\{t_1, \dots, t_m\}$ . Now, since  $t_1, \dots, t_m$  are bounded by a quasi-polynomial in  $n$  and  $m$ , we conclude that

$$|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})| \leq 2^{\text{poly} \log k} \quad \text{where } k = \max\{n, m, |\mathcal{H}|\}. \tag{7}$$

By definition, the family  $\mathcal{B} \subseteq 2^V$  of all minimal feasible sets for (1) is a Sperner hypergraph. Furthermore, the hypergraph  $\mathcal{B}$  has a simple superset oracle: given a set  $X \subseteq V$ , we can determine whether or not  $X$  contains some set  $H \in \mathcal{B}$  by checking the feasibility of  $X$  for (1), i.e., by evaluating  $f_1(X), \dots, f_m(X)$ . As observed in [3,11], for any Sperner hypergraph  $\mathcal{B}$  defined via a 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, consider an arbitrary hypergraph  $\mathcal{H} \subseteq \mathcal{B}$ . 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 (1). If  $X$  is feasible for (1) then  $X$  contains a new minimal solution to (1) which can be found by querying the superset oracle at most  $|X| + 1$  times. If  $X \in \mathcal{I}(\mathcal{H})$  is infeasible for (1), 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  $|\mathcal{I}(\mathcal{B}) \cap \mathcal{I}(\mathcal{H})|$ .)

Combining the above reduction with (7) and the fact that the hypergraph dualization problem can be solved in quasi-polynomial time  $poly(n) + (|\mathcal{H}| + |\mathcal{G}|)^{o(\log(|\mathcal{H}| + |\mathcal{G}|))}$  (see [9]), we readily obtain Theorem 2.

## 5 Proper Mappings of Independent Sets into Binary Trees

Our proof of Theorem 4 makes use of a combinatorial construction which may be of independent interest. Theorem 4 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 rooted 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. 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  $\mathbf{T}$ . 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 conclude that the sets  $S_u$  are non-empty and independent, for all non-root nodes  $u$ .

**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*

$$r(f) \geq t \geq |L(\mathbf{T})|. \tag{8}$$

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  $u$  in  $\mathbf{T}$ . 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 contains 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*

$$|L(\mathbf{T})| \geq |\mathcal{S}|^{c(|V|, |\mathcal{H}|) / \log |\mathcal{H}|}. \tag{9}$$

Clearly, Lemmas 1 and 2 imply Theorem 4, which in turn implies Theorem 3. The proof of Lemmas 1 and 2 is given in the next Section.

## 6 Proof of Main Lemmas

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

$$f(S_w) \leq t - |L(\mathbf{T}(w))| \tag{10}$$



holds for every node  $w$  of the 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 (10), 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 (10) 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 (10) 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 3.** *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

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

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 4.** *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 3 for the Sperner hypergraph  $\mathcal{S}^c \stackrel{\text{def}}{=} \{V \setminus I \mid I \in \mathcal{S}\}$ , and obtain that

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

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 4, 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. □

**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

$$L(\alpha) \geq \alpha^{c/\log \beta} \tag{11}$$

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 (11).

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 4, 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

$$L(\alpha) \geq L\left(\frac{\alpha}{n}\right) + L\left(\frac{\alpha}{n\beta}\right). \tag{12}$$

Applying now our inductive hypothesis, we get

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

where the last equality holds by (3). This proves (11), and hence the lemma follows. □

Note that the right-hand side of (11) is the least possible solution of the recursion (12).

## References

1. R. Agrawal, T. Imielinski and A. Swami, Mining associations between sets of items in massive databases, *Proc. 1993 ACM-SIGMOD Int. Conf. on Management of Data*, pp. 207-216.
2. R. Agrawal, H. Mannila, R. Srikant, H. Toivonen and A. I. Verkamo, Fast discovery of association rules, in U. M. Fayyad, G. Piatetsky-Shapiro, P. Smyth and R. Uthurusamy eds., *Advances in Knowledge Discovery and Data Mining*, pp. 307-328, AAAI Press, Menlo Park, California, 1996.
3. J. C. Bioch and T. Ibaraki, Complexity of identification and dualization of positive Boolean functions, *Information and Computation* 123 (1995) pp. 50-63.
4. E. Boros, K. Elbassioni, V. Gurvich, L. Khachiyan and K. Makino, On generating all minimal integer solutions for a monotone system of linear inequalities, in *ICALP 2001*, LNCS 2076, pp. 92-103. An extended version is to appear in *SIAM Journal on Computing*.

5. E. Boros, K. Elbassioni, V. Gurvich and L. Khachiyan, An inequality for polymatroid functions and its applications, DIMACS Technical Report 2001-14, Rutgers University, <http://dimacs.rutgers.edu/TechnicalReports/2001.html>.
6. E. Boros, V. Gurvich, L. Khachiyan and K. Makino, Dual bounded generating problems: partial and multiple transversals of a hypergraph. *SIAM Journal on Computing* 30 (6) (2001) pp. 2036-2050.
7. E. Boros, V. Gurvich, L. Khachiyan and K. Makino, On the complexity of generating maximal frequent and minimal infrequent sets, in *STACS 2002*, LNCS 2285, pp. 133-141.
8. C. J. Colbourn, *The combinatorics of network reliability*, Oxford Univ. Press, 1987.
9. M. L. Fredman and L. Khachiyan, On the complexity of dualization of monotone disjunctive normal forms, *Journal of Algorithms*, 21 (1996) pp. 618-628.
10. M. Galañ, I. Garcíá-Vargas, F.V. Fernández and A. Rodríguez-Vázquez, A new matroid intersection algorithm for symbolic large circuit analysis, in *Proc. 4th Int. Workshop on Symbolic Methods and Applications to Circuit Design*, Oct. 1996.
11. V. Gurvich and L. Khachiyan, On generating the irredundant conjunctive and disjunctive normal forms of monotone Boolean functions, *Discrete Applied Mathematics*, 96-97 (1999) pp. 363-373.
12. T. Helgason, Aspects of the theory of hypermatroids, in *Hypergraph Seminar*, Lecture Notes in Math. 411 (1975) Springer, pp. 191-214.
13. E. L. Lawler, *Combinatorial Optimization: Networks and Matroids*, Holt, Rinehart and Winston, New York, 1976.
14. E. Lawler, J. K. Lenstra and A. H. G. Rinnooy Kan, Generating all maximal independent sets: NP-hardness and polynomial-time algorithms, *SIAM Journal on Computing*, 9 (1980) pp. 558-565.
15. K. Makino and T. Ibaraki, Interior and exterior functions of Boolean functions, *Discrete Applied Mathematics*, 69 (1996) pp. 209-231.
16. K. G. Ramamurthy, *Coherent Structures and Simple Games*, Kluwer Academic Publishers, 1990.
17. R. C. Read and R. E. Tarjan, Bounds on backtrack algorithms for listing cycles, paths, and spanning trees, *Networks*, 5 (1975) pp. 237-252.
18. R. H. Sloan, K. Takata and G. Turan, On frequent sets of Boolean matrices, *Annals of Mathematics and Artificial Intelligence* 24 (1998) pp. 1-4.
19. D.J.A. Welsh, *Matroid Theory* (Academic Press, London, New York, San Francisco 1976).