

Polymatroids and Submodularity

jackedmonds@rogers.com

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A predicate $p(x)$,
i.e., a statement with a variable input subject, x ,
is called NP if whenever it is true
there is an easy proof (certification) of it
(polynomial time relative to the size of x).

$p(x)$ is called coNP when 'not $p(x)$ ' is NP.
**A theory is called a 'good characterization' of $p(x)$
when it states how $p(x)$ is in $NP \cap coNP$.**

It is a practical conjecture that if $p(x)$ is in $NP \cap coNP$
then there is a polytime algorithm for deciding it.

Linear program: Maximize cx where $x \geq 0$ and $Ax \leq b$.

Dual linear program: Minimize yb where $y \geq 0$ and $yA \geq c$.

Duality Theorem: $\max = \min$.

Polytope: a bounded convex polyhedron, $P = \{x : x \geq 0 \text{ and } Ax \leq b\}$.

Or: all convex combinations of a finite set V .

Caratheodory Theorem. x in P iff

convex combination of a small number from V .

An easy description of system $L = \{x \geq 0 \text{ and } Ax \leq b\}$

and an easy description of V ,

provides $NP \cap coNP$,

and probably polytime algorithms.

There are several known classes of “nice systems”, $L = \{x \geq 0 \text{ and } Ax \leq b\}$, where the matrix A is all 0s & 1s. Let $P_L = \{x \text{ satisfying } L\}$.

Then where $S \subseteq E$ (\equiv the column index set),

$f(S) = \max [x(S)] \equiv \max [\sum x_j : j \text{ in } S], x \in P_L$, is called the set extension.

In fact for $w \in R^E_+$, $f(w) = \max \{wx : x \in P_L\}$, is called the real extension.

Obvious: $P_L = P(E, f) \equiv \{x : x \geq 0; x(S) \leq f(S) \text{ for every } S \subseteq E\}$.

Polymatroids turn out to be $P(E, f)$ where the set function, f , is submodular, non-decreasing, and $f(\emptyset) = 0$.

For most known classes of nice systems, the f has not been studied.

Example Theorem: For polymatroid functions f_1 & f_2 :

Minkowski sum, $P(E, f_1) + P(E, f_2) = P(E, f_1 + f_2)$. **Other systems?**

You know: For any finite set E of vectors over any field, every maximal independent subset (basis of E) is a largest independent subset.

MATROID, $M = (E, F)$:

Abstract set E with “independent” subsets so that every subset of an ind set is ind; and

For every subset S of E , every maximal independent subset of S (basis of S) is largest,

called the rank “ $r(S)$ ”.

Let F denote the family of independent sets.

An **integer polymatroid** F with groundset E is a bounded set F of non-negative integer-valued vectors coordinatized by E

such that F includes the origin; and any non-neg integer vector is a member of F if it is \leq some member of F ; **and for any non-neg integer vector y , every maximal integer vector x in F which is $\leq y$ has the same sum $x(E) \equiv \sum (x_j : j \in E)$, called the rank $r(y)$ of vector y in F .**

The 0–1 vectors in an integral polymatroid are the incidence vectors of the independent sets J (i.e., $J \in F$) of a matroid $M = (E, F)$.

Polymatroid: leave out “integer”.

For any vectors x and y in Z_+ ,

$x \cap y$ denotes the maximal integer vector which is $\leq x$ and $\leq y$;

$x \cup y$ denotes the minimal integer vector which is $\geq x$ and $\geq y$.

Clearly $0 \leq r(0) \leq r(x) \leq r(y)$ for $x \leq y$ (*non-neg & non-decreasing*); AND

Proposition. For any x and y in Z_+ , **$r(x \cup y) + r(x \cap y) \leq r(x) + r(y)$** . (*submodularity*)

Proof: Let v be an P -basis of $x \cap y$. Extend it to P -basis w of $x \cup y$. $r(x \cap y) = |v|$ and $r(x \cup y) = |w|$.

$r(x) + r(y) \geq |w \cap x| + |w \cap y| = |w \cap (x \cap y)| + |w \cap (x \cup y)| = |v| + |w| = r(x \cup y) + r(x \cap y)$. \square

For subsets S of E , define $f_F(S)$ to be $r(x)$ where $x \in Z_+$ is very large for coordinates in S and $= 0$ for coordinates not in S .

Corollary. For sets $S \subseteq E$, $f_F(S)$ is a non-neg, non-decreasing, and submodular set function.

Such an $f_F(S)$ for subsets $S \subseteq E$, is called a ***polymatroid function***.

(We will see how, conversely,

any polymatroid is given by a polymatroid function.

Discussion of **polymatroid** is the same by omitting everywhere the word ‘integer’ (but including only the condition that it be closed and bounded).

It will be theorem that any polymatroid F is in fact a polytope.

However, when it is an integral polymatroid,

$f_{F(S)}$ is integer valued

and $F = \{x: x \geq 0; x(S) \leq f_{F(S)}\}$ has *total dual integrality* (TDI),

meaning F has integer-valued vertices (primal integrality),

and for any integer-valued c ,

the primal l.p., $\max\{cx: x \geq 0; x(S) \leq f_{F(S)}\}$, and

the dual l.p., $\min\{\sum y_S f_{F(S)}: y_S \geq 0; \sum (y_S : S \text{ containing } j) \geq c_j \}$,
 have integer-valued optima $x = (x_j : j \text{ in } E)$ and $y = (y_S : S \subseteq E)$.

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polymatroid function: A real-valued function $f(S)$, $S \subseteq E$, where

(a) $f(\emptyset) = 0$ and $f(A) \geq 0$ for every $A \subseteq E$;

(b) is non-decreasing: $A \leq B \Rightarrow f(A) \leq f(B)$; and

(c) submodular: $f(A \cup B) + f(A \cap B) \leq f(A) + f(B)$ for $A \subseteq E$ and $B \subseteq E$.

1) The Polymatroid Theorem. For a polymatroid function, $f(S)$, $S \subseteq E$, the following polyhedron is a polymatroid:

$P(E, f) \equiv \{ x \geq 0 : x(S) \leq f(S) \text{ for every } S \subseteq E \}$.

The polymatroidal rank function r of $P(E, f)$, for any $a = [a_j : j \in E] \geq 0$, is $r(a) = \min [f(S) + a(E - S) : S \subseteq E]$.

Proof. Clearly $r(a) \leq \min [f(S) + a(E - S) : S \subseteq E]$ (weak l.p. duality).

Let $x_0 \geq 0$ be any maximal $x \geq 0$ such that $x \leq a$ and $x(S) \leq f(S)$ for every $S \subseteq E$.

For any sets A and B such that $x_0(A) = f(A)$ and $x_0(B) = f(B)$, we have

$$x_0(A \cup B) + x_0(A \cap B) \leq f(A \cup B) + f(A \cap B)$$

$$\leq f(A) + f(B) = x_0(A) + x_0(B) = x_0(A \cup B) + x_0(A \cap B),$$

and so $x_0(A \cup B) = f(A \cup B)$ and $x_0(A \cap B) = f(A \cap B)$,

and so where S is the union of all sets where equality holds,

we have $x_0(S) = f(S)$ and $x_0(E - S) = a(E - S)$. Hence $r(a) = f(S) + a(E - S)$.

By the same argument, where f is integer-valued we get an integer polymatroid.

**Theorem. For any polymatroid P : we have $P = P(E, f_P)$.
 Thus, all polymatroids are polytopes,
 and they correspond precisely to polymatroid set functions.**

Submodular set function minimization:

Any submodular set function $f'(S)$, $S \subseteq E$,
 not necessarily non-neg and non-decreasing,
 can, for some easily specified vector $a \geq 0$,
 be expressed up to an additive constant as
 $f' = f(S) + a(E - S)$ where $f(S)$ is a polymatroid function.

And so Thm 1 again: $r(a) = \min [f'(S) : S \subseteq E]$.

That is, if we have an oracle which tells the value of $f'(S)$ for any $S \subseteq E$,
 we can **find the minimum of $f'(S)$**

by getting a maximal vector x

such that $x \in P(E, f) \equiv \{ x \geq 0 : x(S) \leq f(S) \text{ for every } S \subseteq E \}$

and $x \leq a$.

The only difficulty in simply pushing x up arbitrarily in the various coordinates, starting with $x = 0$, until any further push will violate $x \leq a$ in some coordinate or $x(S) \leq f(S)$ for some $S \subseteq E$, is that **there are too many inequalities, $x(S) \leq f(S)$, to check.**

When f is the rank function of a matroid, $x(J) = f(j)$ iff J is independent.

More generally, it is easy to verify that an x **is not in $P(E, f)$**

i.e., not independent,

by showing $x(S) > f(S)$ for one set $S \subseteq E$.

But how can we verify that x satisfies $x(S) \leq f(S)$ for all $S \subseteq E$?

We get $[\min f(S)]$ into $NP \cap coNP$ relative to the oracle for $f(S)$ by getting a good description of when x **is in $P(E, f)$.**

We do that by getting a good description of the vertices of **$P(E, f)$.**

Then we can certify that $x \in P(E, f)$ by using Cartheodory's theorem to show x as a convex combination of a small number of vertices of **$P(E, f)$.**

Let $j(1), j(2), \dots, j(k)$, be an arbitrary ordering of a subset of E .
For each integer i , $1 \leq i \leq k$, let $A_i = \{j(1), j(2), \dots, j(i)\}$.

2) Greedy Vertex Theorem.

The vertices of $P(E, f)$ are the vectors, x , where

$$x_{j(1)} = f(A_1);$$

$$x_{j(i)} = f(A_i) - f(A_{i-1}) \text{ for } 2 \leq i \leq k; \text{ and } x_{j(i)} = 0 \text{ for } k < i \leq |E|.$$

To solve the linear program, maximize $cx = (\sum c_j x_j: j \in E)$
 over $x \in P(E, f) \equiv \{x \geq 0 : x(S) \leq f(S) \text{ for every } S \subseteq E\}$, let
 $j(1), j(2), \dots$ be an ordering of E such that $c_{j(1)} \geq c_{j(2)} \geq \dots \geq c_{j(k)} > 0 \geq c_{j(k+1)} \geq$
 \dots . For each integer $i, 1 \leq i \leq k$, let $A_i = \{j(1), j(2), \dots, j(i)\}$.

The dual l.p. is to Minimize $f y = \sum [f(S)y(S): S \subseteq E]$
 where $y(S) \geq 0$; and for every $j \in E, \sum [y(S): j \in S] \geq c_j$.

2') The Greedy Algorithm:

cx is maximized over $x \in P(E, f)$ by the vector x of the form in Theorem 2:
 $x_{j(1)} = f(A_1)$; $x_{j(i)} = f(A_i) - f(A_{i-1})$ for $2 \leq i \leq k$; and $x_{j(i)} = 0$ for $k < i \leq |E|$.

The Dual Greedy Algorithm:

An optimum solution, $y = [y(A)]$, $A \subseteq E$, to the dual l.p. is
 $y(A_i) = c_{j(i)} - c_{j(i+1)}$ for $1 \leq i \leq k - 1$;
 $y(A_k) = c_{j(k)}$; and $y(A) = 0$ for all other $A \subseteq E$.

2'') Where f is the rank function of a matroid $M = (E, F)$,
 Theorem 2 implies that
 the vertices of $P(E, f)$ are precisely the 0,1 incidence vectors of the
 independent sets $J \in F$ of matroid M .

Such a $P(E, f)$ is called a matroid polytope. For matroids, the
 greedy alg takes a more familiar looking form. (Boruvka 1930s)

2''') Another form of the Greedy Alg Thm.

For any submod, f , with $f(\emptyset) = 0$, (not non-decreasing), and any $c \in \mathbb{R}^E$,

$\max[cx \equiv (\sum c_j x_j : j \in E) \text{ over } x \in P'(E, f)],$

where $P'(E, f) \equiv \{ x : x(S) \leq f(S) \text{ for every } S \subseteq E, x(E) = f(E) \},$

is achieved by

$x_{j(k)} = f[j(1), \dots, j(k)] - f[j(1), \dots, j(k-1)]$ where $c_{j(1)} \geq c_{j(2)} \geq \dots \geq c_{j(n)}$.

$f(w) = \max[w x \text{ over } x \in P'(E, f)],$ (= dual l.p. min), $w \in \mathbb{R}^E,$

is a real extension of $f(S)$, $S \subseteq E$, sometimes called the Lovasz extension of $f(S)$.

Optional. The facets of a polymatroid.

Let f be a polymatroidal function on $S \subseteq E$.

A set $A \subseteq E$ is called f -closed or an f -flat, when $f(A) < f(C)$ for any $C \subseteq E$ which properly contains A .

Theorem. If A and B are f -closed then $A \cap B$ is f -closed.

A set $A \subseteq E$ is called f -separable when $f(A) = f(A_1) + f(A_2)$ for some partition of A into non-empty subsets A_1 and A_2 . Otherwise A is called f -inseparable.

Theorem. Any $A \subseteq E$ partitions in only one way into a family of f -inseparable sets A_i such that $f(A) = \sum f(A_i)$. The A_i 's are called the f -blocks of A .

Theorem. Where f is a polymatroidal function such that the empty set is f -closed, the facets of polymatroid $P(E, f)$ are:

$x_j \geq 0$ for every $j \in E$;

and $x(A) \leq f(A)$ for every $A \subseteq E$ which is f -closed and f -inseparable.

“The Hitchcock transportation problem”, longtime well-known in O.R.:

Let each V_p , $p = 1$ and 2 ,
be a family of mutually disjoint subsets of H .

Where $[a_{ij}]$, $i \in E$, $j \in E$, is the 0–1 incidence matrix of $V_1 \cup V_2$,
the following l.p. is known in O.R.
as the

Maximize $cx = \sum [c_j x_j: j \in E]$, where x satisfies
 $L = [x_j \geq 0$ for every $j \in E$, and $\sum [a_{ij} x_j: j \in E] \leq b_i$ for every $i \in H]$.
 The dual l.p. is Minimize $by = \sum b_i y_i$, where
 $y_i \geq 0$ for every $i \in H$, and $\sum [a_{ij} y_i: i \in H] \geq c_j$ for every $j \in E$.

The following properties of the Hitchcock problem are important in its combinatorial use. This is essentially Egervary’s Theorem, 1931.

Theorem. Where the b_i ’s are integers,
the primal l.p. has integer valued opt.

Where the c_j ‘s are integers,
the dual l.p. has integer value opt.

In other words, the primal system L is TDI.

3) The Integer Polymatroid Intersection Theorem:

For any two integral polymatroids $P_1 = P(E, f_1)$ and $P_2 = P(E, f_2)$, $P_1 \cap P_2$ is TDI. In particular, the vertices of $P_1 \cap P_2$ are integer-valued.

L.P. Dual of Polymatroid Intersection.

$P(E, f_1)$ and $P(E, f_2)$ are polymatroids (not necessarily integral), the dual of the l.p.:

Maximize $cx = \sum [c_j x_j: j \in E]$, where $x \in P(E, f_1) \cap P(E, f_2)$

is the l.p.:

Minimize $fy = \sum [f_1(S)y_1(S) + f_2(S)y_2(S):]$

where for every $S \subseteq E$, $y_1(S) \geq 0$ and $y_2(S) \geq 0$;

and for every $j \in E$, $\sum [y_1(S) + y_2(S): j \in S \subseteq E] \geq c_j$.

TDI: *If the c_j 's are all integers, then, regardless of whether f_1 and f_2 are integral, there is an integer-valued solution y which minimizes fy .*

max $cx = \min fy$ where $x \in P(E, f_1) \cap P(E, f_2)$

and where y is dual feasible.

If f is integral, x can be integral. If c is integral, y can be integral.

3') The Matroid Polytope Intersection Theorem.

Where P_1 and P_2 are the polytopes of any two matroids M_1 and M_2 on E , the vertices of $P_1 \cap P_2$ are precisely the vectors which are vertices of both P_1 and P_2 namely, the incidence vectors of sets which are independent in both M_1 and M_2 . In fact, the inequality system for $P_1 \cap P_2$ is TDI.

Where P_1 , P_2 , and P_3 are the polytopes of three matroids on E , polytope $P_1 \cap P_2 \cap P_3$ generally has many fractional vertices besides those which are vertices of P_1 , P_2 , and P_3 .

Where $c = [c_j]$, $j \in E$,
 is any numerical weighting of the elements of E ,
 Finding a set J , independent in both M_1 and M_2 ,
 that has maximum weight sum, $\sum (c_j: j \in J)$,
 is equivalent to the *l.p.* problem:
 Find a vertex x of $P_1 \cap P_2$ that maximizes cx .

Assuming there is a good algorithm for recognizing
 whether or not a set $J \subseteq E$ is independent in M_1 or in M_2 ,
 there is a good algorithm for this problem.

In particular, where f_1 and f_2 are the rank functions, r_1 and r_2 ,
 of any two matroids, $M_1 = (E, F_1)$ and $M_2 = (E, F_2)$,
 and where every $c_j = 1$, we have

3'') The Matroid Cardinality Intersection Theorem:
For any two matroids $M_1 = (E, F_1)$ and $M_2 = (E, F_2)$,
 $\max |J : J \in F_1 \cap F_2| = \min [r_1(S) + r_2(E - S) : S \subseteq E]$.

Example Application. A branching S rooted at node v_0 in directed graph $G(V, E)$, having node-set V and edge-set E , means a set $S \subseteq E$ such that

- 1) S is the edge-set of a spanning tree of G , and
- 2) For each node $v \neq v_0$, exactly one edge of S enters v .

Condition (1) means S is a basis (maximal independent set) in matroid M_1 .

Condition (2) means S is a basis in matroid M_2 .

So for any weighting $c = (c_j : j \in E)$ finding an optimum branching in $G(V, E)$ means finding an optimum weight common basis of matroids M_1 and M_2 , i.e., an optimum vertex of polytope $P(E, r_1) \cap P(E, r_2)$.

Algorithm sketch:

Choose best edge to each $v \neq v_0$ until a directed cycle is created.

Shrink the cycle to a pseudo-node, and continue. When chosen set not shrunk is a branching in the 'pseudo-node graph', expand the pseudo-nodes in the reverse order they were creating and use chosen edges in the unique way to get a branching in $G(E, V)$.

Using also a 3rd matroid similar to (2) we get the "traveling salesman problem". We don't know a good way to optimize over common bases of 3 matroids.

Matroid Partitioning, another application of Matroid Intersection

Matroid Sums:

Let $\{M_i: i \in I\}$, be a family of matroids, $M_i = (E, F_i)$, having rank functions r_i .

Let $J \subseteq E$ be a member of F iff $|S| \leq \sum(r_i(S): i \in I)$ for every $S \subseteq J$.

Since $f(S) = \sum[r_i(S): i \in I]$ is an integer polymatroid function on $S \subseteq E$, $M = (E, F)$ is a matroid, called the sum of the matroids M_i .

4) The Matroid Partitioning Theorem (also called The Matroid Union Thm):

$J \in F$ iff J can be partitioned into sets J_i such that $J_i \in F_i$.

Can be proved by a polytime algorithm which either finds a partitioning of J into sets $J_i \in F_i$ or else finds one set $S \subseteq E$ such that $|S| > \sum(r_i(S): i \in I)$.

Since $M = (E, F)$ is a matroid we can use the greedy algorithm to find an optimum weight set J which is partitionable into sets independent respectively into given matroids M_i .

In particular given graphs $G_i(V_i, E)$ we can easily find an optimum subset of E which is partitionable into forests, respectively, of the graphs $G_i(V_i, E)$.

Proof of (4) from The Matroid Intersection Theorem:

Obtain for a given family of matroids $M_i = (E, F_i)$, $i \in I$, an “optimum” family of mutually disjoint sets $J_i \in F_i$, by using the matroid intersection algorithm (3'') on the following two matroids $M_1 = (E, F_1)$ and $M_2 = (E, F_2)$.

Let E' consist of all pairs (j, i) , $j \in E$ and $i \in I$.

Let $M_1 = (E', F_1)$ be the matroid such that $J \subseteq E'$ is a member of F_1 iff the corresponding sets J_i are mutually disjoint — that is, if and only if the j 's of the members of J are distinct.

Let $M_2 = (E', F_2)$ be the matroid such that $J \subseteq E'$ is a member of F_2 iff the corresponding sets J_i are such that $J_i \in F_i$.

The Minkowski Sum Theorem for polymatroids:

$$P(E, f_1) + P(E, f_2) = P(E, f_1 + f_2)$$

can be obtained from the intersection of two polymatroids in the same way.

Optimum System of k edge-disjoint branchings in digraph $G(V,E)$, rooted at node v_0 :

an optimum subset $J \subseteq E$, where J is a basis of matroids

$M_1 = (E, F_1)$ and $M_2 = (E, F_2)$,

i.e., $J \in F_1 \cap F_2$ and $|J| = r(E)$,

where for the two matroids $M_1 = (E, F_1)$ and $M_2 = (E, F_2)$,

J is a basis of matroid M_1

when J can be partitioned into k spanning trees of $G(V,E)$;

and J is a basis of matroid M_2

when exactly k edges of J enter each node $\neq v_0$.

A version of the 'Disjoint Branchings Theorem':

J can then be partitioned into k edge-disjoint branchings

in digraph $G(V,E)$, rooted at node v_0 . **Please implement.**
