Polyhedral approach: the *p***-median polytope**

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I. Polyhedral approach

Given a finite set $E = \{e_1, e_2, \dots, e_n\}$. Let \mathcal{F} be a family of some particular subsets of E (feasible solutions), and let c be a cost function, a mapping

$$c: E \longrightarrow I\!\!R.$$

The problem

$$(P): \text{ minimize } \{c(F) \stackrel{\text{def}}{=} \sum_{e \in F} c(e) : F \in \mathcal{F}\},\$$

is called a *combinatorial optimization problem*.

Examples: Traveling Salesman problem, Maximum-Weight Matching problem, Minimum Spanning Tree problem, ...

Integer Polyhedra

For every element $F \in \mathcal{F}$, associate a 0,1 vector $x^F \in \{0,1\}^E$, defined as follows:

$$x^{F}(e) = \begin{cases} 1 & \text{if } e \in F \\ 0 & \text{otherwise,} \end{cases}$$

 x^F is called the *incidence vector of* F. (P) may be rewritten as

$$(P): \text{ minimize } \{c^T x^F : x^F \in S\},\$$

where S is the set of incidence vectors of the elements of \mathcal{F} .

Convex hulls: The convex hull of a finite set S, denoted by conv(S), is the set of all points that are a convex combination of points in S.

Proposition 1. Let $S \subseteq \mathbb{R}^n$ be a finite set and let $c \in \mathbb{R}^n$. Then

$$\min\{c^T x : x \in S\} = \min\{c^T x : x \in conv(S)\}.$$



Polytopes:

- $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ where A is an $m \times n$ matrix and $b \in \mathbb{R}^m$, is called a *polyhedron*.
- A *polytope* is a bounded polyhedron.
- An inequality $\alpha^T x \leq \alpha_0$ is *valid* for a polytope P if $P \subseteq \{x : \alpha^T x \leq \alpha_0\}$.
- The *dimension* of a polytope P, $\dim(P)$, is equal to the maximum number of affinely independent points in P minus one.
- Let $\alpha^T x \leq \alpha_0$ be a valid inequality for the polytope P. $F = \{x \in P : \alpha^T x = \alpha_0\}$ is called a *face* of P. F is a *facet* of P if dim $(F) = \dim(P) 1$.

• $x \in P$ is an *extreme* point of P if x is a face of P of dimension 0.



Two useful characterizations of extreme points are:

Characterization 1. $x \in P$ is an extreme point of P if and only if there do not exist $x^1, x^2 \in P, x^1 \neq x^2$, such that $x = \frac{1}{2}x^1 + \frac{1}{2}x^2$.

Characterization 2. $x \in P$ is an extreme point of P if x is the unique solution of a subsystem of inequalities defining P when replaced by equalities.

Theorem 1. A set P is a polytope if and only if there exists a finite set S such that P is the convex hull of S.

By this theorem, the optimization over conv(S) is equivalent to the optimization over a polytope.

An example

Consider the following combinatorial optimization problem (P) formulated as an integer linear program:

subject to $\begin{pmatrix} -13x_1 + 28x_2 \leq 72 \\ -x_1 + 2x_2 \leq 4 \end{pmatrix}$	$\max x_1 + x$	r_2						
$ \left(\begin{array}{ccc} -13x_1 + 28x_2 & \leq & 72 \end{array} \right) \left(\begin{array}{ccc} -x_1 + x_2 & \leq & 1 \\ -x_1 + 2x_2 & < & 4 \end{array} \right) $	subject to						/	1
$P_{1} = \begin{cases} -5x_{1} + 4x_{2} \leq 4 \\ 9x_{1} - 4x_{2} \leq 24 \\ 6x_{1} - 5x_{2} \leq 9 \\ x_{1} \geq 0 \\ x_{2} \geq 0 \end{cases} D = \begin{cases} x_{1} + 2x_{2} \leq 4 \\ x_{1} - x_{2} \leq 1 \\ x_{1} \leq 4 \\ x_{1} \leq 4 \\ x_{1} \geq 0 \\ x_{2} \geq 0 \end{cases}$	$P_1 = \begin{cases} - \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\$	$\begin{array}{c} 13x_1 + 28x_2 \\ -5x_1 + 4x_2 \\ 9x_1 - 4x_2 \\ 6x_1 - 5x_2 \\ x_1 \\ x_2 \\ \text{nd } x_2 \text{ integers} \end{array}$		$72 \\ 4 \\ 24 \\ 9 \\ 0 \\ 0 \\ 0$	$D = \langle$	$ \begin{array}{c} -x_1 + x_2 \\ -x_1 + 2x_2 \\ x_1 - x_2 \\ x_1 \\ x_1 \\ x_2 \end{array} $	$ \vee \vee \wedge \wedge \wedge \wedge \wedge \wedge \wedge \wedge $	$ \begin{array}{c} 1 \\ 4 \\ 1 \\ 4 \\ 0 \\ 0 \\ 0 \end{array} $

Theorem 1 says that the convex hull of the solutions of (P) is also a polytope; in our example this polytope is denoted D.



The *p*-median problem

Given a direct graph G = (V, A) where each arc (u, v) is associated with a cost c(u, v). The problem is to select p nodes, and assign to them the non-selected one such that the assignment cost is minimized.



The costof this solution is :

2+3+1+6+2

The relation with the stable set problem

An instance of the *p*-median problem : a directed graph G = (V, A), a cost function c associatend with the arc set A and a fixed integer $p \leq |V|$.

- From G = (V, A) define an undirected graph I(G) = (A, E) called the *intersection graph of* G.

- The nodes of I(G) are the arcs of G,
- The edges of I(G) are defined as below:































In green a stable set of size |V|-p=5

- The *p*-median problem may be formulated by the following linear integer

 $P_p(G) = \begin{cases} \sum_{v \in V} y(v) = p, \\ \sum_{v:(u,v) \in A} x(u,v) = 1 - y(u) \quad \forall u \in V, \\ x(u,v) \le y(v) \qquad \forall (u,v) \in A, \\ y(v) \le 1 \qquad \forall v \in V, \\ x(u,v) \ge 0 \\ x \text{ and } y \ge z \end{cases}$

- The following four figures give four fractional extreme points of $P_p(G)$.









An Odd cycle



An Odd Y-cycle



An Odd Y-cycle



An Odd Y-cycle

- The parity of a cycle is the number of green nodes+red nodes.



An Odd Y-cycle



An Odd cycle which is not a Y-cycle

The main result

Theorem 2. Let G = (V, A) be a **directed** graph, then $P_p(G)$ is integral for any p if and only if it does not contain as a subgraph none of the graphs

and does not contain an odd Y-cycle plus an arc (u, v) with u and v not in the cycle.



To prove Theorem 2, we need the following two results.



Theorem 4. Let G = (V, A) be an **oriented** graph, then $P_p(G)$ is integral for any p if and only if it does not contain as subgraph one of the graphs

and an odd Y-cycle plus an arc (u, v) with u and v not in the cycle.

A case in the proof of Theorem 2

- Let G = (V, A) be a directed graph that does not contain the subgraphs of Theorem 2.

- Assume that (x, y) is an extreme fractional point of $P_p(G)$.

An extreme point is the **unique** solution of a system of inequalities when replaced by equalities.

To arrive to a contradiction, we plan to show that (x, y) is not the unique solution of the set of inequalities that are satisfied as equalities by (x, y).

- We can assume that x(u,v) > 0 for all $(u,v) \in A$.

- We may assume that $|\delta^-(v)| \leq 1$ for every pendent node v in G. (a node v is pendent if $\delta^+(v) = \emptyset$, there is no arc leaving v).

The case : G does not contain an odd Y-cycle.

Lemma 1. x(u, v) = y(v) for all (u, v) with v not a pendent node. **Lemma 2.** *G* does not contain a cycle. *Proof:*

- If G contains a cycle, then it contains a Y-cycle:











In blue: a Y-cycle.

A labeling function l assign to each node and variable the value +1, -1 or 0.

From a labeling function l and a solution (x, y) we define a new solution (x', y') as follows: $x'(u, v) = x(u, v) + l(u, v)\epsilon$ and $y'(u) = y(u) + l(u)\epsilon$, for each arc and node.



In blue: a Y-cycle.











Extending the labeling procedure.

The end of the proof

- Assume the contrary : G is a directed graph with no odd Y-cycle and does not contain none of the four following graphs as a subgraph:



- Denote by Pair(G) the set of pair of nodes $\{u, v\}$ with (u, v) and (v, u) in A.

- The proof is by induction on |Pair(G)|. If |Pair(G)| = 0 then G is an *oriented* graph that satisfies the hypothesis of our theorem , so $P_p(G)$ is integral.

From G define the graph G' with |Pair(G')| < |Pair(G)|, as follows:



- It is clear that G' does not contain none of the forbiden subgraphs. Also since by Lemma 2 G does not contain a cycle, so G' too. Thus G' satisfies the induction hypothesis.

From the extreme point (x, y) of $P_p(G)$ define $(x', y') \in P_{p+1}(G')$ as follows:

- x'(u,v) = x(u,v) for all arc (u,v) in G' different from (u,w),
- y'(u) = y(u) for all node in G' different from w,
- x'(u,w) = x(u,v) and y'(w) = 1



- By the induction hypothesis, $P_{p+1}(G')$ is integral, we have that (x', y') is not an extreme point of $P_{p+1}(G')$.

- Thus (x',y') is a convex combination of 0-1 vectors in $P_{p+1}(G')$, (x^i,y^i) $i=1,\ldots,t$,

$$(x',y') = \sum_{i=1}^{t} \lambda_i(x^i,y^i),$$

 $\sum_{i=1}^{t} \lambda_i = 1,$
 $\lambda_i \ge 0 \quad i = 1, \dots, t.$

- Each of the (x^i, y^i) satisfies the same equalities as (x', y'). Since x'(v, u) > 0, there is at least of these vectors, say (x^1, y^1) , such that $x^1(v, u) = 1$.

- In the last step, as shown in the figure below, we construct from (x^1, y^1) a solution $(x^*, y^*) \in P_p(G)$ that satifies the same equalities as (x, y). This contradicts the fact that (x, y) is an extreme point of $P_p(G)$.

