

## *Lecture*

Solving combinatorial optimization problems  
using mathematical programming

### Section 3 : Polyhedral approach

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Hanoi, Vietnam, 2024

1. Introduction and definition

2. Dimension and facet

3. Characterization

## What are the “best” valid inequalities ?

Given the variable set of a formulation,

- What are the “best” valid inequalities ?
- How to have “often” integer Branch&Bound nodes ?

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## What are the “best” valid inequalities ?

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- What are the “best” valid inequalities ?
- How to have “often” integer Branch&Bound nodes ?
- ...
- Can we know when a linear formulation produces integer solutions ?

## Integer polytope

- Solving a (bounded) linear formulation

$$(\tilde{F}) \begin{cases} \max & c^T x \\ & Ax \leq b \end{cases}$$

reduces to find an optimal extreme point  
of polytope  $P = \{x \in \mathbf{R}^n \mid Ax \leq b\}$

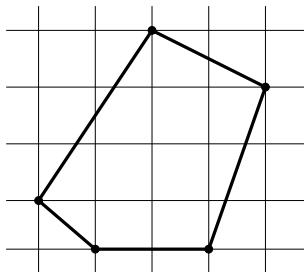
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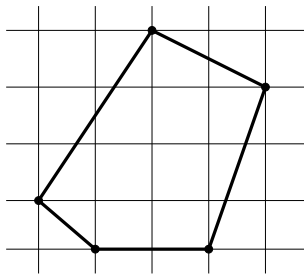
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- A rational polytope  $P$  is integer  $\Leftrightarrow \forall c \in \mathbf{Z}^n, \max\{c^T x \mid x \in P\}$  is integer.

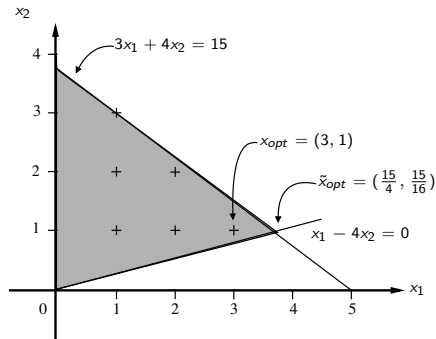
## A 2-dimensional example

$$(F) \left\{ \begin{array}{l} \max z = 2x_1 + x_2 \\ x_1 - 4x_2 \leq 0 \\ 3x_1 + 4x_2 \leq 15 \\ x_1 \geq 0 \\ x_2 \geq 0 \\ x_1, x_2 \in \mathbf{N} \end{array} \right.$$



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$x_{opt}$  optimal integer solution of  $(F)$

$z^{opt}$  optimal integer value : 7

$\tilde{x}_{opt}$  optimal fractional solution of linear relaxation  $(\tilde{F})$

$z_{opt}^*$  optimal fractional value :  $8 + \frac{7}{16}$

Note that  $\tilde{x}_{opt}$  is the (only) optimal extreme point of  $\tilde{F}$ .

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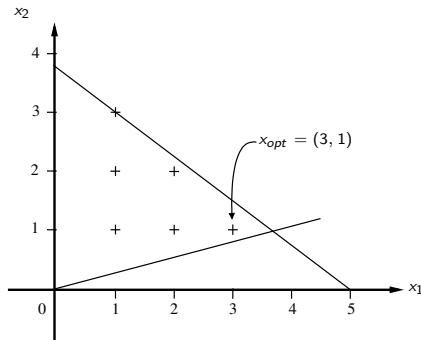
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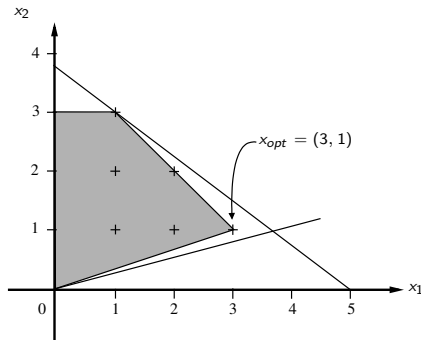
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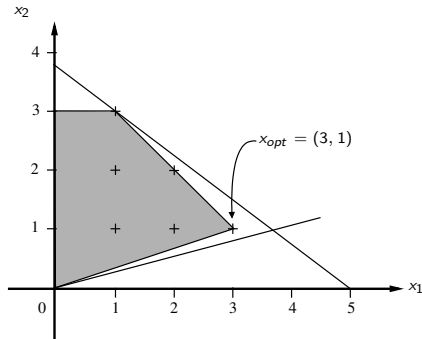


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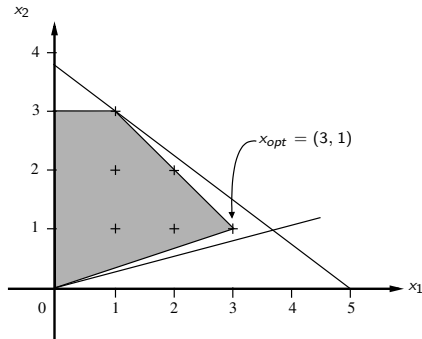
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Let's take an elastic and wrap it around these integer points...

We obtain a **new polytope** ! (the convex hull of the integer points)  
And this polytope is **integer** by construction.

## Convex hull

Given a set  $S$  of points of  $\mathbf{R}^n$ .

the **convex hull** of  $S$ , denoted by  $\text{conv}(S)$   
is the smallest convex set containing  $S$ .

### Theorem (of Minkowski)

*A set  $P \subseteq \mathbf{R}^n$  is a polytope*

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**Consequently :**

- ▶  $\text{conv}(S)$  is a polytope
- ▶ there exists a finite subset of inequalities  $Dx \leq \beta$  such that

$$\text{conv}(S) = \{x \in \mathbf{R}^n \mid Dx \leq \beta\}$$

- ▶  $\max\{c\chi \mid \chi \in \text{conv}(S)\}$  is a linear program



## Combinatorial polytope

Let  $\mathcal{P}$  be a combinatorial optimization problem :

- over  $n$  decisions corresponding to  $n$  integer variables.
- with a function cost  $c$ .

Let  $S$  the set of the incidence vectors of the solutions of  $\mathcal{P}$ .

Problem  $\mathcal{P}$  is

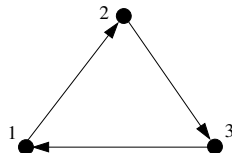
$$\max \{c\chi \mid \chi \in S\}$$

Let us consider the **linear program**

$$\max \{c\chi \mid \chi \in \text{conv}(S)\}$$

## A 3-dimensional example

Let us consider the AISP on a triangle



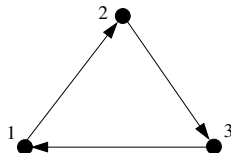
The solutions are

$$\emptyset \quad \{1\} \quad \{2\} \quad \{3\} \quad \{1,2\} \quad \{1,3\} \quad \{2,3\}$$

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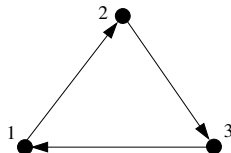
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$$\begin{aligned} \chi^{\emptyset} &= \begin{bmatrix} 0 & 0 & 0 \end{bmatrix} \\ \chi^{\{1\}} &= \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \\ \chi^{\{2\}} &= \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} \\ \chi^{\{3\}} &= \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} \\ \chi^{\{1,2\}} &= \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} \\ \chi^{\{1,3\}} &= \begin{bmatrix} 1 & 0 & 1 \end{bmatrix} \\ \chi^{\{2,3\}} &= \begin{bmatrix} 0 & 1 & 1 \end{bmatrix} \end{aligned}$$

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However point

$$\chi^{\{1,2,3\}} = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$$

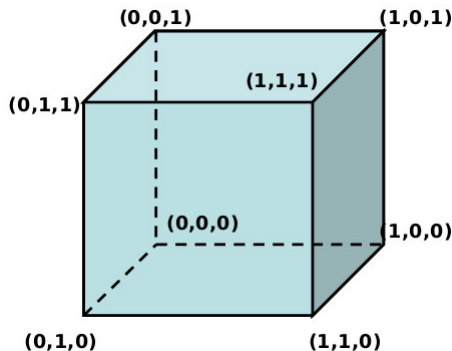
does not correspond to a solution

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**What is the convex hull of these 7 points ?**

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This convex hull is included into the hypercube (of dimension 3)

The hypercube is characterized by

$$x_1 \leq 1$$

$$x_2 \leq 1$$

$$x_3 \leq 1$$

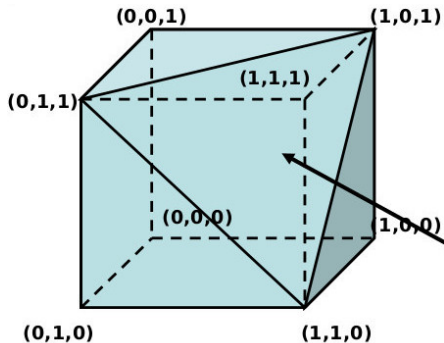
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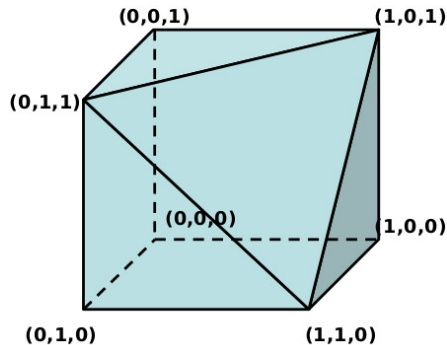
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$$x_1 + x_2 + x_3 \leq 2$$

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The convex hull of these 7 points is characterized by

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 x_1 + x_2 + x_3 &\leq 2 \\
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**Indeed**

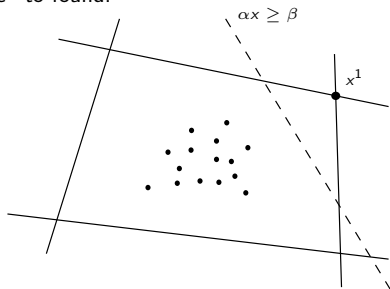
- Every extreme points of the convex hull  $\text{conv}(S)$  are integer by construction.
- The optimal points of  $S$  are among the extreme points of polytope  $\text{conv}(S)$ .

## Get around the combinatorial explosion

Optimizing  $(P)$  reduces to optimizing a linear program on  $\text{conv}(S)$ .

The convex hull is then the “unknown value” to found.

$$(F) \begin{cases} \max cx \\ Ax \leq b \\ \alpha x \leq \beta \\ x \text{ integer} \end{cases}$$



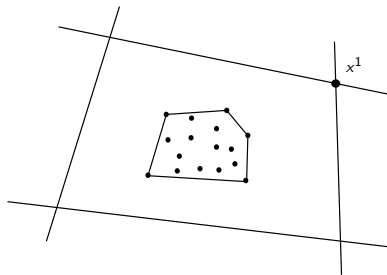
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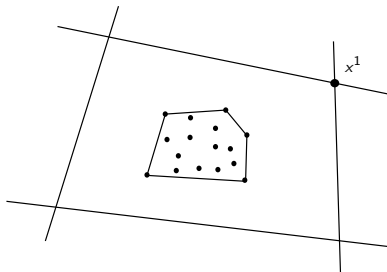
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Unfortunately we cannot use this process in polynomial time...

Unless  $P=NP$ , finding the convex hull of a combinatorial polytope is NP-hard !

But even a “partial” knowledge of polytope  $\text{conv}(S)$  is very useful

1. Introduction and definition

2. Dimension and facet

3. Characterization

## Example : The acyclic induced subgraph polytope

Given a directed graph  $G = (V, A)$ ,

$acycl(G)$  : family of all node subsets inducing an acyclic subgraph of  $G$ .

Then  $acycl(G)$  is the solutions set of the AISP on  $G$  (whatever will be the costs)

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Given a solution  $W \in acycl(G)$ , the **incidence vector**  $\chi^W$  is

$$\chi^W[i] = \begin{cases} 1 & \text{if } i \in W \\ 0 & \text{otherwise} \end{cases}$$

Some solutions :

- Pour  $\emptyset \in acycl(G)$

$$\chi^\emptyset = [ 0 \ 0 \ 0 \ 0 \ 0 \ 0 ]$$

- $\{i\} \in acycl(G) \ \forall i \in V$

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$P(G)$  : the **acyclic induced subgraph polytope** of  $G$

*i.e.* is the convex hull of the incidence vectors of the solutions

*i.e.*  $P(G) = conv\{\chi^W \mid W \in acycl(G)\}$ .

## Dimension

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### Definition

- A set of points  $x^1, \dots, x^k \in \mathbf{R}^n$  are **affinely independent** if vectors  $x^2 - x^1, \dots, x^k - x^1$  are linearly independent.

- A polytope  $P$  in  $\mathbf{R}^n$  is of **dimension  $d$**  (denoted  $\dim(P) = d$ ) if  $P$  contains at least  $d + 1$  affinely independent points.

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Examples :

- a plane in 3D-space is not full dimensional.
- an hypercube  $[0, 1]^n$  is full dimensional in  $\mathbf{R}^n$  but not in  $\mathbf{R}^{n'}$  if  $n' < n$

## Example : The acyclic induced subgraph polytope

### Lemma

*The AIS polytope  $P(G)$  is full-dimensional for every graph  $G$ .*

#### Proof.

It is sufficient to produce  $n + 1$  affinely independent points of  $P(G)$ .

For instance, the incidence vectors of

- the empty set  $\emptyset$
- the singletons  $\{i\} \forall i \in V$ .

Moreover, the vectors  $\chi^{\{u\}} - \chi^\emptyset$  are linearly independent since they form the identity matrix.

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

□

## “Degree of freedom”

Let us assume that the characterization of a polytope  $P \subseteq \mathbf{R}^n$  is given by

$$P = \left\{ x \in \mathbf{R}^n \mid \begin{array}{l} A_i x \leq b_i, \quad i = 1, \dots, m_1 \\ B_j x = d_j, \quad j = 1, \dots, m_2 \end{array} \right\}.$$

where every inequality  $A_i x \leq b_i$  is a “true” inequality,  
*i.e.* there exists  $\tilde{x} \in P$  such that  $A_i \tilde{x} < b_i$ .

### Theorem

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If  $P \neq \emptyset$ , then  $\dim(P) = n - \text{rang}(B)$ .

$\dim(P)$  gives the “degree of freedom” of a problem :

$n - \dim(P)$  variables can be obtained by fixing the  $\dim(P)$  others

## Redundant inequality

### Definition

Let  $P$  a polytope characterized by a system  $Ax \leq b$ .

An inequality  $ax \leq \alpha$  of  $Ax \leq b$  is **redundant** if the system “ $Ax \leq b$  minus  $ax \leq \alpha$ ” still characterizes  $P$ .

A non-redundant inequality is then **essential**.

What are the essential inequalities ?



## Facet of a polytope

Let  $ax \leq \alpha$  is a valid inequality for the problem corresponding to a polytope  $P$ .

### Definition

- The **face** of  $ax \leq \alpha$  is the set of points of  $P$  satisfying  $ax \leq \alpha$  to equality,

$$\text{i.e. } F = \{x \in P \mid ax = \alpha\}$$

- A face  $F$  is a **facet** of  $P$   
if  $\emptyset \neq F \neq P$  and  $\dim(F) = \dim(P) - 1$ .

### Theorem

- If  $P \neq \emptyset$ , then a non-facet inequality of  $P$  is redundant.
- Every facet of  $P$  corresponds to one inequality of a characterization of  $P$ .

## Trivial facet of the AIS polytope $P(G)$

Given a node  $i_0 \in V$ , the trivial inequality

$$x_{i_0} \geq 0$$

defines a facet of  $P(G)$ .

The corresponding face is  $F_{i_0} = \{\chi^W \in \mathbf{R}^n \mid W \in \text{acycl}(G) \text{ and } \chi^W[i_0] = 0\}$ .

- $\chi^\emptyset \in F_{i_0}$  then  $F_{i_0} \neq \emptyset$
  - $\chi^{\{i_0\}} \notin F_{i_0}$  then  $F_{i_0} \neq P(G)$
  - The vectors  $\chi^\emptyset$  and  $\chi^{\{i\}}$ ,  $i \neq i_0$ , are  $n$  affinely independent points of  $F_{i_0}$  then  $\dim(F_{i_0}) = n - 1$
- Hence  $x_{i_0} \geq 0$  is a facet of  $P(G)$ . □

## Clique inequality of the AIS polytope

Given a clique  $K$  of  $G$ , the clique inequality is

$$\sum_{i \in K} x_i \leq 1$$

If there exists  $K'$  a clique of  $G$  such that  $K \subset K'$

Then

$$\sum_{i \in K'} x_i \leq 1$$

$$-x_i \leq 0 \quad \forall i \in K' \setminus K$$

---


$$\sum_{i \in K} x_i \leq 1$$

Thus the clique inequality associated to  $K$  is redundant and not-facet defining.

## Clique inequality of the AIS polytope

### Lemma

*A clique inequality on  $K$  defines a facet if and only if  $K$  is inclusion-wise maximal.*

The consequence of this theorem is that maximal cliques are “better inequalities” to add to strengthen a cutting plane algorithm.

Indeed facet defining inequalities are called “**the deepest cuts**” !

In practice, the heuristic method we present for clique inequalities always produce maximal cliques.

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For some polynomial combinatorial problem  $\mathcal{P}$  and its associated combinatorial polytope  $P$ .

Some methods to show that a system  $Ax \leq b$  characterizes  $P$

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- ▶ to show that there is no fractional extreme point
- ▶ to show that  $A$  is a totally unimodular matrix



## Characterization

For some polynomial combinatorial problem  $\mathcal{P}$  and its associated combinatorial polytope  $P$ .

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- ▶ ... and many others (polyhedral decomposition, extended formulation+projection, critical extreme point study,...)

## Bipartite matching problem

Let  $G = (V_1 \cup V_2, E)$  be a bipartite (undirected) graph

Let  $c \in \mathbf{R}^m$  a cost associated to the edges of  $E$ .

A **matching** of  $G$  is a set of pairwise disjoint edges.

The **matching problem** on bipartite graph  $G$  is to find a matching of maximal cost.

### Theorem

*The following linear program is integer and is equiv. to the bipartite matching problem.*

$$\begin{aligned} \max \quad & \sum_{e \in E} c(e)x(e) \\ & \sum_{e \in \delta(u)} x(e) \leq 1 \quad \forall u \in V_1 \\ & \sum_{e \in \delta(u)} x(e) \leq 1 \quad \forall u \in V_2 \\ & x(e) \geq 0 \quad \forall e \in E. \end{aligned}$$

(The matrix is totally unimodular)

## The matching problem

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The **matching polytope** is the convex hull of the incidence vectors of the matchings  
i.e.

$$P_M(G) = \text{conv}\{\chi^M \in \mathbf{R}^n \mid M \text{ matching of } G\}.$$

### Theorem (Jack Edmonds (1965))

*The matching polytope is characterized by*

$$\begin{aligned} \sum_{e \in \delta(u)} x(e) &\leq 1 && \forall u \in V \\ \sum_{e \in E(S)} x(e) &\leq \frac{|S| - 1}{2} && \forall S \subseteq V \text{ with } |S| \text{ odd} \\ x(e) &\geq 0 && \forall e \in E. \end{aligned}$$