

Extended formulations in combinatorial optimization: constructions and lower bounds

Samuel Fiorini

Université libre de Bruxelles (ULB, Brussels)

VI IPCO Summer School

VIII Escuela de Verano en Matemáticas Discretas

Day 2

Today

- 1 Quick recapitulation
- 2 More examples and techniques
- 3 Communication complexity
- 4 Combinatorial lower bounds

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- **Factorization theorem [Yannakakis'91]**: $\text{xc}(P) = \text{rk}_+(S)$

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Even polytope

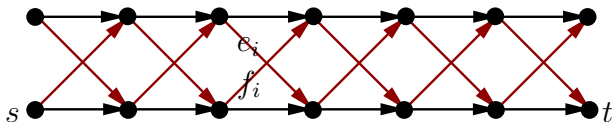
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$$\text{EVEN}(n) = \text{conv} \left\{ x \in \{0, 1\}^n \mid \sum_{i=1}^n x_i \equiv 0 \pmod{2} \right\}$$

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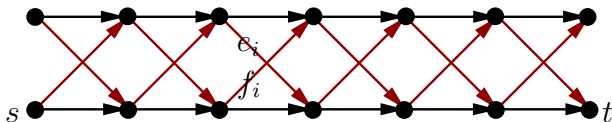
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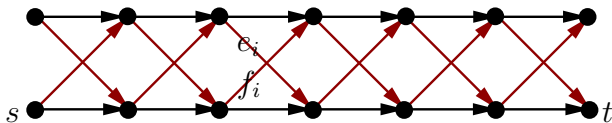


$$\text{EVEN}(n) = \{x \in \mathbb{R}^n \mid \exists y \text{ } s\text{-}t \text{ flow of value } 1 : x_i = y_{e_i} + y_{f_i} \forall i\}$$

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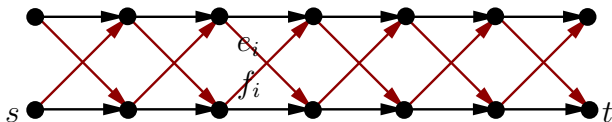
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Exercise. What are the facet-defining inequalities of $\text{EVEN}(n)$?

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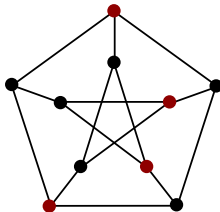
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Exercise. Generalize this to all **regular languages** (that is, languages decided by a deterministic finite automaton).

Stable set polytope

Stable set polytope of graph $G = (V, E)$:

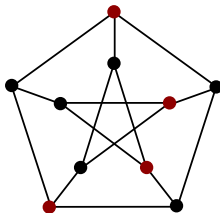
$$\text{STAB}(G) := \text{conv}\{\chi^S \in \{0, 1\}^V \mid S \text{ stable set of } G\}$$



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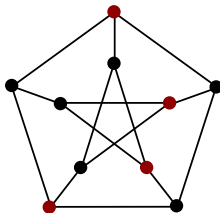
We have $\text{STAB}(G) \subseteq \text{QSTAB}(G)$ where

$$\text{QSTAB}(G) := \{x \in \mathbb{R}^V \mid x \geq 0, x(K) \leq 1 \ \forall \text{ clique } K \text{ of } G\}$$

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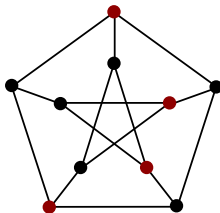
Theorem (**Chvátal'75**)

$\text{STAB}(G) = \text{QSTAB}(G)$ if and only if G is *perfect*

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$\chi(H) = \omega(H)$ for all $H \stackrel{\text{ind}}{\subseteq} G$ (χ = chromatic #; ω = clique #)

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$P = (V, \leqslant_P)$ partially ordered set (**poset**)

$G = G(P)$ **comparability graph** of P (always perfect)

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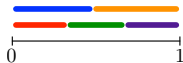
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- $\forall v \in V: (y_{v-}, y_{v+})$ open interval $\subseteq (0, 1)$
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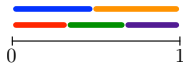
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$$\text{STAB}(G(P)) = \left\{ x \in \mathbb{R}^V \mid \exists \{(y_{v-}, y_{v+})\}_{v \in V} \text{ consistent with } P \right. \\ \left. \text{s.t. } x_v = y_{v+} - y_{v-} \quad \forall v \in V \right\}$$

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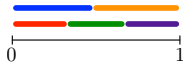
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Exercise. Prove this.

Union of polyhedra

Theorem (**Balas**'98, 85)

Consider k polyhedra $P^i := \{x \in \mathbb{R}^d \mid A^i x \leq b^i\}$. For each i :

- $\text{rec}(P^i) := \{x \mid A^i x \leq 0\} = \text{cone}\{r_1^i, \dots, r_{q_i}^i\}$
- $P^i = \text{conv}\{v_1^i, \dots, v_{p_i}^i\} + \text{cone}\{r_1^i, \dots, r_{q_i}^i\}$ (for $P^i \neq \emptyset$)

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Then

$$\begin{aligned} A^i x^i &\leq \delta^i b^i & \forall i \\ \sum_{i \in K} x^i &= x \\ \sum_{i \in K} \delta^i &= 1 \\ \delta^i &\geq 0 & \forall i \end{aligned}$$

is an EF of

$$P := \text{conv}\left(\bigcup_{i: P^i \neq \emptyset} \{v_1^i, \dots, v_{p_i}^i\}\right) + \text{cone}\left(\bigcup_i \{r_1^i, \dots, r_{q_i}^i\}\right)$$

Union of polyhedra

... an application to vertex interdiction

Theorem (Union of polyhedra, restated)

If P_1, \dots, P_k are polyhedra in \mathbb{R}^d such that $\dim(P_i) \geq 1$ for all i , and $P := \text{conv}(P_1 \cup \dots \cup P_k)$:

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Theorem (Ahmed, Angulo, Dey & Kaibel'13)

Let $V \subseteq \{0, 1\}^d$ and $P = \text{conv}(V)$ be a 0/1-polytope. For a point $v \in V$, let $P - v := \text{conv}(V \setminus \{v\})$. Then:

$$xc(P - v) \leq d \, xc(P)$$

and more generally for $v_1, \dots, v_k \in V$:

$$xc(P - v_1 - v_2 - \dots - v_k) \leq \text{poly}(xc(P), k)$$

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Deterministic communication protocols

$f : A \times B \rightarrow \{0, 1\}$ Boolean function (\equiv binary matrix)

Two players:

- Alice knows $a \in A$
- Bob knows $b \in B$

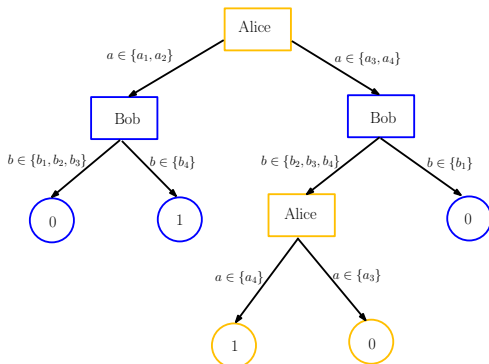
want to **compute** $f(a, b)$ **by exchanging bits**

Goal: Minimize **complexity** $:=$ #bits exchanged

Deterministic communication protocols

Example

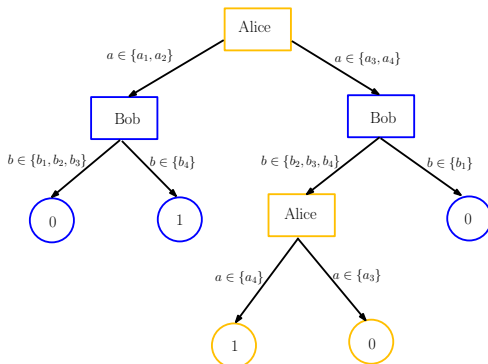
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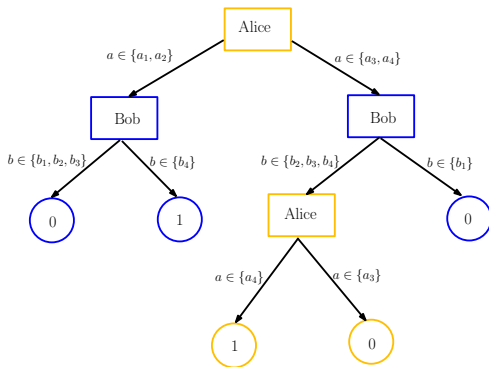
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Observation

\exists complexity c protocol for computing $f \implies \text{rk}_+(f) \leq 2^c$

Deterministic communication protocols

The clique vs stable set problem

G graph with n vertices

$A = \{a \in \{0, 1\}^n \mid a \text{ encodes a clique in } G\}$

$B = \{b \in \{0, 1\}^n \mid b \text{ encodes a stable set in } G\}$

$$f(a, b) = \begin{cases} 1 & \text{if } a, b \text{ are disjoint} \\ 0 & \text{if } a, b \text{ intersect} \end{cases} = 1 - a^\top b = (1 - a^\top b)^2$$

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$\exists O(\log^2 n)$ -complexity protocol for $f = f(G)$

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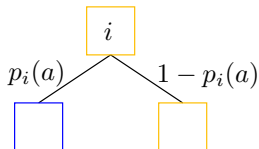
Corollary (**Yannakakis'91**)

\forall perfect graphs G : $xc(STAB(G)) = 2^{O(\log^2 n)} = n^{O(\log n)}$

Computing a function in expectation

The main differences:

- Alice and Bob can use (private) random bits to make choices

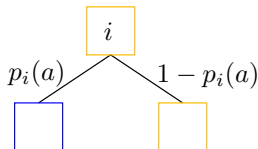


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Theorem (Faenza, F, Grappe & Tiwary'11)

If $c = c(f)$ is the minimum complexity of a randomized communication protocol with nonnegative outputs computing f in expectation, then

$$rk_+(f) = \Theta(2^c)$$

Computing a function in expectation

Proof of $\text{rk}_+(M) = \Omega(2^c)$

$$\text{Write } M = TU, \quad \text{where} \quad \left| \begin{array}{l} T \in \mathbb{R}_+^{m \times r} \text{ row-stochastic (w.l.o.g.)} \\ U \in \mathbb{R}_+^{r \times n} \\ r \leq \text{rk}_+(M) + 1 \end{array} \right.$$

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- Alice gets row index i , Bob gets column index j
- Alice picks random column index $k \in [r]$ w.p. T_{ik} , sends it to Bob
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$$\text{Expected value on input } (i, j): \quad \sum_{k=1}^r T_{ik} U_{kj} = M_{ij}$$

$$\text{Complexity:} \quad \log \text{rk}_+(M) + O(1)$$

The threefold way

Three **equivalent** ways to look at EFs:

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- 3 A $\log r$ -complexity randomized protocol computing S in expectation

- 1 Quick recapitulation
- 2 More examples and techniques
- 3 Communication complexity
- 4 Combinatorial lower bounds

Rectangle covering bound

Taking supports

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Theorem (Yannakakis'91)

$$\text{rk}_+(S) \geq \text{rc}(S)$$

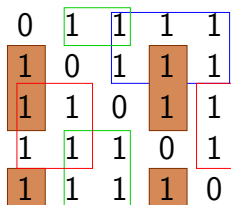
Rectangle covering bound

... also known as

0	1	1	1	1
1	0	1	1	1
1	1	0	1	1
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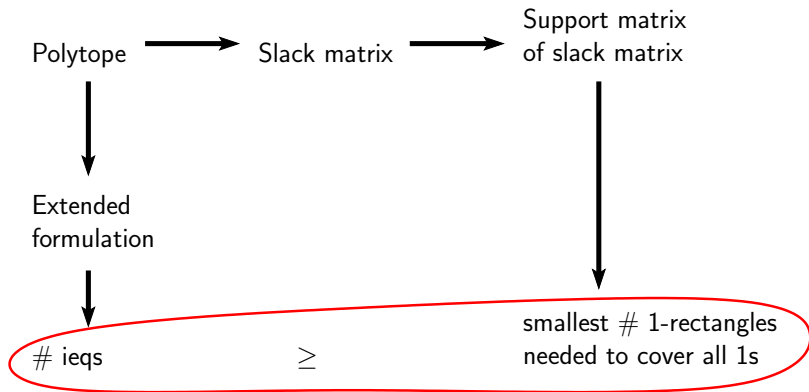
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Observation

$\text{rc}(S)$ is the best lower bound that only depends on $\text{supp}(M)$!

Rectangle covering bound

Summary



Proposition (**Goemans**'09)

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Exercise. Use this to give lower bounds on the extension complexities of n -gons and permutahedra.

The rectangle graph

Definition

M 0/1-matrix

$G_M := \text{rectangle graph}$:

- $V(G_M) :=$ 1-entries of M

Example

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \quad \begin{array}{ccc} \bullet & & \\ & \bullet & \bullet \\ \bullet & \bullet & \\ \bullet & \bullet & \bullet \end{array}$$

The rectangle graph

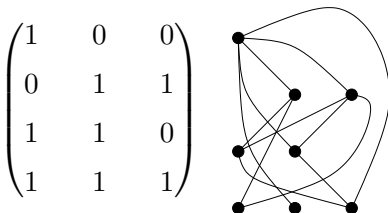
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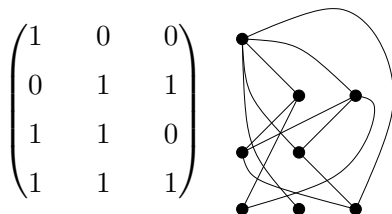
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Lemma (F, Kaibel, Pashkovich, Theis'13)

$$\text{rc}(M) = \chi(G_M)$$

Clique number — aka fooling sets

Chromatic number \geq maximum size of a clique

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In the matrix, clique \equiv fooling set $:=$

- set of 1-entries

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Example: n -gon

rectangle covering number $\approx \log n$

max size of fooling set = 4

Example: Cubes

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Every extension of a d -cube has at least $2d$ facets.

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Support of the slack matrix:

- Columns: $\xi \in \{0, 1\}^d$
- Rows: d for $x_i \geq 0$ (upper part) and d for $x_i \leq 1$ (lower part)
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0	0	0	0	1	1	1	1	1	1	1	1	0	0
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0	0	0	0	0	0	0	1	1	1	1	1	0	0

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1	1	1	0	0	0	0	0	0	0	0	1	1	1
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1	1	1	1	0	0	0	0	0	0	0	0	1	1
1	1	1	1	1	0	0	0	0	0	0	0	0	1
1	1	1	1	1	1	0	0	0	0	0	0	0	0

What about $\log_2(\# \text{faces})$?

- $\# \text{faces} \approx 3^d$
- $\# \text{ facets of an extension}$
 $\geq rc \geq d \log_2 3$ ($\log_2 3 = 1.585\dots$)

Proposition (**F, Kaibel, Pashkovich, Theis'13**)

For M support matrix of slack matrix of d -polytope,

$$\omega(G_M) \leq (d+1)^2$$