

*Recent Developments in Cuts for MILP*  
*2. Theoretical comparisons*

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## *Mixed-Integer Linear Program (MILP)*

$$\begin{array}{ll}\max & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} = \mathbf{b} \\ & x_i \in \mathbb{Z} \quad \text{for } i \in I \\ & x_i \geq 0 \quad \text{for } i \in [n]\end{array}$$

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2-dimensional Relaxed Corner Polyhedron:

$$\begin{aligned}(RCP) : \quad \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} + \sum_{j \in N} \begin{pmatrix} r_1^j \\ r_2^j \end{pmatrix} x_j \\ &x_j \geq 0 \text{ for all } j \in N \\ &x_1, x_2 \in \mathbb{Z}\end{aligned}$$

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$$\Gamma = \{r^j \mid j \in N\}$$

$$R(f, \Gamma) = \text{convex hull of (RCP)}$$

(LRCP): Linear relaxation of (RCP)

## *Description of $R(f, \Gamma)$*

Six relaxations of  $R(f, \Gamma)$ :

- $S(f, \Gamma) = (\text{LRCP}) + \text{all cuts from splits containing } f \text{ in their interior}$
- $\triangle(f, \Gamma) = (\text{LRCP}) + \text{all cuts from triangles containing } f \text{ in their interior}$
- $\triangle_i(f, \Gamma) = (\text{LRCP}) + \text{all cuts from triangles of type } i \text{ containing } f \text{ in their interior, for } i = 1, 2, 3$
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We have:

$$R(f, \Gamma) = S(f, \Gamma) \cap \Delta(f, \Gamma) \cap \square(f, \Gamma)$$

## *Comparison of split, triangle and quadrilateral cuts*

Theoretical comparison:

- Upper and lower bounds on quality of relaxations vs.  $R(f, \Gamma)$
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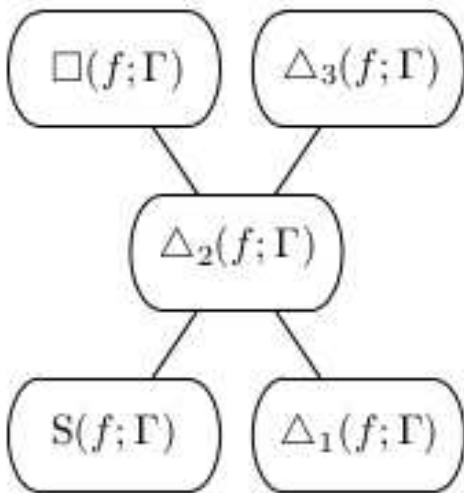
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- Heuristic separation algorithm for some families of cuts
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Probabilistic comparison:

- Over all choices of  $f$  and  $\Gamma$  in a set  $L$ :
  - Probability that  $L$  generates an inequality improving on the split closure
  - Prob. that some inequalities dominates others (coefficients or volume cut off)
- Select  $f$  and  $\Gamma$  uniformly; compare average and worst-case strength of split closure vs. triangle closure over all possible optimization directions

## *Comparing families*



## *Domination between families*

$L \subseteq \mathbb{R}^n$  : lattice-free set;       $\epsilon > 0$

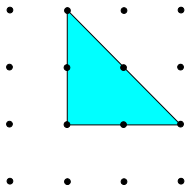
$$\textit{relax}(L, \epsilon) = \{x \in \mathbb{R}^n \mid \|x - \bar{x}\| \leq \epsilon, \text{ for some } \bar{x} \in L\}$$

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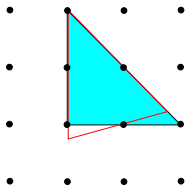


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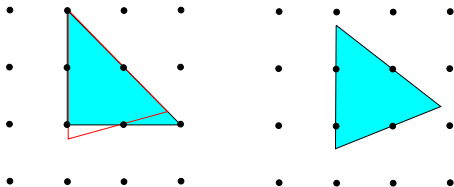
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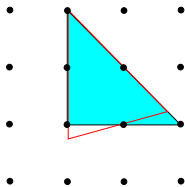
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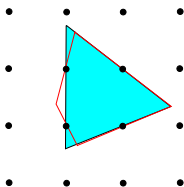
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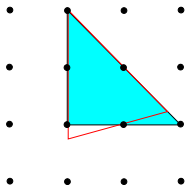
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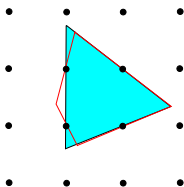
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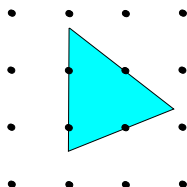
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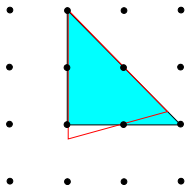


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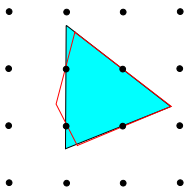
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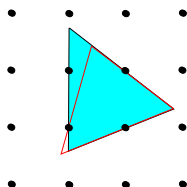
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$$\Delta_3(f, \Gamma) \subseteq \Delta_2(f, \Gamma)$$

## *Comparing relaxations*

$C \subseteq \mathbb{R}_+^n$ : convex set is **monotone** when

$$x \in C \text{ and } y \geq x \quad \Rightarrow \quad y \in C$$

Example:  $R(f, \Gamma)$ ,  $S(f, \Gamma)$ ,  $\Delta(f, \Gamma)$ ,  $\square(f, \Gamma)$

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$C_1, C_2$ : monotone convex sets

How much to inflate  $C_2$  to contain  $C_1$ :

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## *Computing an upper bound on $\rho(\mathcal{L}_1, \mathcal{L}_2)$*

We have

$$\frac{1}{\rho(\mathcal{L}_1, \mathcal{L}_2)} = \inf_{\Gamma, f \in L \in \mathcal{L}_2} \sum_{j \in \Gamma} \psi_L(r^j) x_j$$

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$\Rightarrow \{r^j \mid j \in \Gamma\}$  is exactly  $\Gamma(L) :=$  set of corner rays for  $L$

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Difficulties:

- Need to check all  $L \in \mathcal{L}_2$
- For each  $L \in \mathcal{L}_2$  one inequality for each  $C \in \mathcal{L}_1$

## *Unimodular transformation*

Unimodular transformation:  $\phi : \mathbb{R}^n \rightarrow \mathbb{R}^n$  with

$$\phi(x) = v + Mx$$

where

$$v \in \mathbb{Z}^n$$

$$M \in \mathbb{Z}^{n \times n} \text{ with } \det(M) = \pm 1$$

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Need to check all  $L \in \mathcal{L}_2$  only up to unimodular transformation

## *Computing an upper bound on $\rho(S, \Delta_1)$*

All  $L \in \Delta_1$  are identical up to unimodular transformation.

We can just pick one.

$T$ : vertices  $(0, 0), (0, 2), (2, 0)$

$T_I$ : vertices  $(1, 0), (0, 1), (1, 1)$

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$$\begin{aligned} z = \min \quad & \sum_{j=1}^3 x_j \\ & \sum_{j=1}^3 \psi_S(r^j) x_j \geq 1 \quad \text{for all split } S \text{ containing } f \\ & x_j \in \mathbb{R}_+ \quad \text{for all } j = 1, 2, 3 \end{aligned}$$

Then  $\frac{1}{z} = \rho(S, \Delta_1)$

## *Computing an upper bound on $\rho(S, \Delta_1)$ (cont.)*

Replace the infinite number of inequalities in the LP by a small number:

$S_1$ : split  $0 \leq x_1 \leq 1$

$S_2$ : split  $0 \leq x_2 \leq 1$

$S_3$ : split  $1 \leq x_1 + x_2 \leq 2$

$$\begin{aligned} z = \min \quad & \sum_{j=1}^3 x_j \\ & \sum_{j=1}^3 \psi_{S_t}(r^j) x_j \geq 1 \quad \text{for } t = 1, 2, 3 \\ & \tilde{x}_j \in \mathbb{R}_+ \quad \text{for all } j = 1, 2, 3 \end{aligned}$$

Then  $\frac{1}{z} \geq \rho(S, \Delta_1)$



## Computing an upper bound on $\rho(S, \Delta_1)$ (cont.)

Assume that  $f \in T_I$

$$\begin{aligned}
 z = \min \quad & x_1 + x_2 + x_3 \\
 & \frac{f_1 + f_2}{f_1 + f_2 - 1} x_1 + x_2 + x_3 \geq 1 \\
 & x_1 + \frac{2 - f_1}{1 - f_1} x_2 + x_3 \geq 1 \\
 & x_1 + x_2 + \frac{2 - f_2}{1 - f_2} x_3 \geq 1 \\
 & x \geq 0
 \end{aligned}$$

## Computing an upper bound on $\rho(\mathcal{S}, \Delta_1)$ (cont.)

Assume that  $f \in T_I$

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Optimal solution:

$$x_1^* = \frac{f_1+f_2-1}{2} \quad x_2^* = \frac{1-f_1}{2} \quad x_3^* = \frac{1-f_2}{2}$$

$$\text{with value } z = x_1^* + x_2^* + x_3^* = \frac{1}{2}$$

## *Computing a lower bound on $\rho(S, \Delta_1)$*

**Lemma 6.3** [43]: If  $f$  is in the interior of triangle  $T_I$  then the split closure is defined by  $S_1, S_2, S_3$ .

## *Computing a lower bound on $\rho(S, \Delta_1)$*

**Lemma 6.3** [43]: If  $f$  is in the interior of triangle  $T_I$  then the split closure is defined by  $S_1, S_2, S_3$ .

In general, to prove a lower bound on  $\rho(\mathcal{L}_1, \mathcal{L}_2)$ :

- Select  $L \in \mathcal{L}_2, f$
- Find a point  $\bar{x}$  in  $\mathcal{L}_1(f, \Gamma(L))$
- $z = \sum_{j \in \Gamma(L)} \psi_L(r^j) \bar{x}_j$
- $\frac{1}{z}$  is a lower bound on  $\rho(\mathcal{L}_1, \mathcal{L}_2)$

## *Lower, upper bounds on $\rho$*

Entry  $(i,j)$ : lower bound, upper bound on  $\rho(i,j)$

	$S$	$\Delta_1$	$\Delta_2$	$\Delta_3$	$\square$	$R$
$S$	—	2	$+\infty$	$+\infty$	$+\infty$	$+\infty$
$\Delta_1$	$+\infty$	—	$+\infty$	$+\infty$	$+\infty$	$+\infty$
$\Delta_2$	1	1	—	1.125, 1.5	1.125, 1.5	1.125, 1.5
$\Delta_3$	1	1	1	—	1.125, 1.5	1.125, 1.5
$\square$	1	1	1	1.125, 1.5	—	1.125, 1.5

[Awate, Cornuéjols, Guenin, Tuncel 2013 [ACGT]]

## *Probabilistic comparisons (model (i))*

[Del Pia, Wagner, Weismantel 2011 [DPP]]

- Fix a lattice-free convex set  $L$
- Let  $f$  vary uniformly in  $\text{int}(L)$  and use  $\Gamma(L)$  as rays.
- For any  $z > 1$ , compute  $P^L(z) := \text{Prob}(\rho(S, S \cup L) \leq z)$ .

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Findings:

- $P^L(z)$  tends to 1 when lattice-width of  $L \in \Delta_2$  tends to 1

$$P^L(z) \geq \begin{cases} 0 & \text{if } 1 < z \leq w \\ \frac{(z-w)(2zw-w-z)}{w^2(z-1)^2} & \text{if } w < z \leq \frac{w}{w-1} \\ \frac{(z-w)(2zw-w-z) + (w-1)^2(z-1)^2 - 1}{w^2(z-1)^2} & \text{if } \frac{w}{w-1} < z < +\infty \end{cases}$$

(similar result for  $L \in \Delta_3$ )

## *Probabilistic comparisons (model (ii))*

[Basu, Cornuéjols, Molinaro 2010 [BCM]]

- Let  $f$  and  $\Gamma$  be selected uniformly
- Compare average (avg) and worst-case (wc) gap between closures over all possible cost vector.
- For a cost vector  $c \in \mathbb{R}_+^N$

$$gap(C_1, C_2, c) = \frac{\{\min c x \mid x \in C_1\}}{\{\min c x \mid x \in C_2\}}$$

with value  $+\infty$  if  $C_1 = \emptyset$  or  $\{\min c x \mid x \in C_2\} = 0$ .



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Findings:

- For  $\alpha \geq 1$  and  $|\Gamma| \rightarrow +\infty$ :  $Prob(wc(\triangle, S) \geq \alpha) \approx \frac{1}{\alpha} - \frac{1}{4\alpha^2}$
- For  $\alpha > 1$  and  $\epsilon > 0$  (lower bound on entries in  $c$ ):  
 $Prob(avg(R(f, \Gamma), S(f, \Gamma)) \leq \alpha) \geq 1 - \frac{1}{|N|}$

## *Probabilistic comparisons (model (iii))*

[He, Ahmed, Nemhauser 2011 [HAN]]

- Let  $f$  vary uniformly in the unit square  $U$  with uniformly distributed rays.
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- One of the two splits is more likely to dominate a Type 1 triangle than the opposite. Probability tends fast to 0 as number of rays increases.

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### Findings:

- One of the two splits is more likely to dominate a Type 1 triangle than the opposite. Probability tends fast to 0 as number of rays increases.
- Same conclusion for volume cut off. Probability tends to 1 as number of rays increases.