

The Strength of Multi-Row Relaxations

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August 2012

Motivations

- ▶ Cuts viewed as facets of relaxations of the problem
- ▶ In particular, multi-row relaxations
- ▶ Focus on **exact** separation
- ▶ Evaluate any relaxation

Plan

- A. Separation over arbitrary mixed-integer sets
- B. Application to two-row relaxations

A. Separation over arbitrary mixed-integer sets

Problem

Given

- ▶ a mixed-integer set $P \subseteq \mathbb{R}^n$,
- ▶ a point $x^* \in \mathbb{R}^n$,

find $(\alpha, \alpha_0) \in \mathbb{R}^{n+1}$ such that $\alpha^T x \geq \alpha_0$ is a valid inequality for P that separates x^* ,

or show that $x^* \in \text{conv}(P)$.

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General framework

Solve the optimization problem

$$\begin{aligned} \min \quad & x^{*T} \alpha \\ \text{s.t.} \quad & x^T \alpha \geq \alpha_0 \quad \text{for all } x \in P \\ & \langle \text{norm.} \rangle \end{aligned}$$

Let $(\bar{\alpha}, \bar{\alpha}_0)$ be the optimal solution.

If $x^{*T} \bar{\alpha} < \bar{\alpha}_0$, then $(\bar{\alpha}, \bar{\alpha}_0)$ separates x^* .

If $x^{*T} \bar{\alpha} \geq \bar{\alpha}_0$, then $x^* \in \text{conv}(P)$.

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Row generation

1. Consider the relaxation of the separation problem

$$\begin{array}{ll} \min & x^{*T} \alpha \\ \text{s.t.} & x^T \alpha \geq \alpha_0 \quad \text{for all } x \in S \subseteq P \\ & \langle \text{norm.} \rangle \end{array} \quad (\text{master})$$

Let $(\bar{\alpha}, \bar{\alpha}_0)$ be an optimal solution.

2. Now solve the MIP

$$\begin{array}{ll} \min & \bar{\alpha}^T x \\ \text{s.t.} & x \in P \end{array} \quad (\text{slave})$$

and let x^P be a finite optimal solution.

If $\bar{\alpha}^T x^P \geq \bar{\alpha}_0$, then $(\bar{\alpha}, \bar{\alpha}_0)$ is valid for P .

If $\bar{\alpha}^T x^P < \bar{\alpha}_0$, then $S := S \cup \{x^P\}$.

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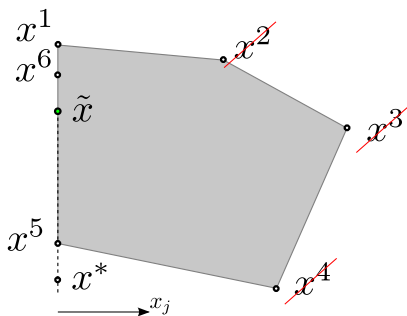
Computational example

Instance: bell13a
Constraints: 123
Variables: 133 (71 integer: 32 general, 39 binaries)
Models: 82 five-row models read from an optimal tableau

Cuts: 37 (17 tight at the end)
Gap closed: 59.02% (from 39.03% by GMIs)

Time:	1615.70s
Iterations:	107647

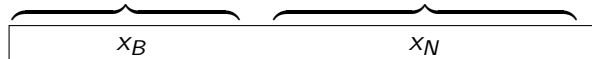
Two-phases: Phase one



x^* between bounds

x^* at bounds

$x :$

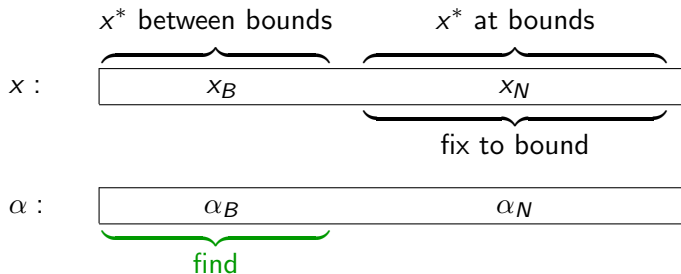
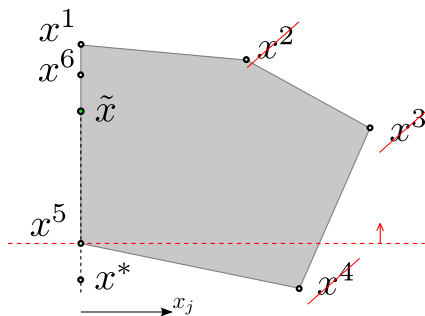


fix to bound

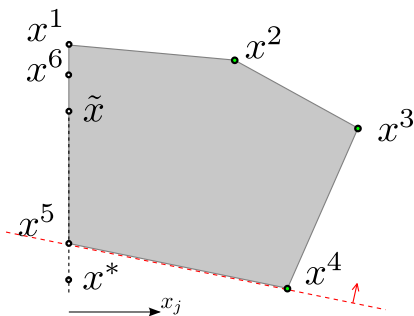
$\alpha :$



Two-phases: Phase one



Two-phases: Phase two



x^* between bounds

x^* at bounds

$x :$



fix to bound

$\alpha :$



fixed

lift

Two-phases summary

- ▶ The feasible region of phase-1 slave is $P \cap \{x : x_N = x_N^*\}$
- ▶ “phase-1 separates” iff “phase-2 separates”
→ whenever $x^* \in \text{conv}(P)$, phase-2 is avoided
- ▶ Optimal objective function values are the same
→ phase-2 master objective function is 0

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Computational example (2-phases)

(bell13a, 82 five-row models, 37 cuts, 59.02%gc)

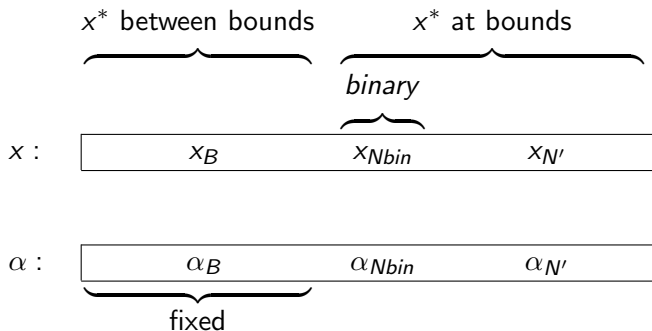
	original	2-phases
Time:	1615.70s	161.15s
Iterations:	107647	23822

Computational example (2-phases)

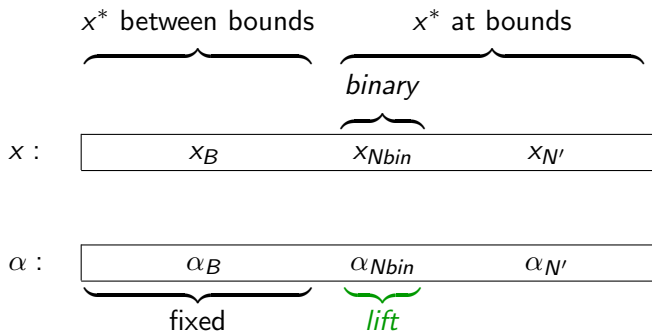
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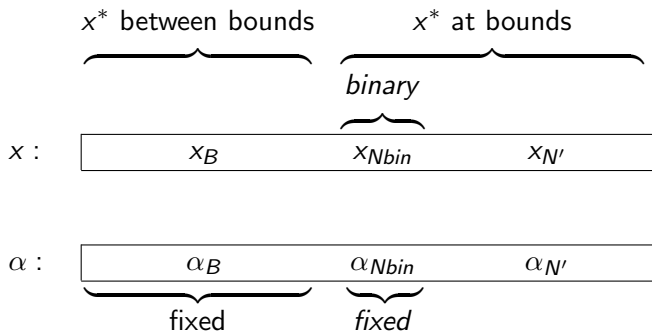
Lifting binary variables



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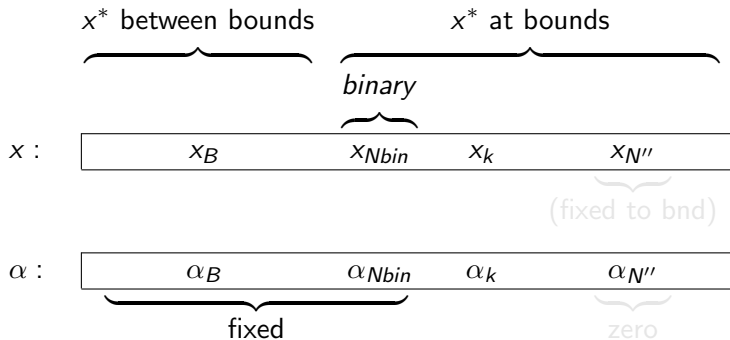
	original	2-phases	lifting
Time:	1615.70s	161.15s	136.54s
Iterations:	107647	23822	23231

Computational example (lifting binaries)

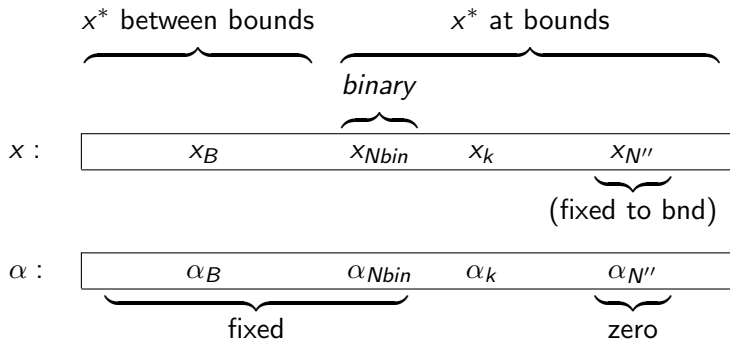
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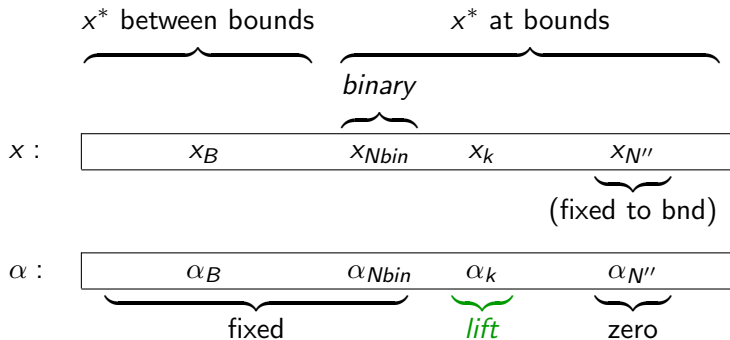
Sequential phase-2 (“phase-S”)



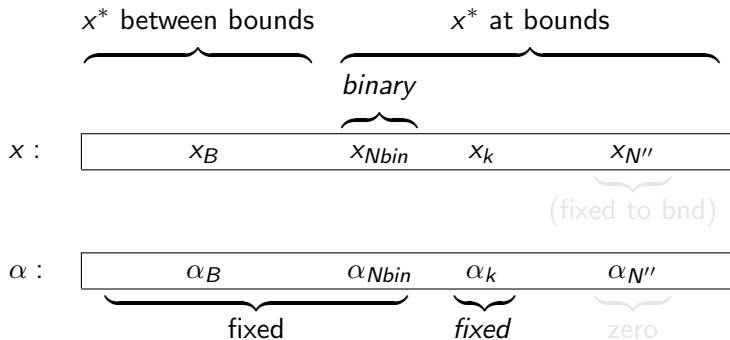
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(bell13a, 82 five-row models, 37 cuts, 59.02%gc)

	original	2-phases	lifting	phase S
Time:	1615.70s	161.15s	136.54s	5.84s
Iterations:	107647	23822	23231	2497

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Solver tricks: callbacks

Solving slave MIPs

$$\begin{array}{ll} \min & \bar{\alpha}^T x \\ \text{s.t.} & x \subseteq P, \end{array}$$

- ▶ Feasible solution \hat{x} with $\bar{\alpha}^T \hat{x} < \bar{\alpha}_0$
→ \hat{x} can be added to S .
- ▶ Dual bound \underline{z} reaches $\bar{\alpha}_0$,
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Computational example (solver tricks)

(bell13a, 82 five-row models, 37 cuts, 59.02%gc)

	original	2-phases	lifting	phase S	cb
Time:	1615.70s	161.15s	136.54s	5.84s	4.65s
Iterations:	107647	23822	23231	2497	2497

Computational example (solver tricks)

(bell13a, 82 five-row models, 37 cuts, 59.02%gc)

	original	2-phases	lifting	phase S	cb
Time:	1615.70s	161.15s	136.54s	5.84s	4.65s
Iterations:	107647	23822	23231	2497	2497

Computational example (summary)

(bell13a, 82 five-row models, 37 cuts, 59.02%gc)

	original	2-phases	lifting	phase S	cb
Time:	347× 1615.70s	35× 161.15s	29× 136.54s	1.26× 5.84s	1 4.65s
Iterations:	107647 43×	23822 10×	23231 9×	2497 1	2497 1

B. Application to two-row relaxations

Objectives

Mainly, evaluate and compare

- ▶ the intersection cut model
- ▶ a few strengthenings of it
- ▶ a full two-row model

Two-row relaxation: which models?

- ▶ We are still far from a closure
- ▶ What reasonable set of two-models can we select?
 - All models read from a simplex tableau
 - $O(m^2)$ two-row models

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“all” two-row models: separation loop

Let $x^* \leftarrow$ LP optimum

Read the two-row models from optimal tableau.

Read and add GMIs from that tableau.

do {

 Let $x^* \leftarrow$ new LP optimum.

 Separate x^* with the two-row models.

} **while** (cuts were found).

“all” two-row models: separation loop

Let $x^* \leftarrow$ LP optimum

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“all” two-row models: results

Computations on the 62 MIPLIB 3.0 (preprocessed) instances for which

- (a). the integrality gap is not zero, and
- (b). the optimal MIP solution is known.

“all” two-row models: results

We have a result for 55/62 instances (4 numerical, 3 memory).

	cuts	gc%
GMI	24.800	22.60%
All 2-row	72.382	37.49%

For 13 instances, the separation is exact.

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Heuristic selection of two-row models

Issue:

- ▶ $O(m^2)$ is already a large number of models

Hypothesis:

- ▶ Not all models are necessary to achieve good separation

Rationale:

- ▶ MIPLIB models are mostly sparse
- ▶ Multi-cuts from rows with no common support are linear combinations of the corresponding one-row cuts

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Heuristic selection of two-row models: results

With an arbitrary limit of m two-row models,
we have a result for 58/62 instances (1 numerical, 3 memory).

On the 55 common results,

	cuts	gc%
GMI	24.800	22.60%
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Heuristic	57.418	35.19%

For 25 instances, the separation is exact.

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Two-row intersection cuts

$$\begin{array}{rcccccc} \underbrace{}_{\text{basic}} & & \underbrace{}_{\text{nonbasic}} & & & & \\ x_1 & & +2 x_3 & - x_4 & + x_5 & +3 x_6 & = 2.4 \\ & x_2 & - x_3 & & - x_5 & +2 x_6 & = -0.2 \end{array}$$
$$x_1, x_2, x_3, x_4, x_5, x_6 \in \mathbb{Z}$$
$$0 \leq x_1 \leq 1$$
$$0 \leq x_2 \leq 1$$
$$0 \leq x_3 \leq 1$$
$$0 \leq x_4 \leq 1$$
$$0 \leq x_5 \leq 1$$
$$0 \leq x_6 \leq 1$$

Two-row intersection cuts

$$\begin{array}{rccclcl} \text{basic} & & \text{nonbasic} & & & & \\ \underbrace{} & & \underbrace{} & & & & \\ x_1 & & +2 x_3 - x_4 + x_5 + 3 x_6 & = & 3 & & \\ & x_2 & - x_3 - x_5 + 2 x_6 & = & -1 & & \\ & & x_1, x_2, x_3, x_4, x_5, x_6 \in \mathbb{Z} & & & & \\ & & 0 \leq x_1 \leq 1 & & & & \\ & & 0 \leq x_2 \leq 1 & & & & \\ & & 0 \leq x_3 \leq 1 & & & & \\ & & 0 \leq x_4 \leq 1 & & & & \\ & & 0 \leq x_5 \leq 1 & & & & \\ & & 0 \leq x_6 \leq 1 & & & & \end{array}$$

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$x_1, -, x_3, x_4, x_5, x_6 \in \mathbb{Z}$

$$\begin{aligned} 0 &\leq x_1 \leq 1 \\ 0 &\leq x_2 \leq 1 \\ 0 &\leq x_3 \leq 1 \\ 0 &\leq x_4 \leq 1 \\ 0 &\leq x_5 \leq 1 \\ 0 &\leq x_6 \leq 1 \end{aligned}$$

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Two-row intersection cuts + strengthening

	basic		nonbasic	
	$\in \mathbb{Z}$	bnd.	$\in \mathbb{Z}$	bnd.
P_I	✓	×	×	B
S -free	✓	✓	×	B
lifting	✓	×	✓	B
P_{IW}	✓	×	×	✓
full 2-row	✓	✓	✓	✓

✓: keep
B: keep binding
×: drop

Two-row intersection cuts + strengthening

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P_{IU}	✓	×	×	✓
full 2-row	✓	✓	✓	✓

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lifting	✓	×	✓	B
P_{IU}	✓	×	×	✓
full 2-row	✓	✓	✓	✓

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full 2-row	✓	✓	✓	✓

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Two-row intersection cuts + strengthening

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P_I	✓	×	×	B
S -free	✓	✓	×	B
lifting	✓	×	✓	B
P_{IU}	✓	×	×	✓
full 2-row	✓	✓	✓	✓

✓: keep
B: keep binding
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Two-row intersection cuts and strengthenings

51 common instances:

	cuts	gc%	exact
GMI	28.240	22.46%	all
P_I	29.420	27.65%	42
S -free	38.380	30.22%	29
lifting	22.700	27.35%	10
P_{IU}	42.640	28.56%	25
full 2-row	55.500	35.66%	22

Two-row intersection cuts and strengthenings

51 common instances:

	cuts	gc%	exact
GMI	28.240	22.46%	all
P_I	29.420	27.65%	42
S-free	38.380	30.22%	29
lifting	22.700	27.35%	10
P_{IU}	42.640	28.56%	25
full 2-row	55.500	35.66%	22

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51 common instances:

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GMI	28.240	22.75%	all
P_I	29.420	27.65%	42
S-free	38.380	30.22%	29
lifting	22.790	27.35%	10
P_{IU}	22.640	26.56%	25
full 2-row	55.500	35.60%	22

Two-row intersection cuts and strengthenings

15 common instances:

	cuts	gc%	exact
GMI	20.667	26.541	all
P_I	20.933	33.535	all
S -free	25.400	35.229	all
P_{IU}	36.600	36.257	all
full 2-row	57.267	43.956	all

Two-row intersection cuts and strengthenings

7 common instances:

[bell15, blend2, egout, khb05250, misc03, misc07, set1ch]

	cuts	gc%	exact
GMI	25.571	24.744	all
P_I	25.143	33.641	all
S-free	28.714	33.836	all
lifting	25.571	33.716	all
P_{IU}	47.857	37.531	all
full 2-row	48.000	37.583	all

Bases

- ▶ We depend on the optimal basis
- ▶ Will the gap closed by two-row cuts survive more GMIs?

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- ▶ Convenient way to explore different (feasible) bases.
- ▶ Now trying to separate a point with a much stronger LP bound (obtained by adding GMIs).

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Relax and cut: results

43 common instances:

	cuts	gc%	exact
GMI	24.814	23.282	all
2-row i.c.	31.884	28.838	42
full 2-row	62.140	36.080	22
relax&cut GMI	60.372	34.970	all
relax&cut 2-row i.c.	63.163	41.951	37
relax&cut full 2-row	76.767	47.277	12

Future

- ▶ More rows
- ▶ More tricks
- ▶ More tests

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