Valid Inequalities for Mixed Integer Linear Programs

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Brief history

First Algorithms

Polynomial Algorithms

Solving systems of linear equations

- Babylonians 1700BC
- Gauss 1801

Edmonds 1967

Solving systems of linear inequalities

- Fourier 1822
- Dantzig 1951

- Khachyan 1979
- Karmarkar 1984

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Solving systems of linear inequalities in integers

• Gomory 1958 • Lenstra 1983

Improvements in MILP software in the last 15 years

Based on Bixby, Gu, Rothberg, Wunderling 2004 and Laundy, Perregaard, Tavares, Tipi, Vazacopoulos 2007

Instances that would have required years of computing time 15 years ago can be solved in seconds today.

- LP Algorithms
- MILP Algorithms
- Computers

1000 times faster 1000 times faster 1000 times faster

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Overall speedup

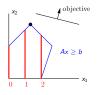
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Sources of improvement for MILP :

- Preprocessors Factor 2
- Heuristics Factor 1,5
- Cutting Planes Factor 300

Mixed Integer Linear Programming min cx $x \in S$

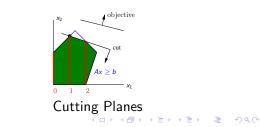
where $S := \{x \in \mathbb{Z}_+^p \times \mathbb{R}_+^{n-p} : Ax \ge b\}$



Linear Relaxation min cx $x \in P$ where $P := \{x \in \mathbb{R}^n_+ : Ax \ge b\}$



Branch-and-bound



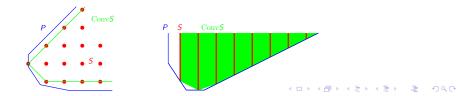
Polyhedral Theory

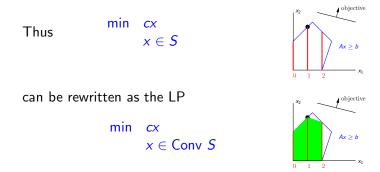
 $P := \{x \in \mathbb{R}^n_+ : Ax \ge b\}$ Polyhedron $S := P \cap (\mathbb{Z}^p_+ \times \mathbb{R}^{n-p}_+)$ Mixed Integer Linear Set $Conv S := \{x \in \mathbb{R}^n : \exists x^1, \dots, x^k \in S, \ \lambda \ge 0, \ \sum \lambda_i = 1$ such that $x = \lambda_1 x^1 + \dots + \lambda_k x^k\}$

THEOREM Meyer 1974

If A, b have rational entries, then Conv S is a polyhedron.

Proof Using a theorem of Minkowski, Weyl : P is a polyhedron if and only if P = Q + C where Q is a polyhedra and C is a polyhedral cone.

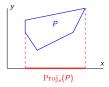




We are interested in the constructive aspects of Conv S.

REMARK The number of constraints of Conv *S* can be exponential in the size of $Ax \ge b$, BUT 1) sometimes a partial representation of Conv *S* suffices (Example : Dantzig, Fulkerson, Johnson 1954); 2) Conv *S* can sometimes be obtained as the projection of a polyhedron with a polynomial number of variables and constraints. Projections

Let $P := \{(x, y) \in \mathbb{R}^n \times \mathbb{R}^k : Ax + Gy \ge b\}$ DEFINITION $\operatorname{Proj}_x(P) := \{x \in \mathbb{R}^n : \exists y \in \mathbb{R}^k \text{ such that } Ax + Gy \ge b\}$



THEOREM $Proj_{x}(P) = \{x \in \mathbb{R}^{n} : vAx \ge vb \text{ for all } v \in Q\}$ where $Q := \{v \in \mathbb{R}^{m} : vG = 0, v \ge 0\}.$

PROOF Let $x \in \mathbb{R}^n$. Farkas's lemma implies that $Gy \ge b - Ax$ has a solution y if and only if $v(b - Ax) \le 0$ for all $v \ge 0$ such that vG = 0 Lift-and-Project Sherali-Adams 1990 Lovász-Schrijver 1991 Balas-Ceria-Cornuéjols 1993

Let
$$S := \{x \in \{0,1\}^p \times \mathbb{R}^{n-p}_+ : Ax \ge b\}$$
$$P := \{x \in \mathbb{R}^n_+ : Ax \ge b\}$$

LIFT-AND-PROJECT PROCEDURE

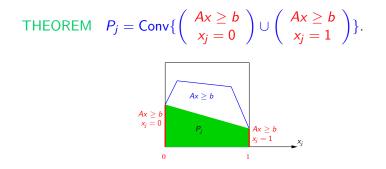
STEP 0 Choose an index $j \in \{1, \ldots, p\}$.

 $\begin{array}{l} \text{STEP 1 Generate the nonlinear system} \quad \begin{array}{l} x_j(Ax-b) \geq 0 \\ (1-x_j)(Ax-b) \geq 0 \end{array}$

STEP 2 Linearize the system by substituting $x_i x_j$ by y_i for $i \neq j$, and x_i^2 by x_j . Denote this polyhedron by M_j .

STEP 3 Project M_i on the x-space. Denote this polyhedron by P_i .

PROPOSITION Conv $(S) \subseteq P_j \subseteq P$.



THEOREM Balas 1979 $\operatorname{Conv}(P) = P_p(\ldots P_2(P_1)\ldots).$

LIFT-AND-PROJECT CUT

Given a fractional solution \bar{x} of the linear relaxation $Ax \geq b$, find a cutting plane $\alpha x \geq \beta$ (namely $\alpha \bar{x} < \beta$) that is valid for P_j (and therefore for S).

CUT GENERATION LINEAR PROGRAM

$$egin{aligned} M_j &:= \{x \in \mathbb{R}^n_+, y \in \mathbb{R}^n_+ : & Ay - bx_j \geq 0, \ Ax + bx_j - Ay \geq b, & y_j = x_j \} \end{aligned}$$

In fact, one does not use the variable *y_j*

To project onto the *x*-space, we use the cone

 $Q := \{u, v \ge 0 : uA_j - vA_j = 0\}$

The first two constraints come from the linearization in STEP 1.

$$M_j := \{x \in \mathbb{R}^n_+, y \in \mathbb{R}^{n-1}_+ : \ B_j x + A_j y \ge 0, \ D_j x - A_j y \ge b\}$$

 $P_j = \{x \in \mathbb{R}^n_+ : (uB_j + vD_j)x \ge vb \\ \text{for all } (u, v) \in Q\}$

DEEPEST CUT

 $\max vb - (uB_j + vD_j)\bar{x}$ $uA_j - vA_j = 0$ $u \ge 0, v \ge 0$ $\sum u_i + \sum v_i = 1$

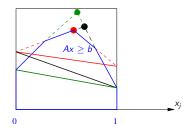
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SIZE OF THE CUT GENERATION LP

 $\max vb - (uB_j + vD_j)\bar{x}$ $uA_j - vA_j = 0$ $\sum u_i + \sum v_i = 1$ $u \ge 0, v \ge 0$

Number of variables : 2mNumber of constraints : n + nonnegativity

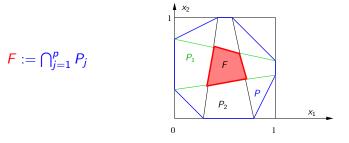
Balas and Perregaard 2003 give a precise correspondance between the basic feasible solutions of the cut generation LP and the basic solutions of the LP $\frac{\min cx}{Ax > b}$



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LIFT-AND-PROJECT CLOSURE OF P

37 %



REMARK Balas and Jeroslow 1980 show how to strengthen cutting planes by using the integrality of the other integer variables (lift-and-project only considers the integrality of one variable x_j at a time).

Experiments of Bonami and Minoux 2005 on MIPLIB 3 instances give the amount of duality gap $= \min_{x \in S} cx - \min_{x \in P} cx$ closed by strengthening P:

Lift-and-project closure Lift-and-project + strengthening

45 %

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Mixed Integer Cuts Gomory 1963

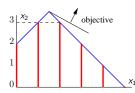
Consider a single constraint : $S := \{x \in \mathbb{Z}^p_+ \times \mathbb{R}^{n-p}_+ : \sum_{j=1}^n a_j x_j = b\}.$ Let $b = \lfloor b \rfloor + f_0$ where $0 < f_0 < 1$, and $a_j = \lfloor a_j \rfloor + f_j$ where $0 \le f_j < 1$.

THEOREM

$$\sum_{f_j \le f_0}^{j \le p:} \frac{f_j}{f_0} x_j + \sum_{f_j > f_0}^{j \le p:} \frac{1 - f_j}{1 - f_0} x_j + \sum_{a_j > 0}^{j \ge p+1:} \frac{a_j}{f_0} x_j - \sum_{a_j < 0}^{j \ge p+1:} \frac{a_j}{1 - f_0} x_j \ge 1$$

is a valid inequality for S.

APPLICATION



 $\max z = x_1 + 2x_2$ $-x_1 + x_2 \le 2$ $x_1 + x_2 \le 5$ $x_1 \in \mathbb{Z}_+$ $x_2 \in \mathbb{R}_+$ $z + 0.5s_1 + 1.5s_2 = 8.5$ $x_1 - 0.5s_1 + 0.5s_2 = 1.5$ $x_2 + 0.5s_1 + 0.5s_2 = 3.5$ Cut $s_1 + s_2 \ge 1$. Or $x_2 \le 3$.

HOW GOOD ARE GOMORY CUTS GENERATED FROM THE OPTIMAL BASIS ?

Bonami and Minoux 2005 MIPLIB 3

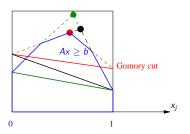
Gomory cuts (optimal basis)

24 %



Lift-and-project

Lift-and-project + strengthening 45 %



CAN ONE IMPROVE GOMORY CUTS BY GENERATING THEM FROM OTHER BASES (Balas, Perregaard 2003) (Balas, Bonami 2007) OR OTHER EQUATIONS ? (Andersen, Cornuéjols, Li 2005) Reduce-and-split cuts Andersen, Cornuéjols, Li 2005

> Perform linear combinaisons of the constraints $\sum_{j=1}^{n} a_j x_j = b$ in order to reduce the coefficients of the continuous variables, and generate the corresponding Gomory cuts.

Why?

Remember the Gomory cut formula :

$$\sum_{f_j \leq f_0}^{j \leq p:} \frac{f_j}{f_0} x_j + \sum_{f_j > f_0}^{j \leq p:} \frac{1 - f_j}{1 - f_0} x_j + \sum_{a_j > 0}^{j \geq p+1:} \frac{a_j}{f_0} x_j - \sum_{a_j < 0}^{j \geq p+1:} \frac{a_j}{1 - f_0} x_j \geq 1$$

ALGORITHM Consider the lines L of the optimal simplex tableau for the basic variables x_i such that $i \leq p$. For every line $\ell \in L$, reduce the norm $||(a_{p+1}^{\ell}, \ldots, a_n^{\ell})||$ by performing integer combinaisons of the other lines of L.

COMPUTATIONAL RESULTS ON MIPLIB INSTANCES

- Reduce-and-split cuts are often very different from Gomory cuts generated directly from the optimal basis.
- Their quality is typically at least as good.
- In some cases, their quality is much better :

Name	20 Times	20 Times	Nodes	Nodes
	Gomory	R&S	Gomory	R&S
flugpl	14 %	100 %	184	0
gesa2	46 %	97 %	743	116
gesa2o	92 %	98 %	9145	75
mod008	47 %	88 %	1409	82
pp08a	83 %	92 %	7467	745
rgn	15 %	100 %	874	0
vpm1	44 %	98 %	7132	1
vpm2	41 %	61 %	38946	4254

GOMORY CLOSURE

$Ax \ge b$ $x \in \mathbb{Z}^p_+ \times \mathbb{R}^{n-p}_+$

- Every valid inequality for P := {x ≥ 0 : Ax ≥ b} (≠ ∅) is of the form uAx + vx ≥ ub − t, where u, v, t ≥ 0.
- Subtract a nonnegative surplus variable $\alpha x s = \beta$.
- Generate a Gomory inequality.
- Eliminate $s = \alpha x \beta$ to get the inequality in the x-space.
- The convex set obtained by intersecting all these inequalities with P is called the Gomory closure.

THEOREM Cook, Kannan, Schrijver 1990

The Gomory closure is a polyhedron.

THEOREM Caprara, Letchford 2002 et Cornuéjols, Li 2002

It is NP-hard to optimize a linear function over the Gomory closure.

Nevertheless,

Balas and Saxena 2006 and Dash, Günlück and Lodi 2007 were able to optimize over the Gomory closure by solving a sequence of parametric MILPs.

DUALITY GAP CLOSED BY DIFFERENT CUTS MIPLIB 3

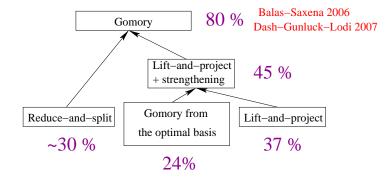
Gomory cuts (optimal basis) 24 %

Reduce-and-split (optimal basis) 30 %

Gomory closure 80 %

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Duality gap closed by different types of cutting planes MIPLIB 3 instances



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Mixed integer rounding Nemhauser-Wolsey 1990 Wolsey 1999 $S := \{x \in \mathbb{Z}^p_+ \times \mathbb{R}^{n-p}_+ : Ax \le b\}$ AGGREGATION $uAx \le ub, u \ge 0, \text{ or } \bar{a}x \le \bar{b}.$ THEOREM $(f, f_0)^+)$

$$\sum_{i \leq p} \left(\lfloor \bar{a}_i \rfloor + \frac{(f_i - f_0)^+}{1 - f_0} \right) x_i + \sum_{\substack{i \geq p+1 \\ \bar{a}_i < 0}} \frac{\bar{a}_i}{1 - f_0} x_i \leq \lfloor \bar{b} \rfloor$$

is a valid inequality for S. It is called MIR.

REMARK It is the same cut as the Gomory inequality obtained from $\bar{a}x + s = \bar{b}$ where $s \ge 0$.

REMARK Stronger inequalities can be obtained by first adding slack variables to $Ax \le b$, and then aggregating the equalities Ax + s = b using multipliers u_i (positive or negative). Bonami and Cornuéjols 2007, and Dash, Günlük and Lodi 2007

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Mixed integer rounding continued

 $S := \{ x \in \mathbb{Z}_{+}^{p} \times \mathbb{R}_{+}^{n-p} : Ax \le b \}$ AGGREGATION uAx + us = ub.

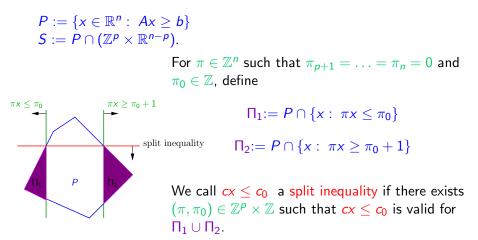
Define the MIR closure as the intersection of the MIR inequalities generated from all the possible multipliers u.

THEOREM The MIR closure is identical to the Gomory closure.

Marchand-Wolsey 2001

- Several classical inequalities for structured problems are MIR inequalities, like "flow cover" inequalities.
- An aggregation heuristic to generate MIR cuts closes around 23 % of the duality gap.

Split Inequalities Cook-Kannan-Schrijver 1990

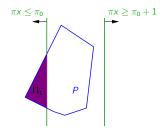


The split closure is the intersection of all split inequalities.

THEOREM Nemhauser-Wolsey 1990, Cornuéjols-Li 2002 The split closure is identical to the Gomory closure.

Chvátal Inequalities Chvátal 1973

A Chvátal inequality is a split inequality where $\Pi_2 = \emptyset$.

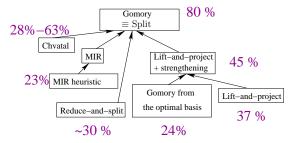


The Chvátal closure is the intersection of all these inequalities.

REMARK Chvátal defined this concept in 1973 in the context of pure integer programs.

The Chvátal closure reduces the duality gap by around 63 % on the pure integer MIPLIB 03 instances (Fischetti-Lodi 2006) and around 28 % on the mixed instances (Bonami-Cornuéjols-Dash-Fischetti-Lodi 2007).

Duality gap closed by different types of cutting planes MIPLIB 3 instances



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Paper available on http://integer.tepper.cmu.edu/