

# A Precise Correspondance Between Lift-and-Project Cuts, Simple Disjunctive Cuts, and Mixed Integer Gomory Cuts

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# Outline of the Paper

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- 2 Simple Disjunctive Cuts and Mixed Integer Gomory Cuts
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- 4 correspondance btw. Lift-and-Project Cuts and Simple Disjunctive Cuts
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# Outline of the Paper (Cont'd)

- 1 bounds on the number of fundamental disjunctive cuts
- 2 bounds on the rank of LP polyhedron wrt. various families of cuts
- 3 an algorithm for solving the cut generating LP
- 4 computational results
- 5 using the algorithm for Gomory Cuts

# Mixed Integer 0-1 Program

(MIP):

$$\begin{aligned} \min \quad & cx \\ \text{s.t.} \quad & \\ Ax \geq & b \\ x \geq & 0 \\ x_j \in & \{0, 1\}, j = 1, \dots, p \end{aligned}$$

# LP Relaxation

(LP):

$$\min\{cx : x \in P\},$$

$$P := \{x \in \mathbb{R}_+^n\}$$

$P$  is sometimes denoted by  $\tilde{A}x \geq \tilde{b}$ , where  $A := \begin{pmatrix} A \\ I \end{pmatrix}$  and

$$b := \begin{pmatrix} b \\ 0 \end{pmatrix}.$$

- $\bar{x}$  denotes the optimum solution to the (LP)
- $S$  is the set of surplus variables and  $N$  is the set of structural variables

## Mixed Integer 0-1 Program (Cont'd)

- the simplex tableau for (LP) can be uniquely determined by the set of variables chosen to be nonbasic.
- the simplex tableau with such a choice can be written as

$$x_i + \sum_{j \in N \cap J} \bar{a}_{ij} x_j + \sum_{j \in S \cap J} \bar{a}_{ij} s_j = \bar{a}_{i0} \text{ for } i \in N \cap I$$

$$s_i + \sum_{j \in N \cap J} \bar{a}_{ij} x_j + \sum_{j \in S \cap J} \bar{a}_{ij} s_j = \bar{a}_{i0} \text{ for } i \in S \cap I$$

$\bar{a}_{ij}$  denotes the coefficient of nonbasic variable  $j$  in the row for the nonbasic variable  $i$ , and  $\bar{a}_{i0}$  is the corresponding RHS

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  - **S.Disj. Cuts and M.I.G. Cuts**
  - Lift-and-Project Cuts
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- 5 Algorithm

# Simple Disjunctive vs. Mixed Int. Gomory

- if we identify the nonbasic variables  $x_j$  with their corresponding surplus variables  $s_j$ , row  $k$  becomes:

$$x_k + \sum_{j \in J} \bar{a}_{kj} s_j = \bar{a}_{k0}$$

- in particular, chose  $x_k$  to be s.t.  $0 \leq \bar{a}_{k0} \leq 1$  and apply disjunction  $x_k \leq 0 \vee x_k \geq 1$  you get  $\pi s_j \geq \pi_0$  where  
 $\pi_0 := \bar{a}_{k0}(1 - \bar{a}_{k0})$  and  
 $\pi_j := \max\{\bar{a}_{k0}(1 - \bar{a}_{k0}), -\bar{a}_{kj}\bar{a}_{k0}\}$
- the cut  $\pi s_j \geq \pi_0$  depends on nonbasic set  $J$ .



## Simple Disjunctive vs. Mixed Int. Gomory(Cont'd)

- if  $p \geq 1$ ,  $\pi s_j \geq \pi_0$  can be strengthened by replacing  $\pi$  with  $\bar{\pi}$ :

$$\bar{\pi} := \begin{cases} \min\{f_{kj}(1 - \bar{a}_{k0}), (1 - f_{kj})\bar{a}_{k0}\} & j \in J \cap \{1, \dots, p\} \\ \pi_j & j \in J - \{1, \dots, p\} \end{cases}$$

with  $f_{kj} := \bar{a}_{kj} - \lfloor \bar{a}_{kj} \rfloor$

- the strengthened version is the same as the Mixed Integer Gomory Cut

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# Lift-and-Project

- Lift and Project cuts are special disjunctive cuts of the form

$$\left( \begin{array}{l} Ax \geq b \\ x \geq 0 \\ -x_k \geq 0 \end{array} \right) \vee \left( \begin{array}{l} Ax \geq b \\ x \geq 0 \\ -x_k \geq 1 \end{array} \right)$$

for some  $k \in \{1, \dots, p\}$  such that  $0 < \bar{x}_k < 1$ .

# Lift-and-Project [Lift]

**Theorem 1([1]):** Let the disjunctive constraints be

$$\bigvee_{h \in Q} (D^h x \geq d_0^h)$$

and let  $A^h = \begin{pmatrix} A \\ D^h \end{pmatrix}$ ,  $a_0^h = \begin{pmatrix} a_0 \\ d_0^h \end{pmatrix}$

Let  $F$  be the feasible set of Disjunctive Program (DP). Then

$$F = \left\{ x \in \mathbb{R}^n : \bigvee_{h \in Q} (A^h x \geq a_0^h, x \geq 0) \right\}$$

Letting  $F_h = \{x \in \mathbb{R}^n : (A^h x \geq a_0^h, x \geq 0)\}$ ,

we note  $F = \bigcup_{h \in Q} F_h$ . Let  $Q^* = \{h \in Q \mid F_h \neq \emptyset\}$

**claim:** If  $F \neq \emptyset$ ,

$$\text{clconv } F = \left\{ x \in \mathbb{R}^n \mid \begin{array}{l} x = \sum_{h \in Q^*} \xi^h, \\ A^h \xi^h - a_0^h \xi_0^h \geq 0, \quad h \in Q^* \\ x = \sum_{h \in Q^*} \xi_0^h = 1 \end{array} \right\}$$

# Lift-and-Project [Lift](Cont'd)

**proof:** Let  $S$  denote the RHS in the claim, so that the theorem is  $F = S$ . If  $Q$  is finite and  $F \neq \emptyset$ , then  $Q^* \neq \emptyset$  and is finite. Moreover,

$$\text{clconv } F = \text{clconv} \left( \bigcup_{h \in Q} F_h \right)$$

**(i)  $F \subseteq S$ :**

If  $x \in \text{conv} F$ , then  $x$  is a convex combination of at most  $|Q^*|$  points, belonging to a different  $F_h$ :

$$x = \sum_{h \in Q^*} \lambda^h u^h, \quad \lambda^h \geq 0, h \in Q^*$$

where  $\sum_{h \in Q^*} \lambda^h = 1$  and for each  $h \in Q^*$ ,  $A^h u^h \geq a_0^h, u^h \geq 0$

# Lift-and-Project [Lift](Cont'd)

- We immediately note that if  $x, \lambda^h, u^h, h \in Q^*$  satisfy the above constraints, then  
 $x, \xi_0^h = \lambda^h, \xi^h = u^h \lambda^h, h \in Q^*$  satisfies S.

$\Rightarrow$ (i)

## Lift-and-Project [Lift](Cont'd)

**(ii)  $S \subseteq clconvF$ :**

Let  $\bar{x} \in S$  with associated vectors  $(\bar{\xi}^h, \bar{\xi}_0^h)$ ,  $h \in Q^*$ . Let's divide the index set of nonempty  $F_h$  sets,  $Q^*$  so that

$$Q_1^* = \{h \in Q^* | \xi_0^h > 0\}, Q_2^* = \{h \in Q^* | \xi_0^h = 0\}$$

**case  $h \in Q_1^*$ :**  $\bar{\xi}^h / \bar{\xi}_0^h$  is a solution to  $A^h x \geq a^h$ ,  $x \geq 0$  (see RHS)

thus  $(\bar{\xi}^h / \bar{\xi}_0^h) \in F_h$ , So

$$(\bar{\xi}^h / \bar{\xi}_0^h) = \sum_{i \in U_h} \mu^{hi} u^{hi} + \sum_{k \in V_h} \nu^{hk} v^{hk}$$

for some  $u^{hi} \in \text{vert}F_h$ ,  $i \in U_h$  and  $v^{hk} \in \text{dir}F_h$ ,  $k \in V_h$  with  $U_h, V_h$  finite index sets,  $\mu^{hi}, \nu^{hk} \geq 0$ , and  $\sum_{i \in U_h} \mu^{hi} = 1$



# Lift-and-Project [Lift](Cont'd)

By setting  $\mu^{hi} \bar{\xi}_0^h = \theta^{hi}$ , and  $\nu^{hk} \bar{\xi}_0^h = \sigma^{hk}$  we get:

$$\bar{\xi}^h = \sum_{i \in U_h} \theta^{hi} u^{hi} + \sum_{k \in V_h} \sigma^{hk} v^{hk}$$

with  $\theta^{hi} \geq 0$ ,  $i \in U_h$ ,  $\sigma^{hk} \geq 0$ ,  $k \in V_h$  and  $\sum_{i \in U_h} \theta^{hi} = \bar{\xi}_0^h$

# Lift-and-Project [Lift](Cont'd)

**case**  $h \in Q_2^*$ : either  $\bar{\xi}^h = 0$ , or  $\bar{\xi}_0^h$  is a solution to  $Ax \geq 0, x \geq 0$   
(extreme ray) thus

$$\bar{\xi}^h = \sum_{k \in V_h} \sigma^{hk} v^{hk}$$

with  $\theta^{hi} \geq 0, k \in V_h$  for some  $v^{hk} \in \text{dir}F_h$

# Lift-and-Project [Lift](Cont'd)

Thus,

$$\begin{aligned}\bar{x} &= \sum_{h \in Q^*} \bar{\xi}^h \\ &= \sum_{h \in Q_1^*} \left( \sum_{i \in U_h} \theta^{hi} u^{hi} + \sum_{k \in V_h} \sigma^{hk} v^{hk} \right) + \sum_{h \in Q_2^*} \left( \sum_{k \in V_h} \sigma^{hk} v^{hk} \right)\end{aligned}$$

- Noting that  $\sum_{h \in Q_1^*} \sum_{i \in U_h} \theta^{hi} u^{hi} = \sum_{h \in Q_1^*} \bar{\xi}_0^h = 1$ , we realize that  $\bar{x}$  is a convex combination of finitely many points and directions of F.

⇒(ii)

## Lift-and-Project [Lift](Cont'd)

- So,  $\text{conv } F \subseteq S \subseteq \text{clconv } F$  and since  $\text{clconv } F$  is the smallest closed set containing  $\text{conv}F$ ,  $\text{clconv}F = S$ .

Lifting in our special case,  $x_j \in \{0, 1\}$ 

$$P_{j0} := \{x \in \mathbb{R}_+^n : Ax \geq b, x_j = 0\}$$

$$P_{j1} := \{x \in \mathbb{R}_+^n : Ax \geq b, x_j = 1\}$$

$$\begin{array}{rcll}
 x & -y & & -z & = & 0 \\
 & Ay & -by_0 & & \geq & 0 \\
 & -y_j & 0y_0 & & = & 0 \\
 & & & Az & -bz_0 & \geq & 0 \\
 & & & z_j & -1z_0 & = & 0 \\
 & & y_0 & & z_0 & = & 1
 \end{array}$$

# Lift-and-Project [Project]

- We want a cut of the form  $\alpha x \geq \beta$ . To get this from the disjunctive constraint set above, let  $A^i$  be  $A$  amended with the unit vector row  $e_j$ . Let  $b^1 = \begin{pmatrix} b \\ 0 \end{pmatrix}$  and  $b^2 = \begin{pmatrix} b \\ 1 \end{pmatrix}$ .
- Then to satisfy the constraints  $A^i x \geq b^i$ , we should have  $\alpha x \geq A^i x \geq b^i \geq \beta$ . In other words,  $\alpha \geq u^i A^i$  and  $\beta \leq u^i b^i$ .

# Lift-and-Project [Project]

the resulting feasible set for  $(\alpha, \beta)$  is thus:

$$\alpha \geq uA - u_0 e_j$$

$$\alpha \geq vA + v_0 e_j$$

$$\beta \leq ub$$

$$\beta \leq vb + v_0$$

$$u, v \geq 0$$

$$(\alpha, \beta) \in \mathbb{R}^{n+1}$$





# Lift-and-Project (Cont'd)

- this program maximizes the cut off
- $\alpha$  and  $\beta$  are urs, so they can be eliminated and can be retrieved anytime given the solution vector for  $u, u_0, v, v_0$ :

$$\beta := ub = vb + v_0$$
$$\alpha := \begin{cases} \max\{ua_j, va_j\} & j \neq k \\ \max\{ua_k - u_0, va_j + v_0\} & j = k \end{cases}$$

# Lift-and-Project (Cont'd)

- this also can be strengthened using the integrality of the  $x_j$ ,  $j \in \{1, \dots, p\} - \{k\}$ :

$$\bar{\alpha} := \begin{cases} \min\{ua_j + u_0 \lceil m_j \rceil, va_j - v_0 \lfloor m_j \rfloor\} & j \in \{1, \dots, p\} - \{k\} \\ \alpha_j, & j \in \{k\} \cup \{p+1, \dots, n\} \end{cases}$$

with  $m_j := \frac{va_j - ua_j}{u_0 + v_0}$ .

## Correspondance btw. the Unstrengthened Cuts

- introduce surplus variables to  $(CGLP)_k$  so that  $u, v$  have the surplus variables included:

$$\begin{array}{rcll}
 \min & \alpha \bar{x} & -\beta & \\
 \text{st} & & & \\
 & \alpha & -uA & +u_0 e_k = 0 \\
 & \alpha & & -vA + v_0 e_k = 0 \\
 & -\beta & +ub & = 0 \\
 & -\beta & & +vb + v_0 = 0 \\
 & & \sum_{i=1}^{m+p} u_i + u_0 & + \sum_{i=1}^{m+p} v_i + v_0 = 1 \\
 & & & u, u_0, v, v_0 \geq 0
 \end{array}$$

## Correspondance btw. the Unstrengthened Cuts (Cont'd)

**Lemma 1:** In any basic solution to the constraint set above that gives  $\alpha \geq \beta$  not dominated by the constraint set of (LP),  $u_0, v_0 > 0$ .  
**proof:**

# Correspondance btw. the Unstrengthened Cuts (Cont'd)

**Lemma 2:** Let  $(\bar{\alpha}, \bar{\beta}, \bar{u}, \bar{u}_0, \bar{v}, \bar{v}_0)$  be a basic solution to the above constraint set,  $\bar{u}_0, \bar{v}_0 > 0$   $(\bar{\alpha}, \bar{\beta})$  basic. (They are URS). Let the basic components of  $\bar{u}$  and  $\bar{v}$  be indexed by  $M_1$  and  $M_2$ . Then  $M_1 \cap M_2 = \emptyset$ ,  $|M_1 \cup M_2| = n$ , and submatrix  $\hat{A}_{n \times n}$  of  $\tilde{A}$  whose rows are indexed by  $M_1 \cup M_2$  is nonsingular.

**proof:**

## Correspondance btw. the Unstrengthened Cuts (Cont'd)

- define  $J := M_1 \cap M_2$
- replace  $n$  inequalities indexed by  $J$  in  $\tilde{A}x \geq \tilde{b}$  this amounts to setting surplus variables to 0. Since  $\hat{A}_{n \times n}$  is nonsingular, these equalities define a basic solution.
- The simplex tableau associated with this solution has its *nonbasic* variables indexed by  $J$ .
- in the  $(CGLP)_k$  solution was the index set of basic components of  $(u, v)$ .

# Correspondance btw. the Unstrengthened Cuts (Cont'd)

- we have

$$\hat{A}x - s_j = \hat{b}$$

, or equivalently

$$x = \hat{A}^{-1}\hat{b} + \hat{A}^{-1}s_j$$

- if we let  $\bar{a}_{k0} = e_k \hat{A}^{-1} \hat{b}$  and  $\bar{a}_{kj} = (\hat{A}^{-1})_{kj}$ , this can be written as

$$x_k = \bar{a}_{k0} - \sum_{j \in J} \bar{a}_{kj} s_j$$

- this is same as the row of (LP) associated with basic variable  $x_k$

## Correspondance btw. the Unstrengthened Cuts (Cont'd)

**Lemma 3:**  $0 < \bar{a}_{k0} < 1$ .  
**proof:**



**Theorem 4A:** Let  $\alpha x \geq \beta$  be the lift-and-project cut associated with a basic solution  $(\alpha, \beta, u, u_0, v, v_0)$  to  $(CGLP)_k$ , with  $u_0, v_0 > 0$  and all components of  $\alpha, \beta$  basic, and the basic components of  $u$  and  $v$  be indexed by  $M_1$  and  $M_2$  respectively. Let  $\pi s_j \geq \pi_0$  be the simple disjunctive cut from the disjunction  $x_k < 0 \vee x_k > 1$  applied to  $x_k = \bar{a}_{k0} - \sum_{j \in J} \bar{a}_{kj} s_j$  with  $J := M_1 \cap M_2$ . Then  $\pi s_j \geq \pi_0$  is equivalent to  $\alpha x \geq \beta$ .

# Correspondance btw. the Unstrengthened Cuts (Cont'd)

## sketch of proof:

Remember that  $x_k < 0 \vee x_k > 1$  applied to  $x_k = \bar{a}_{k0} - \sum_{j \in J} \bar{a}_{kj} s_j$

was defined by

$$\pi_0 := \bar{a}_{k0}(1 - \bar{a}_{k0})$$

and

$$\pi_j := \max\{\pi_j^1, \pi_j^2\}$$

where

$$\pi_j^1 := \bar{a}_{k0}(1 - \bar{a}_{k0}), \quad \pi_j^2 := -\bar{a}_{kj}\bar{a}_{k0} = (\hat{A}^{-1})_{kj}\bar{a}_{k0}$$

# Bounds on the Number of Essential Cuts

- Every valid inequality for  $\{x \in P : (x_k \leq 0) \vee (x_k \geq 1)\}$  is dominated by some lift-and-project cut corresponds to a basic solution of a basic solution of  $(CGLP)_k$
- The number of undominated valid inequalities is bounded by

$$\binom{2(m+p+n+1) + n + 1}{2n + 3}$$

- By using Theorem 4A/4B, the number of bases in a simplex tableau where  $x_k$  is basic, that is, the number of subsets  $J$  of cardinality  $n$  is

$$\binom{m+p+n-1}{n}$$

# Bounds on the Number of Essential Cuts

- Thus the elementary closure  $\bigcap_{k=1}^p P_k$  of  $P$  with respect to the lift-and-project operation has at most

$$p \binom{m + p + n - 1}{n}$$

facets.

- Can we extend these bounds for strengthened lift-and-project cuts?
  - That is OK for strengthened cuts derived from basic solutions
  - But a strengthened cut derived from a nonbasic solution may not be dominated by any strengthened cut derived from a basic solution

## The Rank of P With Respect to Different Cuts

- The rank of P with respect to each of the following families is at most p
  - unstrengthened lift-and-project cuts
  - simple disjunction cuts
  - strengthened lift-and-project cuts
  - mixed integer Gomory cuts

# The Rank of $P$ With Respect to Different Cuts

- Proof:

- $P := \{x \in R^n : \tilde{A}x \geq \tilde{b}\}$
- $P_0 := P$
- $P_D := \text{conv}\{x \in P : x_j \in \{0, 1\}, j = 1, \dots, p\}$
- $P^j := \text{conv}\{P^{j-1} \cap \{x_j \in R^n : x_j \in \{0, 1\}\}$
- then  $P^p = P_D$

## Solving $(CGLP)_k$ on the (LP) Simplex Tableau

- A basic solution to (LP) associated with set J corresponds to a set of basic solutions to  $(CGLP)_k$ .
- The various solutions to  $(CGLP)_k$  differ among themselves by the partition of J into  $M_1$  and  $M_2$ .
- These solutions can be obtained by degenerate pivots in  $(CGLP)_k$
- A single pivot in (LP) differs J with some element with together shifting one ore more elements from  $M_1$  to  $M_2$  vice-versa

Solving  $(CGLP)_k$  on the (LP) Simplex Tableau

- The simple disjunction cut is defined by  $\pi x_J \geq \pi_0$ , where  $\pi_0 = \bar{a}_{k0}(1 - \bar{a}_{k0})$  and

$$\pi_j := \{ \max\{\bar{a}_{kj}(1 - \bar{a}_{k0}), -\bar{a}_{kj}\bar{a}_{k0}\} \quad j \in J$$

- We want to pivot on  $\bar{a}_{ij}$ ,  $i \neq k$   
then row k becomes

$$x_k = \bar{a}_{k0} + \gamma_j \bar{a}_{i0} - \sum_{h \in J \setminus \{j\}} (\bar{a}_{kh} + \gamma_j \bar{a}_{ih}) s_h - \gamma_j x_i$$

where

$$\gamma_j = -\frac{\bar{a}_{kj}}{\bar{a}_{ij}}.$$

- Note that we can pivot on any nonzero  $\bar{a}_{ij}$  since we do not restrict ourselves to feasible bases.



## Solving $(CGLP)_k$ on the (LP) Simplex Tableau

- Pivoting the variable  $x_j$  out of basis corresponds to pivoting into the basis one of the variables  $u_i$  or  $v_i$  on  $(CGLP)_k$
- Such a pivot is improving on  $(CGLP)_k$  only if either  $u_i$  or  $v_i$  have a negative reduced cost
- First, we choose a row  $i$ , some multiple of which is to be added to row  $k$ ,  
second, we choose a column in row  $i$ , which sets the sign and size of the multiplier

# Solving $(CGLP)_k$ on the (LP) Simplex Tableau

The sketch of the algorithm:

- **Step 0.** Solve (LP). Let  $\bar{x}$  be an optimal solution and let  $k$  be such that  $0 < \bar{x}_k < 1$
- **Step 1.** Let  $J$  index the nonbasic variables in the current basis. Compute the reduced costs  $r_{u_i} < 0$  with  $M_1 = \{j \in J : \bar{a}_{kj} < 0 \vee (\bar{a}_{kj} = 0 \wedge \bar{a}_{ij} > 0)\}$ , and  $M_2 = J \setminus M_1$  and  $r_{v_i} < 0$  with  $M_1 = \{j \in J : \bar{a}_{kj} < 0 \vee (\bar{a}_{kj} = 0 \wedge \bar{a}_{ij} < 0)\}$ , and  $M_2 = J \setminus M_1$  of  $u_i, v_i$  corresponding to each row  $i \neq j$  of the simplex tableau of LP.
- **Step 2.** Let  $i_*$  be a row with  $r_{u_{i_*}} < 0$  or  $r_{v_{i_*}} < 0$ . If no such row exists, go to step 5.

# Solving $(CGLP)_k$ on the (LP) Simplex Tableau

- **Step 3.** Identify the most improving pivot column  $j_*$  in row  $i_*$  by minimizing  $f^+(\gamma_j)$  over all  $j \in J$  with  $\gamma_j > 0$  and  $f^-(\gamma_j)$  over all  $j \in J$  with  $\gamma_j < 0$  and choosing the more negative of these two values.
- **Step 4.** Pivot on  $\bar{a}_{i_*j_*}$  and go to Step 1.
- **Step 5.** If row  $k$  has no 0 entries, stop. Otherwise perturb row  $k$  by replacing every 0 entry by  $\xi^t$  for some small  $\xi$  and  $t = 1, 2, \dots$  (different for each entry). Go to step 1.

Solving  $(CGLP)_k$  on the (LP) Simplex Tableau

Let  $(\alpha, \beta, u, u_0, v, v_0)$  be a basic feasible solution to CGLP with  $u_0, v_0 > 0$ , all components of  $\alpha$  and  $\beta$  basic, and the basic components of  $u$  and  $v$  indexed by  $M_1$  and  $M_2$ , respectively. Let  $\bar{s}$  be surplus variables of  $\tilde{A}x \geq \tilde{b}$  corresponding to the solution  $\bar{x}$ . Then the reduced costs of  $u_i$  and  $v_i$ , for  $i \notin J \cup \{k\}$  in this basic solution are, respectively

$$r_{u_i} = \sigma \left( - \sum_{j \in M_1} \bar{a}_{ij} + \sum_{j \in M_2} \bar{a}_{ij} - 1 \right) - \sum_{j \in M_2} \bar{a}_{ij} \bar{s}_j + \bar{a}_{i0} (1 - \bar{x}_k)$$

$$r_{v_i} = \sigma \left( + \sum_{j \in M_1} \bar{a}_{ij} - \sum_{j \in M_2} \bar{a}_{ij} - 1 \right) - \sum_{j \in M_1} \bar{a}_{ij} \bar{s}_j + \bar{a}_{i0} \bar{x}_k$$

where

$$\sigma = \frac{\sum_{j \in M_2} \bar{a}_{kj} \bar{s}_j - \bar{a}_{k0} (1 - \bar{x}_k)}{1 + \sum_{j \in J} |\bar{a}_{kj}|}$$

Solving  $(CGLP)_k$  on the (LP) Simplex Tableau

- Write the objective function,  $\alpha\bar{x} - \beta$ , of  $(CGLP)_k$  in terms of  $u_i$  and  $v_i$
- Then substitute  $u_i$  and  $v_i$  in terms of  $\bar{a}_{ij}$
- During this calculation, they pointed:  
$$u_j = -(u_0 + v_0)\bar{a}_{kj} + (u_i - v_i)\bar{a}_{ij} \text{ for } j \in M_1$$
$$v_j = (u_0 + v_0)\bar{a}_{kj} - (u_i - v_i)\bar{a}_{ij} \text{ for } j \in M_2$$

The pivot column in row  $i$  of the (LP) simplex tableau that is most improving with respect to the cut from row  $k$ , is indexed by that  $l^* \in J$  that minimizes  $f^+(\gamma_l)$  if  $\bar{a}_{kl}\bar{a}_{il} < 0$  or  $f^-(\gamma_l)$  if  $\bar{a}_{kl}\bar{a}_{il} > 0$ , over all  $l \in J$  that satisfies  $\frac{-\bar{a}_{k0}}{\bar{a}_{i0}} < \gamma_l < \frac{1-\bar{a}_{k0}}{\bar{a}_{i0}}$ , where  $\gamma_l := -\frac{\bar{a}_{kl}}{\bar{a}_{il}}$  and for  $0 \leq \gamma < \frac{1-\bar{a}_{k0}}{\bar{a}_{i0}}$

$$f^+(\gamma) :=$$

$$\frac{\sum_{j \in J} (-(\bar{a}_{k0} + \gamma \bar{a}_{i0}) \bar{a}_{kj} + \max\{\bar{a}_{kj}, -\gamma \bar{a}_{ij}\} \bar{x}_j - (1 - \bar{a}_{k0} - \gamma \bar{a}_{i0}) \bar{a}_{k0})}{1 + |\gamma| + \sum_{j \in J} |\bar{a}_{kj} + \gamma \bar{a}_{ij}|}$$

and for  $\frac{-\bar{a}_{k0}}{\bar{a}_{i0}} < \gamma \leq 0$

$$f^-(\gamma) :=$$

$$\frac{\sum_{j \in J} (-(\bar{a}_{k0} + \gamma \bar{a}_{i0}) \bar{a}_{kj} + \max\{\bar{a}_{kj} + \gamma \bar{a}_{ij}, 0\} \bar{x}_j - (1 - \bar{a}_{k0}) (\bar{a}_{k0} + \gamma \bar{a}_{i0}))}{1 + |\gamma| + \sum_{j \in J} |\bar{a}_{kj} + \gamma \bar{a}_{ij}|}$$

- At termination, the simple disjunctive cut from row  $k$  is an optimal lift-and-project cut; the mixed-integer Gomory cut from row  $k$  is an optimal strengthened lift-and-project cut.
- When the algorithm comes to a point where Step 2 finds no row with negative reduced costs, we can not conclude the solution is optimal if there is entries of 0's in row  $k$
- In this case, partition  $(M_1, M_2)$  of set  $J$  is not unique, so different partition of  $(M_1, M_2)$  may lead to a basis change where  $J'$  differs from  $J$  in one element.
- Perturbation in Step 5 eliminates the 0 entries in row  $k$ , thus  $(M_1, M_2)$  will be unique for set  $J$ .

## Using Lift-and-Project to Strengthen Mixed Integer Gomory Cuts

- Steiner triple problem with 15 variables and 35 constraints
- LP with five fractional variables is 35.
- Generating mixed integer Gomory cut for each fractional variables yields a solution of value 39
- Using improved cuts in place of original ones we get a solution of value 41.41
- Iterating this procedure for 10 times yields a value of 42.73 for mixed integer Gomory cuts and a value of 44.85 for strengthened cuts.
- IP optimum is 45.
- Intermediate cuts resulting from the procedure are dominated by the final improved ones for the first iteration.



## Concluding Remarks

- There are numerous attempts to improve mixed integer Gomory cuts but none of these attempts has succeeded in defining a procedure that is guaranteed to find an improved cuts.
- The lift-and-project approach has done that
- Does the gain in the quality of the cuts justify the computation effort for improving them?