

## A PRIMAL CUTTING PLANE ALGORITHM FOR INTEGER FRACTIONAL PROGRAMMING PROBLEMS

HIROAKI ISHII, *Osaka University*

TOSHIHIDE IBARAKI and HISASHI MINE, *Kyoto University*

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### Abstract

A primal cutting plane algorithm is proposed for the integer fractional programming problem:

$$\begin{aligned} &\text{maximize} \quad (c_0 + \sum_{j=1}^n c_j x_j) / (d_0 + \sum_{j=1}^n d_j x_j) \\ &\text{subject to} \quad \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i=1, 2, \dots, m \\ &\quad \quad \quad x_j \geq 0, \quad \text{integer}, \quad j=1, 2, \dots, n \end{aligned}$$

The algorithm is obtained by slightly modifying *Young's simplified primal algorithm* developed for the ordinary integer programming problem, and is based on the parametric programming approach to the fractional problem given by Jagannathan and Dinkelbach.

### 1. Introduction

Research on fractional programming problems has been concentrated on continuous type problems with a linear fractional objective function. This paper, however, discusses the integer programming problem with a linear fractional objective function. In addition to the known methods ([1], [5], [6], [9])

for the integer fractional programming problem or some special cases, this paper proposes a new method, which may be considered as an integer programming version of the well known *Martos' method* for the linear fractional problems with continuous variables [8]. The adaptation to the integer problem is made possible by applying *Young's simplified primal algorithm (SPA)* for the integer programming problem, in a parametric programming manner proposed by Jagannathan [7] and Dinkelbach [2].

In Section 2, the problem description and some assumptions are given. In Section 3, a subsidiary problem  $P(\lambda)$  is defined, and close relations between  $P(\lambda)$  and the original integer fractional problem  $P$  are discussed. Based on these relations, an algorithm is proposed in Section 4. Following an example given in Section 5, Section 6 proves the validity of the algorithm and its finiteness. In Section 7 some further considerations are given.

## 2. Integer Fractional Programming Problem

This paper treats the following integer fractional problem  $P$ :

$$\text{maximize} \quad N(x)/D(x) = \frac{\sum_{j=1}^n c_j x_j + c_0}{\sum_{j=1}^n d_j x_j + d_0}$$

subject to  $x \in S$ ,

where

$$S = \{x = (x_1, \dots, x_n) \mid \sum_{j=1}^n a_{ij} x_j \leq b_i, \quad i=1, 2, \dots, m\}$$

$$\text{and } x_j \geq 0, \text{ integer, } j=1, 2, \dots, n\}$$

The followings are assumed throughout this paper.

- (1) Coefficients  $c_j, d_j, b_i, a_{ij}$  are all integers.
- (2)  $S$  is bounded and nonempty.
- (3)  $D(x) > 0$  for all  $x \in S$ .

## 3. Subsidiary Problem $P(\lambda)$ and its Relations to $P$

Subsidiary Problem  $P(\lambda)$  is defined for a real number  $\lambda$  as follows.

$$P(\lambda) ; \text{ maximize } N(x) - \lambda D(x) = \sum_{j=1}^n (c_j - \lambda d_j) x_j + c_0 - \lambda d_0$$

subject to  $x \in S$ .

Let  $Z(\lambda)$  be the optimal value of  $P(\lambda)$ . It is well known that  $P$  and  $P(\lambda)$  are closely related to each other as follows. (Theorem 3.1 and Theorem 3.2 are due to [2], [7].)

Theorem 3.1. Let  $\lambda^*$  be the optimal value of  $P$ . Then, it holds that

- (i)  $\lambda < \lambda^* \iff Z(\lambda) > 0$ ,
- (ii)  $\lambda = \lambda^* \iff Z(\lambda) = 0$ ,
- (iii)  $\lambda > \lambda^* \iff Z(\lambda) < 0$ .

Theorem 3.2.  $Z(\lambda)$  is a strictly decreasing convex function of  $\lambda$ .

By theorem 3.1, the problem  $P$  reduces to finding a  $\lambda$  such that  $Z(\lambda)=0$ .

Theorem 3.3. For  $x, \bar{x} \in S$  and  $\lambda=N(x)/D(x)$ ,

$$N(\bar{x}) - \lambda D(\bar{x}) > 0$$

if and only if

$$N(\bar{x}) / D(\bar{x}) > N(x) / D(x) .$$

Proof.  $N(\bar{x}) - \lambda D(\bar{x}) > 0 \iff N(\bar{x}) - \frac{N(x)}{D(x)} D(\bar{x}) > 0 \iff \frac{N(\bar{x})}{D(\bar{x})} - \frac{N(x)}{D(x)} > 0$   
 (since  $D(\bar{x}) > 0$ ).  $\square$

The property of Theorem 3.3 plays an important role in the development of our algorithm.

#### 4. An Algorithm for the Integer Fractional Programming Problem

Before describing an algorithm for the integer fractional programming problem, we shall briefly review Young's SPA used to solve the ordinary integer programming problems ([3, 11, 13, 14]). Young's SPA starts with the following integer programming problem as the initial tableau:

$$\begin{aligned}
 &\text{Maximize} && x_0 = c_0 + \sum_{j=1}^n (-c_j)(-x_j) \\
 (4.1) \quad &\text{subject to} && x_{n+i} = b_i + \sum_{j=1}^n a_{ij}(-x_j), \quad i=1,2,\dots,m, \\
 &&& x_k \geq 0, \quad k=1,2,\dots,n+m,
 \end{aligned}$$

where the primal feasibility condition

$$b_i \geq 0, \quad i=1,2,\dots,m$$

is assumed, and  $c_j, a_{ij}, b_i$  are all integer constants.

To deal with a tableau obtained after pivot operations, in general, let us assume that the following is the current tableau.

$$\begin{aligned} \text{maximize} \quad & x_0 = \alpha_{00} + \sum_{j=1}^n \alpha_{0j}(-t_j) \\ \text{subject to} \quad & u_i = \alpha_{i0} + \sum_{j=1}^n \alpha_{ij}(-t_j), \quad i=1,2,\dots,m' \\ (4.2) \quad & t_j \geq 0, \quad j=1,2,\dots,n \\ & u_i \geq 0, \quad i=1,2,\dots,m', \end{aligned}$$

where the primal feasibility

$$\alpha_{i0} \geq 0, \quad j=1,2,\dots,n$$

is also assumed;  $u_i$  denotes the current basic variables and  $t_j$  nonbasic variables.

If this tableau satisfies the dual feasibility condition also, i.e.,

$$\alpha_{0j} \geq 0, \quad j=1,2,\dots,n,$$

then the current tableau provides an optimal solution and computation terminates. Otherwise, a cut is generated according to a certain rule, and a pivot operation is performed on this cut row. ( $m'$  in (4.2) includes the number of the generated cuts.) The resulting tableau satisfies one of the following three conditions:

- (i) It is dual feasible. Then Young's SPA terminates.
- (ii) It is not dual feasible, but will be able to satisfy  $\alpha'_{00} > \alpha_{00}$  in the next tableau, where  $\alpha'_{00}$  is the new coefficient  $\alpha_{00}$  after the pivot operation is executed. (This case is called a transition cycle.) Then the same procedure is repeated by regarding the resulting tableau as (4.2) until case (i) or case (iii) is reached. (If the tableau has L-row (introduced in (iii) below), it is deleted before the pivot operation.)

(iii) It is not dual feasible, and  $\alpha'_{00} = \alpha_{00}$  will hold in the next tableau. (This case is called a stationary cycle.) A pivot element is determined by a rule based on a special row, called *L*-row. (If the current tableau does not have *L*-row, a pivot element is selected after generating *L*-row.) Then pivot operation is performed. After repeating this process finite times, it is known that either case (i) or case (ii) is eventually reached.

By repeating the above cycles, Young's SPA guarantees that case (i) is eventually reached and an optimal solution of (4.1) is obtained. Now note that the problem  $P(\lambda)$  is rewritten as follows.

$$\begin{aligned}
 P(\lambda) : \quad & \text{maximize} \quad x_0 = c_0 - d_0 + \sum_{j=1}^n \{(-c_j) - \lambda(-d_j)\}(-x_j) \\
 & \text{subject to} \quad z_1 = c_0 + \sum_{j=1}^n (-c_j)(-x_j) \\
 (4.3) \quad & z_2 = d_0 + \sum_{j=1}^n (-d_j)(-x_j) \\
 & x_{n+i} = b_i + \sum_{j=1}^n a_{ij}(-x_j), \quad i=1,2,\dots,m \\
 & x_j \geq 0, \text{ integer}, \quad j=1,2,\dots,n+m.
 \end{aligned}$$

This differs from (4.1) only in that the objective function is parametrized by  $\lambda$ , and rows  $z_1$  and  $z_2$  are augmented. The primal feasibility

$$b_i \geq 0, \quad i=1,2,\dots,m,$$

is also assumed. For a fixed  $\lambda$ , our algorithm is exactly the same as Young's SPA. Two rows  $z_1$  and  $z_2$  corresponding to  $N(x)$  and  $D(x)$  respectively, are used to compute the new objective row  $x_0$  when  $\lambda$  is modified. To describe a general step of our algorithm, let the current tableau be as follows:

$$\begin{aligned}
 & \text{maximize} \quad x_0 = \alpha_{00}(\lambda) + \sum_{j=1}^n \alpha_{0j}(\lambda)(-t_j) \\
 & \text{subject to} \quad z_1 = \beta_{00} + \sum_{j=1}^n \beta_{0j}(-t_j)
 \end{aligned}$$

$$z_2 = \gamma_{00} + \sum_{j=1}^n \gamma_{0j}(-t_j)$$

$$u_i = \alpha_{i0} + \sum_{j=1}^n \alpha_{ij}(-t_j) , \quad i=1,2,\dots,m' ,$$

$$t_j \geq 0 , \quad j=1,2,\dots,n$$

$$u_i \geq 0 , \quad i=1,2,\dots,m'$$

This tableau satisfies the primal feasibility

$$\alpha_{i0} \geq 0 , \quad i=1,2,\dots,m'$$

and all coefficients  $\alpha_{ij}$ ,  $\beta_{0j}$ ,  $\gamma_{0j}$  are integers. The objective row  $x_0$  is related to the  $z_1$  and  $z_2$  rows in the following way:

$$(4.4) \quad \alpha_{0j}(\lambda) = \beta_{0j} - \lambda \gamma_{0j} , \quad j=1,2,\dots,n .$$

Our algorithm starts with the initial tableau (4.3) with

$$\lambda = c_0 / d_0$$

(this makes  $\alpha_{00}(\lambda)=0$ ). If the initial tableau does not satisfy the dual feasibility condition, the above Young's SPA is applied until coefficient  $\alpha_{00}(\lambda)$  strictly increases (i.e.,  $\alpha_{00}(\lambda) > 0$ ) or the dual feasibility condition (with  $\alpha_{00}(\lambda)=0$ ) is satisfied. As noted above, Young's SPA always produces one of the two results in finite pivot operations. If a dual feasible tableau is obtained, computation terminates and the resulting tableau provides an optimal solution. If  $\alpha_{00}(\lambda) > 0^+$  is obtained, however,  $\lambda$  is updated to  $\lambda = \beta_{00}/\gamma_{00}$  (this is justified by Theorem 3.3) so that  $\alpha_{00}(\lambda)=0$  and the objective row  $x_0$  is recalculated by (4.4). Then the above procedure is repeated.

As shown in Theorem 6.1, the entire computation eventually terminates and an optimal solution is obtained.

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<sup>†</sup>  $\alpha_{00}(\lambda) > 0$  corresponds to  $\alpha'_{00} > \alpha_{00}$  (the condition to enter the transition cycle).

An Algorithm for the Integer Fractional Programming Problem

Step 1 (Initialize)<sup>††</sup>: Let  $\lambda \leftarrow N(x^I)/D(x^I)$ ,

$$x_j^I = 0, \quad j=1,2,\dots,n.$$

Step 2 (Check the optimality)<sup>†††</sup>: If

$$\alpha_{0j}(\lambda) \geq 0, \quad j=1,2,\dots,n,$$

go to Step 4. Otherwise, add a cut and apply a pivot operation according to the rule for the transition cycle of Young's SPA (see Appendix for details).

Step 3 : If  $\alpha_{00}(\lambda) > 0$ , let  $\lambda \leftarrow \frac{\beta_{00}}{\gamma_{00}}$  and update the  $x_0$ -row by (4.4). Return to Step 2. Otherwise return to Step 2 directly.

Step 4 (Terminate) : Terminate. The current  $\lambda$  is the optimal value of P and

$$u_i = \alpha_{i0}, \quad i=1,2,\dots,m'$$

$$t_j = 0, \quad j=1,2,\dots,n$$

is an optimal solution of P.

<sup>††</sup> We assume that the tableau (4.3) obtained from the above  $\lambda$  is primal feasible, i.e.,

$$b_i \geq 0, \quad i=1,2,\dots,m.$$

It obviously satisfies  $\alpha_{00}(\lambda) = 0$ . If (4.3) does not provide a primal feasible tableau, a primal feasible tableau has to be obtained by some means. This point is not discussed in this paper, since it is the same as the ordinary integer programming problem.

<sup>†††</sup> Whenever Step 2 is entered,  $\alpha_{00}(\lambda) = 0$  is always satisfied.

### 5. Example

Consider the following problem P (see Fig. 5.1).

P :

$$\text{Maximize} \quad \frac{N(x)}{D(x)} = \frac{3 + 7x_1 + 9x_2}{2 + 3x_1 + 4x_2}$$

$$\text{subject to} \quad x_3 = 6 + 2(-x_1) + 3(-x_2)$$

$$x_4 = 5 + 3(-x_1) + 2(-x_2)$$

$$x_1, x_2, x_3, x_4 \geq 0, \quad \text{integer.}$$

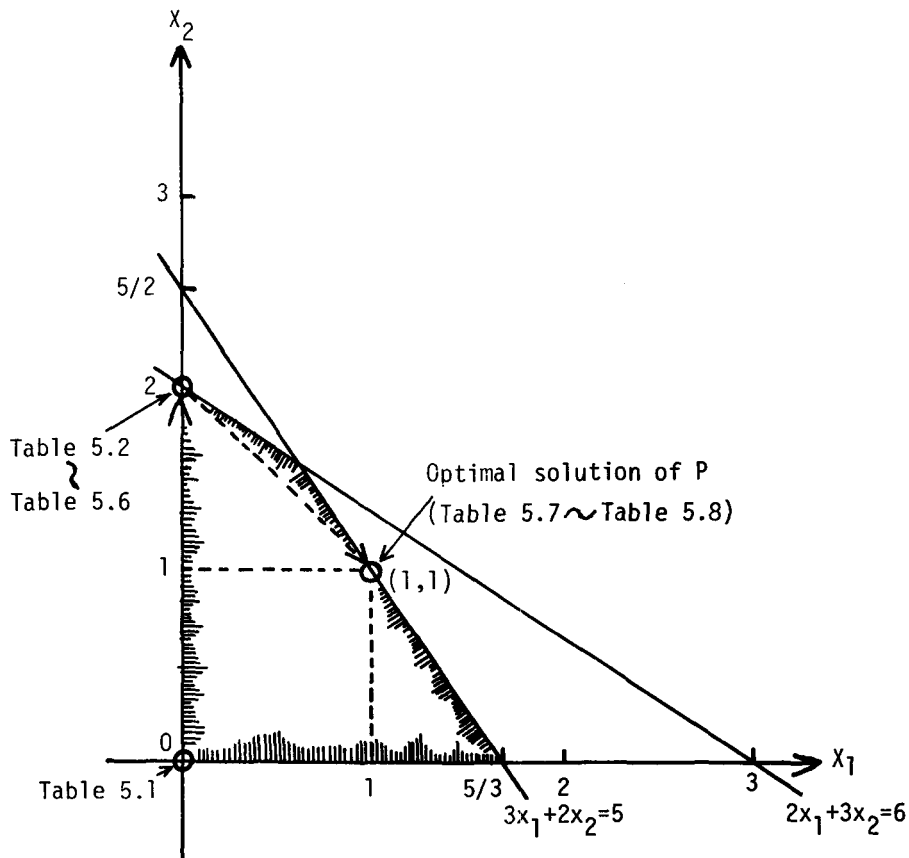


Figure 5.1 Illustration of computation process for the example in Section 5



From the constraints we have

$$(5.1) \quad 0 \leq x_1 \leq \min \left[ \left\lfloor \frac{6}{2} \right\rfloor, \left\lfloor \frac{5}{2} \right\rfloor \right] = 1 (= d_1),$$

$$0 \leq x_2 \leq \min \left[ \left\lfloor \frac{6}{3} \right\rfloor, \left\lfloor \frac{5}{2} \right\rfloor \right] = 2 (= d_2)$$

$$0 \leq x_3 \leq 6 (= d_3), \quad 0 \leq x_4 \leq 5 (= d_4),$$

where  $[x]$  denotes the integer part of  $x$ .

Step 1 : Let  $\lambda \leftarrow \frac{N(x^I)}{D(x^I)} (= \frac{c_0}{d_0} = \frac{3}{2})$  and construct the initial

tableau (Table 5.1) for

$$x^I = (x_1^I, x_2^I, x_3^I, x_4^I) = (0, 0, 6, 5).$$

Step 2 : Since  $\alpha_{01}(\lambda) = -5/2$ ,  $\alpha_{02}(\lambda) = -6/2$ , the dual feasibility condition is not satisfied. We have  $J = \{1, 2\}$  according to Substep 1 of Young's SPA since  $\theta_1 = \min\{6/2, 5/3\} = 5/3 \geq 1$ ,  $\theta_2 = \min\{6/3, 5/2\} = 2 \geq 0$ .

In Substep 2,  $x_2$  column is chosen as a pivot column though  $x_1$  column is also possible. In Substep 5, cut

$$(5.2) \quad s_1 = 2 + 0(-x_1) + 1(-x_2)$$

is generated and added to the tableau (see  $s_1$ -row of Table 5.1) and execute a pivot operation (\* denotes the pivot element). The resulting tableau is given in Table 5.2. Obviously

$$0 \leq s_1 \leq 2$$

follows from (5.1) and (5.2).

Step 3 :  $\alpha_{00}(\lambda) = 6 > 0$ . Thus let  $\lambda \leftarrow \beta_{00}/\gamma_{00} = 21/10$ , and recalculate the  $x_0$ -row of Table 5.2 by using the new  $\lambda$ . Table 5.3 shows the resulting tableau. Return to Step 2.

Step 2 :  $\alpha_{01}(\lambda) = -7/10 < 0$ ; the dual feasibility is not satisfied.

Since  $J = \emptyset$  in this case and the tableau does not have L-row, L-row

$$x_L = 3 + 1(-x_1) + 1(-s_1)$$

Table 5.1. First tableau

	1	$-x_1$	$-x_2$
$x_0 =$	0	$-5/2$	$-6/2$
$z_1 =$	3	-7	-9
$z_2 =$	2	-3	-4
$x_3 =$	6	2	3
$x_4 =$	5	3	2
$s_1 =$	2	0	1*

Table 5.2. Second tableau

	1	$-x_1$	$-s_1$
$x_0 =$	6	$-5/2$	$6/2$
$z_1 =$	21	-7	9
$z_2 =$	10	-3	4
$x_3 =$	0	2	-3
$x_4 =$	1	3	-2
$x_2 =$	2	0	1

Table 5.3. Third tableau

	1	$-x_1$	$-s_1$
$x_0 =$	0	$-7/10$	$6/10$
$z_1 =$	21	-7	9
$z_2 =$	10	-3	4
$x_3 =$	0	2	-3
$x_4 =$	1	3	-2
$x_2 =$	2	0	1
$x_L =$	3	1	1
$s_2 =$	0	1*	-2

Table 5.4. Fourth tableau

	1	$-s_2$	$-s_1$
$x_0 =$	0	$7/10$	$-8/10$
$z_1 =$	21	7	-5
$z_2 =$	10	3	-2
$x_3 =$	0	-2	1*
$x_4 =$	1	-3	4
$x_2 =$	2	0	1
$x_L =$	3	-1	3
$x_1 =$	0	1	-2

is added in *Substep 3* (see  $x_L$ -row of *Table 5.3*;  $\alpha_{L0}=3(=1+2)$  since  $0 \leq x_1 \leq 1$  and  $0 \leq s_1 \leq 2$ ). According to the rule for the stationary cycle of Young's SPA,  $x_1$ -column is selected as the pivot column in *Substep 4*. Then cut row is added to the tableau (see  $s_2$ -row of *Table 5.3*).

$$0 \leq s_2 \leq 4$$

follows from  $s_2$ -row and  $0 \leq s_1 \leq 2$ . The resulting tableau after a pivot operation is given in *Table 5.4*.

Step 3 :  $\alpha_{00}(\lambda)=0$ . Return to Step 2.

Step 2 :  $\alpha_{00}(\lambda)=-8/10<0$ ; the dual feasibility is not satisfied.

Although  $J=\emptyset$ , *Substep 3* is skipped since *Table 5.4* has L-row. In *Substeps 4* and *5*, a cut is generated. In this case, however, the generated cut is the same as  $x_3$ -row, and  $x_3$ -row is used as the pivot row. After a pivot operation, *Table 5.5* is obtained.

Step 3 :  $\alpha_{00}(\lambda)=0$ . Return to Step 2.

Step 2 :  $\alpha_{01}(\lambda)=-9/10<0$ . According to *Substeps 4* and *5*, add a cut

$$s_3 = 0 + 1(-s_2) + (-1)(-x_3)$$

to the tableau, and execute a pivot operation, to obtain *Table 5.6*.

Step 3 :  $\alpha_{00}(\lambda)=0$ . Return to Step 2.

Step 2 :  $\alpha_{02}(\lambda)=-1/10<0$ . Since  $J=\{2\}$ , in this case, a cut is generated according to *Substeps 2* and *5*. (Note that L-row is deleted). The generated cut is the same as  $x_4$ -row. After a pivot operation, *Table 5.7* is obtained.

Step 3 :  $\alpha_{00}(\lambda)=1/10>0$ . Then let  $\lambda+\beta_{00}/\gamma_{00}=19/9$  and recalculate  $x_0$ -row. *Table 5.8* is obtained.

Step 2 :  $\alpha_{01}(\lambda)=3/9>0$ ,  $\alpha_{02}(\lambda)=1/9>0$ ; the dual feasibility is satisfied.

Go to Step 4.

Step 4 : Terminate. An optimal solution  $x^0$  is given by

$$x_1^0 = x_2^0 = x_3^0 = 1, \quad x_4^0 = 0$$

$$N(x^0)/D(x^0)=\lambda=19/9.$$

Table 5.5. Fifth tableau

	1	$-s_2$	$-x_3$
$x_0 =$	0	$-9/10$	$8/10$
$z_1 =$	21	-3	5
$z_2 =$	10	-1	2
$x_4 =$	1	5	4
$x_2 =$	2	2	-1
$x_1 =$	0	-3	2
$x_L =$	3	5	-3
$s_3 =$	0	1*	-1

Table 5.6. Sixth tableau

	1	$-s_3$	$-x_3$
$x_0 =$	0	$9/10$	$-1/10$
$z_1 =$	21	3	2
$z_2 =$	10	1	1
$x_4 =$	1	-5	1*
$x_2 =$	2	-2	1
$x_1 =$	0	3	-1
$x_L =$	3	-5	2

Table 5.7. Seventh tableau

	1	$-s_3$	$-x_4$
$x_0 =$	$1/10$	$4/10$	$1/10$
$z_1 =$	19	13	-2
$z_2 =$	9	6	-1
$x_3 =$	1	-5	1
$x_2 =$	1	3	-1
$x_1 =$	1	-2	1

Table 5.8. Eighth tableau

	1	$-s_3$	$-x_4$
$x_0 =$	0	$3/9$	$1/9$
$z_1 =$	19	13	-2
$z_2 =$	9	6	-1
$x_3 =$	1	-5	1
$x_2 =$	1	3	-1
$x_1 =$	1	-2	1

## 6. Proof of Finiteness and Validity

**Theorem 6.1** (Finiteness and Validity). The procedure given in Section 4 terminates in a finite number of iterations and, upon termination, produces an optimal solution of P.

**Proof.** Finiteness : (a) From Assumption 2 of Section 2, the number of feasible solution in S is finite.

(b) Each  $P(\lambda)$  has the same feasible region S, and each tableau represents a feasible solution since the primal feasibility is assumed. When  $\lambda$  is updated in Step 3 or Step 5, the new  $\lambda$  satisfies,

$$\lambda = \frac{\beta_{00}}{\gamma_{00}} = \frac{N(\bar{x})}{D(\bar{x})},$$

where  $\bar{x}$  is the feasible solution represented by the tableau. By Theorem 3.3, this new  $\lambda$  is strictly greater than the old  $\lambda$ .

(c) It is known that, for each  $\lambda$ , condition  $\alpha_{00}(\lambda) > 0$  (then  $\lambda$  is updated) or a dual feasibility (termination) is satisfied after a finite number of pivot operations. (This property was first proved by Young [13, 14] under the assumption that all coefficients in a tableau are integers. Salkin, Schroff and Mehta [12] generalized this property to the case in which each  $\alpha_{ij}$  is rational. Note that  $\alpha_{0j}(\lambda)$  is rational in our algorithm.) (a) (b) (c) together prove the finiteness.

Validity : When Step 6 is reached, the tableau satisfies both primal and dual feasibility conditions, and hence it gives an optimal solution of  $P(\lambda)$  for the current  $\lambda$ . Since the tableau also satisfies  $\alpha_{00}(\lambda) = 0$ , this solution is an optimal solution of P by Theorem 3.1.  $\square$

## 7. Discussion

It is shown in this paper that the primal cutting plane algorithm, Young's SPA, for the ordinary integer programming problem can be easily modified to handle the integer fractional programming problem. In view of this result, it appears possible to modify other primal algorithms such as *Glover's simplified primal algorithm* [4] (which is favorably compared to Young's SPA in [10] from the view point of computational efficiency) in a similar manner to accept the fractional problem. However, a straightforward application of the present technique to Glover's SPA seems to cause a difficulty that the required property of the reference equation (which plays in Glover's SPA, a role similar to the L-row of Young's SPA) is no longer preserved when  $\lambda$  is updated.

Thus it would be a subject for future research to find a generation method of the reference equation when  $\lambda$  is updated.

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(The Authors' address : Department of Applied Physics, Faculty of Engineering,  
Osaka University, Suita Japan.

Department of Applied Mathematics and Physics, Faculty  
of Engineering, Kyoto University, Kyoto, Japan.)

Appendix. Description of Cut Generation Procedure by Young's SPA ([14])

The generation of a cut and the selection of a pivot element in Young's SPA are done according to the following rule.  $\alpha_j$  ( $j=1,2,\dots,n$ ) is used to denote the  $j$ -th column of a tableau.  $J$  computed during computation determines whether the present cycle is transitive or stationary;  $J \neq \emptyset$  implies that  $\alpha'_{00} > \alpha_{00}$  holds in the next tableau (i.e., it is a transition cycle), while  $J = \emptyset$  implies that  $\alpha'_{00} = \alpha_{00}$  holds in the next tableau (i.e., a stationary cycle).

Substep 1 : (This substep is entered from Step 2 of the main algorithm in Section 4): Calculate  $\theta_j = \min\{\alpha_{i0}/\alpha_{ij} \mid \alpha_{ij} > 0, i \text{ does not correspond to } x_0\text{-row, } z_1\text{-row, } z_2\text{-row or } L\text{-row}\}$  for each  $j$  satisfying  $\alpha_{0j}(\lambda) < 0$ .  $J \leftarrow \{j \mid \theta_j \geq 1\}$ .

Go to Substep 2 if  $J \neq \emptyset$ , and go to Substep 3 if  $J = \emptyset$ .

Substep 2 : If the current tableau has  $L$ -row, delete it from the tableau. Select any column  $\alpha_{j_0}$ ,  $j_0 \in J$ , as the pivot column. Go to Substep 5.

Substep 3 : If the current tableau has  $L$ -row, go to Substep 4. Otherwise add the following  $L$ -row to the tableau, and go to Substep 4.

$$x_L = \alpha_{L0} + \sum_{j=1}^n (-t_j) \quad ,$$

where

$$\alpha_{L0} = \sum_{j=1}^n d_j$$

and  $d_j$  is an upper bound of  $t_j$ , i.e.,  $0 \leq t_j \leq d_j$  for any  $x \in S$ . (The example in Section 5 includes a method for obtaining  $d_j$ .)

Substep 4 : Let  $\alpha_{Lj}$  ( $j=1,2,\dots,n$ ) denote the elements in the  $L$ -row. For each  $\alpha_j$ ,  $j=1,2,\dots,n$ , that has  $\alpha_{Lj} > 0$ , calculate column

$$R_j = (\alpha_{0j} / \alpha_{Lj}, \alpha_{1j} / \alpha_{Lj}, \dots, \alpha_{m'j} / \alpha_{Lj})^T,$$

where  $T$  denotes transpose, and select the lexicographically smallest column among them as the pivot column  $\alpha_{j_0}$ . Go to Substep 5.

Substep 5 : Calculate  $i_0$  such that



$$\alpha_{i_0 0} / \alpha_{i_0 j_0} = \min \{ \alpha_{i_0} / \alpha_{i j_0} \mid \alpha_{i j_0} > 0, i=1, 2, \dots, m' \},$$

and generate the following Gomory cut to adjoin the tableau.

$$s = \lfloor \alpha_{i_0 0} / \mu \rfloor + \sum_{j=1}^n \lfloor \alpha_{i_0 j} / \mu \rfloor (-t_j),$$

where  $\mu = \alpha_{i_0 j_0} (> 0)$  and  $\lfloor y \rfloor$  denotes the integer part of  $y$ .

Execute a pivot operation on the pivot element  $\lfloor \alpha_{i_0 j_0} / \mu \rfloor (=1)$  of the  $s$ -row.

(Then return to Step 3 of the main algorithm).