

A Remark on Solving the Set-Partitioning Problem
by Dual All Integer Algorithm

Kakuzo Iwamura
Mathematics
Josai University
Sakado, Saitama

Abstract

A careful consideration when one solves the set-partitioning problem by dual all integer algorithm is presented.
It saves both computing time and memory size.

[1]. Introduction

A Set-Partitioning Problem,

$$\begin{aligned} \text{minimize } x_0 &= \sum_{j=1}^n c_j x_j \\ \text{subject to } \sum_{j=1}^n a_{ij} x_j &= 1 \quad (1 \leq i \leq m), \quad x_j: \text{binary} \quad (1 \leq j \leq n), \end{aligned} \quad (1.1)$$

where c_j positive integer, $a_{ij}=0$ or 1 can be solved by Dual All Integer Algorithm[1,2]. Salkin & Koncal[4,5,6] transformed this problem to the Set-Covering Problem,

$$\begin{aligned} \text{maximize } u_0 &= \sum_{j=1}^n (c_j + Lh_j) (-x_j) \\ \text{subject to } \sum_{j=1}^n a_{ij} x_j &\geq 1 \quad (1 \leq i \leq m), \quad x_j: \text{binary} \quad (1 \leq j \leq n), \end{aligned} \quad (1.2)$$

where integer L is greater than $\sum_{j=1}^n c_j$, $h_j = \sum_{i=1}^m a_{ij}$ [1,3] and solved the original Set-Partitioning Problem successfully.

Setting $x_{n+i} = \sum_{j=1}^n a_{ij} x_j - 1 \quad (1 \leq i \leq m)$, they applied Dual All Integer Algorithm to the dual feasible all integer tableau as follows[1,2];

$$\begin{array}{rccccccc} & & 1 & -x_1 & -x_2 & \dots & -x_n & & \\ u_0 & & 0 & c_1 + Lh_1 & c_2 + Lh_2 & \dots & c_n + Lh_n & & \\ x_{n+1} & & -1 & -a_{11} & & & & -a_{1n} & \\ x_{n+2} & & -1 & & & & & -a_{2n} & \\ \cdot & & & & & & & & \\ \cdot & = & & & & & & & \\ \cdot & & & & & & & & \\ x_{n+m} & & -1 & -a_{m1} & & & & -a_{mn} & \end{array} \quad (1.3)$$

Maximum tableau size could grow as large as $(m+n+2)(n+1)$ including a cut row.

[2]. Another Transformation

Let's consider another transformation which transforms (1.1) to

$$\begin{aligned} \text{maximize } v_0 &= -\sum_{j=1}^n c_j x_j \\ \text{subject to } \sum_{j=1}^n a_{ij} x_j &= 1 \quad (1 \leq i \leq m), x_j \geq 0, \text{ integer } (1 \leq j \leq n), \end{aligned} \quad (2.1)$$

where $v_0 = -x_0$.

Let M be any integer greater than the minimal value x_0 of (1.1), for example

$$M = \sum_{j=1}^n c_j + 1, \text{ then we see that}$$

$$v(2.1) > -M \quad (2.2)$$

as $v(1.1) = -v(2.1)$, where $v(P)$ denotes the optimal value of the 0-1 integer programming problem (P).

Consider one more problem such as

$$\begin{aligned} \text{maximize } w_0 &= -\sum_{j=1}^n c_j x_j - M \sum_{i=1}^m x_{n+i} \\ \text{subject to } \sum_{j=1}^n a_{ij} x_j - x_{n+i} &= 1 \quad (1 \leq i \leq m), x_u \geq 0 \text{ integer } (1 \leq i \leq m+n). \end{aligned} \quad (2.3)$$

We easily see that the following properties hold.

Property a; (2.3) has a dual feasible integer solution $x_j = 0$ ($1 \leq j \leq n$), $x_{n+i} = -1$ ($1 \leq i \leq m$) with the same dual feasible all integer tableau as (1.3), u_0, L replaced by w_0, M .

Property b; (2.1) has a feasible integer solution if and only if (2.3) has a feasible integer solution whose objective function value w_0 is

greater than $-M$.

$$\text{Property c; } v(2.3) \geq -\sum_{j=1}^n c_j - mM$$

From these properties, we can obtain an optimal integer solution of (2.3) after finite iterations of Dual All Integer Algorithm. Moreover we have,

$$v(2.3) \begin{cases} > -M & \text{iff every optimal integer solution of (2.3) is an optimal} \\ & \text{integer solution of (2.1) \& v(2.3) = v(2.1),} \\ \leq -M & \text{iff (2.3) is infeasible,} \end{cases}$$

so that we get the next Procedure d.

Procedure d; Every time any variable x_u ($n+1 \leq u \leq n+m$) becomes nonbasic in the course of dual pivoting, we can drop x_u and its corresponding column from the tableau.

[3]. Example

I quote Dual All Integer Algorithm from [1].

Step 0: (Preparation) Prepare simplex tableau,

$$x_{B_i} = y_{i0} + \sum_{j \in R} y_{ij} (-x_j), \quad (0 \leq i \leq m), \quad (3.1)$$

where $x_{B_0} = x_0$ = objective function value, x_{B_i} ($1 \leq i \leq m$) are basic variables, x_j ($j \in R$) are nonbasic variables. A vector $v \neq 0$ is called lexicographically positive if its first nonzero component is positive. We use notation $v \stackrel{L}{>} 0$ to denote v lexicographically positive. We use y_j to denote the j -th column of the simplex tableau (3.1). Simplex tableau (3.1) is called dual feasible if $y_j \stackrel{L}{>} 0$ for all $j \in R$, all integer if y_{ij} ($0 \leq i \leq m$, $0 \leq j \leq n$) are all integers. $[u]$ denotes the largest integer less than or equal to u .

Step 1: (Initialization) Begin with a dual feasible all integer tableau (3.1).

Go to Step 2.

Step 2: (Test for optimality) If the solution is primal feasible, it is optimal to (3.1). STOP. If not, go to Step 3.

Step 3: (Cutting and pivoting) Choose a source row ($i \neq 0$) in the tableau with $y_{i0} < 0$, say $i=r$. The topmost row with $y_{i0} < 0$ must be chosen at least periodically. Select the lexicographically smallest column with $y_{rj} < 0$, say $j=k$, as the pivot column. Compute \bar{h} by

$$\bar{h} = \min_{j \in R_r} \frac{\bar{M}_j}{y_{rj}}$$

where $R_r = \{j \in R \mid y_{rj} < 0\}$, $\bar{M}_k = -1$, $\bar{M}_j = \min \{u \mid y_j + u y_k \stackrel{L}{>} 0, u \text{ integer}\}$ for $j \in R_r \setminus \{k\}$.

If $\bar{h}=1$, execute one dual simplex iteration with pivot element y_{rk} .

If $\bar{h}<1$, adjoin the cut

$$s = [hy_{r0}] + \sum_{j \in R} [hy_{rj}](-x_j)$$

with $h = \bar{h}$, to the bottom of the tableau. Execute a dual simplex iteration

with s as the departing variable and x_k as the entering variable. In any case,

if x_k is a slack from a cut, delete the x_k row. Return to Step 2.

To see the power of Procedure d, we take the Example from [1, page 315].

$$\begin{array}{rcl} \text{minimize} & 3x_1 + 7x_2 + 5x_3 + 8x_4 + 10x_5 + 4x_6 + 6x_7 + 9x_8 & \\ & x_1 + x_2 & = 1 \\ & & x_3 + x_4 + x_5 = 1 \\ & & & x_5 + x_6 + x_7 = 1 \\ & & & & x_7 + x_8 = 1 \\ & x_2 & + x_4 & + x_6 & = 1 \end{array}$$

We start with dual feasible all integer tableau (3.2) which is obtained

through replacing u_0, L by $w_0, M = \sum_{j=1}^8 c_j + 1 = 53$.

$$\begin{array}{rcccccccc}
 & & \overset{k}{1} & -x_1 & -x_2 & -x_3 & -x_4 & -x_5 & -x_6 & -x_7 & -x_8 \\
 w_0 & 265 & 56 & 113 & 58 & 114 & 116 & 110 & 112 & 62 & \\
 r)x_9 & -1 & \textcircled{-1} & -1 & & & & & & & \\
 x_{10} & -1 & & & -1 & -1 & -1 & & & & \\
 x_{11} = & -1 & & & & & -1 & -1 & -1 & & \\
 x_{12} & -1 & & & & & & & -1 & -1 & \\
 x_{13} & -1 & & -1 & & -1 & & -1 & & &
 \end{array} \quad (3.2)$$

$r=1$, $R_r = \{1, 2\}$, $k=1$, $\bar{M}_1 = -1$, $\bar{M}_2 = -2$, $y_{rk} = -1$ (circled) gives $\bar{h}=1$. Pivoting on y_{rk} makes x_1 basic, x_9 nonbasic so that we may drop x_9 column from the new tableau (3.3).

$$\begin{array}{rcccccccc}
 & & \overset{k}{1} & -x_2 & -x_3 & -x_4 & -x_5 & -x_6 & -x_7 & -x_8 \\
 w_0 & 209 & 57 & 58 & 114 & 116 & 110 & 112 & 62 & \\
 x_1 & 1 & 1 & & & & & & & \\
 r)x_{10} & -1 & & \textcircled{-1} & -1 & -1 & & & & \\
 x_{11} = & -1 & & & & -1 & -1 & -1 & & \\
 x_{12} & -1 & & & & & & -1 & -1 & \\
 x_{13} & -1 & -1 & & -1 & & -1 & & &
 \end{array} \quad (3.3)$$

$r=2$, $k=2$, $\bar{M}_2 = -1$, $\bar{M}_3 = -1$, $\bar{M}_4 = -2$, $y_{rk} = -1$ (circled) gives $\bar{h}=1$. Pivoting on y_{rk} makes x_3 basic, x_{10} nonbasic so that we may drop x_{10} column from the next tableau (3.4).

$$\begin{array}{rcccccccc}
 & & & & \ell & & & & \\
 & & 1 & -x_2 & -x_4 & -x_5 & -x_6 & -x_7 & -x_8 \\
 w_0 & 151 & 57 & 56 & 58 & 110 & 112 & 62 & \\
 x_1 & & 1 & & & & & & \\
 x_3 & & & 1 & & 1 & & & \\
 r) x_{11} = & -1 & & & \textcircled{-1} & -1 & -1 & & \\
 x_{12} & & -1 & & & & & -1 & -1 \\
 x_{13} & & & -1 & -1 & -1 & & & -1
 \end{array} \tag{3.4}$$

Doing in this way, i.e.,

x_5 basic, x_{11} nonbasic drop x_{11} column;

x_7 basic, x_{12} nonbasic drop x_{12} column;

x_6 basic, x_{13} nonbasic drop x_{13} column;

x_4 basic, x_5 nonbasic drop none,

we get final tableau (3.5) which is optimal.

$$\begin{array}{rcccc}
 & & 1 & -x_2 & -x_5 & -x_8 \\
 w_0 & -17 & 1 & 4 & 4 & \\
 x_1 & & 1 & & & \\
 x_3 & & 0 & -1 & 2 & -1 \\
 x_4 = & 1 & 1 & -1 & 1 & \\
 x_7 & & 1 & & & 1 \\
 x_6 & & 0 & & 1 & -1
 \end{array} \tag{3.5}$$

As $v(3.5) = -17 > -53$, we see that $x_1 = x_4 = x_7 = 1$, $x_j = 0$ (otherwise), $x_0 = 17$ is an optimal solution. Final tableau size is half as large as the original. We also do away with needless calculations for the deleted columns.

References

- [1] Garfinkel, R.S., and Nemhauser, G.L., "Integer Programming", John Wiley & Sons, 1972
- [2] Hu, T.C., "Integer Programming and Network Flows", Addison-Wesley, 1970
- [3] Lenke, C., Salkin, H., and Spielberg, K., "Set Covering by Single Branch Enumeration with Linear Programming Subproblems", Oper. Res. 19 (1971), 998-1022
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