

**[Muthuselvi, 5(8): August 2018]****ISSN 2348 - 8034****DOI- 10.5281/zenodo.1405513****Impact Factor- 5.070**

## **GLOBAL JOURNAL OF ENGINEERING SCIENCE AND RESEARCHES**

### **A STUDY OF REDUNDANT CONSTRAINTS IN LINEAR PROGRAMMING MODELS**

**M. Muthuselvi**

Assistant Professor, Department of Mathematics, Thiruvalluvar College Papanasam

**ABSTRACT**

Linear Programming is a mathematical technique to help plan and to achieve the best outcome. It will make decisions relative to allocate resources and find the minimum or maximum value of the objective. While formulating a linear programming model, decision makers often tend to include all the possible constraints and variables although some of them may not be binding at the optimal solution. In fact, LP models almost contain a significant number of redundant constraints and variables. Therefore it is worthwhile to devote some efforts in solving for considerable reduction in the size of the problem. In this aspect, this paper proposes to identify which constraints and variables most likely to be tight at optimality.

**Keywords** - linear programming models, redundant constraints, variables decrease method.

**I. INTRODUCTION**

Linear programming models consist of an objective function which is to be maximized or minimized subject to a set of constraints. Many researchers have proposed different algorithms to solve linear programming models. The computational complexity of any linear programming model depends on the number of constraints and variables. In solving linear programming models, it is acknowledged that redundancies do exist in most of the practical linear programming models.

Telgen[2] first defined strongly and weakly redundant constraints. Boneh[3] first defined absolute and relative redundancy. Boneh and caron[4] first discussed weakly and strongly redundant constraints. Many researchers have proposed different algorithms to identify the redundancies and removed them to get a reduced model for linear programming. Thompson [5] use techniques for removing nonbinding constraints. Paulraj et al [6] proposed a heuristic method to identify redundant constraints by using the intercept matrix of constraints of a linear programming problem. Paulraj and Sumathi[7] introduced a comparative study of redundant constraints identification methods in linear programming models. Sumathi and Paulraj[9] submitted a new approach proposed for reducing time and more data manipulation by selecting a restrictive constraint in linear programming models to identify the redundant constraints.

**II. DEFINITION AND CLASSIFICATION OF REDUNDANT CONSTRAINT****Redundant Constraints**

A redundant constraint is a constraint that can be removed from a system of linear constraints without changing the feasible region [7].

**Binding Constraint**

Binding constraint is the one which passes through the optimal solution point. It is also called a relevant constraint.

Consider the following system of  $m$  nonnegative linear inequality constraints and  $n$  variables  $m \geq n$  :

$$AX \leq b, X \geq 0 \quad (2.1)$$



**[Muthuselvi, 5(8): August 2018]**

**ISSN 2348 - 8034**

**DOI- 10.5281/zenodo.1405513**

**Impact Factor- 5.070**

where  $A \in R^{m \times n}$ ,  $b \in R^m$ ,  $X \in R^n$

Let  $A_i X \leq b_i$  be the  $i^{\text{th}}$  constraint of the system (2.1) and let  $S = \{X \in R^n / A_i X \leq b_i, X \geq 0\}$ , be the feasible region associated with system (2.1).

Let  $S_k = \{X \in R^n / A_i X \leq b_i, X \geq 0, i \neq k\}$  be the feasible region associated with the system of equations  $A_i X \leq b_i, i = 1, \dots, m, i \neq k$ . The  $k^{\text{th}}$  constraint  $A_k X \leq b_k$  ( $1 \leq k \leq m$ ) is redundant for the system (2.1) if and only if  $S = S_k$ .

### III. ALGORITHM

This method [10] identifies the relevant set of constraints which finds out the optimal solution for the linear programming models.

The steps of the algorithm are as follow:

1. Construct a matrix of intercepts of all the decision variables formed by each of the resource constraints along the respective coordinate axis

$$\Theta_{ji} = \frac{b_i}{a_{ij}}, \quad a_{ij} > 0, i = 1, 2, \dots, m \text{ and } j = 1, 2, \dots, n.$$

Find the  $\min \{\Theta_{ji}\} = \beta_j$  in each row  $j$ .

2. Set  $L = \{i : \min_{\text{at least one } j} \{\Theta_{ji}\} \cup \{\Theta_{ji}\} < \max \beta_j\}$ , for a  $j\}$ ,  $L$  is the index of the selected set of constraints and assumes it contains  $p$  number of constraints,  $p \leq m$ .

3. Set  $J = \{m + j, j = 1, 2, 3, \dots, n\}$

4. Find  $\alpha_k = \frac{a_k C}{\|a_k\|}$ ,  $K \in L \cup J$ .

5. Set  $q = 1$ .

Solve the relaxed problem

$$\text{Maximize } Z = C^T X$$

Subject to the constraints

$$A_r X \leq b_r, A_s X \leq b_s$$

$$x \geq 0,$$

where the  $r^{\text{th}}$  and  $s^{\text{th}}$  constraints are selected by  $r = (\max_k a_k)$  and  $s = \arg(\min_k a_k)$ , provided  $\alpha_k > 0$ . Let the solution obtained, be  $x(q)$ . Check for violated constraints. If none is found, stop. Since  $x(q)$  is the optimal solution. Otherwise

6. Take  $l = \arg(\max \alpha_k)$  of the violated constraints if  $q$  is odd or Take  $l = \arg(\min \alpha_k)$  of the violated constraints if  $q$  is even. Set  $q = q + 1$ . Append the violated constraint corresponding to index  $l$  to the relaxed problem and go to step 5.

### IV. VARIABLE DECREASE METHOD

In this section the use of variable decrease method [8] for linear programming model involving two variables and then, it will extend the algorithm involving 'n' variables.

Consider the following linear programming model involving two variables ((M) denotes given linear programming model)



**[Muthuselvi, 5(8): August 2018]**

**ISSN 2348 - 8034**

**DOI- 10.5281/zenodo.1405513**

**Impact Factor- 5.070**

$$(M) \text{ Maximize } z = \sum_{j=1}^2 c_j x_j$$

Subject to

$$\sum_{j=1}^2 a_{ij} x_j \leq b_i, i=1, 2, 3, \dots, m.$$

$$x_j \geq 0, j=1, 2$$

where  $c_j \geq 0, a_{ij} \geq 0, b_i \geq 0 \forall i$  and  $j$ .

### Algorithm

1. Find the minimum of the maximum possible integral values of  $x_1$  and  $x_2$ . Say  $U$  which corresponds to  $x_1 \leq U$

denotes the largest integral value which is  $\leq$  or  $= k$ .

Now, maximum possible integral value of  $x_j$  in the model (M),  $x_j^0$  is given by

$$x_j^0 = \text{Minimum of } \left\{ \left\lfloor \frac{b_i}{a_{ij}} \right\rfloor, i=1, 2, 3, \dots, n \right\}, j=1, 2$$

2.  $z(M)$  denotes the optimal value of  $z$  in the problem (M).

Consider the following (EM):

(EM): Find the value of  $z$  and the integral values of  $x_1$  and  $x_2$  such that

$$\text{Maximize } \left\{ \begin{array}{l} z = \sum_{j=1}^2 c_j x_j \\ (x_1, x_2) : P \text{ or } Q \end{array} \right\}$$

Where

$$P = \left\{ \begin{array}{l} (x_1, x_2) : x_1 \leq U \\ x_2 = \text{minimum} \left\{ \left\lfloor \frac{b_i - a_{i1} x_1}{a_{i2}} \right\rfloor \right\} \\ i=1, 2, 3, \dots, m \end{array} \right\}$$

Where  $U$  is the maximum possible integral value of  $x_1$  and



**[Muthuselvi, 5(8): August 2018]**

**ISSN 2348 - 8034**

**DOI- 10.5281/zenodo.1405513**

**Impact Factor- 5.070**

$$Q = \left\{ (x_1, x_2) : x_2 = \{1, 2, \dots, V\} \right. \\ \left. x_1 = \text{minimum} \left\{ \left[ \frac{b_i - a_{i2} x_2}{a_{i2}} \right] \right\} \right. \\ \left. i = 1, 2, 3, \dots, m \right\}$$

Where V is the maximum possible integral value of  $x_2$  and

Now, it will prove a relationship between z and  $z^*$ .

Solve the following equivalent model (EM) of the model (M). Let  $(x_1^0, x_2^0, z^0)$  be a solution of the model (EM).

4. It is easy to prove that  $(x_1^0, x_2^0, z^0)$  is a solution of the model (M)  $\Leftrightarrow (x_1^0, x_2^0, z^0)$  is a solution of the model (EM).

#### Algorithm II (for n variables)

1. Find the minimum of the maximum possible integral values of  $x_j = 1, 2, 3, \dots, n$  in the model (P). Say  $U_r$  which corresponds to  $x_r$ .

2. For each  $\hat{x}_r \in \{0, 1, 2, \dots, U_r\}$ , we have a linear programming model (P( $\hat{x}_r$ )) involving n-1 variables  $x_j$ ,  $j = 1, 2, \dots, n$  and  $j \neq r$  where

$$(P(\hat{x}_r)) \text{ Maximize } z = c_r \hat{x}_r + \sum_{\substack{j=1 \\ j \neq r}}^n c_j x_j$$

subject to

$$\sum_{\substack{j=1 \\ j \neq r}}^n a_{ij} x_j \leq b_i - a_{ir} \hat{x}_r, i = 1, 2, 3, \dots, m$$

$$x_j \geq 0, j = 1, 2, 3, \dots, n, j \neq r$$

3. Find the minimum of the maximum possible integral values of  $x_j$ ,  $j = 1, 2, 3, \dots, n$  and  $j \neq r$  in the model (P( $\hat{x}_r$ )),  $x_r = \{0, 1, 2, \dots, U_r\}$ . Say  $U_t$  which corresponds to  $x_t$ .

4. For each  $\hat{x}_t \in \{0, 1, 2, \dots, U_t\}$ , we have a linear programming model (P( $\hat{x}_r, \hat{x}_t$ )) involving n-2 variables  $x_j$ ,  $j = 1, 2, 3, \dots, n$  and  $j \neq r, t$  where

$$(P(\hat{x}_r, \hat{x}_t)) \text{ Maximize } z = c_r \hat{x}_r + c_t \hat{x}_t + \sum_{\substack{j=1 \\ j \neq r, t}}^n c_j x_j$$



[Muthuselvi, 5(8): August 2018]

ISSN 2348 - 8034

DOI- 10.5281/zenodo.1405513

Impact Factor- 5.070

subject to

$$\sum_{\substack{j=1 \\ j \neq r}}^n a_{ij} x_j \leq b_i - a_r \dot{x}_r - a_t \dot{x}_t, i=1,2,3,\dots,m$$

$$x_j \geq 0, j=1,2,3,\dots,n, j \neq r, t$$

5. Do Step 3 and 4 till a set of linear programming models  $(P(\dot{x}_r, \dot{x}_t, \dots, \dot{x}_k))$ ,  $\dot{x}_k \in \{0,1,2,\dots, U_k\}$  involving two variables is obtained.
6. Solve the set of linear programming models  $(P(\dot{x}_r, \dot{x}_t, \dots, \dot{x}_k))$ ,  $\dot{x}_k \in \{0,1,2,\dots, U_k\}$  individually by the algorithm I.
7. Find the maximum of  $z(P(\dot{x}_r, \dot{x}_t, \dots, \dot{x}_k))$ ,  $\dot{x}_k \in \{0,1,2,\dots, U_k\}$ . Say  $z^0$  which corresponds to the problem  $(P(\dot{x}_r, \dot{x}_t, \dots, \dot{x}_k))$  where  $t = \{0,1,2,\dots, U_k\}$ .
8. The solution to the problem (P) is  $z(P) = z^0$ ,  $\dot{x}_k = t$  and solution of the model  $\dot{x}_j$ , for all  $j$ .

## V. CONCLUSION

The computational complexity of any linear programming model depends on the number of constraints and variables of the linear programming model. In this paper, a novel approach is used to identify the redundant constraints and variables. Also, discussed to select the constraints from given constraints and to solve the linear programming models. The variable decrease method is much useful for the decision makers. Computations are very simple and easy. It maintains feasibility as it moves from one vertex to another vertex in the feasible region.

## REFERENCES\

1. Mark Karwan, Valid Lotfi, Jan Telgen, Stanly Zionts, "A study of redundancy in mathematical programming", operational research 1981.
2. J. Telgen, "Identifying redundant constraints and implicit equalities in system of linear constraints," Management Science, vol. 29, no. 10, pp. 1209–1222, 1983.
3. A. Boneh, "Identification of redundancy by a set covering equivalence", operational research, proceeding of the tenth international conference on operational research (Amsterdam, 1984). A. Boneh, S. Boneh, and R. J. Caron, "Constraint classification in mathematical programming," Mathematical Programming, vol. 61, no. 1, pp. 61–73, 1993.
4. G. L. Thompson, F. M. Tonge, and S. Zionts, "Techniques for removing nonbinding constraints and extraneous variables from linear programming problems," Management Science, vol. 12, no. 7, pp. 588–608, 1996.
5. S. Paulraj, C. Chellappan, and T. R. Natesan, "A heuristic approach for identification of redundant constraints in linear programming models," International Journal of Computer Mathematics, vol. 83, no. 8-9, pp. 675–683, 2006.
6. Paulraj S. and Sumathi P. "A Comparative Study of Redundant Constraints Identification Methods in Linear Programming Problems", Mathematical Problems in Engineering, 2010.
7. Pandian, P and Jayalakshmi, M "A new approach for solving a class of pure integer linear programming problems", International Journal of Advanced Engineering Technology, E-ISSN 0976-3945, Vol.3, 2012, 248-251.



**[Muthuselvi, 5(8): August 2018]**

**ISSN 2348 - 8034**

**DOI- 10.5281/zenodo.1405513**

**Impact Factor- 5.070**

8. Paulraj S. and Sumathi P. "Identification of Redundant Constraints in Large Scale Linear Programming Problems with Minimal Computational Effort", *Applied Mathematical Sciences*, Vol. 7, 2013.
9. Sumathi, P and Gangadharan, A " Selection of Constraints with a New Approach in Linear Programming problems, *Applied Mathematical Sciences*, Vol. 8, 2014, no. 27, 1311 – 1321.