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# RANKING OF DODECAGONAL FUZZY NUMBERS FOR SOLVING MULTI OBJECTIVE FUZZY LINEAR PROGRAMMING PROBLEM WITH SIMPLEX METHOD AND GRAPHICAL METHOD

- J. Jon Arockiaraj \*

- N. Sivasankari \*\*

#### Abstract

In this paper, a ranking procedure based on Dodecagonal Fuzzy numbers, is applied to a Multi – Objective Linear Programming Problem (MOLPP) with fuzzy coefficients. By this ranking method any Multi Objective Fuzzy Linear Programming Problem (MOFLPP) can be converted into a crisp value problem to get an optimal solution. This ranking procedure serves as an efficient method wherein a numerical example is taken.

Keywords: Ranking, Dodecagonal Fuzzy Numbers, MOFLPP, Simplex Method, Graphical Method,  $\alpha$  – Level Set.

## INTRODUCTION

Ranking fuzzy number is used in decision – making process in an economic environment. In an organization various activities such as planning, execution, and other process takes place continuously. This requires careful observation of various parameters which are all in uncertain in nature due the competitive business environment globally. In fuzzy environment ranking fuzzy numbers is a very important decision making procedure.

The idea of fuzzy set was first proposed by Bellman and **Zadeh [1]**, as a mean of handling uncertainty that is due to imprecision rather than randomness. The concept of fuzzy linear programming (FLP) was first introduced by **Tanaka [8]** et al. **Zimmerman [2]** introduced fuzzy linear programming in fuzzy environment.

Multi-objective linear programming was introduced by **Zelenly. Lai Y.J – Hawng C.L** considered MOLPP with all parameters having a triangular possibility distribution. They used an auxiliary model and it was solved by MOLPP. **Zimmerman** applied their approach to vector maximum problem by transforming MOFLP problem to a single objective linear programming problem.

Qiu–PengGu, and Bing–Yuan Cao solved fuzzy linear programming based on the representation theorem and on fuzzy number ranking method. In particular, the most convenient methods are based on the concept of comparison of fuzzy numbers by the use ranking function.

# PRELIMINARIES

### Definition

If X is a universe of discourse and x be any particular element of X, then a fuzzy set A defined on X me written as,

\* Head & Asst. Professor, P. G. and Research Department of Mathematics, St. Joseph's College of Arts and Science (Autonomous), Cuddalore, Tamil Nadu, India \*\* Research Scholar, P. G. and Research Department of Mathematics, St. Joseph's College of Arts and Science (Autonomous), Cuddalore, Tamil Nadu, India

Email: jonarockiaraj@gmail.com, prithimahathi@gmail.com, Mobile: 91+ 9443669011

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$$\tilde{A} = \{ (x, \mu_{\tilde{A}}(x)) : x \in X \}$$

The membership function of a fuzzy set  $\tilde{A}$  is denoted by  $\mu_{\tilde{A}}$ , i.e.,  $\mu_{\tilde{A}}$ :  $X \rightarrow [0, 1]$ 

The membership function of a fuzzy set  $\tilde{A}$  has the form,  $\tilde{A}: X \rightarrow [0, 1]$ 

## **Fuzzy Number**

A Fuzzy set  $\tilde{\mathrm{A}}$  of the real line R with membership function

 $\mu_{\tilde{\lambda}}(x): \mathbb{R} \rightarrow [0, 1]$  is called fuzzy number if

- i) A must be normal and convex fuzzy set
- ii) the support of Â, must be bounded
- iii)  $\alpha_{A}$  must be a closed interval for every  $\alpha \in [0, 1]$

## Support

The support of a fuzzy set  $\tilde{A}$ , S ( $\tilde{A}$ ), is the crisp set of all x  $\in$  X such that,

 $S(\tilde{A}) = \mu_{\tilde{A}}(x) > 0$ 

#### $\alpha$ – Cut set or $\alpha$ – Level set:

The  $\alpha$  – Cut set of a fuzzy set  $\tilde{A}$  of the set X is the following crisp set given

 $\widetilde{A}a = \{x \in X \colon \mu_{\widetilde{A}}(x) \ge \alpha$ 

# Normal Fuzzy Set

A Fuzzy set  $\tilde{A}$  of a set X is said to be a normal fuzzy set iff

 $\mu_{\tilde{x}}(x) = 1$  for at least one  $x \in X$ 

#### **DODECAGONAL FUZZY NUMBERS**

A fuzzy number  $\tilde{A}$  is a Dodecagonal Fuzzy Number (DoFN) denoted by

 $\tilde{A} = (a_1, a_2, a_3, a_4, a_5, a_6, a_7, a_8, a_9, a_{10}, a_{11}, a_{12})$ 

Where  $a_1$ ,  $a_2$ ,  $a_3$ ,  $a_4$ ,  $a_5$ ,  $a_6$ ,  $a_7$ ,  $a_8$ ,  $a_9$ ,  $a_{10}$ ,  $a_{11}$ ,  $a_{12}$  are real numbers and its membership function is given below:

$$\begin{pmatrix} 0 & x \le a_1 \\ k_1 \begin{pmatrix} x - a_1 \\ a_1 \le x \le a_2 \end{pmatrix}$$

$$\begin{pmatrix} a_2 - a_1 \end{pmatrix}$$

$$k_1 \qquad \qquad a_2 \le x \le a_3$$

$$\begin{vmatrix} k_1 + (k_2 - k_1) \left( \frac{x - a_3}{a_4 - a_3} \right) & a_2 \le x \le a_3 \\ k_2 & a_4 \le x \le a_4 \end{vmatrix}$$

$$k_{2} + (1 - k_{2}) \left( \frac{x - a_{5}}{a_{6} - a_{5}} \right) \qquad a_{5} \le x \le a_{6}$$

$$\mu_{\tilde{A}}(\mathbf{x}) = \begin{cases} 1 & a_6 \le \mathbf{x} \le a_7 \\ k_1 + (1 - k_2) \left( \frac{a_8 - \mathbf{x}}{2} \right) & a_8 \le \mathbf{x} \le a_7 \end{cases}$$

$$\begin{aligned} \mathbf{k}_{2} + (\mathbf{1} - \mathbf{k}_{2}) \begin{pmatrix} \mathbf{a}_{8} - \mathbf{a}_{7} \end{pmatrix} & \mathbf{a}_{7} \le \mathbf{x} \le \mathbf{a}_{8} \\ \mathbf{k}_{2} & \mathbf{a}_{8} \le \mathbf{x} \le \mathbf{a}_{9} \\ \mathbf{k}_{1} + (\mathbf{k}_{2} - \mathbf{k}_{1}) \begin{pmatrix} \frac{\mathbf{a}_{10} - \mathbf{x}}{\mathbf{a}_{10} - \mathbf{a}_{9}} \end{pmatrix} & \mathbf{a}_{9} \le \mathbf{x} \le \mathbf{a}_{10} \\ \mathbf{k}_{1} & \mathbf{a}_{10} \le \mathbf{x} \le \mathbf{a}_{11} \\ \mathbf{k}_{1} & \mathbf{a}_{10} \le \mathbf{x} \le \mathbf{a}_{11} \end{aligned}$$

 $k_1 \lfloor a_{12} \rfloor$ 

0

$$\frac{-\mathbf{x}}{-\mathbf{a}_{11}} \qquad \mathbf{a}_{11} \le \mathbf{x} \le \mathbf{a}_{12} \\ \mathbf{a}_{12} \le \mathbf{x}$$

Where  $0 < k_1 < k_2 < 1$ 

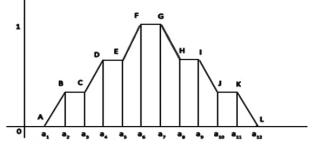


Figure: 9.1. Graphical representation of a dodecagonal fuzzy number for  $x \in [0, 1]$ 

# Arithmetic Operations on Dodecagonal Fuzzy Numbers

Let  $\tilde{A}$  Do =  $(a_1, a_2, a_3, a_4, a_5, a_6, a_7, a_8, a_9, a_{10}, a_{11}, a_{12})$ and  $\tilde{B}$  Do =  $(b_1, b_2, b_3, b_4, b_5, b_6, b_7, b_8, b_9, b_{10}, b_{11}, b_{12})$ be two Dodecagonal Fuzzy Numbers, then

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## *i)* Addition

$$\widetilde{A}_{D_0}(+)\widetilde{B}_{D_0} = (a_1 + b_1, a_2 + b_2, a_3 + b_3, a_4 + b_4, a_5 + b_5, a_6 + b_6, a_7 + b_7, a_8 + b_8, a_9 + b_9, a_{10} + b_{10}, a_{11} + b_{11}, a_{12} + b_{12})$$

$$a_7 + b_{7,} a_8 + b_{8,} a_9 + b_{9,} a_{10} + b_{10,} a_{11} + b_{11,} a_{12}$$

## ii) Subtraction

 $\widetilde{A}_{D_0} (-) \widetilde{B}_{D_0} = (a_1 - b_1, a_2 - b_2, a_3 - b_3, a_4 - b_4, a_5 - b_5, a_6 - b_6, a_7 - b_7, a_8 - b_8, a_9 - b_9, a_{10} - b_{10}, a_{11} - b_{11}, a_{12} - b_{12})$ 

iii) Multiplication

$$\widetilde{A}_{D_{0}}(*) \widetilde{B}_{D_{0}} = (a_{1}*b_{1}, a_{2}*b_{2}, a_{3}*b_{3}, a_{4}*b_{4}, a_{5}*b_{5}$$
$$a_{6}*b_{6},$$
$$a_{7}*b_{7}, a_{8}*b_{8}, a_{9}*b_{9}, a_{10}*b_{10}, a_{11}*b_{11}, a_{12}*b_{12})$$

iv) Division

$$\tilde{A}_{D_{0}}(\div) \ \tilde{B}_{D_{0}} = (a_{1} \div b_{1}, a_{2} \div b_{2}, a_{3} \div b_{3}, a_{4} \div b_{4}, a_{5} \div b_{5}, a_{6} \div b_{6}$$

$$a_7 \div b_{7,} a_8 \div b_{8,} a_9 \div b_{9,} a_{10} \div b_{10,} a_{11} \div b_{11}, a_{12} \div b_{12}$$

#### **Ranking of Dodecagonal Fuzzy Numbers**

A number of approaches have been proposed for the ranking of fuzzy numbers.

In this paper for a dodecagonal fuzzy number  $\tilde{A}_{Do} = (a_{1,} a_{2,} a_{3,} a_{4,} a_{5,} a_{6,} a_{7,} a_{8,} a_{9,} a_{10}, a_{11}, a_{12})$  a ranking method is devised based on the following formula,

$$\mathbf{R} \left( \mathbf{\tilde{A}}_{\mathbf{D}\mathbf{0}} \right) = \begin{pmatrix} 2(a_1 + a_6 + a_7 + a_{12}) + \\ 6(a_2 + a_3 + a_4 + a_9 + a_{10} + a_{11}) \\ +5(a_5 + a_8) \\ \hline 54 \\ \end{pmatrix} \left( \frac{25}{18} \right)$$

Let  $\tilde{\mathbf{A}}_{\text{Do}} = (\mathbf{a}_{1,} \mathbf{a}_{2,} \mathbf{a}_{3,} \mathbf{a}_{4,} \mathbf{a}_{5,} \mathbf{a}_{6,} \mathbf{a}_{7,} \mathbf{a}_{8,} \mathbf{a}_{9,} \mathbf{a}_{10,} \mathbf{a}_{11,} \mathbf{a}_{12})$  and

 $\mathbf{B}_{Do} = (\mathbf{b}_{1,} \mathbf{b}_{2,} \mathbf{b}_{3,} \mathbf{b}_{4,} \mathbf{b}_{5,} \mathbf{b}_{6,} \mathbf{b}_{7,} \mathbf{b}_{8,} \mathbf{b}_{9,} \mathbf{b}_{10,} \mathbf{b}_{11,} \mathbf{b}_{12})$  be two Dodecagonal Fuzzy Numbers, then

$$\widetilde{A}_{D_0} \approx \widetilde{B}_{D_0} R (\widetilde{A}_{D_0}) = R (\widetilde{B}_{D_0})$$
$$\widetilde{A}_{D_0} \ge \widetilde{B}_{D_0} R (\widetilde{A}_{D_0}) \ge R (\widetilde{B}_{D_0})$$

$$\tilde{A}_{Do} \leq \tilde{B}_{Do} R (\tilde{A}_{Do}) \leq R (\tilde{B}_{Do})$$

# METHOD OF SOLVING MULTI OBJECTIVE FUZZY LINEAR PROGRAMMING PROBLEM

This paper we discuss a multi – objective fuzzy linear programming problem in constraints conditions with fuzzy coefficients.

Maximize 
$$Z_1 = f_1 y$$
  
Minimize  $Z_2 = f_2 y$ 

Subject to

$$\tilde{A}_{D_0} X \leq \overline{g}, X \geq 0$$

Where  $f_{ij} = (f_{i1}, f_{i2}, \dots, f_{in})$  is an n – dimensional crisp row vector,

 $\tilde{A}_{Do} = \bar{a}_{ii}$  is an m × n fuzzy matrix,

 $\overline{g} = (g_1, g_2, \dots, g_m)^T$  is an m – dimensional fuzzy line vector and

 $X = (x_1, x_2, x_3, \dots,)^T$  is an n – dimensional decision variable vector.

We now consider a bi – objective fuzzy linear programming problem with constraints having fuzzy coefficients is given by

Maximize 
$$Z_1 = f_{11} x_1 + f_{12} x_2 + \dots + f_{1n} x_n$$
  
Minimize  $Z_2 = f_{11} x_1 + f_{12} x_2 + \dots + f_{1n} x_n$ 

Subject to

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$$\bar{\mathbf{a}}_{i1}\mathbf{x}_1 + \bar{\mathbf{a}}_{i2}\mathbf{x}_2 + \ldots + \bar{\mathbf{a}}_{in}\mathbf{x}_n \le \overline{\mathbf{g}}_i$$

 $x_1, x_2, x_3, \ldots, x_n \ge 0, i = 1, 2, 3, \ldots, m$ 

where fuzzy numbers are dodecagonal,

where

$$\bar{\mathbf{a}}_{i1} = \bar{\mathbf{a}}_{i11}, \ \bar{\mathbf{a}}_{i12}, \ \bar{\mathbf{a}}_{i13}, \ \bar{\mathbf{a}}_{i14}, \ \bar{\mathbf{a}}_{i15}, \ \bar{\mathbf{a}}_{i16}, \ \bar{\mathbf{a}}_{i17}, \ \bar{\mathbf{a}}_{i18}, \ \bar{\mathbf{a}}_{i19}, \ \bar{\mathbf{a}}_{i110}, \ \bar{\mathbf{a}}_{i111}, \ \bar{\mathbf{a}}_{i112} \\ \bar{\mathbf{a}}_{i2} = \bar{\mathbf{a}}_{i21}, \ \bar{\mathbf{a}}_{i22}, \ \bar{\mathbf{a}}_{i23}, \ \bar{\mathbf{a}}_{i24}, \ \bar{\mathbf{a}}_{i25}, \ \dots \ \dots \ , \ \bar{\mathbf{a}}_{i212}$$

$$\begin{split} \boldsymbol{\bar{a}}_{in} &= \boldsymbol{\bar{a}}_{in1}, \, \boldsymbol{\bar{a}}_{in2}, \dots \dots, \boldsymbol{\bar{a}}_{in8}, \boldsymbol{\bar{a}}_{in9}, \boldsymbol{\bar{a}}_{in10}, \boldsymbol{\bar{a}}_{in11}, \boldsymbol{\bar{a}}_{in12} \\ & \boldsymbol{\overline{g}} = \boldsymbol{\overline{g}}_{i1}, \, \boldsymbol{\overline{g}}_{i2}, \boldsymbol{\overline{g}}_{i3}, \dots \dots, \, \boldsymbol{\overline{g}}_{i12} \end{split}$$

By the ranking Algorithm, the above MOFLPP is transformed into a MOLPP is as follows:

Maximize  $Z_1 = f_{11}x_1 + f_{12}x_2 + \dots + f_{1n}x_n$ Minimize  $Z_2 = f_{11}x_1 + f_{12}x_2 + \dots + f_{1n}x_n$ 

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Subject to,  

$$2[(a_{i11} x_1 + a_{i21} x_2 + \dots + a_{in1} x_n) + (a_{i16} x_1 + a_{i26} x_2 + \dots + a_{in6} x_n) + (a_{i17} x_1 + a_{i27} x_2 + \dots + a_{in7} x_n) + (a_{i112} x_1 + a_{i212} x_2 + \dots + a_{in12} x_n)] + 6[(a_{i12} x_1 + a_{i22} x_2 + \dots + a_{in2} x_n) + (a_{i13} x_1 + a_{i23} x_2 + \dots + a_{in3} x_n) + (a_{i14} x_1 + a_{i24} x_2 + \dots + a_{in4} x_n) + (a_{i19} x_1 + a_{i29} x_2 + \dots + a_{in9} x_n) + (a_{i110} x_1 + a_{i210} x_2 + \dots + a_{in10} x_n) + \dots + (a_{in11} x_1 + a_{in11} x_2 + \dots + a_{in11} x_n)] + 5[(a_{i15} x_1 + a_{i25} x_2 + \dots + a_{in5} x_n) + (a_{i18} x_1 + a_{i28} x_2 + \dots + a_{in8} x_n)] + (a_{i19} + 6g_{i11} + 2g_{i12} x_1 + a_{i29} x_2 + \dots + a_{in5} x_n) + (a_{i18} x_1 + a_{i28} x_2 + \dots + a_{in9} x_n)] \leq 2 g_{i1} + 6g_{i2} + 6g_{i3} + 6g_{i4} + 5g_{i5} + 2g_{i6} + 2g_{i7} + 5g_{i8} + 6g_{i9} + 6g_{i10} + 6g_{i11} + 2g_{i12} x_1, x_2, x_3, \dots, x_n \ge 0, i = 1, 2, 3, \dots, m - (*)$$
Using (\*), this can be converted into a single objective

Using (<sub>\*</sub>), this can be converted into a single objective problem subject to the constraints with transformed crisp number coefficients and hence solved accordingly.

Similarly, multi – objective problems with more than two objectives can also be solved using the above procedure, here in the very first stage itself the problem is transformed into a crisp problem and afterwards there will be no more fuzziness in the constraints as well as in the problem.

#### **Simplex Method Algorithm**

Step 1: Determine a starting basic feasible solution.

- **Step 2:** Select an entering variable using the optimality condition Stop if there is no entering variable; the last solution is optimal. Else, go to Step 3.
- **Step 3:** Select a leaving variable using the feasibility condition.
- **Step 4:** Determine the new basic solution. Go to Step 2.

#### **Numerical Example**

Consider,

 $\text{Max } Z=50 \text{ } \text{x}_1+80 \text{x}_2 \\ \tilde{a}_{11} \text{X}_1+\tilde{a}_{12} \text{X}_2 \leq \overline{g}_1$ 

$$\tilde{\boldsymbol{a}}_{_{21}}\boldsymbol{X}_{_{1}}\!+\tilde{\boldsymbol{a}}_{_{22}}\boldsymbol{X}_{_{2}}\!\leq\overline{\boldsymbol{g}}_{_{2}}$$

where

 $\tilde{a}_{11} = (100, 30, 60, 110, 70, 200, 200, 105, 50, 120, 90, 65)$ 

 $\tilde{a}_{12} = (160, 100, 140, 50, 180, 150, 150, 170, 40, 200, 70, 90)$ 

 $\tilde{a}_{21} = (100, 150, 60, 90, 170, 300, 300, 190, 110, 80, 50, 200)$ 

 $\tilde{a}_{22} = (180, 50, 120, 70, 90, 100, 100, 110, 200, 60, 80, 40)$ 

 $\overline{\mathbf{g}}_1 = (800, 1400, 400, 600, 1300, 1000, 1000, 700, 900, 1600, 1200, 1100)$ 

 $\overline{\mathbf{g}}_2 = (500, 1000, 1100, 650, 1250, 800, 1500, 1500, 600, 1300, 400, 700)$ 

Subject to constraints

 $2(100x_1 + 160x_2) + 6(30x_1 + 100x_2) + 6(60x_1 + 140x_2) + 6(110x_1 + 50x_2) +$ 

 $5(70x_1 + 180x_2) + 2(200x_1 + 150x_2) + 2(200x_1 + 150x_2) + 5(105x_1 + 170x_2) +$ 

 $\leq (800 + 1400 + 400 + 600 + 1300 + 1000 + 1000 +$ 700 + 900 + 1600 + 1200 + 1100)

$$2(100x_1 + 180x_2) + 6(150x_1 + 50x_2) + 6(60x_1 + 120x_2) + 6(90x_1 + 70x_2) + 6(90x_1 + 70x_2) + 6(150x_1 + 120x_2) + 6(150x_1 + 12$$

 $5(170x_1 + 90x_2) + 2(300x_1 + 100x_2) + 2(300x_1 + 100x_2) + 5(190x_1 + 110x_2) +$ 

 $6(110x_1 + 200x_2) + 6(80x_1 + 60x_2) + 6(50x_1 + 80x_2) + 2(200x_1 + 40x_2)$ 

 $\leq (500 + 1000 + 1100 + 650 + 1250 + 800 + 1500 + 1500 + 1500 + 600 + 1300 + 400 + 700)$ 

Max 
$$Z = 50x_1 + 80x_2$$
  
Subject to constraints  
4765  $x_1$ +6450  $x_2 \le 12000$ 

$$x_1 + 5320 x_2 \le 11300$$

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## **SIMPLEX METHOD**

### Step 1

			C <sub>j</sub>	50	80	0	0	
	C <sub>B</sub>	X <sub>B</sub>	В	X <sub>1</sub>	X <sub>2</sub>	S <sub>1</sub>	<b>S</b> <sub>2</sub>	RATIO
	0	$S_1$	12000	4765	6450	1	0	12000/6450←
	0	$S_2$	11300	6760	5320	0	1	11300/5320
	Z		0	0	0	0	0	
C	$z_j - Z_j$		-	50	80↑			

## Step 2

Enter X<sub>2</sub> and skip S<sub>1</sub>

		C	50	80	0	0	
C <sub>B</sub>	X <sub>B</sub>	В	X <sub>1</sub>	X <sub>2</sub>	S <sub>1</sub>	S <sub>2</sub>	RATIO
80	v	12000	4765	1	1	0	
80	X <sub>2</sub>	/6450	/6450		/6450		
	0	9045000	18252200	0	-5320	1	
0	S <sub>2</sub>	/6450	/6450		/6450		
Z		148.83	59.10	80	0.01	0	
$C_j - Z_j$		148.83	-9.10	0	-0.01	0	

Maximize Z = 148.83 at  $x_1 = 0$ ;  $x_2 = 1.86$ 

Similarly,

We can calculate Minimize Z = (-Maximize Z)Therefore,

Minimize Z = 84 at  $x_1 = 1.67$ ;  $x_2 = 0$ 

#### **GRAPHICAL METHOD**

Maximize Z=50  $x_1$ +80 $x_2$ Subject to constraints 4765  $x_1$ +6450  $x_2 \le 12000$ 6760  $x_1$ +5320  $x_2 \le 11300$ 

# Solution:

Given: Max Z=50  $x_1$ +80 $x_2$ 

Subject to constraints

4765  $x_1$ +6450  $x_2$ = 12000 -----→(1) 6760  $x_1$ +5320  $x_2$ = 11300 -----→(2)

Put 
$$x_1 = 0$$
 in (1)   
 $6450 x_2 = 12000$   
 $X_2 = 1.86$   
A (0, 1.86)  
Put  $x_2 = 0$  in (1)   
 $4765 x_1 = 12000$   
 $X_1 = 2.51$   
B (2.51, 0)  
Put  $x_1 = 0$  in (2)   
 $5320 x_2 = 11300$   
 $X_2 = 2.12$   
C (0, 2.12)  
Put  $x_2 = 0$  in (2)   
 $6760x_1 = 11300$   
 $X_1 = 1.67$   
D (1.67, 0)  
 $3 x_1 = 1.67$   
D (1.67, 0)  
 $3 x_1 = 1.67$   
D (1.67, 0)

$(x_1, x_2)$	Max Z=50 x <sub>1</sub> +80 x <sub>2</sub>				
A (0,0)	Max Z = 0				
B (0,1.86)	Max Z = 148.8				
C (1.67,0)	Min Z = 83.55				

# COMPARISON OF RESULT OBTAINED BY SIMPLEX METHOD AND GRAPHICAL METHOD

Graphical Method	Simplex Method
Maximize Z=148.8 at	Maximize Z=148.83 at
$x_1 = 0$ and $x_2 = 1.86$	$x_1 = 0$ and $x_2 = 1.86$
Minimize Z=83.55 at	Minimize Z=84 at
$X_1 = 1.67 \text{ and } x_2 = 0$	$X_1 = 1.7$ and $x_2 = 0$

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## RESULT

From the above table, we have obtained the correct value from both Simplex Method and Graphical Method. By comparing the two methods Simplex Method is the best one in which the results are more accurate.

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