

# *Transforming and Solving Multi-objective Quadratic Fractional Programming Problems by Optimal Average of Maximin & Minimax Techniques*

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**Abstract** This paper uses Optimal Average maximin and minimax value techniques to transform multi-objective quadratic fractional programming problems (MOQFPP) to single QFPP. An algorithm is proposed for solving such problems. We also solved the problem by Chandra Sen. technique, mean & median techniques. The results from the latter techniques are obtained and compared to the result of maximin and minimax technique. This work has been tested through several numerical examples; only two of them are presented in this work. The numerical results in this paper indicate that our technique is promising.

**Keywords** MOQFPP, Chandra Sen, Mean, Median, Optimal Average of Maximin & Minimax Techniques

## 1. Introduction

A quadratic fractional form is a mathematical expression of the type  $\frac{c^T x + \frac{1}{2} x^T G_1 x}{d^T x + \frac{1}{2} x^T G_2 x}$  where  $G_1$  and  $G_2$  are  $(n \times n)$  matrix of coefficients and they are symmetric matrixes. All vectors are assumed to be column vectors unless transposed ( $T$ ),  $x$  is an  $n$ -dimensional vector of decision variables,  $c, d$  are the  $n$ -dimensional vector of constants [10]. The term Programming refers to the process of determining a particular program or plan of action. So Quadratic Fractional Programming (QFP) is one of the most important optimization (maximization / minimization) techniques developed in the field of Operations Research (OR).

A new modified simplex method [9] is used to solve special case of a quadratic fractional programming, a solution can be obtained by a set of simultaneous equations. However a unique solution for a set of simultaneous equations in  $n$ -variables  $(x_1, x_2 \dots x_n)$ , at least one of them is non-zero, can be obtained if there are exactly  $n$  relations. When the number of relations is greater than or less than  $n$ , a unique solution cannot exist but a number of trial solutions can be found.

In various practical situations, the problems are seen in which the number of relations is not equal to the number of variables and many of the relations are in the form of inequalities ( $\leq$  or  $\geq$ ) to maximize or minimize a quadratic fractional function of the variables subject to

such conditions. Such problems are known as Quadratic Fractional Programming Problem (QFPP). Quadratic Fractional programming problems have attracted considerable research and interest, since they are useful in production planning, financial and corporate planning, health care and hospital planning.

This work develops a new model of (MOQFPP) and suggested an algorithm to solve it, by using a special case of objective function for (QFPP) as a form  $\frac{(c_1^T x + \alpha)(c_2^T x + \beta)}{(C^T x + \gamma)}$  where  $x, c_1, c_2, C$  are  $(n \times 1)$  column vectors, and  $\alpha, \beta, \gamma$  are scalars and prime ( $T$ ) denoted the transpose of the vector. A new technique optimal average of maximin & minimax is suggested to solve MOQFPP with several methods (Chandra Sen, mean, median) to compare between the results.

In (1983), Chandra Sen [6] defined the multi-objective linear programming problem, and suggested an approach to construct multi-objective function under the limitation that the optimum values of individual problems are greater than zero. This technique is used to solve MOLPP [7] and MOFPP [5] and apply the same technique to solve MOQFPP.

Mean, median methods are used to transform multi-objective to one combined objective function and solve it by previous ways. In (1997) a reference direction approach to multiple objective quadratic-linear programming, that studied by GuangYuan Yu and Pekka Korhonen [2], proposes an interactive procedure for solving multiple criteria problems with one quadratic objective, several linear objectives, and a set of linear constraints. The procedure is based on the use of reference directions and weighted sums. [4] Sulaiman and Sadiq (2006) studied the multi-objective function by solving the multi-objective

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programming problem, using mean and median value. (2006), Salwa and Emad Abass[8] studied and Construct mathematical models for solving multi-objective linear programming (MOLP) problem. Huey-Kuo Chen and Huey - Wen Chou[3] (1996) proposed a fuzzy approach, which induces some methodologies as special cases, for solving the multiple objective linear programming problem. Sulaiman & Abulrahim (2013) studied the transformation technique to solve multi-objective linear fractional programming problem [5]. (2008) Sulaiman and Hamadameen studied optimal transformation technique to solve multi-objective linear programming problem (MOLPP)[7].

In this article, we aim to solve a multi objective quadratic fractional programming problem by optimal average of maximin & minimax (OAxn) which is reported in section (5.3) to minimizes cost and maximizes profit, Irrespective of the number objectives with less computational burden, Computer applications for the algorithm will be discussed by solving numerical examples.

## 2. Quadratic Programming

If the optimization problem assumes the form

$$\max. z \text{ (or min. } z) = \alpha + C^T x + x^T G x$$

subject to:

$$Ax \begin{cases} \leq \\ \geq \\ = \end{cases} b$$

$$x \geq 0$$

Where  $A = (a_{ij})_{m \times n}$ , matrix of coefficients,  $\forall i = 1, 2, \dots, m$  and  $j = 1, 2, \dots, n$ ,

$b = (b_1, b_2, \dots, b_m)^T$ ,  $x = (x_1, x_2, \dots, x_n)^T$ ,  $C^T = (c_1, c_2, \dots, c_n)^T$ ,

and  $G = (g_{ij})_{n \times n}$  is a positive definite or positive semi-definite symmetric square matrix, moreover the constraints are linear and the objective function is quadratic. Such optimization problem is said to be a quadratic programming (QP) problem.[1]

## 3. Mathematical Form of QFPP

The mathematical form of this type of problems is given as follows:

$$\max. z = \frac{(c^T x + \delta + \frac{1}{2} x^T G x)}{(C^T x + \gamma)}$$

Subject to:

$$Ax \begin{cases} \geq \\ \leq \\ = \end{cases} b$$

$$x \geq 0$$

Where  $G$  is  $(n \times n)$  matrix of coefficients with  $G$  is symmetric matrixes. All vectors are assumed to be column vectors unless transposed ( $T$ ). Where  $x$  is an  $n$ -dimensional vector of decision variables,  $c, C$  is the  $n$ -dimensional vector of constants,  $b$  is  $n$ -dimensional vector of constants.  $\gamma, \delta$  are scalars.

In this work the problem that has objective function is tried to be solved has the following form:

$$\max. z = \frac{(c_1^T x + \alpha)(c_2^T x + \beta)}{(C^T x + \gamma)}$$

Subject to:

$$Ax \begin{cases} \geq \\ \leq \\ = \end{cases} b$$

$$x \geq 0$$

$A$  is  $m \times n$  matrix, all vectors are assumed to be column vectors unless transposed ( $T$ ). where  $x$  is an  $n$ -dimensional vector of decision variables,  $c_1, c_2, C$  are the  $n$ -dimensional vector of constants,  $b$  is  $n$ -dimensional vector of constants,  $\alpha, \beta, \gamma$  are scalars.[9]

### 3.1. Solving QFPP by the Following Method

A new Modified Simplex Method is used to solve the numerical example to apply simplex process, first we find  $\Delta k_1^i$ ,  $\Delta k_2^j$  from the coefficient of numerator and denominator of objective function respectively, by using the following formula:

$$\Delta k_1^{ij} = c_{Bi} x_j - c_{ij}, \quad i = 1, 2, \quad j = 1, 2, \dots, m + n$$

$$\Delta k_2^{1j} = C_B x_j - C_{1j}, \quad j = 1, 2, \dots, m + n$$

$$z_1^1 = c_{B1} x_B + \alpha, z_1^2 = c_{B2} x_B + \beta, z_2 = C_B x_B + \gamma$$

$$Z_1 = z_1^1 z_1^2, Z_2 = z_2,$$

$$\Delta \xi_{1j} = z_1^2 \Delta k_1^{1j} + z_1^1 \Delta k_1^{2j}$$

$$\Delta \xi_{2j} = \Delta k_2^{1j}$$

$$Z = Z_1 / Z_2$$

In this approach we define the formula to find  $\Delta_j$  from  $Z_1, Z_2, \Delta \xi_{1j}$  and  $\Delta \xi_{2j}$  as follows:

$$\Delta_j = Z_2 \Delta \xi_{1j} - Z_1 \Delta \xi_{2j}. \text{ For more detail see [9] p. 3756.}$$

### 3.2. Algorithm for Solving QFPP by New Modified Simplex Method

An algorithm to solve QFPP by Modified Simplex Method is presented in reference [9] p. 3756 and p. 3757.

## 4. Multi-Objective Quadratic Fractional Programming Problem

Multi-Objective functions are the ratio of two objective functions that have quadratic objective function in numerator and linear objective function in denominator, this is said to be MOQFPP then can be defined:

$$\left. \begin{aligned}
 \text{Max. } Z_1 &= \frac{(c_{11}^T x + \alpha_1)(c_{21}^T x + \beta_1)}{(c_1^T x + \gamma_1)} \\
 \text{Max. } Z_2 &= \frac{(c_{12}^T x + \alpha_2)(c_{22}^T x + \beta_2)}{(c_2^T x + \gamma_2)} \\
 &\vdots \\
 \text{Max. } Z_r &= \frac{(c_{1r}^T x + \alpha_r)(c_{2r}^T x + \beta_r)}{(c_r^T x + \gamma_r)} \\
 \text{Min. } Z_{r+1} &= \frac{(c_{1r+1}^T x + \alpha_{r+1})(c_{2r+1}^T x + \beta_{r+1})}{(c_{r+1}^T x + \gamma_{r+1})} \\
 &\vdots \\
 \text{Min. } Z_s &= \frac{(c_{1s}^T x + \alpha_s)(c_{2s}^T x + \beta_s)}{(c_s^T x + \gamma_s)}
 \end{aligned} \right\} \quad (4.1)$$

Subject to:

$$Ax = b \quad (4.2)$$

$$x \geq 0 \quad (4.3)$$

Where  $b$  is  $m$ -dimensional vector of constants,  $x$  is  $n$ -dimensional vector of decision variables,  $A$  is a  $m \times n$  matrix of constants,  $r$  is number of objective functions to be maximized,  $s$  is the number of objective functions to be maximized and minimized and  $(s - r)$  is the number of objective functions that is minimized.  $A$  is  $m \times n$  matrix, all vectors are assumed to be column vectors unless transposed ( $T$ ).  $c_{1i}, c_{2i}, C_i$  where  $i=1, \dots, s$  are the  $n$ -dimensional vector of constants,  $\alpha_i, \beta_i, \gamma_i$  where  $i=1, \dots, s$  are scalars.

## 5. Solving MOQFPP by Using the Following Technique

### 5.1. Chandra Sen. Technique

The same approach which was taken by Sen. (1983)[6] is followed here to formulate the constraint objective function for the MOQFPP. Suppose we obtain a single value corresponding to each of the objective functions of the MOQFPP of equation (4.1). They are being optimized individually subject to the constraints (4.2) and (4.3) as follows:

$$\left. \begin{aligned}
 \text{Max. } Z_1 &= \varphi_1 \\
 \text{Max. } Z_2 &= \varphi_2 \\
 &\vdots \\
 \text{Max. } Z_r &= \varphi_r \\
 \text{Min. } Z_{r+1} &= \varphi_{r+1} \\
 &\vdots \\
 \text{Min. } Z_s &= \varphi_s
 \end{aligned} \right\} \quad (5.1.1)$$

Where  $\varphi_1, \varphi_2, \dots, \varphi_s$  are values of the objective functions, which each objective function in 4.1 solved by section 3.3, the level of the decision variable may not necessarily be the same for all optimal solutions in presence of conflicts among objectives. But we require the common set of decision variables to be the best compromising optimal solution that we can determine for the common set of the

decision variables from the following combined objective function, which formulate the MOQFPP given in following equation.

$$\text{Max. } Z = \sum_{i=1}^r \frac{Z_i}{|\varphi_i|} - \sum_{i=r+1}^s \frac{Z_i}{|\varphi_i|} \quad (5.1.2)$$

Where  $\varphi_i \neq 0, i = 1, 2, \dots, s$ . Subject to constraints (4.2) and (4.3), the optimum value of the objective functions  $\varphi_i, i = 1, 2, \dots, s$  may be positive or negative. And  $Z_i$  when  $i=1, \dots, r$  represents the max.z objective functions and when  $i= r+1, \dots, s$  min.z objective function in (4.1). Finally solve equation (5.1.2) with same constraint (4.2) and (4.3) by section 3.2.

### 5.2. Mean and Median Modified Approach

We formulate the combined objective function as follows to determine the common set of decision variables, to solving the MOQFPP by modified approach (using mean and median value), [7], [5].

$$\text{Max. } Z = \sum_{i=1}^r \frac{Z_i}{\text{mean}(MV_i)} - \sum_{i=r+1}^s \frac{Z_i}{\text{mean}(MU_i)} \quad (5.2.1)$$

Subject to the same constraints (4.2), (4.3);

Where  $MV_i = |\varphi_i|$ , for all  $i = 1, 2, \dots, r$ ;

$MU_i = |\varphi_i|$ , for all  $i = r + 1, r + 2, \dots, s$

$$\text{Max. } Z = \sum_{i=1}^r \frac{Z_i}{\text{median}(MV_i)} - \sum_{i=r+1}^s \frac{Z_i}{\text{median}(MU_i)} \quad (5.2.2)$$

Subject to the same constraints (4.2), (4.3);

Where  $MV_i = |\varphi_i|$ , for all  $i = 1, 2, \dots, r$ .

$MU_i = |\varphi_i|$ , for all  $i = r + 1, r + 2, \dots, s$ .

### 5.3. Optimal Average of Maximin & Minimax (OAxn) Techniques

We will define some definitions related with the Optimal Average (OAxn) techniques and introduce an algorithm for it.

**Definition1:** let  $y_1 = \min \{ MV_i \}$ , where  $MV_i = |\varphi_i|$ , and  $\varphi_i$  is the maximum value of  $Z_i$ , for all  $i = 1, 2, \dots, r$ .

**Definition 2:** let  $y_2 = \max \{ MU_i \}$ , where  $MU_i = |\varphi_i|$ , and  $\varphi_i$  is the minimum value of  $Z_i$ , for all  $i = r + 1, r + 2, \dots, s$ .

**Definition 3:** We denote the optimal average of maximin & minimax (OAxn), as follows:

OAxn =  $(|y_1| + |y_2|)/2$ , where  $y_j$  defined by definition(j), for all  $j=1, 2$  respectively

#### 5.3.1. The Algorithm

The following algorithm is to obtain the optimal average of maximin & minimax for the multiobjective quadratic fractional programming problem and can be summarized as follows:-

**Step1:** Write the standard form of the problem, by introducing slack and artificial variables to constraints, and write starting Simplex table.

**Step2:** Calculate the  $\Delta_j$  by the following formula  $\Delta_j = Z2\Delta\xi1j - Z1\Delta\xi2j$ , then write it in the starting Simplex table.

**Step3:** Find the solution of first objective problem by using Simplex process.

**Step4:** Check the solution for feasibility in step3, if it is feasible then go to step5, otherwise use dual Simplex method to remove infeasibility.

**Step5:** Check the solution for optimality if all  $\Delta_j \geq 0$ , then go to step6, otherwise back to step3.

**Step6:** Assign a name to the optimum value of the maximum objective function  $Z_i$  say  $\phi V_i$  where  $\forall i = 1, 2, \dots, r$ , and assign a name to the optimum value of the minimum objective function  $Z_i$  say  $\phi U_i$  where  $\forall i = r + 1, r + 2, \dots, s$ .

**Step7:** Repeat process from the step 3; for  $i = 2, \dots, s$  to be include all the objective problem.

**Step8:** Select  $y_1 = \min \{ \phi V_i \}$ ,  $\forall i = 1, 2, \dots, r$ ,  $y_2 = \max \{ \phi U_i \}$ ,  $\forall i = r + 1, r + 2, \dots, s$  then calculate,  $OA_{xn} = (|y_1| + |y_2|)/2$ ,

**Step9:** Optimize the combined objective function order the same constrains

(4.2),(4.3) as:  $Max.Z = (\sum_{i=1}^r Z_i - \sum_{i=r+1}^s Z_i) / OA_{xn}$ , (5.3.1.1)

### 5.3.2. Program Notation

The following notations, which are used in computer program, are defined as follows:

$\phi V_i$  = The value of objective function which is to be maximized.

$\phi U_i$  = The value of objective function which is to be minimized.

$MV_i = |\phi V_i|$ ;  $\forall i = 1, 2, \dots, r$ .

$MU_i = |\phi U_i|$ ;  $\forall i = r + 1, r + 2, \dots, s$ .

$SM = \sum_{i=1}^r Z_i$ .

$SN = \sum_{i=r+1}^s Z_i$ .

$$Max.Z = (SM - SN) / OA_{xn}.$$

## 6. Numerical Examples

We solved several numerical examples, but only two of them are presented in this work.

**Example 1:**

$$Max.Z_1 = \frac{(2x_1 + x_2 + 1)(2x_1 + x_2 + 2)}{(3x_1 + 3x_2 + 3)}$$

$$Max.Z_2 = \frac{(6x_1 + 3x_2 + 3)(4x_1 + 2x_2 + 4)}{(2x_1 + 2x_2 + 2)}$$

$$Max.Z_3 = \frac{(8x_1 + 4x_2 + 4)(6x_1 + 3x_2 + 6)}{(5x_1 + 5x_2 + 5)}$$

$$Min.Z_4 = \frac{(10x_1 + 5x_2 + 5)(-8x_1 - 4x_2 - 8)}{(7x_1 + 7x_2 + 7)}$$

$$Min.Z_5 = \frac{(-4x_1 - 2x_2 - 2)(6x_1 + 3x_2 + 6)}{(6x_1 + 6x_2 + 6)}$$

$$Min.Z_6 = \frac{(-2x_1 - x_2 - 1)(4x_1 + 2x_2 + 4)}{(9x_1 + 9x_2 + 9)}$$

Subject to:

$$x_1 + 2x_2 \leq 4$$

$$3x_1 + x_2 \leq 6$$

$$x_1, x_2 \geq 0$$

**Solution:** After finding the value of each of individual objective functions by an algorithm of section (3.2) the results are as below in table 1:

**Table 1.** Results of example (1)

$i$	$Z_i$	$x_i$	$\phi_i$	$MV_i =  \phi V_i  \forall i = 1, 2, \dots, r$	$MU_i =  \phi U_i  \forall i = r + 1, r + 2, \dots, s$
1	3.33	(2,0)	3.33	3.33	
2	30	(2,0)	30	30	
3	24	(2,0)	24	24	
4	-28.571	(2,0)	-28.571		28.571
5	-10	(2,0)	-10		10
6	-2.22	(2,0)	-2.22		2.22

(5.3.1.1) in section (5.3.1) is utilized in solving example (1) to find MOQFPP by using optimal average of maximin & minimax techniques:

$$OA_{xn} = (|y_1| + |y_2|)/2 = \frac{|3.33| + |-2.22|}{2} = 2.775$$

$$\text{And } SM = \sum_{i=1}^r Z_i = \frac{(16x_1 + 8x_2 + 8)(43x_1 + 21.5x_2 + 43)}{(30x_1 + 30x_2 + 30)} \text{ and}$$

$$SN = \sum_{i=r+1}^s Z_i = \frac{(-80x_1 - 40x_2 - 40)(77.1x_1 + 38.55x_2 + 77.1)}{(378x_1 + 378x_2 + 378)}$$

Then  $Max.Z = (SM - SN) / OA_{xn}$ ,

$$Max.Z = \left( \frac{(16x_1 + 8x_2 + 8)(43x_1 + 21.5x_2 + 43)}{(30x_1 + 30x_2 + 30)} - \frac{(-80x_1 - 40x_2 - 40)(77.1x_1 + 38.55x_2 + 77.1)}{(378x_1 + 378x_2 + 378)} \right) / 2.775$$

Then write the objective functions as following.

$$Max.Z = \frac{(24x_1+12x_2+12)(73.666x_1+36.833x_2+73.666)}{(125x_1+125x_2+125)} \quad (A1)$$

Subject to given constraint

$$\begin{aligned} x_1 + 2x_2 &\leq 4 \\ 3x_1 + x_2 &\leq 6 \\ x_1, x_2 &\geq 0 \end{aligned}$$

Now (A1) is solved by algorithm (3.2), then the optimal solution is found as follow

$$Max.Z = 35.359, x_1 = 2, x_2 = 0.$$

When (5.1.2) in section (5.1) is used to solve example (1), to find MOQFP by Chandra Sen. technique:

We have  $Max.Z = \sum_{i=1}^r \frac{Z_i}{|\varphi_i|} - \sum_{i=r+1}^s \frac{Z_i}{|\varphi_i|}$ , the combined objective quadratic fractional function is

$$Max.Z = \frac{(4x_1+2x_2+2)(6x_1+3x_2+6)}{(5x_1+5x_2+5)} - \frac{(-6x_1-3x_2-3)(2x_1+x_2+2)}{(10x_1+10x_2+10)}$$

Then

$$Max.Z = \frac{(10x_1+5x_2+5)(6x_1+3x_2+6)}{(10x_1+10x_2+10)} \quad (B1)$$

Therefore. After solving (B1) by given a subject with the same constraints as before, we find

$$Max.Z = 15, x_1 = 2, x_2 = 0.$$

When (5.2.1) and (5.2.2) in section (5.2) are used to solve example (1) to find MOQFP by using mean and median respectively:

**First** when we solve the example by using mean modified approach (5.2.1)

Then the combined objective quadratic fractional function is

$$Max.Z = \frac{(6x_1+3x_2+3)(2x_1+x_2+2)}{(x_1+x_2+1)} - \frac{(-6x_1-3x_2-3)(2x_1+x_2+2)}{(10x_1+10x_2+10)}$$

$$\text{Then } Max.Z = \frac{(8x_1+4x_2+4)(16x_1+8x_2+16)}{(10x_1+10x_2+10)}$$

Therefore. After solving  $Max.Z$  by given a subject with the same constraints as before, we find

$$Max.Z = 32, x_1 = 2, x_2 = 0.$$

**Second** when we solve the example by using median modified approach (5.2.2)

Then the combined objective quadratic fractional function is

$$Max.Z = \frac{(0.46x_1+0.23x_2+0.23)(2x_1+x_2+2)}{(x_1+x_2+1)} - \frac{(-50x_1-25x_2-25)(32.56x_1+16.28x_2+32.56)}{(1000x_1+1000x_2+1000)}$$

$$\text{Then } Max.Z = \frac{(60x_1+30x_2+30)(42.46x_1+21.23x_2+42.46)}{(1000x_1+1000x_2+1000)}$$

Therefore. After solving  $Max.Z$  by given a subject with the same constraints as before, we find

$Max.Z = 6.369, x_1 = 2, x_2 = 0$  when using median modified approach

### Example 2:

$$Max.Z_1 = \frac{(3x_1+3x_2+2)(2x_1+3x_2+3)}{(5x_1+5x_2+5)}$$

$$Max.Z_2 = \frac{(6x_1+6x_2+4)(6x_1+9x_2+9)}{(6x_1+6x_2+6)}$$

$$Max.Z_3 = \frac{(12x_1+12x_2+8)(4x_1+6x_2+6)}{(8x_1+8x_2+8)}$$

$$Max.Z_4 = \frac{(9x_1+9x_2+6)(8x_1+12x_2+12)}{(7x_1+7x_2+7)}$$

$$Min.Z_5 = \frac{(15x_1+15x_2+10)(-12x_1-18x_2-18)}{(3x_1+3x_2+3)}$$

$$Min.Z_6 = \frac{(-21x_1-21x_2-28)(10x_1+15x_2+15)}{(4x_1+4x_2+4)}$$

$$Min.Z_7 = \frac{(-18x_1-18x_2-12)(14x_1+21x_2+21)}{(6x_1+6x_2+6)}$$

Subject to:

$$\begin{aligned}x_1 + 4x_2 &\leq 4 \\ 2x_1 + x_2 &\leq 2 \\ x_1, x_2 &\geq 0\end{aligned}$$

**Solution:** After finding the value of each of individual objective functions by an algorithm of section (3.2) the results as below in table 2:

**Table 2.** Results of example (2)

$i$	$Z_i$	$x_i$	$\varphi_i$	$MV_i =  \varphi V_i  \forall i = 1, 2, \dots, r.$	$MU_i =  \varphi U_i  \forall i = r + 1, r + 2, \dots, s.$
1	3.466	$(\frac{4}{7}, \frac{6}{7})$	3.466	3.466	
2	17.377	$(\frac{4}{7}, \frac{6}{7})$	17.377	17.377	
3	17.378	$(\frac{4}{7}, \frac{6}{7})$	17.378	17.378	
4	29.79	$(\frac{4}{7}, \frac{6}{7})$	29.79	29.79	
5	-173.792	$(\frac{4}{7}, \frac{6}{7})$	-173.792		173.792
6	-200.444	$(\frac{4}{7}, \frac{6}{7})$	-200.444		200.444
7	-121	$(\frac{4}{7}, \frac{6}{7})$	-121		121

(5.3.1.1) in section (5.3.1) is used to solve the example (2) to find MOQFPP by using optimal average of maximin & minimax techniques:

$$Max. Z = \frac{(24x_1 + 24x_2 + 16)(34.25x_1 + 51.375x_2 + 51.375)}{(35x_1 + 35x_2 + 35)} - \frac{(-15x_1 - 15x_2 - 10)(10.3x_1 + 15.45x_2 + 15.45)}{(x_1 + x_2 + 1)} \bigg/ 62.557$$

Then write the objective functions as following.

$$Max. Z = \frac{(36x_1 + 36x_2 + 24)(39.5x_1 + 39.5x_2 + 59.25)}{(500x_1 + 500x_2 + 500)} \quad (B1)$$

Subject to

$$\begin{aligned}x_1 + 4x_2 &\leq 4 \\ 2x_1 + x_2 &\leq 2 \\ x_1, x_2 &\geq 0\end{aligned}$$

Now solving (B1) by algorithm (3.2) we get the optimal solution as follows.

$$Max. Z = 8.237, x_1 = \frac{4}{7}, x_2 = \frac{6}{7}.$$

When (5.1.2) in section (5.1) is used to solve example (2), to find MOQFPP by using Chandra Sen. Technique:

$$Max. Z = \frac{(12x_1 + 12x_2 + 8)(11.5x_1 + 17.25x_2 + 17.25)}{(100x_1 + 100x_2 + 100)} - \frac{(-27x_1 - 27x_2 - 18)(17.554x_1 + 26.331x_2 + 26.331)}{(500x_1 + 500x_2 + 500)}$$

$$Then we have. Max. Z = \frac{(45x_1 + 45x_2 + 30)(25.866x_1 + 38.799x_2 + 38.797)}{(500x_1 + 500x_2 + 500)}$$

Therefore. After solving  $Max. Z$  by given a subject with the same constraints as before, we find

$$Max. Z = 6.742, x_1 = \frac{4}{7}, x_2 = \frac{6}{7}.$$

When (5.2.1) and (5.2.2) in section (5.2) are used to solve example (2) to find MOQFPP by using mean and median respectively:

**First** when we solve the example by using mean modified approach (5.2.1)

Then the combined objective quadratic fractional function is

$$Max. Z = \frac{(12x_1 + 12x_2 + 8)(11.5x_1 + 17.25x_2 + 17.25)}{(100x_1 + 100x_2 + 100)} - \frac{(-12x_1 - 12x_2 - 8)(12.874x_1 + 19.311x_2 + 19.311)}{(165.295x_1 + 165.295x_2 + 165.295)}$$

$$Then Max. Z = \frac{(24x_1 + 24x_2 + 16)(19.25x_1 + 28.875x_2 + 28.875)}{(200x_1 + 200x_2 + 200)}$$

Therefore. After solving  $Max. Z$  by given a subject with the same constraints as before, we find

$$Max. Z = 6.6905, x_1 = \frac{4}{7}, x_2 = \frac{6}{7}.$$

**Second** when we solve the example by using median modified approach (5.2.2)

Then the combined objective quadratic fractional function is

$$Max.Z = \frac{(9x_1 + 9x_2 + 6)(6x_1 + 9x_2 + 9)}{(40x_1 + 40x_2 + 40)} - \frac{(-12x_1 - 12x_2 - 8)(12.874x_1 + 19.311x_2 + 19.311)}{(173.792x_1 + 173.792x_2 + 173.792)}$$

$$\text{Then } Max.Z = \frac{(24x_1 + 24x_2 + 16)(11.75x_1 + 17.625x_2 + 17.625)}{(100x_1 + 100x_2 + 100)}$$

Therefore. After solving  $Max.Z$  by given a subject with the same constraints as before, we find

$$Max.Z = 8.1677, x_1 = \frac{4}{7}, x_2 = \frac{6}{7}$$

## 7. Comparison of the Numerical Results

Now, we are going to compare the numerical results which are obtained of the examples above. The comparison is presented in table 3 below

The table 3 shows the results which solved by optimal Average approach was better than the Results which solved by other approaches.

**Table 3.** Comparison between results of the numerical techniques

Examples	Example 1	Example 2
Chandra Sen. Technique	15	6.742
Mean Technique	32	6.6905
Median Technique	6.369	8.1677
Optimal Average of Maximin & Minimax Techniques	35.359	8.237

## 8. Conclusions

In This work, we have defined and discussed a number of techniques, the comparisons of these methods are based on the value of the objective function. After solving the numerical examples, we found that  $max.z$  which obtained by our technique is better than other techniques (Chandra Sen., Mean & Median).

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