

Duality in Solving Multi-Objective Optimization (MOO) Problems

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Abstract

Multi-Objective Optimization (MOO) techniques often achieve the combination of both maximization and minimization objectives. The study suggests scalarizing the multi-objective functions simpler using duality. An example of four objective functions has been solved using duality with satisfactory results.

Keywords

Duality, Multi-Objective Optimization (MOO), Scalarizing Techniques

1. Introduction

Multi-Objective Optimization helps in making decisions in presence of usually conflicting objectives. Scalarizing techniques have been popularly used for solving multi-objective optimization problems. Several new scalarizing techniques [1]-[11] have been proposed during recent years. These scalarizing techniques are not efficient [12] [13] in optimizing the multiple objectives simultaneously. An improved scalarizing technique is proposed for solving MOO problems. Duality can be used to formulate the multi-objective function easier. The present study explains the utility of duality in solving multi-objective optimization problem with a suitable example.

2. Sen's Multi-Objective Optimization Technique

2.1. Primal Multi-Objective Function

The mathematical form of Sen's MOO technique [12] [13] is described as:

$$\text{Optimize } Z = [\text{Max. } Z_1, \text{Max. } Z_2, \dots, \text{Max. } Z_r, \text{Min. } Z_{r+1}, \dots, \text{Min. } Z_s]$$

Subject to:

$$AX = b \text{ and } X \geq 0$$

The individual optima are obtained by optimizing each objective separately as:

$$Z_{\text{optima}} = [\theta_1, \theta_2, \dots, \theta_s]$$

The Primal Multi-Objective Function is formulated as:

$$\text{Maximize } Z = \frac{\sum_{j=1}^r Z_j}{|\theta_j|} - \frac{\sum_{j=r+1}^s Z_j}{|\theta_{r+1}|}$$

Subject to:

$$\begin{aligned} AX &= b \text{ and } X \geq 0 \\ \theta_j &\neq 0 \text{ for } j = 1, 2, \dots, s. \end{aligned}$$

where, θ_j is the optimal value of j th objective function.

2.2. Dual Multi-Objective Function

All the objective functions are converted into either maximizing or minimizing form as described below:

Maximize Z_j or Minimize Z_j

Subject to:

$$AX = b \text{ and } X \geq 0$$

The minimization objective function can be converted into maximization objective function by multiplying -1 . Similarly the maximization objective can be converted into minimization objective function by multiplying -1 . The Multi-Objective Function is formulated as:

$$\text{Maximize } Z = \frac{\sum_{j=1}^s Z_j}{|\theta_j|}$$

or

$$\text{Minimize } Z = \frac{\sum_{j=1}^s Z_j}{|\theta_j|}$$

Subject to:

$$\begin{aligned} AX &= b \text{ and } X \geq 0 \\ \theta_j &\neq 0 \text{ for } j = 1, 2, \dots, s. \end{aligned}$$

where, θ_j is the optimal value of j th objective function.

3. Algorithm of Proposed Technique

Step I: Convert all the objective functions either maximization or minimization mode.

Step II: Formulate multi-objective function as explained in 2.2

Step III: Optimize the multi-objective function under the same constraints.

4. Multi-Objective Optimization Problem

The following example has been solved with duality technique.

Example

$$\text{Max. } Z_1 = 12500X_1 + 25100X_2 + 16700X_3 + 23300X_4 + 20200X_5$$

$$\text{Max. } Z_2 = 21X_1 + 15X_2 + 13X_3 + 17X_4 + 11X_5$$

$$\text{Min. } Z_3 = 370X_1 + 280X_2 + 350X_3 + 270X_4 + 240X_5$$

$$\text{Min. } Z_4 = 1930X_1 + 1790X_2 + 1520X_3 + 1690X_4 + 1720X_5$$

Subject to:

$$X_1 + X_2 + X_3 + X_4 + X_5 = 4.5$$

$$2X_1 \geq 1.0$$

$$3X_4 \geq 1.5$$

The above problem can be converted with all the four objective functions either maximization or minimization mode as detailed below:

$$\text{Max. } Z_1 = 12500X_1 + 25100X_2 + 16700X_3 + 23300X_4 + 20200X_5$$

$$\text{Max. } Z_2 = 21X_1 + 15X_2 + 13X_3 + 17X_4 + 11X_5$$

$$\text{Max. } Z_3 = -370X_1 - 280X_2 - 350X_3 - 270X_4 - 240X_5$$

$$\text{Max. } Z_4 = -1930X_1 - 1790X_2 - 1520X_3 - 1690X_4 - 1720X_5$$

or

$$\text{Min. } Z_1 = -12500X_1 - 25100X_2 - 16700X_3 - 23300X_4 - 20200X_5$$

$$\text{Min. } Z_2 = -21X_1 - 15X_2 - 13X_3 - 17X_4 - 11X_5$$

$$\text{Min. } Z_3 = 370X_1 + 280X_2 + 350X_3 + 270X_4 + 240X_5$$

$$\text{Min. } Z_4 = 1930X_1 + 1790X_2 + 1520X_3 + 1690X_4 + 1720X_5$$

The problem was solved with multi-objective function of both maximization and minimization mode. It is very clear from **Table 1** that all the four individual optimizations are all different and do not achieve all the objectives simultaneously.

This necessitates the need of multi-objective optimization. Both the solutions of multi-objective optimization are exactly the same and achieving all the four objectives simultaneously. Hence the multi-objective optimization problems can be solved by formulating multi-objective function after converting all the objective functions in either maximizing or minimizing mode.

Table 1. Individual and multi-objective optimization.

Objective Function	Individual Optimization				Multi-Objective Optimization	
	Max. Z_1	Max. Z_2	Min. Z_3	Min. Z_4	Maximization Mode	Minimization Mode
X_i	0.5, 3.5, 0, 0.5, 0	4, 0, 0, 0.5, 0	0.5, 0, 0, 0.5, 3.5	0.5, 0, 3.5, 0.5, 0	0.5, 0, 0, 4, 0	0.5, 0, 0, 4, 0
Z_1	105750	61650	88600	76350	99450	99450
Z_2	71.5	92.5	57.5	64.5	78.5	78.5
Z_3	1300	1615	1160	1545	1265	1265
Z_4	8075	8565	7830	7130	7725	7725

5. Conclusion

One of the important advantages of the duality theory is presented in the paper for solving MOO problems. It is established that duality makes easier the formulation of multi-objective function. However, it is needed only when optimization is done for a set of both maximization and minimization objective functions.

Conflicts of Interest

The author declares no conflicts of interest regarding the publication of this paper.

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