

Solving Definite Quadratic Bi-Objective Programming Problems by KKT Conditions

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Abstract: A bi-objective programming has been proposed for dealing with decision process involving two decision makers. In this paper, a bi-objective programming problem in which both objective functions are definite quadratic is considered. The feasible region is assumed to be a convex polyhedron. Solution methods namely; using KKT Conditions is developed. Illustrative examples for the method are presented and theorems and facts to support the method are also discussed. The solution of the examples are obtained using a LINGO (15.0) mathematical software.

Keywords: Bi-Objective Programming, Definite Quadratic Programming, Quadratic Programming, KKT Conditions

1. Introduction

1.1. Background of the Study

A general optimization problem is to select n decision variables x_1, x_2, \dots, x_n from a feasible region in such a way as to optimize (minimize or maximize) the objective function $f(x_1, x_2, \dots, x_n)$ of the decision variable. The problem is called a non-linear programming problem (NLP) if the objective function is non-linear and/or the feasible region is determined by non-linear constraints.

Interest in non-linear programming has grown simultaneously with the growth of linear programming. Kuhn and Tucker developed a necessary and sufficient condition for the existence of an optimal solution to a non-linear programming problem, which is a basis for further development in the field.

A simple subclass of non-linear programming problem is a one in which the objective function is non-linear but the constraints are all linear. This gives rise to a variety of problems depending up on the nature of the objective function.

When the objective function is given by $Z = d + CX + \frac{1}{2}X^T QX$, the problem is called a quadratic programming problem.

Given linear constraints, optimal solution of a general non-linear programming problem may not always exist at an

extreme point. In fact, a study of the nature of the objective function is necessary to predict this.

A numerical function f defined on a convex set $S \subset R^n$ is said to be convex if for each $x_1, x \in S$ and $\lambda \in [0,1]$ such that $f(\lambda x + (1 - \lambda)x_1) \leq \lambda f(x) + (1 - \lambda)f(x_1)$.

A numerical function f defined on a set $S \subset R^n$ is concave on S if and only if the function $-f$ is convex in on S .

A numerical function f is said to have a local or relative maximum at $x_0 \in S$ if $\exists \delta > 0$ such that f is defined for all points of $\|x - x_0\| < \delta$ and $f(x) \leq f(x_0)$ for all points. f is said to have a global maximum at $x_0 \in S$ if for all $x \in S$, $f(x) \leq f(x_0)$.

Algorithms for solving a general mathematical programming problem approach systematically to a local optimal solution. Under appropriate assumptions, a local optima can be shown to be a global optima. For instance, when the constraint set is a convex polyhedron and the objective function is convex (concave), the local minimum (maximum) is also a global minimum (maximum) [5].

Single objective decision making method reflect an earlier and simpler era. The world has become more complex as one enters the information age. One can find that almost every important real world problem involves more than one objective. Since the goals or objectives might be conflicting with each other, no single optimal solution can be found and the optimization problem becomes finding the best compromise solutions [5].

The general Multi-Objective Programming problem [9] is defined as

$$\begin{aligned} \max Z_1 &= f_1(X) \\ \max Z_2 &= f_2(X) \\ &\cdot \\ &\cdot \\ &\cdot \\ \max Z_k &= f_k(X) \\ \text{subject to } X &\in S \end{aligned} \quad (1)$$

A bi-objective problem formulated as follows:

$$\begin{aligned} \max_X f_1(X) \\ \max_X f_2(X) \\ \text{subject to } X \in S \end{aligned} \quad (2)$$

where f_1 and f_2 are the first and the second objective functions respectively and S is constraint set.

Quadratic Programming Problem

A quadratic program (QP) is an optimization problem wherein one either minimizes or maximizes a quadratic objective function of a finite number of decision variable subject to a finite number of linear inequality and/or equality constraints. A quadratic function of a finite number of variables $X = (x_1, x_2, \dots, x_n)^T$ is any function of the form

$$f(X) = \alpha + \sum_{j=1}^n c_j x_j + \frac{1}{2} \sum_{k=1}^n \sum_{j=1}^n q_{kj} x_k x_j \quad (3)$$

Using matrix notation, this expression simplifies to

$$f(X) = \alpha + C^T X + \frac{1}{2} X^T Q X \quad (4)$$

where $C = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix}$ and $Q = \begin{pmatrix} q_{11} & q_{12} & \cdots & q_{1n} \\ q_{21} & q_{22} & \cdots & q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ q_{n1} & q_{n2} & \cdots & q_{nn} \end{pmatrix}$

without loss of generality, assume that Q is symmetric matrix since

$$\begin{aligned} X^T Q X &= (X^T Q X)^T = X^T Q^T X = \frac{1}{2} (X^T Q X + X^T Q^T X) = \\ &X^T \left(\frac{Q+Q^T}{2} \right) X, \end{aligned}$$

and so it is free to replace the matrix Q by the symmetric matrix $\frac{Q+Q^T}{2}$. Henceforth, the matrix Q is symmetric.

The general quadratic programming problem can be written as

$$\begin{aligned} (QP): \text{maximize } f(X) &= C^T X + \frac{1}{2} X^T Q X \\ \text{subject to } AX &\begin{pmatrix} \leq \\ \geq \end{pmatrix} B \text{ and } X \geq 0 \end{aligned} \quad (5)$$

If constant term exists, it is dropped from the model since

it plies no role in the optimization step. The decision variables are denoted by the n -dimensional column vector X , and the constraints are defined by an $(m \times n)$ matrix A and an m -dimensional column vector B of right hand side.

1.2. Statement of the Problem

Different researchers used different methods for solving BOQP problems, but the method that they used is so lengthy and in addition to this, most of them use linearization technique, this creates an approximation error. So, the researcher was trying to answered the question that; can we solve definite quadratic bi-objective programming problem by KKT condition easily.

1.3. Significance of the Study

This study will give a direction towards definite quadratic bi-objective programming problems to anybody or any organization to solve their real life problems (or whatever any problem) which are modeled as definite quadratic bi-objective programming with constraint region determined by linear constraints.

1.4. Objective of the Study

The general objective of this study is to solve the definite BOQP problem with constraint region determined by linear constraints by KKT conditions.

The study is intended to explore the following specific objectives

- a) To discuss definite quadratic bi-objective programming problem;
- b) To discuss KKT conditions, and to solve the definite BOQP.

2. Research Methodology

This research work involve in collecting the information about solving definite quadratic bi-objective programming problem in which KKT conditions focused, from optimization books, optimization journals, and searching other materials and references from the internet. The collected material and the techniques or methods used by authors in relation to KKT conditions are examined.

- a) Important theorems and facts to support the methods are discussed.
- b) The definite quadratic bi-objective programming problem is changed into single programming problem by using KKT conditions. Then after changing in to single programming problem LINGO (15.0) was applying to solve the problem.

3. Definite Quadratic Bi-Objective Programming Problem

A quadratic program

$$\text{maximize } f(X) = CX + \frac{1}{2}X^T QX$$

Subject to $AX \leq B$ and $X \geq 0$ where Q is symmetric is said to be definite quadratic if $X^T QX$ satisfies one of the following concepts

- i. positive definite if $X^T QX > 0$ for all $X \neq 0$,
- ii. positive semi definite if $X^T QX \geq 0$ for all X ,
- iii. negative definite if $X^T QX < 0$ for all $X \neq 0$,
- iv. negative semi definite if $X^T QX \leq 0$ for all X .

However, it is difficult to determine the value of X easily, so we use the following concepts:

Let $Q = \begin{pmatrix} q_{11} & q_{12} & \dots & q_{1n} \\ q_{21} & q_{22} & \dots & q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ q_{n1} & q_{n2} & \dots & q_{nn} \end{pmatrix}$, the n determinants

$$D_1 = |q_{11}|, D_2 = \begin{vmatrix} q_{11} & q_{12} \\ q_{21} & q_{22} \end{vmatrix}, \dots, D_n =$$

$\begin{vmatrix} q_{11} & q_{12} & \dots & q_{1n} \\ q_{21} & q_{22} & \dots & q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ q_{n1} & q_{n2} & \dots & q_{nn} \end{vmatrix}$ are called the leading principal minors of a matrix Q .

Instead of the above four concepts, we can use the following criteria to determine the definiteness of the quadratic program. i.e.

- i. positive definite if and only if $D_1 > 0, D_2 > 0, \dots, D_n > 0$ that is all principal minors are strictly greater than zero,
- ii. negative definite if and only if $D_1 < 0, D_2 > 0, D_3 < 0, \dots$ that is all principal minors alternate in sign starting with negative one (or the value of the K^{th} leading principal minor has the sign of $(-1)^K$),
- iii. positive semi definite if and only if all principal minors are greater than or equal to zero,
- iv. negative semi definite if and only if all principal minors of odd degree are less than or equal to zero, and all principal minors of even degree are greater than or equal to zero.

A bi-objective programming problem, which both the objective functions are definite quadratic is called definite quadratic bi-objective program. Consider the following definite quadratic bi-objective programming problem.

$$(BOQP): \max_{X,Y} f_1(X, Y) = a^T X + b^T Y + (X^T, Y^T)P \begin{pmatrix} X \\ Y \end{pmatrix} \quad (6)$$

$$\max_Y f_2(X, Y) = p^T X + q^T Y + (X^T, Y^T)Q \begin{pmatrix} X \\ Y \end{pmatrix}$$

Subject to $AX + BY \leq C$ and $X, Y \geq 0$

Where $X \in R^{n_1}, Y \in R^{n_2}; a, p \in R^{n_1}; b, q \in R^{n_2}; C \in R^m; P$ and Q are $(n_1 + n_2) \times (n_1 + n_2)$ real symmetric definite matrices and with $Q = \begin{pmatrix} Q_1 & Q_2^T \\ Q_2 & Q_3 \end{pmatrix}$ A and B are $m \times n_1$ and $m \times n_2$ matrices respectively. The second objective function becomes

$$\max_{X,Y} f_2(X, Y) = p^T X + q^T Y + (X^T, Y^T) \begin{pmatrix} Q_1 & Q_2^T \\ Q_2 & Q_3 \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix}$$

$$\Rightarrow \max_{X,Y} f_2(X, Y) = p^T X + q^T Y + X^T Q_1 X$$

$$+ Y^T Q_2 X + X^T Q_2^T Y + Y^T Q_3 Y$$

$$= p^T X + q^T Y + X^T Q_1 X + 2X^T Q_2^T Y$$

$$+ Y^T Q_3 Y, \text{ since } Y^T Q_2 X = X^T Q_2^T Y.$$

$$\Rightarrow \max_{X,Y} f_2(X, Y) = p^T X + (q + 2Q_2 X)^T Y$$

$$+ X^T Q_1 X + Y^T Q_3 Y.$$

Therefore, f_2 can be written as $Y^T Q_3 Y + (q + 2Q_2 X)^T Y + X^T Q_1 X + p^T X$.

If a constant term exists, it is dropped from the model since it plays no role in the optimization step. So X is fixed prior to the maximization of f_2 , because of the fact that the second objective function the decision variable Y only.

The second objective problem is equivalent to

$$P(X): \max_{X,Y} f_2(X, Y) = Y^T Q_3 Y + (q + 2Q_2 X)^T Y \quad (7)$$

$$\text{Subject to } BY \leq C - AX \text{ and } Y \geq 0$$

Therefore, problem BOQP becomes equivalent to

$$(BOQP_1): \max_{X,Y} f_1(X, Y) = a^T X + b^T Y + (X^T, Y^T)P \begin{pmatrix} X \\ Y \end{pmatrix} \quad (8)$$

$$\max_{X,Y} f_2(X, Y) = Y^T Q_3 Y + (q + 2Q_2 X)^T Y$$

Subject to $BY \leq C - AX$ and $Y \geq 0$.

In this study, the researcher assumed that bi-objective programming problem is considered; the objective functions are convex with positive semi definite matrices P and Q on minimization type of problems and the second objective function controls the decision variable Y only. On the other hand, the researcher was assumed to be concave objective function with negative semi definite matrices P and Q on maximization type of problems and the second objective function controls the decision variable Y only with convex polyhedron region determined by linear constraints.

The aim of doing this research was solving the definite quadratic bi-objective programming problems by using KKT conditions and lingo (15.0) mathematical software. So, it is obligate that knowing the concept of Karush-Khun-Tucker Conditions.

Karush-Kuhn-Tucker Conditions: The Lagrange multipliers can be used to develop optimality criteria for the equality constraint. Khun and Tucker extend this theory to include equality and inequality constraint problems i.e.

$$\max f(X) \quad (9)$$

$$s. t \ g_j(X) \leq 0 \text{ for } j = 1, 2, \dots, J$$

$$h_k(X) = 0 \text{ for } k = 1, 2, \dots, K.$$

Kuhn and Tucker developed the necessary and sufficient conditions for the NLP problem by assuming f, g_j and h_k are differentiable. The general NLP problem is given by

$$\max f(X)$$

$$s. t \ g_j(X) \leq 0 \text{ for } j = 1, 2, \dots, J$$

$$h_k(X) = 0 \text{ for } k = 1, 2, \dots, K,$$

where f is nonlinear or/and g_j and h_k are nonlinear.

To solve this problem, we write the Lagrange function:

$$l(X, U, V) = f(X) - \sum_{j=1}^J U_j g_j(X) - \sum_{k=1}^K V_k h_k(X) \quad (10)$$

To find the critical point of l , $\frac{\partial l}{\partial X_i} = 0$ for $i = 1, 2, \dots, n$.

The KKT conditions are stated as: Find vectors $X_{(NX1)}$, $U_{(1XJ)}$ and $V_{(1XK)}$ that satisfy

$$\nabla f(X) - \sum_{j=1}^J U_j \nabla g_j(X) - \sum_{k=1}^K V_k \nabla h_k(X) = 0 \quad (11)$$

$$g_j(X) \leq 0 \text{ for } j = 1, 2, \dots, J$$

$$h_k(X) = 0 \text{ for } k = 1, 2, \dots, K$$

$$U_j g_j(X) = 0 \text{ for } j = 1, 2, \dots, J$$

$$U_j \geq 0 \text{ for } j = 1, 2, \dots, J$$

Observe that for the minimization problem, all one needs to do is to change the minus sign in the Lagrange to plus, because finding a minimum for f is the same as finding a maximum for $-f$.

Note that the above KKT condition is a necessary condition (in general) but is also sufficient for convex problems.

Theorem 3.1: (Kuhn-Tucker Necessary Theorem)

Consider the NLP problem and let f , g_j and h_k be differentiable functions and X^* be a feasible solution to the NLP problem. Let $I = \{j: g_j(X) = 0\}$.

Further, $\nabla g_j(X^*)$ for $j \in I$ and $\nabla h_k(X^*)$ for $k = 1, 2, \dots, K$ are linearly independent. If X^* is an optimal solution to the NLP, then there exists (U^*, V^*) such that (X^*, U^*, V^*) solves the KT problem.

Proof: The proof of theorem 3.1 found on [7].

The conditions $\nabla g_j(X^*)$ for $j \in I$ and $\nabla h_k(X^*)$ for $k = 1, 2, \dots, K$ are linearly independent at the optimum is known as the constraint qualification and this constraint qualification holds in the following cases:

- i. when all the constraints are linear
- ii. when all the inequality constraints are concave functions and the equality constraints are linear and there exists at least one feasible X that is strictly inside the feasible region of the inequality constraints. In other words, there exists an X such that

$$g_j(X) < 0 \text{ for } j = 1, 2, \dots, J \text{ and } h_k(X) = 0$$

$$\text{for } k = 1, 2, \dots, K.$$

Note that when the constraint qualification is not met at the optimum, there may not exist a solution to the Kuhn-Tucker problem. Therefore, do not apply the Kuhn-Tucker optimality conditions when the constraint qualification is not met.

We know that a quadratic programming problem is a special type of the NLP problem that one either minimize or

maximize a quadratic objective function subject to linear constraints. Therefore, we can apply the Kuhn-Tucker optimality conditions because the constraint qualification is fulfilled since the constraints are linear. Depending on this reason, let us consider the following optimality conditions for the quadratic programming problem.

Consider the quadratic programming problem

$$Q_1 \min_X f(X) = CX + \frac{1}{2}X^T QX \quad (12)$$

$$s. t \ AX \leq b \text{ for } 0 \leq X$$

A pair $(X, U) \in R^n X R^m$ is said to be a karush-kuhn-tucker pair (or KKT pair) for the quadratic program Q_1 if and only if the following conditions are satisfied:

$$0 \leq X, AX \leq b \text{ (primal feasibility)}$$

$$0 \leq U, 0 \leq C + QX + A^T U \text{ (dual feasibility)}$$

$$0 = U^T (AX - b), \text{ and}$$

$$0 = X^T (C + QX + A^T U) \text{ (complementarity)}$$

Theorem 3.2: (Necessary Conditions for Optimality in Quadratic Programming)

Consider the quadratic programming problem Q_1 . If $\bar{X} \in R^n$ solves Q_1 , then there exists a vector U^* such that (\bar{X}, U^*) is a KKT pair for Q_1 .

Before going to solve the optimal solution of the definite quadratic bi-objective programming problem, we have to check whether the solutions of a given program exist, and unique or not. Because of this reason, consider the following existence and uniqueness theorem.

Theorem 3.3: (Weierstrass Theorem: Existence of a Solution)

If $f: R^n \rightarrow R$ is continuous function, and S is a non-empty, closed and bounded subset of R^n , then there exists X_1 and X_2 in S such that $f(X_1) \leq f(X) \leq f(X_2)$ for all $X \in S$.

Proof: The proof of theorem 3.3 found on [1].

Take the idea of Weierstrass Theorem on equation $P(X)$ of the above lower objective function. Let S be a feasible set. Suppose S non-empty, closed, bounded and f_2 is a convex objective function, then there exists a global optimum point for f_2 .

Theorem 3.4: (Uniqueness of a Solution for $P(X)$)

Suppose S is non-empty, closed, bounded and convex subset of R^n and suppose that f_2 is strictly convex (concave) function on S . Then there exists $X^* \in S$ such that $f_2(X^*) < f_2(X)$ (respectively, $f_2(X^*) > f_2(X)$) for all $X \in S$ where $X \neq X^*$.

To assure that KKT optimality conditions are both necessary and sufficient for obtaining the global optimum of the inner problem, consider the following theorem.

Theorem 3.5: (Necessary and Sufficient Condition for Optimality in Convex QP problem $P(X)$)

If Q is symmetric and negative semi-definite, then (\bar{X}, \bar{Y}) solves $P(X)$ if and only if there exists \bar{W} such that $(\bar{X}, \bar{Y}, \bar{W})$ is a KKT pair for $P(X)$.

Proof: The proof of Theorem 3.5 found on [4].

Consider $BOQP_1$ and apply KKT conditions on the second objective function programming problem by considering X as fixed. i.e.

The lagrangian function of the second objective function programming problem is

$$L(X, Y, W) = Y^T Q_3 Y + (q + 2Q_2 X)^T Y - W^T (AX + BY - C) \quad (13)$$

Then

$$\begin{aligned} \frac{\partial L(X, Y, W)}{\partial Y} &= 0 \\ W^T (AX + BY - C) &= 0 \\ AX + BY &\leq C \\ \Rightarrow \begin{cases} 2Q_2 X + 2Q_3 Y + q = B^T W \\ W^T (AX + BY) = W^T C \\ AX + BY \leq C \end{cases} & \quad (14) \end{aligned}$$

where W is Lagrange multiplier

By considering (14) as a constraint function for the first objective function programming problem, we get

$$\begin{aligned} \max_{X, Y} f_1(X, Y) &= a^T X + b^T Y + (X^T, Y^T) P \begin{pmatrix} X \\ Y \end{pmatrix} \\ M: \quad \begin{cases} AX + BY \leq C \\ W^T (AX + BY) = W^T C \\ 2Q_2 X + 2Q_3 Y + q = B^T W \\ X, Y, W \geq 0 \end{cases} & \quad (15) \end{aligned}$$

We have that (M) is a single level maximization problem with one quadratic objective function and the constraints are nonlinear. Now we can solve problem M by Lingo (15.0) software because of the fact that this software is appropriate for solving the optimal solution for such likes of problems.

Therefore, the following necessary and sufficient optimality for $BOQP_1$ (theorem 3.6) is clarified that the point that solves the problem (M) is an optimal solution to the quadratic programming problem $BOQP_1$

Note: An optimal solution is a feasible solution that has a maximum objective function value for maximization type of problem or a minimum objective function value for minimization type of problems.

Let $S = \{(X, Y): AX + BY \leq C; X, Y \geq 0\}$, and S assumed to be closed and bounded.

For problem $BOQP_1$, the first objective function solution space is given by

$$T = \{X \in R^{n_1}: \text{there exists } Y \text{ such that } (X, Y) \in S\}.$$

The second level solution space given by

$$S(X) = \{Y \in R^{n_2}: (X, Y) \in S\}.$$

For each $X \in T, Y(X) = \{Y \in S(X): f_2(X, Y) = \max_Y f_2(X, Y)\}$ for a given X denotes the set of optimal solutions to the second objective programming problem.

$\bar{S} = \{(X, Y): (X, Y) \in S, Y \in Y(X)\}$ denotes the set of solution points to $BOQP_1$ and is called the induced region of

problem $BOQP_1$.

A point (X^*, Y^*) is said to be optimal to problem $BOQP_1$ if

- i. $(X^*, Y^*) \in \bar{S}$
- ii. $f_1(X^*, Y^*) \geq f_1(X, Y) \forall (X, Y) \in \bar{S}$

Theorem 3.6: (Optimality Condition for $BOQP_1$)

(X^*, Y^*) is an optimal solution to the quadratic programming problem $BOQP_1$ if and only if there exists a vector $W^* \in R^m$ such that (X^*, Y^*, W^*) solves the following nonlinear programming problem.

M:

$$\max_{X, Y} f_1(X, Y) = a^T X + b^T Y + (X^T, Y^T) P \begin{pmatrix} X \\ Y \end{pmatrix}$$

$$AX + BY \leq C \quad (16)$$

$$W^T (AX + BY) = W^T C \quad (17)$$

$$2Q_2 X + 2Q_3 Y + q = B^T W \quad (18)$$

$$X, Y, W \geq 0 \quad (19)$$

Proof: Let (X^*, Y^*) be an optimal solution to $BOQP_1$. Then $(X^*, Y^*) \in \bar{S}$ and

$f_1(X^*, Y^*) \geq f_1(X, Y) \forall (X, Y) \in \bar{S}$. This implies that Y^* solves the relaxed second objective function given by

$$P_r(x): \max_Y f_2(X^*, Y) = Y^T Q_3 Y + (q + 2Q_2 X^*)^T Y$$

Subject to

$$BY \leq C - AX^* \text{ and } Y \geq 0$$

Therefore, by Kuhn Tucker necessary conditions, there exist a $W^* \geq 0$ such that

$$W^{*T} (AX^* + BY^* - C) = 0$$

$$2Q_2 X^* + 2Q_3 Y^* + q - B^T W^* = 0$$

$$AX^* + BY^* \leq C$$

which implies that (X^*, Y^*, W^*) is a feasible solution to problem M. We will show that it is an optimal solution to M. If (X^*, Y^*, W^*) is not an optimal solution to problem M, then there exists $(\bar{X}, \bar{Y}, \bar{W})$ satisfying (16)-(19) and $f_1(\bar{X}, \bar{Y}) \geq f_1(X^*, Y^*)$. As Q is a negative semi-definite problem $P_r(x)$ is a concave maximization problem. Therefore (16)-(19) are sufficient conditions for \bar{Y} to be an optimal solution. Hence $\bar{Y} \in Y(\bar{X})$ and $f_1(\bar{X}, \bar{Y}) \geq f_1(X^*, Y^*)$. This contradicts that (X^*, Y^*) is optimal to problem $BOQP_1$. Therefore, (X^*, Y^*, W^*) is optimal to problem M. Conversely, assume that there exists a $W^* \in R^m$ such that (X^*, Y^*, W^*) solves M. We will show that (X^*, Y^*) solves $BOQP_1$. As shown above $Y^* \in Y(X^*)$. For any $(X, Y) \in \bar{S}, Y \in Y(X)$ implies that Y solves $P(X)$. So, by Kuhn Tucker necessary conditions, there exists a $W \geq 0$ such that (16)-(19) hold. Thus (X, Y, W) is feasible to M and hence $f_1(X^*, Y^*) \geq f_1(X, Y)$, which proves that (X^*, Y^*) solves $BOQP_1$.

The stepwise description of the method that of solving the optimal solution of definite quadratic bi-objective programming problem using KKT conditions is as follows:

Given a definite $BOQP$;

$$\max_{X,Y} f_1(X,Y) = a^T X + b^T Y + (X^T, Y^T) P \begin{pmatrix} X \\ Y \end{pmatrix} \quad (20)$$

$$\max_Y f_2(X,Y) = p^T X + q^T Y + (X^T, Y^T) Q \begin{pmatrix} X \\ Y \end{pmatrix}$$

Subject to $AX + BY \leq C$ and $X, Y \geq 0$

Step 1: Formulate the single objective problem by applying KKT conditions on the second objective function by considering X as fixed

$$M: \begin{cases} \max_{X,Y} f_1(X,Y) = a^T X + b^T Y + (X^T, Y^T) P \begin{pmatrix} X \\ Y \end{pmatrix} \\ AX + BY \leq C \\ W^T (AX + BY) = W^T C \\ 2Q_2 X + 2Q_3 Y + q = B^T W \\ X, Y, W \geq 0 \end{cases} \quad (21)$$

where, W is Lagrange multiplier

Step 2: Solve problem M by LINGO (15.0).

Example 3.1. Consider the following definite quadratic bi-objective programming problem

$$\max_{X,Y} f_1(X,Y) = 6X + 3Y - X^2 - Y^2$$

$$\max_Y f_2(X,Y) = X + 5Y - Y^2$$

$$X + Y \leq 5$$

$$3X + 2Y \leq 9$$

$$\Rightarrow \begin{cases} 2Y + W_1 + 2W_2 + W_3 = 5 \\ W_1(X + Y) + W_2(3X + 2Y) + W_3(2X + Y) = 5W_1 + 9W_2 + 6W_3 \\ X + Y \leq 5 \\ 3X + 2Y \leq 9 \\ 2X + Y \leq 6 \end{cases} \quad (22)$$

By taking (22) as a constraint function, then we can get the following problem

$$\begin{cases} \max_{X,Y} f_1(X,Y) = 6X + 3Y - X^2 - Y^2 \\ 2Y + W_1 + 2W_2 + W_3 = 5 \\ W_1(X + Y) + W_2(3X + 2Y) + W_3(2X + Y) = 5W_1 + 9W_2 + 6W_3 \\ X + Y \leq 5 \\ 3X + 2Y \leq 9 \\ 2X + Y \leq 6 \end{cases} \quad (23)$$

If X, Y, W_1, W_2, W_3 solves (23), then X, Y is the optimal solution of the definite BOQP problem. By using LINGO (15.0) software, we can get the optimal solution of problem (23), so the solution of (23) is $(X, Y, W_1, W_2, W_3) = (2.308, 1.038, 0, 1.462, 0)$ and it really satisfies all systems of (23). Therefore, $X = 2.308$ and $Y = 1.038$ are the optimal solution of the definite quadratic bi-objective programming problem and the value of the objective functions are

$$f_2 = 6.421, \text{ and } f_1 = 10.558$$

Example 3.2. Consider the following quadratic bi-objective programming problem

$$\min_{X,Y} f_1(X,Y) = 3X^2 + 4Y^2 - 2X - 2Y$$

$$2X + Y \leq 6 \text{ where } X, Y \in R \text{ and } X, Y \geq 0$$

The lagrangian function for the second objective function is

$$L(X, Y, W_1, W_2, W_3) = X + 5Y - Y^2 - W_1(X + Y - 5) - W_2(3X + 2Y - 9) - W_3(2X + Y - 6)$$

First, apply KKT conditions on the second objective function by considering X as a fixed.

$$\frac{\partial L(Y, W_1, W_2, W_3)}{\partial Y} = 0$$

$$W_1(X + Y - 5) = 0$$

$$W_2(3X + 2Y - 9) = 0$$

$$W_3(2X + Y - 6) = 0$$

$$X + Y \leq 5$$

$$3X + 2Y \leq 9$$

$$2X + Y \leq 6$$

$X, Y, W_1, W_2, W_3 \geq 0$, where W_1, W_2, W_3 are language multipliers.

$$\min_Y f_2(X,Y) = 5X^2 + 2Y^2 - X - 2Y$$

$$2X + Y \geq 2$$

$$X + 3Y \leq 7$$

$$X, Y \geq 0$$

Applying KKT conditions on the second objective problem, then the corresponding quadratic programming problem is

$$\min_{X,Y} f_1(X,Y) = 3X^2 + 4Y^2 - 2X - 2Y$$

$$4Y - W_1 + 3W_2 = 2$$

$$W_1(-2X - Y) + W_2(X + 3Y) = 7W_2 - 2W_1$$

$$2X + Y \geq 2$$

$$X + 3Y \leq 7$$

$$X, Y, W_1, W_2 \geq 0$$

Where W_1 and W_2 are Lagrange multipliers. Then by using LINGO software we have, $(X, Y, W_1, W_2) = (0.75, 0.5, 0, 0)$, from this $X = 0.75$ and $Y = 0.5$ are the optimal solution of the definite quadratic bi-objective programming problem and the value of the objective functions are $f_2 = 1.563$ and $f_1 = 0.188$.

4. Conclusions and Recommendation

4.1. Conclusion

In this paper, the researcher used KKT conditions. Finally, using LINGO (15.0) software the problem was solved. Numerical examples were considered to show the efficiency of the proposed algorithm. The KKT conditions are simple to find the optimal solution of BOQP Problems.

4.2. Recommendation

In this paper, KKT conditions are proposed to solve definite quadratic bi-objective programming problems. These techniques can easily extend for definite quadratic bi-objective fractional programming problems.

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