A Weaker Constraint Qualification of Globally Convergent Homotopy Method for a Multiobjective Programming Problem

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ABSTRACT

In this paper, we prove that the combined homotopy interior point method for a multiobjective programming problem introduced in Ref. [1] remains valid under a weaker constrained qualification—the Mangasarian-Fromovitz constrained qualification, instead of linear independence constraint qualification. The algorithm generated by this method associated to the Karush-Kuhn-Tucker points of the multiobjective programming problem is proved to be globally convergent.

Keywords: Multiobjective Programming Problem; Homotopy Method; KKT Condition; Efficient Solution; MFCQ

1. Introduction

Let \( \mathbb{R}^n \) be the \( n \)-dimensional Euclidean space, and let \( \mathbb{R}^n_+ \) and \( \mathbb{R}^n_+ \) denote the nonnegative and positive \( \mathbb{R}^n \), respectively. For any two vectors \( y = (y_1, y_2, \ldots, y_n) \) and \( z = (z_1, z_2, \ldots, z_n) \) in \( \mathbb{R}^n \), we use the following conventions: \( y \leq z \) if \( y_i = z_i, i = 1, 2, \ldots, n \). Similarly, we can define \( y \leq z, y < z \), and \( y \leq z \).

Consider the following multiobjective programming problem (MOP)

\[
\begin{align*}
\min & \quad f(x) \\
\text{s.t.} & \quad g(x) \leq 0, h(x) = 0,
\end{align*}
\]

where \( x \in \mathbb{R}^n, f = (f_1, f_2, \ldots, f_p) \),

\[
g = (g_1, g_2, \ldots, g_m)^T, h = (h_1, h_2, \ldots, h_s)^T.
\]

We assume that all \( f_i, i = 1, \ldots, p, g_j, j \in I \) and \( h_k, k \in J \) are twice continuously differentiable functions, where \( I = \{1, 2, \ldots, m\}, J = \{1, 2, \ldots, s\} \). Let

\[
\Omega = \{x \in \mathbb{R}^n : g(x) < 0, h(x) = 0\},
\]

\[
\Omega_r = \{x \in \mathbb{R}^n : g(x) < 0\},
\]

\[
A^{++} = \{\lambda \in \mathbb{R}^n_+ : \sum_{i=1}^p \lambda_i = 1\},
\]

\[
I(x) := \{j \in I : g_j(x) = 0\}.
\]

It is well known that if \( x \) is an efficient solution of (MOP), under some constraint qualifications, such as the Kuhn and Tucker constraint qualification (see Ref. [2]) or the Abadie constraint qualification (see Ref. [3]), then the following Karush-Kuhn-Tucker (KKT) condition at \( x \) for (MOP) holds (see Refs. [4,5]):

\[
\begin{align*}
\nabla f(x)^T \lambda + \nabla g(x)^T u + \nabla h(x)^T v & = 0 \\
Ug(x) & = 0 \\
h(x) & = 0
\end{align*}
\]

where \( \lambda \in \mathbb{R}^n_+ \setminus \{0\}, u \in \mathbb{R}^m_+, v \in \mathbb{R}^s, \) and

\[
U = \text{diag}\{u_1, u_2, \ldots, u_m\}.
\]

We say that \( x \) is a KKT point of (MOP) if it satisfies the KKT condition.

Since the remarkable papers of Kellogg et al. (Ref. [6]) and Chow et al. (Ref. [7]) have been published, more and more attention has been paid to the homotopy method. As a globally convergent method, the homotopy method (or path-following method) now becomes an important tool for numerically solving nonlinear problems including nonlinear mathematical programming and complementarily problems (see Refs. [3,4]).

In 1988, Megiddo (see Ref. [8]) and Kojima et al. (see Ref. [9]) discovered that the Karmakar interior point method was a kind of path-following method for solving linear programming. Since then, the interior path-following method has been generalized to convex programming, and becomes one of the main methods for solving mathematical programming problems. Among most interior
methods, one of the main ideas is numerically tracing the center path generated by the optimal solution set of the so-called logarithmic barrier function. Usually, the strict convexity of the logarithmic barrier function or non-emptiness and boundedness of the feasible set (see Ref. [10]) are needed. In 1997, Lin, Yu and Feng (see Ref. [11]) presented a new interior point method—combined homotopy interior point method (CHIP method)—for convex nonlinear programming without such assumptions. Subsequently, Lin, Li and Yu (see Ref. [12]) generalized CHIP method to general nonlinear programming where, instead of convexity condition, they used a more general “normal cone condition”.

In 2003, Lin, Zhu and Sheng (see Ref. [13]) generalized CHIP method to convex multiobjective programming (CMOP) with only inequality constraints. Instead of (CMOP), they considered an associated non-convex nonlinear scalar optimization problem and constructed the homotopy mapping.

In Refs. [1,14], we considered a combined homotopy interior point method for the multiobjective programming (MOP) under the condition linearly independent constraint qualification (LICQ). To find a KKT point of (MOP), we construct a homotopy as follows

\[
0 = H(\omega, \omega^0, \mu) = \begin{bmatrix}
(1-\mu)\left(\nabla f(x)^T + \nabla g(x)^T u + \nabla h(x)^T v + \mu(x-x^0)\right)

h(x)

U g(x) - \mu U g(x^0)

(1-\mu)\left(1-\sum_{i=1}^{k} \lambda_i \right) e - \mu \left(\lambda^{2/5} - (\lambda^0)^{2/5}\right)
\end{bmatrix}
\]

where \(\omega^0 = (x^0, \lambda^0, u^0, v^0) \in \Omega \times \Lambda^{++} \times R_+^m \times \{0\}, e = (1, \ldots, 1)^T \in R^e, \omega = (x, \lambda, u, v) \in \bar{\Omega} \times R_+^{rm} \times R^e, \mu \in [0, 1], \lambda^{2/5} = (\lambda_1^{2/5}, \ldots, \lambda_p^{2/5}), \) and \(\lambda^{0,2/5} = (\lambda_1^{0}, \ldots, \lambda_p^{0})^{2/5}.\)

Let \(H^+_{\omega^0}(0) = \{\omega, \mu \in \Omega \times R_+^{rm} \times R^e \times \{0, 1\} : H(\omega, \omega^0, \mu) = 0\}.\)

Let \(A \subset R^n\) be a nonempty closed set and \(x \in A.\)

We recall that the Fréchet normal cone of \(A\) at \(x\) is defined as

\[
N_A(x) = \left\{ x^* \in R^n : \limsup_{x' \to x} \frac{\langle x^*, x' - x \rangle}{\|x' - x\|} \leq 0 \right\}
\]

We used the following basic assumptions which are commonly used in that literature:

(A1) \(\Omega\) is nonempty (Slater condition) and bounded;

(A2) (LICQ) \(\forall x \in \bar{\Omega},\) the matrix

\[
\left(\nabla h(x)^T, \nabla g_j(x)^T : j \in I(x)\right)
\]

is a matrix of full column rank;

(A3) Normal condition:

\[
\forall x \in \bar{\Omega}, \{x + \hat{N}_\Omega(x)\} \cap \bar{\Omega} = \{x\}.
\]

It is well known that if condition (A2) holds, then

\[
T \times \{0\} \subset \bar{\Omega} \times R_+^{rm} \times \{0\}
\]

of \(\Gamma_{\omega^0}\) is nonempty, and the \(x\)-component of every point in \(T\) is a KKT point of (MOP).

Recently, many researchers extended and improved the results in Ref. [1] to convex multiobjective programming problem, see Ref. [14-17]. The purpose of this paper is to show that Theorem 1.1 remains true under the condition MFCQ instead of LICQ. The paper is organized as following. In Section 2, we prove the existence and convergence of a smooth homotopy path from almost any interior initial point \(\omega^0\) to a solution of
the KKT system of (MOP) under the condition MFCQ.

2. Main Results

We need the following elementary condition.

(2.1') (MFCQ) For every \( x \in \Omega \), the following conditions hold:

- \( \nabla h_j(x), j \in J \), are linear independent;

- there exists a \( p \in R^n \) such that

\[
\nabla g_i(x)p > 0, i \in I(x) \quad \text{and} \quad \nabla h_j(x)p = 0, j \in J.
\]

Clearly, condition (A2) implies (A2'). It is also known that if (A2') holds, then (3) remains valid.

By using an analogue argument as in Ref. [1], we can prove the following two theorems.

Theorem 2.1 Suppose that \( \Omega \neq \emptyset \) and conditions (A1), (A2'), and \( (A3) \) hold. Then for almost all initial points \( o^0 \in \Omega \times \Lambda^++ \times R_+^{m} \times \{0\}, 0 \) is a regular value of \( H_{o^0} \)

and \( H_{o^0}^{\ast}(0) \) consists of some smooth curves. Among them, a smooth curve, say \( \Gamma_{o^0} \) starts from \( (o^0,1) \).

Theorem 2.2 Suppose that \( \Omega \neq \emptyset \) and conditions (A1), (A2'), and \( (A3) \) hold. For a given \( o^0 \in \Omega \times \Lambda^++ \times R_+^{m} \times R^n \), if \( 0 \) is a regular value of \( H_{o^0} \), then the projection of the smooth curve \( \Gamma_{o^0} \) on the \( \lambda^0 \) component is bounded.

We next prove that \( \Gamma_{o^0} \) is a bounded curve.

Theorem 2.3 (Boundedness) Suppose that the conditions (A1), (A2'), and \( (A3) \) hold. Then for a given \( o^0 \in \Omega \times \Lambda^++ \times R_+^{m} \times R^n \), if \( 0 \) is a regular value of \( H_{o^0} \), then \( \Gamma_{o^0} \) is a bounded curve.

Proof: By Theorem 2.2, it is sufficient to show that the \((u, v)\)-component of smooth curve is bounded. Suppose that there exists a sequence \( \{(o^k, \mu_k)\} \in \Gamma_{o^0} \), such that

\[
x^k \to x^*, \quad \mu_k \to \mu^*, \quad \lambda^k \to \lambda^*
\]

and

\[
\left\| u^k, v^k \right\| \to \infty (k \to \infty),
\]

where \( x^* \in \Omega, \mu_* \in [0,1], \lambda^* \geq 0 \). Since closed unit circle of \( R_+^{m*} \) is compact, without loss of generality we can assume that

\[
\lim_{k \to \infty} \left\| u^k, v^k \right\| = (u^*, v^*).
\]

Clearly, \( u^* \geq 0, \left\| u^*, v^* \right\| = 1 \). By (2), we have

\[
(1-\mu_i) \left[ \nabla f(x^*)^T \lambda^k + \nabla g_i(x^*)^T u^k \right] + \nabla h(x^*)^T v^k + \mu_i (x^k - x^0) = 0 \quad \text{(4)}
\]

\[
U^k g(x^k) - \mu_i U^0 g(x^0) = 0 \quad \text{(5)}
\]

Let

\[
I_i(x^*) = \{ i \in I: \lim_{k \to \infty} u_i^k = \infty \},
\]

\[
I_i(x^*) = \{ j \in J: \lim_{k \to \infty} v_i^k = \infty \}.
\]

By (5), we know \( I_i(x^*) \subset I(x^*) \). Rewrite (4) as

\[
(1-\mu_i) \left[ \nabla f(x^*)^T \lambda^k + \sum_{i \in I(x^*)} u_i^k \nabla g_i(x^*) \right] + \sum_{i \in I(x^*)} u_i^k \nabla g_i(x^*) + \nabla h(x^*)^T v^k + \mu_i (x^k - x^0) = 0. \quad \text{(6)}
\]

Divide (6) by \( \left\| u^k, v^k \right\| \to \infty \), since \( \left\| u^k, v^k \right\| \to \infty \), (6) becomes

\[
\nabla g_i(x^*)^T \left( (1-\mu_i) u_i^k \right) + \nabla h(x^*)^T v^k = 0, \quad \text{(7)}
\]

where

\[
\nabla g_i(x^*) = \left\{ \left( g_i(x^*) : i \in I(x^*) \right) \right\},
\]

\[
u_i^k \rightarrow \left\{ \left( u_i^k : i \in I(x^*) \right) \right\}.
\]

1) If \( u_i^k = 0 \), then \( u_i = 0 \), and \( \nabla h(x^*)^T v^k = 0 \). By (A2'), \( v^* = 0 \). This is a contradiction with \( \left\| u^*, v^* \right\| = 1 \).

2) If \( u_i^k \neq 0 \), we consider the following two cases:

- If \( \mu_i \in (0,1) \), then \( (1-\mu_i) u_i^k \geq 0 \) because of \( u_i^k = 0 \). By (A2'), there exists a nonzero vector \( p \in R^n \), such that

\[
p^T \nabla g_i(x^*)^T > 0 \quad \text{and} \quad p^T \nabla h(x^*)^T = 0. \quad \text{(8)}
\]

This, together with (7), implies that \( (1-\mu_i) u_i^k = 0 \), which is a contradiction.

- If \( \mu_i = 1 \), by (7) and (A2'), we know \( v^* = 0 \). So,

\[
\left\| v^* \right\| = 1 \quad \text{since} \quad \left\| u^*, v^* \right\| = 1.
\]

Because of (5), \( I_i(x^*) = I(x^*) \neq \emptyset \). Thus

\[
x^k \in \partial \Omega, \left\| u_i^k \right\| \to \infty (k \to \infty), \forall i \in I.
\]

Without loss of generality, we can assume that

\[
u_i^k > 0, \mu_i \neq 1 (\forall k \in N) \), and \( \lim_{k \to \infty} u_i^k = \mu_i \). \quad \text{(9)}
\]

Hence \( u_i \geq 0, \left\| u_i \right\| = 1 \).

a) If \( I_i(x^*) = \emptyset \), then \( \left\{ v_i^k \right\} \) is bounded. We may assume \( \lim_{k \to \infty} v_i^k = v \). Divide (6) by \( (1-\mu_i) u_i^k \) and let \( k \to \infty \), (6) becomes

\[
\left\| u_i^k \right\| = 1.
\]
\[ \nabla g_t \left( x^\ast \right)^T u + \lim_{k \to \infty} \left[ \frac{1}{\left\| (1 - \mu_k) u_i^k \right\|^2} \nabla h \left( x^\ast \right)^T v^k + \mu_k \left( x^\ast - x^0 \right) \right] = 0. \]  
(9)

This implies that

\[ \lim_{k \to \infty} \frac{1}{\left\| (1 - \mu_k) u_i^k \right\|^2} = \mu. \]

Assume \( \lim_{k \to \infty} \frac{1}{\left\| (1 - \mu_k) u_i^k \right\|^2} = \mu. \)

Then \( \mu \geq 0. \)

If \( \mu > 0, (9) \) becomes

\[ \nabla g_t \left( x^\ast \right)^T \frac{u}{\mu} + \nabla h \left( x^\ast \right)^T v + x^\ast = x^0. \]

This contradicts to condition (A3).

If \( \mu = 0, (9) \) becomes

\[ \nabla g_t \left( x^\ast \right)^T u = 0. \]

By (A2'), there exists a nonzero vector \( p \in \mathbb{R}^n \) such that

\[ \nabla g_t \left( x^\ast \right)^T p > 0, \]

and \( p^T \nabla g_t \left( x^\ast \right)^T u = 0. \)

Thus \( u = 0, \) which contradicts \( \|u\| = 1. \)

b) If \( I_2 \left( x^\ast \right) \neq \emptyset, \) without loss of generality, we can assume that

\[ \lim_{k \to \infty} \frac{1}{\left\| (1 - \mu_k) u_i^k \right\|^2} \left\| (1 - \mu_k) u_i^k, v^k \right\| = (\hat{u}, \hat{v}), \]

where \( \| (\hat{u}, \hat{v}) \| = 1. \)

Since

\[ \left\| (1 - \mu_k) u_i^k, v^k \right\| \to \infty \] as \( k \to \infty, \]

we divide (4) by \( \left\| (1 - \mu_k) u_i^k, v^k \right\|, \) and let \( k \to \infty, \)

we have that

\[ \nabla g_t \left( x^\ast \right)^T \hat{u} + \nabla h \left( x^\ast \right)^T \hat{v} = 0. \]  
(11)

If \( \hat{u} = 0, \) then

\[ \nabla h \left( x^\ast \right)^T \hat{v} = 0. \]

By condition (A2'), \( \hat{v} = 0. \) This is a contradiction since \( \| (\hat{u}, \hat{v}) \| = 1. \)

If \( \hat{u} \neq 0, \) by (A2'), there is a nonzero vector \( p \in \mathbb{R}^n \) such that

\[ p^T \nabla g_t \left( x^\ast \right)^T > 0, \] and \( p^T \nabla h \left( x^\ast \right)^T = 0. \]

This, together with (11), implies \( \hat{u} = 0. \) This is a contradiction.

Therefore, \( \Gamma_{x^0} \) is a bounded curve.

By an analogue argument as in Ref. [1], it is easy to show the following result.

**Theorem 2.4 (Convergence of the method)** Suppose that the conditions (A1), (A2'), and (A3) hold. Then for almost all \( x^0 \in \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^r,\) the zero-set point set \( H_{x^0}^t(0) \) of the homotopy map (2) contains a smooth curve \( \Gamma_{x^0} \in \mathbb{R}^m \times \mathbb{R}^r \times \mathbb{R}^1 \) which starts from \( (x^0,1). \) As \( \mu \to 0, \) the limit set \( T \times \{0\} \subset \mathbb{R}^m \times \mathbb{R}^r \times \{0\} \) of \( \Gamma_{x^0} \) is nonempty, and every point in \( T \) is a solution of (1).

Therefore, Theorem 2.4 shows that for almost all \( x^0 \in \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^r \times \{0\}, \) the homotopy Equation (2) generates a smooth curve \( \Gamma_{x^0} \) starts from \( (x^0,1), \) which is called the homotopy path, the limit set \( T \times \{0\} \subset \mathbb{R}^m \times \mathbb{R}^r \times \{0\} \) of \( \Gamma_{x^0} \) is nonempty, and the \( x \)-component of every point in \( T \) is a KKT point of (MOP), the \( \omega \)-component of the homotopy path is the solution of (1) as \( \mu \) goes to 0.

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