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On ε -optimality conditions for multiobjective fractional optimization problems

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Abstract

A multiobjective fractional optimization problem (MFP), which consists of more than two fractional objective functions with convex numerator functions and convex denominator functions, finitely many convex constraint functions, and a geometric constraint set, is considered. Using parametric approach, we transform the problem (MFP) into the non-fractional multiobjective convex optimization problem (NMCP)_v with parametric $v \in \mathbb{R}^p$, and then give the equivalent relation between (weakly) ε -efficient solution of (MFP) and (weakly) $\bar{\varepsilon}$ -efficient solution of (NMCP)_v. Using the equivalent relations, we obtain ε -optimality conditions for (weakly) ε -efficient solution for (MFP). Furthermore, we present examples illustrating the main results of this study.

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1 Introduction

We need constraint qualifications (for example, the Slater condition) on convex optimization problems to obtain optimality conditions or ε -optimality conditions for the problem.

To get optimality conditions for an efficient solution of a multiobjective optimization problem, we often formulate a corresponding scalar problem. However, it is so difficult that such scalar program satisfies a constraint qualification which we need to derive an optimality condition. Thus, it is very important to investigate an optimality condition for an efficient solution of a multiobjective optimization problem which holds without any constraint qualification.

Jeyakumar et al. [1,2], Kim et al. [3], and Lee et al. [4], gave optimality conditions for convex (scalar) optimization problems, which hold without any constraint qualification. Very recently, Kim et al. [5] obtained ε -optimality theorems for a convex multiobjective optimization problem. The purpose of this article is to extend the ε -optimality theorems of Kim et al. [5] to a multiobjective fractional optimization problem (MFP).

Recently, many authors [5-15] have paid their attention to investigate properties of (weakly) ε -efficient solutions, ε -optimality conditions, and ε -duality theorems for multiobjective optimization problems, which consist of more than two objective functions and a constrained set.

In this article, an MFP, which consists of more than fractional objective functions with convex numerator functions, and convex denominator functions and finitely many convex constraint functions and a geometric constraint set, is considered. We discuss ε -efficient solutions and weakly ε -efficient solutions for (MFP) and obtain ε -optimality theorems for such solutions of (MFP) under weakened constraint qualifications. Furthermore, we prove ε -optimality theorems for the solutions of (MFP) which hold without any constraint qualifications and are expressed by sequences, and present examples illustrating the main results obtained.

2 Preliminaries

Now, we give some definitions and preliminary results. The definitions can be found in [16-18]. Let $g : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ be a convex function. The subdifferential of g at a is given by

$$\partial g(a) := \{v \in \mathbb{R}^n \mid g(x) \geq g(a) + \langle v, x - a \rangle, \quad \forall x \in \text{dom}g\},$$

where $\text{dom}g := \{x \in \mathbb{R}^n \mid g(x) < \infty\}$ and $\langle \cdot, \cdot \rangle$ is the scalar product on \mathbb{R}^n . Let $\varepsilon \geq 0$. The ε -subdifferential of g at $a \in \text{dom}g$ is defined by

$$\partial_\varepsilon g(a) := \{v \in \mathbb{R}^n \mid g(x) \geq g(a) + \langle v, x - a \rangle - \varepsilon, \quad \forall x \in \text{dom}g\}.$$

The conjugate function of $g : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ is defined by

$$g^*(v) = \sup\{\langle v, x \rangle - g(x) \mid x \in \mathbb{R}^n\}.$$

The epigraph of g , epig , is defined by

$$\text{epig} = \{(x, r) \in \mathbb{R}^n \times \mathbb{R} \mid g(x) \leq r\}.$$

For a nonempty closed convex set $C \subset \mathbb{R}^n$, $\delta_C : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ is called the indicator of C if $\delta_C(x) = \begin{cases} 0 & \text{if } x \in C, \\ +\infty & \text{otherwise} \end{cases}$.

Lemma 2.1 [19] *If $h : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ is a proper lower semicontinuous convex function and if $a \in \text{dom}h$, then*

$$\text{epih}^* = \bigcup_{\varepsilon \geq 0} \{(v, \langle v, a \rangle + \varepsilon - h(a)) \mid v \in \partial_\varepsilon h(a)\}.$$

Lemma 2.2 [20] *Let $h : \mathbb{R}^n \rightarrow \mathbb{R}$ be a continuous convex function and $u : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ be a proper lower semicontinuous convex function. Then*

$$\text{epi}(h + u)^* = \text{epih}^* + \text{epiu}^*.$$

Now, we give the following Farkas lemma which was proved in [2,5], but for the completeness, we prove it as follows:

Lemma 2.3 *Let $h_i : \mathbb{R}^n \rightarrow \mathbb{R}$, $i = 0, 1, \dots, l$ be convex functions. Suppose that $\{x \in \mathbb{R}^n \mid h_i(x) \leq 0, i = 1, \dots, l\} \neq \emptyset$. Then the following statements are equivalent:*

- (i) $\{x \in \mathbb{R}^n \mid h_i(x) \leq 0, i = 1, \dots, l\} \subseteq \{x \in \mathbb{R}^n \mid h_0(x) \geq 0\}$
- (ii) $0 \in \text{epih}_0^* + \text{cl} \bigcup_{\lambda_i \geq 0} \text{epi}(\sum_{i=1}^l \lambda_i h_i)^*$.

Proof. Let $Q = \{x \in \mathbb{R}^n \mid h_i(x) \leq 0, i = 1, \dots, l\}$. Then $Q \neq \emptyset$ and by Lemma 2.1 in [2], $\text{epi}\delta_Q^* = \text{cl} \bigcup_{\lambda_i \geq 0} \text{epi}(\sum_{i=1}^l \lambda_i h_i)^*$. Hence, by Lemma 2.2, we can verify that (i) if and only if (ii).

Lemma 2.4 [16] *Let $h_i : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$, $i = 1, \dots, m$ be proper lower semi-continuous convex functions. Let $\varepsilon \geq 0$. if $\bigcap_{i=1}^m \text{ri dom} h_i \neq \emptyset$, where $\text{ri dom} h_i$ is the relative interior of $\text{dom} h_i$, then for all $x \in \bigcup_{i=1}^m \text{dom} h_i$*

$$\partial_\varepsilon \left(\sum_{i=1}^m h_i \right) (x) = \bigcup_{i=1}^m \left\{ \sum_{i=1}^m \partial_{\varepsilon_i} h_i(x) \mid \varepsilon_i \geq 0, i = 1, \dots, m, \sum_{i=1}^m \varepsilon_i = \varepsilon \right\}.$$

3 ε -optimality theorems

Consider the following MFP:

$$\begin{aligned} \text{(MFP)} \quad & \text{Minimize} \quad \frac{f(x)}{g(x)} := \left(\frac{f_1(x)}{g_1(x)}, \dots, \frac{f_p(x)}{g_p(x)} \right) \\ & \text{subject to} \quad x \in Q := \{x \in \mathbb{R}^n \mid h_j(x) \leq 0, j = 1, \dots, m\}. \end{aligned}$$

Let $f_i : \mathbb{R}^n \rightarrow \mathbb{R}$, $i = 1, \dots, p$ be convex functions, $g_i : \mathbb{R}^n \rightarrow \mathbb{R}$, $i = 1, \dots, p$, concave functions such that for any $x \in Q$, $f_i(x) \geq 0$ and $g_i(x) > 0$, $i = 1, \dots, p$, and $h_j : \mathbb{R}^n \rightarrow \mathbb{R}$, $j = 1, \dots, m$, convex functions. Let $\varepsilon = (\varepsilon_1, \dots, \varepsilon_p)$, where $\varepsilon_i \geq 0$, $i = 1, \dots, p$.

Now, we give the definition of ε -efficient solution of (MFP) which can be found in [11].

Definition 3.1 *The point $\bar{x} \in Q$ is said to be an ε -efficient solution of (MFP) if there does not exist $x \in Q$ such that*

$$\begin{aligned} \frac{f_i(x)}{g_i(x)} &\leq \frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i, \text{ for all } i = 1, \dots, p, \\ \frac{f_j(x)}{g_j(x)} &< \frac{f_j(\bar{x})}{g_j(\bar{x})} - \varepsilon_j, \text{ for some } j \in \{1, \dots, p\}. \end{aligned}$$

When $\varepsilon = 0$, then the ε -efficiency becomes the efficiency for (MFP) (see the definition of efficient solution of a multiobjective optimization problem in [21]).

Now, we give the definition of weakly ε -efficient solution of (MFP) which is weaker than ε -efficient solution of (MFP).

Definition 3.2 *A point $\bar{x} \in Q$ is said to be a weakly ε -efficient solution of (MFP) if there does not exist $x \in Q$ such that*

$$\frac{f_i(x)}{g_i(x)} < \frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i, \text{ for all } i = 1, \dots, p.$$

When $\varepsilon = 0$, then the weak ε -efficiency becomes the weak efficiency for (MFP) (see the definition of efficient solution of a multiobjective optimization problem in [21]).

Using parametric approach, we transform the problem (MFP) into the nonfractional multiobjective convex optimization problem (NMCP) $_\nu$ with parametric $\nu \in \mathbb{R}^p$:

$$\begin{aligned} \text{(NMCP)}_\nu \quad & \text{Minimize} \quad (f(x) - \nu g(x)) := (f_1(x) - \nu_1 g_1(x), \dots, f_p(x) - \nu_p g_p(x)) \\ & \text{subject to} \quad x \in Q. \end{aligned}$$

Adapting Lemma 4.1 in [22] and modifying Proposition 3.1 in [12], we can obtain the following proposition:

Proposition 3.1 *Let $\bar{x} \in Q$. Then the following are equivalent:*

- (i) \bar{x} is an ε -efficient solution of (MFP).
- (ii) \bar{x} is an $\bar{\varepsilon}$ -efficient solution of (NMCP) $_{\bar{v}}$, where $\bar{v} := \left(\frac{f_1(\bar{x})}{g_1(\bar{x})} - \varepsilon_1, \dots, \frac{f_p(\bar{x})}{g_p(\bar{x})} - \varepsilon_p \right)$ and $\bar{\varepsilon} = (\varepsilon_1 g_1(\bar{x}), \dots, \varepsilon_p g_p(\bar{x}))$.
- (iii) $Q \cap S(\bar{x}) = \emptyset$ or

$$\sum_{i=1}^p \left[f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \right] \geq 0 = \sum_{i=1}^p \left[f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) \right] - \sum_{i=1}^p \varepsilon_i g_i(\bar{x}) \text{ for any } x \in Q \cap S(\bar{x}),$$

where $S(\bar{x}) = \{x \in \mathbb{R}^n \mid f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \leq 0 = f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) - \bar{\varepsilon}_i, i = 1, \dots, p\}$.

Proof. (i) \Leftrightarrow (ii): It follows from Lemma 4.1 in [22].

(ii) \Rightarrow (iii): Let \bar{x} be an $\bar{\varepsilon}$ -efficient solution of (NMCP) $_{\bar{v}}$, where $\bar{v} := \left(\frac{f_1(\bar{x})}{g_1(\bar{x})} - \varepsilon_1, \dots, \frac{f_p(\bar{x})}{g_p(\bar{x})} - \varepsilon_p \right)$ and $\bar{\varepsilon} = (\varepsilon_1 g_1(\bar{x}), \dots, \varepsilon_p g_p(\bar{x}))$. Then $Q \cap S(\bar{x}) = \emptyset$ or $Q \cap S(\bar{x}) \neq \emptyset$. Suppose that $Q \cap S(\bar{x}) \neq \emptyset$. Then for any $x \in Q \cap S(\bar{x})$ and all $i = 1, \dots, p$,

$$f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \leq f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) - \bar{\varepsilon}_i.$$

Hence the $\bar{\varepsilon}$ -efficiency of \bar{x} yields

$$f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) = f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) - \bar{\varepsilon}_i$$

for any $x \in Q \cap S(\bar{x})$ and all $i = 1, \dots, p$. Thus we have, for all $x \in Q \cap S(\bar{x})$,

$$\sum_{i=1}^p \left[f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \right] = \sum_{i=1}^p \left[f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) \right] - \sum_{i=1}^p \bar{\varepsilon}_i.$$

(iii) \Rightarrow (ii): Suppose that $Q \cap S(\bar{x}) = \emptyset$. Then there does not exist $x \in Q$ such that $x \in S(\bar{x})$; that is, there does not exist $x \in Q$ such that

$$f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \leq f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) - \bar{\varepsilon}_i$$

for all $i = 1, \dots, p$. Hence, there does not exist $x \in Q$ such that

$$\begin{aligned} f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) &\leq f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) - \bar{\varepsilon}_i, \quad i = 1, \dots, p, \\ f_j(x) - \left(\frac{f_j(\bar{x})}{g_j(\bar{x})} - \varepsilon_j \right) g_j(x) &< f_j(\bar{x}) - \left(\frac{f_j(\bar{x})}{g_j(\bar{x})} - \varepsilon_j \right) g_j(\bar{x}) - \bar{\varepsilon}_j, \quad \text{for some } j \in \{1, \dots, p\}. \end{aligned}$$

Therefore, \bar{x} is an $\bar{\varepsilon}$ -efficient solution of (NMCP) $_{\bar{v}}$, where $\bar{v} := \left(\frac{f_1(\bar{x})}{g_1(\bar{x})} - \varepsilon_1, \dots, \frac{f_p(\bar{x})}{g_p(\bar{x})} - \varepsilon_p \right)$.

Assume that $Q \cap S(\bar{x}) \neq \emptyset$. Then, from this assumption

$$\sum_{i=1}^p \left[f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \right] \geq \sum_{i=1}^p \left[f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) \right] - \sum_{i=1}^p \bar{\varepsilon}_i, \quad (3.1)$$

for any $x \in Q \cap S(\bar{x})$. Suppose to the contrary that \bar{x} is not an $\bar{\varepsilon}$ -efficient solution of $(\text{NMCP})_{\bar{v}}$. Then, there exist $\hat{x} \in Q$ and an index j such that

$$\begin{aligned} f_i(\hat{x}) - \bar{v}_i g_i(\hat{x}) &\leq f_i(\bar{x}) - \bar{v}_i g_i(\bar{x}) - \bar{\varepsilon}_i, \quad i = 1, \dots, p, \\ f_j(\hat{x}) - \bar{v}_j g_j(\hat{x}) &< f_j(\bar{x}) - \bar{v}_j g_j(\bar{x}) - \bar{\varepsilon}_j, \quad \text{for some } j \in \{1, \dots, p\}. \end{aligned}$$

Therefore, $\hat{x} \in Q \cap S(\bar{x})$ and $\sum_{i=1}^p \left[f_i(\hat{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\hat{x}) \right] < \sum_{i=1}^p \left[f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) \right] - \sum_{i=1}^p \bar{\varepsilon}_i$, which contradicts the above inequality. Hence, \bar{x} is an $\bar{\varepsilon}$ -efficient solution of $(\text{NMCP})_{\bar{v}}$.

We can easily obtain the following proposition:

Proposition 3.2 *Let $\bar{x} \in Q$ and suppose that $f_i(\bar{x}) \geq \varepsilon_i g_i(\bar{x})$, $i = 1, \dots, p$. Then the following are equivalent:*

- (i) \bar{x} is a weakly ε -efficient solution of (MFP).
- (ii) \bar{x} is a weakly $\bar{\varepsilon}$ -efficient solution of $(\text{NMCP})_{\bar{v}}$, where $\bar{\varepsilon} = (\varepsilon_1 g_1(\bar{x}), \dots, \varepsilon_p g_p(\bar{x}))$ and $\bar{\varepsilon} = (\varepsilon_1 g_1(\bar{x}), \dots, \varepsilon_p g_p(\bar{x}))$.
- (iii) there exists $\bar{\lambda} := (\bar{\lambda}_1, \dots, \bar{\lambda}_p) \in \mathbb{R}_+^p \setminus \{0\}$ such that

$$\begin{aligned} &\sum_{i=1}^p \bar{\lambda}_i \left[f_i(x) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x) \right] \\ &\geq 0 = \sum_{i=1}^p \bar{\lambda}_i \left[f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(\bar{x}) \right] - \sum_{i=1}^p \bar{\lambda}_i \varepsilon_i g_i(\bar{x}) \text{ for any } x \in Q. \end{aligned}$$

Proof. (i) \Leftrightarrow (ii): The proof is also following the similar lines of Proposition 3.1.

(ii) \Rightarrow (iii): Let $\phi(x) = (\phi_1(x), \dots, \phi_p(x))$, $\forall x \in Q$, where $\phi_i(x) = f_i(\bar{x}) - \left(\frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i \right) g_i(x)$, $i = 1, \dots, p$. Then, $\phi_i(x)$, $i = 1, \dots, p$, are convex. Since $\bar{x} \in Q$ is a weakly ε -efficient solution of $(\text{NMCP})_{\bar{v}}$, where $(\varphi(Q) + \mathbb{R}_+^p) \cap (-\text{int} \mathbb{R}_+^p) = \emptyset$, $(\varphi(Q) + \mathbb{R}_+^p) \cap (-\text{int} \mathbb{R}_+^p) = \emptyset$, and hence, it follows from separation theorem that there exist $\bar{\lambda}_i \geq 0$, $i = 1, \dots, p$, $(\bar{\lambda}_1, \dots, \bar{\lambda}_p) \neq 0$ such that

$$\sum_{i=1}^p \bar{\lambda}_i \phi_i(x) \geq 0 \quad \forall x \in Q.$$

Thus (iii) holds.

(iii) \Rightarrow (ii): If (ii) does not hold, that is, \bar{x} is not a weakly $\bar{\varepsilon}$ -efficient solution of $(\text{NMCP})_{\bar{v}}$, then (iii) does not hold. \square

We present a necessary and sufficient ε -optimality theorem for ε -efficient solution of (MFP) under a constraint qualification, which will be called the closedness assumption.

Theorem 3.1 *Let $\bar{x} \in Q$ and assume that $Q \cap S(\bar{x}) \neq \emptyset$ and $f_i(\bar{x}) \geq \varepsilon_i g_i(\bar{x})$, $i = 1, \dots, p$. Suppose that*

$$\bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^* + \bigcup_{\mu_i \geq 0} \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*]$$

is closed, where $\bar{v}_i = \frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i$, $i = 1, \dots, p$. Then the following are equivalent.

(i) \bar{x} is an ε -efficient solution of (MFP).

$$\begin{aligned} \begin{pmatrix} 0 \\ 0 \end{pmatrix}^T &\in \sum_{i=1}^p [\text{epi}f_i^* + \text{epi}(-\bar{v}_i g_i)^*] + \bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^* \\ \text{(ii)} \quad &+ \bigcup_{\mu_i \geq 0} \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*]. \end{aligned}$$

(iii) there exist $\alpha_i \geq 0$, $u_i \in \partial_{\alpha_i} f_i(\bar{x})$, $\beta_i \geq 0$, $\gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$, $\lambda_j \geq 0$, $\gamma_j \geq 0$, $w_j \in \partial_{\gamma_j}(\lambda_j h_j)(\bar{x})$, $j = 1, \dots, m$, $\mu_i \geq 0$, $q_i \geq 0$, $s_i \in \partial_{q_i}(\mu_i f_i)(\bar{x})$, $z_i \geq 0$, $t_i \in \partial_{z_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$ such that

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \sum_{j=1}^m w_j + \sum_{i=1}^p (s_i + t_i)$$

and

$$\sum_{i=1}^p (\alpha_i + \beta_i + q_i + z_i) + \sum_{j=1}^m \gamma_j = \sum_{i=1}^p \varepsilon_i (1 + \mu_i) g_i(\bar{x}) + \sum_{j=1}^m \lambda_j h_j(\bar{x}).$$

Proof. Let $h_0(x) = \sum_{i=1}^p [f_i(x) - \bar{v}_i g_i(x)]$.

(i) \Leftrightarrow (by Proposition 3.1) $h_0(x) \geq 0, \forall x \in Q \cap S(\bar{x})$.

$\Leftrightarrow \{x | f_i(x) - \bar{v}_i g_i(x) \leq 0, i = 1, \dots, p, h_j(x) \leq 0, j = 1, \dots, m\} \subset \{x | h_0(x) \geq 0\}$.

\Leftrightarrow (by lemma 2.3)

$$\begin{aligned} \begin{pmatrix} 0 \\ 0 \end{pmatrix}^T &\in \sum_{i=1}^p [\text{epi}f_i^* + \text{epi}(-\bar{v}_i g_i)^*] + \text{cl} \left\{ \bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^* \right. \\ &\left. + \bigcup_{\mu_i \geq 0} \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] \right\}. \end{aligned}$$

Thus by the closedness assumption, (i) is equivalent to (ii).

(ii) \Leftrightarrow (iii): (ii) \Leftrightarrow (by Lemma 2.1), there exist $\alpha_i \geq 0$, $u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x})$, $i = 1, \dots, p$, $\beta_i \geq 0$, $\gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$, $\lambda_j \geq 0$, $\gamma_j \geq 0$, $w_j \in \partial_{\gamma_j}(\lambda_j h_j)(\bar{x})$, $j = 1, \dots, m$, $\mu_i \geq 0$, $q_i \geq 0$, $s_i \in \partial_{q_i}(\mu_i f_i)(\bar{x})$, $i = 1, \dots, p$, $z_i \geq 0$, $t_i \in \partial_{z_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$ such that

$$\begin{aligned} \begin{pmatrix} 0 \\ 0 \end{pmatrix}^T &= \sum_{i=1}^p \left[\begin{pmatrix} u_i \\ \langle u_i, \bar{x} \rangle + \alpha_i - f_i(\bar{x}) \end{pmatrix}^T + \begin{pmatrix} \gamma_i \\ \langle \gamma_i, \bar{x} \rangle + \beta_i - (-\bar{v}_i \mu_i g_i)(\bar{x}) \end{pmatrix}^T \right] \\ &+ \sum_{j=1}^m \begin{pmatrix} w_j \\ \langle w_j, \bar{x} \rangle + \gamma_j - (\lambda_j h_j)(\bar{x}) \end{pmatrix}^T \\ &+ \sum_{i=1}^p \left[\begin{pmatrix} s_i \\ \langle s_i, \bar{x} \rangle + q_i - (\mu_i f_i)(\bar{x}) \end{pmatrix}^T + \begin{pmatrix} t_i \\ \langle t_i, \bar{x} \rangle + z_i - (-\bar{v}_i \mu_i g_i)(\bar{x}) \end{pmatrix}^T \right]. \end{aligned}$$

\Leftrightarrow there exist $\alpha_i \geq 0$, $u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x})$, $\beta_i \geq 0$, $\gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$, $\lambda_j \geq 0$, $\gamma_j \geq 0$, $w_j \in \partial_{\gamma_j}(\lambda_j h_j)(\bar{x})$, $j = 1, \dots, m$, $\mu_i \geq 0$, $q_i \geq 0$, $s_i \in \partial_{q_i}(\mu_i f_i)(\bar{x})$, $z_i \geq 0$, $t_i \in \partial_{z_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$ such that

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \sum_{j=1}^m w_j + \sum_{i=1}^p (s_i + t_i)$$

$$\text{and } \sum_{i=1}^p (\alpha_i + \beta_i + q_i + z_i) + \sum_{j=1}^m \gamma_j = \sum_{i=1}^p \left[f_i(\bar{x}) - \bar{v}_i g_i(\bar{x}) + (\mu_i f_i)(\bar{x}) - (\bar{v}_i \mu_i g_i)(\bar{x}) + \sum_{j=1}^m \lambda_j h_j(\bar{x}) \right].$$

⇔ (iii) holds. □

Now we give a necessary and sufficient ε -optimality theorem for ε -efficient solution of (MFP) which holds without any constraint qualification.

Theorem 3.2 *Let $\bar{x} \in Q$. Suppose that $Q \cap S(\bar{x}) \neq \emptyset$ and $f_i(\bar{x}) \geq \varepsilon_i g_i(\bar{x})$, $i = 1, \dots, p$, $i = 1, \dots, p$. Then \bar{x} is an ε -efficient solution of (MFP) if and only if there exist $\alpha_i \geq 0$, $u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x})$, $i = 1, \dots, p$, $\beta_i \geq 0$, $\gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$, $\lambda_j^n \geq 0$, $\gamma_j^n \geq 0$, $w_j^n \in \partial_{\gamma_j^n}(\lambda_j^n h_j)(\bar{x})$, $j = 1, \dots, m$, $\mu_k^n \geq 0$, $q_k^n \geq 0$, $s_k^n \in \partial_{q_k^n}(\mu_k^n f_k)(\bar{x})$, $z_k^n \geq 0$, $t_k^n \in \partial_{z_k^n}(-\bar{v}_k \mu_k^n g_k)(\bar{x})$, $k = 1, \dots, p$ such that*

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \lim_{n \rightarrow \infty} \left[\sum_{j=1}^m w_j^n + \sum_{k=1}^p (s_k^n + t_k^n) \right]$$

and

$$\begin{aligned} \sum_{i=1}^p \varepsilon_i g_i(\bar{x}) &= \sum_{i=1}^p (\alpha_i + \beta_i) + \lim_{n \rightarrow \infty} \left\{ \sum_{j=1}^m [\gamma_j^n - (\lambda_j^n h_j)(\bar{x})] \right. \\ &\quad \left. + \sum_{k=1}^p [q_k^n + z_k^n - \mu_k^n \varepsilon_k g_k(\bar{x})] \right\}. \end{aligned}$$

Proof. \bar{x} is an ε -efficient solution of (MFP)

⇔ (from the proof of Theorem 3.1)

$$\begin{aligned} \begin{pmatrix} 0 \\ 0 \end{pmatrix}^T &\in \sum_{i=1}^p [\text{epi} f_i^* + \text{epi}(-\bar{v}_i g_i)^*] + \text{cl} \left\{ \bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^* \right. \\ &\quad \left. + \bigcup_{\mu_i \geq 0} \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] \right\}. \end{aligned}$$

⇔ (by Lemma 2.1) there exist $\alpha_i \geq 0$, $u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x})$, $i = 1, \dots, p$, $\beta_i \geq 0$, $\gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x})$, $i = 1, \dots, p$, $\lambda_j^n \geq 0$, $\gamma_j^n \geq 0$, $w_j^n \in \partial_{\gamma_j^n}(\lambda_j^n h_j)(\bar{x})$, $j = 1, \dots, m$, $\mu_k^n \geq 0$, $s_k^n \in \partial_{q_k^n}(\mu_k^n f_k)(\bar{x})$, $z_k^n \in \partial_{z_k^n}(\mu_k^n f_k)(\bar{x})$, $z_k^n \geq 0$, $t_k^n \in \partial_{z_k^n}(-\bar{v}_k \mu_k^n g_k)(\bar{x})$, $k = 1, \dots, p$, such that

$$\begin{aligned} \begin{pmatrix} 0 \\ 0 \end{pmatrix}^T &= \sum_{i=1}^p \left[\begin{pmatrix} u_i \\ \langle u_i, \bar{x} \rangle + \alpha_i - f_i(\bar{x}) \end{pmatrix}^T + \begin{pmatrix} \gamma_i \\ \langle \gamma_i, \bar{x} \rangle + \beta_i - (-\bar{v}_i g_i)(\bar{x}) \end{pmatrix}^T \right] \\ &\quad + \lim_{n \rightarrow \infty} \left\{ \sum_{j=1}^m \begin{pmatrix} w_j^n \\ \langle w_j^n, \bar{x} \rangle + \gamma_j^n - (\lambda_j^n h_j)(\bar{x}) \end{pmatrix}^T \right. \\ &\quad \left. + \sum_{k=1}^p \left[\begin{pmatrix} s_k^n \\ \langle s_k^n, \bar{x} \rangle + q_k^n - (\mu_k^n f_k)(\bar{x}) \end{pmatrix}^T + \begin{pmatrix} t_k^n \\ \langle t_k^n, \bar{x} \rangle + z_k^n - (-\bar{v}_k \mu_k^n g_k)(\bar{x}) \end{pmatrix}^T \right] \right\}. \end{aligned}$$

\Leftrightarrow there exist $\alpha_i \geq 0, u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x}), i = 1, \dots, p, \beta_i \geq 0, \gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x}), i = 1, \dots, p, \lambda_j^n \geq 0, \gamma_j^n \geq 0, w_j^n \in \partial_{\gamma_j^n}(\lambda_j^n h_j)(\bar{x}), j = 1, \dots, m, \mu_k^n \geq 0, q_k^n \geq 0, s_k^n \in \partial_{q_k^n}(\mu_k^n f_k)(\bar{x}), t_k^n \in \partial_{z_k^n}(-\bar{v}_k \mu_k^n g_k)(\bar{x}), t_k^n \in \partial_{z_k^n}(-\bar{v}_k \mu_k^n g_k)(\bar{x}), k = 1, \dots, p,$ such that

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \lim_{n \rightarrow \infty} \left[\sum_{j=1}^m w_j^n + \sum_{k=1}^p (s_k^n + t_k^n) \right]$$

and

$$\sum_{i=1}^p \varepsilon_i g_i(\bar{x}) = \sum_{i=1}^p (\alpha_i + \beta_i) + \lim_{n \rightarrow \infty} \left\{ \sum_{j=1}^m [\gamma_j^n - (\lambda_j^n h_j)(\bar{x})] + \sum_{k=1}^p [q_k^n + z_k^n - \mu_k^n \varepsilon_k g_k(\bar{x})] \right\}.$$

We present a necessary and sufficient ε -optimality theorem for weakly ε -efficient solution of (MFP) under a constraint qualification.

Theorem 3.3 *Let $\bar{x} \in Q$ and assume that $f_i(\bar{x}) \geq \varepsilon_i g_i(\bar{x}), i = 1, \dots, p, i = 1, \dots, p,$ and $\bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^*$ is closed. Then the following are equivalent.*

- (i) \bar{x} is a weakly ε -efficient solution of (MFP).
- (ii) there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1$ such that

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}^T \in \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] + \bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^*,$$

where $\bar{v}_i = \frac{f_i(\bar{x})}{g_i(\bar{x})} - \varepsilon_i, i = 1, \dots, p.$

- (iii) there exist $\mu_i \geq 0, \sum_{i=1}^p \mu_i = 1, \alpha_i \geq 0, u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x}), \beta_i \geq 0, \gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x}), i = 1, \dots, p, \lambda_j \geq 0, \gamma_j \geq 0, w_j \in \partial_{\gamma_j}(\lambda_j h_j)(\bar{x}), j = 1, \dots, m,$ such that

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \sum_{j=1}^m w_j$$

and

$$\sum_{i=1}^p \mu_i \varepsilon_i g_i(\bar{x}) = \sum_{i=1}^p (\alpha_i + \beta_i) + \sum_{j=1}^m [\gamma_j - (\lambda_j h_j)(\bar{x})].$$

Proof. (i) \Leftrightarrow (ii): \bar{x} is a weakly ε -efficient solution of (MFP)

\Leftrightarrow (by Proposition 3.2) there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1$ such that

$$\sum_{i=1}^p \mu_i [f_i(x) - \bar{v}_i g_i(x)] \geq 0 \quad \forall x \in Q$$

\Leftrightarrow there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1$ such that

$$\{x | h_j(x) \leq 0, j = 1, \dots, m\} \subset \{x | \sum_{i=1}^p \mu_i [f_i(x) - \bar{v}_i g_i(x)] \geq 0\}$$

⇔ (by Lemma 2.3) there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1$ such that

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}^T \in \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] + \text{cl} \left\{ \bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^* \right\}.$$

Thus, by the closedness assumption, (i) is equivalent to (ii).

(ii) ⇔ (iii): (ii) ⇔ (by Lemma 2.1) there exist $\mu_i \geq 0, \sum_{i=1}^p \mu_i = 1, \alpha_i \geq 0, u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x}), \beta_i \geq 0, \gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x}), i = 1, \dots, p, \lambda_j \geq 0, \gamma_j \geq 0, w_j \in \partial_{\gamma_j}(\lambda_j h_j)(\bar{x}), j = 1, \dots, m,$ such that

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}^T = \sum_{i=1}^p \left[\begin{pmatrix} u_i \\ \langle u_i, \bar{x} \rangle + \alpha_i - (\mu_i f_i)(\bar{x}) \end{pmatrix}^T + \begin{pmatrix} \gamma_i \\ \langle \gamma_i, \bar{x} \rangle + \beta_i - (-\bar{v}_i \mu_i g_i)(\bar{x}) \end{pmatrix}^T \right] + \sum_{j=1}^m \begin{pmatrix} w_j \\ \langle w_j, \bar{x} \rangle + \gamma_j - (\lambda_j h_j)(\bar{x}) \end{pmatrix}^T.$$

⇔ (iii) holds. □

Now, we propose a necessary and sufficient ε -optimality theorem for weakly ε -efficient solution of (MFP) which holds without any constraint qualification.

Theorem 3.4 *Let $\bar{x} \in Q$ and assume that $f_i(\bar{x}) \geq \varepsilon_i g_i(\bar{x}), i = 1, \dots, p.$ Then \bar{x} is a weakly ε -efficient solution of (MFP) if and only if there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1, \alpha_i \geq 0, u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x}), i = 1, \dots, p, \beta_i \geq 0, \gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x}), i = 1, \dots, p, \gamma_j^n \geq 0, \gamma_j^n \geq 0, w_j^n \in \partial_{\gamma_j^n}(\lambda_j^n h_j)(\bar{x}), j = 1, \dots, m,$ such that*

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \lim_{n \rightarrow \infty} \sum_{j=1}^m w_j^n$$

and

$$\sum_{i=1}^p \mu_i \varepsilon_i g_i(\bar{x}) = \sum_{i=1}^p (\alpha_i + \beta_i) + \lim_{n \rightarrow \infty} \sum_{j=1}^m [\gamma_j^n - (\lambda_j^n h_j)(\bar{x})].$$

Proof. \bar{x} is a weakly ε -efficient solution of (MFP)

⇔ ((from the proof of Theorem 3.3) there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1$ such that

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}^T \in \sum_{i=1}^p [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] + \text{cl} \left\{ \bigcup_{\lambda_j \geq 0} \sum_{j=1}^m \text{epi}(\lambda_j h_j)^* \right\}.$$

⇔ (by Lemma 2.1) there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1, \alpha_i \geq 0, u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x}), i = 1, \dots, p, \beta_i \geq 0, \gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x}), i = 1, \dots, p, \lambda_j^n \geq 0, \gamma_j^n \geq 0, w_j^n \in \partial_{\gamma_j^n}(\lambda_j^n h_j)(\bar{x}), j = 1, \dots, m,$ such that

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}^T = \sum_{i=1}^p \left[\begin{pmatrix} u_i \\ \langle u_i, \bar{x} \rangle + \alpha_i - (\mu_i f_i)(\bar{x}) \end{pmatrix}^T + \begin{pmatrix} \gamma_i \\ \langle \gamma_i, \bar{x} \rangle + \beta_i - (-\bar{v}_i \mu_i g_i)(\bar{x}) \end{pmatrix}^T \right] + \lim_{n \rightarrow \infty} \left\{ \sum_{j=1}^m \begin{pmatrix} w_j^n \\ \langle w_j^n, \bar{x} \rangle + \gamma_j^n - (\lambda_j^n h_j)(\bar{x}) \end{pmatrix}^T \right\}.$$

\Leftrightarrow there exist $\mu_i \geq 0, i = 1, \dots, p, \sum_{i=1}^p \mu_i = 1, \alpha_i \geq 0, u_i \in \partial_{\alpha_i}(\mu_i f_i)(\bar{x}), i = 1, \dots, p, \beta_i \geq 0, \gamma_i \in \partial_{\beta_i}(-\bar{v}_i \mu_i g_i)(\bar{x}), i = 1, \dots, p, \lambda_j^n \geq 0, \gamma_j^n \geq 0, w_j^n \in \partial_{\gamma_j^n}(\lambda_j^n h_j^n)(\bar{x}), j = 1, \dots, m$, such that

$$0 = \sum_{i=1}^p (u_i + \gamma_i) + \lim_{n \rightarrow \infty} \sum_{j=1}^m w_j^n$$

and

$$\sum_{i=1}^p \mu_i \varepsilon_i g_i(\bar{x}) = \sum_{i=1}^p (\alpha_i + \beta_i) + \lim_{n \rightarrow \infty} \sum_{j=1}^m [\gamma_j^n - (\gamma_j^n h_j)(\bar{x})].$$

□

Now, we give examples illustrating Theorems 3.1, 3.2, 3.3, and 3.4.

Example 3.1 Consider the following MFP:

$$\begin{aligned} \text{(MFP)}_1 \text{ Minimize } & \begin{pmatrix} x_1 \\ x_2 \\ x_1 \end{pmatrix} \\ \text{subject to } & (x_1, x_2) \in Q := \{(x_1, x_2) \in \mathbb{R}^2 \mid -x_1 + 1 \leq 0, \quad -x_2 + 1 \leq 0\}. \end{aligned}$$

Let $\varepsilon = (\varepsilon_1, \varepsilon_2) = (\frac{1}{2}, \frac{1}{2})$, and $f_1(x_1, x_2) = x_1, g_1(x_1, x_2) = 1, f_2(x_1, x_2) = x_2, g_2(x_1, x_2) = x_1, h_1(x_1, x_2) = -x_1 + 1$ and $h_2(x_1, x_2) = -x_2 + 1$.

(1) Let $(\bar{x}_1, \bar{x}_2) = (\frac{3}{2}, \frac{3}{4})$. Then (\bar{x}_1, \bar{x}_2) is an ε -efficient solution of (MFP)₁.

Let $\bar{v}_1 = \frac{f_1(\bar{x}_1, \bar{x}_2)}{g_1(\bar{x}_1, \bar{x}_2)} - \varepsilon_1$ and $\bar{v}_2 = \frac{f_2(\bar{x}_1, \bar{x}_2)}{g_2(\bar{x}_1, \bar{x}_2)} - \varepsilon_2$. Then $\bar{v}_1 = \bar{v}_2 = 1$, and

$$\begin{aligned} & Q \cap S(\bar{x}_1, \bar{x}_2) \\ &= Q \cap \{(\bar{x}_1, \bar{x}_2) \in \mathbb{R}^2 \mid f_1(\bar{x}_1, \bar{x}_2) - \bar{v}_1 g_1(\bar{x}_1, \bar{x}_2) \leq 0, f_2(\bar{x}_1, \bar{x}_2) - \bar{v}_2 g_2(\bar{x}_1, \bar{x}_2) \leq 0\} \\ &= \{(1, 1)\}. \end{aligned}$$

Thus, $Q \cap S(\bar{x}_1, \bar{x}_2) \neq \emptyset$. It is clear that $f_1(\bar{x}_1, \bar{x}_2) \geq \varepsilon_1 g_1(\bar{x}_1, \bar{x}_2)$ and $f_2(\bar{x}_1, \bar{x}_2) \geq \varepsilon_2 g_2(\bar{x}_1, \bar{x}_2)$. Let $A = \bigcup_{\substack{\lambda_1 \geq 0, \\ \lambda_2 \geq 0}} \sum_{j=1}^2 \text{epi}(\lambda_j h_j)^* + \bigcup_{\substack{\mu_1 \geq 0, \\ \mu_2 \geq 0}} \sum_{i=1}^2 [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*]$.

Then

$$\begin{aligned} A &= \bigcup_{\substack{\lambda_1 \geq 0, \lambda_2 \geq 0 \\ \mu_1 \geq 0, \mu_2 \geq 0}} \text{epi} \left(\sum_{j=1}^2 \lambda_j h_j + \sum_{i=1}^2 \mu_i (f_i - \bar{v}_i g_i) \right)^* \\ &= \text{cone co}\{(-1, 0, -1), (0, -1, -1), (1, 0, 1), (-1, 1, 0), (0, 0, 1)\}, \end{aligned}$$

where $\text{co}D$ is the convexhull of a set D and $\text{cone co}D$ is the cone generated by $\text{co}D$.

Thus A is closed. Let $B = \sum_{i=1}^2 [\text{epi} f_i^* + \text{epi}(-\bar{v}_i g_i)^*] + A$. Then

$B = \{(1, 0)\} \times [0, \infty) + \{(0, 0)\} \times [1, \infty) + \{(0, 1)\} \times [0, \infty) + \{(-1, 0)\} \times [0, \infty) + A$. Since $(0, -1, -1) \in A, (0, 0, 0) \in B$. Thus (ii) of Theorem 3.1 holds. Let $\alpha_1 = \beta_1 = \gamma_1 = q_1 = z_1 = \alpha_2 = \beta_2 = \gamma_2 = q_2 = z_2 = 0$, and let $\mu_1 = \mu_2 = 1$, and $\lambda_1 = 0$ and $\lambda_1 = 2$. Moreover, $\partial f_2(\bar{x}_1, \bar{x}_2) = \{(0, 1)\}, \partial f_2(\bar{x}_1, \bar{x}_2) = \{(0, 1)\}, \partial(-\bar{v}_1 \mu_1 g_1)(\bar{x}_1, \bar{x}_2) = \{(0, 0)\}, \partial(-\bar{v}_2 \mu_2 g_2)(\bar{x}_1, \bar{x}_2) = \{(-1, 0)\}, \partial(\lambda_2 h_2)(\bar{x}_1, \bar{x}_2) = \{(0, -2)\}, \partial(\lambda_2 h_2)(\bar{x}_1, \bar{x}_2) = \{(0, -2)\}, \partial(\mu_1 f_1)(\bar{x}_1, \bar{x}_2) = \{(1, 0)\}, \partial(-\bar{v}_1 \mu_1 g_1)(\bar{x}_1, \bar{x}_2) = \{(0, 0)\}, \partial(-\bar{v}_1 \mu_1 g_1)(\bar{x}_1, \bar{x}_2) = \{(0, 0)\}, \partial(-\bar{v}_2 \mu_2 g_2)(\bar{x}_1, \bar{x}_2) = \{(-1, 0)\}$.

Thus, $\sum_{i=1}^2 \partial(f_i - \bar{v}_i g_i)(\bar{x}_1, \bar{x}_2) + \sum_{i=1}^2 \partial(\lambda_i h_i)(\bar{x}_1, \bar{x}_2) + \sum_{i=1}^2 \partial(\mu_i f_i - \bar{v}_i \mu_i g_i)(\bar{x}_1, \bar{x}_2) = \{(0, 0)\}$ and $\sum_{i=1}^2 (\alpha_i + \beta_i + q_i + z_i) + \sum_{j=1}^2 \gamma_j = 0 = \sum_{i=1}^2 \varepsilon_i (1 + \mu_i) g_i(\bar{x}_1, \bar{x}_2) + \sum_{i=1}^2 \lambda_i h_i(\bar{x}_1, \bar{x}_2)$.

Thus, (iii) of Theorem 3.1 holds.

(2) Let $(\tilde{x}_1, \tilde{x}_2) = (\frac{3}{2}, \frac{15}{4})$. Then $(\tilde{x}_1, \tilde{x}_2)$ is not an ε -efficient solution of $(MFP)_1$, but $(\tilde{x}_1, \tilde{x}_2)$ is a weakly ε -efficient solution of $(MFP)_1$.

Let $C = \bigcup_{\substack{\lambda_1 \geq 0 \\ \lambda_2 \geq 0}} \sum_{i=1}^2 \text{epi}(\lambda_i h_i)^*$. Then

$$C = \text{cone co}\{(-1, 0, -1), (0, -1, -1), (0, 0, 1)\}.$$

Hence, C is closed. Moreover, $f_1(\tilde{x}_1, \tilde{x}_2) - \varepsilon_1 g_1(\tilde{x}_1, \tilde{x}_2) = 1 \geq 0$, and $f_2(\tilde{x}_1, \tilde{x}_2) - \varepsilon_2 g_2(\tilde{x}_1, \tilde{x}_2) = 3 \geq 0$. Let $\bar{v}_1 = \frac{f_1(\tilde{x}_1, \tilde{x}_2)}{g_1(\tilde{x}_1, \tilde{x}_2)} - \varepsilon_1$ and $\bar{v}_2 = \frac{f_2(\tilde{x}_1, \tilde{x}_2)}{g_2(\tilde{x}_1, \tilde{x}_2)} - \varepsilon_2$. Then, $\bar{v}_2 = 2, \tilde{v}_2 = 2$. Let $\mu_1 = 1$ and $\mu_2 = 1$. Then,

$$\begin{aligned} & \sum_{i=1}^2 [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] \\ &= \{(1, 0)\} \times \mathbb{R}_+ + \{(0, 0)\} \times [1, \infty) + \{(0, 0)\} \times \mathbb{R}_+. \end{aligned}$$

Since $(-1, 0, -1) \in C, (0, 0, 0) \in \sum_{i=1}^2 [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*] + C$. So, (ii) of Theorem 3.3 holds. Let $\alpha_1 = \beta_1 = \gamma_1 = \alpha_2 = \beta_2 = \gamma_2 = 0, \lambda_1 = 1$ and $\lambda_2 = 0$. Then,

$$\sum_{i=1}^2 \partial(\mu_i f_i)(\tilde{x}_1, \tilde{x}_2) + \sum_{i=1}^2 \partial(-\bar{v}_i \mu_i g_i)(\tilde{x}_1, \tilde{x}_2) + \sum_{j=1}^2 \partial(\lambda_j h_j)(\tilde{x}_1, \tilde{x}_2) = \{(0, 0)\}$$

and

$$\sum_{i=1}^2 \mu_i \varepsilon_i g_i(\tilde{x}_1, \tilde{x}_2) = \frac{1}{2} = \sum_{i=1}^2 (\alpha_i + \beta_i) + \sum_{j=1}^2 [\gamma_j - (\lambda_j h_j)(\tilde{x}_1, \tilde{x}_2)].$$

Thus, (iii) of Theorem 3.3 holds.

Example 3.2 Consider the following MFP:

$$\begin{aligned} (MFP)_2 \text{ Minimize } & \left(-x_1 + 1, \frac{x_2}{-x_1 + 1}\right) \\ \text{subject to } & [\max\{0, x_1\}]^2 \leq 0, \quad -x_2 + 1 \leq 0. \end{aligned}$$

Let $\varepsilon = (\varepsilon_1, \varepsilon_2) = (\frac{1}{2}, \frac{1}{2})$, and $f_1(x_1, x_2) = -x_1 + 1, g_1(x_1, x_2) = 1, f_2(x_1, x_2) = x_2, g_2(x_1, x_2) = -x_1 + 1, h_1(x_1, x_2) = [\max\{0, x_1\}]^2$ and $h_2(x_1, x_2) = -x_2 + 1$.

(1) Let $(\bar{x}_1, \bar{x}_2) = (-\frac{1}{2}, \frac{9}{4})$. Then, (\bar{x}_1, \bar{x}_2) is an ε -efficient solution of $(MFP)_2$. Let $A = \bigcup_{\substack{\lambda_1 \geq 0 \\ \lambda_2 \geq 0}} \sum_{j=1}^2 \text{epi}(\lambda_j h_j)^* + \bigcup_{\substack{\mu_1 \geq 0 \\ \mu_2 \geq 0}} \sum_{i=1}^2 [\text{epi}(\mu_i f_i)^* + \text{epi}(-\bar{v}_i \mu_i g_i)^*]$. Then, $\text{cl}A = \text{cone co}\{(0, -1), (-1), (1, 0, 0), (-1, 0, 0), (1, 1, 1), (0, 0, 1)\}$. Here, $(1, 0, 0) \in \text{cl}A$, but $(1, 0, 0) \in A$, where $\text{cl}A$ is the closure of the set A . Thus, A is not closed. Let $Q = \{(x_1, x_2) \in \mathbb{R}^n \mid h_1(x_1, x_2) \leq 0, h_2(x_1, x_2) \leq 0\}$. Then, $Q \cap S(\bar{x}_1, \bar{x}_2) = \{(1, 1)\}$. Let $v_i = \frac{f_i(\bar{x}_1, \bar{x}_2)}{g_i(\bar{x}_1, \bar{x}_2)} - \varepsilon_i, i = 1, 2$. Then, $\bar{v}_1 = \bar{v}_2 = 1$. Let $\alpha_1 = \beta_1 = \alpha_2 = \beta_2 = 0, \lambda_1^n = 0, \lambda_2^n = 1, \gamma_1^n = \gamma_2^n = 0, w_1^n = (0, 0), w_2^n = (0, -1)$. Let $u_1 = (-1, 0), u_2 = (0, 1), y_1 = (0, 0)$ and $y_2 = (1, 0)$. Let $q_1^n = q_2^n = z_1^n = z_2^n = 0$, and $\mu_1^n = \mu_2^n = 0$. Let $s_1^n = s_2^n = (0, 0)$ and $t_1^n = t_2^n = \{(0, 0)\}$. Then, $u_i \in \partial f_i(\bar{x}_1, \bar{x}_2), i = 1, 2, \gamma_i \in \partial(-\bar{v}_i g_i)(\bar{x}_1, \bar{x}_2), i = 1, 2, w_j^n \in \partial(\lambda_j^n h_j)(\bar{x}_1, \bar{x}_2), j = 1, 2, s_k^n \in \partial(\mu_k^n f_k)(\bar{x}_1, \bar{x}_2), k = 1, 2$, and $t_k^n \in \partial(-\bar{v}_k \mu_k^n g_k)(\bar{x}_1, \bar{x}_2), k = 1, 2$. Moreover,

$$0 = \sum_{i=1}^2 (u_i + \gamma_i) + \lim_{n \rightarrow \infty} \left[\sum_{j=1}^2 w_j^n + \sum_{i=1}^2 (s_k^n + t_k^n) \right]$$

and

$$\begin{aligned} & \sum_{i=1}^2 \varepsilon_i g_i(\tilde{x}_1, \tilde{x}_2) \\ &= \sum_{i=1}^2 (\alpha_i + \beta_i) + \lim_{n \rightarrow \infty} \sum_{j=1}^2 [\gamma_j^n - (\lambda_j^n h_j)(\tilde{x}_1, \tilde{x}_2)] + \sum_{k=1}^2 [q_k^n + z_k^n - \mu_k^n \varepsilon_k g_k(\tilde{x}_1, \tilde{x}_2)]. \end{aligned}$$

Thus, Theorem 3.2 holds.

(2) Let $(\tilde{x}_1, \tilde{x}_2) = (-\frac{1}{2}, \frac{15}{4})$. Then, $(\tilde{x}_1, \tilde{x}_2)$ is a weakly ε -efficient solution of $(MFP)_2$, but

not an ε -efficient solution of $(MFP)_2$. Let $B = \bigcup_{\substack{\lambda_1 \geq 0, \\ \lambda_2 \geq 0}} \text{epi}(\sum_{i=1}^2 \lambda_i h_i)^*$. Then, $\text{cl}B = \text{cone co} \{(0, -1, -1), (1, 0, 0), (0, 0, 1)\}$. However, $(1, 0, 0) \notin B$. Thus, B is not closed. Moreover, $f_2(\tilde{x}_1, \tilde{x}_2) - \varepsilon_2 g_2(\tilde{x}_1, \tilde{x}_2) = 3 \geq 0$, $f_2(\tilde{x}_1, \tilde{x}_2) - \varepsilon_2 g_2(\tilde{x}_1, \tilde{x}_2) = 3 \geq 0$. Let

$$\tilde{v}_2 = \frac{f_2(\tilde{x}_1, \tilde{x}_2)}{g_2(\tilde{x}_1, \tilde{x}_2)} - \varepsilon_2 \text{ and } \tilde{v}_2 = \frac{f_2(\tilde{x}_1, \tilde{x}_2)}{g_2(\tilde{x}_1, \tilde{x}_2)} - \varepsilon_2. \text{ Then, } \tilde{v}_1 = 1 \text{ and } \tilde{v}_2 = 2. \text{ Let } \mu_1 = 1, \mu_2 = 0,$$

$\alpha_1 = \beta_1 = \alpha_2 = \beta_2 = 0$ and $r_2^n = 0, \lambda_2^n = 0$. Let $\gamma_1^n = \frac{1}{2} + \frac{1}{4n}, \lambda_1^n = n, \gamma_2^n = 0, \lambda_2^n = 0, n \in \mathbb{N}$. Then, $\partial(\mu_1 f_1)(\tilde{x}_1, \tilde{x}_2) = \{(-1, 0)\}, \partial(\mu_2 f_2)(\tilde{x}_1, \tilde{x}_2) = \{(0, 0)\}, \partial(-\tilde{v}_1 \mu_1 g_1)(\tilde{x}_1, \tilde{x}_2) = \{(0, 0)\},$

$$\partial_{\gamma_1^n}(\lambda_1^n h_1)(\tilde{x}_1, \tilde{x}_2) = \left[0, -n + \sqrt{n^2 + 4n(\frac{1}{2} + \frac{1}{4n})} \right] \times \{0\} = [0, 1] \times \{0\},$$

$\partial_{\gamma_2^n}(\lambda_2^n h_2)(\tilde{x}_1, \tilde{x}_2) = \{(0, 0)\}, \partial_{\lambda_2^n}(\lambda_2^n h_2)(\tilde{x}_1, \tilde{x}_2) = \{(0, 0)\}$. Let $u_1 = (-1, 0)$ and $u_2 = y_1 = y_2 = (0, 0)$. Then, $u_1 \in \partial(\mu_1 f_1)(\tilde{x}_1, \tilde{x}_2), u_2 \in \partial(\mu_2 f_2)(\tilde{x}_1, \tilde{x}_2), \gamma_1 \in \partial(-\tilde{v}_1 \mu_1 g_1)(\tilde{x}_1, \tilde{x}_2), \gamma_2 \in \partial(-\tilde{v}_2 \mu_2 g_2)(\tilde{x}_1, \tilde{x}_2)$. Let $w_1^n = (1, 0)$ and $w_2^n = (0, 0)$. Then, $w_1^n \in \partial_{\gamma_1^n}(\lambda_1^n h_1)(\tilde{x}_1, \tilde{x}_2)$ and $w_2^n \in \partial_{\gamma_2^n}(\lambda_2^n h_2)(\tilde{x}_1, \tilde{x}_2)$. Thus, $\sum_{i=1}^2 (u_i + \gamma_i) + \lim_{n \rightarrow \infty} \sum_{j=1}^2 w_j^n = (-1, 0) + (1, 0) = (0, 0),$

$$\lim_{n \rightarrow \infty} \sum_{i=1}^2 \left[\gamma_j^n - (\lambda_j^n h_j)(\tilde{x}_1, \tilde{x}_2) \right] = \lim_{n \rightarrow \infty} \left(\frac{1}{2} + \frac{1}{4n} \right) = \frac{1}{2} \text{ and}$$

$$\lim_{n \rightarrow \infty} \sum_{i=1}^2 \left[\gamma_j^n - (\lambda_j^n h_j)(\tilde{x}_1, \tilde{x}_2) \right] = \lim_{n \rightarrow \infty} \left(\frac{1}{2} + \frac{1}{4n} \right) = \frac{1}{2}. \text{ Hence, Theorem 3.4 holds.}$$

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Authors' contributions

The authors, together discussed and solved the problems in the manuscript. All Authors read and approved the final manuscript.

Competing interests

The authors declare that they have no competing interests.

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