

SUFFICIENCY IN NONSMOOTH MULTIOBJECTIVE FRACTIONAL PROGRAMMING

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ABSTRACT. In this paper, Karush-Kuhn-Tucker type sufficient optimality conditions are obtained for a feasible point of a nonsmooth multiobjective fractional programming problem to be an efficient or properly efficient by using generalized (F, ρ, σ) -type I functions.

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1. Introduction

Consider the following nonsmooth multiobjective programming problem:

$$(NP) \text{ Minimize } f(x) = [f_1(x), f_2(x), \dots, f_k(x)] \\ \text{subject to } x \in X = \{x \in S : g(x) \leq 0\},$$

where $S \subseteq R^n$, the functions $f = (f_1, f_2, \dots, f_k) : S \rightarrow R^k$ and $g = (g_1, g_2, \dots, g_m) : S \rightarrow R^m$ are locally Lipschitz functions.

Zhao [16] obtained Karush-Kuhn-Tucker type sufficient conditions and duality results for a nonsmooth scalar optimization assuming Clarke [4] generalized subgradients under type I functions. Kuk and Tanino [8] considered a nonsmooth multiobjective program (NP) and established sufficient optimality conditions and duality theorems involving generalized type I vector-valued functions. Gulati and Agarwal [7] defined generalized (F, α, ρ, d) -type I functions for (NP) and obtained sufficiency and duality results. In [12], Nobakhtian used the concept of infine functions to establish optimality conditions and duality results for (NP). Ahmad and Sharma [1] introduced a new class of (F, ρ, σ) -type I functions for a nonsmooth multiobjective program and derived optimality conditions and duality theorems. Recently, Nobakhtian [15] introduced generalized (F, ρ) -convexity for (NP) and proved duality results for a mixed type dual.

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In this paper, we consider the following multiobjective fractional programming problem:

$$\begin{aligned} \text{(FP) Minimize } & \left[\frac{f_1(x)}{h_1(x)}, \frac{f_2(x)}{h_2(x)}, \dots, \frac{f_k(x)}{h_k(x)} \right] \\ \text{subject to } & x \in X = \{x \in S : g(x) \leq 0\}, \end{aligned}$$

where the functions $f = (f_1, f_2, \dots, f_k) : S \rightarrow R^k$, $h = (h_1, h_2, \dots, h_k) : S \rightarrow R^k$ and $g = (g_1, g_2, \dots, g_m) : S \rightarrow R^m$ are locally Lipschitz on S . Let $f_i(x) \geq 0$ and $h_i(x) > 0$ for each $i = 1, 2, \dots, k$ and $x \in S$.

We derive sufficient optimality conditions for (FP) by using the concept of generalized (F, ρ, σ) -type I functions. Our results improve the results appeared in [9, 10, 11, 13, 14, 15].

2. Definitions and preliminaries

The following conventions of vectors in R^n will be followed throughout this paper: $x \geq y \Leftrightarrow x_p \geq y_p, p = 1, 2, \dots, n$; $x \geq y \Leftrightarrow x \geq y, x \neq y$; $x > y \Leftrightarrow x_p > y_p, p = 1, 2, \dots, n$. Let $K = \{1, 2, \dots, k\}$, $M = \{1, 2, \dots, m\}$ be the index sets. A function $f : R^n \rightarrow R$ is said to be locally Lipschitz at $\bar{x} \in R^n$, if there exist scalars $\delta > 0$ and $\epsilon > 0$ such that

$$|f(x^1) - f(x^2)| \leq \delta \|x^1 - x^2\|, \text{ for all } x^1, x^2 \in \bar{x} + \epsilon B,$$

where $\bar{x} + \epsilon B$ is the open ball of radius ϵ about \bar{x} .

The generalized directional derivative [4] of a locally Lipschitz function f at x in the direction v , denoted by $f^\circ(x; v)$, is as follows:

$$f^\circ(x; v) = \limsup_{\substack{y \rightarrow x \\ t \downarrow 0}} \frac{f(y + tv) - f(y)}{t}.$$

The generalized gradient [4] of f at x is denoted by

$$\partial f(x) = \{\xi \in R^n : f^\circ(x; v) \geq \xi^t v, \text{ for all } v \in R^n\}.$$

Now consider the following multiobjective optimization problem:

$$\begin{aligned} \text{(MP) Minimize } & f(x) = [f_1(x), f_2(x), \dots, f_k(x)] \\ \text{subject to } & x \in X. \end{aligned}$$

The following definitions are from Geoffrion [6]:

Definition 1. A point $\bar{x} \in X$ is said to be an efficient solution of (MP), if there exists no $x \in X$ such that $f(x) \leq f(\bar{x})$.

Definition 2. A point $\bar{x} \in X$ is said to be a weakly efficient solution of (MP), if there exists no $x \in X$ such that $f(x) < f(\bar{x})$.

Definition 3. An efficient solution \bar{x} of (MP) is said to be properly efficient, if there exists a scalar $N > 0$ such that for each $i, f_i(x) < f_i(\bar{x})$ and $x \in X$ imply that

$$\frac{f_i(\bar{x}) - f_i(x)}{f_j(x) - f_j(\bar{x})} \leq N$$

for at least one j satisfying $f_j(\bar{x}) < f_j(x)$.

Definition 4. A functional $F : S \times S \times R^n \rightarrow R$ is sublinear in its third argument, if for all $x, \bar{x} \in S$,

- (i) $F(x, \bar{x}; a + b) \leq F(x, \bar{x}; a) + F(x, \bar{x}; b)$, for all $a, b \in R^n$,
- (ii) $F(x, \bar{x}; \alpha a) = \alpha F(x, \bar{x}; a)$, for all $\alpha \in R, \alpha \geq 0$ and $a \in R^n$.

We recall the following generalized (F, ρ, σ) -type I functions [1]. Let $f : S \rightarrow R^k$ and $g : S \rightarrow R^m$ be locally Lipschitz at a given point $\bar{x} \in S$, $\rho = (\rho_1, \rho_2, \dots, \rho_k) \in R^k$, $\sigma = (\sigma_1, \sigma_2, \dots, \sigma_m) \in R^m$, and $d(\cdot, \cdot) : S \times S \rightarrow R$. Also, for $\bar{x} \in X$, $J(\bar{x}) = \{j \in M : g_j(\bar{x}) = 0\}$ and g_J will denote the vector of active constraints at \bar{x} .

Definition 5. For each $i \in K$ and $j \in M$, (f_i, g_j) is said to be (F, ρ, σ) -type I at $\bar{x} \in S$, if for all $x \in X$, we have

$$f_i(x) - f_i(\bar{x}) \geq F(x, \bar{x}; \xi_i) + \rho_i d^2(x, \bar{x}), \text{ for all } \xi_i \in \partial f_i(\bar{x}), \quad (2.1)$$

$$-g_j(\bar{x}) \geq F(x, \bar{x}; \eta_j) + \sigma_j d^2(x, \bar{x}), \text{ for all } \eta_j \in \partial g_j(\bar{x}).$$

If (2.1) is a strict inequality, then we say that (f_i, g_j) is (F, ρ, σ) -semistrictly-type I at \bar{x} .

Definition 6. For each $i \in K$ and $j \in M$, (f_i, g_j) is said to be (F, ρ, σ) -prestrictquasi-strictlypseudo-type I at $\bar{x} \in S$, if for all $x \in X$, we have

$$\begin{aligned} f_i(x) < f_i(\bar{x}) &\implies F(x, \bar{x}; \xi_i) \leq -\rho_i d^2(x, \bar{x}), \text{ for all } \xi_i \in \partial f_i(\bar{x}), \\ F(x, \bar{x}; \eta_j) &\geq -\sigma_j d^2(x, \bar{x}) \implies -g_j(\bar{x}) > 0, \text{ for all } \eta_j \in \partial g_j(\bar{x}). \end{aligned}$$

Definition 7. For each $i \in K$ and $j \in M$, (f_i, g_j) is said to be (F, ρ, σ) -pseudoquasi-type I at $\bar{x} \in S$, if for all $x \in X$, we have

$$F(x, \bar{x}; \xi_i) \geq -\rho_i d^2(x, \bar{x}) \implies f_i(x) \geq f_i(\bar{x}), \text{ for all } \xi_i \in \partial f_i(\bar{x}), \quad (2.2)$$

$$-g_j(\bar{x}) \leq 0 \implies F(x, \bar{x}; \eta_j) \leq -\sigma_j d^2(x, \bar{x}), \text{ for all } \eta_j \in \partial g_j(\bar{x}).$$

If (2.2) is satisfied as

$$F(x, \bar{x}; \xi_i) \geq -\rho_i d^2(x, \bar{x}) \implies f_i(x) > f_i(\bar{x}), \text{ for all } \xi_i \in \partial f_i(\bar{x}),$$

then we say that (f_i, g_j) is (F, ρ, σ) -strictly-pseudoquasi-type I at \bar{x} .

In order to derive sufficient optimality conditions, we will invoke the following results. We use Dinkelbach-type [5] approach to get the following auxiliary parametric problem:

$$\begin{aligned} \text{(FP)}^\lambda \text{ Minimize } f(x) &= [f_1(x) - \bar{\lambda}_1 h_1(x), f_2(x) - \bar{\lambda}_2 h_2(x), \dots, f_k(x) - \bar{\lambda}_k h_k(x)] \\ &\text{subject to } x \in X. \end{aligned}$$

where $\bar{\lambda}_i, i \in K$, are parameters. The problem is equivalent to (FP) in the sense that for particular choices of $\bar{\lambda}_i, i \in K$, the two problems have the same set of efficient solutions.

In relation to $(FP)^\lambda$, we consider the following scalar minimization problem on the lines of Geoffrion [6]:

$$(FP)^\mu \text{ Minimize } \sum_{i \in K} \mu_i (f_i(x) - \bar{\lambda}_i h_i(x))$$

subject to $x \in X$.

Lemma 1 ([6]). *If \bar{x} is an optimal solution of $(FP)^\mu$, for some $\mu \in R^k$, with strictly positive components, where $\bar{\lambda}_i = \frac{f_i(\bar{x})}{h_i(\bar{x})}$, $i \in K$, then \bar{x} is a properly efficient solution of (FP) .*

Lemma 2 ([3]). *\bar{x} is an efficient solution of $(FP)^\lambda$, if and only if \bar{x} solves $(FP)_r$, $r \in K$:*

$$(FP)_r \text{ Minimize } f_r(x) - \bar{\lambda}_r h_r(x)$$

subject to $f_i(x) - \bar{\lambda}_i h_i(x) \leq f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})$, for all $i \neq r$,
 $g(x) \leq 0$, $x \in S$.

Proposition 1 (Karush-Kuhn-Tucker type necessary conditions) ([9]). *If \bar{x} is an efficient solution of (FP) , then there exist $\bar{\mu} \in R^k$ and $\bar{\nu} \in R^m$ such that*

$$0 \in \sum_{i \in K} \bar{\mu}_i [\partial f_i(\bar{x}) - \bar{\lambda}_i \partial h_i(\bar{x})] + \sum_{j \in J(\bar{x})} \bar{\nu}_j \partial g_j(\bar{x}),$$

$$\bar{\nu}_j g_j(\bar{x}) = 0, j \in M,$$

$$\bar{\mu} > 0, \bar{\nu} \geq 0,$$

where $\bar{\lambda}_i = \frac{f_i(\bar{x})}{h_i(\bar{x})}$, $i \in K$.

3. Sufficiency

In this section, we obtain sufficient conditions for a feasible point of (FP) to be efficient and properly efficient.

Theorem 1. *Let $\bar{x} \in X$, and let there exist scalars $\bar{\mu}_i > 0, i \in K$ and $\bar{\nu}_j \geq 0, j \in J(\bar{x})$ such that*

$$0 \in \sum_{i \in K} \bar{\mu}_i [\partial f_i(\bar{x}) - \bar{\lambda}_i \partial h_i(\bar{x})] + \sum_{j \in J(\bar{x})} \bar{\nu}_j \partial g_j(\bar{x}), \quad (3.1)$$

$$\bar{\nu}_j g_j(\bar{x}) = 0, j \in M, \quad (3.2)$$

where $\bar{\lambda}_i = \frac{f_i(\bar{x})}{h_i(\bar{x})}$, $i \in K$.

If

- (i) $[f_i - \bar{\lambda}_i h_i, g_j]$, $i \in K, j \in J(\bar{x})$ is (F, ρ, σ) -type I at \bar{x} ; and
- (ii) $\sum_{i \in K} \bar{\mu}_i \rho_i + \sum_{j \in J(\bar{x})} \bar{\nu}_j \sigma_j \geq 0$,

then \bar{x} is a properly efficient solution of (FP) .

Proof. By (3.1) we obtain that there exist $\xi_i \in \partial f_i(\bar{x})$, $\zeta_i \in \partial h_i(\bar{x})$, $i \in K$ and $\eta_j \in \partial g_j(\bar{x})$, $j \in J(\bar{x})$ satisfying

$$\sum_{i \in K} \bar{\mu}_i [\xi_i - \bar{\lambda}_i \zeta_i] + \sum_{j \in J(\bar{x})} \bar{\nu}_j \eta_j = 0. \quad (3.3)$$

From hypothesis (i), we get

$$[f_i(x) - \bar{\lambda}_i h_i(x)] - [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})] \quad (3.4)$$

$$\geq F(x, \bar{x}; \xi_i - \bar{\lambda}_i \zeta_i) + \rho_i d^2(x, \bar{x}) \text{ for all } \xi_i \in \partial f_i(\bar{x}), \zeta_i \in \partial h_i(\bar{x}),$$

$$-g_j(\bar{x}) \geq F(x, \bar{x}; \eta_j) + \sigma_j d^2(x, \bar{x}) \text{ for all } \eta_j \in \partial g_j(\bar{x}). \quad (3.5)$$

On multiplying (3.4) by $\bar{\mu}_i > 0$, $i \in K$, and (3.5) by $\bar{\nu}_j \geq 0$, $j \in J(\bar{x})$, using the sublinearity of F ; and taking summation over i and j , respectively, we get

$$\begin{aligned} & \sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] - \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})] \\ & \geq F(x, \bar{x}; \sum_{i \in K} \bar{\mu}_i (\xi_i - \bar{\lambda}_i \zeta_i)) + \sum_{i \in K} \bar{\mu}_i \rho_i d^2(x, \bar{x}) \text{ for all } \xi_i \in \partial f_i(\bar{x}), \zeta_i \in \partial h_i(\bar{x}), \end{aligned}$$

$$0 \geq - \sum_{j \in J(\bar{x})} \bar{\nu}_j g_j(\bar{x}) \geq F(x, \bar{x}; \sum_{j \in J(\bar{x})} \bar{\nu}_j \eta_j) + \sum_{j \in J(\bar{x})} \bar{\nu}_j \sigma_j d^2(x, \bar{x}) \text{ for all } \eta_j \in \partial g_j(\bar{x}).$$

Combining these inequalities, and using the sublinearity of F , we obtain

$$\begin{aligned} & \sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] - \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})] \\ & \geq F(x, \bar{x}; \sum_{i \in K} \bar{\mu}_i (\xi_i - \bar{\lambda}_i \zeta_i) + \sum_{j \in J(\bar{x})} \bar{\nu}_j \eta_j) + (\sum_{i \in K} \bar{\mu}_i \rho_i + \sum_{j \in J(\bar{x})} \bar{\nu}_j \sigma_j) d^2(x, \bar{x}) \\ & \geq F(x, \bar{x}; \sum_{i \in K} \bar{\mu}_i (\xi_i - \bar{\lambda}_i \zeta_i) + \sum_{j \in J(\bar{x})} \bar{\nu}_j \eta_j), \text{ (by hyp. (ii)),} \end{aligned}$$

which on using (3.3) with the sublinearity of F , yields

$$\sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] - \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})] \geq 0,$$

or

$$\sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] \geq \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})]. \quad (3.6)$$

The inequality (3.6) shows that \bar{x} is an optimal solution of (FP)^k. Hence by Lemma 1, we can conclude that \bar{x} is a properly efficient solution of (FP). \square

Theorem 2. Let $\bar{x} \in X$, and let there exist scalars $\bar{\mu}_i > 0$, $i \in K$ and $\bar{\nu}_j \geq 0$, $j \in J(\bar{x})$ satisfying (3.1) and (3.2). If

(i) $[\sum_{i \in K} \bar{\mu}_i (f_i - \bar{\lambda}_i h_i), \bar{\nu}_j g_j]$ is (F, ρ_1, σ_1) -pseudoquasi-type I at \bar{x} ; and

(ii) $\rho_1 + \sigma_1 \geq 0$,

then \bar{x} is a properly efficient solution of (FP).

Proof. From (3.2), we get

$$\bar{\nu}_j g_j(\bar{x}) = 0, \quad j \in J(\bar{x}), \quad \text{or} \quad - \sum_{j \in J(\bar{x})} \bar{\nu}_j g_j(\bar{x}) \leq 0.$$

Then the second part of hypothesis (i) implies

$$F(x, \bar{x}; \sum_{j \in J(\bar{x})} \bar{\nu}_j \eta_j) + \sigma_1 d^2(x, \bar{x}) \leq 0,$$

which in view of (3.3), hypothesis (ii), and the sublinearity of F , gives

$$F(x, \bar{x}; \sum_{i \in K} \bar{\mu}_i (\xi_i - \bar{\lambda}_i \zeta_i) + \rho_1 d^2(x, \bar{x}) \geq 0.$$

The above inequality along with the first part of assumption (i) yields

$$\sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] \geq \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})],$$

which is precisely (3.6). Hence, \bar{x} is a properly efficient solution of (FP). \square

Theorem 3. Let $\bar{x} \in X$, and let there exist scalars $\bar{\mu}_i > 0, i \in K$ and $\bar{\nu}_j \geq 0, j \in J(\bar{x})$ satisfying (3.1) and (3.2). If

- (i) $[\sum_{i \in K} \bar{\mu}_i (f_i - \bar{\lambda}_i h_i), \bar{\nu}_J g_J]$ is (F, ρ_2, σ_2) -prestrictly quasi-strictly pseudo-type I at \bar{x} ; and
- (ii) $\rho_2 + \sigma_2 \geq 0$,

then \bar{x} is a properly efficient solution of (FP).

Proof. The proof follows on the similar lines of Theorem 2. \square

Remark 1. If we replace $\bar{\mu}_i > 0, i \in K$ by $\bar{\mu}_i \geq 0, i \in K, \sum_{i \in K} \bar{\mu}_i = 1$ in the above theorems and other conditions are imposed on $[\sum_{i \in K} \bar{\mu}_i (f_i - \bar{\lambda}_i h_i), \bar{\nu}_J g_J]$, we get stronger conclusion that \bar{x} is an efficient solution of (FP). The results are shown below:

Theorem 4. Let $\bar{x} \in X$, and let there exist scalars $\bar{\mu}_i \geq 0, i \in K, \sum_{i \in K} \bar{\mu}_i = 1$ and $\bar{\nu}_j \geq 0, j \in J(\bar{x})$ satisfying (3.1) and (3.2). If

- (i) $[\sum_{i \in K} \bar{\mu}_i (f_i - \bar{\lambda}_i h_i), \bar{\nu}_J g_J]$ is (F, ρ_3, σ_3) -semistrictly-type I at \bar{x} ; and
- (ii) $\rho_3 + \sigma_3 \geq 0$,

then \bar{x} is an efficient solution of (FP).

Proof. Suppose to the contrary that \bar{x} is not an efficient solution of (FP), then there exists $x \in X$ such that

$$\left[\frac{f_1(x)}{h_1(x)}, \frac{f_2(x)}{h_2(x)}, \dots, \frac{f_k(x)}{h_k(x)} \right] \leq \left[\frac{f_1(\bar{x})}{h_1(\bar{x})}, \frac{f_2(\bar{x})}{h_2(\bar{x})}, \dots, \frac{f_k(\bar{x})}{h_k(\bar{x})} \right]. \quad (3.7)$$

By hypothesis (i), we have

$$\begin{aligned} & \sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] - \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})] \\ & > F(x, \bar{x}; \sum_{i \in K} \bar{\mu}_i (\xi_i - \bar{\lambda}_i \zeta_i)) + \rho_3 d^2(x, \bar{x}) \text{ for all } \xi_i \in \partial f_i(\bar{x}), \zeta_i \in \partial h_i(\bar{x}), \\ 0 & \geq - \sum_{j \in J(\bar{x})} \bar{\nu}_j g_j(\bar{x}) \geq F(x, \bar{x}; \sum_{j \in J(\bar{x})} \bar{\nu}_j \eta_j) + \sigma_3 d^2(x, \bar{x}) \text{ for all } \eta_j \in \partial g_j(\bar{x}). \end{aligned}$$

Now following the proof of Theorem 1, we reach at

$$\sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] > \sum_{i \in K} \bar{\mu}_i [f_i(\bar{x}) - \bar{\lambda}_i h_i(\bar{x})].$$

As $\bar{\lambda}_i = \frac{f_i(\bar{x})}{h_i(\bar{x})}$, $i \in K$, it follows that

$$\sum_{i \in K} \bar{\mu}_i [f_i(x) - \bar{\lambda}_i h_i(x)] > 0.$$

Since $\bar{\mu}_i \geq 0$, $i \in K$, $\sum_{i \in K} \bar{\mu}_i = 1$, we get

$$(f_1(x) - \bar{\lambda}_1 h_1(x), f_2(x) - \bar{\lambda}_2 h_2(x), \dots, f_k(x) - \bar{\lambda}_k h_k(x)) > 0,$$

which in turn yields

$$\left[\frac{f_1(x)}{h_1(x)}, \frac{f_2(x)}{h_2(x)}, \dots, \frac{f_k(x)}{h_k(x)} \right] > (\bar{\lambda}_1, \bar{\lambda}_2, \dots, \bar{\lambda}_k),$$

or

$$\left[\frac{f_1(x)}{h_1(x)}, \frac{f_2(x)}{h_2(x)}, \dots, \frac{f_k(x)}{h_k(x)} \right] > \left[\frac{f_1(\bar{x})}{h_1(\bar{x})}, \frac{f_2(\bar{x})}{h_2(\bar{x})}, \dots, \frac{f_k(\bar{x})}{h_k(\bar{x})} \right],$$

which is a contradiction to (3.7). Hence \bar{x} is an efficient solution of (FP). \square

The following theorem can be proved along the lines of Theorem 4.

Theorem 5. Let $\bar{x} \in X$, and let there exist scalars $\bar{\mu}_i \geq 0$, $i \in K$, $\sum_{i \in K} \bar{\mu}_i = 1$ and $\bar{\nu}_j \geq 0$, $j \in J(\bar{x})$ satisfying (1) and (2). If

- (i) $[\sum_{i \in K} \bar{\mu}_i (f_i - \bar{\lambda}_i h_i), \bar{\nu}_J g_J]$ is (F, ρ_4, σ_4) -strictly pseudoquasi-type I at \bar{x} ; and
- (ii) $\rho_4 + \sigma_4 \geq 0$,

then \bar{x} is an efficient solution of (FP).

Remark 2. It may be noted that Theorems 4 and 5 also hold for weakly efficient solution of (FP).

REFERENCES

1. I. Ahmad and S Sharma, *Optimality conditions and duality in nonsmooth multiobjective optimization*, J. Nonlinear Convex Anal. **8** (2007), 417-430 .
2. C.R. Bector, S Chandra and I. Husain, *Optimality conditions and subdifferentiable multi-objective fractional programming*, J. Optim. Theory Appl. **79** (1993), 105-125 .
3. V. Chankong and Haimes, *Multiobjective Decision Making: Theory and Methods*, North Holland, New York (1983).
4. F.H. Clarke, *Optimization and Nonsmooth Analysis*, John Wiley and Sons, New York (1983).
5. W. Dinkelbach, *On nonlinear fractional programming*, Manag. Sci. **13** (1967), 492-498 .
6. A.M. Geoffrion, *Proper efficiency and theory of vector maximization*, J. Math. Anal. Appl. **22**, 618-630 (1968).
7. T.R. Gulati and D. Agarwal, *Sufficiency and duality in nonsmooth multiobjective optimization involving generalized (F, α, ρ, d) -type I functions*, Comput. Math. Appl. **52** (2006), 81-94 .
8. H. Kuk and T. Tanino, *Optimality and duality in nonsmooth multiobjective optimization involving generalized type I functions*, Comput. Math. Appl. **45** (2003), 1497-1506 .
9. H. Kuk, G.M. Lee and T. Tanino, *Optimality and duality for nonsmooth multiobjective fractional with generalized invexity*, J. Math. Anal. Appl. **262** (2001), 365-375 .
10. J.C. Liu, *Optimality and duality for multiobjective fractional programming involving nonsmooth (F, ρ) -convex functions*, Optimization **36**, 333-346 .
11. J.C. Liu, *Optimality and duality for multiobjective fractional programming involving nonsmooth pseudoinvex functions*, Optimization **37** (1996), 27-39.
12. S. Nobakhtian, *Infine functions and Nonsmooth multiobjective optimization problems*, Comp. Math. Appl. **51** (2006), 1385-1394 .
13. S. Nobakhtian, *Sufficiency in nonsmooth multiobjective programming involving generalized (F, ρ) -convexity*, J. Optim. Theory Appl. **130** (2006), 361-367 .
14. S. Nobakhtian, *Optimality and duality for nonsmooth multiobjective fractional programming with mixed constraints*, J. Global Optim. **41** (2008), 103-115 .
15. S. Nobakhtian, *Generalized (F, ρ) -convexity and duality in nonsmooth problems of multiobjective optimization*, J. Optim. Theory Appl. **136** (2008), 61-68 .
16. F. Zhao, *On sufficiency of the Kuhn-Tucker conditions in nondifferentiable programming*, Bull. Austral. Math. Soc. **46** (1992), 385-389 .

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