

On Multiple-Objective Optimization with Generalized Univexity

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Submitted by Koichi Mizukami

Received July 1, 1997

A multiple-objective optimization problem involving generalized univex functions is considered. Kuhn–Tucker type sufficient optimality conditions are obtained for a feasible point to be an efficient or properly efficient solution. Mond–Weir type duality results are obtained. Further, a vector-valued Lagrangian is introduced and certain vector saddlepoint results are presented. © 1998 Academic Press

Key Words: univexity; type I function; pseudo-type I function; quasi-type I function; optimality; duality; efficient solutions; properly efficient solutions.

1. INTRODUCTION

Hanson and Mond [5] introduced two new classes of functions called type I and type II functions, which are not only sufficient but also are necessary for optimality in primal and dual problems, respectively. Consider the following nonlinear programming problem,

$$\begin{aligned} & \text{Min } f(x), \\ & \text{subject to } g(x) \leq 0; \end{aligned}$$

$f(x)$ and $g(x)$ are type I objective and constraint functions, respectively, with respect to η at x_0 [5], if there exists a vector function $\eta(x)$, defined for all $x, x_0 \in P = \{x; g(x) \leq 0\}$, such that

$$\begin{aligned} f(x) - f(x_0) & \geq [\nabla f(x_0)]^t \eta(x, x_0), \\ -g(x_0) & \geq [\nabla g(x_0)]^t \eta(x, x_0). \end{aligned}$$

Reuda, Hanson [7] further extended type I functions to pseudo-type I and quasi-type I functions and have obtained sufficient optimality criteria for a nonlinear programming problem involving these functions.

Bector and Singh [1] introduced a new class of functions, called b -vex functions. Optimality and duality results for these functions were proved by Bector, Suneja, and Lalitha [3]. A further generalization was defined by Bector, Suneja, and Gupta [2], called univex functions.

Let X be a nonempty open set in \mathbb{R}^n , $f: X \rightarrow \mathbb{R}$, $\eta: X \times X \rightarrow \mathbb{R}^n$, $\phi: \mathbb{R} \rightarrow \mathbb{R}$, and $b: X \times [0, 1] \rightarrow \mathbb{R}_+$, $b = b(x, u, \lambda)$. If the function f is differentiable then b does not depend on λ [1, 2].

DEFINITION 1.1. A differentiable function f is said to be univex at $x_0 \in X$ with respect to η , ϕ , and b if $\forall x \in X$ we have

$$b(x, x_0)\phi[f(x) - f(x_0)] \geq [\nabla f(x_0)]^t \eta(x, x_0).$$

DEFINITION 1.2. A functional $f: X \rightarrow \mathbb{R}$ is sublinear if $F(x + y) \leq F(x) + F(y) \forall x, y \in X$ and $F(\alpha x) = \alpha F(x) \forall x \in X$ and every nonnegative real number α .

Recently, Rueda, Hanson, and Singh [8] obtained optimality and duality results for several mathematical programs by combining the concepts of type I and univex functions.

In this article, we consider a multiple objective nonlinear programming problem and we obtain optimality and duality results by combining the concepts of type I, type II, pseudo-type I, quasi-type I, quasi-pseudo-type I, pseudo-quasi-type I, strictly pseudo-quasi-type I, and univex functions.

2. OPTIMALITY CRITERIA

Throughout this article we consider the following multiple-objective primal problem,

$$\begin{aligned} \text{(VP)} \quad \text{Min } f(x) &= (f_1(x), f_2(x), \dots, f_p(x)), \quad x \in X \subseteq \mathbb{R}^n, \\ &\text{subject to } g(x) \leq 0, \end{aligned}$$

when $f: X \rightarrow \mathbb{R}^p$ and $g: X \rightarrow \mathbb{R}^m$ are differentiable functions on a set $X \subseteq \mathbb{R}^n$ and minimization means obtaining efficient solution of (VP).

Let $P := \{x: x \in X, g(x) \leq 0\}$. For a feasible point $x^* \in P$, we denote by $I(x^*)$ the set,

$$I(x^*) = \{i: g_i(x^*) = 0\}.$$

A feasible solution x^* for (VP) is efficient for (VP) if and only if there is no other feasible x for (VP) such that, for some $i \in \{1, 2, \dots, p\}$,

$$\begin{aligned} f_i(x) &< f_i(x^*), \\ f_j(x) &\leq f_j(x^*), \quad \forall j \neq i. \end{aligned}$$

An efficient solution x^* for (VP) is properly efficient for (VP) if there exists a scalar $M > 0$ such that, for each i ,

$$\frac{f_i(x^*) - f_i(x)}{f_j(x) - f_j(x^*)} \leq M,$$

for some j such that $f_j(x) > f_j(x^*)$ whenever x is feasible for (VP) and $f_i(x) < f_i(x^*)$.

Let $x, y \in \mathbb{R}^n$. By $x \leq y$, we mean $x_i \leq y_i \quad \forall i$; by $x \leq y$, we mean $x_i \leq y_i \quad \forall i$ and $x_j < y_j$ for at least one j , $1 \leq j \leq n$; by $x < y$, we mean $x_i < y_i \quad \forall i$.

In this section, we obtain sufficient optimality conditions for a feasible solution x^* to be efficient or properly efficient for (VP) in the form of the following theorems.

THEOREM 2.1. *Let x^* be (VP)-feasible. Suppose that there exist η, ϕ_0, b_0 , and $\phi_1, b_1, \lambda_0^* \geq 0, i = 1, 2, \dots, p, \sum_{i=1}^p \lambda_i^* = 1, \mu_i^* \geq 0, i \in I$ such that*

$$\begin{aligned} b_0(x, x^*) \phi_0 \left[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \right] \\ > \sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*), \end{aligned} \tag{2.1}$$

$$\begin{aligned} -b_1(x, x^*) \phi_1 \left[\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \right] \geq \sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)] \eta(x, x^*), \\ \forall (VP)\text{-feasible } x \end{aligned} \tag{2.2}$$

and

$$\sum_{i=1}^p \lambda_i^* \nabla f_i(x^*) + \sum_{i \in I(x^*)} \mu_i^* \nabla g_i(x^*) = 0. \tag{2.3}$$

Further suppose

$$a \leq 0 \Rightarrow \phi_0(a) \leq 0, \tag{2.4}$$

$$\phi_1(a) \leq 0 \Rightarrow a > 0, \tag{2.5}$$

$$b_0(x, x^*) > 0, \quad b_1(x, x^*) \geq 0, \tag{2.6}$$

for all feasible x . Then x^* is an efficient solution for (VP).

Proof. Suppose that x^* is not an efficient solution for (VP). Then, there exists a feasible x for (VP) and an index j such that

$$\begin{aligned} f_j(x) &< f_j(x^*), \\ f_i(x) &\leq f_i(x^*), \quad \forall i \neq j. \end{aligned}$$

These two inequalities lead to

$$0 \geq \sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*).$$

From (2.4) and (2.5) it follows that

$$b_0(x, x^*) \phi_0 \left[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \right] \leq 0.$$

Therefore, by (2.1), we have

$$\sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*) < 0. \quad (2.7)$$

Then, by (2.3), we have

$$\sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)]^T \eta(x, x^*) \geq 0. \quad (2.8)$$

From (2.2) and (2.8), we obtain

$$b_1(x, x^*) \phi_1 \left[\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \right] \leq 0. \quad (2.9)$$

By (2.5), (2.6), and (2.9) it follows that

$$\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) > 0,$$

which is a contradiction to the (VP) feasibility of x^* , because $\mu_i^* \geq 0$, $i \in I$. Therefore, x^* is an efficient solution for (VP). ■

THEOREM 2.2. Let x^* be (VP)-feasible. Suppose that there exist $\lambda_i^* > 0$, $i = 1, 2, \dots, p$, $\mu_i^* \geq 0$, $i \in I(x^*)$, η , b_0 , b_1 , ϕ_0 , and ϕ_1 such that

$$\begin{aligned} \sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*) &\geq 0 \\ \Rightarrow b_0(x, x^*) \phi_0 \left[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \right] &\geq 0, \quad (2.10) \end{aligned}$$

and

$$\begin{aligned}
 & -b_1(x, x^*) \phi_1 \left[\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \right] \leq 0 \\
 & \Rightarrow \sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)]^T \eta(x, x^*) \leq 0, \tag{2.11}
 \end{aligned}$$

for all (VP)-feasible x , and (2.3) of Theorem 2.1 hold. Further, suppose

$$a \geq 0 \Rightarrow \phi_1(a) \geq 0, \tag{2.12}$$

$$\phi_0(a) \geq 0 \Rightarrow a \geq 0, \tag{2.13}$$

$$b_1(x, x^*) > 0, \quad b_0(x, x^*) \geq 0, \tag{2.14}$$

\forall feasible x . Then x^* is a properly efficient solution for (VP).

Proof. Because $g_I(x^*) = 0$, $\mu_i^* \geq 0$, $i \in I(x^*)$, $\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \geq 0$, and $b_1(x, x^*) \geq 0$ and (2.12) and (2.11), we have $\sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)]^T \eta(x, x^*) \leq 0$, which on using (2.3) and (2.10) yields $b_0(x, x^*) \phi_0[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*)] \geq 0$. By (2.13) and (2.14), we get $\sum_{i=1}^p \lambda_i^* f_i(x) \geq \sum_{i=1}^p \lambda_i^* f_i(x^*)$. Therefore, by Theorem 1 of Geoffrion [4], x^* is a properly efficient solution for (VP). ■

THEOREM 2.3. Let x^* be (VP)-feasible. Suppose that there exist $\lambda_i^* \geq 0$, $i = 1, 2, \dots, p$, $\sum_{i=1}^p \lambda_i^* = 1$, $\mu_i^* \geq 0$, $i \in I(x^*)$, η , b_0 , b_1 , ϕ_0 , and ϕ_1 such that

$$\begin{aligned}
 & \sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*) \geq 0 \\
 & \Rightarrow b_0(x, x^*) \phi_0 \left[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \right] > 0, \tag{2.15}
 \end{aligned}$$

or equivalently,

$$\begin{aligned}
 & b_0(x, x^*) \phi_0 \left[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \right] \leq 0 \\
 & \Rightarrow \sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*) < 0. \tag{2.16}
 \end{aligned}$$

and

$$\begin{aligned}
 & -b_1(x, x^*) \phi_1 \left[\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \right] \leq 0 \\
 & \Rightarrow \sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)]^T \eta(x, x^*) \leq 0, \tag{2.17}
 \end{aligned}$$

for all (VP)-feasible x , and (2.8) of Theorem 2.1 hold. Further, suppose,

$$a \leq 0 \quad \Rightarrow \quad \phi_0(a) \leq 0, \quad (2.18)$$

$$a \geq 0 \quad \Rightarrow \quad \phi_1(a) \geq 0, \quad (2.19)$$

$$b_0(x, x^*) > 0, \quad b_1(x, x^*) \geq 0. \quad (2.20)$$

Then, x^* is an efficient solution for (VP).

Proof. Suppose that x^* is not efficient for (VP). Then, there exist a feasible x for (VP) such that $\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \leq 0$, which on using (2.18), (2.20), and (2.16) yields,

$$\sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*) < 0. \quad (2.21)$$

Because $\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \geq 0$, by (2.20), (2.19), and (2.17), we have

$$\sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)]^T \eta(x, x^*) \leq 0. \quad (2.22)$$

Now, on adding (2.21) and (2.22), we obtain a contradiction to (2.3). Hence, x^* is an efficient solution for (VP). ■

Remark 2.1. Proceeding along similar lines as in Theorem 2.2, it can be easily seen that x^* becomes properly efficient for (VP) in the preceding theorem if $\lambda_i > 0, \forall i = 1, 2, \dots, p$.

THEOREM 2.4. Let x^* be (VP)-feasible. Suppose that there exist, $\lambda_i^* \geq 0, i = 1, 2, \dots, p, \sum_{i=1}^p \lambda_i^* = 1, \mu_i^* \geq 0, i \in I(x^*), b_0, b_1, \phi_0, \phi_1$, and η such that (2.3) of Theorem 2.1 holds, and if $I(x^*) \neq \emptyset$ and

$$\begin{aligned} b_0(x, x^*) \phi_0 \left[\sum_{i=1}^p \lambda_i^* f_i(x) - \sum_{i=1}^p \lambda_i^* f_i(x^*) \right] &\leq 0 \\ \Rightarrow \sum_{i=1}^p \lambda_i^* [\nabla f_i(x^*)]^T \eta(x, x^*) &\leq 0, \end{aligned} \quad (2.23)$$

and

$$\begin{aligned} \sum_{i \in I(x^*)} \mu_i^* [\nabla g_i(x^*)]^T \eta(x, x^*) &\geq 0 \\ \Rightarrow -b_1(x, x^*) \phi_1 \left[\sum_{i \in I(x^*)} \mu_i^* g_i(x^*) \right] &> 0, \end{aligned} \quad (2.24)$$

for all feasible x . Further suppose,

$$a \leq 0 \Rightarrow \phi_0(a) \leq 0, \tag{2.25}$$

$$\phi_1(a) < 0 \Rightarrow a > 0, \tag{2.26}$$

$$b_0(x, x^*) > 0, \quad b_1(x, x^*) \geq 0. \tag{2.27}$$

Then x^* is an efficient solution for (VP).

Proof. The proof is similar to the proof of Theorem 2.3 ■

Remark 2.2. Proceeding along similar lines as in Theorem 2.2, it can be easily seen that x^* becomes properly efficient for (VP) in the previous theorem if $\lambda_i^* > 0, \forall i = 1, 2, \dots, p$.

3. DUALITY

In this section we consider the Mond–Weir type dual and generalize duality results of Rueda, Hanson, and Singh [8] as well as Kaul, Suneja, and Srivastava [6] under weaker univexity assumptions.

Consider the following Mond–Weir type dual of (VP),

$$(VD) \quad \text{Max } f(u),$$

$$\text{subject to } \sum_{i=1}^p \lambda_i \nabla f_i(u) + \sum_{j=1}^m \mu_j \nabla g_j(u) = 0, \tag{3.1}$$

$$\sum_{j=1}^m \mu_j g_j(u) \geq 0, \tag{3.2}$$

$$\mu_j \geq 0, \quad j = 1, 2, \dots, m, \tag{3.3}$$

$$\lambda_i \geq 0, \quad i = 1, 2, \dots, p,$$

$$\sum_{i=1}^p \lambda_i = 1, \tag{3.4}$$

where $e = (1, 1, \dots, 1) \in \mathbb{R}^p$.

Assuming (f_i, g) to type I, semistrictly type I pseudo-quasi-type I etc. for the same η , Kaul, Suneja, and Srivastava [6] established various duality results for (VP) and (VD). We shall generalize various duality results for (VP) and (VD) by combining univex and type I and its generalizations, these generalizing results of Rueda, Hanson, and Singh [8] and as a byproduct the results of Kaul, Suneja, and Srivastava [6].

THEOREM 3.1 (Weak Duality). *Let x be feasible for (VP) and let the triplet (u, λ, μ) be feasible for (VD). Let for $i = 1, 2, \dots, p$, $\lambda_i > 0$, $\mu_j \geq 0$, $j = 1, 2, \dots, m$, $\eta, b_0, b_1, \phi_0, \phi_1$ such that*

$$b_0(x, u) \phi_0 \left[\sum_{i=1}^p \lambda_i f_i(x) - \sum_{i=1}^p \lambda_i f_i(u) \right] \geq \sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u), \quad (3.5)$$

and

$$-b_1(x, u) \phi_1 \left[\sum_{j=1}^m \mu_j g_j(u) \right] \geq \sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u), \quad (3.6)$$

at u over P ;

Further suppose,

$$a \leq 0 \quad \Rightarrow \quad \phi_0(a) \leq 0, \quad (3.7)$$

$$\phi_1(a) \leq 0 \quad \Rightarrow \quad a > 0, \quad (3.8)$$

$$b_0(x, u) > 0, \quad b_1(x, u) \geq 0. \quad (3.9)$$

Then $f(x) \not\leq f(u)$.

Proof. The proof is similar to the proof of Theorem 2.1 ■

THEOREM 3.2 (Weak Duality). *Let x be feasible for (VP) and let the triplet (u, λ, μ) be feasible for (VD). Let either (a) or (b) Hold:*

(a) for $i = 1, 2, \dots, p$, $\lambda_i > 0$, and $j = 1, 2, \dots, m$, $\mu_j \geq 0$ and there exist η, b_0, b_1, ϕ_0 , and ϕ_1 such that

$$\begin{aligned} & \sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) \geq 0 \\ \Rightarrow & b_0(x, u) \phi_0 \left[\sum_{i=1}^p \lambda_i f_i(x) - \sum_{i=1}^p \lambda_i f_i(u) \right] \geq 0, \quad (3.10) \end{aligned}$$

(or equivalently,

$$\begin{aligned} & b_0(x, u) \phi_0 \left[\sum_{i=1}^p \lambda_i f_i(x) - \sum_{i=1}^p \lambda_i f_i(u) \right] < 0 \\ \Rightarrow & \sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) < 0, \end{aligned}$$

and

$$\begin{aligned}
 & -b_1(x, u) \phi_1 \left[\sum_{j=1}^m \mu_j g_j(u) \right] \leq \mathbf{0} \\
 & \Rightarrow \sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) \leq \mathbf{0}, \tag{3.11}
 \end{aligned}$$

at u over p ;

Further, suppose

$$a < \mathbf{0} \Rightarrow \phi_0(a) < \mathbf{0}, \tag{3.12}$$

$$a \geq \mathbf{0} \Rightarrow \phi_1(a) \geq \mathbf{0}, \tag{3.13}$$

$$b_0(x, u) > \mathbf{0}, \quad b_1(x, u) \geq \mathbf{0}. \tag{3.14}$$

(b) for $i = 1, 2, \dots, p$, $\lambda_i > \mathbf{0}$, and $j = 1, 2, \dots, m$, $\mu_j \geq \mathbf{0}$, and there exist η , b_0 , b_1 , ϕ_0 , and ϕ_1 such that

$$\begin{aligned}
 & b_0(x, u) \phi_0 \left[\sum_{i=1}^p \lambda_i f_i(x) - \sum_{i=1}^p \lambda_i f_i(u) \right] \leq \mathbf{0} \\
 & \Rightarrow \sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) \leq \mathbf{0}, \tag{3.15}
 \end{aligned}$$

and

$$\begin{aligned}
 & \sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) \geq \mathbf{0} \\
 & \Rightarrow -b_1(x, u) \phi_1 \left[\sum_{j=1}^m \mu_j g_j(u) \right] > \mathbf{0}, \tag{3.16}
 \end{aligned}$$

(or equivalently,

$$\begin{aligned}
 & -b_1(x, u) \phi_1 \left[\sum_{j=1}^m \mu_j g_j(u) \right] \leq \mathbf{0} \\
 & \Rightarrow \sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) < \mathbf{0}),
 \end{aligned}$$

at u over P .

Further suppose,

$$a \leq 0 \Rightarrow \phi_0(a) \leq 0, \quad (3.17)$$

$$a \geq 0 \Rightarrow \phi_1(a) \geq 0, \quad (3.18)$$

$$b_0(x, u) \geq 0, \quad b_1(x, u) \geq 0. \quad (3.19)$$

Then $f(x) \not\leq f(u)$.

Proof. (a) If possible, let $f(x) \leq f(u)$. Then, there exists an index i_0 such that

$$\begin{aligned} f_{i_0}(x) &< f_{i_0}(u), \\ f_i(x) &\leq f_i(u), \quad \forall i \neq i_0. \end{aligned}$$

Because $\lambda_i > 0$, $i = 1, 2, \dots, p$, the foregoing inequalities yield

$$\sum_{i=1}^p \lambda_i f_i(x) < \sum_{i=1}^p \lambda_i f_i(u).$$

On using (3.12) and (3.14), from the earlier strict inequality, we have

$$b_0(x, u) \phi_0 \left[\sum_{i=1}^p \lambda_i f_i(x) - \sum_{j=1}^p \lambda_j f_j(u) \right] < 0.$$

Using (3.10) from the previous strict inequality, we have

$$\sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) < 0, \quad \forall x \in P. \quad (3.20)$$

Using (3.2), (3.13), and (3.14), we obtain

$$-b_1(x, u) \phi_1 \left[\sum_{j=1}^m \mu_j g_j(u) \right] \leq 0.$$

Now, by using (3.11), the foregoing inequality yields

$$\sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) \leq 0. \quad (3.21)$$

On adding (3.20) and (3.21), we obtain

$$\sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) + \sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) < 0,$$

which contradicts (3.1). Hence,

$$f(x) \not\leq f(u).$$

(b) If possible, let $f(x) \leq f(u)$. Then, there exists an index i_0 such that

$$\begin{aligned} f_{i_0}(x) &< f_{i_0}(u), \\ f_i(x) &\leq f_i(u), \quad \forall i \neq i_0. \end{aligned}$$

Now $\lambda_i \geq 0$, and $\sum_{i=1}^p \lambda_i = 1$, therefore, from the preceding inequalities, we get

$$\sum_{i=1}^p \lambda_i f_i(x) \leq \sum_{i=1}^p \lambda_i f_i(u),$$

i.e.,

$$\sum_{i=1}^p \lambda_i f_i(x) - \sum_{i=1}^p \lambda_i f_i(u) \leq 0.$$

Then by (3.17), (3.19), and the previous inequality, we have

$$b_0(x, u) \phi_0 \left[\sum_{i=1}^p \lambda_i f_i(x) - \sum_{i=1}^p \lambda_i f_i(u) \right] \leq 0.$$

Using (3.15), the foregoing inequality yields

$$\sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) \leq 0. \tag{3.22}$$

Again, from (3.2), (3.18), and (3.19), we obtain

$$-b_1(x, u) \phi_1 \left[\sum_{j=1}^m \mu_j g_j(u) \right] \leq 0.$$

Using (3.16), we get

$$\sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) < 0. \tag{3.23}$$

Now, on adding (3.22) and (3.23), we obtain,

$$\sum_{i=1}^p \lambda_i [\nabla f_i(u)]^T \eta(x, u) + \sum_{j=1}^m \mu_j [\nabla g_j(u)]^T \eta(x, u) < 0,$$

which contradicts (3.1). Hence,

$$f(x) \not\leq f(u).$$

■

THEOREM 3.3 (Strong Duality). *If x^* is a properly efficient solution for (VP) at which a constraint qualification is satisfied, then there exists $(\lambda^*, \mu^*) \in \mathbb{R}^p \times \mathbb{R}^m$ such that (x^*, λ^*, μ^*) is (VD)-feasible and the values of the objective functions for (VP) and (VD) are equal at x^* and (x^*, λ^*, μ^*) , respectively. Furthermore, if for all (VP)-feasible x and (VD)-feasible (u, λ, μ) the hypotheses of Theorem 3.2(a) are satisfied, then (x^*, λ^*, μ^*) is properly efficient for (VD).*

Proof. Because a constraint qualification is satisfied at x^* then there exist scalars $\lambda_i^* \geq 0$, $i = 1, 2, \dots, p$, $\sum_{i=1}^p \lambda_i^* = 1$, $\mu_j^* \geq 0$, $j = 1, 2, \dots, m$ such that

$$\sum_{i=1}^p \lambda_i^* \nabla f_i(x^*) + \sum_{j=1}^m \mu_j^* \nabla g_j(x^*) = 0, \quad (3.24)$$

$$\sum_{j=1}^m \mu_j^* g_j(x^*) = 0. \quad (3.25)$$

Therefore (x^*, λ^*, μ^*) is feasible for (VD).

Now, the proper efficiency of (x^*, λ^*, μ^*) follows as in Theorem 4 of Weir [9] by using part (a) of the weak duality Theorem 3.2. ■

THEOREM 3.4. *Let x^* be an efficient solution for (VP) at which the Kuhn–Tucker constraint qualification of (VD), $(\sum_{i=1}^p \lambda_i f_i, \sum_{j=1}^m \mu_j g_j)$ satisfies strict inequalities (3.10) and (3.11) at u over P , then there exists $(\lambda^*, \mu^*) \in \mathbb{R}^p \times \mathbb{R}^m$ such that (x^*, λ^*, μ^*) is efficient for (VD) and the objective function values of (VP) and (VD) are equal.*

Proof. Because the Kuhn–Tucker constraint qualification is satisfied at x^* then there exist scalars $\lambda_i^* > 0$, $i = 1, 2, \dots, p$ and $\mu_i^* \geq 0$, $i \in I(x^*)$, such that (3.24) and (3.25) hold. The scalars $\lambda_i^* > 0$ may be normalized according to $\sum_{i=1}^p \lambda_i^* = 1$. Setting $\mu_i^* = 0$, $i \notin I(x^*)$, then gives that the triplet (x^*, λ^*, μ^*) is not efficient, then there exist a feasible (u, λ, μ) for (VD) and an index i_0 such that

$$f_{i_0}(x^*) < f_{i_0}(u),$$

$$f_i(x) \leq f_i(u), \quad \forall i \neq i_0.$$

On using (3.25), we obtain a contradiction to part (b) of the weak duality Theorem 3.2 for feasible solutions x^* for (VP) and (u, λ, μ) for (VD). Hence, (x^*, λ^*, μ^*) is efficient for (VD). ■

THEOREM 3.5. *Suppose that there exists a feasible x^* for (VP) and (x^*, λ^*, μ^*) for (VD) such that*

$$f_i(x^*) = f_i(u^*), \quad \forall i = 1, 2, \dots, p. \quad (3.26)$$

If $\lambda_i^* > 0$, for $i = 1, 2, \dots, p$ and $(\sum_{i=1}^p \lambda_i^* f_i, \sum_{j=1}^m \mu_j g_j)$ satisfy (3.10) and (3.11) at u^* over P , then x^* is properly efficient for (VP). Also if for each feasible (u, λ, μ) of (VD), $(\sum_{i=1}^p \lambda_i^* f_i, \sum_{j=1}^m \mu_j^* g_j)$ satisfy (3.10) and (3.11) at u over P and suppose (3.12)–(3.14) hold, then (x^*, λ^*, μ^*) is properly efficient for (VD).

Proof. The proof of this theorem is similar to that of Theorem 4.5 of Kaul, Suneja, and Srivastava [6]. ■

4. VECTOR LAGRANGIAN AND SADDLE POINT ANALYSIS

In this section we give as a consequence of Theorem 2.1, a Lagrange multipliers theorem and consider saddle point of the vector Lagrangian function.

THEOREM 4.1. *If Theorem 2.1 holds, then equivalent multiobjective programming problem (EVP) for (VP) is given by*

$$\begin{aligned} \text{(EVP)} \quad & V\text{-Minimize } (f_1(x) + \mu^T g(x), \dots, f_p(x) + \mu^T g(x)) \\ & \text{subject to } \mu_j g_j(x^*) = 0, \quad j = 1, 2, \dots, m, \\ & \mu_j \geq 0, \quad j = 1, 2, \dots, m. \end{aligned}$$

Proof. Let x^* be an efficient solution of (VP), from the (KT) optimality conditions, we have

$$\sum_{i=1}^p \lambda_i \nabla f_i(x^*) + \sum_{j=1}^m \mu_j \nabla g_j(x^*) = 0, \tag{4.1}$$

and

$$\mu_j g_j(x^*) = 0, \quad j = 1, 2, \dots, m. \tag{4.2}$$

Using (4.2) in (4.1), we get

$$\sum_{i=1}^p \lambda_i [\nabla f_i(x^*) + \mu^T g(x^*)] + \sum_{j=1}^m \mu_j \nabla g_j(x^*) = 0. \tag{4.3}$$

If (x^*, λ^*, μ^*) is not properly efficient for (VD), then there exists a feasible (u, λ, μ) of (VD) and an index i such that

$$f_i(u) - f_i(u^*) > M(f_j(u^*) - f_j(u)),$$

$\forall M > 0$ and $\forall j$ such that

$$f_j(u^*) > f_j(u),$$

whenever

$$f_i(u^*) < f_i(u).$$

On using (3.26), we get

$$f_i(u) - f_i(x^*) > M(f_j(x^*) - f_j(u)),$$

$\forall M > 0$ and $\forall j$ such that

$$f_j(x^*) > f_j(u),$$

whenever

$$f_i(x^*) < f_i(u).$$

Now

$$\sum_{i=1}^p \lambda_i^* [f_i(x^*) - f_i(u)] < 0,$$

which contradicts the weak duality for feasible solutions x^* of (VP) and feasible solution (u, λ, μ) of (VD) by Theorem 3.2(a). Thus, (x^*, λ^*, μ^*) is properly efficient for (VD). Now, applying the arguments of Theorem 2.1 by replacing f_i by $f_i(\cdot) + \mu^T G(\cdot)$ yields the result. ■

We now introduce the vector valued Lagrangian function and we study its saddle point. Theorem 4.1 suggests the vector valued Lagrangian function $L(x, \mu)$ as $L: X \times \mathbb{R}_+^m \rightarrow \mathbb{R}^p$ given by

$$L(x, \mu) = (L_1(x, \mu), L_2(x, \mu), \dots, L_p(x, \mu)),$$

where

$$L_i(x, \mu) = f_i(x) + \mu^T g(x), \quad i = 1, 2, \dots, p.$$

DEFINITION 4.1. A point $(x^*, \mu^*) \in X \times \mathbb{R}_+^m$ is said to be a vector saddle point of the vector valued Lagrangian function $L(x, \mu)$ if it satisfies the following conditions,

$$L(x^*, \mu) \not\preceq L(x^*, \mu^*), \quad \forall \mu \in \mathbb{R}_+^m, \quad (4.4)$$

$$L(x^*, \mu^*) \not\preceq L(x, \mu^*), \quad \forall x \in X. \quad (4.5)$$

THEOREM 4.2. If (x^*, μ^*) is a vector saddle point of $L(x, \mu)$, then x^* is a properly efficient solution of (VP).

Proof. Because (x^*, μ^*) is a vector saddle point of $L(x, \mu)$, therefore, (4.4) implies that $L_i(x^*, \mu) \leq L_i(x^*, \mu^*)$ for at least one $i = 1, 2, \dots, p$, $\forall \mu \in \mathbb{R}_+^m$,

$$\begin{aligned} \Rightarrow f_i(x^*) + \mu^T g(x^*) &\leq f_i(x^*) + \mu^{*T} g(x^*), \\ &\text{for at least one } i \text{ and } \forall \mu \in \mathbb{R}_+^m, \\ \Rightarrow (\mu - \mu^*)^T g(x^*) &\leq 0, \quad \forall \mu \in \mathbb{R}_+^m. \end{aligned} \tag{4.6}$$

For any $j = 1, 2, \dots, m$, set

$$\begin{aligned} \mu_k &= \bar{\mu}_k, \quad \text{for } k = 1, 2, \dots, j - 1, j + 1, \dots, m, \\ \mu_j &= \bar{\mu}_{j+1}. \end{aligned}$$

From which it follows that

$$g_j(x^*) \leq 0.$$

Repeating this process for $j = 1, 2, \dots, m$, we have

$$g(x^*) \leq 0.$$

Hence, x^* is feasible for (VP). Again, because $\mu^* \in \mathbb{R}_+^m$ and $g(x^*) \leq 0$, we have $\mu^{*T} g(x^*) \leq 0$. But from (4.6), we have by setting $\mu = 0$, that $\mu^{*T} g(x^*) \geq 0$. Thus, $\mu^{*T} g(x^*) \leq 0$ and $\mu^{*T} g(x^*) \geq 0$ yield

$$\mu^{*T} g(x^*) = 0. \tag{4.7}$$

Now, we assume that x^* is not an efficient solution of the problem (VP). Therefore, there exists feasible x with $g(x) \leq 0$, such that

$$f_i(x) \leq f_i(x^*), \quad \forall i = 1, 2, \dots, p,$$

and

$$f_{i_0}(x) < f_{i_0}(x^*), \quad \text{for at least one } i_0 \in \{1, 2, \dots, p\}.$$

These along with (4.2) and (4.7) yield

$$\begin{aligned} f_i(x) + \mu^{*T} g(x) &\leq f_i(x^*) + \mu^{*T} g(x^*), \\ &\forall i = 1, 2, \dots, p \text{ and } \forall x \in X, \end{aligned}$$

and

$$f_{i_0}(x) + \mu^{*T}g(x) < f_{i_0}(x^*) + \mu^{*T}g(x^*)$$

for at least one $i_0 \in \{1, 2, \dots, p\}$ and $\forall x \in X$.

That is

$$L_i(x, \mu^*) \leq L_i(x^*, \mu^*), \quad \forall i = 1, 2, \dots, p \text{ and } \forall x \in X,$$

$$L_{i_0}(x, \mu^*) < L_{i_0}(x^*, \mu^*),$$

for at least one $i_0 \in \{1, 2, \dots, p\}$ and $\forall x \in X$.

which is a contradiction to (4.5). Hence, x^* is an efficient solution of (VP).

We now suppose that x^* is not a properly efficient solution of (VP). Therefore, there exists a feasible point x for (VP) and an index i_0 such that for every positive function $M > 0$, we have

$$f_i(x) - f_i(x^*) > M(f_j(x^*) - f_j(x)),$$

for all j satisfying

$$f_j(x) > f_j(x^*),$$

wherever

$$f_i(x) < f_i(x^*).$$

This along with (4.2) and (4.7) yields

$$f_i(x) + \mu^{*T}g(x) < f_i(x^*) + \mu^{*T}g(x^*),$$

$\forall i = 1, 2, \dots, p \text{ and } x \in X,$

i.e., $L_i(x, \mu^*) < L_i(x^*, \mu^*)$, which is a contradiction to (4.5). Hence, x^* is a properly efficient solution of (VP).

Similarly by assuming (4.5) we can get a contradiction to (4.4). ■

THEOREM 4.3. *Let x^* be a properly efficient solution of (VP) and let an x^* slater type constraint qualification be satisfied. If $(\sum_{i=1}^p \lambda_i f_i, \sum_{j=1}^m \mu_j g_j)$ satisfy (2.1) and (2.2) and (2.4)–(2.6) hold, then there exists $\mu^* \in \mathbb{R}_+^m$ such that (x^*, μ^*) is a vector saddle point of $L(x, \mu)$.*

Proof. Because x^* is a properly efficient solution of (VP), therefore, x^* is also an efficient solution of (VP) and because at x^* , slater type

constraint qualification is satisfied, therefore, by (KT) conditions, there exist $\lambda > 0$, $\lambda \in \mathbb{R}^p$, $\mu^* \in \mathbb{R}_+^m$, such that the following hold

$$\sum_{i=1}^p \lambda_i \nabla f_i(x^*) + \sum_{j=1}^m \mu_j^* \nabla g_j(x^*) = 0, \tag{4.8}$$

$$\mu_j^* g_j(x^*) = 0, \quad j = 1, 2, \dots, m. \tag{4.9}$$

Now for all $i = 1, 2, \dots, p$, $\forall x \in X$, we have

$$\begin{aligned} L_i(x, \mu^*) - L_i(x^*, \mu^*) &= f_i(x) - f_i(x^*) + \mu^{*T} [g(x) - g(x^*)] \\ &\geq -\eta(x, x^*)^T \left[\sum_{i=1}^p \mu^{*T} \nabla g_j(x^*) \right] \\ &\geq 0, \text{ (by (2.2)).} \end{aligned}$$

Because $\lambda_i \in \mathbb{R}^p$, $\lambda > 0$, therefore,

$$L_i(x^*, \mu^*) \not\leq L_i(x, \mu^*), \quad \forall x \in X.$$

The other part,

$$L_i(x^*, \mu) \not\leq L_i(x, \mu^*), \quad \forall x \in \mathbb{R}_+^m,$$

of the vector saddle point inequality follows from

$$L(x^*, \mu) - L(x^*, \mu^*) = (\mu - \mu^*)^T g(x^*) \leq 0.$$

Hence (x^*, μ^*) is a vector saddle point of $L(x, \mu)$. ■

Remark 4.1. Theorem 4.3 can be established under weaker assumptions used in previous sections.

ACKNOWLEDGMENTS

The author acknowledges his indebtedness to Dr. Norma Rueda for providing [8] in preprint form. The author thanks the Department of Applied Mathematics, Institute of Technology, Banaras Hindu University, Varanasi, for providing academic atmosphere during the preparation of this article. The author also thanks an anonymous referee for his suggestions and his careful reading of an earlier version of the manuscript.

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