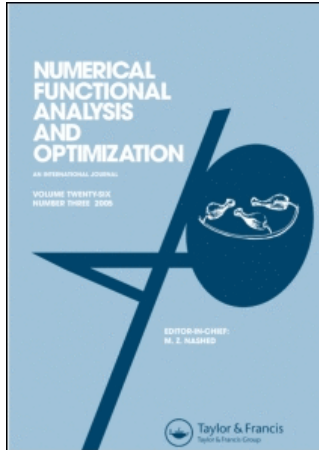


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SECOND-ORDER DUALITY FOR NONDIFFERENTIABLE MULTIOBJECTIVE PROGRAMMING PROBLEMS

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□ *This paper is concerned with second-order duality for a class of nondifferentiable multiobjective programming problems. Usual duality theorems are proved for Mangasarian type and general Mond–Weir type vector duals under generalized bonvexity assumptions.*

Keywords Generalized bonvexity; Nondifferentiable multiobjective programming; Properly efficient solution; Second-order duality.

AMS Subject Classification 90C29; 90C30; 90C46.

1. INTRODUCTION

Preda [13] introduced the notion of (F, ρ) -convexity, an extension of F -convexity defined by Hanson and Mond [7] and ρ -convexity introduced by Vial [14], and he used this concept to obtain multiobjective duality results for efficient solutions. Gulati and Islam [5] derived sufficiency and duality theorems for efficient and properly efficient solutions of a multiobjective nonlinear programming problem under the assumptions taken by Hanson and Mond [7]. In [2], Ahmad obtained a number of sufficiency theorems for efficient and properly efficient solutions of a multiobjective programming problem under various generalized convexity assumptions, and he discussed duality results also for a general Mond–Weir type dual.

Second-order duality was first introduced by Mangasarian [8] for a nonlinear programming problem, which involves second-order derivatives of the objective and the constraint functions. He established duality results involving somewhat complicated assumptions. Mond [11] reproved the second-order duality results under different and less restricted assumptions

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than those previously considered in [8]. Zhang and Mond [16] discussed duality results for nondifferentiable programs under generalized invexity.

Mishra [9] formulated a second-order Mond–Weir type multiobjective dual and derived weak and strong duality theorems under generalized type I functions. Aghezzaf [1] formulated a second-order multiobjective mixed type dual and obtained various duality results involving a new class of generalized second-order (F, ρ) -convex functions. In [6], Hachimi and Aghezzaf proposed a new class of second-order generalized type I functions and established multiobjective duality results for a mixed type dual. Recently, Ahmad and Husain [3] studied a Mond–Weir type multiobjective dual and derived duality results by defining second-order (F, α, ρ, d) -convex function and its generalizations.

In this paper, we consider the following nondifferentiable vector optimization problem:

$$\begin{aligned} &\text{Minimize} && \left(f_1(x) + (x^t B_1 x)^{\frac{1}{2}}, f_2(x) + (x^t B_2 x)^{\frac{1}{2}}, \dots, f_k(x) + (x^t B_k x)^{\frac{1}{2}} \right) \quad (\text{P}) \\ &\text{subject to} && x \in S = \{x \in X : g(x) \leq 0\}, \end{aligned}$$

where X is an open subset of R^n , $f_j : X \rightarrow R$, $j = 1, 2, \dots, k$, $g : X \rightarrow R^m$ and B_j , $j = 1, 2, \dots, k$ is an $n \times n$ positive semidefinite symmetric matrix. We formulate Mangasarian type and general Mond–Weir type second-order duals for (P) and prove weak, strong, and strict converse duality theorems under bonvexity/generalized bonvexity.

2. NOTATIONS AND PRELIMINARIES

The following conventions for vectors $u, v \in R^n$ will be used: $u \geq v \Leftrightarrow u_i \geq v_i$, $i = 1, 2, \dots, n$; $u > v \Leftrightarrow u_i > v_i$, $i = 1, 2, \dots, n$, but $u \neq v$; $u \geq v \Leftrightarrow u_i \geq v_i$, $i = 1, 2, \dots, n$, but $u \neq v$; $u > v \Leftrightarrow u_i > v_i$, $i = 1, 2, \dots, n$. The index sets are $K = \{1, 2, \dots, k\}$ and $M = \{1, 2, \dots, m\}$. For each $j \in K$, $K_j = K - \{j\}$.

Consider the following vector optimization problem:

$$\begin{aligned} &\text{Minimize} && f(x) = [f_1(x), f_2(x), \dots, f_k(x)] \\ &\text{subject to} && x \in S = \{x \in X : g(x) \leq 0\}. \end{aligned} \quad (\text{VOP})$$

Definition 2.1. A point $x^\circ \in S$ is said to be an efficient solution of (VOP) if there exists no other $x \in S$ such that

$$f_j(x) \leq f_j(x^\circ), \quad \text{for all } j \in K,$$

and

$$f_i(x) < f_i(x^\circ), \quad \text{for at least one } i \in K_j.$$

Definition 2.2. An efficient solution x° is said to be a properly efficient solution of (VOP) if there exists a scalar $M > 0$ such that for each $j \in K$, $f_j(x) < f_j(x^\circ)$ and $x \in S$ imply that

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

for at least one $i \in K_j$ such that $f_i(x^\circ) < f_i(x)$.

The concept of so-called second-order convex functions was first introduced by Mond [11]. Later on, Bector and Chandra [4] named these functions as bonvex functions. They also introduced their generalizations.

Definition 2.3. A real-valued twice-differentiable function $f_j: X \rightarrow R$, $j \in K$, is said to be bonvex at x° if for every $x \in X$ and $p \in R^n$, we have

$$f_j(x) - f_j(x^\circ) + \frac{1}{2}p^t \nabla^2 f_j(x^\circ) p \geq (\nabla f_j(x^\circ) + \nabla^2 f_j(x^\circ) p)(x - x^\circ).$$

Definition 2.4. A real-valued twice-differentiable function $f_j: X \rightarrow R$, $j \in K$, is said to be strictly bonvex at x° if for every $x \in X$, $x \neq x^\circ$, and $p \in R^n$, we have

$$f_j(x) - f_j(x^\circ) + \frac{1}{2}p^t \nabla^2 f_j(x^\circ) p > (\nabla f_j(x^\circ) + \nabla^2 f_j(x^\circ) p)(x - x^\circ).$$

Definition 2.5. A real-valued twice-differentiable function $f_j: X \rightarrow R$, $j \in K$, is said to be quasibonvex at x° if for every $x \in X$ and $p \in R^n$, we have

$$f_j(x) \leq f_j(x^\circ) - \frac{1}{2}p^t \nabla^2 f_j(x^\circ) p \Rightarrow (\nabla f_j(x^\circ) + \nabla^2 f_j(x^\circ) p)(x - x^\circ) \leq 0.$$

Definition 2.6. A real-valued twice-differentiable function $f_j: X \rightarrow R$, $j \in K$, is said to be pseudobonvex at x° if for every $x \in X$ and $p \in R^n$, we have

$$f_j(x) < f_j(x^\circ) - \frac{1}{2}p^t \nabla^2 f_j(x^\circ) p \Rightarrow (\nabla f_j(x^\circ) + \nabla^2 f_j(x^\circ) p)(x - x^\circ) < 0.$$

Definition 2.7. A real-valued twice-differentiable function $f_j: X \rightarrow R$, $j \in K$, is said to be strictly pseudobonvex at x° if for every $x \in X$, $x \neq x^\circ$, and $p \in R^n$, we have

$$(\nabla f_j(x^\circ) + \nabla^2 f_j(x^\circ) p)(x - x^\circ) \geq 0 \Rightarrow f_j(x) > f_j(x^\circ) - \frac{1}{2}p^t \nabla^2 f_j(x^\circ) p.$$

We shall make use of the following generalized Schwartz inequality:

$$x^t A z \leq (x^t A x)^{\frac{1}{2}} (z^t A z)^{\frac{1}{2}},$$

where $x, z \in R^n$, and A is a positive semidefinite symmetric matrix of order n .

The following theorem will be needed in the sequel.

Theorem 2.8 [12]. *Let x° be a properly efficient solution of (P) at which a constraint qualification [10, 15] is satisfied. Then there exist $\lambda^\circ \in R^k$, $u^\circ \in R^m$ and $v_j^\circ \in R^n, j \in K$ such that*

$$\sum_{j=1}^k \lambda_j^\circ (\nabla f_j(x^\circ) + B_j v_j^\circ) + \nabla u^{\circ t} g(x^\circ) = 0,$$

$$u^{\circ t} g(x^\circ) = 0,$$

$$(x^{\circ t} B_j x^\circ)^{\frac{1}{2}} = x^{\circ t} B_j v_j^\circ, \quad j \in K,$$

$$v_j^{\circ t} B_j v_j^\circ \leq 1, \quad j \in K,$$

$$\lambda^\circ > 0, \quad \sum_{j=1}^k \lambda_j^\circ = 1, \quad u^\circ \geq 0.$$

3. MANGASARIAN TYPE DUALITY

In this section, we propose the following Mangasarian type dual to (P) and prove weak, strong, and strict converse duality theorems.

$$\begin{aligned} \text{Maximize} \quad & \left(f_1(y) + u^t g(y) + y^t B_1 v_1 - \frac{1}{2} p^t \nabla^2 \{f_1(y) + u^t g(y)\} p, \dots, \right. \\ & \left. f_k(y) + u^t g(y) + y^t B_k v_k - \frac{1}{2} p^t \nabla^2 \{f_k(y) + u^t g(y)\} p \right) \quad (\text{MSD}) \end{aligned}$$

subject to

$$\sum_{j=1}^k \lambda_j (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j) + \nabla u^t g(y) + \nabla^2 u^t g(y) p = 0, \quad (3.1)$$

$$v_j^t B_j v_j \leq 1, \quad j \in K, \quad (3.2)$$

$$\lambda > 0, \quad \sum_{j=1}^k \lambda_j = 1, \quad u \geq 0, \quad (3.3)$$

where $y, v_j, p \in R^n, j \in K$, and $u \in R^m$.

Theorem 3.1 (Weak Duality). *Let x and $(y, u, v_1, v_2, \dots, v_k, \lambda, p)$ be feasible solutions of (P) and (MSD), respectively. Suppose $(f_j(\cdot) + (\cdot)^t B_j v_j), j \in K$, is bonvex at y , and $g_i(\cdot), i \in M$, is bonvex at y . Then, the following cannot hold:*

$$f_j(x) + (x^t B_j x)^{\frac{1}{2}} \leq f_j(y) + u^t g(y) + y^t B_j v_j - \frac{1}{2} p^t \nabla^2 \{f_j(y) + u^t g(y)\} p, \quad \text{for all } j \in K, \quad (3.4)$$

and

$$f_i(x) + (x^t B_i x)^{\frac{1}{2}} < f_i(y) + u^t g(y) + y^t B_i v_i - \frac{1}{2} p^t \nabla^2 \{f_i(y) + u^t g(y)\} p, \quad \text{for at least one } i \in K_j. \quad (3.5)$$

Proof. Suppose to the contrary that (3.4) and (3.5) hold, that is,

$$f_j(x) + (x^t B_j x)^{\frac{1}{2}} \leq f_j(y) + u^t g(y) + y^t B_j v_j - \frac{1}{2} p^t \nabla^2 \{f_j(y) + u^t g(y)\} p, \quad \text{for all } j \in K,$$

and

$$f_i(x) + (x^t B_i x)^{\frac{1}{2}} < f_i(y) + u^t g(y) + y^t B_i v_i - \frac{1}{2} p^t \nabla^2 \{f_i(y) + u^t g(y)\} p, \quad \text{for at least one } i \in K_j.$$

Because $\lambda > 0$ and $\sum_{j=1}^k \lambda_j = 1$, the above inequalities yield

$$\sum_{j=1}^k \lambda_j (f_j(x) + (x^t B_j x)^{\frac{1}{2}}) < \sum_{j=1}^k \lambda_j \left(f_j(y) + y^t B_j v_j - \frac{1}{2} p^t \nabla^2 f_j(y) p \right) + u^t g(y) - \frac{1}{2} p^t \nabla^2 u^t g(y) p. \quad (3.6)$$

As $(f_j(\cdot) + (\cdot)^t B_j v_j), j \in K$ and $g_i(\cdot), i \in M$, are bonvex at y , we have

$$f_j(x) + x^t B_j v_j - f_j(y) - y^t B_j v_j + \frac{1}{2} p^t \nabla^2 f_j(y) p \geq (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j)(x - y), \quad (3.7)$$

and

$$g_i(x) - g_i(y) + \frac{1}{2} p^t \nabla^2 g_i(y) p \geq (\nabla g_i(y) + \nabla^2 g_i(y) p)(x - y). \quad (3.8)$$

On multiplying (3.7) by $\lambda_j > 0$, $j \in K$ and (3.8) by $u_i \geq 0$, $i \in M$, and then summing up to get

$$\begin{aligned} & \sum_{j=1}^k \lambda_j \left(f_j(x) + x^t B_j v_j - f_j(y) - y^t B_j v_j + \frac{1}{2} p^t \nabla^2 f_j(y) p \right) \\ & \quad + u^t g(x) - u^t g(y) + \frac{1}{2} p^t \nabla^2 u^t g(y) p \\ & \geq \left[\sum_{j=1}^k \lambda_j (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j) + \nabla u^t g(y) + \nabla^2 u^t g(y) p \right] (x - y). \end{aligned}$$

The above inequality, in view of (3.2), (3.6), the generalized Schwartz inequality, and $u^t g(x) \leq 0$, gives

$$\left[\sum_{j=1}^k \lambda_j (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j) + \nabla u^t g(y) + \nabla^2 u^t g(y) p \right] (x - y) < 0,$$

which is a contradiction to (3.1). \square

Theorem 3.2 (Strong Duality). *Let x° be a properly efficient solution of (P) at which a constraint qualification [10, 15] is satisfied. Then there exist $\lambda^\circ \in R^k$, $u^\circ \in R^m$, $v_j^\circ \in R^n$, $j \in K$, and $p^\circ \in R^n$ such that $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is feasible for (MSD) and the corresponding objective values of (P) and (MSD) are equal. If, in addition, the assumptions of weak duality (Theorem 3.1) are satisfied, then $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is a properly efficient solution of (MSD).*

Proof. Because x° is a properly efficient solution of (P) at which a constraint qualification [10, 15] is satisfied, by Theorem 2.8, there exist $\lambda^\circ \in R^k$, $u^\circ \in R^m$ and $v_j^\circ \in R^n$, $j \in K$ such that

$$\sum_{j=1}^k \lambda_j^\circ (\nabla f_j(x^\circ) + B_j v_j^\circ) + \nabla u^{\circ t} g(x^\circ) = 0,$$

$$u^{\circ t} g(x^\circ) = 0,$$

$$(x^{\circ t} B_j x^\circ)^{\frac{1}{2}} = x^{\circ t} B_j v_j^\circ, \quad j \in K,$$

$$v_j^{\circ t} B_j v_j^\circ \leq 1, \quad j \in K,$$

$$\lambda^\circ > 0, \quad \sum_{j=1}^k \lambda_j^\circ = 1, \quad u^\circ \geq 0.$$

Thus, $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is feasible for (MSD) and the corresponding objective values of (P) and (MSD) are equal. Now, we show that $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is an efficient solution of (MSD). Suppose that it is not efficient, then there exists a feasible solution $(y^*, u^*, v_1^*, v_2^*, \dots, v_k^*, \lambda^*, p^*)$ such that

$$\begin{aligned} f_j(y^*) + u^{*t}g(y^*) + y^{*t}B_jv_j^* - \frac{1}{2}p^{*t}\nabla^2[f_j(y^*) + u^{*t}g(y^*)]p^* \\ \geq f_j(x^\circ) + u^{\circ t}g(x^\circ) + x^{\circ t}B_jv_j^\circ, \quad \text{for all } j \in K, \end{aligned}$$

and

$$\begin{aligned} f_i(y^*) + u^{*t}g(y^*) + y^{*t}B_iv_i^* - \frac{1}{2}p^{*t}\nabla^2[f_i(y^*) + u^{*t}g(y^*)]p^* \\ > f_i(x^\circ) + u^{\circ t}g(x^\circ) + x^{\circ t}B_iv_i^\circ, \quad \text{for at least one } i \in K_j. \end{aligned}$$

Using $(x^{\circ t}B_jx^\circ)^{\frac{1}{2}} = x^{\circ t}B_jv_j^\circ$, $j \in K$, and $u^{\circ t}g(x^\circ) = 0$, the above inequalities give

$$\begin{aligned} f_j(x^\circ) + (x^{\circ t}B_jx^\circ)^{\frac{1}{2}} \leq f_j(y^*) + u^{*t}g(y^*) + y^{*t}B_jv_j^* \\ - \frac{1}{2}p^{*t}\nabla^2[f_j(y^*) + u^{*t}g(y^*)]p^*, \quad \text{for all } j \in K, \end{aligned}$$

and

$$\begin{aligned} f_i(x^\circ) + (x^{\circ t}B_ix^\circ)^{\frac{1}{2}} < f_i(y^*) + u^{*t}g(y^*) + y^{*t}B_iv_i^* \\ - \frac{1}{2}p^{*t}\nabla^2[f_i(y^*) + u^{*t}g(y^*)]p^*, \quad \text{for at least one } i \in K_j, \end{aligned}$$

which is a contradiction to weak duality (Theorem 3.1). Hence, $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is an efficient solution of (MSD). Now assume that it is not properly efficient, then for each scalar $M > 0$, there exists a feasible solution $(y^*, u^*, v_1^*, v_2^*, \dots, v_k^*, \lambda^*, p^*)$ and some $j \in K$, such that

$$\begin{aligned} f_j(y^*) + u^{*t}g(y^*) + y^{*t}B_jv_j^* - \frac{1}{2}p^{*t}\nabla^2[f_j(y^*) + u^{*t}g(y^*)]p^* \\ > f_j(x^\circ) + u^{\circ t}g(x^\circ) + x^{\circ t}B_jv_j^\circ, \end{aligned}$$

implies

$$\begin{aligned} \left[f_j(y^*) + u^{*t}g(y^*) + y^{*t}B_jv_j^* - \frac{1}{2}p^{*t}\nabla^2[f_j(y^*) + u^{*t}g(y^*)]p^* \right] \\ - [f_j(x^\circ) + u^{\circ t}g(x^\circ) + x^{\circ t}B_jv_j^\circ] \end{aligned}$$

$$\begin{aligned}
&> M \left[[f_i(x^\circ) + u^{\circ t} g(x^\circ) + x^{\circ t} B_i v_i^\circ] \right. \\
&\quad \left. - \left[f_i(y^*) + u^{*t} g(y^*) + y^{*t} B_i v_i^* - \frac{1}{2} p^{*t} \nabla^2 [f_i(y^*) + u^{*t} g(y^*)] p^* \right] \right],
\end{aligned}$$

for all $i \in K_j$ satisfying

$$\begin{aligned}
&[f_i(x^\circ) + u^{\circ t} g(x^\circ) + x^{\circ t} B_i v_i^\circ] \\
&> \left[f_i(y^*) + u^{*t} g(y^*) + y^{*t} B_i v_i^* - \frac{1}{2} p^{*t} \nabla^2 [f_i(y^*) + u^{*t} g(y^*)] p^* \right].
\end{aligned}$$

Again, by $(x^{\circ t} B_j x^\circ)^{\frac{1}{2}} = x^{\circ t} B_j v_j^\circ$, $j \in K$, and $u^{\circ t} g(x^\circ) = 0$, we get

$$\begin{aligned}
&\left[f_j(y^*) + u^{*t} g(y^*) + y^{*t} B_j v_j^* - \frac{1}{2} p^{*t} \nabla^2 [f_j(y^*) + u^{*t} g(y^*)] p^* \right] \\
&\quad - \left[f_j(x^\circ) + (x^{\circ t} B_j x^\circ)^{\frac{1}{2}} \right] \\
&> M \left[\left[f_i(x^\circ) + (x^{\circ t} B_i x^\circ)^{\frac{1}{2}} \right] \right. \\
&\quad \left. - \left[f_i(y^*) + u^{*t} g(y^*) + y^{*t} B_i v_i^* - \frac{1}{2} p^{*t} \nabla^2 [f_i(y^*) + u^{*t} g(y^*)] p^* \right] \right].
\end{aligned}$$

This means that $[f_j(y^*) + u^{*t} g(y^*) + y^{*t} B_j v_j^* - \frac{1}{2} p^{*t} \nabla^2 [f_j(y^*) + u^{*t} g(y^*)] p^*]$ is infinitely better than $[f_j(x^\circ) + (x^{\circ t} B_j x^\circ)^{\frac{1}{2}}]$ for some $j \in K$, whereas $[f_j(x^\circ) + (x^{\circ t} B_j x^\circ)^{\frac{1}{2}}]$ is at most finitely better than $[f_i(y^*) + u^{*t} g(y^*) + y^{*t} B_i v_i^* - \frac{1}{2} p^{*t} \nabla^2 [f_i(y^*) + u^{*t} g(y^*)] p^*]$ for all $i \in K_j$. Therefore,

$$\begin{aligned}
f_j(x^\circ) + (x^{\circ t} B_j x^\circ)^{\frac{1}{2}} &\leq f_j(y^*) + u^{*t} g(y^*) + y^{*t} B_j v_j^* \\
&\quad - \frac{1}{2} p^{*t} \nabla^2 [f_j(y^*) + u^{*t} g(y^*)] p^*, \quad \text{for all } j \in K,
\end{aligned}$$

and

$$\begin{aligned}
f_i(x^\circ) + (x^{\circ t} B_i x^\circ)^{\frac{1}{2}} &< f_i(y^*) + u^{*t} g(y^*) + y^{*t} B_i v_i^* \\
&\quad - \frac{1}{2} p^{*t} \nabla^2 [f_i(y^*) + u^{*t} g(y^*)] p^*, \quad \text{for at least one } i \in K_j,
\end{aligned}$$

which again contradicts weak duality (Theorem 3.1). Hence, $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is a properly efficient solution of (MSD). \square

Theorem 3.3 (Strict Converse Duality). *Let x° and $(y^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ)$ be feasible solutions of (P) and (MSD), respectively, such that*

$$\sum_{j=1}^k \lambda_j^\circ [f_j(x^\circ) + x^{\circ t} B_j v_j^\circ] \leq \sum_{j=1}^k \lambda_j^\circ \left[f_j(y^\circ) + y^{\circ t} B_j v_j^\circ - \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ \right] + u^{\circ t} g(y^\circ) - \frac{1}{2} p^{\circ t} \nabla^2 u^{\circ t} f(y^\circ) p^\circ. \tag{3.9}$$

Suppose $(f_j(\cdot) + (\cdot)^t B_j v_j^\circ)$, $j \in K$ is strictly bonvex at y° , and $g_i(\cdot)$, $i \in M$ is bonvex at y° , then $x^\circ = y^\circ$.

Proof. We assume that $x^\circ \neq y^\circ$ and exhibit a contradiction. The strict bonvexity of $(f_j(\cdot) + (\cdot)^t B_j v_j^\circ)$, $j \in K$ at y° and bonvexity of $g_i(\cdot)$, $i \in M$ at y° , imply

$$\begin{aligned} & \left(f_j(x^\circ) + x^{\circ t} B_j v_j^\circ - f_j(y^\circ) - y^{\circ t} B_j v_j^\circ + \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ \right) \\ & > (\nabla f_j(y^\circ) + \nabla^2 f_j(y^\circ) p^\circ + B_j v_j^\circ)(x^\circ - y^\circ), \end{aligned} \tag{3.10}$$

and

$$g_i(x^\circ) - g_i(y^\circ) + \frac{1}{2} p^{\circ t} \nabla^2 g_i(y^\circ) p^\circ \geq (\nabla g_i(y^\circ) + \nabla^2 g_i(y^\circ) p^\circ)(x^\circ - y^\circ). \tag{3.11}$$

On multiplying (3.10) by $\lambda_j^\circ > 0$, $j \in K$ and (3.11) by $u_i^\circ \geq 0$, $i \in M$, and then summing up to get

$$\begin{aligned} & \sum_{j=1}^k \lambda_j^\circ (f_j(x^\circ) + x^{\circ t} B_j v_j^\circ) + u^{\circ t} g(x^\circ) - \sum_{j=1}^k \lambda_j^\circ (f_j(y^\circ) + y^{\circ t} B_j v_j^\circ) \\ & - \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ - u^{\circ t} g(y^\circ) + \frac{1}{2} p^{\circ t} \nabla^2 u^{\circ t} g(y^\circ) p^\circ \\ & > \left[\sum_{j=1}^k \lambda_j^\circ (\nabla f_j(y^\circ) + \nabla^2 f_j(y^\circ) p^\circ + B_j v_j^\circ) + \nabla u^{\circ t} g(y^\circ) + \nabla^2 u^{\circ t} g(y^\circ) p^\circ \right] (x^\circ - y^\circ). \end{aligned}$$

The above inequality on using (3.1) and $u^{\circ t} g(x^\circ) \leq 0$, gives

$$\begin{aligned} \sum_{j=1}^k \lambda_j^\circ [f_j(x^\circ) + x^{\circ t} B_j v_j^\circ] & > \sum_{j=1}^k \lambda_j^\circ \left[f_j(y^\circ) + y^{\circ t} B_j v_j^\circ - \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ \right] \\ & + u^{\circ t} g(y^\circ) - \frac{1}{2} p^{\circ t} \nabla^2 u^{\circ t} g(y^\circ) p^\circ, \end{aligned}$$

which is a contradiction to (3.9). Hence, $x^\circ = y^\circ$. □

4. GENERAL MOND-WEIR TYPE DUALITY

For (P), we now formulate the following general Mond-Weir type second-order dual:

$$\begin{aligned} \text{Maximize } & \left(f_1(y) + y^t B_1 v_1 + \sum_{i \in I_0} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \left[f_1(y) + \sum_{i \in I_0} u_i g_i(y) \right] p, \dots, \right. \\ & \left. f_k(y) + y^t B_k v_k + \sum_{i \in I_0} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \left[f_k(y) + \sum_{i \in I_0} u_i g_i(y) \right] p \right) \end{aligned} \quad (\text{GMD})$$

subject to

$$\sum_{j=1}^k \lambda_j (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j) + \nabla u^t g(y) + \nabla^2 u^t g(y) p = 0, \quad (4.1)$$

$$\sum_{i \in I_\alpha} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \sum_{i \in I_\alpha} u_i g_i(y) p \geq 0, \quad \alpha = 1, 2, \dots, r, \quad (4.2)$$

$$v_j^t B_j v_j \leq 1, \quad j \in K, \quad (4.3)$$

$$\lambda > 0, \quad \sum_{j=1}^k \lambda_j = 1, \quad u \geq 0, \quad (4.4)$$

where $I_\alpha \subseteq M$, $\alpha = 0, 1, 2, \dots, r$, with $\bigcup_{\alpha=0}^r I_\alpha = M$ and $I_\alpha \cap I_\beta = \emptyset$, if $\alpha \neq \beta$.

Remark 4.1. If $I_\alpha = \emptyset$, $\alpha = 1, 2, \dots, r$, and $I_0 = M$, then (GMD) reduces to (MSD).

Theorem 4.2 (Weak Duality). Let x and $(y, u, v_1, v_2, \dots, v_k, \lambda, p)$ be feasible solutions of (P) and (GMD), respectively. Suppose $[\sum_{j=1}^k \lambda_j (f_j(\cdot) + (\cdot)^t B_j v_j) + \sum_{i \in I_0} u_i g_i(\cdot)]$ is pseudobonvex at y , and $\sum_{i \in I_\alpha} u_i g_i(\cdot)$, $\alpha = 1, 2, \dots, r$, is quasibonvex at y . Then, the following cannot hold:

$$\begin{aligned} f_j(x) + (x^t B_j x)^{\frac{1}{2}} & \leq f_j(y) + y^t B_j v_j + \sum_{i \in I_0} u_i g_i(y) \\ & - \frac{1}{2} p^t \nabla^2 \left[f_j(y) + \sum_{i \in I_0} u_i g_i(y) \right] p, \quad \text{for all } j \in K, \end{aligned} \quad (4.5)$$

and

$$f_l(x) + (x^t B_l x)^{\frac{1}{2}} < f_l(y) + y^t B_l v_l + \sum_{i \in I_o} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \left[f_i(y) + \sum_{i \in I_o} u_i g_i(y) \right] p,$$

for at least one $l \in K_j$. (4.6)

Proof. Because x and $(y, u, v_1, v_2, \dots, v_k, \lambda, p)$ are feasible solutions of (P) and (GMD), respectively, then

$$\sum_{i \in I_x} u_i g_i(x) \leq 0 \leq \sum_{i \in I_x} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \sum_{i \in I_x} u_i g_i(y) p, \quad \alpha = 1, 2, \dots, r,$$

which by the quasibonvexity of $\sum_{i \in I_x} u_i g_i(\cdot)$, $\alpha = 1, 2, \dots, r$, at y yields

$$\left(\nabla \sum_{i \in I_x} u_i g_i(y) + \nabla^2 \sum_{i \in I_x} u_i g_i(y) p \right) (x - y) \leq 0, \quad \alpha = 1, 2, \dots, r. \quad (4.7)$$

Now, suppose to the contrary that (4.5) and (4.6) hold, that is,

$$f_j(x) + (x^t B_j x)^{\frac{1}{2}} \leq f_j(y) + y^t B_j v_j + \sum_{i \in I_o} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \left[f_j(y) + \sum_{i \in I_o} u_i g_i(y) \right] p, \quad \text{for all } j \in K,$$

and

$$f_l(x) + (x^t B_l x)^{\frac{1}{2}} < f_l(y) + y^t B_l v_l + \sum_{i \in I_o} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \left[f_i(y) + \sum_{i \in I_o} u_i g_i(y) \right] p, \quad \text{for at least one } l \in K_j.$$

Because $\lambda > 0$ and $\sum_{j=1}^k \lambda_j = 1$, we get

$$\begin{aligned} \sum_{j=1}^k \lambda_j (f_j(x) + (x^t B_j x)^{\frac{1}{2}}) &< \sum_{j=1}^k \lambda_j \left(f_j(y) + y^t B_j v_j - \frac{1}{2} p^t \nabla^2 f_j(y) p \right) \\ &+ \sum_{i \in I_o} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \sum_{i \in I_o} u_i g_i(y) p, \end{aligned}$$

which on using the generalized Schwartz inequality along with (4.3) and the feasibility of (P), implies

$$\begin{aligned} & \sum_{j=1}^k \lambda_j (f_j(x) + (x^t B_j v_j) + \sum_{i \in I_0} u_i g_i(x) \\ & < \sum_{j=1}^k \lambda_j \left(f_j(y) + y^t B_j v_j - \frac{1}{2} p^t \nabla^2 f_j(y) p \right) + \sum_{i \in I_0} u_i g_i(y) - \frac{1}{2} p^t \nabla^2 \sum_{i \in I_0} u_i g_i(y) p. \end{aligned}$$

The pseudobonvexity of $[\sum_{j=1}^k \lambda_j (f_j(\cdot) + (\cdot)^t B_j v_j) + \sum_{i \in I_0} u_i g_i(\cdot)]$ at y , implies

$$\left[\sum_{j=1}^k \lambda_j (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j) + \nabla \sum_{i \in I_0} u_i g_i(y) + \nabla^2 \sum_{i \in I_0} u_i g_i(y) p \right] (x - y) < 0. \quad (4.8)$$

On combining (4.7) and (4.8), we obtain

$$\left[\sum_{j=1}^k \lambda_j (\nabla f_j(y) + \nabla^2 f_j(y) p + B_j v_j) + \nabla u^t g(y) + \nabla^2 u^t g(y) p \right] (x - y) < 0,$$

which is a contradiction to (4.1). \square

Theorem 4.3 (Strong Duality). *Let x° be a properly efficient solution of (P) at which a constraint qualification [10, 15] is satisfied. Then there exist $\lambda^\circ \in R^k$, $u^\circ \in R^m$, $v_j^\circ \in R^n$, $j \in K$ and $p^\circ \in R^n$ such that $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is feasible for (GMD) and the corresponding objective values of (P) and (GMD) are equal. If, in addition, the assumptions of weak duality (Theorem 4.2) are satisfied, then $(x^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ = 0)$ is a properly efficient solution of (GMD).*

Proof. The proof follows on the similar lines of Theorem 3.2.

Theorem 4.4 (Strict Converse Duality). *Let x° and $(y^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ)$ be the feasible solutions of (P) and (GMD), respectively, such that*

$$\begin{aligned} \sum_{j=1}^k \lambda_j^\circ [f_j(x^\circ) + x^{\circ t} B_j v_j^\circ] & \leq \sum_{j=1}^k \lambda_j^\circ \left[f_j(y^\circ) + y^{\circ t} B_j v_j^\circ - \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ \right] \\ & + \sum_{i \in I_0} u_i^\circ g_i(y^\circ) - \frac{1}{2} p^{\circ t} \nabla^2 \sum_{i \in I_0} u_i^\circ g_i(y^\circ) p^\circ. \quad (4.9) \end{aligned}$$

If $[\sum_{j=1}^k \lambda_j^\circ (f_j(\cdot) + (\cdot)^t B_j v_j^\circ) + \sum_{i \in I_0} u_i^\circ g_i(\cdot)]$ is strictly pseudobonvex at y° , and $\sum_{i \in I_\alpha} u_i^\circ g_i(\cdot)$, $\alpha = 1, 2, \dots, r$, is quasibonvex at y° , then $x^\circ = y^\circ$.

Proof. We assume that $x^\circ \neq y^\circ$ and exhibit a contradiction. Because x° and $(y^\circ, u^\circ, v_1^\circ, v_2^\circ, \dots, v_k^\circ, \lambda^\circ, p^\circ)$ are feasible solutions of (P) and (GMD), respectively, we have

$$\sum_{i \in I_\alpha} u_i^\circ g_i(x^\circ) \leq 0 \leq \sum_{i \in I_\alpha} u_i^\circ g_i(y^\circ) - \frac{1}{2} p^{\circ t} \nabla^2 \sum_{i \in I_\alpha} u_i^\circ g_i(y^\circ) p^\circ, \quad \alpha = 1, 2, \dots, r.$$

By the quasiconvexity of $\sum_{i \in I_\alpha} u_i^\circ g_i(\cdot), \alpha = 1, 2, \dots, r$ at y° , it follows that

$$\left(\nabla \sum_{i \in I_\alpha} u_i^\circ g_i(y^\circ) + \nabla^2 \sum_{i \in I_\alpha} u_i^\circ g_i(y^\circ) p^\circ \right) (x^\circ - y^\circ) \leq 0.$$

The above inequality together with (4.1) gives

$$\left[\sum_{j=1}^k \lambda_j^\circ (\nabla f_j(y^\circ) + \nabla^2 f_j(y^\circ) p^\circ + B_j v_j^\circ) + \nabla \sum_{i \in I_0} u_i^\circ g_i(y^\circ) + \nabla^2 \sum_{i \in I_0} u_i^\circ g_i(y^\circ) p^\circ \right] (x^\circ - y^\circ) \geq 0,$$

which on using the strict pseudobonvexity of $[\sum_{j=1}^k \lambda_j^\circ (f_j(\cdot) + (\cdot)^t B_j v_j^\circ) + \sum_{i \in I_0} u_i^\circ g_i(\cdot)]$ at y° , yields

$$\begin{aligned} & \sum_{j=1}^k \lambda_j^\circ [f_j(x^\circ) + x^{\circ t} B_j v_j^\circ] + \sum_{i \in I_0} u_i^\circ g_i(x^\circ) \\ & > \sum_{j=1}^k \lambda_j^\circ \left[f_j(y^\circ) + y^{\circ t} B_j v_j^\circ - \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ \right] \\ & \quad + \sum_{i \in I_0} u_i^\circ g_i(y^\circ) - \frac{1}{2} p^{\circ t} \nabla^2 \sum_{i \in I_0} u_i^\circ g_i(y^\circ) p^\circ, \end{aligned}$$

which by $\sum_{i \in I_0} u_i^\circ g_i(x^\circ) \leq 0$ implies

$$\begin{aligned} \sum_{j=1}^k \lambda_j^\circ [f_j(x^\circ) + x^{\circ t} B_j v_j^\circ] & > \sum_{j=1}^k \lambda_j^\circ \left[f_j(y^\circ) + y^{\circ t} B_j v_j^\circ - \frac{1}{2} p^{\circ t} \nabla^2 f_j(y^\circ) p^\circ \right] \\ & \quad + \sum_{i \in I_0} u_i^\circ g_i(y^\circ) - \frac{1}{2} p^{\circ t} \nabla^2 \sum_{i \in I_0} u_i^\circ g_i(y^\circ) p^\circ, \end{aligned}$$

a contradiction to (4.9). Hence $x^\circ = y^\circ$. □

5. CONCLUSION

In this paper, we have discussed the second-order duality results for nondifferentiable multiobjective Mangasarian type and general Mond–Weir type duals involving bonconvexity/generalized bonconvexity. The current results can be further extended for the following nondifferentiable multiobjective fractional programming problem:

$$\text{Minimize } \left(\frac{f_1(x) + (x^t B_1 x)^{\frac{1}{2}}}{h_1(x) - (x^t D_1 x)^{\frac{1}{2}}}, \frac{f_2(x) + (x^t B_2 x)^{\frac{1}{2}}}{h_2(x) - (x^t D_2 x)^{\frac{1}{2}}}, \dots, \frac{f_k(x) + (x^t B_k x)^{\frac{1}{2}}}{h_k(x) - (x^t D_k x)^{\frac{1}{2}}} \right)$$

subject to $x \in S = \{x \in X : g(x) \leq 0\}$, (\bar{P})

where $f_j : X \rightarrow R$, $h_j : X \rightarrow R$, $j = 1, 2, \dots, k$, $g : X \rightarrow R^m$; B_j and D_j , $j = 1, 2, \dots, k$ are $n \times n$ positive semidefinite symmetric matrices.

REFERENCES

1. B. Aghezzaf (2003). Second order mixed type duality in multiobjective programming problems. *J. Math. Anal. Appl.* 285:97–106.
2. I. Ahmad (2005). Sufficiency and duality in multiobjective programming with generalized (F, ρ) -convexity. *J. Appl. Anal.* 11:19–33.
3. I. Ahmad and Z. Husain (2006). Second order (F, α, ρ, d) -convexity and duality in multiobjective programming. *Inform. Sci.* 176:3094–3103.
4. C.R. Bector and S. Chandra (1987). Generalized-bonconvexity and higher order duality for fractional programming. *Opsearch.* 24:143–154.
5. T.R. Gulati and M.A. Islam (1994). Sufficiency and duality in multiobjective programming involving generalized F -convex functions. *J. Math. Anal. Appl.* 183:181–195.
6. M. Hachimi and B. Aghezzaf (2004). Second order duality in multiobjective programming involving generalized type-I functions. *Numer. Funct. Anal. Optim.* 25:725–736.
7. M.A. Hanson and B. Mond (1982). Further generalizations of convexity in mathematical programming. *J. Inform. Optim. Sci.* 3:25–32.
8. O.L. Mangasarian (1975). Second and higher order duality in nonlinear programming. *J. Math. Anal. Appl.* 51:607–620.
9. S.K. Mishra (1997). Second order generalized invexity and duality in mathematical programming. *Optimization.* 42:51–69.
10. B. Mond (1974). A class of nondifferentiable mathematical programming problems. *J. Math. Anal. Appl.* 46:169–174.
11. B. Mond (1974). Second order duality for nonlinear programs. *Opsearch.* 11:90–99.
12. B. Mond, I. Husain, and M.V. Durgaprasad (1988). Duality for a class of nondifferentiable multiple objective programming problems. *J. Inform. Optim. Sci.* 9:331–341.
13. V. Preda (1992). On efficiency and duality for multiobjective programs. *J. Math. Anal. Appl.* 166:365–377.
14. J.P. Vial (1983). Strong and weak convexity of sets and functions. *Math. Oper. Res.* 8:231–259.
15. H. Wolkowitz (1983). An optimality condition for a nondifferentiable convex program. *Naval Res. Logistics Quart.* 30:415–418.
16. J. Zhang and B. Mond (1997). Duality for a nondifferentiable programming problem. *Bull. Austral. Math. Soc.* 55:29–44.