

SECOND ORDER MIXED SYMMETRIC DUALITY IN NON-DIFFERENTIABLE MULTI-OBJECTIVE MATHEMATICAL PROGRAMMING

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Abstract. A pair of Mond-Weir type second order mixed symmetric duals is presented for a class of non-differentiable multi-objective non-linear programming problems with multiple arguments. We establish duality theorems for the new pair of dual models under second order generalized convexity assumptions. This mixed second order dual formulation unifies the two existing second order symmetric dual formulations in the literature. Many recent works on symmetric duality are obtained as special cases of the results established in the present paper.

1. INTRODUCTION

Dorn [7] introduced symmetric duality in nonlinear programming by defining a program and its dual to be symmetric if the dual of the dual

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is the original problem. The symmetric duality for scalar programming has been studied extensively in the literature, one can refer to Dantzig et al. [5], Mishra [14]–[16], Mond [18], Mond and Weir [21].

Mond and Schechter [20] studied non-differentiable symmetric duality for a class of optimization problems in which the objective function consists of support functions. Following Mond and Schechter [20], Chen [4], Hou and Yang [11], and Yang et al. [26], [28], studied symmetric duality for such problems.

The study of second order duality is significant due to the computational advantage over the first order duality as it provides tighter bounds for the value of the objective function when approximations are used (see [9], [13], [17], [19]). Mond [19] was the first one to present second order symmetric dual models and proved second order symmetric duality theorems under convexity assumptions. Mishra [15] established Wolfe type and Mond-Weir type second order symmetric duality for nonlinear programming problems under second order generalized convexity assumptions. Hou and Yang [11] and Yang et al. [26] extended the results of Mishra [15] to the non-differentiable case. Suneja et al. [24] extended the results of Mishra [15] to the multiobjective case. Some more results on symmetric duality can be seen in [6], [21], [22], [25].

Yang et al. [27] presented a mixed symmetric dual formulation for a non-differentiable nonlinear programming problem. Bector et al. [3] introduced a mixed symmetric dual model for a class of nonlinear multiobjective programming problems. However, the models given by Bector et al. [3] as well as by Yang et al. [27] do not allow the further weakening of generalized convexity assumptions on a part of the objective functions.

Very recently, Yang et al. [28] established second order duality results for second order symmetric dual models for a class of non-differentiable multi-objective programming problems as an extension of the results of Suneja et al. [24].

In this paper, a pair of Mond-Weir type second order mixed symmetric dual is presented for a class of non-differentiable multi-objective nonlinear programming problems with multiple arguments. We establish duality theorems for the new pair of dual models under second order generalized convexity assumptions. This mixed second order dual formulation unifies the two existing second order symmetric dual formulations in the literature. Many recent works on symmetric duality are obtained as special cases of the results established in the present paper.

2. PRELIMINARIES

Let \mathbb{R}^n denote the n -dimensional Euclidean space and let \mathbb{R}_+^n be its non-negative orthant. The following ordering relations for vectors in \mathbb{R}^n will be used in this paper:

For $x, y \in \mathbb{R}^n$, by $x < y$ we mean $y - x \in \text{int } \mathbb{R}_+^n$; by $x \leq y$ we mean $y - x \in \mathbb{R}_+^n \setminus \{0\}$; and by $x \leq y$ we mean $y - x \in \mathbb{R}_+^n$. The negation of $x \leq y$ is denoted by $x \not\leq y$.

Let $f(x, y)$ be real valued twice differentiable function defined on $\mathbb{R}^n \times \mathbb{R}^m$. Let $\nabla_x f(\bar{x}, \bar{y})$ and $\nabla_y f(\bar{x}, \bar{y})$ denote the partial derivatives of $f(x, y)$ with respect to x and y at (\bar{x}, \bar{y}) . Let $\nabla_{xx} f(\bar{x}, \bar{y})$ denote the $n \times n$ symmetric Hessian matrix at (\bar{x}, \bar{y}) , $\nabla_{xy} f(\bar{x}, \bar{y})$, $\nabla_{yx} f(\bar{x}, \bar{y})$ and $\nabla_{yy} f(\bar{x}, \bar{y})$ are defined similarly. Consider the following multiobjective programming problem (VP):

$$\text{Min } (f_1(x), f_2(x), \dots, f_p(x)) \quad \text{subject to } x \in X_0,$$

where $f_i: \mathbb{R}^n \rightarrow \mathbb{R}$, $i = 1, 2, \dots, p$ and $X_0 \subseteq \mathbb{R}^n$.

For problem (VP), an efficient solution and a properly efficient solution are defined as follows:

Definition 1. A feasible solution x^0 is said to be an efficient solution for (VP) if there exists no other feasible solution x such that

$$f(x) \leq f(x^0).$$

Definition 2. A feasible solution x^0 is said to be properly efficient solution for (VP) if it is efficient for (VP) and there exists a scalar $M > 0$ such that, for each i , we have

$$\frac{f_i(x^0) - f_i(x)}{f_j(x) - f_j(x^0)} \leq M$$

for some j such that $f_j(x) > f_j(x^0)$ whenever $x \in X_0$ and $f_i(x) < f_i(x^0)$.

Let C be a compact convex set in \mathbb{R}^n . The support function of C is defined by

$$s(x|C) = \max \{x^T y : y \in C\}.$$

A support function, being convex and everywhere finite, has a subdifferential [23], that is, there exists $z \in \mathbb{R}^n$ such that

$$s(y|C) \geq s(x|C) + z^T(y - x), \quad \forall y \in C.$$

The subdifferential of $s(x|C)$ is given by

$$\partial s(x|C) = \{z \in C : z^T x = s(x|C)\}.$$

For any set $D \subset \mathbb{R}^n$, the normal cone to D at a point $x \in D$ is defined by

$$N_D(x) = \{y \in \mathbb{R}^n : y^T(z - x) \leq 0, \forall z \in D\}.$$

It is obvious that for a compact convex set C , $y \in N_C(x)$ if and only if $s(y|C) = x^T y$, or equivalently, $x \in \partial s(y|C)$.

The following definitions will be needed in the sequel:

Definition 3. Let $X \subset \mathbb{R}^n$. A functional $F: X \times X \times \mathbb{R}^n \rightarrow \mathbb{R}$ is said to be sublinear with respect to its third argument if for any $x, y \in X$

- (A) $F(x, y; a_1 + a_2) = F(x, y; a_1) + F(x, y; a_2)$ for any $a_1, a_2 \in \mathbb{R}^n$;
- (B) $F(x, y; \alpha a) = \alpha F(x, y; a)$ for any $\alpha \in \mathbb{R}_+$ and $a \in \mathbb{R}^n$.

Definition 4. Let $X \subset \mathbb{R}^n$, $Y \subset \mathbb{R}^m$ and $F: X \times Y \times \mathbb{R}^n \rightarrow \mathbb{R}$ be sublinear with respect to its third argument. $f(\cdot, y)$ is said to be *second order F -convex* at $\bar{x} \in X$, with respect to $p \in \mathbb{R}^n$, for fixed $y \in Y$, if

$$f(x, y) - f(\bar{x}, y) + \frac{1}{2}p^T \nabla_{xx} f(\bar{x}, y) p \geq F(x, \bar{x}; \nabla_x f(\bar{x}, y) + \nabla_{xx} f(\bar{x}, y) p), \quad \forall x \in X.$$

f is said to be *second order F -concave* at $\bar{x} \in X$, with respect to $p \in \mathbb{R}^n$, for fixed $y \in Y$, if $-f$ is *second order F -convex* at $\bar{x} \in X$, with respect to $p \in \mathbb{R}^n$.

Definition 5. Let $X \subset \mathbb{R}^n$, $Y \subset \mathbb{R}^m$ and $F: X \times Y \times \mathbb{R}^n \rightarrow \mathbb{R}$ be sublinear with respect to its third argument. $f(x, \cdot)$ is said to be *second order F -pseudo-convex* at $\bar{x} \in X$, with respect to $p \in \mathbb{R}^n$, for fixed $y \in Y$, if

$$\begin{aligned} F(x, \bar{x}; \nabla_x f(\bar{x}, y) + \nabla_{xx} f(\bar{x}, y) p) &\geq 0 \\ \Rightarrow f(x, y) &\geq f(\bar{x}, y) - \frac{1}{2}p^T \nabla_{xx} f(\bar{x}, y) p, \quad \forall x \in X. \end{aligned}$$

f is said to be *second order F -pseudo-concave* at $\bar{x} \in X$, with respect to $p \in \mathbb{R}^n$, for fixed $y \in Y$, if $-f$ is *second order F -pseudo-convex* at $\bar{x} \in X$, with respect to $p \in \mathbb{R}^n$.

Remark 1.

- (i) The second order F -pseudo-convexity reduces to the F -pseudo-convexity introduced by Hanson and Mond [10] when $p = 0$.
- (ii) For $F(x, \bar{x}; a) = \eta(x, \bar{x})^T a$, and $p = 0$, where $\eta: X \times X \rightarrow \mathbb{R}^n$, the second order F -convexity reduces to the invexity introduced by Hanson [8], and second order F -pseudo-convexity reduces to the pseudo-invexity introduced by Hanson [8].

3. MIXED TYPE MULTI-OBJECTIVE SYMMETRIC DUALITY

For $N = \{1, 2, \dots, n\}$ and $M = \{1, 2, \dots, m\}$ let $J_1 \subset N$, $K_1 \subset M$ and $J_2 = N \setminus J_1$ and $K_2 = M \setminus K_1$. Let $|J_1|$ denote the number of elements in the set J_1 . The numbers $|J_2|$, $|K_1|$ and $|K_2|$ are defined similarly. Notice that if $J_1 = \Phi$, then $J_2 = N$, that is, $|J_1| = 0$ and $|J_2| = n$. Hence, $\mathbb{R}^{|J_1|}$ is zero dimensional Euclidean space and $\mathbb{R}^{|J_2|}$ is n -dimensional Euclidean space. It is clear that any $x \in \mathbb{R}^n$ can be written as $x = (x^1, x^2)$, $x^1 \in \mathbb{R}^{|J_1|}$, $x^2 \in \mathbb{R}^{|J_2|}$. Similarly, any $y \in \mathbb{R}^m$ can be written as $y = (y^1, y^2)$, $y^1 \in \mathbb{R}^{|K_1|}$, $y^2 \in \mathbb{R}^{|K_2|}$. Let $f: \mathbb{R}^{|J_1|} \times \mathbb{R}^{|K_1|} \rightarrow \mathbb{R}^l$ and $g: \mathbb{R}^{|J_2|} \times \mathbb{R}^{|K_2|} \rightarrow \mathbb{R}^l$ be twice differentiable functions and $e = (1, 1, \dots, 1)^T \in \mathbb{R}^l$.

Now we can introduce the following pair of non-differentiable multi-objective programs and discuss duality theorems under some mild assumptions of generalized convexity.

Primal Problem (SMP)

$$\begin{aligned} & \text{Min } H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \\ & = (H_1(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda), \dots, H_l(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda)) \end{aligned}$$

subject to

$$\begin{aligned} & (x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \\ & \in \mathbb{R}^{|J_1|} \times \mathbb{R}^{|J_2|} \times \mathbb{R}^{|K_1|} \times \mathbb{R}^{|K_2|} \times \mathbb{R}^{|K_1|} \times \mathbb{R}^{|K_2|} \times \mathbb{R}^{|K_1|} \times \mathbb{R}^{|K_2|} \times \mathbb{R}_+^l \end{aligned}$$

$$\sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1] \leq 0 \quad (1)$$

$$\sum_{i=1}^l \lambda_i [\nabla_{y^2} g_i(x^2, y^2) - z_i^2 + \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2] \leq 0 \quad (2)$$

$$(y^1)^T \sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1] \geq 0 \quad (3)$$

$$(y^2)^T \sum_{i=1}^l \lambda_i [\nabla_{y^2} g_i(x^2, y^2) - z_i^2 + \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2] \geq 0 \quad (4)$$

$$(x^1, x^2) \geq 0, \quad (5)$$

$$z_i^1 \in D_i^1, \text{ and } z_i^2 \in D_i^2, \quad i = 1, 2, \dots, l \quad (6)$$

$$\lambda > 0, \quad \sum_{i=1}^l \lambda_i = 1. \quad (7)$$

Dual Problem (SMD)

$$\begin{aligned} & \text{Max } G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda) \\ & = (G_1(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda), \dots, G_l(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda)) \end{aligned}$$

subject to

$$\begin{aligned} & (u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda) \\ & \in \mathbb{R}^{|J_1|} \times \mathbb{R}^{|J_2|} \times \mathbb{R}^{|K_1|} \times \mathbb{R}^{|K_2|} \times \mathbb{R}^{|J_1|} \times \mathbb{R}^{|J_2|} \times \mathbb{R}^{|J_1|} \times \mathbb{R}^{|J_2|} \times \mathbb{R}_+^l \end{aligned}$$

$$\sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \geq 0 \quad (8)$$

$$\sum_{i=1}^l \lambda_i [\nabla_{u^2} g_i(u^2, v^2) + w_i^2 + \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2] \geq 0 \quad (9)$$

$$(u^1)^T \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \leq 0 \quad (10)$$

$$(u^2)^T \sum_{i=1}^l \lambda_i [\nabla_{u^2} g_i(u^2, v^2) + w_i^2 + \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2] \leq 0 \quad (11)$$

$$(v^1, v^2) \geq 0, \quad (12)$$

$$w_i^1 \in C_i^1, \text{ and } w_i^2 \in C_i^2, \quad i = 1, 2, \dots, l \quad (13)$$

$$\lambda > 0, \quad \sum_{i=1}^l \lambda_i = 1 \quad (14)$$

where

$$\begin{aligned} & H_i(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \\ & = f_i(x^1, y^1) + g_i(x^2, y^2) + s(x^1 | C_i^1) + s(x^2 | C_i^2) - (y^1)^T z_i^1 \\ & \quad - (y^2)^T z_i^2 - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 - \frac{1}{2} (q_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) q_i^2 \\ & G_i(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda) \\ & = f_i(u^1, v^1) + g_i(u^2, v^2) - s(v^1 | D_i^1) - s(v^2 | D_i^2) + (u^1)^T w_i^1 \\ & \quad + (u^2)^T w_i^2 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2 \end{aligned}$$

and C_i^1 is a compact and convex subsets of $\mathbb{R}^{|J_1|}$ for $i = 1, 2, \dots, l$ and C_i^2 is a compact and convex subsets of $\mathbb{R}^{|J_2|}$ for $i = 1, 2, \dots, l$, similarly, D_i^1 is a compact and convex subsets of $\mathbb{R}^{|K_1|}$ for $i = 1, 2, \dots, l$ and D_i^2 is a compact and convex subsets of $\mathbb{R}^{|K_2|}$ for $i = 1, 2, \dots, l$.

For the first model, we can prove the following weak duality theorem.

Theorem 1 (Weak duality). *Let $(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda)$ be feasible for (SMP) and $(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda)$ be feasible for (SMD). Suppose there exist sublinear functionals F_1, F_2, G_1 and G_2 satisfying the following conditions:*

- (i) $F_1(x^1, u^1; a) + (u^1)^T a \geq 0$ if $a \geq 0$;
- (ii) $G_1(x^2, u^2; b) + (u^2)^T b \geq 0$ if $b \geq 0$;
- (iii) $F_2(v^1, y^1; c) + (y^1)^T c \geq 0$ if $c \geq 0$;
- (iv) $G_2(v^2, y^2; d) + (y^2)^T d \geq 0$ if $d \geq 0$.

Furthermore, assume that for $i = 1, 2, \dots, l$, $f_i(\cdot, v^1) + \cdot^T w_i^1$ is second order F_1 -convex for fixed v^1 , with respect to $q_i^1 \in \mathbb{R}^{|J_1|}$, $f_i(x^1, \cdot) - \cdot^T z_i^1$ is second order F_2 -concave for fixed x^1 , with respect to $p_i^1 \in \mathbb{R}^{|K_1|}$, $g_i(\cdot, v^2) + \cdot^T w_i^2$ is second order G_1 -convex for fixed v^2 with respect to $q_i^2 \in \mathbb{R}^{|J_2|}$ and $g_i(x^2, \cdot) - \cdot^T z_i^2$ is second order G_2 -concave for fixed x^2 , with respect to $p_i^2 \in \mathbb{R}^{|K_2|}$.

Then

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \not\leq G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

Proof. Assume that the result is not true, that is,

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \leq G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

Then, since $\lambda > 0$, we have

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) + g_i(x^2, y^2) + s(x^1 | C_i^1) + s(x^2 | C_i^2) - (y^1)^T z_i^1 \right. \\ & \quad - (y^2)^T z_i^2 - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \\ & \quad \left. - \frac{1}{2} (p_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2 \right] \\ & < \sum_{i=1}^l \lambda_i \left[f_i(u^1, v^1) + g_i(u^2, v^2) - s(v^1 | D_i^1) - s(v^2 | D_i^2) + (u^1)^T w_i^1 \right. \\ & \quad \left. + (u^2)^T w_i^2 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2 \right] \end{aligned} \quad (15)$$

By the second order F_1 -convexity of $f_i(\cdot, v^1) + \cdot^T w_i^1$ at x^1 with respect to $q_i^1 \in \mathbb{R}^{|J_1|}$, we have

$$\begin{aligned} & f_i(x^1, v^1) + (x^1)^T w_i^1 - f_i(u^1, v^1) - (u^1)^T w_i^1 + \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 \\ & \geq F_1(x^1, u^1; \nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1). \end{aligned}$$

for $i = 1, 2, \dots, l$.

From $\lambda > 0$ and sublinearity of F_1 , we have

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, v^1) + (x^1)^T w_i^1 - f_i(u^1, v^1) - (u^1)^T w_i^1 \right. \\ & \quad \left. + \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 \right] \\ & \geq F_1 \left(x^1, u^1; \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \right). \end{aligned} \quad (16)$$

By the duality constraint (8), it follows that

$$a = \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \in \mathbb{R}_+^{|J_1|}.$$

Thus, by condition (i) given in Theorem 1, we have

$$F_1(x^1, u^1; a) + (u^1)^T a \geq 0,$$

that is,

$$\begin{aligned} & F_1 \left(x^1, u^1; \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \right) \\ & \geq - (u^1)^T \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1]. \end{aligned} \quad (17)$$

From (10), (16) and (17), we get

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, v^1) + (x^1)^T w_i^1 - f_i(u^1, v^1) - (u^1)^T w_i^1 \right. \\ & \quad \left. + \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 \right] \geq 0. \end{aligned} \quad (18)$$

By second order F_2 -concavity of $f_i(x^1, \cdot) - \cdot^T z_i^1$ for fixed x^1 , with respect to $p_i^1 \in \mathbb{R}^{|K_1|}$, we have

$$\begin{aligned} & -f_i(x^1, v^1) + (v^1)^T z_i^1 + f_i(x^1, y^1) - (y^1)^T z_i^1 - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \\ & \geq F_2(v^1, y^1; -\nabla_{y^1} f_i(x^1, y^1) + z_i^1 - \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1). \end{aligned}$$

From $\lambda > 0$ and sublinearity of F_2 , we have

$$\sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) - (y^1)^T z_i^1 + (v^1)^T z_i^1 - f_i(x^1, v^1) \right]$$

$$\begin{aligned}
& -\frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \Big] \\
& \geq F_2 \left(v^1, y^1; \sum_{i=1}^l \lambda_i [-\nabla_{y^1} f_i(x^1, y^1) + z_i^1 - \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1] \right).
\end{aligned} \tag{19}$$

By the primal constraint (1), it follows that

$$c = -\sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1] \geq 0.$$

Thus, by condition (iii) given in Theorem 1, we have

$$F_2(v^1, y^1; c) + (y^1)^T c \geq 0,$$

that is,

$$\begin{aligned}
& F_2 \left(v^1, y^1; -\sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1] \right) \\
& \geq (y^1)^T \sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1].
\end{aligned} \tag{20}$$

From (3), (19) and (20), we get

$$\begin{aligned}
& \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) - (y^1)^T z_i^1 + (v^1)^T z_i^1 - f_i(x^1, v^1) \right. \\
& \left. - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \right] \geq 0.
\end{aligned} \tag{21}$$

Using $(x^1)^T w_i^1 \leq s(x^1 | C_i^1)$ and $(v^1)^T z_i^1 \leq s(v^1 | D_i^1)$, it follows from (18) and (21), that

$$\begin{aligned}
& \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) + s(x^1 | C_i^1) - (y^1)^T z_i^1 \right. \\
& \left. - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \right] \\
& \geq \sum_{i=1}^l \lambda_i \left[f_i(u^1, v^1) - s(v^1 | D_i^1) + (u^1)^T w_i^1 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 \right]
\end{aligned} \tag{22}$$

Similarly, using remaining hypotheses and conditions given in Theorem 1 and using constraints of the primal and dual problems for functions

$g_i(\cdot, v^2) + \cdot^T w_i^2$ and $g_i(x^2, \cdot) - \cdot^T z_i^2$, we get

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[g_i(x^2, y^2) + s(x^2 | C_i^2) - (y^2)^T z_i^2 \right. \\ & \quad \left. - \frac{1}{2} (p_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2 \right] \\ & \geq \sum_{i=1}^l \lambda_i \left[g_i(u^2, v^2) - s(v^2 | D_i^2) + (u^2)^T w_i^2 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} f_i(u^2, v^2) q_i^2 \right]. \end{aligned} \quad (23)$$

From (22) and (23), we get

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) + g_i(x^2, y^2) + s(x^1 | C_i^1) + s(x^2 | C_i^2) - (y^1)^T z_i^1 \right. \\ & \quad \left. - (y^2)^T z_i^2 - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 - \frac{1}{2} (p_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2 \right] \\ & \geq \sum_{i=1}^l \lambda_i \left[f_i(u^1, v^1) + g_i(u^2, v^2) - s(v^1 | D_i^1) - s(v^2 | D_i^2) + (u^1)^T w_i^1 \right. \\ & \quad \left. + (u^2)^T w_i^2 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2 \right], \end{aligned}$$

which is a contradiction to (15).

Hence

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \not\leq G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

□

Theorem 2 (Weak duality). *Let $(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda)$ be feasible for (SMP) and $(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda)$ be feasible for (SMD). Suppose there exist sublinear functionals F_1, F_2, G_1 and G_2 satisfying the following conditions:*

- (i) $F_1(x^1, u^1; a) + (u^1)^T a \geq 0$ if $a \geq 0$;
- (ii) $G_1(x^2, u^2; b) + (u^2)^T b \geq 0$ if $b \geq 0$;
- (iii) $F_2(v^1, y^1; c) + (y^1)^T c \geq 0$ if $c \geq 0$;
- (iv) $G_2(v^2, y^2; d) + (y^2)^T d \geq 0$ if $d \geq 0$.

Furthermore, assume that for $i = 1, 2, \dots, l$, $f_i(\cdot, v^1) + \cdot^T w_i^1$ is second order F_1 -pseudo-convex for fixed v^1 , with respect to $q_i^1 \in \mathbb{R}^{|J_1|}$, $f_i(x^1, \cdot) - \cdot^T z_i^1$ is second order F_2 -pseudo-concave for fixed x^1 , with respect to $p_i^1 \in \mathbb{R}^{|K_1|}$, $g_i(\cdot, v^2) + \cdot^T w_i^2$ is second order G_1 -pseudo-convex for fixed v^2 with respect

to $q_i^2 \in \mathbb{R}^{|J_2|}$ and $g_i(x^2, \cdot) - \cdot^T z_i^2$ is second order G_2 -pseudo-concave for fixed x^2 , with respect to $p_i^2 \in \mathbb{R}^{|K_2|}$.

Then

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \not\leq G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

Proof. Assume that the result is not true, that is,

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \leq G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

Then, since $\lambda > 0$, we have

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) + g_i(x^2, y^2) + s(x^1 | C_i^1) + s(x^2 | C_i^2) - (y^1)^T z_i^1 \right. \\ & \quad - (y^2)^T z_i^2 - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \\ & \quad \left. - \frac{1}{2} (p_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2 \right] \\ & < \sum_{i=1}^l \lambda_i \left[f_i(u^1, v^1) + g_i(u^2, v^2) - s(v^1 | D_i^1) - s(v^2 | D_i^2) + (u^1)^T w_i^1 \right. \\ & \quad \left. + (u^2)^T w_i^2 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2 \right] \end{aligned} \quad (24)$$

By the duality constraint (8), it follows that

$$a = \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \in \mathbb{R}_+^{|J_1|}.$$

Thus, by condition (i) given in Theorem 2, we have

$$F_1(x^1, u^1; a) + (u^1)^T a \geq 0,$$

that is,

$$\begin{aligned} & F_1 \left(x^1, u^1; \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \right) \\ & \geq - (u^1)^T \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \end{aligned} \quad (25)$$

From (10) and (25), we get

$$F_1 \left(x^1, u^1; \sum_{i=1}^l \lambda_i [\nabla_{u^1} f_i(u^1, v^1) + w_i^1 + \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1] \right) \geq 0. \quad (26)$$

From (26) and second order F_1 -pseudo-convexity of $f_i(\cdot, v^1) + \cdot^T w_i^1$ for fixed v^1 , with respect to $q_i^1 \in \mathbb{R}^{|J_1|}$, we get

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, v^1) + (x^1)^T w_i^1 - f_i(u^1, v^1) - (u^1)^T w_i^1 \right. \\ & \quad \left. + \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 \right] \geq 0. \end{aligned} \quad (27)$$

By primal constraint (1), it follows that

$$c = - \sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1] \geq 0.$$

Thus, by condition (iii) given in Theorem 2, we have

$$F_2(v^1, y^1; c) + (y^1)^T c \geq 0,$$

that is,

$$\begin{aligned} & F_2\left(v^1, y^1; - \sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1]\right) \\ & \geq (y^1)^T \sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1]. \end{aligned} \quad (28)$$

From (3) and (28), we get

$$F_2\left(v^1, y^1; - \sum_{i=1}^l \lambda_i [\nabla_{y^1} f_i(x^1, y^1) - z_i^1 + \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1]\right) \geq 0. \quad (29)$$

From (29) and the second order F_2 -pseudo-concavity of $f_i(x^1, \cdot) - \cdot^T z_i^1$ for fixed x^1 , with respect to $p_i^1 \in \mathbb{R}^{|K_1|}$, we have

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) - (y^1)^T z_i^1 + (v^1)^T z_i^1 - f_i(x^1, v^1) \right. \\ & \quad \left. - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \right] \geq 0. \end{aligned} \quad (30)$$

Using $(x^1)^T w_i^1 \leq s(x^1 | C_i^1)$ and $(v^1)^T z_i^1 \leq s(v^1 | D_i^1)$, it follows from (27) and (30), that

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) + s(x^1 | C_i^1) - (y^1)^T z_i^1 \right. \\ & \quad \left. - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 \right] \end{aligned} \quad (31)$$

$$\geq \sum_{i=1}^l \lambda_i \left[f_i(u^1, v^1) - s(v^1 | D_i^1) + (u^1)^T w_i^1 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 \right].$$

Similarly, using remaining hypotheses and conditions given in Theorem 2 and using constraints of the primal and dual problems for functions $g_i(\cdot, v^2) + \cdot^T w_i^2$ and $g_i(x^2, \cdot) - \cdot^T z_i^2$, we get

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[g_i(x^2, y^2) + s(x^2 | C_i^2) - (y^2)^T z_i^2 \right. \\ & \quad \left. - \frac{1}{2} (p_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2 \right] \end{aligned} \quad (32)$$

$$\geq \sum_{i=1}^l \lambda_i \left[g_i(u^2, v^2) - s(v^2 | D_i^2) + (u^2)^T w_i^2 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2 \right].$$

From (31) and (32), we get

$$\begin{aligned} & \sum_{i=1}^l \lambda_i \left[f_i(x^1, y^1) + g_i(x^2, y^2) + s(x^1 | C_i^1) + s(x^2 | C_i^2) - (y^1)^T z_i^1 \right. \\ & \quad \left. - (y^2)^T z_i^2 - \frac{1}{2} (p_i^1)^T \nabla_{y^1 y^1} f_i(x^1, y^1) p_i^1 - \frac{1}{2} (p_i^2)^T \nabla_{y^2 y^2} g_i(x^2, y^2) p_i^2 \right] \\ & \geq \sum_{i=1}^l \lambda_i \left[f_i(u^1, v^1) + g_i(u^2, v^2) - s(v^1 | D_i^1) - s(v^2 | D_i^2) + (u^1)^T w_i^1 \right. \\ & \quad \left. + (u^2)^T w_i^2 - \frac{1}{2} (q_i^1)^T \nabla_{u^1 u^1} f_i(u^1, v^1) q_i^1 - \frac{1}{2} (q_i^2)^T \nabla_{u^2 u^2} g_i(u^2, v^2) q_i^2 \right] \end{aligned}$$

which is a contradiction to (24).

Hence

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) \not\leq G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

□

Theorem 3 (Strong duality). Let $(\bar{x}^1, \bar{x}^2, \bar{y}^1, \bar{y}^2, \bar{z}^1, \bar{z}^2, \bar{p}^1, \bar{p}^2, \bar{\lambda})$ be a properly efficient solution for (SMP). Let $\bar{\lambda} = \lambda$ be fixed in (SMD). Suppose that the Hessian matrix $\nabla_{y^1}^2 f_i(\bar{x}^1, \bar{y}^1)$ is positive definite for $i = 1, 2, \dots, l$ and $\sum_{i=1}^l \lambda_i (p_i^1)^T [\nabla_{y^1} f_i - \bar{z}_i^1] \geq 0$; and $\nabla_{y^2}^2 g_i(\bar{x}^2, \bar{y}^2)$ is positive definite for $i = 1, 2, \dots, l$ and $\sum_{i=1}^l \lambda_i (p_i^2)^T [\nabla_{y^2} g_i - \bar{z}_i^2] \geq 0$; or $\nabla_{y^1}^2 f_i(\bar{x}^1, \bar{y}^1)$ is negative definite for $i = 1, 2, \dots, l$ and $\sum_{i=1}^l \lambda_i (p_i^1)^T [\nabla_{y^1} f_i - \bar{z}_i^1] \leq 0$;

and $\nabla_{y^2}^2 g_i(\overline{x^2}, \overline{y^2})$ is negative definite for $i = 1, 2, \dots, l$ and $\sum_{i=1}^l \lambda_i (p_i^2)^T [\nabla_{y^2} g_i - \bar{z}_i^2] \leq 0$. Also suppose that the sets

$$\{\nabla_{y^1} f_1 - \bar{z}_1^2 + \nabla_{y^1 y^1} f_1 \bar{p}_1^1, \dots, \nabla_{y^1} f_l - \bar{z}_l^2 + \nabla_{y^1 y^1} f_l \bar{p}_l^1\}$$

and

$$\{\nabla_{y^2} g_1 - \bar{z}_1^2 + \nabla_{y^2 y^2} g_1 \bar{p}_1^2, \dots, \nabla_{y^2} g_l - \bar{z}_l^2 + \nabla_{y^2 y^2} g_l \bar{p}_l^2\}$$

are linearly independent. Then there exists $w_i^1 \in C_i^1$ and $w_i^2 \in C_i^2$ such that

$$(\overline{x^1}, \overline{x^2}, \overline{y^1}, \overline{y^2}, \overline{w^1}, \overline{w^2}, \overline{q^1} = \overline{q^2} = 0, \overline{\lambda})$$

is feasible for (SMD) and

$$H(x^1, x^2, y^1, y^2, z^1, z^2, p^1, p^2, \lambda) = G(u^1, u^2, v^1, v^2, w^1, w^2, q^1, q^2, \lambda).$$

Moreover, if the generalized convexity hypotheses and conditions (i)-(iv) of Theorem 1 or 2 are satisfied, then $(\overline{x^1}, \overline{x^2}, \overline{y^1}, \overline{y^2}, \overline{w^1}, \overline{w^2}, \overline{q^1}, \overline{q^2}, \overline{\lambda})$ is a properly efficient solution for (SMD).

Proof. The proof follows the lines of the proof of Theorem 2 in Yang et al. [28] in light of the discussions above in this paper. \square

Remark 2. The converse duality theorem can also be established for the problems considered in this paper.

4. SPECIAL CASES

In this section, we consider some special cases of the problem (SMP) and (SMD) by choosing particular forms of the sublinear functionals and the compact convex sets involved in the problems.

If $|J_2| = |K_2| = 0$, then (SMP) and (SMD) reduce to the problems (P) and (D) studied by Yang et al. [28].

If $|J_2| = |K_2| = 0$, and $C_i^1 = C_i^2 = D_i^1 = D_i^2 = \{0\}$, $i = 1, 2, \dots, l$, then (SMP) and (SMD) reduce to the problems (P) and (D) studied by Suneja et al. [24].

If $C_i^1 = C_i^2 = D_i^1 = D_i^2 = \{0\}$, $i = 1, 2, \dots, l$, and $l = 1$, then (SMP) and (SMD) reduce to the problems (MP) and (MD) studied by Mishra [15].

If $|J_2| = |K_2| = 0$, and $l = 1$, then (SMP) and (SMD) reduce to the pair of problems studied by Hou and Yang [11].

If $|J_2| = |K_2| = 0$, $l = 1$, and $p^1 = p^2 = q^1 = q^2 = 0$ then (SMP) and (SMD) reduce to the problems (P1) and (D1) studied by Mond and Schechter [20].

If $l = 1$, $p^1 = p^2 = q^1 = q^2 = 0$ and $\nabla_2 f(x^1, y^1) = z^1$, then (SMP) and (SMD) reduce to the problems (MP) and (MD) studied by Yang et al. [27].

If $p_i^1 = p_i^2 = q_i^1 = q_i^2 = 0$ and $C_i^1 = C_i^2 = D_i^1 = D_i^2 = \{0\}$, $i = 1, 2, \dots, l$, and then (SMP) and (SMD) reduce to a general form of the problems (P1) and (D1) studied by Bector et al. [3].

5. CONCLUSION

In this paper, we have extended the results of Yang et al. [28], Suneja et al. [24], Hou and Yang [11] and Mond and Schechter [20] to mixed symmetric second order duality, the results of Bector et al. [3] to the case of non-differentiable problems and the results of Yang et al. [27] to the second order case. The results obtained in this paper can be extended to the class of functions introduced by Antczak [2] and Aghezzaf and Hachimi [1]. Moreover, these results can be further extended to higher order case as an extension of the results of Chen [4], to the case of complex functions and to the case of continuous-time problems as well. Some of these problems will be the subject of research of the present author in very near future.

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