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# ON SECOND-ORDER CONVERSE DUALITY FOR A NONDIFFERENTIABLE PROGRAMMING PROBLEM

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Certain shortcomings are described in the second order converse duality results in the recent work of (J. Zhang and B. Mond, Bull. Austral. Math. Soc. 55(1997) 29-44). Appropriate modifications are suggested.

# 1. Introduction

A second-order dual for a nonlinear programming problem was introduced by Mangasarian ([1]). Later, Mond [2] proved duality theorems under a condition which is called "second-order convexity". This condition is much simpler than that used by Mangasarian. Later, Mond and Weir [3] reformulated the second-order dual.

In [4], Mond considered the class of nondifferentiable mathematical programming problems

(P) minimize 
$$f(x) + (x^T B x)^{1/2}$$

(1) subject to 
$$g(x) \ge 0$$
,

where  $x \in \mathbb{R}^n$ , f and g are twice differentiable functions from  $\mathbb{R}^n$  into  $\mathbb{R}$  and  $\mathbb{R}^m$ , respectively, and B is an  $n \times n$  positive semi-definite (symmetric) matrix.

Recently, Zhang and Mond in [5] formulate a general second-order dual model for nondifferentiable programming problems (P):

(GD) maximize 
$$f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w - \frac{1}{2} p^T \left[ \nabla^2 f(u) - \nabla^2 \sum_{i \in I_0} y_i g_i(u) \right] p,$$

(2) subject to 
$$\nabla f(u) - +\nabla (y^T g(u)) + Bw + \nabla^2 f(u)p - \nabla^2 y^T g(u)p = 0,$$

(3) 
$$\sum_{i \in I_{\alpha}} y_i g_i(u) - \frac{1}{2} p^T \nabla^2 \sum_{i \in I_{\alpha}} y_i g_i(u) p \leqslant 0, \alpha = 1, 2, \dots, r,$$

$$(4) w^T B w \leqslant 1,$$

$$(5) y \geqslant 0,$$

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where  $u, w, p \in \mathbb{R}^n$ ,  $y \in \mathbb{R}^m$ ,  $I_{\alpha} \subset M = \{1, 2, ..., m\}$ ,  $\alpha = 0, 1, 2, ..., r$  with  $\bigcup_{\alpha=0}^r I_{\alpha} = M$  and  $I_{\alpha} \cap I_{\beta} = \emptyset$  if  $\alpha \neq \beta$ .

Zhang and Mond in [5] give weak, strong and converse duality theorems for first order and second order nondifferentiable dual models under generalised convexity. In particular, they prove the following second order converse duality Theorem.

**THEOREM 1.** Converse duality (see [5, Theorem 6]). Let  $(x^*, y^*, w^*, p^*)$  be an optimal solution of (GD) at which

- (A1) the  $n \times n$  Hessian matrix  $\nabla \left[ \nabla^2 f(x^*) \nabla^2 (y^{*T} g(x^*)) \right] p^*$  is positive or negative definite,
- (A2) the vectors

$$\left\{ \left[ \nabla^2 f(x^*) - \nabla^2 \sum_{i \in I_0} y^*_{i} g_i(x^*) \right]_j, \left[ \nabla^2 \sum_{i \in I_\alpha} y^*_{i} g_i(x^*) \right]_j, \alpha = 1, 2, \dots, r, j = 1, 2, \dots, n \right\}$$

are linearly independent, where  $[\cdot]_j$  denotes the  $j^{\text{th}}$  row.

If for all feasible (x, u, y, w, p),  $f(\cdot) - \sum_{i \in I_0} y_i g_i(\cdot) + (\cdot)^T Bw$  is second order pseudoinvex and  $\sum_{i \in I_\alpha} y_i g_i(\cdot)$ ,  $\alpha = 1, 2, \ldots, r$  is second order quasincave with respect to the same  $\eta$ , then  $x^*$  is an optimal solution to (P).

We note that the matrix  $\nabla \left[ \nabla^2 f(x^*) - \nabla^2 \left( y^{*T} g(x^*) \right) \right] p^*$  is positive or negative definite in the assumption  $(A_1)$  of Theorem 1, and the result of Theorem 1 implies  $p^* = 0$ , see [5, proof of Theorem 6]. It is obvious that the assumption and the result are inconsistent. In this note, we shall give appropriate modifications for the deficiency in Theorem 1.

## 2. Second order converse duality

In the section, we shall present a second order converse duality theorem which corrects Theorem 1.

**THEOREM 2.** (Converse duality.) Let  $(x^*, y^*, w^*, p^*)$  be an optimal solution of (GD) at which

(A1) for all  $\alpha = 1, 2, ..., r$ , either (a) the  $n \times n$  Hessian matrix  $\nabla^2 \sum_{i \in I_{\alpha}} y^*_{i} g_{i}(x^*)$  is positive definite and  $p^{*T} \nabla \sum_{i \in I_{\alpha}} y^*_{i} g_{i}(x^*) \geqslant 0$  or (b) the  $n \times n$  Hessian matrix  $\nabla^2 \sum_{i \in I_{\alpha}} y^*_{i} g_{i}(x^*)$  is negative definite and  $p^{*T} \nabla \sum_{i \in I_{\alpha}} y^*_{i} g_{i}(x^*) \leqslant 0$ ,

(A2) the vectors

$$\left\{ \left[ \nabla^2 f(x^*) - \nabla^2 \sum_{i \in I_0} y^*_{i} g_i(x^*) \right]_j, \left[ \nabla^2 \sum_{i \in I_0} y^*_{i} g_i(x^*) \right]_j, \alpha = 1, 2, \dots, r, j = 1, 2, \dots, n \right\}$$

are linearly independent, where

(A3) the vectors  $\left\{ \nabla \sum_{i \in I_{\alpha}} y^*_{i} g_{i}(x^*), \ \alpha = 1, 2, \dots, r \right\}$  are linearly independent.

If, for all feasible (x, u, y, w, p),  $f(\cdot) - \sum_{i \in I_0} y_i g_i(\cdot) + (\cdot)^T Bw$  is second order pseudoinvex and  $\sum_{i \in I_\alpha} y_i g_i(\cdot)$ ,  $\alpha = 1, 2, \ldots, r$  is second order quasincave with respect to the same  $\eta$ , then  $x^*$  is an optimal solution to (P).

PROOF: Since  $(x^*, y^*, w^*, p^*)$  is an optimal solution of (GD), by the generalised Fritz John necessary conditions, there exists,  $\tau_0 \in \mathbb{R}$ ,  $v \in \mathbb{R}^n$ ,  $\tau_\alpha \in \mathbb{R}$ ,  $\alpha = 1, 2, \ldots, r$ ,  $\beta \in \mathbb{R}$ ,  $\gamma \in \mathbb{R}^m$ , such that

$$\tau_0 \left\{ -\nabla f(x^*) + \sum_{i \in I_0} \nabla y_i^* g_i(x^*) - Bw^* + \frac{1}{2} p^{*T} \nabla \left[ \nabla^2 f(x^*) - \nabla^2 \sum_{i \in I_0} y_i^* g_i(x^*) p^* \right] \right\} + v^T \left\{ \nabla^2 f(x^*) - \nabla^2 y^{*T} g(x^*) + \nabla \left[ \nabla^2 f(x^*) p^* - \nabla^2 y^{*T} g(x^*) p^* \right] \right\}$$

(6) 
$$+ \sum_{\alpha=1}^{r} \tau_{\alpha} \left\{ \nabla \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) - \frac{1}{2} p^{*T} \nabla \left[ \nabla^{2} \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) p^{*} \right] \right\} = 0, y$$

(7) 
$$\tau_0 \left\{ g_i(x^*) - \frac{1}{2} p^{*T} \nabla^2 g_i(x^*) p^* \right\} - v^T \left\{ g_i(x^*) + \nabla^2 g_i(x^*) p^* \right\} - \gamma_i = 0, \ i \in I_0,$$
$$\tau_\alpha \left\{ g_i(x^*) - \frac{1}{2} p^{*T} \nabla^2 g_i(x^*) p^* \right\}$$

(8) 
$$-v^{T} \left\{ \nabla g_{i}(x^{*}) + \nabla^{2} g_{i}(x^{*}) p^{*} \right\} - \gamma_{i} = 0, i \in I_{\alpha}, \alpha = 1, 2, \dots, r,$$

(9) 
$$\tau_0 B x^* - v^T B - 2\beta^T (B w^*) = 0,$$

$$(\tau_0 p^* + v)^T \left\{ \nabla^2 f(x^*) - \nabla^2 \sum_{i \in I_0} y^*_{i} g_i(x^*) \right\}$$

(10) 
$$-\sum_{\alpha=1}^{r} (\tau_{\alpha} p^* + v) T \left\{ \nabla^2 \sum_{i \in I_{\alpha}} y^*_{i} g_i(x^*) \right\} = 0,$$

(11) 
$$\tau_{\alpha} \left\{ \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(x^{*}) - \frac{1}{2} p^{*T} \nabla^{2} \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(x^{*}) p^{*} \right\} = 0, \ \alpha = 1, 2, \dots, r,$$

$$\beta(w^*Bw^* - 1) = 0,$$

$$\gamma^T y^* = 0,$$

$$(14) (\tau_0, \tau_1, \tau_2, \dots, \tau_r, \beta, \gamma) \geqslant 0,$$

(15) 
$$(\tau_0, \tau_1, \tau_2, \dots, \tau_r, \beta, \gamma, v) \neq 0.$$

Because of Assumption (A2), (10) gives

(16) 
$$\tau_{\alpha} p^* + v = 0 \quad \alpha = 0, 1, 2, \dots, r.$$

Multiplying (8) by  $y_i^*$ ,  $i \in I_\alpha$ ,  $\alpha = 1, 2, ..., r$  and using (11), we have

$$\tau_{\alpha} \left\{ y_{i}^{*} g_{i}(x^{*}) - \frac{1}{2} p^{*T} \nabla^{2} y_{i}^{*} g_{i}(x^{*}) p^{*} \right\}$$
$$- v^{T} \left\{ \nabla y_{i}^{*} g(x^{*}) + \nabla^{2} y_{i}^{*} g(x^{*}) p^{*} \right\} = 0, i \in I_{\alpha}, \alpha = 1, 2, \dots, r,$$

thus

$$\tau_{\alpha} \left\{ \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(x^{*}) - \frac{1}{2} p^{*T} \sum_{i \in I_{\alpha}} \nabla^{2} y_{i}^{*} g_{i}(x^{*}) p^{*} \right\} - v^{T} \left\{ \sum_{i \in I_{\alpha}} \nabla y_{i}^{*} g(x^{*}) + \sum_{i \in I_{\alpha}} \nabla^{2} y_{i}^{*} g(x^{*}) p^{*} \right\} = 0, \alpha = 1, 2, \dots, r.$$

From (11), it follows that

(17) 
$$v^{T} \left\{ \sum_{i \in I_{\alpha}} \nabla y_{i}^{*} g(x^{*}) + \sum_{i \in I_{\alpha}} \nabla^{2} y_{i}^{*} g(x^{*}) p^{*} \right\} = 0, \alpha = 1, 2, \dots, r.$$

Using (2) in (6), we have

$$(\tau_{\alpha}p^{*} + v)^{T} \left\{ \nabla^{2} f(x^{*}) - \nabla^{2} \sum_{i \in I_{0}} y^{*}_{i} g_{i}(x^{*}) + \nabla \left[ \nabla^{2} f(x^{*}) - \nabla^{2} \sum_{i \in I_{0}} y^{*}_{i} g_{i}(x^{*}) \right] p^{*} \right\}$$

$$- \sum_{\alpha=1}^{r} (\tau_{\alpha} p^{*} + v)^{T} \left\{ \nabla^{2} \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) + \nabla \left[ \nabla^{2} \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) \right] p^{*} \right\}$$

$$- \tau_{0} \left\{ \nabla \sum_{i \in M \setminus I_{0}} y^{*}_{i} g_{i}(x^{*}) + \nabla^{2} \sum_{i \in M \setminus I_{0}} y^{*}_{i} g_{i}(x^{*}) p^{*} \right\}$$

$$- \frac{1}{2} \tau_{0} p^{*T} \left\{ \nabla \left[ \nabla^{2} f(x^{*}) - \nabla^{2} \sum_{i \in I_{0}} y^{*}_{i} g_{i}(x^{*}) \right] p^{*} \right\}$$

$$+ \sum_{\alpha=1}^{r} \tau_{\alpha} \left\{ \nabla \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) + \nabla^{2} \left[ \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) \right] p^{*} \right\}$$

$$+ \sum_{\alpha=1}^{r} \frac{1}{2} \tau_{\alpha} p^{*T} \left\{ \nabla \left[ \nabla^{2} \sum_{i \in I} y^{*}_{i} g_{i}(x^{*}) \right] p^{*} \right\} = 0.$$

From (16), it follows that

$$\begin{split} \sum_{\alpha=1}^{r} (\tau_{\alpha} - \tau_{0}) \bigg\{ \nabla \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) + \nabla^{2} \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) p^{*} \bigg\} \\ + \frac{1}{2} v^{T} \bigg\{ \nabla \bigg[ \nabla^{2} f(x^{*}) - \nabla^{2} \sum_{i \in I_{0}} y^{*}_{i} g_{i}(x^{*}) \bigg] p^{*} - \nabla \bigg[ \nabla^{2} \sum_{i \in M \setminus I_{0}} y^{*}_{i} g_{i}(x^{*}) \bigg] p^{*} ) \bigg\} = 0. \end{split}$$

That is

(18) 
$$\sum_{\alpha=1}^{r} (\tau_{\alpha} - \tau_{0}) \left\{ \nabla \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) + \nabla^{2} \sum_{i \in I_{\alpha}} y^{*}_{i} g_{i}(x^{*}) p^{*} \right\}$$

$$+ \frac{1}{2}v^{T} \left\{ \nabla \left[ \nabla^{2} f(x^{*}) - \nabla^{2} y^{*T} g(x^{*}) \right] p^{*} \right\} = 0.$$

If for all  $\alpha = 0, 1, 2, \ldots, r$ ,  $\tau_{\alpha} = 0$ , then v = 0 from (16),  $\gamma = 0$  from (7) and (8), and  $\beta = 0$  from (9) and (12); that is,  $(\tau_0, \tau_1, \tau_2, \ldots, \tau_r, \beta, \gamma, v) = 0$ , contradicts (15). Thus, there exists an  $\overline{\alpha} \in \{0, 1, 2, \ldots, r\}$ , such that  $\tau_{\overline{\alpha}} > 0$ .

We claim that  $p^* = 0$ . Indeed, if  $p^* \neq 0$ , then (16) gives

$$(\tau_{\alpha} - \tau_{\overline{\alpha}})p^* = 0, \alpha = 1, 2, \dots, r, .$$

This implies  $\tau_{\alpha} = \tau_{\overline{\alpha}} > 0, \alpha = 1, 2, \dots, r$ . So, (17) yields

$$p^{*T} \left\{ \sum_{i \in I_{\alpha}} \nabla y_{i}^{*} g(x^{*}) + \sum_{i \in I_{\alpha}} \nabla^{2} y_{i}^{*} g(x^{*}) p^{*} \right\} = 0, \alpha = 1, 2, \dots, r,$$

which contradicts to assumption  $(A_1)$ . Hence,  $p^* = 0$ . Based on (16) and  $p^* = 0$ , we have v = 0. In view of (A3), (16),  $p^* = 0$  and  $\tau_{\overline{\alpha}} > 0$  for some  $\overline{\alpha} \in \{0, 1, 2, ..., r\}$ , (18) implies  $\tau_{\alpha} = \tau_{\overline{\alpha}} > 0$ . Now from (7) and (8), it follows that

(19) 
$$\tau_0 g_i(x^*) - \gamma_i = 0, \ i \in I_0,$$

(20) 
$$\tau_{\alpha} g_i(x^*) - \gamma_i = 0, \ i \in I_{\alpha}, \alpha = 1, 2, \dots, r,$$

Therefore  $g(x^*) \ge 0$  since  $\gamma \ge 0$  and  $\tau_{\alpha} > 0, \alpha = 0, 1, 2, ..., r$ . Thus,  $x^*$  is feasible for (P), and the objective functions of (P) and (GD) are equal.

Multiplying (19) by  $y_i^*$ ,  $i \in I_0$  and using (13) gives

$$\tau_0 y_i^* q_i(x^*) = 0, i \in I_0.$$

By  $\tau_0 > 0$ , it follows that

$$y^*_{i}g_i(x^*) = 0, i \in I_0.$$

Also, v = 0,  $\tau_0 > 0$  and (9) give

$$(22) Bx^* = (2\beta \tau_0) Bw^*.$$

Hence

(23) 
$$x^{*T}Bx^* = (x^{*T}Bx^*)^{1/2}(w^{*T}Bw^*)^{1/2}.$$

If  $\beta > 0$ , then (12) gives  $w^{*T}Bw^* = 1$ , and so (23) yields

$$x^{*T}Bw^* = (x^{*T}Bx^*)^{1/2}.$$

If  $\beta = 0$ , then (22) gives  $Bx^* = 0$ . So we still get

$$x^{*T}Bw^* = (x^{*T}Bx^*)^{1/2}.$$

Thus, in either case, we have

$$(24) x^{*T}Bw^* = (x^{*T}Bx^*)^{1/2}.$$

Therefore from (21), (24) and  $p^* = 0$ , we have

$$f(x^*) + (x^{*T}Bx^*)^{1/2} = f(x^*) - \sum_{i \in I_0} y^*_{i}g_i(x^*) + u^{*T}Bw^* - \frac{1}{2}p^{*T} \left[ \nabla^2 f(x^*) - \nabla^2 \sum_{i \in I_0} y^*_{i}g_i(x^*) \right] p^*.$$

If, for all feasible (x, u, y, w, p),  $f(\cdot) - \sum_{i \in I_0} y_i g_i(\cdot) + (\cdot)^T B w$  is second order pseudoinvex and  $\sum_{i \in I_\alpha} y_i g_i(\cdot)$ ,  $\alpha = 1, 2, \ldots, r$  is second order quasincave with respect to the same  $\eta$ , by [5, Theorem 4], then  $x^*$  is an optimal solution to (P).

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