# DUALITY FOR A NON-DIFFERENTIABLE PROGRAMMING PROBLEM

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A generalised dual to a non-differentiable programming problem is given and duality established under general convexity and invexity conditions. A second order dual is also given and duality theorems proved under generalised second order invexity conditions.

### 1. INTRODUCTION

In [11], Mond considered the class of non-differentiable mathematical programming problems

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

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

where f and g are twice differentiable functions from  $\mathbb{R}^n$  to  $\mathbb{R}$  and  $\mathbb{R}^m$  respectively, and B is an  $n \times n$  positive semi-definite (symmetric) matrix. Let  $x_0$  satisfy (1); Mond [11] defined the set

$$Z_{0} = \left\{ z \mid z^{T} \nabla g_{i}(x_{0}) \geq 0 \quad (\forall i \in Q_{0}), \text{ and} \\ z^{T} \nabla f(x_{0}) + z^{T} B x_{0} / \left(x_{0}^{T} B x_{0}\right)^{1/2} < 0, \text{ if } x_{0}^{T} B x_{0} > 0; \\ z^{T} \nabla f(x_{0}) + \left(z^{T} B z\right)^{1/2} < 0, \text{ if } x_{0}^{T} B x_{0} \doteq 0 \right\}$$

where  $Q_0 = \{i \mid g_i(x_0) = 0\}$ , and established the following necessary conditions for  $x_0$  to be an optimal solution to (P).

**PROPOSITION 1.** If  $x_0$  is an optimal solution of (P) and the corresponding set  $Z_0$  is empty, then there exist  $y \in \mathbb{R}^m$ ,  $y \ge 0$ , and  $w \in \mathbb{R}^n$  such that

$$y^T g(x_0) = 0, \quad \nabla y^T g(x_0) = \nabla f(x_0) + Bw, \quad w^T Bw \leq 1, \quad (x_0^T Bx_0)^{1/2} = x_0^T Bw.$$

(Mond and Schechter [12] gave a constraint qualification which assures that  $Z_0$  is empty. Additional constraint qualifications were given by Wolkowitz [18].)

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Using these necessary conditions, a Wolfe type dual problem [17] was formulated in [11]:

(WD)  
Maximise 
$$f(u) - y^T g(u) + u^T [\nabla y^T g(u) - \nabla f(u)]$$
  
subject to  $\nabla f(u) + Bw = \nabla y^T g(u)$   
 $w^T Bw \leq 1$   
 $y \geq 0.$ 

(WD) is a dual to (P) assuming that f is convex and g is concave.

Chandra, Craven and Mond [4] weakened the convexity requirements for duality by giving a Mond-Weir type dual [14]

(M-WD)  
Maximise 
$$f(u) + u^T [\nabla y^T g(u) - \nabla f(u)]$$
  
subject to  $\nabla f(u) - \nabla y^T g(u) + Bw = 0$   
 $y^T g(u) \leq 0$   
 $w^T Bw \leq 1$   
 $y \geq 0$ 

and established duality theorems assuming that  $f(.) + (.)^T Bw$  is pseudo-convex for all  $w \in \mathbb{R}^n$  and that  $y^T g$  is quasi-concave.

Mond and Smart [13] later generalised the results obtained by Mond [11] and Chandra, Craven and Mond [4] to invexity conditions ([3, 5, 7]). Bector and Chandra [2] recently presented two different second order duals to (P), which extended the results obtained by Mangasarian [8], Mond [10] and Mond and Weir [15] for second order duality and the results obtained by Mond [11], Mond and Weir [14] and Chandra, Craven and Mond [4] for first order duality.

In this paper, we propose a general Mond-Weir type dual [14] to (P) and establish the duality theorems under both convexity and invexity conditions. A general second order Mond-Weir dual [15] to (P) will also be proposed and duality results established under generalised second order invexity conditions [1].

We shall make use of the generalised Schwarz inequality ([6] and [16])

(2) 
$$(x^T B w) \leq (x^T B x)^{1/2} (w^T B w)^{1/2}$$

Note that equality holds if, for  $\lambda \ge 0$ ,  $Bx = \lambda Bw$ .

#### 2. DUALITY

We propose the following general dual (GD) to (P).

(GD) Maximise 
$$f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w$$

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

(4) 
$$\sum_{i\in I_{\alpha}}y_{i}g_{i}(u)\leq 0, \quad \alpha=1,2,\ldots,r$$

(5) 
$$w^T B w \leq 1$$
  
 $y \geq 0$ 

where  $I_{\alpha} \subseteq M = \{1, 2, \dots, m\}, \quad \alpha = 0, 1, 2, \dots, r$  with

$$\bigcup_{\alpha=0}^{\prime} I_{\alpha} = M \text{ and } I_{\alpha} \cap I_{\beta} = \phi \text{ if } \alpha \neq \beta.$$

**THEOREM 1.** (Weak Duality) Let x be feasible for (P) and (u, y, w) feasible for (GD). If, for all feasible (x, u, y, w),  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  is pseudo-invex and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is quasi-incave with respect to the same  $\eta$ , then

infimum  $(P) \ge$  supremum (GD).

**PROOF:** Since x is feasible for (P) and (u, y, w) is feasible for (GD), we have

$$\sum_{i\in I_{oldsymbol{lpha}}}y_ig_i(x)-\sum_{i\in I_{oldsymbol{lpha}}}y_ig_i(u)\geqslant 0, \quad lpha=1,2,\ldots,r$$

By the quasi-incavity of  $\sum_{i\in I_{\alpha}} y_i g_i$ ,  $\alpha = 1, 2, ..., r$ , it follows that

$$\eta(x,u)^T \nabla \sum_{i \in I_{\alpha}} y_i g_i(u) \ge 0, \quad \alpha = 1, 2, \ldots, r.$$

Hence

$$\eta(\boldsymbol{x}, \boldsymbol{u})^T \nabla \Big( \sum_{i \in M \setminus I_0} y_i g_i(\boldsymbol{u}) \Big) \ge 0,$$

then from (3), it follows that

$$\eta(\boldsymbol{x}, \boldsymbol{u})^T \Big[ \nabla f(\boldsymbol{u}) - \nabla \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) + B w \Big] \ge 0.$$

The pseudo-invexity of  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  then yields

$$f(x) - \sum_{i \in I_0} y_i g_i(x) + x^T B w \ge f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w$$

Thus

$$f(x) + x^T B w \geqslant f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w$$
, from  $y \geqslant 0$  and  $g(x) \geqslant 0$ .

Since  $w^T B w \leq 1$ , by the generalised Schwarz inequality (2), it follows that

$$f(\boldsymbol{x}) + (\boldsymbol{x}^T B \boldsymbol{x})^{1/2} \geq f(\boldsymbol{u}) - \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) + \boldsymbol{u}^T B \boldsymbol{w}.$$

**THEOREM 2.** (Strong Duality) If  $x_0$  is an optimal solution of (P) and the corresponding set  $Z_0$  is empty, then there exist  $y \in \mathbb{R}^m$  and  $w \in \mathbb{R}^n$  such that  $(x_0, y, w)$  is feasible for (GD) and the corresponding values of (P) and (GD) are equal. If, also,  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  is pseudo-invex for all  $w \in \mathbb{R}^n$  and  $\sum_{i \in I_\alpha} y_i g_i(.), \alpha = 1, 2, ..., r$  is quasi-incave with respect to the same  $\eta$ , then  $(x_0, y, w)$  is optimal for (GD).

**PROOF:** Since  $x_0$  is an optimal solution to (P) and the corresponding set  $Z_0$  is empty, then from Proposition 1, there exist  $y \in \mathbb{R}^m$  and  $w \in \mathbb{R}^n$  such that

$$y^{T}g(x_{0}) = 0, \ \nabla y^{T}g(x_{0}) = \nabla f(x_{0}) + Bw, \ w^{T}Bw \leq 1, \ (x_{0}^{T}Bx_{0})^{1/2} = x_{0}^{T}Bw, \ y \geq 0.$$

So,  $(x_0, y, w)$  is feasible for (GD) and the corresponding values of (P) and (GD) are equal. If  $f(.) - \sum_{i \in I_0} y_i g_i (.) + (.)^T B w$  is pseudo-invex for all  $w \in \mathbb{R}^n$  and  $\sum_{i \in I_\alpha} y_i g_i (.)$ ,  $\alpha = 1, 2, ..., r$ , is quasi-incave with respect to the same  $\eta$ , then from Theorem 1,  $(x_0, y, w)$  must be an optimal solution for (GD).

We now consider some special cases of the dual (GD) and Theorems 1 and 2.

If  $I_0 = M$ , then (GD) becomes (WD) and from Theorems 1 and 2, (WD) is a dual to (P) if  $f(.) - y^T g(.) + (.)^T Bw$  is pseudo-invex with respect to  $\eta$ .

In the case  $I_0 = \phi$  and  $I_{\alpha} = M$  (for some  $\alpha \in \{1, 2, ..., r\}$ ) then (GD) becomes (M-WD) and from Theorems 1 and 2, (M-WD) is a dual to (P) if  $f(.) + (.)^T Bw$  is pseudo-invex and  $y^T g$  is quasi-incave with respect to the same  $\eta$ . This extends the results obtained in [4] because pseudo-convex and quasi-concave functions are pseudo-invex and quasi-incave functions respectively.

If 
$$I_0 = \phi, I_1 = \{1\}, \dots, I_m = \{m\} \ (r = m)$$
, then (GD) becomes

(M-WD1)  
Maximise 
$$f(u) + u^T Bw$$
  
subject to  $\nabla f(u) - \nabla y^T g(u) + Bw = 0$   
 $y_i g_i(u) \leq 0, \quad i = 1, 2, ..., m$   
 $w^T Bw \leq 1$   
 $y \geq 0$ 

and (M-WD1) is a dual to (P) if  $f(.) + (.)^T Bw$  is pseudo-invex and each  $y_i g_i$ , i = 1, 2, ..., m is quasi-incave with respect to the same  $\eta$ . Note that if  $g_i$  is quasi-incave with respect to  $\eta$ ,  $y_i \ge 0$ , then  $y_i g_i$  is quasi-incave with respect to the same  $\eta$ ; thus (M-WD1) is a dual to (P) if  $f(.) + (.)^T Bw$  is pseudo-invex and each  $g_i$ , i = 1, 2, ..., m is quasi-incave with respect to the same  $\eta$ .

The following corollaries obviously hold because pseudo-convex and quasi-concave functions are, respectively, pseudo-invex and quasi-incave functions.

**COROLLARY 1.** (Weak Duality) Let x be feasible for (P) and (u, y, w) feasible for (GD). If, for all feasible (x, u, y, w),  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T B w$  is pseudo-convex and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is quasi-concave, then

infimum (P)  $\geq$  supremum (GD).

**COROLLARY 2.** (Strong Duality) If  $x_0$  is an optimal solution to (P) and the corresponding set  $Z_0$  is empty, then there exist  $y \in \mathbb{R}^m$  and  $w \in \mathbb{R}^n$  such that  $(x_0, y, w)$  is feasible for (GD) and the corresponding values of (P) and (GD) are equal. If, also,  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  is pseudo-convex for all  $w \in \mathbb{R}^n$  and  $\sum_{i \in I_\alpha} y_i g_i(.), \alpha = 1, 2, \ldots, r$  is quasi-concave, then  $(x_0, y, w)$  is an optimal for (GD).

**THEOREM 3.** (Converse Duality) Let  $(x^*, y^*, w^*)$  be optimal to (GD) at which the matrix

$$\nabla^2 f(x^*) - \nabla^2 y^{*T} g(x^*)$$

is positive or negative definite and the vectors

$$\left\{\sum_{i\in I_{oldsymbol{lpha}}}
abla y_{i}^{*}g_{i}(x^{*}), \quad lpha=1,2,\ldots,r
ight\}$$

are linearly independent. If, for all feasible (x, u, y, w),  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  is pseudo-convex and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is quasi-concave, or  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$ 

 $(.)^{T}Bw$  is pseudo-invex and  $\sum_{i\in I_{\alpha}} y_{i}g_{i}(.), \alpha = 1, 2, ..., r$  is quasi-incave with respect to the same  $\eta$ , then  $x^{*}$  is an optimal to (P).

PROOF: Since  $(x^*, y^*, w^*)$  is an optimal solution to (GD), by the generalised Fritz-John theorem [9], there exist  $\tau_0 \in R$ ,  $\nu \in R^n$ ,  $\tau_\alpha \in R$ ,  $\alpha = 1, 2, ..., r$ ,  $\beta \in R$  and  $\gamma \in R^m$  such that

(6)  
$$\tau_0 \left( -\nabla f(\boldsymbol{x}^*) + \sum_{i \in I_0} \nabla y_i^* g_i(\boldsymbol{x}^*) - B \boldsymbol{w}^* \right) + \nu^T \left( \nabla^2 f(\boldsymbol{x}^*) - \nabla^2 y^{*T} g(\boldsymbol{x}^*) \right) + \sum_{\alpha=1}^r \tau_\alpha \left( \sum_{i \in I_\alpha} \nabla y_i^* g_i(\boldsymbol{x}^*) \right) = 0$$

(7) 
$$\tau_0 g_i(\boldsymbol{x}^*) - \boldsymbol{\nu}^T \nabla g_i(\boldsymbol{x}^*) - \gamma_i = 0, \quad i \in I_0$$

(8) 
$$\nu^T \nabla g_i(x^*) - \tau_\alpha g_i(x^*) + \gamma_i = 0, \quad i \in I_\alpha, \ \alpha = 1, 2, \dots, r,$$

(9) 
$$\tau_0(Bx^*) - \nu^T B - 2\beta(Bw^*) = 0$$

(10) 
$$\tau_{\alpha}\left(\sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(\boldsymbol{x}^{*})\right)=0, \quad \alpha=1,2,\ldots,r$$

(11) 
$$\beta\left(w^{*T}Bw^*-1\right)=0$$

(12) 
$$\gamma^T y^* = 0$$

(13) 
$$(\tau_0, \tau_1, \ldots, \tau_{\alpha}, \beta, \gamma) \ge 0$$

(14) 
$$(\tau_0, \tau_1, \ldots, \tau_{\alpha}, \beta, \gamma, \nu) \neq 0.$$

Multiplying (8) by  $y_i^* \ge 0$ ,  $i \in I_{\alpha}$ ,  $\alpha = 1, 2, ..., r$  and using (12) yields

$$u^T 
abla y_i^* g_i(x^*) - au_lpha y_i^* g_i(x^*) = 0, \quad i \in I_lpha, \quad lpha = 1, 2, \dots, r.$$

Hence

$$\nu^T \sum_{i \in I_{\alpha}} \nabla y_i^* g_i(x^*) - \tau_{\alpha} \sum_{i \in I_{\alpha}} y_i^* g_i(x^*) = 0, \quad \alpha = 1, 2, \dots, r.$$

From (10), it follows that

(15) 
$$\nu^T \sum_{i \in I_{\alpha}} \nabla y_i^* g_i(\boldsymbol{x}^*) = 0, \quad \alpha = 1, 2, \dots, r.$$

Using the equation constraint (3), (6) becomes

(16) 
$$\sum_{\alpha=1}^{r} (\tau_{\alpha}-\tau_{0}) \Big( \sum_{i\in I_{\alpha}} \nabla y_{i}^{*}g_{i}(\boldsymbol{x}^{*}) \Big) + \nu^{T} \Big( \nabla^{2}f(\boldsymbol{x}^{*}) - \nabla^{2}y^{*T}g(\boldsymbol{x}^{*}) \Big) = 0.$$

Multiplying (16) by  $\nu$  and using (15) gives

$$\nu^T \Big( \nabla^2 f(x^*) - \nabla^2 y^{*T} g(x^*) \Big) \nu = 0.$$

By assuming that  $\nabla^2 f - \nabla^2 y^T g$  is positive or negative definite at  $(x^*, y^*, w^*)$  it follows that

(17) 
$$\boldsymbol{\nu}=\boldsymbol{0}.$$

Then (16) gives

(18) 
$$\sum_{\alpha=1}^{r} (\tau_{\alpha} - \tau_{0}) \Big( \sum_{i \in I_{\alpha}} \nabla y_{i}^{*} g_{i}(\boldsymbol{x}^{*}) \Big) = 0.$$

Since the vectors  $\left\{\sum_{i\in I_{\alpha}} \nabla y_i^* g_i(x^*), \alpha = 1, 2, \ldots, r\right\}$  are linearly independent, (18) then yields

(19) 
$$\tau_{\alpha}-\tau_{0}=0, \quad \alpha=1,2,\ldots,r.$$

If  $\tau_0 = 0$ , then  $\tau_{\alpha} = 0$ ,  $\alpha = 1, 2, ..., r$  from (19),  $\gamma = 0$  from (7) and (8), and  $\beta = 0$  from (9) and (11), but  $(\tau_0, \tau_1, \tau_2, ..., \tau_r, \nu, \gamma, \beta) = 0$  contradicts (14). So  $\tau_0 > 0$ . This gives  $\tau_{\alpha} > 0$ ,  $\alpha = 1, 2, ..., r$ . Then (7), (8), (13) and  $\tau_{\alpha} > 0$ ,  $\alpha = 0, 1, 2, ..., r$  yield  $g(\boldsymbol{x}^*) \ge 0$ . Therefore,  $\boldsymbol{x}^*$  is feasible for (P).

Multiplying (7) by  $y_i^*$ ,  $i \in I_0$  and using (12) gives

$$au_0 y_i^* g_i(x^*) = 0, \; i \in I_0.$$

Then from  $\tau_0 > 0$ , it follows that

(20) 
$$y_i g_i(x^*) = 0, \ i \in I_0.$$

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

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

Hence

(22) 
$$(x^{*T}Bw^{*}) = (x^{*T}Bx^{*})^{1/2}(w^{*T}Bw^{*})^{1/2}$$

If  $\beta > 0$ , then (11) gives  $w^{*T}Bw^* = 1$  and so (22) yields

$$\left(\boldsymbol{x}^{*T}\boldsymbol{B}\boldsymbol{w}^{*}\right)=\left(\boldsymbol{x}^{*T}\boldsymbol{B}\boldsymbol{x}^{*}\right)^{1/2}.$$

If  $\beta = 0$  then (21) gives  $Bx^* = 0$ . So we still get

$$\left(x^{*T}Bw^{*}\right) = \left(x^{*T}Bx^{*}\right)^{1/2}$$

Thus in either case, we obtain

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

Therefore from (20) and (23), we have

$$f(x^*) + (x^{*T}Bx^*)^{1/2} = f(x^*) - \sum_{i \in I_0} y_i^* g_i(x^*) + x^{*T}Bw^*.$$

If, for all feasible (x, u, y, w),  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T B w$  is pseudo-convex and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is quasi-concave, or  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T B w$  is pseudo-invex and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is quasi-incave with respect to the same  $\eta$ , then from Theorem 1 or Corollary 1,  $x^*$  is an optimal solution to (P).

## 3. SECOND ORDER DUALITY

In this section, we present a general non-differentiable second order Mond-Weir dual [15] to (P). We shall make use of the following definitions.

DEFINITION 1: [1] f is second order pseudo-invex if for all  $p \in \mathbb{R}^n$ , there exists an  $\eta(x, u)$  such that

$$\eta(x,u)^T \Big[ \nabla f(u) + \nabla^2 f(u) p \Big] \ge 0 \implies f(x) \ge f(u) - \frac{1}{2} p^T \nabla^2 f(u) p.$$

DEFINITION 2: [1] f is second order quasi-invex if for all  $p \in \mathbb{R}^n$ , there exists an  $\eta(x, u)$  such that

$$f(x) \leqslant f(u) - rac{1}{2}p^T 
abla^2 f(u)p \implies \eta(x,u)^T \Big[ 
abla f(u) + 
abla^2 f(u)p \Big] \leqslant 0.$$

A function g is said to be second order pseudo-incave or second order quasi-incave if -g is second order pseudo-invex and second order qausi-invex respectively.

The second order Mangasarian type [8] and Mond-Weir type [15] duals to (P) were regarded in [2] as the following problems:

(2MD)

Maximise 
$$f(u) - y^T g(u) + u^T B w - \frac{1}{2} p^T \nabla^2 \Big[ f(u) - y^T g(u) \Big] p$$
  
subject to  $\nabla f(u) - \nabla y^T g(u) + B w + \nabla^2 f(u) p - \nabla^2 y^T g(u) p = 0$   
 $w^T B w \leq 1$   
 $y \geq 0$ 

[8]

where  $u, w, p, \in \mathbb{R}^n$  and  $y \in \mathbb{R}^m$ .

Maximise 
$$f(u) + u^T B w - \frac{1}{2} p^T \nabla^2 f(u) p$$
  
subject to  $\nabla f(u) - \nabla y^T g(u) + B w + \nabla^2 f(u) p - \nabla^2 y^T g(u) p = 0$   
 $y^T g(u) - \frac{1}{2} p^T \nabla^2 y^T g(u) p \leq 0$   
 $w^T B w \leq 1$   
 $y \geq 0.$ 

Using the second order convexity conditions (called bonvexity in [2]), Bector and Chandra established duality theorems between (P) and (2MD) and (2M-WD), respectively.

Following Mond-Weir [15] we now propose a general second order dual (2GD) to (P)

$$\text{Maximise } f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w - \frac{1}{2} p^T \Big[ \nabla^2 f(u) - \nabla^2 \sum_{i \in I_0} y_i g_i(u) \Big] p$$

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

(25) 
$$\sum_{i\in I_{\alpha}}y_{i}g_{i}(u)-\frac{1}{2}p^{T}\nabla^{2}\sum_{i\in I_{\alpha}}y_{i}g_{i}(u)p\leqslant 0, \ \alpha=1,2,\ldots,r,$$

 $(26) w^T B w \leqslant 1$ 

 $(27) y \ge 0$ 

where  $I_{\alpha} \subseteq M = \{1, 2, \dots, m\}, \ \alpha = 0, 1, 2, \dots, r$  with

$$\bigcup_{\alpha=0}^{\prime} I_{\alpha} = M \text{ and } I_{\alpha} \cap I_{\beta} = \phi \text{ if } \alpha \neq \beta.$$

**THEOREM 4.** (Weak Duality) Let x be feasible for (P) and (u, y, w, p) feasible for (2GD). If, for all feasible (x, u, y, w, p),  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  is second order pseudo-invex, and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is second order quasi-incave with respect to the same  $\eta$ , then

infimum (P)  $\geq$  supremum (2GD).

**PROOF:** Since x is feasible for (P) and (u, y, w, p) is feasible for (2GD), we have

$$\sum_{i\in I_{\alpha}}y_ig_i(x)-\sum_{i\in I_{\alpha}}y_ig_i(u)-\frac{1}{2}p^T\nabla^2\sum_{i\in I_{\alpha}}y_ig_i(u)p\geq 0, \quad \alpha=1,2,\ldots,r$$

By the second order quasi-incavity of  $\sum_{i\in I_{\alpha}} y_i g_i$ ,  $\alpha = 1, 2, \ldots, r$ , it follows that

$$\eta(x,u)^T igg( 
abla \sum_{i \in I_{oldsymbol{lpha}}} y_i g_i(u) + 
abla^2 \sum_{i \in I_{oldsymbol{lpha}}} y_i g_i(u) p igg) \geqslant 0, \quad lpha = 1, 2, \dots, r.$$

Hence

(28) 
$$\eta(x,u)^T \left( \nabla \sum_{i \in M \setminus I_0} y_i g_i(u) + \nabla^2 \sum_{i \in M \setminus I_0} y_i g_i(u) p \right) \ge 0.$$

Then from (24), (28) yields

$$\eta(\boldsymbol{x},\boldsymbol{u})^T \left( \nabla f(\boldsymbol{u}) + \nabla^2 f(\boldsymbol{u}) p - \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) - \nabla^2 \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) p + B w \right) \ge 0.$$

Since  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T B w$  is second order pseudo-invex, it follows that

$$f(x) - \sum_{i \in I_0} y_i g_i(x) + x^T B w$$
  
$$\geq f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w - \frac{1}{2} p^T \nabla^2 \Big[ f(u) - \sum_{i \in I_0} y_i g_i(u) + u^T B w \Big] p.$$

Thus, from  $y \ge 0$ ,  $g(x) \ge 0$ , we have

$$(29) \quad f(\boldsymbol{x}) + \boldsymbol{x}^T B \boldsymbol{w} \geq f(\boldsymbol{u}) - \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) + \boldsymbol{u}^T B \boldsymbol{w} - \frac{1}{2} \boldsymbol{p}^T \Big[ \nabla^2 f(\boldsymbol{u}) - \nabla^2 \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) \Big] \boldsymbol{p}.$$

Since  $w^T B w \leq 1$ , by the generalised Schwarz inequality (2), (29) gives that

$$f(\boldsymbol{x}) + \left(\boldsymbol{x}^T B \boldsymbol{x}\right)^{1/2} \geq f(\boldsymbol{u}) - \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) + \boldsymbol{u}^T B \boldsymbol{w} - \frac{1}{2} p^T \Big[ \nabla^2 f(\boldsymbol{u}) - \nabla^2 \sum_{i \in I_0} y_i g_i(\boldsymbol{u}) \Big] p.$$

**THEOREM 5.** (Strong Duality) If  $x_0$  is an optimal solution to (P) and the corresponding set  $Z_0$  is empty, then there exist  $y \in \mathbb{R}^m$  and  $w \in \mathbb{R}^n$  such that  $(x_0, y, w, p = 0)$  is feasible for (2GD), and the corresponding values of (P) and (2GD) are

equal. If, for all feasible (x, u, y, p, w),  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T Bw$  is second order pseudoinvex, and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$  is second order quasi-incave, then  $(x_0, y, w, p = 0)$  is an optimal solution for (2GD).

**PROOF:** Since  $x_0$  is an optimal solution to (P) and the corresponding set  $Z_0$  is empty, then from Proposition 1, there exist  $y \in \mathbb{R}^m$  and  $w \in \mathbb{R}^n$  such that

So,  $(x_0, y, w, p = 0)$  is feasible for (2GD) and the corresponding values of (P) and (2GD) are equal. If  $f(.) - \sum_{i \in I_0} y_i g_i(.) + (.)^T B w$  is second order pseudo-invex for all  $w \in \mathbb{R}^n$  and  $\sum_{i \in I_\alpha} y_i g_i(.)$ ,  $\alpha = 1, 2, ..., r$ , is second order quasi-incave with respect to the same  $\eta$ , then from Theorem 4,  $(x_0, y, w, p = 0)$  must be an optimal solution for (2GD).

We now consider some special cases of (2GD) and Theorems 4 and 5.

If  $I_0 = M$ , then (2GD) becomes (2MD), and from Theorems 4 and 5, (2MD) is a second order dual to (P) if  $f(.) - y^T g(.) + (.)^T Bw$  is second order pseudo-invex, which extends the results obtained in [2] because second order pseudo-convex and second order quasi-concave are second order pseudo-invex and second order quasi-incave respectively [1].

If  $I_0 = \phi$  and  $I_{\alpha} = M$  (for some  $\alpha \in \{1, 2, ..., r\}$ ), then (2GD) becomes (2M-WD), and from Theorems 4 and 5, (2M-WD) is a second order dual to (P) if  $f(.)+(.)^T Bw$  is second order pseudo-invex and  $y^T g$  is second order quasi-incave, which extends the results obtained in [2].

We now assume that f and g are three times differentiable.

**THEOREM 6.** (Converse Duality) Let  $(x^*, y^*, w^*, p^*)$  be an optimal solution to (2GD) at which the matrix

$$abla ig[ 
abla^2 f(x^*) - 
abla^2 y^{*T} g(x^*) ig] p^*$$

is positive or negative definite and the vectors

$$\Big\{\Big[\nabla^2 f(\boldsymbol{x}^*) - \nabla^2 \sum_{i \in I_0} y_i^* g_i(\boldsymbol{x}^*)\Big]_j, \quad \Big[\nabla^2 \sum_{i \in I_\alpha} y_i^* g_i(\boldsymbol{x}^*)\Big]_j, \quad \alpha = 1, 2, \ldots, r, \quad j = 1, 2, \ldots, n\Big\}$$

are linearly independent, where  $\left[\nabla^2 f - \nabla^2 \sum_{i \in I_0} y_i g_i\right]_j$  is the j-th row of  $\nabla^2 f - \nabla^2 \sum_{i \in I_0} y_i g_i$  and  $\left[\nabla^2 \sum_{i \in I_\alpha} y_i g_i\right]_j$  is the j-th row of  $\nabla^2 \sum_{i \in I_\alpha} (y_i g_i)$ . If, for all feasible

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

PROOF: Since  $(x^*, y^*, w^*, p^*)$  is an optimal solution to (2GD), by the generalised Fritz-John theorem [9], there exist  $\tau_0 \in R$ ,  $\nu \in R^n$ ,  $\tau_\alpha \in R$ ,  $\alpha = 1, 2, ..., r$ ,  $\beta \in R$ and  $\gamma \in R^m$  such that

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

$$(30)$$

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

(31)  

$$\tau_{0}\left\{g_{i}(x^{*})-\frac{1}{2}p^{*T}\nabla^{2}g_{i}(x^{*})p^{*}\right\}-\nu^{T}\left\{\nabla g(x^{*})+\nabla^{2}g(x^{*})p^{*}\right\}-\gamma_{i}=0, \quad i\in I_{0},$$

$$-\nu^{T}\left\{\nabla g(x^{*})+\nabla^{2}g(x^{*})p^{*}\right\}+\tau_{\alpha}\left\{g_{i}(x^{*})-\frac{1}{2}p^{*T}\nabla^{2}g_{i}(x^{*})p^{*}\right\}-\gamma_{i}=0,$$
(32)  

$$i\in I_{\alpha}, \quad \alpha=1,2,\ldots,r,$$

(33) 
$$\tau_0 B \boldsymbol{x}^* - \boldsymbol{\nu}^T B - 2\beta (B \boldsymbol{w}^*) = 0$$

(34)

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

(35) 
$$\tau_{\alpha}\left\{\sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(x^{*})-\frac{1}{2}p^{*T}\nabla^{2}\sum_{\substack{i\in I_{\alpha}\\ i\in I_{\alpha}}}y_{i}^{*}g_{i}(x^{*})p^{*}\right\}=0, \quad \alpha=1,2,\ldots,r,$$

(36) 
$$\beta(w^{*T}Bw^*-1)=0$$

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

(38) 
$$(\tau_0, \tau_1, \ldots, \tau_r, \beta, \gamma) \ge 0$$

(39) 
$$(\tau_0, \tau_1, \ldots, \tau_r, \beta, \gamma, \nu) \neq 0.$$

Since

$$\left\{\left[\nabla^2 f(u) - \nabla^2 \sum_{i \in I_0} y_i g_i(u)\right]_j, \quad \left[\nabla^2 \sum_{i \in I_\alpha} y_i g_i(u)\right]_j, \quad \alpha = 1, 2, \ldots, r, \quad j = 1, 2, \ldots, n\right\}$$

are linearly independent at  $(x^*, y^*, w^*, p^*)$ , (34) then gives

(40) 
$$\tau_{\alpha}p^* + \nu = 0, \quad \alpha = 0, 1, 2, \dots, r$$

Multiplying (32) by  $y_i^*$ ,  $i \in I_{\alpha}$ ,  $\alpha = 1, 2, \ldots, r$  and using (37) yields

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

thus

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

From (35), it follows that

(41) 
$$\nu^T \Big\{ \sum_{i \in I_\alpha} \nabla y_i^* g_i(x^*) + \sum_{i \in I_\alpha} \nabla^2 y_i^* g_i(x^*) p^* \Big\} = 0, \quad \alpha = 1, 2, \ldots, r.$$

Using (24), (30) gives

$$\begin{aligned} (\tau_{\alpha}p^{*}+\nu)^{T} \Big\{ \nabla^{2}f(x^{*}) - \nabla^{2}\sum_{i\in I_{0}}y_{i}^{*}g_{i}(x^{*}) + \nabla \Big[ \nabla^{2}f(x^{*}) - \nabla^{2}\sum_{i\in I_{0}}y_{i}^{*}g_{i}(x^{*}) \Big] p^{*} \Big\} \\ &- \sum_{\alpha=1}^{r} (\tau_{\alpha}p^{*}+\nu)^{T} \Big\{ \nabla^{2} \Big[ \sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(x^{*}) \Big] + \nabla \Big[ \nabla^{2}\sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(x^{*}) \Big] p^{*} \Big\} \\ &- \tau_{0} \Big\{ \nabla \sum_{i\in M\setminus I_{0}}y_{i}^{*}g_{i}(x^{*}) + \nabla^{2}\sum_{i\in M\setminus I_{0}}y_{i}^{*}g_{i}(x^{*}) p^{*} \Big\} \\ &- \frac{1}{2}\tau_{0}p^{*T} \Big\{ \nabla \Big[ \nabla^{2}f(x^{*}) - \nabla^{2}\sum_{i\in I_{0}}y_{i}^{*}g_{i}(x^{*}) \Big] p^{*} \Big\} \\ &+ \sum_{\alpha=1}^{r} \tau_{\alpha} \Big\{ \nabla \sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(x^{*}) + \nabla^{2} \Big[ \sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(x^{*}) \Big] p^{*} \Big\} \\ &+ \sum_{\alpha=1}^{r} \frac{1}{2}\tau_{\alpha}p^{*T} \Big\{ \nabla \Big[ \nabla^{2}\sum_{i\in I_{\alpha}}y_{i}^{*}g_{i}(x^{*}) \Big] p^{*} \Big\} = 0. \end{aligned}$$

From (40), it follows that

$$\sum_{\alpha=1}^{r} (\tau_{\alpha} - \tau_{0}) \Big\{ \nabla \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(\boldsymbol{x}^{*}) + \nabla^{2} \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(\boldsymbol{x}^{*}) p^{*} \Big\} \\ + \frac{1}{2} \nu^{T} \Big\{ \nabla \Big[ \nabla^{2} f(\boldsymbol{x}^{*}) - \nabla^{2} \sum_{i \in I_{0}} y_{i}^{*} g_{i}(\boldsymbol{x}^{*}) \Big] p^{*} - \nabla \Big[ \nabla^{2} \sum_{i \in M \setminus I_{0}} y_{i}^{*} g_{i}(\boldsymbol{x}^{*}) \Big] p^{*} \Big\} = 0.$$

That is

(42) 
$$\sum_{\alpha=1}^{r} (\tau_{\alpha} - \tau_{0}) \Big\{ \nabla \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(x^{*}) + \nabla^{2} \sum_{i \in I_{\alpha}} y_{i}^{*} g_{i}(x^{*}) p^{*} \Big\} + \frac{1}{2} \nu^{T} \Big\{ \nabla \Big[ \nabla^{2} f(x^{*}) - \nabla^{2} y^{*T} g(x^{*}) \Big] p^{*} \Big\} = 0.$$

Multiplying (42) by  $\nu$  and using (41) yields

$$\nu^T \Big\{ \nabla \Big[ \nabla^2 f(\boldsymbol{x}^*) - \nabla^2 \boldsymbol{y}^{*T} g(\boldsymbol{x}^*) \Big] \boldsymbol{p}^* \Big\} \nu = 0.$$

By assuming that  $\nabla [\nabla^2 f(x^*) - \nabla^2 y^{*T} g(x^*)] p^*$  is positive or negative definite, it follows that

$$(43) \nu = 0,$$

so (40) becomes

$$au_{oldsymbol{lpha}} p^* = - 
u = 0, \quad oldsymbol{lpha} = 0, 1, 2, \dots, r_{oldsymbol{lpha}}$$

If  $\tau_{\alpha} = 0$ ,  $\alpha = 0, 1, 2, ..., r$ , we get  $\gamma = 0$  from (31) and (32), and  $\beta = 0$  from (33) and (36); but  $(\tau_0, \tau_1, ..., \tau_r, \beta, \gamma, \nu) = 0$  contradicts (39). Thus  $\tau_{\alpha} > 0$ ,  $\alpha = 0, 1, 2, ..., r$ ; this gives  $p^* = 0$ . Hence from (31) and (32), it follows that

(44) 
$$\tau_0 g_i(x^*) - \gamma_i = 0, \quad i \in I_0$$

(45) 
$$\tau_{\alpha}g_{i}(x^{*})-\gamma_{i}=0, \quad i\in I_{\alpha}, \quad \alpha=1,2,\ldots,r$$

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

Multiplying (44) by  $y_i$ ,  $i \in I_0$  and using (37) gives

$$au_0 y_i^* g_i(x^*) = 0, \quad i \in I_0.$$

By  $\tau_0 > 0$ , it follows that

(46) 
$$y_i^* g_i(x^*) = 0, \quad i \in I_0.$$

Also,  $\nu = 0$ ,  $\tau_0 > 0$  and (33) give

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

Hence

(48) 
$$(x^{*T}Bw^{*}) = (x^{*T}Bx^{*})^{1/2}(w^{*T}Bw^{*})^{1/2}.$$

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If  $\beta > 0$ , then (36) gives  $w^{*T}Bw^* = 1$ , and so (48) yields

$$\left(\boldsymbol{x^{*T}}\boldsymbol{B}\boldsymbol{w^{*}}\right) = \left(\boldsymbol{x^{*T}}\boldsymbol{B}\boldsymbol{x^{*}}\right)^{1/2}.$$

If  $\beta = 0$ , then (47) gives  $Bx^* = 0$ . So we still get

$$\left(\boldsymbol{x}^{*T}\boldsymbol{B}\boldsymbol{w}^{*}\right)=\left(\boldsymbol{x}^{*T}\boldsymbol{B}\boldsymbol{x}^{*}\right)^{1/2}$$

Thus, in either case, we have

(49) 
$$(x^{*T}Bw^{*}) = (x^{*T}Bx^{*})^{1/2}$$

Therefore from (46), (49) and  $p^* = 0$ , we have

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

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

Note that if p = 0, then (2GD) becomes (GD), (2MD) becomes (MD) and (2M-WD) becomes (M-WD). This means that second order duality implies first order duality, and so Theorems 4, 5 and 6 imply Theorems 1, 2 and 3 respectively.

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