

Higher-Order Duality for Minimax Fractional Type Programming Involving Symmetric Matrices

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Received September 1, 2011; revised October 16, 2011; accepted October 23, 2011

Abstract

Convexity and generalized convexity play important roles in optimization theory. With the development of programming problem, there has been a growing interest in the higher-order dual problem and a lot of related generalized convexities are given. In this paper, we give the convexity of $(F, \alpha, \rho, d, b, \phi)_\beta$ vector-pseudo-quasi-Type I and formulate a higher-order duality for minimax fractional type programming involving symmetric matrices, and give the weak, strong and strict converse duality theorems under the condition of higher-order $(F, \alpha, \rho, d, b, \phi)_\beta$ vector-pseudoquasi-Type I.

Keywords: Higher-Order $(F, \alpha, \rho, d, b, \phi)_\beta$ Vector-Pseudoquasi-Type I, Higher-Order Duality, Minimax Fractional Type Programming, Positive Semidefinite Symmetric Matrix

1. Introduction

In this paper, we focus on the following nondifferentiable minimax fractional programming problem:

$$\minsup_{x \in R^n, y \in Y} \frac{f(x, y) + (x^T Bx)^{\frac{1}{2}}}{h(x, y) - (x^T Cx)^{\frac{1}{2}}} \quad (P)$$

subject to $g(x) \leq 0, x \in R^n,$

where Y is a compact subset of $R^l, f, h: R^n \times R^l \rightarrow R$ and $g: R^n \rightarrow R^m$ are continuously differentiable functions on $R^n \times R^l$ and R^n , respectively, and

$$f(x, y) + (x^T Bx)^{\frac{1}{2}} \geq 0,$$

$h(x, y) - (x^T Cx)^{\frac{1}{2}} > 0, \forall (x, y) \in R^n \times R^l, B$ and C are two positive semidefinite $n \times n$ symmetric matrices.

When $B = C = 0, (P)$ is a differentiable minimax fractional programming problem.

The duality of programming problem involving symmetric matrix has been investigated widely. Schmitendorf [1] established necessary and sufficient optimality conditions for a particular case of the following problem (P^*) under convexity conditions.

$$\minsup_{y \in Y} f(x, y) + (x^T Bx)^{\frac{1}{2}} \quad (P^*)$$

subject to $g(x) \leq 0.$

Under the optimality conditions of [1], Tanimoto [2] defined a first-order dual problem of (P^*) , which generalized the duality theorems for convex minimax programming problems considered by Weir [3] and relaxed the convexity assumptions in the sufficient optimality of [1]. Mishra and Rueda [4] introduced generalized second-order type I functions and considered the minimax programming problem (P^*) involving those functions and established second-order duality theorems for problem (P^*) . Husian, Anurag Jaysural and Ahmad [5] established two types of second-order dual models for problem (P^*) , which extends some previously known results on minimax programming.

With the development of programming problem, there has been a growing interest in the higher-order dual problem. Mangasarian [6] first formulated a class of second and higher-order dual problems for a nonlinear programming problem involving twice differentiable functions. In [7], Zhang considered the following nondifferentiable mathematical programming problem:

$$\text{Minimize } f(x) + (x^T Bx)^{\frac{1}{2}} \quad (P^{**})$$

subject to $g(x) \geq 0,$

under higher-order invexity assumptions. Mishra and Rueda [8] generalized the results of Zhang [7] to higher-

order type I functions.

In [9], Ahmad, Husain and Sharma considered the nondifferentiable minimax programming problem (P^*) , and formulated a unified higher-order dual of (P^*) , and established weak, strong and strict converse duality theorems under higher-order (F, α, ρ, d) -Type I assumptions. In [10], Jayswal and Stancu-Minasian formulated the weak, strong and strict converse duality of (P^*) under generalized convexity of higher-order (F, α, ρ, d) -Type I.

For problem (P) , H. C. Lai and K. Tanaka gave the necessary and sufficient conditions under the conditions of pseudo-convex, strictly pseudo-convex and quasi-convex [11].

In this paper, we will establish a higher-order dual of (P) and give the weak, strong and strict converse duality theorems under $(F, \alpha, \rho, d, b, \phi)_\beta$ vector-pseudoquasi-Type I assumptions. The convexity conditions in this paper generalized the convexity in [8], and hence, presents an answer of a question raised in [10].

2. Preliminaries

Let R^n be the n -dimensional Euclidean space, R_+^n be its nonnegative orthant and X be an open subset of R^n . Let S be the set of all feasible solutions of (P) . Denote $M = \{1, 2, \dots, m\}$. For each $(x, y) \in S \times Y$, we define

$$J(x) = \{j \in M : g_j(x) = 0\}$$

$$Y(x) = \left\{ y \in Y : \frac{f(x, y) + (x^T Bx)^{\frac{1}{2}}}{h(x, y) - (x^T Cx)^{\frac{1}{2}}} \right.$$

$$\left. = \sup_{z \in Y} \frac{f(x, z) + (x^T Bx)^{\frac{1}{2}}}{h(x, z) - (x^T Cx)^{\frac{1}{2}}} \right\}$$

$$K(x) = \left\{ (s, t, \tilde{y}) \in N \times R_+^s \times R^{ls} : 1 \leq s \leq n+1, \right.$$

$$t = (t_1, t_2, \dots, t_s) \in R_+^s \text{ with}$$

$$\left. \sum_{i=1}^s t_i = 1, \tilde{y} = (\bar{y}_1, \bar{y}_2, \dots, \bar{y}_s) \text{ and } \bar{y}_i \in Y(x), i = 1, 2, \dots, s \right\}$$

Definition 1: A function $F : X \times X \times R^n \rightarrow R$ is said to be sublinear in its third argument, if $\forall x, \bar{x} \in X$,

- 1) (subadditivity) $\forall a_1, a_2 \in R^n$,

$$F(x, \bar{x}; a_1 + a_2) \leq F(x, \bar{x}; a_1) + F(x, \bar{x}; a_2);$$
- 2) (positive homogeneous) $\forall \alpha \in R_+, a \in R^n$,

$$F(x, \bar{x}; \alpha a) = \alpha F(x, \bar{x}; a).$$

Let $f : R^n \times R^l \rightarrow R$, $\rho = (\rho_1, \rho_2) \in R^2$, $\beta \in R^n$, $\alpha_1, \alpha_2 : X \times X \rightarrow R_+ \setminus \{0\}$, $b, d : X \times X \rightarrow R$,

$e : R^n \rightarrow R$ and $\phi : R \rightarrow R$. Let $w : R^n \times R^l \times R^n \rightarrow R$, $l : R^n \times R^n \rightarrow R$ and $k : R^n \times R^n \rightarrow R^m$ be three differentiable functions. We assume that F is a sublinear functional throughout this paper.

Definition 2: (f, e) is said to be higher-order $(F, \alpha, \rho, d, b, \phi)_\beta$ vector-pseudoquasi-Type I at $\bar{x} \in X$ with respect to $p \in R^n$, if for all $x \in S$ and $\bar{y} \in Y(x)$,

$$b(x, \bar{x})\phi \left[f(x, \bar{y}) + x^T \beta - f(\bar{x}, \bar{y}) - \bar{x}^T \beta - w(\bar{x}, \bar{y}, p) \right.$$

$$\left. + p^T \nabla_p w(\bar{x}, \bar{y}, p) \right] < 0$$

$$\Rightarrow F(x, \bar{x}, \alpha_1(x, \bar{x})(\nabla_p w(\bar{x}, \bar{y}, p) + \beta))$$

$$< -\rho_1 d^2(x, \bar{x})$$

$$- \left[e(\bar{x}) + l(\bar{x}, p) - p^T \nabla_p l(\bar{x}, p) \right] \leq 0$$

$$\Rightarrow F(x, \bar{x}, \alpha_2(x, \bar{x})(\nabla_p l(\bar{x}, p))) \leq -\rho_2 d^2(x, \bar{x}).$$

Definition 3: (f, e) is said to be strictly higher-order $(F, \alpha, \rho, d, b, \phi)_\beta$ vector-pseudoquasi-Type I at $\bar{x} \in X$ with respect to $p \in R^n$, if for all $\bar{x} \neq x \in S$ and $\bar{y} \in Y(x)$,

$$b(x, \bar{x})\phi \left[f(x, \bar{y}) + x^T \beta - f(\bar{x}, \bar{y}) - \bar{x}^T \beta - w(\bar{x}, \bar{y}, p) \right.$$

$$\left. + p^T \nabla_p w(\bar{x}, \bar{y}, p) \right] \leq 0$$

$$\Rightarrow F(x, \bar{x}, \alpha_1(x, \bar{x})(\nabla_p w(\bar{x}, \bar{y}, p) + \beta))$$

$$< -\rho_1 d^2(x, \bar{x})$$

$$- \left[e(\bar{x}) + l(\bar{x}, p) - p^T \nabla_p l(\bar{x}, p) \right] \leq 0$$

$$\Rightarrow F(x, \bar{x}, \alpha_2(x, \bar{x})(\nabla_p l(\bar{x}, p))) \leq -\rho_2 d^2(x, \bar{x}).$$

Obviously, when ϕ is subadditive function and satisfies $a \leq 0 \Rightarrow \phi(a) \leq 0$, higher-order $(F, \alpha, \rho, d, b, \phi)_{Bu}$ vector-pseudoquasi-Type I is the convexity condition of Theorem 3.1 in [10].

In the following section, we will use Lemma 1 and Lemma 2 which were given in [11].

Lemma 1: (Necessary Condition) If x^* is an optimal solution of problem (P) satisfying

$x^{*T} Bx^* > 0, x^{*T} Cx^* > 0$ and $\nabla g_j(x^*), j \in J(x^*)$ are linear independent, then there exist $(s^*, t^*, \bar{y}^*) \in K(x^*)$, $u^*, v^* \in R^n$ and $\mu^* \in R_+^m, \lambda^* \in R_+$ such that

$$\sum_{i=1}^{s^*} t_i^* \left\{ \nabla f(x^*, \bar{y}_i^*) - \lambda^* (\nabla h(x^*, \bar{y}_i^*)) + Bu^* + \lambda^* Cv^* \right\}$$

$$+ \nabla \sum_{j=1}^{m} \mu_j^* g_j(x^*) = 0 \tag{2.1}$$

$$f(x^*, \bar{y}_i^*) + (x^{*T} Bx^*)^{\frac{1}{2}} - \lambda^* h(x^*, \bar{y}_i^*) + \lambda^* (x^{*T} Cx^*)^{\frac{1}{2}} = 0,$$

$$i = 1, 2, \dots, s^* \tag{2.2}$$

$$\sum_{i=1}^m \mu_i^* g_i(x^*) = 0 \tag{2.3}$$

$$\sum_{j=1}^{s^*} t_j^* = 1, t_i^* \geq 0, i = 1, 2, \dots, s^* \tag{2.4}$$

$$\begin{cases} u^{*T} B u^* \leq 1, v^{*T} C v^* \leq 1, \\ x^{*T} B u^* = (x^{*T} B x^*)^{\frac{1}{2}}, \\ x^{*T} C v^* = (x^{*T} B x^*)^{\frac{1}{2}}. \end{cases} \tag{2.5}$$

Lemma 2: Let B be a positive semidefinite symmetric matrix of order n . Then for all $x, u \in R^n$,

$$x^T B u \leq (x^T B x)^{\frac{1}{2}} \times (u^T B u)^{\frac{1}{2}}.$$

The equality holds when $Bx = \gamma Bu$ for some $\gamma \geq 0$.

Evidently, if $(u^T B u)^{\frac{1}{2}} \leq 1$, we have

$$x^T B u \leq (x^T B x)^{\frac{1}{2}}.$$

3. Duality Model

We consider the following dual model (WD).

$$\max_{(s,t,\tilde{y}) \in K(z)} \sup_{(z,u,v,\lambda,\mu,p) \in H(s,t,\tilde{y})} \lambda,$$

where $H(s,t,\tilde{y})$ denote the set of all $(z,u,v,\lambda,\mu,p) \in R^n \times R^n \times R^n \times R_+ \times R_+ \times R^n$ satisfying

$$\sum_{i=1}^s t_i \nabla_p w(z, \bar{y}_i, p) + Bu + \lambda Cv + \sum_{j=1}^m \nabla_p (\mu_j k_j(z, p)) = 0 \tag{3.11}$$

$$u^T B u \leq 1, v^T C v \leq 1, \tag{3.12}$$

$$\sum_{i=1}^s t_i \{f(z, \bar{y}_i) - \lambda h(z, \bar{y}_i)\} + z^T B u + \lambda z^T C v, \tag{3.13}$$

$$+ \sum_{i=1}^s t_i [w(z, \bar{y}_i, p) - p^T \nabla_p w(z, \bar{y}_i, p)] \geq 0,$$

$$\sum_{j=1}^m \{ \mu_j g_j(z) + \mu_j k_j(z, p) - p^T \nabla_p (\mu_j k_j(z, p)) \} \geq 0. \tag{3.14}$$

If for a triplet $(s,t,\tilde{y}) \in K(z)$, the set $H(s,t,\tilde{y}) = \emptyset$, then we define the supremum over $H(s,t,\tilde{y})$ to be $-\infty$.

Next, we establish the duality of type (WD).

Theorem 3.1 (Weak Duality) Let x and $(z,u,v,\lambda,\mu,s,t,\tilde{y},p)$ be feasible solutions of (P) and (WD), respectively. Assume that

1) $\left[\sum_{i=1}^s t_i (f(\cdot, \bar{y}_i) - \lambda h(\cdot, \bar{y}_i)), \sum_{j=1}^m \mu_j g_j(\cdot) \right]$ is higher-order

$(F, \alpha, \rho, d, b, \phi)_{Bu + \lambda Cv}$ vector-pseudoquasi Type I at z ,

2) $\phi(a) \geq 0 \Rightarrow a \geq 0, b(x, z) > 0,$

$$\frac{\rho_1}{\alpha_1(x, z)} + \frac{\rho_2}{\alpha_2(x, z)} \geq 0.$$

Then

$$\sup_{y \in Y} \frac{f(x, y) + (x^T B x)^{\frac{1}{2}}}{h(x, y) - (x^T C x)^{\frac{1}{2}}} \geq \lambda.$$

Proof: From (3.14), we know that

$$-\left[\sum_{j=1}^m \{ \mu_j g_j(z) + \mu_j k_j(z, p) - p^T \nabla_p (\mu_j k_j(z, p)) \} \right] \leq 0,$$

then follows from 1) and $\alpha_2(x, z) > 0$, we have

$$F \left(x, z; \sum_{j=1}^m \nabla_p (\mu_j k_j(z, p)) \right) \leq -\frac{\rho_2}{\alpha_2(x, z)} d^2(x, z).$$

Since F is sublinear in its third argument, by (3.11) we can get

$$\begin{aligned} 0 &= F \left(x, z; \sum_{i=1}^s t_i \nabla_p w(z, \bar{y}_i, p) + Bu \right. \\ &\quad \left. + \lambda Cv + \sum_{j=1}^m \nabla_p (\mu_j k_j(z, p)) \right) \\ &= F \left(x, z; \sum_{i=1}^s t_i \nabla_p w(z, \bar{y}_i, p) + Bu + \lambda Cv \right) \\ &\quad \left. + F \left(x, z; \sum_{j=1}^m \nabla_p (\mu_j k_j(z, p)) \right) \right) \\ &\leq F \left(x, z; \sum_{i=1}^s t_i \nabla_p w(z, \bar{y}_i, p) + Bu + \lambda Cv \right) \\ &\quad - \frac{\rho_2}{\alpha_2(x, z)} d^2(x, z). \end{aligned}$$

Furthermore, by $\frac{\rho_1}{\alpha_1(x, z)} + \frac{\rho_2}{\alpha_2(x, z)} \geq 0$ and

$\alpha_1(x, z) > 0$, we have

$$\begin{aligned} &F \left(x, z; \alpha_1(x, z) \left(\sum_{i=1}^s t_i \nabla_p w(z, \bar{y}_i, p) + Bu + \lambda Cv \right) \right) \\ &\geq -\rho_1 d^2(x, z), \end{aligned}$$

which implies that

$$\begin{aligned} &b(x, z) \phi \left(\left(\sum_{i=1}^s t_i \{f(x, \bar{y}_i) - \lambda h(x, \bar{y}_i)\} + x^T B u + \lambda x^T C v \right) \right. \\ &\quad \left. - \left(\sum_{i=1}^s t_i \{f(z, \bar{y}_i) - \lambda h(z, \bar{y}_i)\} + z^T B u + \lambda (z^T C v) \right) \right. \\ &\quad \left. - \sum_{i=1}^s t_i [w(z, \bar{y}_i, p) - p^T \nabla_p w(z, \bar{y}_i, p)] \right) \\ &\geq 0. \end{aligned}$$

From 2) and (3.13), we can get

$$\begin{aligned} & \sum_{i=1}^s t_i \{f(x, \bar{y}_i) - \lambda h(x, \bar{y}_i)\} + x^T B u + \lambda x^T C v \\ & \geq \sum_{i=1}^s t_i \{f(z, \bar{y}_i) - \lambda h(z, \bar{y}_i)\} + z^T B u + \lambda(z^T C v) \\ & \quad + \sum_{i=1}^s t_i [w(z, \bar{y}_i, p) - p^T \nabla_p w(z, \bar{y}_i, p)] \\ & \geq 0. \end{aligned}$$

Therefore, following from (3.12) and Lemma 2,

$$\begin{aligned} & \sum_{i=1}^s t_i \left[f(x, \bar{y}_i) + (x^T B x)^{\frac{1}{2}} - \lambda \left(h(x, \bar{y}_i) - (x^T C x)^{\frac{1}{2}} \right) \right] \\ & \geq \sum_{i=1}^s t_i \{f(x, \bar{y}_i) - \lambda h(x, \bar{y}_i)\} + x^T B u + \lambda x^T C v \\ & \geq 0. \end{aligned}$$

Since $t = (t_1, t_2, \dots, t_s) \neq 0$, $t_i \geq 0, i = 1, 2, \dots, s$, $\bar{y}_i \in Y$

and $h(x, y) - (x^T C x)^{\frac{1}{2}} > 0, \forall (x, y) \in R^n \times R^l$, at least exists one $q \in \{1, 2, \dots, s\}$, such that

$$\frac{f(x, \bar{y}_q) + (x^T B x)^{\frac{1}{2}}}{h(x, \bar{y}_q) - (x^T C x)^{\frac{1}{2}}} \geq \lambda,$$

which implies that

$$\sup_{y \in Y} \frac{f(x, y) + (x^T B x)^{\frac{1}{2}}}{h(x, y) - (x^T C x)^{\frac{1}{2}}} \geq \lambda.$$

Theorem 3.2 (Strong duality) Let x^* be an optimal solution of (P) satisfying $x^{*T} B x^* > 0, x^{*T} C x^* > 0$ and let $\nabla g_j(x^*), j \in J(x^*)$ be linear independent. Assume that for any $i = 1, 2, \dots, s^*$

$$w(x^*, \bar{y}_i^*, 0) = 0,$$

$$\nabla_p w(x^*, \bar{y}_i^*, 0) = \nabla f(x^*, \bar{y}_i^*) - \lambda^* (\nabla h(x^*, \bar{y}_i^*)),$$

and for any $j \in J(x^*)$

$$k_j(x^*, 0) = 0, \quad \nabla_p k_j(x^*, 0) = \nabla g_j(x^*).$$

Then there exist $(s^*, t^*, \tilde{y}^*) \in K(x^*)$ and $(x^*, u^*, v^*, \lambda^*, \mu^*, p^*) \in H(s^*, t^*, \tilde{y}^*)$ such that $(x^*, u^*, v^*, \lambda^*, \mu^*, s^*, t^*, \tilde{y}^*, p^* = 0)$ is a feasible solution of (WD) and the two objectives have the same values. Furthermore, if the assumptions of weak duality hold for all feasible solutions of (P) and (WD), then $(x^*, u^*, v^*, \lambda^*, \mu^*, s^*, t^*, \tilde{y}^*, p^* = 0)$ is an optimal solutions of (WD).

Proof: Since x^* is an optimal solution of (P) satisfying $x^{*T} B x^* > 0, x^{*T} C x^* > 0$ and $\nabla g_j(x^*), j \in J(x^*)$ is

linearly independent, by Lemma 1, there exist $(s^*, t^*, \tilde{y}^*) \in K(x^*), u^*, v^* \in R^n$ and $\mu^* \in R_+^m, \lambda^* \in R_+$ such that

$$\sum_{i=1}^{s^*} t_i^* \left\{ \nabla f(x^*, \bar{y}_i^*) - \lambda^* (\nabla h(x^*, \bar{y}_i^*)) + B u^* + \lambda^* C v^* \right\} \tag{2.1}$$

$$+ \nabla \sum_{j=1}^m \mu_j^* g_j(x^*) = 0$$

$$\begin{aligned} & f(x^*, \bar{y}_i^*) + (x^{*T} B x^*)^{\frac{1}{2}} - \lambda^* h(x^*, \bar{y}_i^*) + \lambda^* (x^{*T} C x^*)^{\frac{1}{2}} = 0, \\ & i = 1, 2, \dots, s^* \end{aligned} \tag{2.2}$$

$$\sum_{i=1}^m \mu_j^* g_j(x^*) = 0 \tag{2.3}$$

$$\sum_{j=1}^{s^*} t_i^* = 1, t_i^* \geq 0, i = 1, 2, \dots, s^* \tag{2.4}$$

$$\begin{cases} u^{*T} B u^* \leq 1, v^{*T} C v^* \leq 1, \\ x^{*T} B u^* = (x^{*T} B x^*)^{\frac{1}{2}}, \\ x^{*T} C v^* = (x^{*T} C x^*)^{\frac{1}{2}}. \end{cases} \tag{2.5}$$

By (2.1) (2.2) (2.3) (2.5) and the conditions of theorem 3.2, we know that $(x^*, u^*, v^*, \lambda^*, \mu^*, p^* = 0) \in H(s^*, t^*, \tilde{y}^*)$, that is $(x^*, u^*, v^*, \lambda^*, \mu^*, s^*, t^*, \tilde{y}^*, p^* = 0)$ is an feasible solutions of (WD). Furthermore, (3.2) implies that $(x^*, u^*, v^*, \lambda^*, \mu^*, s^*, t^*, \tilde{y}^*, p^* = 0)$ is an optimal solutions of (WD).

Theorem 3.3 (Strict Converse Duality) Let x^* be a feasible solution of (P) and $(z^*, u^*, v^*, \lambda^*, \mu^*, p^*)$ be a feasible solution of (WD). Suppose that

1)'

$$\sup_{y^* \in Y} \frac{f(x^*, y^*) + (x^{*T} B x^*)^{\frac{1}{2}}}{h(x^*, y^*) - (x^{*T} C x^*)^{\frac{1}{2}}} = \lambda^*;$$

$$2)' \left[\sum_{i=1}^{s^*} t_i^* (f(\cdot, \bar{y}_i^*) - \lambda^* h(\cdot, \bar{y}_i^*)), \sum_{j=1}^m \mu_j^* g_j(\cdot) \right] \text{ is strictly}$$

higher-order $(F, \alpha, \rho, d, b, \phi)_{B u^* + \lambda^* C v^*}$ vector-pseudoquasi-Type I at z^* and;

$$3)' \phi(a) > 0 \Rightarrow a > 0, \quad b(x^*, z^*) > 0,$$

$$\frac{\rho_1}{\alpha_1(x^*, z^*)} + \frac{\rho_2}{\alpha_2(x^*, z^*)} \geq 0$$

Then

$$z^* = x^*;$$

that is z^* is an optimal solution of (WD).

Proof: Suppose that the contradiction is not true, that is $z^* \neq x^*$. Similar to the proof of Theorem 3.1 we

obtain

$$F\left(x^*, z^*; \alpha_1(x^*, z^*) \left(\sum_{i=1}^{s^*} t_i^* \nabla_p w(z^*, \bar{y}_i^*, p^*) + Bu^* + \lambda^* Cv^* \right) \right)$$

$$\geq -\rho_1 d^2(x^*, z^*),$$

which implies that

$$b(x^*, z^*) \phi \left(\left(\sum_{i=1}^{s^*} t_i^* \{ f(x^*, \bar{y}_i^*) - \lambda^* h(x^*, \bar{y}_i^*) \} + x^{*T} Bu^* + \lambda^* x^{*T} Cv^* \right) - \left(\sum_{i=1}^{s^*} t_i^* \{ f(z^*, \bar{y}_i^*) - \lambda^* h(z^*, \bar{y}_i^*) \} + z^{*T} Bu^* + \lambda^* (z^{*T} Cv^*) \right) - \sum_{i=1}^{s^*} t_i^* \left[w(z^*, \bar{y}_i^*, p^*) - p^{*T} \nabla_p w(z^*, \bar{y}_i^*, p^*) \right] \right) > 0.$$

From 3)' and (3.13), we can get

$$\sum_{i=1}^{s^*} t_i^* \{ f(x^*, \bar{y}_i^*) - \lambda^* h(x^*, \bar{y}_i^*) \} + x^{*T} Bu^* + \lambda^* x^{*T} Cv^* > \sum_{i=1}^{s^*} t_i^* \{ f(z^*, \bar{y}_i^*) - \lambda^* h(z^*, \bar{y}_i^*) \} + z^{*T} Bu^* + \lambda^* (z^{*T} Cv^*) + \sum_{i=1}^{s^*} t_i^* \left[w(z^*, \bar{y}_i^*, p^*) - p^{*T} \nabla_p w(z^*, \bar{y}_i^*, p^*) \right] \geq 0.$$

Therefore, following from (3.12) and Lemma 2,

$$\sum_{i=1}^{s^*} t_i^* \left[f(x^*, \bar{y}_i^*) + (x^{*T} Bx^*)^{\frac{1}{2}} - \lambda^* \left(h(x^*, \bar{y}_i^*) - (x^{*T} Cx^*)^{\frac{1}{2}} \right) \right] \geq \sum_{i=1}^{s^*} t_i^* \{ f(x^*, \bar{y}_i^*) - \lambda^* h(x^*, \bar{y}_i^*) \} + x^{*T} Bu^* + \lambda^* x^{*T} Cv^* > 0.$$

Since $t^* = (t_1^*, t_2^*, \dots, t_s^*) \neq 0$, $t_i^* \geq 0, i = 1, 2, \dots, s^*$, $\bar{y}_i^* \in Y$ and

$h(x^*, y^*) - (x^{*T} Cx^*)^{\frac{1}{2}} > 0, \forall (x^*, y^*) \in R^n \times R^l$, at least exists one $q^* \in \{1, 2, \dots, s\}$, such that

$$\frac{f(x^*, \bar{y}_{q^*}^*) + (x^{*T} Bx^*)^{\frac{1}{2}}}{h(x^*, \bar{y}_{q^*}^*) - (x^{*T} Cx^*)^{\frac{1}{2}}} > \lambda^*,$$

which implies that

$$\sup_{y^* \in Y} \frac{f(x^*, y) + (x^{*T} Bx^*)^{\frac{1}{2}}}{h(x^*, y) - (x^{*T} Cx^*)^{\frac{1}{2}}} > \lambda^*$$

which contradicts with 1)'.
Remark: If we take place condition 2)' of this theorem by

$$2)" \left[\sum_{i=1}^{s^*} t_i^* (f(\cdot, \bar{y}_i^*) - \lambda^* h(\cdot, \bar{y}_i^*)), \sum_{j=1}^m \mu_j^* g_j(\cdot) \right]$$

is higher-order $(F, \alpha, \rho, d, b, \phi)_{Bu^* + \lambda^* Cv^*}$ vector-pseudoquasi-Type

I at z^* , and take place condition 3)' by

$$3)" \phi(a) \geq 0 \Rightarrow a > 0, \quad b(x^*, z^*) > 0,$$

$\frac{\rho_1}{\alpha_1(x^*, z^*)} + \frac{\rho_2}{\alpha_2(x^*, z^*)} \geq 0$, the strict converse duality holds too.

4. Acknowledgements

This work is supported by Youth Foundation of Beijing University of Technology (X1006011201002).

5. References

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