

On a Control Problem Containing Support Functions

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Abstract

A control problem containing support functions in the integrand of the objective of the functional as well as in the inequality constraint function is considered. For this problem, Fritz John and Karush-Kuhn-Tucker type necessary optimality conditions are derived. Using Karush-Kuhn-Tucker type optimality conditions, Wolfe type dual is formulated and usual duality theorems are established under generalized convexity conditions. Special cases are generated. It is also shown that our duality results have linkage with those of nonlinear programming problems involving support functions.

Keywords

Control Problem, Support Function, Optimality Conditions, Generalized Convexity, Wolfe Type Duality, Nonlinear Programming Problem

1. Introduction

Optimal control theory, which is an extension of calculus of variations is a mathematical optimization method for deriving control policies. In essence, an optimal control is set of differential equations describing the path of the control variables that minimize the cost functional. Mond and Hanson [1] were the first to formulate a control problem as a mathematical programming problem and studied Wolfe type duality for the same under convexity of the function involved in the formulation. Subsequently a number of duality results for a control problem involving differentiable functions were obtained, for example, in the references [2]-[5]. There exist applications of optimal control with nondifferentiable terms which appear in the problem of friction. This motivated Chandra *et al.* [2] to study optimality and duality for a class of nondifferentiable control problem containing the square root of certain quadratic form in the integrand of the objective functional. The popularity of this type of

mathematical programming problem seems to originate from the fact that even though the objective functions and/or constraint functions are nonsmooth, a simple representation for the dual may be found. Non smooth mathematical programming theory deals with much more general functions by means of generalized subdifferential [6] and quasidifferential [7]. However, the square root of a positive semidefinite quadratic form and support function are of the few cases of a nondifferentiable function for which subdifferentials can explicitly be written.

In this research we introduce a control problem with a support function in the integrand of the objective functional and each inequality constraint function. Optimality conditions for this nondifferentiable control problem are derived and Wolfe type duality is investigated under pseudoconvexity. Special cases are generated. The linkage between our results and those of nonlinear programming problem containing support function is also indicated.

2. Control Problem and Preliminaries

We introduce the following control problem involving support functions:

$$(\text{CP}): \text{Minimize: } \int_I \left\{ f(t, x, u) + s(u(t)|K) \right\} dt.$$

Subject to

$$x(a) = \alpha, \quad x(b) = \beta \quad (1)$$

$$\dot{x}(t) = h(t, x, u), \quad t \in I \quad (2)$$

$$g^j(t, x, u) + s(x(t)|C^j) \leq 0, \quad t \in I, \quad j = 1, 2, \dots, m \quad (3)$$

where

1) $x: I \rightarrow R^n$ is a differentiable state vector function with its derivative \dot{x} and $u: I \rightarrow R^m$ is a smooth control vector function.

2) R^n denotes an n -dimensional Euclidean space and $I = [a, b]$ is a real interval, and

3) $f: I \times R^n \times R^m \rightarrow R$, $g^j: I \times R^n \times R^m \rightarrow R$, $j = 1, 2, \dots, m$ and $h: I \times R^n \times R^m \rightarrow R^n$ are continuously differentiable.

4) $s(x(t)|K)$ and $s(x(t)|C^j)$, $j = 1, 2, \dots, m$ are the support function of the compact set K and C^j ($j = 1, 2, \dots, m$) respectively.

Denote the partial derivatives of f by f_t , f_x and f_u ,

$$f_t = \frac{\partial f}{\partial t}, \quad f_x = \left[\frac{\partial f}{\partial x^1}, \frac{\partial f}{\partial x^2}, \dots, \frac{\partial f}{\partial x^n} \right], \quad f_u = \left[\frac{\partial f}{\partial u^1}, \frac{\partial f}{\partial u^2}, \dots, \frac{\partial f}{\partial u^m} \right],$$

where superscript denote the vector components. Further X represents the space of continuously differentiable state functions $x: I \rightarrow R^n$ such that $x(a) = 0$ and $x(b) = 0$ and is equipped with the norm $\|x\| = \|x\|_\infty + \|D_x\|_\infty$, and U , the space of piecewise continuous control vector functions $u: I \rightarrow R^m$ having the uniform norm $\|u\|_\infty$.

The differential Equation (2) with initial conditions expressed as $x(t) = x(a) + \int_a^t h(s, x(s), y(s)) ds$ $t \in I$ may

be written as $H_x = H(x, y)$, where $H: X \times U \rightarrow C(I, R^n)$, $C(I, R^n)$ being the space of continuous function from I to R^n defined as $H(x, y)(t) = h(t, x(t), y(t))$. In the derivation of these optimality condition, some constraint qualification to make the equality constraint locally solvable [2] is needed for this and hence, the Fréchet derivative of $D_x - H(x, u) = Q(x, u)$, (say) with respect to (x, u) , namely $Q' = Q'(x, u) = [D - H_x(x, u), -H_u(x, u)]$ are required to be subjective. We review some well known facts about a support function for easy reference. Let Γ be a compact convex set in R^n . Then the support function of Γ denoted by $s(x(t)|\Gamma)$ is defined as, $s(x(t)|\Gamma) = \text{Max} \{x(t)^T v(t) : v(t) \in \Gamma, t \in I\}$.

A support function, being convex and everywhere finite, has a subdifferential in the sense of convex analysis, that is, there exists z such that $s(y(t)|\Gamma) \geq s(x(t)|\Gamma) + z(t)^T (y(t) - x(t))$ for all x .

As in [8] the subdifferential of $s(x(t)|\Gamma)$ is given by $s(x(t)|\Gamma) = \{z(t) \in \Gamma : z(t)^T x(t) = s(x(t)|\Gamma)\}$. Let $N_\Gamma(x(t))$ be normal cone at a point $x(t) \in \Gamma$. Then $y(t) \in N_\Gamma(x)$ if and only if $s(y(t)|\Gamma) = x(t)^T y(t)$ or equivalently, $x(t)$ is in the subdifferential of s at $y(t)$.

3. Optimality Conditions

In this section, we derive necessary optimality conditions of both Fritz John and Karush-Kuhn-Tucker type for the control problem (CP) stated in the preceding section.

Theorem 1. (Fritz John Conditions): If (\bar{x}, \bar{u}) is an optimal solution of (CP) and the Fréchet derivative Q' is surjective, then there exist Lagrange multipliers $\alpha \in R$ and piecewise smooth $\lambda : I \rightarrow R^m$, $\mu : I \rightarrow R^n$, $z : I \rightarrow R^m$ and $\omega^j : I \rightarrow R^n$ such that

$$\alpha f_x(t, \bar{x}, \bar{u}) + \sum_{j=1}^m \lambda^j(t) (g_x^j(t, \bar{x}, \bar{u}) + \omega^j(t)) + \mu(t)^T h_x(t, \bar{x}, \bar{u}) + \dot{\mu}(t) = 0, \quad t \in I \quad (4)$$

$$\alpha f_u(t, \bar{x}, \bar{u}) + \lambda^T(t) g_u(t, \bar{x}, \bar{u}) + \mu(t)^T h_u(t, \bar{x}, \bar{u}) = 0, \quad t \in I \quad (5)$$

$$\sum_{j=1}^m \lambda^j(t) (g_x^j(t, \bar{x}, \bar{u}) + x^T(t) \omega^j(t)) = 0, \quad t \in I \quad (6)$$

$$\mu^T(t) z(t) = s(x(t)|K), \quad t \in I \quad (7)$$

$$x(t) \omega^j(t) = s(x(t)|C^j), \quad j = 1, 2, \dots, m, \quad t \in I \quad (8)$$

$$z(t) \in K, \quad \omega^j(t) \in C^j, \quad j = 1, 2, \dots, m \quad (9)$$

$$(\alpha, \lambda(t)) \geq 0, \quad t \in I \quad (10)$$

$$(\alpha, \lambda(t), \mu(t)) \neq 0, \quad t \in I. \quad (11)$$

Proof: The problem (CP) may be expressed in its abstract version as

$$\text{(ECD): Minimize } \phi(x, u) = F(x, u) + \psi(x)$$

subject to

$$D_x = H(x, u)$$

$$G(x, u) \in S$$

where $F(x, u) = \int_I f(t, x, u)$, $\psi(u) = \int_I s(u(t)|K) dt$, $G : (X \times U) \rightarrow C(I, R^m)$ is given by (for all $t \in I$),

$$G^j(x, u)(t) = g^j(t, x, u) + s(x(t)|C^j), \quad j = 1, 2, \dots, m \quad \text{and} \quad S = C_+(I, R^m); \text{ the nonnegative orthant of } C(I, R^n).$$

By the result of [9] it follows that there exist Lagrange multipliers $\alpha \in R$, $\rho \in S^*$ (the dual of S) and v in the dual space of $C_+(I, R^m)$ satisfying

$$0 \in \alpha \partial \phi(\bar{x}, \bar{u}) + \partial(\rho^T G(\bar{x}, \bar{u})) + \partial v^T(D\bar{x} - H(\bar{x}, \bar{u})) \quad (12)$$

$$\rho^T G(\bar{x}, \bar{u}) = 0, \quad (13)$$

$$(\alpha, \rho) \geq 0 \quad (14)$$

$$(\alpha, \rho, v) \neq 0. \quad (15)$$

The condition (12) reduces to

$$0 \in \{\alpha F_x(x, u)\} + \{\rho^T G_x^1(x, u)\} + \partial \rho^T G_x^2(x, u) + \{v^T(D - H_x(\bar{x}, \bar{u}))\} \quad (16)$$

$$0 \in \{\alpha F_u(\bar{x}, \bar{u})\} + \partial(\alpha \psi(u)) + \rho^T G_u^1(\bar{x}, \bar{u}) + \{v^T H_u(\bar{x}, \bar{u})\}. \quad (17)$$

Since f is continuously differentiable function of x and u , $F(x, u)$ is Fréchet differentiable with respect to (x, u) . The partial derivatives of F with respect to \bar{x} and u , denoted by $F_x(\bar{x}, \bar{u})$ and $F_u(\bar{x}, \bar{u})$ respectively, are given by

$$(\forall p \in X) \quad F_x(x, u)p = \int_I f_x(t, \bar{x}, \bar{u})p(t) + f_{\dot{x}}(t, \bar{x}, \bar{u})\dot{p}(t) dt \quad (18)$$

$$(\forall q \in X) \quad F_u(x, u)q = \int_I f_u(t, x, u)q(t) dt. \quad (19)$$

Similar results for g and h as for f can be given. Assume now subject to later validation that, $\rho \in S^*$ can be represented by measurable function $\lambda: I \rightarrow R^m$ with $\lambda = (\lambda^1, \lambda^2, \dots, \lambda^m)$ satisfying

$$(\forall \xi \in C(I, R^m)), \quad \langle \rho, \xi \rangle = \int_I \lambda(t)^T \xi(t) dt. \quad (20)$$

Define the convex function $\eta_t: R^n \rightarrow R$ by $\eta_t(u) = s(u|K)$. From [2] its subdifferential,

$$\partial \eta_t(v) = \{z | z \in K, \eta_t(u) = u^T z\}.$$

Now $Q(x) = \int_I \eta_t(u) dt$. From ([6], Theorem 3), we have

$$y \in \partial Q(\bar{u}) \Leftrightarrow \left\{ (\forall t \in I), \sigma(t) \in \partial \eta_t(\bar{u}), \langle y, v \rangle = \int_I \sigma(t)^T u(t) dt \right\} \quad (21)$$

with $\sigma: I \rightarrow R^n$ measurable, namely $\sigma(t)^T = z(t)$, $t \in I$ from (16).

Let $\xi(t, x(t)) = s(x(t)|C^{(i)})$, where $s(x(t)|C^{(i)})$ denotes the vector support function whose j^{th} component is $s(x(t)|C^{(j)})$. Then

$$\rho^T \xi(., x) = \int_I \lambda(t)^T \xi(t, x(t)) dt. \quad (22)$$

Denoted by ∂_c the Clarke generalized gradient [6] with respect to x . Then

$$\begin{aligned} \partial_c(\lambda(t)^T \xi(t, x(t))) &\subset \sum_{j=1}^m \partial_c(\lambda^j(t)(\xi^j(t, x(t)))) = \sum_{j=1}^m |\lambda^j(t)| \partial_c(\text{sgn}(\lambda^j(t)) \xi^j(t, x(t))) \\ &= \sum_{j=1}^m |\lambda^j(t)| \text{sgn}(\lambda^j(t)) \partial_c(\xi^j(t, x(t))). \end{aligned} \quad (23)$$

The above is possible by using the representation of $\partial_c(\cdot)$ as the convex hull of limit of points of gradients at smooth points near x . Here \sum denotes the algebraic sum of sets. Since $\xi^j(t, \cdot) = s(\cdot|C^j)$ is convex, we have for each $j \in \{1, 2, \dots, m\}$,

$$\partial_c(\cdot) \xi^j(t, x(t)) = \partial_c \xi^j(t, x(t)) = \{\omega^j(t) | \omega^j(t) \in C^j, \xi^j(t, x(t)) = \bar{x}(t) \omega^j(t), t \in I\}. \quad (24)$$

From [10], it implies that $q \in \partial(\rho^T \xi)(., x)$ if and only if there exists a measurable function $\rho: I \rightarrow R^m$ such that

$$(\forall t \in I), q(t) \in \partial_c \xi(t, x(t)); \quad (\forall p \in X), \langle q, p(t) \rangle = \int_I \lambda(t)^T \rho(t) p(t) dt.$$

Now

$$\begin{aligned} \partial_x(\rho^T G)(\bar{x}, \bar{u}) &= \{\theta^T p | \theta \in \partial(\rho^T G)(\bar{x})\} \\ &= \sum_{j=1}^m \int_I \lambda^j(t) \left[g_x^j(t, \bar{x}(t), \dot{\bar{x}}(t)) p(t) + g_{\dot{x}}^j(t, \bar{x}(t), \dot{\bar{x}}(t)) \dot{p}(t) + \omega^j(t)^T p(t) \right] dt. \end{aligned} \quad (25)$$

Consider,

$$\begin{aligned}
v(H_x - D)p &= \int_I \left(\mu(t)^T h_x p - \mu(t) \dot{p} \right) dt = \int_I \mu(t) h_x p dt - \mu(t) p \Big|_a^b + \int_I \dot{\mu}(t) p dt \\
&= \int_I \left(\mu(t)^T h_x + \dot{\mu}(t) \right) p dt \quad (\text{using } \mu(a) = 0 = \mu(b)).
\end{aligned} \tag{26}$$

Using (18), (25), (26), we have

$$\int_I \left[\alpha f(t, \bar{x}, \bar{u}) + \sum_{j=1}^m \lambda^j(t) \left(g_x^j(t, \bar{x}, \bar{u}) + \omega^j(t) \right) + \mu(t)^T h_x(t, \bar{x}, \bar{u}) + \dot{\mu}(t) \right] p(t) dt = 0. \tag{27}$$

Since the integral values for any $v \in X$, by Lemma 2 ([11], p. 500), it follows that

$$\alpha f_x(t, \bar{x}, \bar{u}) + \sum_{j=1}^m \lambda^j(t) \left(g_x^j(t, \bar{x}, \bar{u}) + \omega^j(t) \right) + \mu(t)^T h_x(t, \bar{x}, \bar{u}) + \dot{\mu}(t) = 0 \quad t \in I. \tag{28}$$

The cited lemma assumes that the expression in the square bracket of (27) is piecewise continuous, but this readily extends to measurable. This validates (4). On the basis of analysis needed to validate (28), we can easily establish

$$f_x(t, \bar{x}, \bar{u}) + z(t) + \lambda(t)^T g_u(t, \bar{x}, \bar{u}) + \mu(t)^T h_u(t) = 0, \quad t \in I.$$

Also $\sum_{j=1}^m \bar{\rho}^j G^j(\bar{x}) = 0$ along with $\bar{x}(t)^T \bar{\omega}^j(t) = (\bar{x}(t) | C^j)$ of (24) yields

$$\sum_{j=1}^m \bar{\lambda}^j(t) \left[g^j(t, \bar{x}(t), \dot{\bar{x}}(t)) + \bar{x}(t)^T \bar{\omega}^j(t) \right] v(t) dt = 0.$$

By the application of the above-cited lemma, this gives (6) i.e.

$$\sum_{j=1}^m \lambda^j(t) \left[g^j(t, \bar{x}(t), \dot{\bar{x}}(t)) + \bar{x}(t)^T \bar{\omega}^j(t) \right] = 0, \quad t \in I.$$

The remaining proof of the theorem easily follow on the lines of the proof of Theorem 4.1 of [2].

Hence the above analysis established the theorem fully.

Chandra et al. [2] pointed out if the optimal solution for (CP) is normal, then the Fritz John type optimal conditions reduce to the following Karush-Kuhn-Tucker optimal conditions:

Theorem 2: If (\bar{x}, \bar{u}) is an optimal solution and is normal and Q' is surjective, there exist piecewise smooth $\lambda^j : I \rightarrow R^m$, $j = 1, 2, \dots, m$ with $\lambda^T = (\lambda_1, \lambda_2, \dots, \lambda_m)$, $\mu : I \rightarrow R^n$, $z : I \rightarrow R^n$ and $\omega^j : I \rightarrow R^n$, $j = 1, 2, \dots, m$.

Such that

$$f_x(t, \bar{x}, \bar{u}) + \sum_{j=1}^m \lambda^j(t) \left(g_x^j(t, \bar{x}, \bar{u}) + \omega^j(t) \right) + \mu(t)^T h_x(t, \bar{x}, \bar{u}) = \dot{\mu}(t) \tag{29}$$

$$f_u(t, \bar{x}, \bar{u}) + \lambda^T(t) g_u(t, \bar{x}, \bar{u}) + \mu(t)^T h_u(t, \bar{x}, \bar{u}) = 0, \quad t \in I \tag{30}$$

$$\sum_{j=1}^m \lambda^j(t) \left(g_x^j(t, \bar{x}, \bar{u}) + x^T(t) \omega^j(t) \right) = 0, \quad t \in I \tag{31}$$

$$u(t)^T z(t) = s(x(t) | K) \tag{32}$$

$$x(t)^T \omega^j(t) = s(x(t) | C^j), \quad j = 1, 2, \dots, m \tag{33}$$

$$\lambda^j(t) \geq 0, \quad t \in I, \quad j = 1, 2, \dots, m \tag{34}$$

$$z(t) \in K, \quad \omega^j(t) \in C^j, \quad j = 1, 2, \dots, m. \tag{35}$$

4. Wolfe Type Duality

We propose the following dual as the Wolfe type dual and validate duality results amongst (CP) and (WCD).

(WCD): Maximize

$$\int_I \left\{ f(t, x, u) + u(t)^T z(t) + \sum_{j=1}^m \lambda^j(t) \left(g^j(t, x, u) + x(t)^T \omega^j(t) \right) + \mu(t)^T (h(t, x, u) - \dot{x}(t)) \right\} dt$$

subject to

$$x(a) = 0, \quad \alpha(b) = 0 \quad (36)$$

$$f_x(t, x, u) + \sum_{j=1}^m \lambda^j(t) (g_x^j(t, x, u) + \omega^j(t)) + \mu(t)^T h_x(t, x, u) + \dot{\mu}(t) = 0, \quad t \in I \quad (37)$$

$$f_u(t, x, u) + \lambda^T(t) g(t, x, u) + \mu(t)^T h_u(t, x, u) = 0, \quad t \in I \quad (38)$$

$$\lambda^i(t) \geq 0, \quad t \in I, \quad i = 1, 2, \dots, m \quad (39)$$

$$z(t) \in K, \quad \omega^j(t) \in C^j, \quad j = 1, 2, \dots, m. \quad (40)$$

Theorem 3 (Weak Duality): Assume that

- 1) (\bar{x}, \bar{u}) is feasibility for (CP)
- 2) $(x, u, \lambda, u, z, \omega^1, \omega^2, \dots, \omega^m)$ is feasible for (CD) and
- 3) for all feasible, $(\bar{x}, \bar{u}, x, u, \lambda, u, z, \omega^1, \omega^2, \dots, \omega^m)$

$$\int_I \left\{ f + \sum_{j=1}^m \lambda^j(t) \left(g^j(.,.) + (.)^T \omega^j(.) \right) + \mu^T(t) (h - \dot{x}(t)) \right\} dt$$

is pseudo convex in (x, u) for all $z(t) \in R^m$ and $\omega^j(t) \in R^n$, $j = 1, 2, \dots, m$.

Then

$$\inf(\text{CP}) \geq \sup(\text{WCD}).$$

Proof: Combining (37) and (38), we have

$$\int_I \left[(\bar{x} - x)^T \left\{ f_x + \sum_{j=1}^m \lambda^j(t) (g_x^j + \omega^j(t)) + \mu(t)^T (h_x - \dot{\mu}(t)) \right\} + (\bar{u} - u) \left\{ f_u + z(t) + \lambda(t)^T g_u + \mu(t)^T h_u \right\} \right] dt = 0.$$

By the pseudoconvexity hypothesis 3), this yields

$$\begin{aligned} & \int_I \left\{ f(t, \bar{x}, \bar{u}) + \bar{x}(t)^T z(t) + \sum_{j=1}^m \lambda^j(t) (g^j(t, \bar{x}, \bar{u})) + \bar{x}(t)^T \omega^j(t) + \mu(t)^T (h(t, \bar{x}, \bar{u}) - \dot{\bar{x}}(t)) \right\} dt \\ & \geq \int_I \left\{ f(t, x, u) + x(t)^T z(t) + \sum_{j=1}^m \lambda^j(t) (g^j(t, x, u)) + x(t)^T \omega^j(t) + \mu(t)^T (h(t, x, u) - \dot{x}(t)) \right\} dt. \end{aligned} \quad (41)$$

Since (\bar{x}, \bar{u}) is feasible for (CP), we have

$$h(t, \bar{x}, \bar{u}) = \dot{\bar{x}}, \quad t \in I$$

implying

$$\mu(t)^T (h(t, \bar{x}, \bar{u}) - \dot{\bar{x}}(t)) = 0, \quad t \in I$$

and

$$g^i(t, \bar{x}, \bar{u}) + s(\bar{x}(t) | C^j) \leq 0, \quad t \in I$$

implying

$$\Rightarrow \sum \lambda(t) \left(g^j(t, \bar{x}, \dot{u}) + s(x(t)|C^j) \right) \leq 0, \quad t \in I, \quad t \in I.$$

Since $\bar{x}(t)^T \omega^j(t) \leq s(x(t)|C^j)$, $t \in I$, we have

$$\sum_{j=1}^m \lambda^j(t) \left(g^j(t, \bar{x}, \bar{u}) + x(t)^T \omega^j(t) \right) \leq \sum_{j=1}^m \lambda^j(t) \left(g^j(t, \bar{x}, \bar{u}) + s(\bar{x}(t)|C^j) \right) \leq 0, \quad t \in I.$$

From (41), we have

$$\begin{aligned} & \int_I \left\{ f(t, \bar{x}, \bar{u}) + s(\bar{u}(t)|K) + \sum_{j=1}^m \lambda^j(t) \left(g^j(t, \bar{x}, \bar{u}) + s(\bar{x}(t)|C^j) + \bar{\mu}(t) \left(h(t, \bar{x}, \bar{u}) - \dot{\bar{x}} \right) \right) \right\} dt \\ & \geq \int_I \left\{ f(t, x, u) + \sum_{j=1}^m \lambda^j(t) \left(g^j(t, x, u) + x(t)^T \omega^j(t) + \mu(t) \left(h(t, x, u) - \dot{x}(t) \right) \right) \right\} dt. \end{aligned}$$

This implies

$$\int_I \left\{ f(t, \bar{x}, \bar{u}) + S(\bar{x}(t)|K) \right\} dt \geq \int_I \left\{ f(t, x, u) + \sum_{j=1}^m \lambda^j(t) \left(g^j(t, x, u) + x(t)^T \omega^j(t) + \mu(t) \left(h(t, x, u) - \dot{x}(t) \right) \right) \right\} dt.$$

That is,

$$\inf(\text{CP}) \geq \sup(\text{CD}).$$

Theorem 4 (Strong duality): If (\bar{x}, \bar{u}) is an optimal solution of (CP) and is normal, there exist piecewise smooth $\lambda: I \rightarrow R^m$ where $(\lambda = \lambda_1, \lambda_2, \dots, \lambda_m)$, $z: I \rightarrow R^n$, $\mu: I \rightarrow R^n$ and $\omega^j: I \rightarrow R^n$, $(j = 1, 2, \dots, m)$ such that $(\bar{x}, \bar{u}, z, \bar{\lambda}, \bar{\mu}, \bar{\omega}^1, \dots, \bar{\omega}^m)$ is feasible for (WCD) and the optimal values of the problem (CP) and (WCD) are equal. If also the hypotheses of Theorem 1 hold, then $(\bar{x}, \bar{u}, z, \bar{\lambda}, \bar{\mu}, \bar{\omega}^1, \dots, \bar{\omega}^m)$ is an optimal solution of the problem (WCD).

Proof: Since (\bar{x}, \bar{u}) is an optimal solution of (CP) and is normal, by Theorem 1, it implies that there exist piecewise smooth $\lambda^j: I \rightarrow R$, $j = 1, 2, \dots, m$, $u: I \rightarrow R^n$, $z: I \rightarrow R^n$ and $\omega^j: I \rightarrow R^n$ ($j = 1, 2, \dots, m$) such that conditions (4)-(10) of the theorem are satisfied. The conditions (4)-(6) together with (9) and (10) imply the feasibility of $(\bar{x}, \bar{u}, z, \bar{\lambda}, \bar{\mu}, \bar{\omega}^1, \dots, \bar{\omega}^m)$ for (WCD). The condition (6)-(8) yield the equality of objective functionals of the two problem. In view of this equality and the hypotheses of Theorem 3, the optimality of $(\bar{x}, \bar{u}, z, \bar{\lambda}, \bar{\mu}, \bar{\omega}^1, \dots, \bar{\omega}^m)$ for (WCD) is obtained.

Theorem 5: (Strict Converse Duality): Assume

(H₁): (\bar{x}, \bar{u}) is an optimal solution and is normal;

(H₂): $(\hat{x}, \hat{u}, \hat{\lambda}, \hat{z}, \hat{\omega}^1, \dots, \hat{\omega}^m, \hat{\mu})$ is an optimal solution;

(H₃): $\int_I \left[\sum_{j=1}^m \hat{\lambda}^j(t) \left(g^j(t, \dots) + (\cdot)^T \omega^j(t) \right) + \hat{u}(t) \left(h(t, \dots) - \dot{\hat{x}}(t) \right) \right] dt$ is strictly pseudo convex.

then $(\bar{x}, \bar{u}) = (\hat{x}, \hat{u})$, i.e. \hat{u} is an optimal solution of (CP).

Proof: Assume that $(\bar{x}, \bar{u}) \neq (\hat{x}, \hat{u})$. By Theorem 4, there exist piecewise smooth $\bar{\lambda}: I \rightarrow R^m$ with, $\bar{\lambda}(t) = (\lambda^1(t), \dots, \lambda^m(t))$, $\bar{z}(t) \in K$, $t \in I$, $\bar{\mu}(t) \in R^n$ and $\bar{\omega}^j = \bar{\omega}^j(t)$, $t \in I$, $j = 1, 2, \dots, m$ such that $(\bar{x}, \bar{u}, \bar{\lambda}, \bar{u}, \bar{z}, \bar{\omega}^1, \dots, \bar{\omega}^m)$ is an optimal to (CD) and

$$\begin{aligned} & \int_I \sum_{j=1}^m \hat{\lambda}^j(t) \left(g^j(t, \bar{x}, \bar{u}) + \bar{x}(t)^T \bar{\omega}^j(t) + \bar{\mu}(t)^T \left(h(t, \bar{x}, \bar{u}) - \dot{\bar{x}}(t) \right) \right) dt \\ & = \int_I \sum_{j=1}^m \hat{\lambda}^j(t) \left(g^j(t, \hat{x}, \hat{u}) + \hat{x}(t)^T \hat{\omega}^j(t) + \hat{\mu}(t)^T \left(h(t, \hat{x}, \hat{u}) - \dot{\hat{x}}(t) \right) \right) dt. \end{aligned}$$

From the feasibility of $(\hat{x}, \hat{u}, \hat{\lambda}, \hat{u}, \hat{z}, \hat{\omega}^1, \dots, \hat{\omega}^m)$ for (WCD), we have

$$\int_I \left\{ (\bar{x} - \hat{x})^T \left(f_x(t, \hat{x}, \hat{u}) + \sum_{j=1}^m \hat{\lambda}^j(t) (g_x^j(t, \hat{x}, \hat{u}) + \hat{\omega}^j(t)) + \mu(t)^T h_x(t, \hat{x}, \hat{u}) - \hat{\mu}(t) \right) \right. \\ \left. + (\bar{u} - \hat{u}) \left(f_u(t, \hat{x}, \hat{u}) + \hat{\mu}(t) \hat{z}(t) + \hat{\lambda}(t)^T g_u(t, \hat{x}, \hat{u}) + \hat{\mu}(t)^T h_u(t, \hat{x}, \hat{u}) \right) \right\} dt = 0.$$

This by strict pseudoconvexity hypothesis (H₃) yields,

$$\int_I \left[f(t, \bar{x}, \bar{u}) + \bar{u}(t)^T \hat{z}(t) + \sum_{j=1}^m \hat{\lambda}^j(t) (g^j(t, \bar{x}, \bar{u}) + \bar{x}(t) \hat{\omega}^j(t)) + \hat{\mu}(t)^T (h(t, \bar{x}, \bar{u}) - \dot{\bar{x}}(t)) \right] dt \\ \geq \int_I \left[f(t, \hat{x}, \hat{u}) + \hat{u}(t)^T \hat{z}(t) + \sum_{j=1}^m \hat{\lambda}^j(t) (g^j(t, \hat{x}, \hat{u}) + \hat{x}(t) \hat{\omega}^j(t)) + \hat{\mu}(t)^T (h(t, \hat{x}, \hat{u}) - \dot{\hat{x}}(t)) \right] dt \\ = \int_I \left[f(t, \bar{x}, \bar{u}) + \bar{u}(t)^T \bar{z}(t) + \sum_{j=1}^m \hat{\lambda}^j(t) (g^j(t, \bar{x}, \bar{u}) + \bar{x}(t)^T \bar{\omega}^j(t)) + \bar{\mu}(t)^T (h(t, \bar{x}, \bar{u}) - \dot{\bar{x}}(t)) \right] dt.$$

Since $\sum_{j=1}^m \bar{\lambda}^j(t) (g^j(t, \bar{x}, \bar{u}) + \bar{x}(t)^T \bar{\omega}^j(t)) = 0$, and $\bar{x}(t)^T \hat{\omega}(t) \leq s(\bar{x}(t)|C^j)$, this yields,

$$\int_I \left[f(t, \bar{x}, \bar{u}) + \bar{u}(t)^T \hat{z}(t) \right] dt > \int_I \left[f(t, \bar{x}, \bar{u}) + \bar{u}(t)^T \bar{z}(t) \right] dt \\ \Rightarrow \int_I \bar{u}(t) \hat{z}(t) dt > \int_I \bar{u}(t) \bar{z}(t) dt \\ \Rightarrow \int_I s(\bar{u}(t)|K) dt > \int_I s(\bar{u}(t)|K) dt.$$

This is absurd. Hence (\bar{x}, \bar{u}) is an optimal solution of (CP).

5. Converse Duality

The problem (WCD) can be written as the follows:

Maximize: $\psi(x, \mu, z, \lambda^1, \dots, \lambda^m, \omega^1, \dots, \omega^m)$.

Subject to $x(a) = 0, x(b) = 0$

$$\theta^1(t, x(t)u(t), \lambda^1(t), \dots, \lambda^m(t), \mu(t), \omega^1(t), \omega^2(t), \dots, \omega^m(t)) = 0, \quad t \in I.$$

$$\theta^2(t, x(t)u(t), z(t), \lambda^1(t), \dots, \lambda^m(t), \mu(t), \omega^1(t), \omega^2(t), \dots, \omega^m(t)) = 0, \quad t \in I$$

$$z(t) \in K, \quad t \in I$$

$$\omega^j(t) \in C^j, \quad t \in I, \quad j = 1, 2, \dots, m$$

$$\lambda(t) \geq 0, \quad t \in I, \quad j = 1, 2, \dots, m$$

where

$$\theta^1 = \theta^2(t, x, u, z, \lambda, \mu) = f_x + \sum_{j=1}^m \lambda^j(t) (g_x^j + \omega^j(t)) + \mu(t)^T h_x + \dot{\mu}(t),$$

$$\theta^2 = \theta^2(t, x, u, z, \lambda, \mu) = f_u + z + \lambda^T g_x + \mu^T h_u.$$

Consider $\theta^1(., x(.), u(.), \lambda(.), \omega_2^1(.), \dots, \omega^m(.), \mu(.))$ and $\theta^2(t, x(.), u(.), \lambda(.), z(.), \mu(.))$ as defining a mappings $Q^1 : X \times u \times \Lambda \times W^1 \times W^2 \times \dots \times W^m \times V \rightarrow B^1$ and $Q^2 : X \times u \times Z \times \Lambda \times V \rightarrow B^2$ respectively where Λ is the space of piecewise smooth λ , V is space of piecewise smooth μ , W^j is the space of piecewise of smooth W^j , $j = 1, 2, \dots, m$, B^1 and B^2 are Banach spaces. $\theta^1 = (x, u, \lambda, \mu, \omega^1, \omega^2, \dots, \omega^m)$ and $\theta^2 = (x, u, \lambda, \mu, z)$ with $\lambda = (\lambda^1, \dots, \lambda^m)$. Here some restrictions are required on the equality constraints. For

this it suffices that if the Fréchet derivatives

$$Q'^1 = (\theta_x^1(\cdot), \theta_u^1(\cdot), \theta_\lambda^1(\cdot), \theta_\mu^1(\cdot), \theta_{\omega^1}^1(\cdot), \dots, \theta_{\omega^m}^1(\cdot))$$

and

$$Q'^2 = (\theta_x^2(\cdot), \theta_u^2(\cdot), \theta_\lambda^2(\cdot), \theta_\mu^2(\cdot), \theta_{\omega^1}^2(\cdot), \dots, \theta_{\omega^m}^2(\cdot)),$$

have weak * closed range.

Theorem 6. (Converse Duality): Assume

(A₁): f , g and h are twice continuously differentiable.

(A₂): $(x, u, \lambda, \mu, z, \omega^1, \dots, \omega^m)$ is an optimal solution of (CP).

(A₃): Q'^1 and Q'^2 have weak * closed range.

(A₄): The matrix
$$\begin{pmatrix} f_{xx} + \lambda^T(t) g_{xx} + \mu(t)^T h_{xx} & f_{ux} + \lambda^T(t) g_{ux} + \mu(t)^T h_{ux} \\ f_{ux} + \lambda^T(t) g_{ux} + \mu(t)^T h_{ux} & f_{uu} + \lambda^T(t) g_{uu} + \mu(t)^T h_{uu} \end{pmatrix}$$
 is nonsingular.

Then \bar{x} is an optimal solution of (CP) and the optimal values of (CP) and (WCD) are equal.

Proof: Since $(x, u, \lambda, \mu, z, \omega^1, \dots, \omega^m)$ is an optimal solution of (WCD), by Theorem 1 there exists $\lambda_0 \in R$, and piecewise smooth functions $\theta: I \rightarrow R^m$, $\phi: I \rightarrow R^m$, and $\eta: I \rightarrow R^m$ such that

$$\begin{aligned} \alpha \left\{ f_x + \sum \lambda^j(t) (g_x^j + \omega^j(t)) + \mu(t)^T h_x + \dot{u}(t) \right\} + \theta(t)^T (f_{xx} + \lambda(t)^T g_{xx} + \mu(t)^T h_{xx}) \\ + \phi(t)^T (f_{ux} + \lambda^T(t) g_{ux} + \mu(t)^T h_{ux}) = 0, \quad t \in I. \end{aligned} \quad (42)$$

$$\begin{aligned} \alpha (f_u + z(t) + \lambda^T(t) g_u + \mu^T(t) h_u) + \theta(t)^T (f_{xu} + \lambda^T(t) g_{xu} + \bar{\mu}^T(t) h_{xu}) \\ + \phi(t)^T (f_{uu} + \lambda^T(t) g_{uu} + \bar{\mu}^T(t) h_{uu}) = 0, \quad t \in I. \end{aligned} \quad (43)$$

$$\alpha (g^j + x(t)^T \omega^j(t)) + \theta(t)^T (g_x^j + \omega^j(t)) + \phi(t)^T g_u^j + \eta^j(t) = 0, \quad t \in I. \quad (44)$$

$$\alpha (h - \dot{x}(t)) + \theta(t)^T h_x + \phi(t)^T h_u + \dot{\theta}(t) = 0, \quad t \in I \quad (45)$$

$$\alpha \lambda^j(t) + x(t) + \theta(t)^T \lambda^j(t) \in N_{c_j}(\omega^j(t)), \quad j = 1, 2, \dots, m. \quad (46)$$

$$\alpha \mu(t) + \phi(t) \in N_K(z) \quad (47)$$

$$\eta^T(t) \lambda(t) = 0, \quad t \in I \quad (48)$$

$$(\alpha, \theta(t), \phi(t), \eta(t)) \geq 0, \quad t \in I \quad (49)$$

$$(\alpha, \theta(t), \phi(t), \eta(t)) \neq 0, \quad t \in I. \quad (50)$$

Using (36) and (37) in (42) and (43) respectively, we obtain

$$\begin{aligned} \theta(t)^T (f_{xx} + \lambda(t)^T g_{xx} + \mu(t)^T h_{xx}) + \phi(t)^T (f_{ux} + \lambda^T(t) g_{ux} + \mu(t)^T h_{ux}) = 0, \\ \theta(t)^T (f_{xu} + \lambda^T(t) g_{xu} + \bar{\mu}^T(t) h_{xu}) + \phi(t)^T (f_{uu} + \lambda^T(t) g_{uu} + \bar{\mu}^T(t) h_{uu}) = 0. \end{aligned}$$

The equations can be combined in the matrix form as,

$$\begin{pmatrix} f_{xx} + \lambda^T(t) g_{xx} + \mu(t)^T h_{xx} & f_{ux} + \lambda^T(t) g_{ux} + \mu(t)^T h_{ux} \\ f_{ux} + \lambda^T(t) g_{ux} + \mu(t)^T h_{ux} & f_{uu} + \lambda^T(t) g_{uu} + \mu(t)^T h_{uu} \end{pmatrix} \begin{pmatrix} \theta(t) \\ \phi(t) \end{pmatrix} = 0, \quad t \in I.$$

This, due to the hypothesis (A₄) yields

$$\theta(t) = 0 = \phi(t), \quad t \in I. \quad (51)$$

Let $\alpha = 0$, then (44) implies $\eta(t) = 0$, $t \in I$, consequently we get $(\alpha, \eta(t), \theta(t), \phi(t) = 0)$, $t \in I$, contradicting (50), hence $\alpha > 0$.

The relations (44) together with (48) and (45) respectively imply

$$g^j + x(t)^T \omega^j(t) = \frac{-\eta^j(t)}{\alpha}, \quad t \in I \quad (52)$$

$$h' - \dot{x}(t) = 0, \quad t \in I. \quad (53)$$

From (52) and $\lambda^j(t) \geq 0$, $t \in I$, we have

$$\sum_{j=1}^m \lambda^j(t) (g^j + x(t)^T \omega^j(t)) = 0, \quad t \in I. \quad (54)$$

From (53) along with $\mu(t) \geq 0$, $t \in I$, we obtain

$$\mu(t)^T (h - \dot{x}(t)) = 0, \quad t \in I. \quad (55)$$

In view of (51) and definition of a normal cone (50) and (51), we have $x(t) \in N_{c^j}(\omega^j(t))$, $t \in I$, $j = 1, 2, \dots, m$ and $u(t) \in N_K(z(t))$ implying

$$x^T(t) \omega^j(t) = s(x(t) | C^j), \quad j = 1, 2, \dots, m$$

and

$$\bar{u}(t)^T z(t) = s(u(t) | K). \quad (56)$$

From (52) together with (56) and

$$\eta^j(t) \geq 0, \quad t \in I, \quad i = 1, 2, \dots, m \quad (57)$$

imply

$$g^j + s(x(t) | C^j) \leq 0, \quad t \in I. \quad (58)$$

From (53) and (57), the feasibility of \bar{x} for (CP) follows.

Consider

$$\begin{aligned} & \int_I \left(f(t, \bar{x}, \bar{u}) + \bar{u}^T z + \sum_{j=1}^m \bar{\lambda}^j (g^j(t, \bar{x}, \bar{u}) + \bar{x}^T \bar{\omega}^j) + \bar{\mu}(t)^T (h(t, \bar{x}, \bar{u}) - \bar{x}) \right) dt \\ &= \int_I \left(f(t, \bar{x}, \bar{u}) + \bar{u}(t)^T \bar{z}(t) \right) dt = \int_I f(t, \bar{x}, \bar{u}) + s(\bar{u}(t) | K) dt \end{aligned}$$

(by using (54), (55) and (56)).

This implies that the values of objective functionals of the problem are equal. Consequently in view of the hypothesis of Theorem 1 it implies that (\bar{x}, \bar{u}) is an optimal solution of (CP).

6. Special Cases

Let for $t \in I$. $B(t)$ and $D^j(t)$, ($j = 1, 2, \dots, m$) be positive semidefinite matrices and continuous on I . Then

$$(u(t)^T B(t) u(t))^{\frac{1}{2}} = s(u(t) | K),$$

where

$$K = \{B(t)z(t) | z(t)^T B(t)z(t) \leq 1, t \in I\}$$

and

$$\left(x(t)^T D^j(t) x(t)\right)^{\frac{1}{2}} = s(x(t)|C^j), \quad j=1,2,\dots,m$$

where

$$C^j = \left\{ D^j(t) \omega^j(t) \left| \omega^j(t)^T D^j(t) \omega^j(t) \leq 1, t \in I \right. \right\}.$$

The control problems of the preceding section becomes as the following:

(WCD₀): Maximize

$$\int_I \left[f(t, x, u) + u(t)^T B(t) z(t) + \sum_{j=1}^m \lambda^j(t) \left(g^j(t, x, u) + x(t)^T D^j(t) \omega^j(t) \right) + \mu(t)^T (h(t, x, u) - \dot{x}) \right] dt.$$

Subject to

$$x(a) = x(b)$$

$$f_x(t, x, u) + \sum_{j=1}^m \lambda^j(t) \left(g_x^j(t, x, u) + D^j(t) \omega^j(t) \right) + \mu(t)^T h(t, x, u) = \dot{\mu}(t), \quad t \in I$$

$$f_u(t, x, u) + B(t) z(t) + \lambda(t)^T g_u(t, x, u) + \mu(t)^T h_u(t, x, u) = 0, \quad t \in I$$

$$\lambda^j(t) \geq 0, \quad t \in I, \quad j=1,2,\dots,m$$

$$z(t) \in K \text{ and } \omega^j(t) \in C^j, \quad j=1,2,\dots,m.$$

If $s(x|C^j)$, $j=1,2,\dots,m$ are deleted and $s(u|K)$ is replaced by $\left(u(t)^T B(t) u(t)\right)^{\frac{1}{2}}$, the problem (CP₀) and (WCD₀) reduce to those studied by Chandra *et al.* [2].

7. Related Nonlinear Programming Problems

If the functions appearing (CP) and (WCD) are independent, of t then these problems reduce to the following nonlinear programming problem with support functions not reported explicitly in the literature.

(CP₀): Minimize $f(x, u) + s(u|K)$

subject to $g^j(x, u) + s(x|C^j) \leq 0, \quad j=1,2,\dots,m.$

(WCD₀): Maximize $f(x, u) + u^T z + \sum_{j=1}^m \lambda^j \left(g^j(x, u) + \mu^T \omega^j \right) + \mu^T h(x, u)$

subject to $f_x(x, u) + \sum_{j=1}^m \lambda^j \left(g_x^j(x, u) + \omega^j \right) + \mu^T h_x(x, u) = 0$

$$f_u(x, u) + z + \lambda^T g_u(x, u) + \mu^T h_u(x, u) = 0$$

$$\lambda^j \geq 0, \quad j=1,2,\dots,m$$

$$z \in K, \quad \omega^j \in C^j, \quad j=1,2,\dots,m.$$

If $f(x, u)$ and $s(u|K)$ are replaced by $f(x)$ and $s(x|K)$ respectively, the above problem reduce to the following problem studied by Husain *et al.* [12].

(NP₁): Minimize $f(x) + s(x|K).$

Subject to $g^j(x) + s(x|C^j) \leq 0, \quad j=1,2,\dots,m.$

(WNP₁): Maximize $f(x) + x^T z + \sum_{j=1}^m \lambda^j \left(g^j(x) + x^T \omega^j \right) + \mu^T h(x).$

$$\begin{aligned} \text{Subject to } f_x(x) + z + \sum_{j=1}^m \lambda^j (g_x^j(x) + \omega^j) + \mu^T h_x &= 0 \\ \lambda^j &\geq 0, \quad j = 1, 2, \dots, m \\ z &\in K, \quad \omega^j \in C^j, \quad j = 1, 2, \dots, m. \end{aligned}$$

8. Conclusion

Fritz John and Karush-Kuhn-Tucker type necessary optimality conditions for class of nondifferentiable control problems are derived. As an application of Karush-Kuhn-Tucker type necessary optimality conditions, Wolfe type dual is formulated and various duality theorems under generalized convexity conditions are proved. The linkage between our duality results and those of a nonlinear programming problem with support functions is indicated.

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