

# The Equivalence between a Complementarity Control System and a Differential Inclusion \*

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**Abstract:** This paper is devoted to the equivalence between a complementarity control system with subdifferentiable constraint function and a differential inclusion. We mainly use tools of convex analysis to verify the equivalence through constructing the tangent cone and the normal cone.

**Key words:** complementarity system; differential inclusion; tangent cone; normal cone; sub-differentiable function

## 1 Introduction

In recent years, the control systems with complementarity constraints are widely studied, see for instance [3] and references therein. It goes without saying that transforming a constrained control system into an unconstrained system is meaningful. In [3], the following systems are considered:

$$\begin{aligned} -\dot{x}(t) &= f(x(t)) + g(t) + \nabla h(x(t))^T \lambda(t), \quad a.e. \\ 0 &\geq h(x(t)) \perp \lambda(t) \geq 0, \quad a.e. \end{aligned} \quad (1)$$

and

$$-\dot{x}(t) \in f(x(t)) + g(t) + N^h(x(t)), \quad a.e. \quad (2)$$

where  $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $g : \mathbb{R}_+ \rightarrow \mathbb{R}^n$  are measurable,  $\lambda : \mathbb{R}_+ \rightarrow \mathbb{R}^m$ ,  $h(x) = (h_1(x), \dots, h_m(x))^T$ ,  $h_i : \mathbb{R}^n \rightarrow \mathbb{R}$  for  $i = 1, \dots, m$  are smooth, and  $N^h(x(t))$  is a cone (see [3]). The equivalence between (1) and (2) is obtained. In the engineering systems, the complementarity systems are the subject of many recent years. Indeed, if we convert a original control system into a complementarity control system, then the stability and reliability of the system can be improved. Furthermore, if we convert a control system with complementarity constraints into a control system without constraints, we can solve the problem easily. The method used in this paper can be applied to some problems in [5,8,10,11].

In previous papers, see for instance [3], the constraint function  $h$  is smooth. The present paper will generalize the related work, nonsmooth constraint function  $h$  is considered. We will suppose that  $h$  is subdifferentiable.

## 2 Preliminaries

In this section, the notions about a subdifferentiable function and a quasiconvex function are given.

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Let  $h : \mathfrak{R}^n \rightarrow \mathfrak{R}^n$  be directional differentiable. As in [4], if there exists a convex and compact set  $\partial h(x)$ , such that the directional derivative of  $h$  can be written as

$$h'(x; d) = \lim_{t \rightarrow 0^+} \frac{1}{t}(h(x + td) - h(x)) = \max_{\xi \in \partial h(x)} \xi^T d, \quad \forall d \in \mathfrak{R}^n,$$

then, we call that  $h$  is subdifferentiable,  $\partial h(x)$  is called the subdifferential of  $h$ .

Subdifferentiable function is widely used in control and optimization. Continuous function, convex function, and maximum function are subdifferentiable. Next, we give an example of a subdifferentiable function. Let

$$\psi(x) = G(\max_{j \in J_1} h_{1j}(x), \dots, \max_{j \in J_m} h_{mj}(x)),$$

where  $G$  and  $h_{ij}$  are continuous and differentiable on  $\mathfrak{R}^m$  and  $\mathfrak{R}^n$ , respectively,  $J_i$  for  $i = 1, \dots, m$  are finite index sets. Denote

$$J_i(x) = \{k \in J_i | h_{kj}(x) = \max_{j \in J_i} h_{ij}(x)\}.$$

As in [6], the subdifferential of  $\psi$  is written by

$$\partial \psi(x) = \text{co}\{\zeta \in \mathfrak{R}^n | \zeta = \sum_{i=1}^m \frac{\partial G(y_1, \dots, y_m)}{\partial y_i} \Big|_{y_i = \max_{j \in J_i} h_{ij}(x)} \nabla h_{j_i}(x), j_i \in J_i\}.$$

Evidently,  $\partial \psi(x)$  is a convex hull of finite number of points.

If the function  $h_i, i \in I$  are subdifferentiable, then  $h(x) = \max_{i \in I} h_i(x)$  also is subdifferentiable with the subdifferential

$$\partial h(x) = \text{co}\left\{ \bigcup_{i \in I(x)} \partial h_i(x) \right\},$$

where  $I(x) = \{i \in I | h_i(x) = h(x)\}$ , see [4]. Therefore if  $\partial h_i(x)$  is a convex hull of finite number of points, then  $\partial h(x)$  is a convex hull of finite number of points also .

**Definition 1** ([2]). A function  $h : \mathfrak{R}^n \rightarrow \mathfrak{R}$  is called quasiconvex, if  $h(q_1 x^1 + q_2 x^2) \leq \max[h(x^1), h(x^2)]$ ,  $\forall x^1, x^2 \in \mathfrak{R}^n$  and  $q_1, q_2 \geq 0$  with  $q_1 + q_2 = 1$ .

**Theorem 2** ([2]). A function  $h : \mathfrak{R}^n \rightarrow \mathfrak{R}$  is called quasiconvex, if and only if the level set of  $h$  is convex for all  $\alpha \in \mathfrak{R}$ .

Suppose  $h_i, i \in I$  are a quasiconvex function. In what follows we show that the maximum function  $h(x) = \max_{i \in I} h_i(x)$  is a quasiconvex function, too.

By deducing, we have

$$\begin{aligned} h(q_1 x^1 + q_2 x^2) &\leq \max_{i \in I} h_i(q_1 x^1 + q_2 x^2) \leq \max_{i \in I} \max[h_i(x^1), h_i(x^2)] \\ &= \max_{i \in I} [\max h_i(x^1), \max h_i(x^2)] = \max[h(x^1), h(x^2)], \\ &\quad \forall x^1, x^2 \in \mathfrak{R}^n, q_1, q_2 \geq 0, q_1 + q_2 = 1. \end{aligned}$$

According to Definition 1 ,  $h(x) = \max_{i \in I} h_i(x)$  is a quasiconvex function.

The normal cone to a nonempty closed convex set  $K \subset \mathfrak{R}^n$  at  $x \in K$  is defined as

$$N_K(x) := \{s \in \mathfrak{R}^n | \langle s, y - x \rangle \leq 0, \quad \forall y \in K\},$$

while the tangent cone is the polar of the normal cone, which means

$$T_K(x) := [N_K(x)]^\circ := \{d \in \mathfrak{R}^n | \langle s, d \rangle \leq 0, \quad \forall s \in N_K(x)\}.$$

### 3 An Equivalence

If there are multiconstraints, for instance  $h_j(x) \leq 0, j = 1, \dots, m$ . Then, we reformulate the constraints by  $h(x) = \max_{1 \leq j \leq m} h_j(x) \leq 0$ . In Section 2, we know that if  $h_j$  for  $j = 1, \dots, m$  are subdifferentiable, quasiconvex and the subdifferential of  $h_j$  is a convex hull of finite number of points, then the maximum function  $h$  also is subdifferentiable, quasiconvex and its subdifferential is a convex hull of finite number of points. Therefore, without loss of generalization, we consider only one constraint function. That is, the constraint function is a maximum function  $h(x) = \max_{1 \leq j \leq m} h_j(x)$ .

Throughout this section, we suppose that  $h_j$  for  $j = 1, \dots, m$  are subdifferentiable, quasiconvex and  $\partial h_j(x)$  for  $j = 1, \dots, m$  are a convex hull of finite number of points. Therefore,  $h$  also is subdifferentiable. Since the subdifferential of  $h_j$  is a convex hull of finite number of points, we have

$$\partial h(x) = \text{co}\{v_1, \dots, v_p\}.$$

Since  $h_j$  is quasiconvex,  $h$  also is quasiconvex. By virtue of Theorem 1,  $C := \{x \in \mathbb{R}^n | h(x) \leq 0\}$  is convex.

Constraint qualifications (CQ for short) of set  $C$  at  $x$  are given as follows:

**CQ 1** ([1]). There exists  $y_0 \in \mathbb{R}^n$ , such that  $h'(x; y_0) < 0$ .

**CQ 2** ([9]).  $\text{cl}\gamma(x) = \Gamma(x)$ , where  $\text{cl}$  is a closure,

$$\gamma(x) = \{y \in \mathbb{R}^n | h'(x; y) < 0\},$$

$$\Gamma(x) = \{y \in \mathbb{R}^n | h'(x; y) \leq 0\}.$$

**Proposition 1** ([9]). If CQ1 or CQ2 is satisfied at  $x \in C$ , then  $T_C(x) = \Gamma(x)$ .

**Proposition 2** If CQ1 or CQ2 is satisfied, then

$$N_C(x) = [T_C(x)]^\circ = \left\{ \sum_{i \in I_0} \lambda_i v_i | \lambda_i \geq 0, i \in I_0 \right\},$$

where  $I_0 = \{i \in \{1, \dots, p\} | v_i^T y = 0\}$ .

**Proof.** According to [7], we obtain

$$T_C = \{y \in \mathbb{R}^n | v_i^T y \leq 0, 1 \leq i \leq p\}.$$

Let

$$D = \left\{ \sum_{i \in I_0} \lambda_i v_i | \lambda_i \geq 0, i \in I_0 \right\}.$$

We will verify

$$D^\circ = \{y \in \mathbb{R}^n | v_i^T y \leq 0, 1 \leq i \leq p\}.$$

If  $v_i^T y \leq 0$  for all  $i$ , then  $x^T y \leq 0$  for all  $x \in D$ . Therefore,  $y \in D^\circ$  and

$$D^\circ \supset \{y \in \mathbb{R}^n | v_i^T y \leq 0, 1 \leq i \leq p\}.$$

Conversely, if  $y \in D^\circ$ , i.e.,  $x^T y \leq 0$  for all  $x \in D$ , and since  $v_i \in D$  for all  $i$ , then we have  $v_i^T y \leq 0$  for all  $i$ . Thus, it follows that

$$D^\circ \subset \{y \in \mathbb{R}^n | v_i^T y \leq 0, 1 \leq i \leq p\}.$$

Since  $D$  is closed and convex, we have  $(D^\circ)^\circ = D$ , in other words,

$$[T_C(x)]^\circ = \left\{ \sum_{i \in I_0} \lambda_i v_i | \lambda_i \geq 0, i \in I_0 \right\}.$$

Furthermore, the convexity and closeness of  $C$  implies

$$N_C(x) = [T_C(x)]^\circ = \left\{ \sum_{i \in I_0} \lambda_i v_i | \lambda_i \geq 0, i \in I_0 \right\}.$$

■

Consider the following complementarity system

$$\begin{aligned} -\dot{x}(t) &= f(x(t)) + g(t) + \lambda(t)V, \quad a.e. \\ (v_i, y)\lambda_i &= 0, \quad (v_i, y) \leq 0, \quad \lambda_i \geq 0, \quad a.e. \end{aligned} \tag{3}$$

where  $V = (v_1, \dots, v_p)^T$ ,  $\lambda(t) = (\lambda_1, \dots, \lambda_p)$  and  $y \in T_C(x)$ , and the differential inclusion

$$-\dot{x}(t) \in f(x(t)) + g(t) + N_C(x(t)). \quad a.e. \tag{4}$$

**Proposition 3**  $x = x(t)$  and  $\dot{x} = \dot{x}(t)$  in  $\mathbb{R}^n$  satisfy (3) if and only if they satisfy (4).

**Proof.** If  $x(t) \notin C$ ,  $x(t)$  does not solve (3) or (4). We suppose  $x(t) \in C$ .  $(v_i, y)\lambda_i = 0, (v_i, y) \leq 0, \lambda_i \geq 0$  mean that if  $(v_i, y) < 0$ , i.e. if  $i \notin I_0$ , then  $\lambda_i = 0$ . From the definition of  $N_C(x)$ , (3) is equivalent to (4). ■

**Remark 1** When the constraint function  $h : \mathbb{R}^n \rightarrow \mathbb{R}$  is smooth, the system (3) degenerates to the form of (1). In fact, (3) is written as

$$\begin{aligned} -\dot{x}(t) &\in f(x(t)) + g(t) + \lambda(t)\partial h(x(t)), \quad a.e. \\ (v_i, y)\lambda_i &= 0, (v_i, y) \leq 0, \lambda_i \geq 0. \quad a.e. \end{aligned} \tag{5}$$

We know that  $v_i \in \partial h(x(t))$  for  $i = 1, \dots, p$ . Thus,  $v_i$  has the form

$$v_i = (v_i^1, v_i^2, \dots, v_i^n)^T.$$

Obviously, if  $h$  is smooth, then  $\partial h(x) = \{\nabla h(x)\}$  and  $p = 1$ . Therefore, (5) can be rewritten by

$$\begin{aligned} -\dot{x}(t) &= f(x(t)) + g(t) + \lambda(t)\nabla h(x(t)), \quad a.e. \\ (\nabla h(x), y)\lambda &= 0, (\nabla h(x), y) \leq 0, \lambda \geq 0, \quad a.e. \end{aligned} \tag{6}$$

where  $h : \mathbb{R}^n \rightarrow \mathbb{R}$  is a smooth function. At the same time, (4) can be rewritten by

$$-\dot{x}(t) \in f(x(t)) + g(t) + N_C(x(t)), \quad a.e. \tag{7}$$

where  $N_C = \{\lambda\nabla h(x) | \lambda \geq 0, (\nabla h(x), y) = 0, y \in T_C\}$  or  $N_C = \emptyset$  (according to proposition 3).

If one of the qualifications (CQ1-4) is satisfied in [3], then according to proposition 3 of [3], we have (7) and (2) are equivalent. Thus, (6) and (1) are equivalent. Namely, (1) is the degeneration of (3).

**Example 1** In the set  $C$ , let  $h(x) = \max\{-x_1, -x_2, x_1^2 + x_2^2 - 1\}$ ,  $x = (x_1, x_2)^T$ . Evidently,  $h$  is subdifferentiable and  $C$  is convex and closed. Consider  $x_0 = (0, 1)^T$ . According to the definition and operation of differential, we have

$$\partial h(x_0) = \text{co}\{(0, -1)^T, (2, 0)^T\}.$$

Denoting

$$I_0 = \{i \in \{1, 2\} | v_i^T y = 0, \},$$

since

$$\begin{aligned} T_C &= \{y \in \mathbb{R}^2 | \max_{1 \leq i \leq 2} v_i^T y \leq 0\} \\ &= \{y \in \mathbb{R}^2 | v_i^T y \leq 0, 1 \leq i \leq 2\}, \end{aligned}$$

we obtain

$$N_C(x) = [T_C(x)]^\circ = \left\{ \sum_{i \in I_0} \lambda_i v_i | \lambda_i \geq 0, i \in I_0 \right\}.$$

Setting  $v_1 = (0, -1)^T, v_2 = (2, 0)^T$ , it follows that

$$T_C = \{y \in \mathbb{R}^2 | y_1 \leq 0, y_2 \geq 0\}.$$

Therefore, it is obtained that

- if  $y = (y_1 \neq 0, y_2 \neq 0)^T$ , then  $I_0 = \emptyset, \lambda_1 = \lambda_2 = 0$ ;
- if  $y = (y_1 \neq 0, y_2 = 0)^T$ , then  $I_0 = \{1\}, \lambda_1 \geq 0, \lambda_2 = 0$ ;
- if  $y = (y_1 = 0, y_2 = 0)^T$ , then  $I_0 = \{1, 2\}, \lambda_1 \geq 0, \lambda_2 \geq 0$ ;
- if  $y = (y_1 = 0, y_2 \neq 0)^T$ , then  $I_0 = \{2\}, \lambda_1 = 0, \lambda_2 \geq 0$ .

Similarly, we can get the equivalence between (3) and (4) at other points. Therefore, the complementarity system (3) and the differential inclusion (4) are equivalent.

## 4 Conclusions

In nonlinear control systems, constraint functions usually are nonsmooth. However, subdifferentiable function is a wide class of nonsmooth functions. It is meaningful to consider the equivalence between a complementarity system with subdifferentiable constraint function and a differential inclusion .

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