Nonsmooth Continuous-time Optimization Problems via Invexity

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Abstract

We introduce the notion of KKT-invexity for nonsmooth continuous-time nonlinear optimization problems and prove that this notion is a necessary and sufficient condition for global optimality of a Karush-Kuhn-Tucker point.

Key words: Nonsmooth continuous optimization, KKT conditions, KKT-invexity.

1 Introduction

We regard the continuous-time nonlinear programming problem below.

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¹ Adilson J. V. Brandão was partially supported by Instituto do Milenio-IM-AGIMB.

 $^{^2~}$ Valeriano A. de Oliveira is a Ph-D student and he was supported by CNPq-Brazil, Grant 141168/2003-0.

 $^{^3}$ Marko A. Rojas-Medar was partially supported by CNPq - Brazil, Grant 301354/03-0 and project BFM 2003-06446-CO-01, Spain.

Minimize
$$\phi(x) = \int_{0}^{T} f(t, x(t)) dt$$
,
subject to $g(t, x(t)) \le 0$ a.e. in $[0, T]$,
 $x \in X$.
(CNP)

Here X is a nonempty open convex subset of the Banach space $L_{\infty}^{n}[0,T]$, $\phi: X \to \mathbb{R}$, $g(t,x(t)) = \gamma(x)(t)$, $f(t,x(t)) = \xi(x)(t)$, $\gamma: X \to \Lambda_{1}^{m}[0,T]$ and $\xi: X \to \Lambda_{1}^{1}[0,T]$, where $L_{\infty}^{n}[0,T]$ denotes the space of all *n*-dimensional vector valued Lebesgue measurable functions, which are essentially bounded, defined on the compact interval $[0,T] \subset \mathbb{R}$, with norm $\|\cdot\|_{\infty}$ defined by

$$||x||_{\infty} = \max_{1 \le j \le n} \operatorname{ess\,sup}\{|x_j(t)|, \ 0 \le t \le T\},\$$

where for each $t \in [0, T]$, $x_j(t)$ is the *j*-th component of $x(t) \in \mathbb{R}^n$ and $\Lambda_1^m[0, T]$ denotes the space of all *m*-dimensional vector functions which are essentially bounded and Lebesgue measurable, defined on [0, T], with the norm $\|\cdot\|_1$ defined by

$$||y||_1 = \max_{1 \le j \le m} \int_0^T |y_j(t)| dt.$$

This class of problems was introduced in 1953 by Bellman [2] in connection with production-inventory "botleneck processes". Optimality conditions in the spirit of Kuhn-Tucker type for continuous nonlinear problems were first investigated by Hanson and Mond [7]. Farr and Hanson [6] obtained necessary and sufficient optimality conditions for a more general class of continuoustime nonlinear problems (both cost function and constraints were nonlinear). Assuming some kind of constraint qualifications and using direct methods, further generalizations of the theory of optimality conditions for continuoustime nonlinear problems are to be found in Scott and Jefferson [11], Abraham and Buie [1], Reiland and Hanson [9] and Zalmai [12], [13], [14], [15]. The development of nonsmooth necessary optimality conditions for problem (CNP) was given in [3]. The sufficient conditions for the nonsmooth case was given in [10]. Related results can be found in Craven [5]. However, his arguments are via approximation of smooth functions rather than alternative theorems. None the above works established necessary and sufficient conditions for a Karush-Kuhn-Tucker point be a global solution of (CNP). We observe that in the case of mathematical programming these results was given by Martin [8]. In this work we obtain a similar result for (CNP).

2 Preliminaries

Let \mathbb{F} be the set of all feasible solutions to Problem (CNP) (which we suppose nonempty), i.e., $\mathbb{F} = \{x \in X : g(t, x(t)) \leq 0 \text{ a.e. in } [0, T]\}.$

Let V be an open subset of \mathbb{R}^n containing the set $\{x(t) \in \mathbb{R}^n : x \in X, t \in [0,T]\}$. We assume that f and g_i (the *i*-th component of g), $i \in I = \{1, 2, \ldots, m\}$, are real functions defined on $V \times [0,T]$. The functions $t \mapsto f(x(t),t)$ and $t \mapsto g(x(t),t)$ are assumed to be Lebesgue measurable and integrable for all $x \in X$.

We assume that, given $a \in V$, there exist an $\varepsilon > 0$ and a positive number k such that for all $t \in [0, T]$, and for all $x, y \in a + \varepsilon B$ (B denotes de unit ball of \mathbb{R}^n) we have $|f(t, x) - f(t, y)| \leq k ||x - y||$. Similar hypotheses are assumed for g_i , $i \in I$. Hence, $f(t, \cdot)$ and $g_i(t, \cdot)$, $i \in I$, are locally Lipschitz on V throughout [0, T].

Let $x \in X$ and $h \in L_{\infty}^{n}[0, T]$. We denote by $\phi^{\circ}(x; h)$ and $g_{i}^{\circ}(t, x(t); h(t))$, $i \in I$, the Clarke generalized directional derivative of ϕ and g_{i} , $i \in I$, at x on the direction h, respectively. See [4] for more details.

Given $x \in \mathbb{F}$, we define for each $i \in I$, the sets $A_i(x) = \{t \in [0,T] : g_i(x(t),t) = 0\}$ and $A(x) = \bigcup_{i \in I} A_i(x)$.

3 Invex characterization of KKT points

In [8] Martin introduced the notion of KKT-invexity for the mathematical programming problems. In this section we extend this concept to the (CNP) problem.

Definition 3.1 The problem (CNP) is called Karush-Kuhn-Tucker invex (or KKT-invex) if there exists a function $\eta: V \times V \to \mathbb{R}^n$ such that $\eta(x(t), y(t)) \in L^n_{\infty}[0,T]$ and

$$\phi(x) - \phi(y) \ge \phi^{\circ}(y; \eta(x, y)), \tag{1}$$

$$-g_i^{\circ}(t, y(t); \eta(y(t), x(t)) \ge 0 \quad a.e. \text{ in } A_i(y), \ i \in I.$$
(2)

for all $x, y \in \mathbb{F}$.

Definition 3.2 We say that a point $y \in \mathbb{F}$ is a Karush-Kuhn-Tucker point (or KKT point) for (CNP) if there exist $\lambda_i \in L_{\infty}[0,T]$, $i \in I$, such that

$$\phi^{\circ}(y;h) + \int_{0}^{T} \sum_{i \in I} \lambda_i(t) g_i^{\circ}(t, y(t); h(t)) dt \ge 0, \ \forall h \in L_{\infty}^n[0, T],$$
(3)

$$\lambda_i(t)g_i(t, y(t)) = 0 \ a.e. \ in \ [0, T], \ i \in I,$$
(4)

$$\lambda_i(t) \ge 0 \ a.e. \ in \ [0,T], \ i \in I.$$
(5)

Definition 3.3 We say that the constraint g satisfies the constraint qualification at $y \in \mathbb{F}$ if there do not exist $u_i \in L_{\infty}[0,T]$, $u_i \ge 0, i \in I$, not all zero, such that

$$\int_{A(y)} u_i(t)g_i^{\circ}(t,y(t);h(t))dt \ge 0 \text{ for all } h \in L_{\infty}^n(A(y)).$$

Lemma 3.4 Let $y \in \mathbb{F}$ and assume that g satisfies the constraint qualification at y. If y is not a KKT point of (CNP) then there exists $h \in L_{\infty}^{n}[0,T]$ such that

$$\phi^{\circ}(y;h)) < 0, \tag{6}$$

$$g_i^{\circ}(t, y(t); h(t)) < 0 \quad a.e. \text{ in } A_i(x), \ i \in I.$$
 (7)

Proof. In fact, if such solution does not exist then, by the Generalized Gordan Theorem (see [13]), there exist $u_0 \in \mathbb{R}$ and $u_i \in L_{\infty}[0,T]$, $i \in I$, with $u_0 \geq 0$, $u_i(t) \geq 0$ a.e. in $A(y), i \in I$, not all zero, such that

$$u_0\phi^{\circ}(y;h) + \int_{A(y)} \sum_{i \in I} u_i(t)g_i^{\circ}(t,y(t);h(t))dt \ge 0, \ \forall h \in L_{\infty}^n(A(y)).$$

If $u_0 = 0$ we have a contradiction with the constraint qualification. Hence $u_0 > 0$. Setting $\lambda_i = u_i/u_0$, $i \in I$, and defining $\lambda_i(t) = 0$, $t \in [0, T] \setminus A$, $i \in I$, we obtain

$$\phi^{\circ}(y;h) + \int_{0}^{T} \sum_{i \in I} \lambda_i(t) g_i^{\circ}(t,y(t);h(t)) dt \ge 0, \forall h \in L_{\infty}^n[0,T].$$

So we have

$$\phi^{\circ}(y;h) + \int_{0}^{T} \sum_{i \in I} \lambda_{i}(t) g_{i}^{\circ}(t, y(t); h(t)) dt \geq 0, \forall h \in L_{\infty}^{n}[0, T]$$
$$\lambda_{i}(t) g_{i}(y(t), t) = 0 \text{ a.e. in } [0, T], \ i \in I,$$
$$\lambda_{i}(t) \geq 0 \text{ a.e. in } [0, T], \ i \in I.$$

Then y is a KKT point, which contradicts the hypothesis. The contradiction has occurred because we suppose that does not exist $h \in L^n_{\infty}[0,T]$ satisfying (6) and (7).

Theorem 3.5 We assume that g satisfies the constraint qualification at each $y \in \mathbb{F}$. Then, every KKT point of (CNP) is a global minimizer if and only if (CNP) is KKT-invex.

Proof. (Necessity) First we suppose that $x, y \in \mathbb{F}$ and $\phi(x) < \phi(y)$. Thence y is not a global minimizer, and so, by hypothesis, y is not a KKT point of (CNP). Then, by Lemma 3.4, there exists $h \in L^n_{\infty}[0,T]$ satisfying (6) and (7). Set $\alpha = \phi^{\circ}(y;h)$ and $\eta(x(t), y(t)) = \{\phi(x) - \phi(y)\}\alpha^{-1}h(t)$. Because of (6) we know that $\{\phi(x) - \phi(y)\}\alpha^{-1} > 0$. Hence

$$\phi^{\circ}(y;\eta(x,y)) = \phi^{\circ}(y;\{\phi(x) - \phi(y)\}\alpha^{-1}h) = \{\phi(x) - \phi(y)\}\alpha^{-1}\phi^{\circ}(y;h),$$

and therefore

$$\phi^{\circ}(y;\eta(x,y)) = \phi(x) - \phi(y). \tag{8}$$

Because of (7) we get

$$g_i^{\circ}(y(t), t; \eta(x(t), y(t))) = \{\phi(x) - \phi(y)\}\alpha^{-1}g_i^{\circ}(y(t), t; h(t)) \\< 0 \text{ a.e. in } A_i(y), \ i \in I.$$

Therefore

$$-g_i^{\circ}(y(t), t; \eta(x(t), y(t))) \ge 0 \text{ a.e. in } A_i(y), \ i \in I.$$

$$\tag{9}$$

By (8) and (9) we conclude that for $\phi(x) < \phi(y)$ (CNP) is KKT-invex. The cases $\phi(x) \ge \phi(y)$ or $x \notin \mathbb{F}$ or $y \notin \mathbb{F}$ are covered taking $\eta \equiv 0$.

(Sufficiency) It follows from (4) that $\lambda_i(t) = 0, t \in [0,T] \setminus A_i(y), i \in I$. Then by (1), (2) and (5) we have

$$\phi(x) - \phi(y) - \phi^{\circ}(y; \eta(x, y)) - \int_{0}^{T} \sum_{i \in I} \lambda_{i}(t) g_{i}^{\circ}(t, y(t); \eta(x(t), y(t))) dt \ge 0,$$

for all $x \in \mathbb{F}$. So

$$\phi(x) - \phi(y) \ge \phi^{\circ}(y; \eta(x, y)) + \int_{0}^{T} \sum_{i \in I} \lambda_i(t) g_i^{\circ}(t, y(t); \eta(x(t), y(t))) dt.$$

By (3) and it follows that $\phi(x) \ge \phi(y), \forall x \in \mathbb{F}$.

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