



Correction to: Convexifactors, Generalized Convexity, and Optimality Condition

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In this article the below author names are included.

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1 Introduction

We correct an error in the proof of Theorem 3.5 in [1].

The original article can be found online at <https://doi.org/10.1023/A:1014853129484>.

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2 Mathematical Details

Theorem 3.5 in [1] says, if $\bar{x} \in S$ is a local minimum of (P) and f is locally Lipschitz which admits an USRC $\partial^* f(\bar{x})$ at \bar{x} , then

$$0 \in \text{cl}(\overline{\text{co}}(\partial^* f(\bar{x})) + T^\circ(S, \bar{x})). \tag{1}$$

The following example justifies that the above theorem fails to hold.

Example 2.1 Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ and the feasible set S be defined as $f(x_1, x_2) = -x_1 + |x_2|$ and $S = \{(x_1, x_2) \in \mathbb{R}^2 : 0 \leq x_1 \leq |x_2|\}$. Clearly, $\bar{x} = (0, 0)$ is a global minimum, $T(S, \bar{x}) = S$ and $\text{co}(T(S, \bar{x})) = \text{co}(S) = \{(x_1, x_2) \in \mathbb{R}^2 : x_1 \geq 0\}$. Now, for any $v = (v_1, v_2) \in \mathbb{R}^2$, $f_d^+(\bar{x}, v) = -v_1 + |v_2|$. Also it can be seen that $\partial^* f(\bar{x}) = \{(-1, 1), (-1, -1)\}$ is an upper semi-regular convexificator of f at \bar{x} . As $T(S, \bar{x}) = S$, it follows that $f_d^+(\bar{x}, v) \geq 0$, for all $v \in T(S, \bar{x})$. Thus $\sup_{\zeta \in \partial^* f(\bar{x})} \langle \zeta, v \rangle \geq 0$, for all $v \in T(S, \bar{x})$. Clearly, $\sup_{\zeta \in \partial^* f(\bar{x})} \langle \zeta, \bar{v} \rangle = f_d^+(\bar{x}, \bar{v}) = -1 < 0$ for $\bar{v} = (2, 1) \in \text{co}(T(S, \bar{x}))$. Moreover, $T^\circ(S, \bar{x}) = \{(x, 0) \in \mathbb{R}^2 : x \leq 0\}$ and $\overline{\text{co}}(\partial^* f(\bar{x})) = \{(-1, t) \in \mathbb{R}^2 : -1 \leq t \leq 1\}$. Hence $0 \notin \text{cl}(\overline{\text{co}}(\partial^* f(\bar{x})) + T^\circ(S, \bar{x}))$. Thus Theorem 3.5 in [1] fails to hold for f at \bar{x} .

We rectify the error in the above theorem by assuming the tangent cone to be convex. For this we first recall the notion of support functions from [2]. The *support function* $\sigma_A(x) : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ of a nonempty set $A \subseteq \mathbb{R}^n$ is defined as $\sigma_A(x) := \sup_{a \in A} \langle x, a \rangle$.

A correct statement of Theorem 3.5 in [1] should be as follows.

Theorem 2.1 *If $\bar{x} \in S$ is a local minimum of (P), $T(S, \bar{x})$ is a convex cone and f is locally Lipschitz which admits an USRC $\partial^* f(\bar{x})$ at \bar{x} , then (1) holds. Further, if $\partial^* f(\bar{x})$ is bounded, then*

$$0 \in \text{co}(\partial^* f(\bar{x})) + T^\circ(S, \bar{x}). \tag{2}$$

Proof As \bar{x} is a local minimum of (P), there exists $\epsilon > 0$ such that $f(\bar{x}) \leq f(x)$ for all $x \in B(\bar{x}, \epsilon) \cap S$. For $v \in T(S, \bar{x})$, there exist sequences $(t_k)_{k \in \mathbb{N}}$ and $(v_k)_{k \in \mathbb{N}}$ with $t_k \downarrow 0$ and $v_k \rightarrow v$ such that $\bar{x} + t_k v_k \in S$. Thus there exists $k_0 \in \mathbb{N}$ such that $\bar{x} + t_k v_k \in B(\bar{x}, \epsilon) \cap S$ for all $k \geq k_0$, which implies $f(\bar{x}) \leq f(\bar{x} + t_k v_k)$ for all $k \geq k_0$. As f is locally Lipschitz with Lipschitz constant say, L , hence for every $v \in T(S, \bar{x})$ we have

$$\begin{aligned} f_d^+(\bar{x}, v) &= \limsup_{t \downarrow 0} \frac{f(\bar{x} + tv) - f(\bar{x})}{t} \\ &\geq \limsup_{k \rightarrow \infty} \left[\frac{f(\bar{x} + t_k v) - f(\bar{x} + t_k v_k)}{t_k} + \frac{f(\bar{x} + t_k v_k) - f(\bar{x})}{t_k} \right] \\ &\geq \lim_{k \rightarrow \infty} [-L \|v_k - v\|] + \limsup_{k \rightarrow \infty} \left[\frac{f(\bar{x} + t_k v_k) - f(\bar{x})}{t_k} \right] \geq 0. \end{aligned}$$

As $\partial^* f(\bar{x})$ is an USRC of f at \bar{x} , it follows from [2, Proposition 2.2.1 (p. 211)] that

$$\sigma_{\text{co}(\partial^* f(\bar{x}))}(v) = \sigma_{\partial^* f(\bar{x})}(v) = \sup_{\zeta \in \partial^* f(\bar{x})} \langle \zeta, v \rangle \geq 0, \text{ for all } v \in T(S, \bar{x}). \quad (3)$$

As $T(S, \bar{x})$ is convex, hence by applying [2, Example 2.3.1 (p. 215)] for $K = T^\circ(S, \bar{x})$ we deduce that

$$\sigma_{T^\circ(S, \bar{x})}(v) = \begin{cases} 0, & \text{if } v \in T(S, \bar{x}), \\ +\infty, & \text{otherwise.} \end{cases} \quad (4)$$

In view of [2, Theorem 3.3.3(i) (p. 226)] the support function of the set $U = \text{cl}(\text{co}(\partial^* f(\bar{x})) + T^\circ(S, \bar{x}))$ is

$$\sigma_U(v) = \sigma_{\text{co}(\partial^* f(\bar{x}))}(v) + \sigma_{T^\circ(S, \bar{x})}(v), \text{ for all } v \in T(S, \bar{x}). \quad (5)$$

Using (3)–(5) and the fact that $\sigma_K(v)$ is infinite for $v \notin T(S, \bar{x})$, we conclude that $\sigma_U(v) \geq 0$ for all $v \in \mathbb{R}^n$. Thus, by [2, Theorem 2.2.2 (p. 211)], $0 \in U = \text{cl}(\text{co}(\partial^* f(\bar{x})) + T^\circ(S, \bar{x}))$.

If $\partial^* f(\bar{x})$ is a bounded set then $\text{co}(\partial^* f(\bar{x}))$ is compact as $\partial^* f(\bar{x})$ is a closed set. Hence (1) reduces to (2) as $\text{co}(\partial^* f(\bar{x})) + T^\circ(S, \bar{x})$ is a closed set. \square

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