

**CHARACTERIZATION OF CONVEXITY FOR A PIECEWISE  
 $C^2$  FUNCTION BY THE LIMITING SECOND-ORDER  
SUBDIFFERENTIAL**

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**Abstract.** We prove in this paper that a piecewise  $C^2$  function  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  is convex if and only if for every  $(x, y) \in \text{gph}\partial\varphi$ , the limiting second-order subdifferential mapping  $\partial^2\varphi(x, y) : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  has the so-called positive semi-definiteness (PSD) - in analogy with the notion of positive semi-definiteness of symmetric real matrices. As a by-product, characterization for strong convexity of  $\varphi$  is established.

1. INTRODUCTION

Due to its important role in mathematical economics, engineering, management science, and optimization theory, convexity of functions and sets has been studied intensively; see [1, 3, 5-7, 9, 11, 13, 14] and the references therein.

First-order characterizations for the convexity of extended real-valued functions via the monotonicity of the Fréchet derivative and the monotonicity of the Fréchet subdifferential mapping or the limiting subdifferential mapping can be found, e.g., in [6, 11, 12] and [7, Theorem 3.56].

The classical second-order characterization of convexity of real-valued functions (see for instance [11, 12]) says that a  $C^2$  function  $\varphi : U \rightarrow \mathbb{R}$  where  $U$  is an open convex subset of  $\mathbb{R}^n$  is convex if and only if for every  $x \in U$  the Hessian  $\nabla^2 f(x)$  is a positive semidefinite matrix. To relax the assumption on the  $C^2$  smoothness of the function under consideration, several authors have characterized the convexity by using various kinds of generalized second-order directional derivatives. The reader is referred to [1, 4, 5, 13, 14] for results in this direction.

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Recently, the authors in [3] have found that to a certain extent convexity of functions can be characterized by second-order subdifferential mappings. Among other things, they obtained some characterizations for convexity of piecewise linear functions and of piecewise  $C^2$  functions of a special type via the limiting second-order subdifferential. The purpose of this paper is to *characterize the convexity of piecewise  $C^2$  functions by the limiting second-order subdifferential*.

We will show that a piecewise  $C^2$  function  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  is convex if and only if for every  $(x, y) \in \text{gph}\partial\varphi$ , the limiting second-order subdifferential mapping  $\partial^2\varphi(x, y) : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  has the so-called positive semi-definiteness (PSD) - in analogy with the notion of positive semi-definiteness of symmetric real matrices. Since strong convexity of functions plays a remarkable role in theory of algorithms [12] and stability theory of optimization problems [2], by using the limiting second-order subdifferential we derive a necessary and sufficient condition for strong convexity of piecewise  $C^2$  functions.

The rest of the paper is organized as follows. Section 2 contains some definitions and results which are needed in the sequel. Section 3 is devoted to the necessary and sufficient condition for convexity of a piecewise  $C^2$  function by its limiting second-order subdifferential. As a by-product, the second-order necessary and sufficient condition for strong convexity of piecewise  $C^2$  functions is given.

## 2. PRELIMINARIES

We start by recalling some notions related to generalized differentiation. The notions and related results of generalized differentiation can be found in [7].

For a set  $\Omega \subset \mathbb{R}^n$  and an extended real-valued function  $\varphi : \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ , the symbols  $x \xrightarrow{\Omega} \bar{x}$  and  $x \xrightarrow{\varphi} \bar{x}$  mean that  $x \rightarrow \bar{x}$  with  $x \in \Omega$  and  $x \rightarrow \bar{x}$  with  $\varphi(x) \rightarrow \varphi(\bar{x})$ , respectively. Given a set-valued mapping  $F : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$ , we denote by

$$\text{Lim sup}_{x \xrightarrow{\Omega} \bar{x}} F(x) := \left\{ x^* \in \mathbb{R}^n \mid \exists \text{ sequences } x_k \xrightarrow{\Omega} \bar{x} \text{ and } x_k^* \rightarrow x^* \right. \\ \left. \text{with } x_k^* \in F(x_k) \text{ for all } k \in \mathbb{N} \right\}$$

the *sequential Painlevé-Kuratowski upper limit* of the mapping  $F$  as  $x \xrightarrow{\Omega} \bar{x}$ .

Let  $\varphi : \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$  be finite at  $\bar{x} \in \mathbb{R}^n$  and let  $\varepsilon \geq 0$ . The  $\varepsilon$ -*subdifferential* of  $\varphi$  at  $\bar{x}$  is the set  $\widehat{\partial}_\varepsilon\varphi(\bar{x})$  defined by

$$\widehat{\partial}_\varepsilon\varphi(\bar{x}) = \left\{ x^* \in \mathbb{R}^n : \liminf_{x \rightarrow \bar{x}} \frac{\varphi(x) - \varphi(\bar{x}) - \langle x^*, x - \bar{x} \rangle}{\|x - \bar{x}\|} \geq -\varepsilon \right\}.$$

We put  $\widehat{\partial}_\varepsilon\varphi(\bar{x}) = \emptyset$  if  $|\varphi(\bar{x})| = \infty$ . When  $\varepsilon = 0$  the set  $\widehat{\partial}_0\varphi(\bar{x})$ , denoted by  $\widehat{\partial}\varphi(\bar{x})$ , is called the Fréchet subdifferential of  $\varphi$  at  $\bar{x}$ . The *limiting subdifferential* (or *Mordukhovich subdifferential*) of  $\varphi$  at  $\bar{x}$  is given by

$$(2.1) \quad \partial\varphi(\bar{x}) = \text{Lim sup}_{\substack{x \xrightarrow{\varphi} \bar{x}; \\ \varepsilon \downarrow 0}} \widehat{\partial}_\varepsilon\varphi(x),$$

that is,  $x^* \in \partial\varphi(\bar{x})$  if and only if there exist sequences  $\varepsilon_k \downarrow 0$ ,  $x_k \xrightarrow{\varphi} \bar{x}$  and  $x_k^* \rightarrow x^*$  such that  $x_k^* \in \widehat{\partial}_{\varepsilon_k}\varphi(x_k)$ . Note that  $\widehat{\partial}_\varepsilon\varphi(\cdot)$  can be replaced by  $\widehat{\partial}\varphi(\cdot)$  in (2.1) when  $\varphi$  is lower semicontinuous around  $\bar{x}$ .

Given  $\Omega \subset \mathbb{R}^n$  with its *indicator function*  $\delta(x; \Omega) = 0$  if  $x \in \Omega$  and  $\delta(x; \Omega) = \infty$  otherwise, the *Fréchet normal cone* and the *limiting normal cone* to  $\Omega$  at  $x$  are defined, respectively, by

$$\widehat{N}(x; \Omega) = \widehat{\partial}\delta(x; \Omega) \text{ and } N(x; \Omega) = \partial\delta(x; \Omega).$$

Obviously,  $\widehat{N}(x; \Omega) \subset N(x; \Omega)$  and

$$x^* \in \widehat{N}(x; \Omega) \Leftrightarrow \limsup_{u \xrightarrow{\Omega} x} \frac{\langle x^*, u - x \rangle}{\|u - x\|} \leq 0.$$

Let  $F: \mathbb{R}^n \rightrightarrows \mathbb{R}^m$  be a set-valued mapping with the *graph*

$$\text{gph } F = \{(x, y) \in \mathbb{R}^n \times \mathbb{R}^m : y \in F(x)\}.$$

The *limiting coderivative*  $D^*F(\bar{x}, \bar{y}): \mathbb{R}^m \rightrightarrows \mathbb{R}^n$  of  $F$  at  $(\bar{x}, \bar{y}) \in \text{gph } F$  is defined by

$$D^*F(\bar{x}, \bar{y})(y^*) = \{x^* \in \mathbb{R}^n : (x^*, -y^*) \in N((\bar{x}, \bar{y}); \text{gph } F)\}.$$

We omit  $\bar{y} = f(\bar{x})$  in the above coderivative notion if  $F = f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is single-valued. If  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is *strictly differentiable* at  $\bar{x}$  in the sense that

$$\lim_{x, u \rightarrow \bar{x}} \frac{f(x) - f(u) - \langle \nabla f(\bar{x}), x - u \rangle}{\|x - u\|} = 0$$

with the derivative operator  $\nabla f(\bar{x}): \mathbb{R}^n \rightarrow \mathbb{R}^m$ , being linear continuous, then  $D^*f(\bar{x})(y^*) = \{\nabla f(\bar{x})^* y^*\}$  for all  $y^* \in \mathbb{R}^m$ . Therefore, the limiting coderivative is an extension of the *adjoint derivative* operator of the classical derivative to nonsmooth functions and set-valued mappings.

Let  $\varphi: \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$  be an extended real-valued function with a finite value at  $\bar{x}$ . Given  $\bar{y} \in \partial\varphi(\bar{x})$ , the mapping  $\partial^2\varphi(\bar{x}, \bar{y}): \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  defined by

$$\partial^2\varphi(\bar{x}, \bar{y})(u) = (D^*\partial\varphi)(\bar{x}, \bar{y})(u), \quad u \in \mathbb{R}^n,$$

is called the *limiting second-order subdifferential* of  $\varphi$  at  $\bar{x}$  relative to  $\bar{y}$ . If  $\varphi$  is twice continuously differentiable at  $\bar{x}$  and  $\bar{y} \in \partial\varphi(\bar{x})$  (actually,  $\bar{y} = \nabla\varphi(\bar{x})$ ), then

$$\partial^2\varphi(\bar{x}, \bar{y})(u) = \{\nabla^2\varphi(\bar{x})(u)\} \text{ for all } u \in \mathbb{R}^n,$$

which is known as the symmetric Hessian matrix. The reader can find various properties and calculus rules for the limiting second-order subdifferential with a number of applications in [7, 8, 10] and the references therein.

**Theorem 2.1.** (see [3, Theorem 3.2]). *Let  $\varphi : \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$  be proper lower semicontinuous. If  $\varphi$  is convex, then*

$$\langle z, u \rangle \geq 0 \text{ for all } u \in \mathbb{R}^n \text{ and } z \in \partial^2\varphi(x, y)(u) \text{ with } (x, y) \in \text{gph}\partial\varphi;$$

that is, for every  $(x, y) \in \text{gph}\partial\varphi$ , the mapping  $\partial^2\varphi(x, y) : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  is positive semi-definite (PSD).

### 3. CHARACTERIZATION OF CONVEXITY

Recall that a function  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  is said to be *piecewise  $C^2$*  if there exist families  $\{P_1, \dots, P_k\}$  of polyhedral convex sets in  $\mathbb{R}^n$  and twice continuously differentiable functions  $\varphi_i : \mathbb{R}^n \rightarrow \mathbb{R}$  such that  $\mathbb{R}^n = \bigcup_{i=1}^k P_i$ ,  $\text{int}P_i \cap \text{int}P_j = \emptyset$  for all  $i \neq j$ , and

$$(3.1) \quad \varphi(x) = \varphi_i(x) \text{ for any } x \in P_i, \quad i \in \{1, \dots, k\}.$$

From (3.1) it follows that  $\varphi_i(x) = \varphi_j(x)$  whenever  $x \in P_i \cap P_j$  and  $i, j \in \{1, \dots, k\}$ .

We need the following two lemmas taken from [3].

**Lemma 3.1.** *If  $I := \{i \in \{1, 2, \dots, k\} \mid \text{int}P_i \neq \emptyset\}$ , then  $\bigcup_{i \in I} P_i = \mathbb{R}^n$ .*

**Lemma 3.2.** *Let  $[x, y]$  be an interval in  $\mathbb{R}^n$  ( $x \neq y$ ),  $0 = \tau_0 < \tau_1 < \dots < \tau_{m-1} < \tau_m = 1$  ( $m \in \mathbb{N}$ ,  $m > 1$ ), and  $x_i := x + \tau_i(y - x)$  ( $i = 0, 1, \dots, m$ ). Suppose that  $\varphi$  is nonconvex and continuous on  $[x, y]$ . Then there must exist  $i \in \{0, 1, \dots, m-2\}$  such that  $\varphi$  is nonconvex on  $[x_i, x_{i+2}]$ .*

We are now ready to state and prove the main result of this paper.

**Theorem 3.3.** *Suppose that  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  is a piecewise  $C^2$  function. Then  $\varphi$  is convex if and only if*

$$(3.2) \quad \langle z, u \rangle \geq 0 \text{ for all } u \in \mathbb{R}^n, z \in \partial^2\varphi(x, y)(u) \text{ with } (x, y) \in \text{gph}\partial\varphi.$$

*Proof.* The necessary condition is due to Theorem 2.1. It remains to prove the sufficient condition. By Lemma 3.1, we can assume that  $\text{int}P_i \neq \emptyset$  for all  $i \in \{1, 2, \dots, k\}$ . Suppose that (3.2) holds but  $\varphi$  is nonconvex. Since  $\varphi$  is twice continuously differentiable on  $\text{int}P_i$ ,  $\partial^2\varphi(x, y)(u) = \{\nabla^2\varphi(x)(u)\}$  for all  $x \in \text{int}P_i$ ,  $y \in \partial\varphi(x)$  and  $u \in \mathbb{R}^n$ . Together with (3.2) this implies that  $\nabla^2\varphi(x)$  is positive semi-definite on  $\text{int}P_i$ . By the classical result on characterizing the convexity of  $C^2$  functions,  $\varphi$  is convex on  $P_i$  ( $i = 1, 2, \dots, k$ ). We consider the following two cases.

**Case 1.**  $k = 2$ . Let  $P_1 = \{x \in \mathbb{R}^n : \langle a, x \rangle \leq \alpha\}$ ,  $P_2 = \{x \in \mathbb{R}^n : \langle a, x \rangle \geq \alpha\}$  ( $a \in \mathbb{R}^n \setminus \{0\}$ ,  $\alpha \in \mathbb{R}$ ),  $P_{12} = P_1 \cap P_2$  and

$$\varphi(x) = \begin{cases} \varphi_1(x) & \text{if } x \in P_1, \\ \varphi_2(x) & \text{if } x \in P_2, \end{cases}$$

where  $\varphi_1, \varphi_2 \in C^2$  and  $\varphi_1(x) = \varphi_2(x)$  for all  $x \in P_{12}$ . Observe that  $\mathbb{R}^n$  is the union of disjoint nonempty sets  $\text{int}P_1$ ,  $\text{int}P_2$ ,  $P_{12}$ , and

$$(3.3) \quad \partial\varphi(x) = \widehat{\partial}\varphi(x) = \begin{cases} \{\nabla\varphi_1(x)\} & \text{if } x \in \text{int}P_1, \\ \{\nabla\varphi_2(x)\} & \text{if } x \in \text{int}P_2. \end{cases}$$

Since  $\varphi$  is convex on each one of the convex sets  $P_1$  and  $P_2$  but it is nonconvex on  $\mathbb{R}^n = P_1 \cup P_2$ , there exist  $x_0 \in \text{int}P_1$ ,  $y_0 \in \text{int}P_2$  and  $t_1 \in (0, 1)$  such that

$$(3.4) \quad \varphi(z_1) > (1 - t_1)\varphi(x_0) + t_1\varphi(y_0),$$

where  $z_1 = (1 - t_1)x_0 + t_1y_0$ . We will prove that

$$(3.5) \quad \varphi(z_0) > (1 - t_0)\varphi(x_0) + t_0\varphi(y_0)$$

with  $z_0 = (1 - t_0)x_0 + t_0y_0 \in P_{12}$  ( $t_0 \in (0, 1)$ ). If  $t_0 = t_1$  then (3.5) follows from (3.4), because  $z_1 = z_0$ . If  $t_0 \in (0, t_1)$  then  $z_1 = (1 - \lambda)y_0 + \lambda z_0$  with  $\lambda = (1 - t_1)/(1 - t_0) \in (0, 1)$ . Since  $\varphi$  is convex on  $[z_0, y_0] \subset P_2$ ,  $\varphi(z_1) \leq (1 - \lambda)\varphi(y_0) + \lambda\varphi(z_0)$ . Combining this fact with (3.4), we obtain

$$\begin{aligned} \varphi(z_0) &> \lambda^{-1}[(1 - t_1)\varphi(x_0) + t_1\varphi(y_0) - (1 - \lambda)\varphi(y_0)] \\ &= (1 - t_0)\varphi(x_0) + t_0\varphi(y_0), \end{aligned}$$

which gives (3.5). Similarly, (3.5) is also valid if  $t_0 \in (t_1, 1)$ . Therefore (3.5) holds. Since  $x_0 \in P_1$ ,  $y_0 \in P_2$  and  $z_0 \in P_{12}$ , by (3.5), we have

$$(1 - t_0)\varphi_1(z_0) + t_0\varphi_2(z_0) > (1 - t_0)\varphi_1(x_0) + t_0\varphi_2(y_0)$$

or in other words,

$$(3.6) \quad (1 - t_0)(\varphi_1(z_0) - \varphi_1(x_0)) + t_0(\varphi_2(z_0) - \varphi_2(y_0)) > 0.$$

According to the mean value theorem, we have

$$\varphi_1(z_0) - \varphi_1(x_0) = \langle \nabla \varphi_1(a_1), z_0 - x_0 \rangle \quad \text{and} \quad \varphi_2(z_0) - \varphi_2(y_0) = \langle \nabla \varphi_2(a_2), z_0 - y_0 \rangle,$$

for some  $a_1 \in (x_0, z_0)$  and  $a_2 \in (z_0, y_0)$ . Note that  $z_0 = (1 - t_0)x_0 + t_0y_0$  and  $t_0 \in (0, 1)$ . By (3.6),

$$(3.7) \quad \langle \nabla \varphi_1(a_1) - \nabla \varphi_2(a_2), y_0 - x_0 \rangle > 0.$$

Our next task is to show

$$(3.8) \quad \langle \nabla \varphi_1(z_0) - \nabla \varphi_2(z_0), y_0 - x_0 \rangle > 0.$$

Assume by contradiction that  $\langle \nabla \varphi_1(z_0) - \nabla \varphi_2(z_0), y_0 - x_0 \rangle \leq 0$ . Since  $\varphi_1$  is convex on  $[a_1, z_0]$ ,

$$\langle \nabla \varphi_1(z_0) - \nabla \varphi_1(a_1), z_0 - a_1 \rangle \geq 0$$

from which we get

$$\langle \nabla \varphi_1(z_0) - \nabla \varphi_1(a_1), y_0 - x_0 \rangle \geq 0.$$

Similarly,  $\langle \nabla \varphi_2(a_2) - \nabla \varphi_2(z_0), y_0 - x_0 \rangle \geq 0$ . Hence

$$\begin{aligned} \langle \nabla \varphi_1(a_1) - \nabla \varphi_2(a_2), y_0 - x_0 \rangle &\leq \langle \nabla \varphi_1(z_0) - \nabla \varphi_2(z_0), y_0 - x_0 \rangle \\ &\leq 0. \end{aligned}$$

This contradicts (3.7) and thus (3.8) is valid.

We claim that  $\widehat{\partial}\varphi(z_0) = \emptyset$ . Suppose that it is not true. Then there exists  $x^* \in \mathbb{R}^n$  satisfying

$$(3.9) \quad \liminf_{u \rightarrow z_0} \frac{\varphi(u) - \varphi(z_0) - \langle x^*, u - z_0 \rangle}{\|u - z_0\|} \geq 0.$$

Let  $u_j := z_0 - \frac{1}{j}(y_0 - x_0)$ . Then  $u_j \rightarrow z_0$  as  $j \rightarrow \infty$ . It is easy to see that  $u_j \in P_1$  for all  $j \in \mathbb{N}$ . Together with (3.9) this gives

$$\liminf_{j \rightarrow \infty} \frac{\varphi_1(u_j) - \varphi_1(z_0) - \langle x^*, u_j - z_0 \rangle}{\|u_j - z_0\|} \geq 0.$$

By the mean value theorem,

$$\liminf_{j \rightarrow \infty} \frac{\langle \nabla \varphi_1(\xi_j), -\frac{1}{j}(y_0 - x_0) \rangle - \langle x^*, -\frac{1}{j}(y_0 - x_0) \rangle}{\frac{1}{j}\|y_0 - x_0\|} \geq 0,$$

where  $\xi_j \in (u_j, z_0)$ . Since  $\nabla \varphi_1(\cdot)$  is continuous and  $\xi_j \rightarrow z_0$  as  $j \rightarrow \infty$ , we have

$$\langle \nabla \varphi_1(z_0), y_0 - x_0 \rangle \leq \langle x^*, y_0 - x_0 \rangle.$$

Similarly, by taking  $u_j = z_0 + \frac{1}{j}(y_0 - x_0)$  we obtain

$$\langle x^*, y_0 - x_0 \rangle \leq \langle \nabla \varphi_2(z_0), y_0 - x_0 \rangle.$$

Consequently,  $\langle \nabla \varphi_1(z_0) - \nabla \varphi_2(z_0), y_0 - x_0 \rangle \leq 0$  which contradicts (3.8). Hence  $\widehat{\partial} \varphi(z_0) = \emptyset$  and  $\nabla \varphi_1(z_0) \neq \nabla \varphi_2(z_0)$  by (3.8). By virtual of (3.5), we can find a positive number  $\gamma$  such that for each  $u \in P_{12} \cap (z_0 + \gamma \mathbb{B})$  there exist  $x_u \in \text{int} P_1$ ,  $y_u \in \text{int} P_2$  satisfying  $u = (1 - t_0)x_u + t_0 y_u$  and

$$\varphi(u) > (1 - t_0)\varphi(x_u) + t_0\varphi(y_u),$$

where  $\mathbb{B} := \{x \in \mathbb{R}^n \mid \|x\| < 1\}$ . Then as in the proof of the claim  $\widehat{\partial} \varphi(z_0) = \emptyset$ , we can show that  $\widehat{\partial} \varphi(u) = \emptyset$  and  $\nabla \varphi_1(u) \neq \nabla \varphi_2(u)$  for all  $u \in P_{12} \cap (z_0 + \gamma \mathbb{B})$ . By the continuity of  $\nabla \varphi_1(\cdot)$  and of  $\nabla \varphi_2(\cdot)$ , together with (3.3) this gives

$$\begin{aligned} \partial \varphi(x) &= \text{Lim sup}_{u \rightarrow x} \widehat{\partial} \varphi(u) \\ &= \text{Lim sup}_{\substack{\text{int} P_1 \\ u \rightarrow x}} \widehat{\partial} \varphi(u) \cup \text{Lim sup}_{\substack{\text{int} P_2 \\ u \rightarrow x}} \widehat{\partial} \varphi(u) \cup \text{Lim sup}_{\substack{P_{12} \\ u \rightarrow x}} \widehat{\partial} \varphi(u) \\ &= \{\nabla \varphi_1(x), \nabla \varphi_2(x)\}, \end{aligned}$$

for all  $x \in P_{12} \cap (z_0 + \gamma \mathbb{B})$ . Hence

$$(3.10) \quad \partial \varphi(x) = \begin{cases} \{\nabla \varphi_1(x)\} & \text{if } x \in \text{int} P_1, \\ \{\nabla \varphi_2(x)\} & \text{if } x \in \text{int} P_2, \\ \{\nabla \varphi_1(x), \nabla \varphi_2(x)\} & \text{if } x \in P_{12} \cap (z_0 + \gamma \mathbb{B}). \end{cases}$$

For  $x \in P_{12} \cap (z_0 + \gamma \mathbb{B})$ ,  $y = \nabla \varphi_1(x)$ , and  $u \in \mathbb{R}^n$ , it holds

$$\partial^2 \varphi(x, y)(u) = \nabla^2 \varphi_1(x)(u) + \mathbb{R}_+ a.$$

Indeed, let  $z = \nabla^2 \varphi_1(x)(u) + \lambda a$  for some  $\lambda \geq 0$ . Since  $\nabla \varphi_1(\cdot)$ ,  $\nabla \varphi_2(\cdot)$  are continuous and  $y = \nabla \varphi_1(x) \neq \nabla \varphi_2(x)$ , by (3.10) for all  $(x', y') \in \text{gph} \partial \varphi$  near

$(x, y)$  we have  $x' \in P_1$  and  $y' = \nabla\varphi_1(x')$ . Hence

$$\begin{aligned}
& \limsup_{(x', y') \xrightarrow{\text{gph}\partial\varphi} (x, y)} \frac{\langle z, x' - x \rangle - \langle u, y' - y \rangle}{\|x' - x\| + \|y' - y\|} \\
&= \limsup_{x' \xrightarrow{P_1} x} \frac{\langle z, x' - x \rangle - \langle u, \nabla\varphi_1(x') - \nabla\varphi_1(x) \rangle}{\|x' - x\| + \|\nabla\varphi_1(x') - \nabla\varphi_1(x)\|} \\
&= \limsup_{x' \xrightarrow{P_1} x} \frac{\langle z - \nabla^2\varphi_1(\xi_{x'})(u), x' - x \rangle}{\|x' - x\| + \|\nabla\varphi_1(x') - \nabla\varphi_1(x)\|} \\
&= \limsup_{x' \xrightarrow{P_1} x} \frac{\langle \nabla^2\varphi_1(x)(u) - \nabla^2\varphi_1(\xi_{x'})(u), x' - x \rangle + \lambda\langle a, x' - x \rangle}{\|x' - x\| + \|\nabla\varphi_1(x') - \nabla\varphi_1(x)\|} \\
&\leq \|u\| \limsup_{x' \xrightarrow{P_1} x} \|\nabla^2\varphi_1(x) - \nabla^2\varphi_1(\xi_{x'})\| = 0,
\end{aligned}$$

where  $\xi_{x'} \in (x', x)$ . This implies that  $z \in \partial^2\varphi(x, y)(u)$  and thus,

$$\nabla^2\varphi_1(x)(u) + \mathbb{R}_+a \subset \partial^2\varphi(x, y)(u).$$

To prove the reverse inclusion, take any  $z \in \partial^2\varphi(x, y)(u)$ . Then there exist  $(z_i, u_i) \rightarrow (z, u)$  and  $(x_i, y_i) \rightarrow (x, y)$  with  $(x_i, y_i) \in \text{gph}\partial\varphi$  such that  $(z_i, -u_i) \in \widehat{N}((x_i, y_i); \text{gph}\partial\varphi)$  for all  $i$ . Note that  $\nabla\varphi_1(\cdot), \nabla\varphi_2(\cdot)$  are continuously differentiable functions satisfying  $\nabla\varphi_1(x) \neq \nabla\varphi_2(x)$  for all  $x \in P_{12} \cap (z_0 + \gamma\mathbb{B})$ . By (3.10), we may assume that  $x_i \in P_1 \cap (z_0 + \gamma\mathbb{B})$ ,  $y_i = \nabla\varphi_1(x_i)$  for all  $i$ . Hence,

$$\begin{aligned}
& (z_i, -u_i) \in \widehat{N}((x_i, y_i); \text{gph}\partial\varphi) \\
&\Leftrightarrow \limsup_{x' \xrightarrow{P_1} x_i} \frac{\langle z_i, x' - x_i \rangle - \langle u_i, \nabla\varphi_1(x') - \nabla\varphi_1(x_i) \rangle}{\|x' - x_i\| + \|\nabla\varphi_1(x') - \nabla\varphi_1(x_i)\|} \leq 0 \\
&\Rightarrow \limsup_{x' \xrightarrow{P_1} x_i} \frac{\langle z_i - \nabla^2\varphi_1(\xi_{x'})(u_i), x' - x_i \rangle}{(1 + \sup_{\xi \in z_0 + \gamma\mathbb{B}} \|\nabla^2\varphi_1(\xi)\|)\|x' - x_i\|} \leq 0 \quad (\text{for some } \xi_{x'} \in (x', x_i)) \\
&\Rightarrow \langle z_i - \nabla^2\varphi_1(x_i)(u_i), x' \rangle \leq 0 \quad \text{whenever } \langle a, x' \rangle \leq 0.
\end{aligned}$$

Taking  $i \rightarrow \infty$ , we have  $\langle z - \nabla^2\varphi_1(x)(u), x' \rangle \leq 0$  if  $\langle a, x' \rangle \leq 0$ . By the Farkas lemma, there exists  $\lambda \geq 0$  such that  $z - \nabla^2\varphi_1(x)(u) = \lambda a$  which proves  $\partial^2\varphi(x, y)(u) \subset \nabla^2\varphi_1(x)(u) + \mathbb{R}_+a$ . Therefore,  $\partial^2\varphi(x, y)(u) = \nabla^2\varphi_1(x)(u) + \mathbb{R}_+a$  for all  $x \in P_{12} \cap (z_0 + \gamma\mathbb{B})$ ,  $y = \nabla\varphi_1(x)$  and  $u \in \mathbb{R}^n$ . Let  $x \in P_{12} \cap (z_0 + \gamma\mathbb{B})$ ,  $y = \nabla\varphi_1(x)$ ,  $z = -\nabla^2\varphi_1(x)(a) + ta$  ( $t \geq 0$ ) and  $u = -a$ . We have  $z \in \partial^2\varphi(x, y)(u)$  and  $\langle z, u \rangle = \langle \nabla^2\varphi_1(x)(a), a \rangle - t\|a\|^2 < 0$  for  $t \geq 0$  large enough. This contradicts (3.2).

**Remark.** As can be seen from the above proof, we also obtain the contradiction if it is only supposed that  $\varphi$  is nonconvex on some ball  $\bar{x} + \varepsilon\mathbb{B}$  ( $\bar{x} \in \mathbb{R}^n$ ,  $\varepsilon > 0$ ) and (3.2) is replaced by the condition:

$$\langle z, u \rangle \geq 0 \text{ for all } u \in \mathbb{R}^n, z \in \partial^2\varphi(x, y)(u) \text{ with } (x, y) \in \text{gph}\partial\varphi \text{ and } x \in \bar{x} + \varepsilon\mathbb{B}.$$

This remark will be used in the sequel.

**Case 2.**  $k > 2$ . Since  $\varphi$  is nonconvex on  $\mathbb{R}^n = \bigcup_{j=1}^k P_j$  but it is convex on each one of the polyhedrals  $P_j$  ( $j = 1, 2, \dots, k$ ), there exist  $x, y \in \mathbb{R}^n$  ( $x \neq y$ ),  $0 = \tau_0 < \tau_1 < \dots < \tau_{m-1} < \tau_m = 1$  ( $m \in \mathbb{N}$ ,  $m > 1$ ), and  $x_i := x + \tau_i(y - x)$  ( $i = 0, 1, \dots, m$ ) such that  $\varphi$  is convex on  $[x_i, x_{i+1}]$  ( $i = 0, 1, \dots, m-1$ ) but it is nonconvex on  $[x, y]$ . By Lemma 3.2 we can find  $i \in \{0, 1, \dots, m-2\}$  such that  $\varphi$  is nonconvex on  $[x_i, x_{i+2}]$ . Thus, without loss of generality we can assume that there exists  $\bar{x} \in (x, y)$  such that  $\varphi$  is convex on each one of intervals  $[x, \bar{x}]$  and  $[\bar{x}, y]$  but it is nonconvex on  $[x, y]$ . For each  $u \in \mathbb{R}^n$ , we put  $I(u) = \{i \in \{1, 2, \dots, k\} : u \in P_i\}$ . Let  $\varepsilon > 0$  such that  $(\bar{x} + \varepsilon\mathbb{B}) \cap P_i = \emptyset$  for all  $i \in \{1, 2, \dots, k\} \setminus I(\bar{x})$ . We may assume that  $x, y \in \mathbb{B}(\bar{x}, \varepsilon)$ . Since  $\varphi$  is convex on  $[x, \bar{x}]$  and on  $[\bar{x}, y]$  and it is nonconvex on  $[x, y]$ ,  $|I(\bar{x})| \geq 2$  and  $\bar{x} \notin \text{int}P_i$  for all  $i$ . If  $|I(\bar{x})| = 2$ , then we obtain a contradiction by using the above remark. If  $|I(\bar{x})| > 2$ , then  $\dim L < n-1$  where  $L := \text{aff}(\bigcap_{i \in I(\bar{x})} P_i)$  denotes the affine hull of  $\bigcap_{i \in I(\bar{x})} P_i$ . Indeed, without loss of generality we may assume that  $\{1, 2, 3\} \subset I(\bar{x})$ . Since  $\text{int}P_i \neq \emptyset$ ,  $\text{int}P_j \neq \emptyset$  and  $\text{int}P_i \cap \text{int}P_j = \emptyset$  ( $\forall i \neq j$ ), by the separation theorem, there exists a hyperplane  $L_{ij}$  separating the sets  $\text{int}P_i$  and  $\text{int}P_j$  ( $1 \leq i < j \leq 3$ ). Since it is impossible that  $L_{12} = L_{13} = L_{23}$ ,  $\dim(L_{12} \cap L_{13} \cap L_{23}) < n-1$ . Noting that  $L \subset (L_{12} \cap L_{13} \cap L_{23})$ , we have  $\dim L < n-1$ . In the case where  $y \in L$ , by invoking the last property we can find  $\tilde{y} \in \mathbb{R}^n \setminus L$  as close to  $y$  as desired. Let  $t \in (0, 1)$  be such that  $\bar{x} = (1-t)x + ty$ . Define  $\tilde{x}_{\tilde{y}}$  by the condition  $\bar{x} = (1-t)\tilde{x}_{\tilde{y}} + t\tilde{y}$ . Clearly,  $\tilde{x}_{\tilde{y}} \notin L$  and  $\tilde{x}_{\tilde{y}} \rightarrow x$  as  $\tilde{y} \rightarrow y$ . Since  $\varphi$  is continuous and nonconvex on  $[x, y]$ , there exists  $\tilde{y} \in \mathbb{R}^n \setminus L$  as close to  $y$  as desired such that  $\varphi$  is nonconvex on  $[\tilde{x}_{\tilde{y}}, \tilde{y}]$ . Thus, replacing  $(x, y)$  by  $(\tilde{x}_{\tilde{y}}, \tilde{y})$  if necessary, we can assume that  $y \notin L$  and  $x \notin L$ . (Note that such replacement may destroy the property of  $\varphi$  of being convex on each one of the segments  $[x, \bar{x}]$  and  $[\bar{x}, y]$ . But this property will not be employed in the sequel.) Take  $\rho > 0$  such that  $(y + \rho\mathbb{B}) \subset (\bar{x} + \varepsilon\mathbb{B})$ ,  $(y + \rho\mathbb{B}) \cap L = \emptyset$ ,  $x \notin (y + \rho\mathbb{B})$ , and  $\varphi$  is nonconvex on  $[x, z]$  for each  $z \in (y + \rho\mathbb{B})$ . Our aim now is to show that there exists  $z \in (y + \rho\mathbb{B})$  such that  $[x, z] \cap L = \emptyset$ . Suppose that this is not true. Then  $[x, z] \cap L \neq \emptyset$  for all  $z \in (y + \rho\mathbb{B})$ . Choose  $y_i \in (y + \rho\mathbb{B})$  ( $i = 1, 2, \dots, n-1$ ) such that  $\{x - y, y_1 - y, \dots, y_{n-1} - y\}$  is linearly independent. For each  $i \in \{1, 2, \dots, n-1\}$ , we can take a vector  $\bar{x}_i \in [x, y_i] \cap L$  because  $[x, z] \cap L \neq \emptyset$  for all  $z \in (y + \rho\mathbb{B})$  and  $y_i \in (y + \rho\mathbb{B})$  ( $i = 1, 2, \dots, n-1$ ). Note that  $\bar{x}_i - \bar{x} = \alpha_i(x - y) + \beta_i(y_i - y)$  for

some  $\alpha_i \in \mathbb{R}, \beta_i \in \mathbb{R} \setminus \{0\}$  and  $\{x - y, y_1 - y, \dots, y_{n-1} - y\}$  is linearly independent. Hence the system  $\{\bar{x}_1 - \bar{x}, \dots, \bar{x}_{n-1} - \bar{x}\}$  is linearly independent from which we get  $\dim L \geq n - 1$ . This contradicts the fact  $\dim L < n - 1$  derived above and thus there exists  $z \in (y + \rho\mathbb{B})$  satisfying  $[x, z] \cap L = \emptyset$ . Since  $\varphi$  is nonconvex on  $[x, z]$ , we can find  $[x', y'] \subset [x, z]$  and  $\bar{x}' \in (x', y')$  such that  $\varphi$  is convex on each of the two intervals  $[x', \bar{x}']$  and  $[\bar{x}', y']$  and it is nonconvex on  $[x', y']$ . Observing that  $\bar{x}' \in (\bar{x} + \varepsilon\mathbb{B}) \setminus [\bigcap_{i \in I(\bar{x})} P_i]$  and  $(\bar{x} + \varepsilon\mathbb{B}) \cap P_i = \emptyset$  for all  $i \in \{1, 2, \dots, k\} \setminus I(\bar{x})$ , we have  $|I(\bar{x}')| < |I(\bar{x})|$ . Hence if  $|I(\bar{x})| > 2$ , then there exist  $[x', y']$  and  $\bar{x}' \in (x', y')$  such that  $\varphi$  is convex on each of the segments  $[x', \bar{x}']$  and  $[\bar{x}', y']$  but it is nonconvex on  $[x', y']$  and  $|I(\bar{x}')| < |I(\bar{x})|$ . Thus, by repeating this procedure after finitely many times, we can reduce the case where  $|I(\bar{x})| = 2$  and obtain a contradiction. The proof is now completed.  $\blacksquare$

Recall that a function  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$  is said to be *strongly convex* on a convex subset  $\Omega \subset \text{dom}\varphi$  if there exists a constant  $\rho > 0$  such that

$$\varphi((1-t)x + ty) \leq (1-t)\varphi(x) + t\varphi(y) - \rho t(1-t)\|x - y\|^2$$

for any  $x, y \in \Omega$  and  $t \in (0, 1)$ . It is well known (see e.g. [12, Lemma 1, p. 184]) that the above condition is fulfilled if and only if the function  $\tilde{\varphi}(x) := \varphi(x) - \rho\|x\|^2$  is convex on  $\Omega$ .

We now have the following characterization of strong convexity for piecewise  $C^2$  functions.

**Theorem 3.4.** *Suppose that  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  is a piecewise  $C^2$  function. Then  $\varphi$  is strongly convex on  $\mathbb{R}^n$  with the constant  $\rho > 0$  if and only if for any  $(x, y) \in \text{gph}\partial\varphi$  the second-order subdifferential mapping  $\partial^2\varphi(x, y) : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  satisfies the condition*

$$(3.11) \quad \langle z, u \rangle \geq 2\rho\|u\|^2 \text{ for all } u \in \mathbb{R}^n \text{ and } z \in \partial^2\varphi(x, y)(u) \text{ with } (x, y) \in \text{gph}\partial\varphi.$$

*Proof.* It is well-known that  $\varphi$  is strongly convex on  $\mathbb{R}^n$  with the constant  $\rho > 0$  if and only if the function  $\tilde{\varphi} := \varphi + \psi$  where  $\psi(x) = -\rho\|x\|^2$  is convex. By [7, Proposition 1.107(ii)],

$$(3.12) \quad \partial\tilde{\varphi}(x) = \partial\varphi(x) - 2\rho x \quad \forall x \in \mathbb{R}^n.$$

Now, applying the coderivative sum rule with equality [7, Proposition 1.62(ii)] to the case where  $F(x) = \partial\varphi(x)$  and  $f(x) = -2\rho x$ , we have

$$D^*(F + f)(x, y)(u) = D^*F(x, y - f(x))(u) - 2\rho u$$

for any  $x \in \mathbb{R}^n$ ,  $y \in \mathbb{R}^n$  with  $y - f(x) \in F(x)$  and for any  $u \in \mathbb{R}^n$ . Together with (3.12) this gives

$$(3.13) \quad \partial^2 \tilde{\varphi}(x, y)(u) = \partial \varphi^2(x, y)(u) - 2\rho u \quad \forall x \in \mathbb{R}^n, \forall y \in \partial \varphi(x).$$

According to Theorem 3.3, the convexity of  $\tilde{\varphi}$  is equivalent to the PSD of the second-order subdifferential mapping  $\partial^2 \tilde{\varphi}(\cdot)$ . Hence, by (3.13) we obtain  $\langle v - 2\rho u, u \rangle \geq 0$  for any  $v \in \partial \varphi^2(x, y)(u)$  which yields (3.11). ■

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