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Variational Analysis Approach for Composite Optimality Problems

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Classical optimality conditions

Preliminaries from variational analysis

Twice epi-Differentiability for composite functions

The second order optimality conditions for composite optimization



1. Classical optimality conditions

1. Classical optimality conditions

$f: S \subset \mathbb{R}^n \rightarrow \mathbb{R}$. Consider the following minimization problem:

$$(P) \quad \min_{x \in S} f(x).$$

- A point $\bar{x} \in S$ is called a local minimum of (P) if $\exists \delta > 0$ s. t.

$$f(x) \geq f(\bar{x}) \quad \text{for all } x \in S \text{ such that } \|x - \bar{x}\| \leq \delta.$$

The point \bar{x} is called the global minimum if $f(x) \geq f(\bar{x})$ for all $x \in S$.

- We say that f satisfying the γ -order growth condition at \bar{x} , if there exists some $\kappa, \delta > 0$ such that

$$f(x) - f(\bar{x}) \geq \kappa \|x - \bar{x}\|^\gamma \quad \text{for all } \|x - \bar{x}\| \leq \delta.$$

In particular, if $\gamma = 1, 2$ we say f satisfying the first (second) order growth condition at \bar{x} .

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When the space is \mathbb{R}^1 , the problem (P) becomes the classical extreme problem, and the first necessary condition is given by Fermat 1638 and Newton 1670.

Lemma 1.1 (Fermat 1638; Newton 1670)

Let $\varphi: O \subset \mathbb{R} \rightarrow \mathbb{R}$ be a differentiable function defined on an open set O . If φ attains its local minimum(or maximum) at \bar{x} , then

$$\varphi'(\bar{x}) = 0.$$

If f is differentiable at a point $x \in S$, then the gradient and Hessian matrix are denoted by

$$\nabla f(x) = \left(\frac{\partial f(x)}{\partial x_1}, \dots, \frac{\partial f(x)}{\partial x_n} \right),$$
$$\nabla^2 f(x) = \left[\frac{\partial^2 f(x)}{\partial x_i \partial x_j} \right], \quad i, j = 1, \dots, n.$$

Theorem 1.1 (Euler 1755)(The first order necessary condition)

Let f be differentiable at $\bar{x} \in \text{int}S$. If f attains its local minimum (or maximum) at \bar{x} , then

$$\nabla f(\bar{x}) = 0. \quad (1.1)$$

If f is convex, but not necessarily differentiable, then the necessary and sufficient condition of a local minimizer becomes

$$0 \in \partial f(\bar{x})$$

When \bar{x} is not an interior of S , in order to characterize the necessary property of local minimizer, we need the concepts of the tangent cone of S to \bar{x} .

Definition 1.1

$S \subset \mathbb{R}^n$ nonempty, $\bar{x} \in S$. The contingent (Bouligand) cone of S to \bar{x} is defined as

$$\begin{aligned} T_S(\bar{x}) : &= \{d \in X : \exists t_k \downarrow 0, \exists S \ni x_k \rightarrow \bar{x}, \frac{x_k - \bar{x}}{t_k} \rightarrow d\} \\ &= \{d \in X : \exists t_k \downarrow 0, d_k \rightarrow d \text{ s.t. } \bar{x} + t_k d_k \in S\}. \end{aligned}$$

Theorem 1.2 (the first order optimality condition)

Assume that \bar{x} is a local minimum of problem (P) and that f is differentiable at \bar{x} . Then

$$\nabla f(\bar{x})^T d \geq 0, \quad \forall d \in T_S(\bar{x}) \Leftrightarrow -\nabla f(\bar{x}) \in [T_S(\bar{x})]^o$$

- Usually, the negative polar of $T_S(\bar{x})$ is called Fréchet normal cone of S to \bar{x} , denoted by $\hat{N}_S(\bar{x})$. Equivalently,

$$\hat{N}_S(\bar{x}) = \{x^* \in X^* \mid \limsup_{x' \xrightarrow{S} \bar{x}} \frac{\langle x^*, x' - \bar{x} \rangle}{\|x' - \bar{x}\|} \leq 0\}.$$

- When S is convex,

$$T_S(\bar{x}) = \overline{\text{cone}(S - \bar{x})}$$

and

$$\hat{N}_S(\bar{x}) = N_S(\bar{x}) = \{s \in \mathbb{R}^n : \langle s, x - \bar{x} \rangle \leq 0, \quad \forall x \in S\}.$$

When S is defined by some constraint functions, for instance,

$$S := \{x \in \mathbb{R}^n : g_i(x) \leq 0, i = 1, 2, \dots, p; \\ h_j(x) = 0, j = 1, 2, \dots, q\}, \quad (1.2)$$

where g_i, h_j are differentiable functions defined on \mathbb{R}^n , or more general form

$$S := \{x \in \mathbb{R}^n : F(x) \in K\}, \quad (1.3)$$

where $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a differentiable mapping, $K \subset \mathbb{R}^m$.

Question:

How to characterize the tangent and the normal cone of S to $\bar{x} \in S$?

It is not hard to know

$$T_S(\bar{x}) \subset \{d \in \mathbb{R}^n : \nabla g_i(\bar{x})^T d \leq 0, i \in I(\bar{x}); \\ \nabla h_j(\bar{x})^T d = 0, j = 1, 2, \dots, q\} =: L_S(\bar{x})$$

or

$$T_S(\bar{x}) \subset \{d \in \mathbb{R}^n : DF(\bar{x})^T d \in T_K(\bar{x})\} =: L_S(\bar{x}).$$

Under what condition, the above inclusions become as equality?

$$T_S(\bar{x}) = L_S(\bar{x})$$

Such condition is called constraint qualification, for example, Mangasarian-Fromovitz, Robinson condition, metric regularity condition and so on.

Constraint qualification

Mangasarian-Fromovitz constraint qualification:

$$\begin{aligned} & \nabla h_j(\bar{x}), j = 1, \dots, q, \text{ are linearly independent,} \\ & \exists d \in X : \nabla h_j(\bar{x})^T d = 0, j = 1, \dots, q, \nabla g_i(\bar{x})^T d < 0, \forall i \in I(\bar{x}), \end{aligned} \quad (1.4)$$

where $I(\bar{x})$ denotes the index set of active at \bar{x} inequality constraints.

Robinson constraint qualification:

$$0 \in \text{int}\{F(\bar{x}) + DF(\bar{x})(\mathbb{R}^n) - K\}. \quad (1.5)$$

Furthermore, In the case of the equality holds, **how to compute the normal cone** $[T_S(\bar{x})]^\circ$?

This is related to Farkas lemma.

From the normal analysis point of view, we easily show that

$$\hat{N}_S(\bar{x}) \supset \{DF(\bar{x})^*y^* | y^* \in \hat{N}_K(F(\bar{x}))\},$$

and

$$N_S(\bar{x}) \subset \{DF(\bar{x})^*y^* | y^* \in N_K(F(\bar{x}))\},$$

whenever the Robinson's condition holds, where $N_S(\bar{x})$ denote the (Morduchovich) limit normal cone to S at \bar{x} .

On the other hand, the problem (P) can be equivalent to write as

$$(P') \quad \min_{x \in X} f(x) + I_S(x).$$

The necessary condition of \bar{x} being a local minimizer of (P') is

$$0 \in \hat{\partial}(f + I_S)(\bar{x}) \subset \partial(f + \delta_S)(\bar{x}),$$

where $\hat{\partial}\phi(\bar{x})$ and $\partial\phi(\bar{x})$ denotes the Fréchet (regular) and Morukhovich subdifferential of ϕ at \bar{x} , respectively.

When f is a continuous convex function and S is a convex set, we have

$$\partial(f + I_S)(\bar{x}) = \partial f(\bar{x}) + \partial I_S(\bar{x}) = \partial f(\bar{x}) + N_S(\bar{x}).$$

How to calculate the subdifferential of the sum of two functions?

In optimization, second derivatives help significantly in the understanding of optimality, especially the formulation of sufficient conditions for local optimality in the absence of convexity. Such conditions form the basis for numerical methodology and assist in studies of what happens to optimal solutions when the parameters on which a problem depends are perturbed.

Theorem 1.3 (The second optimality condition)

Suppose that f is twice differentiable at $\bar{x} \in \text{int}S$. If f attains its local minimum at \bar{x} , then

$$\nabla f(\bar{x}) = 0, \quad \langle d, \nabla^2 f(\bar{x})d \rangle \geq 0, \quad \forall d \in \mathbb{R}^n.$$

Consider the second order tangent vector to a set S at $\bar{x} \in S$.

$$\begin{aligned} T_S^2(\bar{x}, d) &= \{w \in X : \exists t_k \downarrow 0, \exists S \ni x_k \text{ s.t. } \frac{x_k - \bar{x} - t_k d}{\frac{1}{2}t_k^2} \rightarrow w\} \\ &= \{w \in X : \exists t_k \downarrow 0, w_k \rightarrow w \text{ s.t. } \bar{x} + t_k d + \frac{1}{2}t_k^2 w_k \in S\}. \end{aligned}$$

Theorem 1.4

Assume that \bar{x} is a local minimum of problem (P) and that f is twice continuously differentiable at \bar{x} . Then for every $d \in T_S(\bar{x})$ with $\nabla f(\bar{x})^T d = 0$, we have

$$\nabla f(\bar{x})w + \langle d, \nabla^2 f(\bar{x})d \rangle \geq 0, \quad \forall w \in T_S^2(\bar{x}, d), \quad (1.6)$$

Question:

How to calculate the second order tangent set and how to reformulate the inequality (1.6)?

When S is defined by some constraint functions, for instance, has the form of (1.2) or (1.3)(i.e.

$$S := \{x \in \mathbb{R}^n : g_i(x) \leq 0, i = 1, 2, \dots, p; h_j(x) = 0, j = 1, 2, \dots, q\},$$

where g_i, h_j are differentiable functions defined on \mathbb{R}^n , or more general form

$$S := \{x \in \mathbb{R}^n : F(x) \in K\},$$

where $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a differentiable mapping, $K \subset \mathbb{R}^m$), we have the following result.

Lemma 1.2

Assume that Mangasarian-Fromovicz condition is satisfied at $\bar{x} \in S$. Then for every $d \in T_S(\bar{x})$,

$$T_S^2(\bar{x}, d) = \{w \in \mathbb{R}^n : \langle \nabla g_i(\bar{x}), w \rangle \leq -\langle d, \nabla^2 g_i(\bar{x})d \rangle, i \in I^{00}(\bar{x}, d), \\ \langle \nabla h_j(\bar{x}), w \rangle = -\langle d, \nabla^2 h_j(\bar{x})d \rangle, j = 1, \dots, q.\} \quad (1.7)$$

with $I^{00}(\bar{x}, d) = \{i \in I_0(\bar{x}) : \langle \nabla g_i(\bar{x}), d \rangle = 0\}$.

Similarly, if Robinson's condition is satisfied at $\bar{x} \in S$, then

$$T_S^2(\bar{x}, d) = DF(\bar{x})^{-1}[T_K^2(F(\bar{x}), DF(\bar{x})d) - D^2F(\bar{x})(d, d)]. \quad (1.8)$$

It follows by Lemma 1.2 that the inequality (1.6) becomes

$$\begin{aligned} \inf_w \quad & \nabla f(\bar{x})w + \langle d, \nabla^2 f(\bar{x})d \rangle \geq 0, \\ \text{s.t.} \quad & \langle \nabla g_i(\bar{x}), w \rangle \leq -\langle d, \nabla^2 g_i(\bar{x})d \rangle, i \in I^{00}(\bar{x}, d), \\ & \langle \nabla h_j(\bar{x}), w \rangle = -\langle d, \nabla^2 h_j(\bar{x})d \rangle, j = 1, \dots, q. \end{aligned}$$

or

$$\begin{aligned} \inf_w \quad & \nabla f(\bar{x})w + \langle d, \nabla^2 f(\bar{x})d \rangle \geq 0, \\ \text{s.t.} \quad & DF(\bar{x})w + D^2F(\bar{x})(d, d) \in T_K^2(F(\bar{x}), DF(\bar{x})d). \end{aligned}$$

By using duality of linear optimization problem, we can reformulate the second order necessary condition.

2. Preliminaries from variational analysis

2. Preliminaries from variational analysis

A great amount of functions involved in optimization problems are not differentiable. Maximization and minimization are often useful in constructing new functions and mappings from given ones, but, in contrast to addition and composition, they commonly fail to preserve smoothness. For example, the maximization of a finite many affine functions is convex, but not differentiable. However, there are directional differentiable.

Let $f: \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ be a function,

$$\text{dom} f = \{x \in \mathbb{R}^n \mid f(x) < \infty\}, \quad \text{epi} f = \{(x, r) \in \mathbb{R}^n \times \mathbb{R} \mid f(x) \leq r\}. \quad (2.1)$$

Assume that $f(\bar{x})$ is finite. The direction derivative of f at \bar{x} in the direction w is defined as

$$f'(\bar{x}, w) = \lim_{t \rightarrow 0} \frac{f(\bar{x} + tw) - f(\bar{x})}{t} \quad (2.2)$$

If f is convex, then the limit in (2.2) is

$$f'(\bar{x}, w) = \inf_{t > 0} \frac{f(\bar{x} + tw) - f(\bar{x})}{t}.$$

The directional derivative $f'(\bar{x}, w)$ in (2.2) depends only in the direction of w but not others. Instead, we can consider

$$f'(\bar{x}, w) = \lim_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(\bar{x} + tw') - f(\bar{x})}{t}. \quad (2.3)$$

If the limit exists in (2.3), we say that f is semidifferentiable (Hadamard directional differentiable) at \bar{x} in the direction w . In this case $f'(\bar{x}, w)$ is continuous and positively homogeneous in w .

Clearly, the existence of the limit in (2.3) if and only if

$$\liminf_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(\bar{x} + tw') - f(\bar{x})}{t} = \limsup_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(\bar{x} + tw') - f(\bar{x})}{t}.$$

When f is semidifferentiable at \bar{x} , its continuity there, forces it to be finite on a neighborhood of \bar{x} . So, $\bar{x} \in \text{int}(\text{dom}f)$. Therefore, semidifferentiability can't assist in the study of situation where, for instance \bar{x} is a boundary point of $\text{dom}f$.

Different choices of a mode of convergence for the difference quotient functions will lead to different kinds of derivatives. The situation can be avoid when the differentiability introduced by an approach through epi-convergence instead of continuous convergence.

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2.1 Epi-derivatives and epi-differentiability

Let us first recall the following notions of *upper* and *lower limits*, in the sense of Painlevé-Kuratowski, of a parameterized family A_t of subsets of \mathbb{R}^n , where t can be real valued or, more generally, an element of a metric space.

Definition 2.1

The following sets are called the upper (outer) and lower (inner) limits of a parameterized family A_t , of subsets of \mathbb{R}^n ,

$$\operatorname{Lim\,sup}_{t \rightarrow t_0} A_t := \{x \in X : \liminf_{t \rightarrow t_0} d(x, A_t) = 0\}$$

$$\operatorname{Lim\,inf}_{t \rightarrow t_0} A_t := \{x \in X : \limsup_{t \rightarrow t_0} d(x, A_t) = 0\}$$

respectively.

The upper and lower limit sets are both closed. These sets can be also described in terms of sequences as follows.

$$\limsup_{t \rightarrow t_0} A_t = \{x \mid \exists t_k \rightarrow t_0, \exists x_k \in A_{t_k}, x_k \rightarrow x\},$$

$$\liminf_{t \rightarrow t_0} A_t = \{x \mid \forall t_k \rightarrow t_0, \exists x_k \in A_{t_k}, x_k \rightarrow x\}.$$

If the equality in the above holds, we say A_t has a limit at t_0 .

Now let $\varphi_t: \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ be a family of extended real valued functions. The lower and upper *epi-limits* of φ_t , as $t \rightarrow t_0$, are defined as

$$\text{epi}(\text{e-}\liminf_{t \rightarrow t_0} \varphi_t(\cdot)) = \text{Lim sup}_{t \rightarrow t_0} \text{epi} \varphi_t, \quad (2.4)$$

$$\text{epi}(\text{e-}\limsup_{t \rightarrow t_0} \varphi_t(\cdot)) = \text{Lim inf}_{t \rightarrow t_0} \text{epi} \varphi_t. \quad (2.5)$$

Note that since the lower and upper set-limits are closed sets, the lower and upper epi-limit functions have closed epigraphs and hence are lower semicontinuous.

The *lower* and *upper directional epiderivatives* of $f : \mathbb{R}^n \rightarrow \mathbb{R}$, at a point $x \in \mathbb{R}^n$ such that $f(x)$ is finite, as follows

$$f_-^\downarrow(x, \cdot) := e\text{-}\liminf_{t \downarrow 0} \frac{f(x + t \cdot) - f(x)}{t}, \quad (2.6)$$

$$f_+^\downarrow(x, \cdot) := e\text{-}\limsup_{t \downarrow 0} \frac{f(x + t \cdot) - f(x)}{t}. \quad (2.7)$$

Equivalently,

$$f_-^\downarrow(x, w) = \liminf_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(x + tw') - f(x)}{t}, \quad (2.8)$$

$$f_+^\downarrow(x, w) = \sup_{\{t_k\} \in \Sigma_0} \left(\liminf_{\substack{k \rightarrow \infty \\ w' \rightarrow w}} \frac{f(x + t_k w') - f(x)}{t_k} \right), \quad (2.9)$$

where Σ_0 denotes the set of positive real sequences $\{t_k\}$ converging to zero.

Since epi-limit functions are lower semicontinuous, we have that $f_{-}^{\downarrow}(x, \cdot)$ and $f_{+}^{\downarrow}(x, \cdot)$ are l.s.c. positively homogeneous functions. We also have that

$$f_{-}^{\downarrow}(x, w) \leq f_{+}^{\downarrow}(x, w), \quad f_{-}^{\downarrow}(x, w) \leq f'_{-}(x, w), \quad f_{+}^{\downarrow}(x, w) \leq f'_{+}(x, w). \quad (2.10)$$

We say that f is directionally epidifferentiable at x , in a direction w , if $f_{-}^{\downarrow}(x, w) = f_{+}^{\downarrow}(x, w)$, and in that case we denote $f^{\downarrow}(x, w)$ the common value. Note that $f^{\downarrow}(x, w)$ can be different from $f'(x, w)$ even if f is convex.

If f is directionally differentiable, its second order directional derivative is defined as

$$f''(x; h, w) := \lim_{t \downarrow 0} \frac{f(x + th + \frac{1}{2}t^2w) - f(x) - tf'(x, h)}{\frac{1}{2}t^2}, \quad (2.11)$$

provided the above limit exists. We can also define

$$f''(x; h, w) := \lim_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(x + th + \frac{1}{2}t^2w') - f(x) - tf'(x, h)}{\frac{1}{2}t^2}. \quad (2.12)$$

Note that if f has the second-order Taylor expansion at x ,

$$f(x + h) = f(x) + \nabla f(x)h + \frac{1}{2}\nabla^2 f(x)(h, h) + o(\|h\|^2), \quad (2.13)$$

then

$$f''(x; h, w) = \nabla f(x)w + \nabla^2 f(x)(h, h). \quad (2.14)$$

Definition 2.2

Assuming that $f(x)$ and the respective directional epiderivatives $f_-^\downarrow(x, h)$ and $f_+^\downarrow(x, h)$ are finite, we call

$$f_-^{\downarrow\downarrow}(x; h, \cdot) := \mathbf{e}\text{-}\liminf_{t \downarrow 0} \frac{f(x + th + \frac{1}{2}t^2 \cdot) - f(x) - tf_-^\downarrow(x, h)}{\frac{1}{2}t^2}, \quad (2.15)$$

$$f_+^{\downarrow\downarrow}(x; h, \cdot) := \mathbf{e}\text{-}\limsup_{t \downarrow 0} \frac{f(x + th + \frac{1}{2}t^2 \cdot) - f(x) - tf_+^\downarrow(x, h)}{\frac{1}{2}t^2} \quad (2.16)$$

the lower and upper second order epiderivatives. The lower second order epiderivatives can be characterized pointwisely (see Ben-Tal and Zowe, 1982)

$$f_-^{\downarrow\downarrow}(x; h, w) := \liminf_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(x + th + \frac{1}{2}t^2 w') - f(x) - tf_-^\downarrow(x, h)}{\frac{1}{2}t^2}. \quad (2.17)$$

We say that f is parabolically epidifferentiable at x , in a direction h , if $f_-^{\downarrow\downarrow}(x; h, \cdot) = f_+^{\downarrow\downarrow}(x; h, \cdot)$.

f is parabolically epidifferentiable at x in a direction $h \Leftrightarrow \forall w \in \mathbb{R}^n$,
 $\forall t_k \downarrow 0, \exists w_k \rightarrow w$ s. t.

$$f_{-}^{\downarrow\downarrow}(x; h, w) = \lim_{k \rightarrow \infty} \frac{f(x + t_k h + \frac{1}{2} t_k^2 w_k) - f(x) - t_k f_{-}^{\downarrow}(x, h)}{\frac{1}{2} t_k^2}. \quad (2.18)$$

Note again that if $f(\cdot)$ is Lipschitz continuous and directionally differentiable at x , then for all $h, w \in X$ we have $f_{-}^{\downarrow\downarrow}(x; h, w) = f_{-}''(x; h, w)$ and $f_{+}^{\downarrow\downarrow}(x; h, w) = f_{+}''(x; h, w)$.

We can also consider another kind of epi-derivatives. Denote by

$$\Delta_t^2 f(x)(h) = \frac{f(x + th) - f(x) - tf_-^\downarrow(x, h)}{\frac{1}{2}t^2},$$

$$\Delta_t^2 f(x, v)(h) = \frac{f(x + th) - f(x) - t\langle v, h \rangle}{\frac{1}{2}t^2}, \quad \text{for } v \in \mathbb{R}^n.$$

Definition 2.3

The second subderivative of f at x is defined by

$$d^2 f(x)(\cdot) := e - \liminf_{t \downarrow 0} \Delta_t^2 f(x)(\cdot), \quad (2.19)$$

the second subderivative of f at x for v is defined by

$$d^2 f(x, v)(\cdot) = e - \liminf_{t \downarrow 0} \Delta_t^2 f(x, v)(\cdot). \quad (2.20)$$

If the second order difference quotient function $h \rightarrow \Delta_t^2 f(x)(h)$ (resp. $h \rightarrow \Delta_t^2 f(x, v)(h)$), epi-converges to some function as $t \downarrow 0$, we say that f is twice epi-differentiable at x (resp. for v); it is properly twice epi-differentiable at x (for v) if $d^2 f(x)(\cdot)$ (resp. $d^2 f(x, v)(\cdot)$) is proper.

The second subderivative can be equivalently written as

$$d^2 f(x)(h) = \liminf_{\substack{t \downarrow 0 \\ h' \rightarrow h}} \frac{f(x + th') - f(x) - tf_-^\downarrow(x, h')}{\frac{1}{2}t^2}, \quad (2.21)$$

and

$$d^2 f(x, v)(h) = \liminf_{\substack{t \downarrow 0 \\ h' \rightarrow h}} \frac{f(x + th') - f(x) - t\langle v, h' \rangle}{\frac{1}{2}t^2}. \quad (2.22)$$

f is twice epi-differentiable at x (for v) $\Leftrightarrow \forall t_k \downarrow 0, \forall h \in \mathbb{R}^n, \exists h_k \rightarrow h$ s. t.

$$d^2 f(x)(h) = \lim_{k \rightarrow \infty} \frac{f(x + t_k h_k) - f(x) - t_k f_-^\downarrow(x, h_k)}{\frac{1}{2} t_k^2}, \quad (2.23)$$

and

$$d^2 f(x, v)(h) = \lim_{k \rightarrow \infty} \frac{f(x + t_k h_k) - f(x) - t_k \langle v, h_k \rangle}{\frac{1}{2} t_k^2}. \quad (2.24)$$

If f is twice epi-differentiable at x relative to v , then the second-order epi-derivative function $d^2 f(x, v)(\cdot)$ is lower semicontinuous and positively homogeneous of degree 2.

$f_-^\downarrow(x, \cdot)$ is also called the subderivative of f at x and write as $df(x)$.

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and

$$d^2 f(x, v)(h) = \lim_{k \rightarrow \infty} \frac{f(x + t_k h_k) - f(x) - t_k \langle v, h_k \rangle}{\frac{1}{2} t_k^2}. \quad (2.24)$$

If f is twice epi-differentiable at x relative to v , then the second-order epi-derivative function $d^2 f(x, v)(\cdot)$ is lower semicontinuous and positively homogeneous of degree 2.

$f_-^\downarrow(x, \cdot)$ is also called the subderivative of f at x and write as $df(x)$.

- 1 Classical optimality conditions
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2.2 Tangent and normal cones

Definition 2.4

For $S \subset \mathbb{R}^n$ and a point $\bar{x} \in S$
the contingent (Bouligand) cone

$$T_S(\bar{x}) := \limsup_{t \downarrow 0} \frac{S - \bar{x}}{t} = \{h \in \mathbb{R}^n : \exists t_k \downarrow 0, d(\bar{x} + t_k h, S) = o(t_k)\}, \quad (2.25)$$

the inner tangent cone

$$T_S^i(\bar{x}) := \liminf_{t \downarrow 0} \frac{S - \bar{x}}{t} = \{h \in \mathbb{R}^n : d(\bar{x} + th, S) = o(t), t \geq 0\}, \quad (2.26)$$

Clarke tangent cone

$$T_S^c(\bar{x}) := \liminf_{\substack{x' \xrightarrow{S} \bar{x} \\ t \downarrow 0}} \frac{S - x'}{t} \quad (2.27)$$

It is clear that if $\bar{x} \in S$, then $0 \in T_S^c(\bar{x}) \subset T_S^i(\bar{x}) \subset T_S(\bar{x})$. In general, these cones can be different, and the Clarke tangent cone are convex, but the inner tangent cones and the contingent cone can be nonconvex.

If $T_S(\bar{x}) = T_S^i(\bar{x})$, we say that S is geometrically derivable at \bar{x}
 $\Leftrightarrow \forall w \in T_S(\bar{x}), \exists \epsilon > 0$ and an arc $\xi: [0, \epsilon] \rightarrow S$ such that $\xi(0) = \bar{x}$
and $\xi'_+(0) = w$.

For convex sets, however, the contingent, inner and Clarke tangent cones are equal to each other, so must be geometrically derivable.

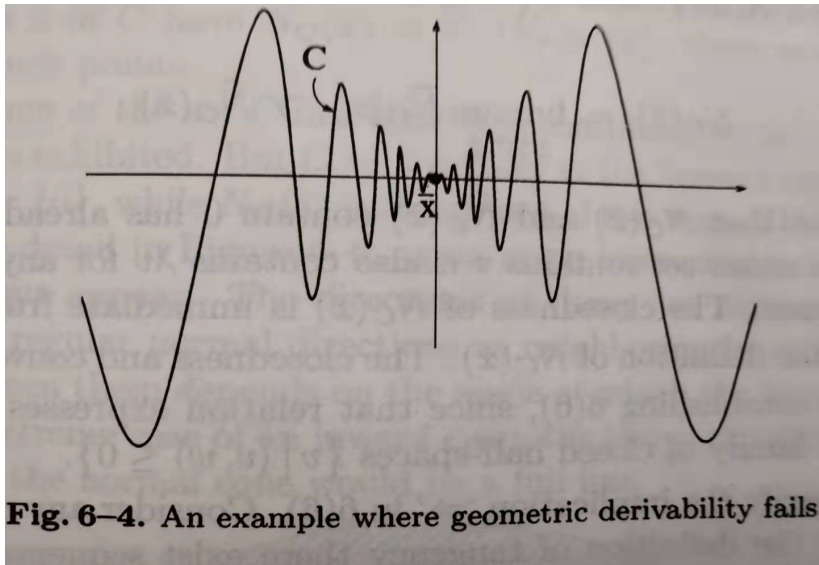


Fig. 6–4. An example where geometric derivability fails.

Definition 2.5

$$T_S^{i,2}(\bar{x}, h) := \liminf_{t \downarrow 0} \frac{S - \bar{x} - th}{\frac{1}{2}t^2}, \quad (2.28)$$

$$T_S^2(\bar{x}, h) := \limsup_{t \downarrow 0} \frac{S - \bar{x} - th}{\frac{1}{2}t^2}. \quad (2.29)$$

are called the inner and outer second order tangent sets, respectively, to the set S at the point \bar{x} and in the direction h .

Alternatively these tangent sets can be written in the form

$$T_S^{i,2}(\bar{x}, h) = \{w \in \mathbb{R}^n : d(\bar{x} + th + \frac{1}{2}t^2w, S) = o(t^2), t \geq 0\}, \quad (2.30)$$

$$T_S^2(\bar{x}, h) = \{w \in \mathbb{R}^n : \exists t_k \downarrow 0 \text{ s.t. } d(\bar{x} + t_k h + \frac{1}{2}t_k^2w, S) = o(t_k^2)\}. \quad (2.31)$$

Clearly, $T_S^{i,2}(\bar{x}, h) \subset T_S^2(\bar{x}, h)$.

If $T_S^{i,2}(\bar{x}, h) = T_S^2(\bar{x}, h)$ for all h , we say that S be parabolically derivable at \bar{x} for h . \Leftrightarrow if $\forall w \in T_S^2(\bar{x}, h)$, $\exists \epsilon > 0$ and an arc $\xi: [0, \epsilon] \rightarrow S$ such that $\xi(0) = \bar{x}$ and $\xi'_+(0) = h$, $\xi''_+(0) = w$.

If $T_S^{i,2}(x, h) \neq \emptyset$ (resp, $T_S^2(x, h) \neq \emptyset$) only if $h \in T_S^i(x)$ (resp. $h \in T_S(x)$). Let $S = \{(x_1, x_2) \in \mathbb{R}^2 | x_2 \geq |x_1|^{\frac{3}{2}}\}$. Then S is a closed and convex. Let $\bar{x} = (0, 0), h = (1, 0)$. Then $h \in T_S(\bar{x}) = \{(x_1, x_2) | x_2 \geq 0\}$ and

$$\lim_{t \downarrow 0} \frac{d(\bar{x} + th, S)}{t^2} = +\infty,$$

and hence both $T_S^{i,2}(x, h)$ and $T_S^2(x, h)$ are empty.

Consider the convex piecewise linear function $y = \eta(x)$ with $\eta(x) = \eta(-x)$, $\eta(0) = 0$, oscillating between two parabolas $y = 3x^2$ and $y = 3.5x^2$. That is, we construct $\eta(x)$ in such a way that and for some sequence x_k monotonically decreasing to zero, the function $\eta(x)$ is linear on every interval $[x_{k+1}, x_k]$, $\eta(x_k) = 3x_k^2$ and the straight line passing through the points $(x_k, \eta(x_k))$ and $(x_{k+1}, \eta(x_{k+1}))$ is tangent to the curve $y = 3.5x^2$. It is quite clear how such a function can be constructed. define $g(x_1, x_2) := \eta(x_1) - x_2$, and it is not parabolically epidifferentiable at $\bar{x} = (0, 0)$ in the direction $h := (1, 0)$. It is easy to show that $g(\bar{x}) = 0$, $g^\downarrow(\bar{x}, h) = 0$. Let $S = \text{epi}\eta$. Consequently, the corresponding second order inner and outer tangent sets are different. And indeed, it is not difficult to verify that

$$T_S^{i,2}(0, h) = \{w | g_+^{\downarrow}(\bar{x}; h, w) \leq 0\} = \{(x_1, x_2) | x_2 \geq 7\},$$

$$T_S^2(0, h) = \{w | g_-^{\downarrow}(\bar{x}; h, w) \leq 0\} \{(x_1, x_2) | x_2 \geq 6\}.$$

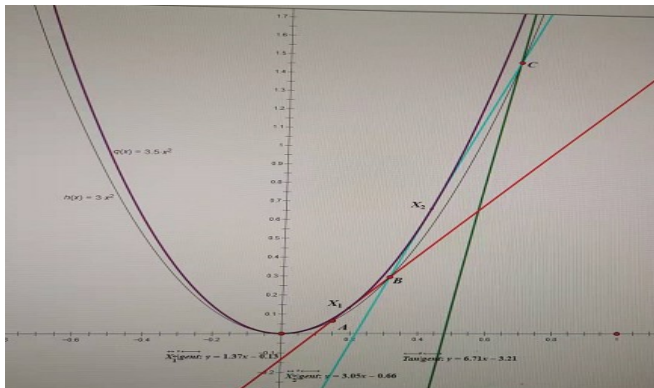


Figure: picture

If C is convex, then $T_C^{i,2}(x, h)$ is convex, but $T_C^2(x, h)$ could be nonconvex. If, in addition, it is parabolically derivable at \bar{x} for h , then the second order tangent set $T_C^2(x, h)$ is a convex set in \mathbb{R}^n .

For a convex set C

$$T_C^{i,2}(x, h) + T_{T_C(x)}(h) \subset T_C^{i,2}(x, h) \subset T_{T_C(x)}(h), \quad (2.32)$$

$$T_C^2(x, h) + T_{T_C(x)}(h) \subset T_C^2(x, h) \subset T_{T_C(x)}(h). \quad (2.33)$$

It follows that if $0 \in T_C^2(x, h)$, then $T_C^2(x, h) = T_{T_C(x)}(h)$. Moreover, if $0 \in T_C^{i,2}(x, h)$, i.e. $d(x + th, C) = o(t^2)$, all three sets coincide, that is

$$T_C^{i,2}(x, h) = T_C^2(x, h) = T_{T_C(x)}(h).$$

Given the function $f: \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$, $f(\bar{x})$ is finite. The regular subdifferential of f at $\bar{x} \in \text{dom} f$ is defined by

$$\hat{\partial}f(\bar{x}) = \{v \in \mathbb{R}^n \mid \liminf_{x \rightarrow \bar{x}} \frac{f(x) - f(\bar{x}) - \langle v, x - \bar{x} \rangle}{\|x - \bar{x}\|} \geq 0\} \quad (2.34)$$

$$v \in \hat{\partial}f(\bar{x}) \Leftrightarrow \forall \sigma > 0 \exists \delta > 0 \text{ s. t.}$$

$$\langle v, x - \bar{x} \rangle \leq f(x) - f(\bar{x}) + \sigma \|x - \bar{x}\|, \quad \forall x \in B(\bar{x}, \delta). \quad (2.35)$$

The dual representation for the regular subgradients

$$\hat{\partial}f(\bar{x}) = \{v \in \mathbb{R}^n \mid \langle v, w \rangle \leq f_-^\downarrow(\bar{x}, w), \quad \forall w \in \mathbb{R}^n\}. \quad (2.36)$$

The subdifferential of f at \bar{x} is given by

$$\partial f(\bar{x}) = \{v \in \mathbb{R}^n \mid \exists x_k \xrightarrow{f} \bar{x}, v_k \rightarrow v \text{ with } v_k \in \hat{\partial}f(x_k)\}, \quad (2.37)$$

where $x_k \xrightarrow{f} \bar{x}$ stands for $x_k \rightarrow \bar{x}$ and $f(x_k) \rightarrow f(\bar{x})$.

The proximal subdifferential of f at \bar{x} is given by

$$\begin{aligned}\partial^p f(\bar{x}) &= \{v \in \mathbb{R}^n \mid \exists \sigma > 0, \delta > 0 \text{ s.t.} \\ &\quad f(x) \geq f(\bar{x}) + \langle v, x - \bar{x} \rangle - \frac{\sigma}{2} \|x - \bar{x}\|^2, \forall \|x - \bar{x}\| < \delta\}\end{aligned}$$

It is well-known that the inclusions $\partial^p f(\bar{x}) \subset \hat{\partial} f(\bar{x}) \subset \partial f(\bar{x})$ always hold and

$$\partial f(\bar{x}) = \limsup_{x \xrightarrow{f} \bar{x}} \partial^p f(x).$$

Given a nonempty set $S \subset \mathbb{R}^n$, the proximal and regular normal cones to S at $\bar{x} \in S$ are defined, respectively, by

$$N_S^p(\bar{x}) = \partial^p \delta_S(\bar{x}), \quad (2.38)$$

$$\hat{N}_S(\bar{x}) = \hat{\partial} \delta_S(\bar{x}). \quad (2.39)$$

Similarly, we define the (limiting/Mordukhovich) normal cone of S at \bar{x} by $N_S(\bar{x}) := \partial \delta_S(\bar{x})$.

$$\begin{aligned} N_S^p(\bar{x}) &= \{v \in H \mid \exists t > 0 \text{ s.t. } d(\bar{x} + tv, S) = t\|v\|\} \\ &= \{v \in H \mid \exists \sigma, \delta > 0 \text{ s.t. } \langle v, x - \bar{x} \rangle \leq \sigma \|x - \bar{x}\|^2 \forall x \in B(\bar{x}, \delta)\} \end{aligned}$$

The subdifferential can be defined by the corresponding normal cone. For instance

$$\partial^p f(\bar{x}) = \{v \in H \mid (v, -1) \in N_{\text{epi}f}^p(\bar{x}, f(\bar{x}))\}$$

If f is differentiable at \bar{x} , then $\partial f(\bar{x}) = \{\nabla f(\bar{x})\}$ and if f is second order continuously differentiable or convex and L -Lipschitz smooth, i.e., $\nabla f(x)$ L -Lipschitz, then $\partial^p f(\bar{x}) = \{\nabla f(\bar{x})\}$. Indeed, since f is convex and L -Lipschitz smooth, we have

$$f(x) \geq f(\bar{x}) + \langle \nabla f(\bar{x}), x - \bar{x} \rangle + \frac{1}{2L} \|\nabla f(x) - \nabla f(\bar{x})\|^2,$$

and this in turn implies $\nabla f(\bar{x}) \in \partial^p f(\bar{x})$.

For a nonconvex set C , there can be regular normals that aren't proximal normals, even when C is defined by smooth inequalities. This is illustrated by

$$C = \{x = (x_1, x_2) \in \mathbb{R}^2 \mid x_2 \geq x_1^{3/5}, x_2 \geq 0\},$$

where the vector $v = (1, 0)$ is a regular normal vector at $\bar{x} = (0, 0)$ but no point of $\{\bar{x} + \tau v \mid \tau > 0\}$ projects onto \bar{x} , cf. Figure 6-12(a).

The proximal normals at \bar{x} always form a cone, and this cone is convex; these facts are evident from the description of proximal normals just provided. But in contrast to the cone of regular normals the cone of proximal normals needn't be closed—as Figure 6-12(a) likewise makes clear. Nor is it true that the closure of the cone of proximal normals always equals the cone of regular normals, as seen from the similar example in Figure 6-12(b), where only the zero vector is a proximal normal at \bar{x} .

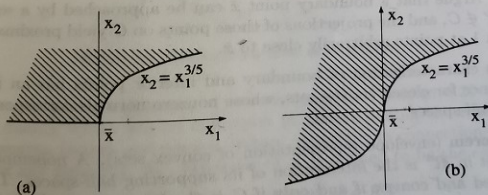


Fig. 6-12. Regular normals versus proximal normals.

For example, $f(x) = -x^{\frac{5}{3}}$, $0 \in \partial^F f(0)$, but $0 \notin \partial^p f(0)$.

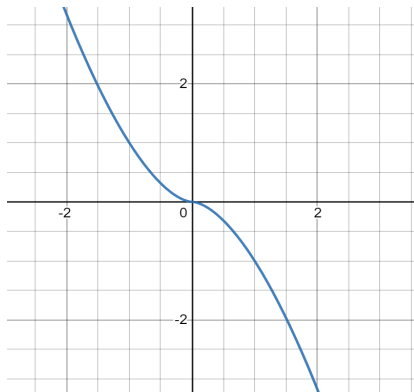


Figure: picture

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2.3 Preliminary properties

Proposition 2.1

Let $f: \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ be an extended real valued function and let $x \in X$ be a point such that $f(x)$ is finite. Then

$$T_{\text{epi}f}(x, f(x)) = \text{epi}f_{-}^{\downarrow}(x, \cdot) \quad (2.40)$$

$$T_{\text{epi}f}^i(x, f(x)) = \text{epi}f_{+}^{\downarrow}(x, \cdot). \quad (2.41)$$

If f is convex, then $T_{\text{epi}f}(x, f(x)) = \text{cl } \text{epi}f'(x, \cdot)$ and then $f_{-}^{\downarrow}(x, \cdot) = \text{lsc}f'(x, \cdot)$.

Proof. By the definition, we have that

$$\begin{aligned} \text{epi}f_{-}^{\downarrow}(x, \cdot) &= \text{Lim sup}_{t \downarrow 0} \text{epi} \left\{ \frac{f(x + t \cdot) - f(x)}{t} \right\} \\ &= \text{Lim sup}_{t \downarrow 0} \frac{\text{epi}f - (\bar{x}, f(\bar{x}))}{t}. \end{aligned}$$

Together with the definition of the contingent cones, this implies the first equation. The second equation can be proved similarly. \square

Proposition 2.2

Let $f : \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ be an extended real valued function taking a finite value at a point $x \in X$. Then

$$T_{\text{epi}f}^{i,2}[(x, f(x)), (h, f_+^\downarrow(x, h))] = \text{epi}f_+^{\downarrow\downarrow}(x; h, \cdot), \quad (2.42)$$

$$T_{\text{epi}f}^2[(x, f(x)), (h, f_-^\downarrow(x, h))] = \text{epi}f_-^{\downarrow\downarrow}(x; h, \cdot), \quad (2.43)$$

provided the respective values $f_-^\downarrow(x, h)$ and $f_+^\downarrow(x, h)$ are finite.

$\text{epi}f$ is geometrically derivable at $(x, f(x))$ iff f is directionally epi-differentiable at \bar{x} in h and $\text{epi}f$ is parabolically derivable at $(x, f(x))$ for $(h, f_-^\downarrow(x, h))$ iff f is parabolically epi-differentiable at \bar{x} in h .

We say a function $f : \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ is called Lipschitz continuous around \bar{x} relative to $C \subset \text{dom} f$ with constant $\ell > 0$ if $\bar{x} \in C$ and there exists a neighborhood U of \bar{x} such that

$$\|f(x_1) - f(x_2)\| \leq \ell \|x_1 - x_2\| \quad \forall x_1, x_2 \in U \cap C. \quad (2.44)$$

Such a function is called locally Lipschitz continuous relative to C if it is Lipschitz continuous around \bar{x} relative to C for all $\bar{x} \in C$. Piecewise linear-quadratic functions (not necessarily convex) and an indicator function of a nonempty set are important examples of functions that are locally Lipschitz continuous relative to their domains.

Proposition 2.3

Suppose that $f: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is Lipschitz continuous around \bar{x} relative to its domain. Then

(i) $\text{dom} f_{-}^{\downarrow}(\bar{x}, \cdot) = T_{\text{dom} f}(\bar{x})$. In particular, for every $h \in T_{\text{dom} f}(\bar{x})$, $f_{-}^{\downarrow}(\bar{x}, h)$ is finite.

(ii) If, in addition, f is parabolically epi-differentiable at \bar{x} for h , then $\text{dom} f$ is parabolically derivable at \bar{x} for h and

$$T_{\text{dom} f}^2(\bar{x}, h) = \text{dom} f_{-}^{\downarrow\downarrow}(x, h, \cdot). \quad (2.45)$$

(i) We always have $\text{dom} f_-^\downarrow(\bar{x}, \cdot) \subset T_{\text{dom} f}(\bar{x})$. Indeed, for every $h \in \text{dom} f_-^\downarrow(\bar{x}, \cdot)$, $f_-^\downarrow(\bar{x}, h) < \infty$ and there exist $t_k \downarrow 0, h_k \rightarrow h$ such that

$$f_-^\downarrow(\bar{x}, h) = \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h_k) - f(\bar{x})}{t_k}.$$

Without loss of generality, we may assume that $\bar{x} + t_k h_k \in \text{dom} f$ and this implies that $h \in T_{\text{dom} f}(\bar{x})$.

Let $h \in T_{\text{dom} f}(\bar{x})$. Then there exist $t_k \downarrow 0$ and $h_k \rightarrow h$ such that $\bar{x} + t_k h_k \in \text{dom} f$. It follows by the Lipschitzian property of f relative to $\text{dom} f$ that

$$\left| \frac{f(\bar{x} + t_k h_k) - f(\bar{x})}{t_k} \right| \leq \ell \|h_k\|. \quad (2.46)$$

Hence $f_-^\downarrow(\bar{x}, h)$ is finite.

(ii)

$$\text{dom} f_{-}^{\downarrow\downarrow}(x, h, \cdot) \subset T_{\text{dom} f}^2(\bar{x}, h)$$

always hold. To prove the converse, take $w \in T_{\text{dom} f}^2(\bar{x}, h)$. Then there exist $t_k \downarrow 0$, $w_k \rightarrow w$ such that $\bar{x} + t_k h + \frac{1}{2} t_k^2 w_k \in \text{dom} f$. Since f is parabolically epi-differentiable at \bar{x} for h , there exists $w' \in \text{dom} f_{-}^{\downarrow\downarrow}(x, h, \cdot)$ such that $f_{-}^{\downarrow\downarrow}(x; h, w') < \infty$. Hence, corresponding to t_k , there exist $w'_k \rightarrow w'$ such that

$$f_{-}^{\downarrow\downarrow}(x; h, w') = \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h + \frac{1}{2} t_k^2 w'_k) - f(\bar{x}) - t_k f_{-}^{\downarrow}(x, h)}{\frac{1}{2} t_k^2} < \infty. \quad (2.47)$$

Without loss of generality, we assume that $\bar{x} + t_k h_k + \frac{1}{2} t_k^2 w'_k \in \text{dom} f$. Using these together with the Lipschitz property of f , we have

$$\begin{aligned} & \frac{f(\bar{x} + t_k h + \frac{1}{2} t_k^2 w'_k) - f(\bar{x}) - t_k f_{-}^{\downarrow}(x, h)}{\frac{1}{2} t_k^2} \\ \leq & \frac{f(\bar{x} + t_k h + \frac{1}{2} t_k^2 w'_k) - f(\bar{x}) - t_k f_{-}^{\downarrow}(x, h)}{\frac{1}{2} t_k^2} + \ell \|w_k - w'_k\|. \end{aligned}$$

passing to the limit in the inequality, we get

$$f_{-}^{\downarrow\downarrow}(x; h, w) \leq f_{-}^{\downarrow\downarrow}(x; h, w') + \ell \|w - w'\|,$$

which in turn implies that $w \in \text{dom} f_{-}^{\downarrow\downarrow}(x, h, \cdot)$. As a direct result of Proposition 2.2 that $\text{dom} f$ is parabolically derivable at \bar{x} for h .

Example 2.1

Let S be a nonempty closed convex subset of \mathbb{R}^n , $\delta_S(\cdot)$ is a proper l.s.c. convex function. consider the set $K := \text{epi}\delta_S = S \times \mathbb{R}_+$ and $x \in S$. Then $\delta^\downarrow(x, \cdot) = \delta_{T_S(x)}(\cdot)$. Given a vector $h \in T_S(x)$. It is not difficult to see that

$$\delta_+^{\downarrow\downarrow}(x; h, w) = \begin{cases} 0, & \text{if } w \in T_S^{i,2}(x, h), \\ +\infty, & \text{otherwise.} \end{cases} \quad (2.48)$$

$$\delta_-^{\downarrow\downarrow}(x; h, w) = \begin{cases} 0, & \text{if } w \in T_S^2(x, h), \\ +\infty, & \text{otherwise.} \end{cases} \quad (2.49)$$

Moreover, we have

$$T_K^{i,2}((x, 0), (h, \gamma)) = \begin{cases} T_S^{i,2}(x, h) \times \mathbb{R}, & \text{if } \gamma > 0, \\ T_S^{i,2}(x, h) \times \mathbb{R}_+, & \text{if } \gamma = 0, \\ \emptyset, & \text{if } \gamma < 0, \end{cases} \quad (2.50)$$

and

$$T_K^2((x, 0), (h, \gamma)) = \begin{cases} T_S^2(x, h) \times \mathbb{R}, & \text{if } \gamma > 0, \\ T_S^2(x, h) \times \mathbb{R}_+, & \text{if } \gamma = 0, \\ \emptyset, & \text{if } \gamma < 0, \end{cases} \quad (2.51)$$

Therefore the following conditions are equivalent:

- (i) the set $K := \text{epi}\delta_S$ is parabolically derivable at $(x, 0)$;
- (ii) the set S is parabolically derivable at x ;
- (iii) the function δ_S is parabolically epidifferentiable at \bar{x} .

Proposition 2.4

Let \bar{x} be such that $f(\bar{x})$ is finite, for a given $h \in \mathbb{R}^n$, let $v \in \mathbb{R}^n$ be such that $\langle v, h \rangle = f_-^\downarrow(\bar{x}, h)$. Then

$$\inf_{w \in X} \{f_-^{\downarrow\downarrow}(\bar{x}; h, w) - \langle v, w \rangle\} \geq d^2 f(\bar{x}, v)(h). \quad (2.52)$$

Proof. Let $w_k \rightarrow w$, $t_k \downarrow 0$ be such that

$$f_{-}^{\downarrow\downarrow}(x; h, w) = \lim_{k \rightarrow \infty} \frac{f(x + t_k h + \frac{1}{2} t_k^2 w_k) - f(x) - t_k f_{-}^{\downarrow}(x, h)}{\frac{1}{2} t_k^2} \quad (2.53)$$

and take $h'_k := h + \frac{1}{2} t_k w_k$ in the definition of $d^2 f(x, v)(h)$. We obtain then that

$$\begin{aligned} d^2 f(x, v)(h) &\leq \liminf_{k \rightarrow \infty} \frac{f(x + t_k h_k) - f(x) - t_k \langle v, h_k \rangle}{\frac{1}{2} t_k^2} \\ &= \liminf_{k \rightarrow \infty} \frac{f(x + t_k h + \frac{1}{2} t_k^2 w_k) - f(x) - t_k f_{-}^{\downarrow}(x, h) - \frac{1}{2} t_k^2 \langle v, w_k \rangle}{\frac{1}{2} t_k^2}. \end{aligned}$$

It follows that for any $w \in \mathbb{R}^n$,

$$d^2 f(x, v)(h) \leq f_{-}^{\downarrow\downarrow}(x; h, w) - \langle v, w \rangle.$$

By taking the infimum of the right hand side of the above inequality over all $w \in \mathbb{R}^n$, we obtain the desired result. \square

Proposition 2.5 (properties of second subderivative)

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ taking finite value at $\bar{x}, \bar{v} \in \mathbb{R}^n$. Then the following conditions hold:

(i) if $d^2 f(\bar{x}, \bar{v})$ is a proper function, then we always have

$$\text{dom} d^2 f(\bar{x}, \bar{v}) \subset \{h \in \mathbb{R}^n \mid f_{-}^{\downarrow}(\bar{x}, h) = \langle \bar{v}, h \rangle\}, \quad (2.54)$$

Moreover, the equality holds if, in addition, $\text{dom} f_{-}^{\downarrow}(x; h, \cdot) \neq \emptyset$;

(ii) if $\bar{v} \in \hat{\partial} f(\bar{x})$, then for any $w \in \mathbb{R}^n$ we have

$$d^2 f(\bar{x}, \bar{v})(h) \geq -\sigma \|h\|.$$

In particular, $d^2 f(\bar{x}, \bar{v})$ is a proper function.

Proof. Note that

$$\begin{aligned} d^2 f(\bar{x}, \bar{v})(w) &= \liminf_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{f(\bar{x} + tw') - f(\bar{x}) - t\langle \bar{v}, w' \rangle}{\frac{1}{2}t^2} \\ &= \liminf_{\substack{t \downarrow 0 \\ w' \rightarrow w}} \frac{\frac{f(\bar{x} + tw') - f(\bar{x})}{t} - \langle \bar{v}, w' \rangle}{\frac{1}{2}t} \end{aligned}$$

It is easily to see that (i) holds by the definition and (2.52).

By the definition of regular subdifferential (2.35), as $t \downarrow 0$, $w' \rightarrow w$, we have $\|\bar{x} + tw' - \bar{x}\| < \delta$, the assertion (ii) follows. \square

Proposition 2.2 implies that $f_{-}^{\downarrow\downarrow}(x; h, \cdot)$ and $f_{+}^{\downarrow\downarrow}(x; h, \cdot)$ are lower semicontinuous functions. If, in addition, a convex function f is parabolically epi-differentiable at \bar{x} for h , then $f_{-}^{\downarrow\downarrow}(x; h, \cdot) = f_{+}^{\downarrow\downarrow}(x; h, \cdot)$ is lower semicontinuous and convex. It follows from Proposition 2.4 and Proposition 2.5 that $d^2f(x, \cdot)$ is also proper whenever $\text{dom}f_{-}^{\downarrow\downarrow}(x; h, \cdot) \neq \emptyset$ and there exists $\bar{v} \in \hat{\partial}f(\bar{x})$ such that $\langle \bar{v}, h \rangle = f_{-}^{\downarrow}(\bar{x}, h)$.

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3.1 Parabolic regularity

Let $f : \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ taking finite value at $\bar{x} \in \mathbb{R}^n$ and pick $\bar{v} \in \mathbb{R}^n$. The critical cone of f at (\bar{x}, \bar{v}) is defined by

$$K_f(\bar{x}, \bar{v}) := \{h \in \mathbb{R}^n \mid df(\bar{x})(h) = \langle \bar{v}, h \rangle\}. \quad (3.1)$$

Definition 3.1

It is said that the function f is parabolically regular at \bar{x} for \bar{v} in $h \in \mathbb{R}^n$ if

$$\inf_{w \in \mathbb{R}^n} \{f_{-}^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle\} = d^2 f(\bar{x}, \bar{v})(h). \quad (3.2)$$

If (3.2) is satisfied for every $h \in \mathbb{R}^n$, f is called parabolically regular at \bar{x} for \bar{v} .

The parabolic regularity of f at \bar{x} can be equivalent characterized as following.

Proposition 3.1

f is parabolically regular at \bar{x} for $\bar{v} \in \hat{\partial}f(\bar{x})$ if for every $h \in \text{dom}d^2f(\bar{x}, \bar{v})$, there exist, among the sequences $t_k \downarrow 0$ and $h_k \rightarrow h$ with $\Delta_{t_k}^2 f(x, v)(h_k) \rightarrow d^2f(\bar{x}, \bar{v})(h)$, ones with additional property that

$$\limsup_{k \rightarrow \infty} \frac{\|h_k - h\|}{t_k} < \infty. \quad (3.3)$$

Moreover, for every $h \in \text{dom}d^2f(\bar{x}, \bar{v})$, there exists $w \in \mathbb{R}^n$ such that

$$d^2f(\bar{x}, \bar{v})(h) = f_{-}^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle.$$

Proof. Since $\bar{v} \in \hat{\partial}f(\bar{x})$, we know that

$$d^2f(\bar{x}, \bar{v})(h) \geq -\sigma\|h\| > -\infty.$$

If $h \in K_f(\bar{x}, \bar{v}) \setminus \text{dom } d^2f(\bar{x}, \bar{v})$, then, $d^2f(\bar{x}, \bar{v})(h) = +\infty$, by (2.52), we have that (3.2) holds.

Assume that $d^2f(\bar{x}, v)(h)$ is finite. Let $h_k \rightarrow h$ and $t_k \rightarrow 0$ be sequences such that $\Delta_{t_k}^2 f(x, v)(h_k) \rightarrow d^2f(\bar{x}, \bar{v})(h)$ with $\limsup_{k \rightarrow \infty} t_k^{-1}\|h_k - h\| < \infty$. Consider $w_k := (\frac{1}{2}t_k)^{-1}(h_k - h)$, i.e., $h_k = h + \frac{1}{2}t_k w_k$, and $x_k = \bar{x} + t_k h + \frac{1}{2}t_k^2 w_k$. Then

$$\begin{aligned} d^2f(\bar{x}, \bar{v})(h) &= \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h_k) - f(\bar{x}) - t_k \langle \bar{v}, h_k \rangle}{\frac{1}{2}t_k^2} \\ &= \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h + \frac{1}{2}t_k^2 w_k) - f(\bar{x}) - t_k \langle \bar{v}, h \rangle}{\frac{1}{2}t_k^2} - \langle \bar{v}, w_k \rangle. \end{aligned}$$

Since $\{w_k\}$ is bounded, without loss of generality, we may assume that $w_k \rightarrow w$. Hence

$$d^2 f(\bar{x}, \bar{v})(h) \geq f_-^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle.$$

Combining Proposition 2.4, we get

$$d^2 f(\bar{x}, \bar{v})(h) = \inf_w \{f_-^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle\}.$$

The left side of (3.2) is identical to the lowest limit attainable for

$$\lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h + \frac{1}{2} t_k^2 w_k) - f(\bar{x}) - t_k \langle \bar{v}, h \rangle}{\frac{1}{2} t_k^2} - \langle \bar{v}, w_k \rangle.$$

relative to $t_k \downarrow 0$ and a bounded sequence of vectors w_k (as seen from the cluster points w of such a sequence). In terms of $h_k = h + \frac{1}{2} t_k w_k$, which corresponds to $\{\frac{h_k - h}{t_k}\}$ is bounded. \square

Definition 3.2

K is outer second order regular at a point $y \in K$ in direction $d \in T_K(y)$, if for any sequence $y_k \in K$ of the form $y_k = y + t_k d + \frac{1}{2} t_k^2 w_k$, satisfying $t_k \downarrow 0$, $t_k w_k \rightarrow 0$, the following condition holds:

$$\lim_{k \rightarrow \infty} d(w_k, T_K^2(y, d)) = 0. \quad (3.4)$$

We say that K is second order regular at y if K is parabolically derivable and outer second order regular at y in all direction $d \in T_K(y)$.

For example, the union of a finite polyhedral sets, the cone of symmetric positive semidefinite matrices, the second order cone are second order regular.

Definition 3.3

We say that a function $f: X \rightarrow \bar{\mathbb{R}}$ is (outer) second order regular at \bar{x} in the direction h if $df(\bar{x})(h)$ is finite and the set $K = \text{epi} f$ is (outer) second order at the point $(\bar{x}, f(\bar{x}))$ in the direction $(h, df(\bar{x})(h))$.

For example, the leading eigenvalue functions of symmetric matrix, a function whose epigraph is the union of a finite polyhedral sets and the indicate function of a second order cone are the second order regular function.

Proposition 3.2

Let $f: X \rightarrow \bar{\mathbb{R}}$, $\bar{v} \in \hat{\partial}f(\bar{x}, \cdot)$, and $h \in K_f(\bar{x}, \bar{v})$. Then f is parabolically regular at x for \bar{v} in the direction h , if the function f is outer second order regular at \bar{x} in the direction h .

Proof. If $d^2f(\bar{x}, v)(h) = +\infty$, then, by (2.52), we have that (3.2) holds. Therefore, we can assume that $d^2f(\bar{x}, v)(h)$ is finite. Note that because of $\langle v, h \rangle = df(\bar{x})(h)$, we have that $df(\bar{x})(h)$ is finite.

Suppose that f is outer second order regular at \bar{x} in the direction h . Let $h_k \rightarrow h$ and $t_k \rightarrow 0$ be sequences at which the limit in the definition of $d^2f(\bar{x}, v)(h)$ is attained. Consider $w_k := (\frac{1}{2}t_k)^{-1}(h_k - h)$, i.e., $h_k = h + \frac{1}{2}t_k w_k$, and $x_k = \bar{x} + t_k h + \frac{1}{2}t_k^2 w_k$. Then $t_k w_k \rightarrow 0$ and

$$\begin{aligned} d^2f(\bar{x}, v)(h) &= \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h_k) - f(\bar{x}) - t_k \langle v, h_k \rangle}{\frac{1}{2}t_k^2} \\ &= \lim_{k \rightarrow \infty} \frac{f(x_k) - f(\bar{x}) - t_k df(\bar{x})(h) - \frac{1}{2}t_k^2 \langle v, w_k \rangle}{\frac{1}{2}t_k^2}. \end{aligned} \quad (3.5)$$

Consider

$$c_k := \frac{f(x_k) - f(\bar{x}) - t_k df(\bar{x})(h)}{\frac{1}{2}t_k^2}.$$

Hence

$$f(x_k) = f(\bar{x}) + t_k df(\bar{x})(h) + \frac{1}{2}t_k^2 c_k.$$

Since $t_k w_k \rightarrow 0$ and $d^2 f(x, v)(h)$ is finite, because of (3.5) we have that $t_k c_k \rightarrow 0$. Then by formula

$$T_{\text{epi}f}^2[(\bar{x}, f(x)), (h, df(\bar{x})(h))] = \text{epi}f_{-}^{\downarrow\downarrow}(\bar{x}; h, \cdot),$$

it follows from the outer second order regularity of f that

$$d((w_k, c_k), \text{epi}f_{-}^{\downarrow\downarrow}(\bar{x}, h, \cdot)) \rightarrow 0.$$

There exists $(w'_k, c'_k) \in \text{epi}f_{-}^{\downarrow\downarrow}(\bar{x}, h, \cdot)$ such that $(w_k, c_k) - (w'_k, c'_k) \rightarrow (0, 0)$. This implies that

$$\begin{aligned} \frac{f(x_k) - f(\bar{x}) - t_k df(\bar{x})(h)}{\frac{1}{2}t_k^2} &= c'_k + (c_k - c'_k) \\ &\geq f_{-}^{\downarrow\downarrow}(\bar{x}, h, w'_k) + (c_k - c'_k). \end{aligned}$$

It follows from (3.5) that

$$d^2 f(\bar{x}, v)(h) \geq \liminf_{k \rightarrow \infty} \{f_{-}^{\downarrow\downarrow}(\bar{x}, h, w'_k) - \langle v, w'_k \rangle + \langle v, w_k - w'_k \rangle\}.$$

Since $\langle v, w_k - w'_k \rangle \rightarrow 0$, it follows that $d^2 f(\bar{x}, v)(h)$ is greater than or equal to the left hand side of (3.2), and hence the equality follows. \square

Theorem 3.1 (twice epi-differentiability of parabolically regular functions)

Let $\bar{v} \in \hat{\partial}f(\bar{x})$ and let f be parabolically epi-differentiable at \bar{x} for every $h \in K_f(\bar{x}, \bar{v})$. If f is parabolically regular at \bar{x} for \bar{v} , then it is properly twice epi-differentiable at \bar{x} for \bar{v} with

$$d^2f(\bar{x}, \bar{v})(h) = \begin{cases} \min_{w \in \mathbb{R}^n} \{f_-^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle\} & \text{if } h \in K_f(\bar{x}, \bar{v}), \\ +\infty, & \text{otherwise.} \end{cases} \quad (3.6)$$

Proof. It follows from the parabolic epi-differentiability of f at \bar{x} for every $h \in K_f(\bar{x}, \bar{v})$ and Proposition 2.5 that $\text{dom}d^2f(\bar{x}, \bar{v}) = K_f(\bar{x}, \bar{v})$. This together with Proposition 3.1 justifies the second subderivative formula (3.6).

To establish the twice epi-differentiability of f at \bar{x} for \bar{v} , we need show that $\forall h \in \mathbb{R}^n, \forall t_k \downarrow 0, \exists h_k \rightarrow h$ s. t.

$$d^2f(x, v)(h) = \lim_{k \rightarrow \infty} \frac{f(x + t_k h_k) - f(x) - t_k \langle v, h_k \rangle}{\frac{1}{2} t_k^2}.$$

Pick $h \in K_f(\bar{x}, \bar{v})$ and $\forall t_k \downarrow 0$. Since f is parabolically regular at \bar{x} for \bar{v} , by Proposition 3.1, we find $w \in \mathbb{R}^n$ such that

$$d^2f(\bar{x}, \bar{v})(h) = f_-^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle. \quad (3.7)$$

By the parabolic epi-differentiability of f at \bar{x} for h , $\exists w_k \rightarrow w$ s.t. have

$$f_-^{\downarrow\downarrow}(\bar{x}; h, w) = \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h + \frac{1}{2} t_k^2 w_k) - f(\bar{x}) - t_k df(\bar{x})(h)}{\frac{1}{2} t_k^2}. \quad (3.8)$$

Define $h_k := h + \frac{1}{2}w_k$ for all k . We obtain

$$\begin{aligned}\Delta_{t_k}^2 f(\bar{x}, \bar{v})(h_k) &= \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h_k) - f(\bar{x}) - t_k \langle \bar{v}, h_k \rangle}{\frac{1}{2} t_k^2} \\ &= \lim_{k \rightarrow \infty} \frac{f(\bar{x} + t_k h + \frac{1}{2} t_k^2 w_k) - f(\bar{x}) - t_k \langle \bar{v}, h \rangle}{\frac{1}{2} t_k^2} - \langle \bar{v}, w_k \rangle.\end{aligned}$$

This together with (3.7) and (3.8) results in

$$\lim_{k \rightarrow \infty} \Delta_{t_k}^2 f(\bar{x}, \bar{v})(h_k) = f_-^{\downarrow\downarrow}(\bar{x}; h, w) - \langle \bar{v}, w \rangle = d^2 f(\bar{x}, \bar{v})(h)$$

which justifies (2.24) for every $h \in K_f(\bar{x}, \bar{v})$. Finally, we are going to show the validity of (2.24) for every $h \notin K_f(\bar{x}, \bar{v})$. For any such a h , we see that $d^2 f(\bar{x}, \bar{v})(h) = \infty$. Hence

$$d^2 f(\bar{x}, \bar{v})(h) = e - \limsup_{t \downarrow 0} \Delta_t^2 f(\bar{x}, \bar{v})(h) = \infty.$$

This completes the proof of the Theorem. □

Proposition 3.3 (conjugate of twice parabolic epiderivatives)

Let $f: \mathbb{R}^n \rightarrow \mathbb{R}$ be a proper l.s.c. and convex function, and $\bar{v} \in \hat{\partial}f(\bar{x})$, and let f be parabolically epi-differentiable at \bar{x} for every $h \in K_f(\bar{x}, \bar{v})$. If f is parabolically regular at \bar{x} for \bar{v} , then $\phi(w) := f_{-}^{\downarrow\downarrow}(\bar{x}; h, w)$ is proper l.s.c. and convex functions and its conjugate function is given by

$$\phi^*(v) = \begin{cases} -d^2 f(\bar{x}, \bar{v})(h) & \text{if } v \in \mathcal{A}(\bar{x}, h), \\ +\infty, & \text{otherwise,} \end{cases} \quad (3.9)$$

where $\mathcal{A}(\bar{x}, h) = \{v \in \partial f(\bar{x}) \mid df(\bar{x})(h) = \langle v, h \rangle\}$.

Proof. We know that ϕ is proper lower semicontinuous and convex function. Pick $v \in \mathcal{A}(\bar{x}, h)$, the formula (3.9) is clearly true due to Theorem 3.1.

Assume now that $v \notin \mathcal{A}(\bar{x}, h)$. This means either $\bar{v} \notin \partial f(\bar{x})$ or $\mathrm{d}f(\bar{x})(h) \neq \langle \bar{v}, h \rangle$. Define

$$\Delta_{t, \bar{x}, h} f(w) = \frac{f(\bar{x} + th + \frac{1}{2}t^2 w) - f(\bar{x}) - \mathrm{d}f(\bar{x})(h)}{\frac{1}{2}t^2}, \quad w \in \mathbb{R}^n, t > 0.$$

It is not hard to see that $\Delta_{t, \bar{x}, h} f(w)$ are proper, convex, and

$$(\Delta_{t, \bar{x}, h} f)^*(v) = \frac{f(\bar{x}) + f^*(v) - \langle v, \bar{x} \rangle}{\frac{1}{2}t^2} + \frac{\mathrm{d}f(\bar{x})(h) - \langle v, h \rangle}{\frac{1}{2}t}, \quad v \in \mathbb{R}^n$$

The parabolic epi-differentiability of f at \bar{x} for h amounts to the sets $\mathrm{epi} \Delta_{t, \bar{x}, h} f(\cdot)$ converging to $\mathrm{epi} \phi$ as $t \downarrow 0$.

Appealing to [14, Theorem 11.34] tells us that the former is equivalent to the sets $\text{epi}(\Delta_{t,\bar{x},h}f)^*$ converging to $\text{epi}\phi^*$ as $t \downarrow 0$. This, in particular, means that $\forall t_k \downarrow 0, \exists v_k \rightarrow \bar{v}$ s. t.

$$\phi^*(\bar{v}) = \lim_{k \rightarrow \infty} (\Delta_{t_k, \bar{x}, h} f)^*(v_k).$$

If $\bar{v} \notin \partial f(\bar{x})$, then we have

$$f(\bar{x}) + f^*(v) - \langle v, \bar{x} \rangle > 0.$$

Since f^* is l.s.c., we get

$$\liminf_{k \rightarrow \infty} \frac{f(\bar{x}) + f^*(v_k) - \langle v_k, \bar{x} \rangle}{\frac{1}{2}t_k} + \frac{df(\bar{x})(h) - \langle v_k, h \rangle}{\frac{1}{2}} \geq \infty$$

which in turn confirms that

$$\phi^*(\bar{v}) = \lim_{k \rightarrow \infty} (\Delta_{t_k, \bar{x}, h} f)^*(v_k) = \infty.$$

If $\bar{v} \in \partial f(\bar{x})$ but $\langle v, h \rangle < \mathrm{d}f(\bar{x})(h)$. Since we always have

$$f(\bar{x}) + f^*(v_k) - \langle v_k, \bar{x} \rangle \geq 0,$$

we arrive at

$$\phi^*(\bar{v}) \geq \lim_{k \rightarrow \infty} \frac{\mathrm{d}f(\bar{x})(h) - \langle v_k, h \rangle}{\frac{1}{2}t_k} = \infty.$$



Example 3.1 (piecewise linear-quadratic functions)

$f: \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ is convex piecewise linear-quadratic. That is if $\text{dom} f = \cup_{i=1}^p C_i$ with C_i being polyhedral convex sets for $i = 1, \dots, p$, and if f has the form

$$f(x) = \frac{1}{2} \langle A_i x, x_i \rangle + \langle a_i, x_i \rangle + \alpha_i \quad \text{for all } x \in C_i, \quad (3.10)$$

Where A_i is an $n \times n$ symmetric matrix, $a_i \in \mathbb{R}^n$, and $\alpha_i \in \mathbb{R}$ for $i = 1, \dots, p$. It was proved in [14, Propsoition 13.9] that the second sub-derivative of f at \bar{x} for $\bar{v} \in \partial f(\bar{x})$ can be calculated by

$$d^2 f(\bar{x}, \bar{v})(h) = \begin{cases} \langle A_i, h \rangle & \text{if } h \in T_{C_i}(\bar{x}) \cap \{\bar{v}_i\}^\perp \\ \infty & \text{otherwise,} \end{cases} \quad (3.11)$$

where $\bar{v}_i = \bar{v} - A_i \bar{x} - a_i$.

To prove the parabolic regularity of f at \bar{x} for \bar{v} , pick $h \in \mathbb{R}^n$ with $d^2f(\bar{x}, \bar{v})(h) < \infty$. Then $\exists i$ s.t. $h \in T_{C_i}(\bar{x}) \cap \{\bar{v}_i\}^\perp$. Since C_i is a polyhedral convex set, $\exists \tau > 0$ s.t. $\bar{x} + th \in C_i \forall t \in [0, \tau]$. Pick $t_k \downarrow 0$ s.t. $t_k \in [0, \tau]$ and let $h_k := h \forall k = 1, 2, \dots$. Thus

$$\Delta_{t_k}^2 f(\bar{x}, \bar{v})(h_k) = \langle A_i h, h \rangle + \frac{\langle h_k, \bar{v} - A_i \bar{x} - a_i \rangle}{\frac{1}{2} t_k^2} = \langle A_i h, h \rangle, \quad (3.12)$$

which implies that $\Delta_{t_k}^2 f(\bar{x}, \bar{v})(h_k) \rightarrow d^2f(\bar{x}, \bar{v})(h)$ as $k \rightarrow \infty$ with $\limsup_{k \rightarrow \infty} \frac{\|h_k - h\|}{t_k} = 0 < \infty$. Hence, f is parabolic regular at \bar{x} for \bar{v} .

This function is second order regular at \bar{x} and so parabolical epi-differentiable at \bar{x} for \bar{v} .

Another example for parabolic regularity and parabolical epi-differentiable function is the sum of the k largest eigenvalues functions for symmetric matrix.

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3.2 First/second order chain rules of subderivatives

Definition 3.4

$\Psi: \mathbb{R}^n \rightrightarrows \mathbb{R}^m$ is metric regular at a point $(\bar{x}, \bar{y}) \in \text{gr}\Psi$, at a rate c , if $\exists U$ of \bar{x} and V of \bar{y} ,

$$d(x, \Psi^{-1}(y)) \leq cd(y, \Psi(x)) \quad \forall x \in U, y \in V. \quad (3.13)$$

If $y = \bar{y}$ in (3.13), Ψ is said to be metric subregular (\bar{x}, \bar{y}) . The infimum of the set of values c for which this holds is the modulus of metric regularity, denoted by $\text{reg}\Psi(\bar{x}, \bar{y})$.

Consider the constraint system

$$\Psi(x) = F(x) - K,$$

where $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a continuously differentiable mapping and $K \subset \mathbb{R}^m$ is a closed set.

Definition 3.5

Robinson's constraint qualification holds at a point $\bar{x} \in \mathbb{R}^n$ with $F(\bar{x}) \in K$, with respect to the mapping $F(\cdot)$ and the set K , if the following condition holds:

$$0 \in \text{int}\{F(\bar{x}) + DF(\bar{x})\mathbb{R}^n - K\} \quad (3.14)$$

$\Psi(x) = F(x) - K$ is metric regular at $(\bar{x}, 0)$, i.e., $\exists U$ of \bar{x} and V of 0 and $\exists c > 0$ s.t.

$$d(x, F^{-1}(K - y)) \leq cd(F(x) + y, K) \quad \forall x \in U, y \in V. \quad (3.15)$$

Observe that Ψ is metric regular at $(\bar{x}, 0)$ if and only if

$$y^* \in N_K(F(\bar{x})), DF(\bar{x})^* y^* = 0 \Rightarrow y^* = 0 \quad (3.16)$$

and both are equivalent to (3.14) whenever K is convex.

Set $C := \{x \in \mathbb{R}^n | F(x) \in K\}$ and $\bar{x} \in C$. Then we always have

$$T_C(\bar{x}) \subset \{h \in \mathbb{R}^n | DF(\bar{x})h \in T_K(F(\bar{x}))\}; \quad (3.17)$$

$$T_C^2(\bar{x}, h) \subset \{w \in \mathbb{R}^n | DF(\bar{x})w + \langle h, DF(\bar{x})h \rangle \in T_K^2(F(\bar{x}), DF(\bar{x})h)\},$$

and the equality hold whenever $\Psi(x) = F(x) - K$ is metric regular at $(\bar{x}, 0)$. C is parabolically derivable at x for h whenever K is Clarke regular at $F(\bar{x})$ and parabolically derivable at $F(\bar{x})$ for $DF(\bar{x})h$ (in particular K is convex) under the assumption of (3.16).

Proposition 3.4

Let $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a continuously differentiable mapping and $g: \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ be l.s.c. convex function taking a finite value at point $\bar{y} = F(\bar{x})$. Suppose that Robinson's constraint qualification condition

$$0 \in \text{int}\{F(\bar{x}) + DF(\bar{x})\mathbb{R}^n - \text{dom}g\} \quad (3.18)$$

holds. Then the lower and upper directional epiderivatives of the composite function $\psi := g \circ F$ coincide at \bar{x} , and

$$\psi^\downarrow(\bar{x}, h) = g^\downarrow(F(\bar{x}), DF(\bar{x})h). \quad (3.19)$$

Proof. Let $\hat{K} := \text{epi}g$ and $\hat{F}(x, \alpha) := (F(x), \alpha)$, $\alpha \in \mathbb{R}$. Then $\hat{F}^{-1}(\hat{K}) = \text{epi}(g \circ F)$ and that (3.18) implies that Robinson's constraint qualification for the set \hat{K} and the mapping \hat{F} , at $(\bar{x}, g(F(\bar{x})))$, i.e.

$$0 \in \text{int} \left\{ \begin{bmatrix} F(\bar{x}) \\ g(F(\bar{x})) \end{bmatrix} + \begin{bmatrix} DF(\bar{x})\mathbb{R}^n \\ \mathbb{R} \end{bmatrix} - \text{epi}g \right\}.$$

It follows that

$$T_{\text{epi}(g \circ F)}(\bar{x}, (g \circ F)(\bar{x})) = D\hat{F}(\bar{x}, \alpha)^{-1} T_{\text{epi}(g)}(\hat{F}(\bar{x}), g(F(\bar{x}))). \quad (3.20)$$

□

The following second order chain rules can be proved in a similar way.

Theorem 3.2

Let $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a twice continuously differentiable mapping and $g : \mathbb{R}^m \rightarrow \overline{\mathbb{R}}$ be a l.s.c. convex function taking a finite value at a point $\bar{y} := F(\bar{x})$. Suppose that Robinson's constraint qualification $0 \in \text{int}\{F(\bar{x}) + DF(\bar{x})\mathbb{R}^n - \text{dom}g\}$ holds. Then, provided $g^\downarrow(F(\bar{x}), DF(\bar{x})h)$ is finite, we have

$$\psi_-^{\downarrow\downarrow}(\bar{x}; h, w) = g_-^{\downarrow\downarrow}(F(\bar{x}); DF(\bar{x})h, DF(\bar{x})w + D^2F(\bar{x})(h, h)), \quad (3.21)$$

$$\psi_+^{\downarrow\downarrow}(\bar{x}; h, w) = g_+^{\downarrow\downarrow}(F(\bar{x}); DF(\bar{x})h, DF(\bar{x})w + D^2F(\bar{x})(h, h)), \quad (3.22)$$

We have the following result concerning outer second order regularity of composite function

Proposition 3.5

Let $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a twice continuously differentiable mapping, $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ a l.s.c. convex function taking a finite value at a point $\bar{y} := F(\bar{x})$, and $h \in \mathbb{R}^n$ and $\lambda \in \partial g(F(\bar{x}))$ satisfying $DF(\bar{x})h \in K_g(F(\bar{x}), \lambda)$. Suppose that g is outer second order regular at \bar{y} in the direction $DF(\bar{x})h$, and that Robinson's constraint qualification $0 \in \text{int}\{\bar{y} + DF(\bar{x})X - \text{dom}g\}$ holds. Then the composite function $\psi = g \circ F$ is outer second order regular at \bar{x} in the direction h .

Theorem 3.3

Let $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a twice continuously differentiable mapping, $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ a l.s.c. convex, $g(F(\bar{x}))$ is finite, $h \in \mathbb{R}^n$ and $\lambda \in \partial g(\bar{y})$ satisfying $DF(\bar{x})h \in K_g(\bar{x}, \lambda)$. Suppose that g is outer second order regular at \bar{y} in the direction $DF(\bar{x})h$, and that Robinson's constraint qualification $0 \in \text{int}\{F(\bar{x}) + DF(\bar{x})\mathbb{R}^n - \text{dom}g\}$ holds. Then the composite function $\psi = g \circ F$ is parabolically regular at \bar{x} in the direction h for $v := [DF(\bar{x})]^*\lambda$, and

$$d^2\psi(\bar{x}, v)(h) = \inf_{w \in X} \{g_{-}^{\downarrow\downarrow}(\bar{y}, DF(\bar{x})h, DF(\bar{x})w + D^2F(\bar{x})(h, h)) - \langle v, w \rangle\}. \quad (3.23)$$

Proof. By Proposition 3.5, it follows from outer second order regularity of g that the composite function $g \circ F$ is outer second order regular at \bar{x} in direction h . Also, since $\lambda \in \partial g(\bar{y})$ and $F(\bar{x} + th') = F(\bar{x}) + tDF(\bar{x})h' + o(t)$ for $h' \rightarrow h$ and $t \downarrow 0$, we have

$$g(F(\bar{x} + th')) - g(F(\bar{x})) - t\langle \lambda, DF(\bar{x})h' \rangle \geq o(t),$$

and hence $d^2(g \circ F)(\bar{x}, v)(h) \geq 0$. By Proposition 3.2, the parabolic regularity of $g \circ F$ then follows. Formula (3.23) follows from (low-com-der) and the corresponding formula from Theorem 3.2 for the lower second order epiderivative of the composite function. \square

Consider the composite function $\psi(x) = g(F(x))$, where $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ a twice differentiable at \bar{x} and $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is proper, l.s.c., convex, and Lipschitz continuous around $F(\bar{x})$ relative to its domain. It is easy to see that

$$\text{dom}\psi = \{x \in \mathbb{R}^n \mid F(x) \in \text{dom}g\}.$$

$\Psi(x) = F(x) - \text{dom}g$ is metric subregular at $(\bar{x}, 0)$, i.e., if $\exists c > 0$, $\exists U$ of \bar{x} ,

$$d(x, \text{dom}\psi) \leq \kappa d(F(x), \text{dom}g) \quad \forall x \in U. \quad (3.24)$$

Robinson condition (3.18) implies that $\Psi(x) = F(x) - \text{dom}g$ is metric subregular at $(\bar{x}, 0)$.

The following results about the chain rules for epiderivatives of composite function in this section comes from:

1. A. Mohammadi and B.S. Mordukhovich, Variational analysis in normed spaces with applications to constrained optimization, SIAM Journal on Optimization, 31(2021):1, 569-603.
2. A. Mohammadi and M.E. Sarabi, Twice epi-Differentiability of extended-real-valued functions with applications in composite optimization, SIAM J. Optim., 30(2020):3, 2379-2409.

Theorem 3.4

Let $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be differentiable at \bar{x} and $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ be Lipschitz continuous around $F(\bar{x})$ relative to its domain. If the metric subregularity constraint qualification (3.24) holds, then the following hold:

(i) for any $h \in \mathbb{R}^n$, the following subderivative chain rule for ψ at \bar{x} holds:

$$d\psi(\bar{x})(h) = dg(F(\bar{x}))(DF(\bar{x})h);$$

(ii) we have the chain rules

$$T_{\text{dom}\psi}(\bar{x}) = \{h \in \mathbb{R}^n \mid DF(\bar{x})h \in T_{\text{dom}g}(F(\bar{x}))\}.$$

If, in addition, F is continuously differentiable at \bar{x} , g is convex, then

$$\partial\psi(\bar{x}) = DF(\bar{x})^* \partial g(F(\bar{x})).$$

Proof. Since $F(\bar{x})$ is continuously differentiable at \bar{x} ,

$$F(\bar{x} + th) = F(\bar{x}) + tDF(\bar{x})h + o(t\|h\|).$$

It is easy to show that $T_{\text{dom}\psi}(\bar{x}) \subset \{h \in \mathbb{R}^n \mid DF(\bar{x})h \in T_{\text{dom}g}(F(\bar{x}))\}$.

Take h being such that $DF(\bar{x})h \in \text{dom } dg(F(\bar{x}))$, there exist sequences $t_k \downarrow 0$ and $v_k \rightarrow DF(\bar{x})h$ such that $F(\bar{x}) + t_k v_k \in \text{dom}g$ for all $k \geq 1$. Then the (3.24) yields

$$d(\bar{x} + t_k h, \text{dom}\psi) \leq \kappa d(F(\bar{x} + t_k h), \text{dom}g), \quad k \geq 1, \quad (3.25)$$

which in turn implies

$$\begin{aligned} \frac{d(\bar{x} + t_k h, \text{dom}\psi)}{t_k} &\leq \frac{\kappa}{t_k} d(F(\bar{x}) + t_k DF(\bar{x})h + o(t_k), \text{dom}g) \\ &\leq \frac{\kappa}{t_k} \|F(\bar{x}) + t_k DF(\bar{x})h + o(t_k) - F(\bar{x}) - t_k v_k\| \\ &= \kappa \|DF(\bar{x})h - v_k + \frac{o(t_k)}{t_k}\| \quad \forall k \geq 1. \end{aligned} \quad (3.26)$$

This implies that $h \in T_{\text{dom}\psi}(\bar{x})$.

For any $h \in \mathbb{R}^n$, we have

$$\begin{aligned} d\psi(\bar{x})(h) &= \liminf_{\substack{t \downarrow 0 \\ h' \rightarrow h}} \frac{g(F(\bar{x} + th')) - g(F(\bar{x}))}{t} \\ &= \liminf_{\substack{t \downarrow 0 \\ h' \rightarrow h}} \frac{g(F(\bar{x}) + tDF(\bar{x})h' + o(t\|h\|)) - g(F(\bar{x}))}{t} \\ &\geq dg(F(\bar{x}))(DF(\bar{x})h). \end{aligned}$$

Take any $h \in \mathbb{R}^n$ and observe from the Lipschitz continuity of g around $F(\bar{x})$ relative to its domain that $dg(F(\bar{x}))(DF(\bar{x})h) > -\infty$. Since the converse inequality is obvious if $dg(F(\bar{x}))(DF(\bar{x})h) = \infty$, we may assume that the value $dg(F(\bar{x}))(DF(\bar{x})h)$ is finite, i.e., By definition, there exist sequences $t_k \downarrow 0$ and $v_k \rightarrow DF(\bar{x})h$ such that

$$dg(F(\bar{x}))(DF(\bar{x})h) = \lim_{k \rightarrow \infty} \frac{g(F(\bar{x}) + t_k v_k) - g(F(\bar{x}))}{t_k} < \infty.$$

Suppose without loss of generality that $F(\bar{x}) + t_k v_k \in \text{dom}g$ for all $k \geq 1$. From the above proof, we see that there exist $h_k \in \frac{\text{dom}\psi - \bar{x}}{t_k}$ satisfying

$$\|h_k - h\| \leq \kappa \|DF(\bar{x})h - v_k + \frac{o(t_k)}{t_k}\|.$$

It follows that $h_k \rightarrow h \in$ as $k \rightarrow \infty$. Therefore,

$$\begin{aligned} & dg(F(\bar{x}))(DF(\bar{x})h) \\ = & \lim_{k \rightarrow \infty} \left[\frac{g(F(\bar{x} + t_k h_k)) - g(F(\bar{x}))}{t_k} + \frac{g(F(\bar{x}) + t_k v_k) - g(F(\bar{x} + t_k h_k))}{t_k} \right] \\ \geq & \liminf_{k \rightarrow \infty} \frac{g(F(\bar{x} + t_k h_k)) - g(F(\bar{x}))}{t_k} - \ell \lim_{k \rightarrow \infty} \left\| \frac{F(\bar{x} + t_k h_k) - F(\bar{x})}{t_k} - v_k \right\| \\ \geq & d\psi(\bar{x})(h) \end{aligned}$$

We remain to prove that

$$\partial\psi(\bar{x}) = DF(\bar{x})^* \partial g(F(\bar{x}))$$

whenever F is continuously differentiable at \bar{x} , g is convex.

The inclusion $\partial\psi(\bar{x}) \supset DF(\bar{x})^* \partial g(F(\bar{x}))$ always holds. The inclusion

$$\partial\psi(\bar{x}) \subset DF(\bar{x})^* \partial g(F(\bar{x}))$$

is a very fundamental calculus formula in variational analysis.

It is well known that this is true under Robinson's condition (3.18). It was proved under the weaker condition (3.24) in [6]¹ very recently.

¹[6]. A. Mohammadi and B.S. Mordukhovich, Variational analysis in normed spaces with applications to constrained optimization, SIAM Journal on Optimization, 31(2021):1, 569-603.

This Theorem remains true if we replace the metric subregularity condition (3.24) by the assumption that Ψ is metric-regular at $(\bar{x}, 0)$ and $\text{epi}g$ is Clarke regular at $(\bar{x}, g(\bar{x}))$.

Theorem 3.5

Suppose that $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is twice differentiable at \bar{x} and $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is proper, l.s.c., convex and Lipschitz continuous around $F(\bar{x})$ relative to its domain and suppose that the metric subregularity constraint qualification (3.24) holds. If g is parabolically epi-differentiable at $\bar{y} := F(\bar{x})$ in the direction $DF(\bar{x})h$ with $h \in T_{\text{dom}f}(\bar{x})$, then ϕ is parabolically epi-differentiable at \bar{x} for h and the following conditions hold:

(i) for every $w \in \mathbb{R}^n$,

$$\psi_{-}^{\downarrow\downarrow}(\bar{x}; h, w) = g_{-}^{\downarrow\downarrow}(F(\bar{x}); DF(\bar{x})h, DF(\bar{x})w + D^2F(\bar{x})(h, h)); \quad (3.27)$$

(ii)

$$\begin{aligned} \text{dom}\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, \cdot) &= T_{\text{dom}\psi}^2(\bar{x}, h) \\ &= \{w \in \mathbb{R}^n \mid DF(\bar{x})w + \langle h, DF(\bar{x})h \rangle \in T_{\text{dom}g}^2(F(\bar{x}), DF(\bar{x})h)\}. \end{aligned}$$

Proof. It is easy to show that

$$T_{\text{dom}\psi}^2(\bar{x}, h) \subset \{w \in \mathbb{R}^n \mid DF(\bar{x})w + \langle h, DF(\bar{x})h \rangle \in T_{\text{dom}g}^2(F(\bar{x}), DF(\bar{x})h)\}.$$

Let w be satisfying

$$z := DF(\bar{x})w + \langle h, DF(\bar{x})h \rangle \in T_{\text{dom}g}^2(F(\bar{x}), DF(\bar{x})h). \quad (3.28)$$

Then there exists $t_k \downarrow 0$, $z_k \rightarrow z$ such that

$$F(\bar{x}) + t_k DF(\bar{x})h + \frac{1}{2}t_k^2 z_k \in \text{dom}g. \quad (3.29)$$

Let $x_k := \bar{x} + t_k h + \frac{1}{2}t_k^2 w$. Then by (3.24) we have

$$\begin{aligned} d(x_k, \text{dom}\psi) &\leq cd(F(x_k), \text{dom}g) \\ &\leq c\|F(x_k) - F(\bar{x}) - t_k DF(\bar{x})h - \frac{1}{2}t_k^2 z_k\| \\ &= c\|\frac{1}{2}t_k^2(DF(\bar{x})w + \langle h, DF(\bar{x})h \rangle) + o(t_k^2) - \frac{1}{2}t_k^2 z_k\| \end{aligned}$$

It follows that $\frac{d(x_k, \text{dom}\psi)}{t_k^2} \rightarrow 0$ and so $w \in T_{\text{dom}\psi}^2(\bar{x}, h)$.

Since F is twice continuously differentiable at \bar{x} , it is not hard to prove that for every $w \in \mathbb{R}^n$, $z = DF(\bar{x})w + \langle h, DF(\bar{x})h \rangle$,

$$g_-^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z) \leq \psi_-^{\downarrow\downarrow}(\bar{x}, h, w). \quad (3.30)$$

For every $w \in T_{\text{dom}\psi}^2(\bar{x}, h)$, there exists $t_k \downarrow 0$, $w_k \rightarrow w$ such that

$$x_k := \bar{x} + t_k h + \frac{1}{2} t_k^2 w_k \in \text{dom}\psi.$$

Since g is parabolically epi-differentiable at $\bar{y} := F(\bar{x})$ in the direction $DF(\bar{x})h$, we have $\text{dom}g_-^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, \cdot) \neq \emptyset$ and for z and the above $t_k \downarrow 0$, there exist $z_k \rightarrow z$ such that

$$g_-^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z) = \lim_{k \rightarrow \infty} \frac{g(F(\bar{x}) + t_k DF(\bar{x})h + \frac{1}{2} t_k^2 z_k) - g(F(\bar{x})) - t_k dg(F(\bar{x}), DF(\bar{x})h)}{\frac{1}{2} t_k^2}.$$

Since $z \in T_{\text{dom}g}^2(F(\bar{x}), DF(\bar{x})h) = \text{dom}g_{-}^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, \cdot)$ due to Proposition 2.3, we have $g_{-}^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z) < \infty$. It follows that $y_k := F(\bar{x}) + t_k DF(\bar{x})h + \frac{1}{2}t_k^2 z_k \in \text{dom}g$ for k sufficiently large. By the Lipschitz property of g we obtain

$$\begin{aligned} \psi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) &= \liminf_{k \rightarrow \infty} \frac{\psi(\bar{x} + t_k h + \frac{1}{2}w_k) - \psi(\bar{x}) - t_k d\psi(\bar{x}, h)}{\frac{1}{2}t_k^2} \\ &\leq \limsup_{k \rightarrow \infty} \frac{g(F(x_k)) - g(F(\bar{x})) - t_k dg(F(\bar{x}), DF(\bar{x})h)}{\frac{1}{2}t_k^2} \\ &\leq \limsup_{k \rightarrow \infty} \frac{g(y_k) - g(F(\bar{x})) - t_k dg(F(\bar{x}), DF(\bar{x})h)}{\frac{1}{2}t_k^2} + \limsup_{k \rightarrow \infty} \frac{g(F(x_k)) - g(y_k)}{\frac{1}{2}t_k^2} \\ &\leq g^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z). \end{aligned}$$

Therefore, we have proved that

$$\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) = g^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z)$$

for the case $w \in T_{\text{dom}\psi}^2(\bar{x}, h)$.

For $w \notin T_{\text{dom}\psi}^2(\bar{x}, h)$, since $\text{dom}\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, \cdot) \subset T_{\text{dom}\psi}^2(\bar{x}, h)$ always holds, we have $\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) = +\infty$ and

$$z \notin T_{\text{dom}g}^2(F(\bar{x}), DF(\bar{x})h) = \text{dom}g_{-}^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, \cdot).$$

This implies that $g_{-}^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z) = +\infty$ and $w \notin \text{dom}\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, \cdot)$ due to (3.30). This shows that

$$\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) = g_{-}^{\downarrow\downarrow}(F(\bar{x}), DF(\bar{x})h, z) = +\infty$$

for the case $w \notin T_{\text{dom}\psi}^2(\bar{x}, h)$ and $\text{dom}\psi_{-}^{\downarrow\downarrow}(\bar{x}, h, \cdot) = T_{\text{dom}\psi}^2(\bar{x}, h)$. Hence, we complete the proofs of (i) and (ii). \square

It is worth mentioning that a chain rule for parabolic subderivatives for the composite form (3.27) was achieved in [14, Exercise 13.63]² and [2, Proposition 3.42]³ when g is merely a proper l.s.c. function and the assumption that $\Psi(x) = F(x) - \text{dom}g$ is metric regular at $(\bar{x}, 0)$.

²Rockafellar and R.J.-B. Wets, Variational Analysis

³J.F. Bonnans and A. Shapiro, Perturbation Analysis of Optimization Problems

Given $\bar{v} \in \partial\psi(\bar{x})$, we define the set of Lagrangian multipliers associated with (\bar{x}, \bar{v}) by

$$\Lambda(\bar{x}, \bar{v}) := \{\lambda \in \mathbb{R}^m \mid DF(\bar{x})^* \lambda = \bar{v}, \lambda \in \partial g(F(\bar{x}))\}. \quad (3.31)$$

It is easy to see that

$$h \in K_\psi(\bar{x}, \bar{v}) \Leftrightarrow DF(\bar{x})h \in K_g(F(\bar{x}), \lambda), \quad \forall \lambda \in \Lambda(\bar{x}, \bar{v})$$

whenever either the Robinson's condition (3.18) holds or the metric subregularity (3.24) and g is Lipschitz continuous around $F(\bar{x})$ relative to its domain.

In the following we always assume that $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is twice differentiable at \bar{x} and $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is proper, l.s.c., convex and Lipschitz continuous around $\bar{y} = F(\bar{x})$ (with $g(\bar{y})$ finite) relative to its domain.

Proposition 3.6

Suppose that the metric subregularity constraint qualification (3.24) holds. If for every $\bar{v} \in \partial\psi(\bar{x})$ and $\lambda \in \Lambda(\bar{x}, \bar{v})$, g is parabolically epi-differentiable at $F(\bar{x})$ in every direction $d \in K_g(F(\bar{x}), \lambda)$, then for every $h \in \mathbb{R}^n$ we have the lower estimate

$$d^2\psi(\bar{x}, \bar{v})(h) \geq \sup_{\lambda \in \Lambda(\bar{x}, \bar{v})} \{ \langle \lambda, D^2F(\bar{x})(h, h) \rangle + d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h) \}; \quad (3.32)$$

while for every $h \in K_\phi(\bar{x}, \bar{v})$ we obtain the upper estimate

$$d^2\psi(\bar{x}, \bar{v})(h) \leq \inf_{w \in \mathbb{R}^n} \{ g^{\downarrow\downarrow}(F(\bar{x}); DF(\bar{x})h, DF(\bar{x})w + D^2F(\bar{x})(h, h)) - \langle \bar{v}, w \rangle \} < \infty. \quad (3.33)$$

Proof. For any $\lambda \in \Lambda(\bar{x}, \bar{v})$, we can write

$$\begin{aligned} \Delta_t^2 \psi(\bar{x}, \bar{v})(h) &= \frac{g(F(\bar{x} + th)) - g(F(\bar{x})) - t\langle \bar{v}, h \rangle}{\frac{1}{2}t^2} \\ &= \frac{g(F(\bar{x}) + t\frac{F(\bar{x}+th)-F(\bar{x})}{t}) - g(F(\bar{x})) - t\langle \lambda, DF(\bar{x})h \rangle}{\frac{1}{2}t^2} \\ &= \frac{g(F(\bar{x}) + t\frac{F(\bar{x}+th)-F(\bar{x})}{t}) - g(F(\bar{x})) - t\langle \lambda, \frac{F(\bar{x}+th)-F(\bar{x})}{t} \rangle}{\frac{1}{2}t^2} \\ &+ \frac{\langle \lambda, F(\bar{x} + th) - F(\bar{x}) - tDF(\bar{x})h \rangle}{\frac{1}{2}t^2} \end{aligned}$$

Because $\frac{F(\bar{x}+th)-F(\bar{x})}{t} \rightarrow DF(\bar{x})h$ as $t \rightarrow 0$ and $h' \rightarrow h$, while at the same time

$$\frac{\langle \lambda, F(\bar{x} + th') - F(\bar{x}) - tDF(\bar{x})h' \rangle}{\frac{1}{2}t^2} \rightarrow \langle \lambda, D^2F(\bar{x})(h, h) \rangle,$$

we have $d^2\psi(\bar{x}, \bar{v})(h) \geq d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h) + \langle \lambda, D^2F(\bar{x})(h, h) \rangle$. This complete the proof of (3.32). The upper estimation (3.33) follows from Proposition 3.5 and Proposition 2.4. \square

The above result carries important information by which we can achieve a chain rule for the second subderivative. To do so, we should look for conditions under which the lower and upper estimates (3.32) and (3.33), respectively, coincide. This motivates us to consider the unconstrained optimization problem

$$\min_{w \in \mathbb{R}^n} \{g_-^{\downarrow\downarrow}(F(\bar{x}); DF(\bar{x})h, DF(\bar{x})w + D^2F(\bar{x})(h, h)) - \langle \bar{v}, w \rangle\} \quad (3.34)$$

Proposition 3.7

Suppose that the metric subregularity constraint qualification (3.24) holds. If for every $\bar{v} \in \partial\psi(\bar{x})$ and $\lambda \in \Lambda(\bar{x}, \bar{v})$, g is parabolically epi-differentiable at $F(\bar{x})$ in every direction $d \in K_g(F(\bar{x}), \lambda)$, and parabolically regular at $F(\bar{x})$ for λ , then for every $h \in K_\psi(\bar{x}, \bar{v})$, the dual problem of (3.34) is given by

$$d^2\psi(\bar{x}, \bar{v})(h) = \max_{\lambda \in \Lambda(\bar{x}, \bar{v})} \langle \lambda, D^2F(\bar{x})(h, h) \rangle + d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h), \quad (3.35)$$

Proof. By the classical Fenchel-Rockafellar duality theorem, we know that the dual problem of (3.34) is given by

$$\max_{\lambda \in \Lambda(\bar{x}, \bar{v})} \langle \lambda, D^2F(\bar{x})(h, h) \rangle + d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h). \quad (3.36)$$



Theorem 3.6

Suppose that the metric subregularity constraint qualification (3.24) holds. Suppose that for every $\bar{v} \in \partial\psi(\bar{x})$ and $\lambda \in \Lambda(\bar{x}, \bar{v})$, g is parabolically epi-differentiable at $F(\bar{x})$ in the direction $d \in K_g(F(\bar{x}), \lambda)$, and parabolically regular at $F(\bar{x})$ for λ . Then ψ is parabolically regular at \bar{x} for \bar{v} . Furthermore, for every $h \in \mathbb{R}^n$, the second subderivative of ψ at \bar{x} for \bar{v} is calculated by

$$\begin{aligned} d^2\psi(\bar{x}, \bar{v})(h) & \qquad \qquad \qquad (3.37) \\ &= \max_{\lambda \in \Lambda(\bar{x}, \bar{v})} \{ \langle \lambda, D^2F(\bar{x})(h, h) \rangle + d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h) \} \end{aligned}$$

Proof. If $h \in K_\psi(\bar{x}, \bar{v})$, It follows from Proposition 3.6 and Proposition 3.7 and Theorem 3.5 that (3.37) is satisfied and

$$d^2\psi(\bar{x}, \bar{v})(h) = \inf_{w \in \mathbb{R}^n} \{\psi_-^{\downarrow\downarrow}(\bar{x}, h, w) - \langle \bar{v}, w \rangle\}. \quad (3.38)$$

By Theorem 3.5, ψ is parabolic epi-differentiable at \bar{x} for every $h \in K_\psi(\bar{x}, \bar{v})$ and $\text{dom}\psi_-^{\downarrow\downarrow}(\bar{x}, h, \cdot) \neq \emptyset$ for every $h \in K_\psi(\bar{x}, \bar{v})$. So, since $d^2\psi(\bar{x}, \bar{v})$ is proper, by Proposition 2.5, $\text{dom}d^2\psi(\bar{x}, \bar{v}) = K_\psi(\bar{x}, \bar{v})$. Thus, if $h \notin K_\psi(\bar{x}, \bar{v})$, then $d^2\psi(\bar{x}, \bar{v})(h) = \infty$.

On the other hand, since

$$h \notin K_\psi(\bar{x}, \bar{v}) \Leftrightarrow DF(\bar{x})h \notin K_g(F(\bar{x}), \lambda) \quad \forall \lambda \in \Lambda(\bar{x}, \bar{v}),$$

we have $d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h) = \infty$. Therefore, both side in (??) are ∞ . \square

Let $f = \max\{f_1, f_2, \dots, f_m\}$ for $f_i \in C^2$. Then f is properly twice epi-differentiable at every point \bar{x} for $\forall v \in \partial f(\bar{x})$. Set $I(\bar{x}) = \{i | f(\bar{x}) = f_i(\bar{x})\}$, $I'(\bar{x}, h) = \{i \in I(\bar{x}) | \langle \nabla f_i(\bar{x}), h \rangle = df(\bar{x})(h)\}$. Then $v \in \partial f(\bar{x}) \Leftrightarrow v \in \text{con}\{\nabla f_i(\bar{x}) | i \in I(\bar{x})\}$, and then

$$\begin{aligned} d^2\psi(\bar{x}, \bar{v})(h) &= \delta_{K_f(\bar{x}, v)}(h) \\ &+ \max\left\{ \sum_{i \in I(\bar{x})} \alpha_i \langle h, \nabla^2 f_i(\bar{x}) h \rangle \right. \\ &\left. \alpha_i \geq 0, \sum_i \alpha_i = 1, v = \sum_{i \in I(\bar{x})} \alpha_i \nabla f_i(\bar{x}) \right\}. \end{aligned}$$

with $h \in K_f(\bar{x}, v)$ iff $v \in \text{con}\{\nabla f_i(\bar{x}) | i \in I'(\bar{x}, h)\}$.

When g is a convex piecewise linear-quadratic (or more case where g is fully amenable), the parabolic regularity of the composite $\psi = g \circ F$ and chain rule (3.37) were established in [14, Theorem 13.67] under the stronger condition of the metric regularity.

Corollary 3.1 (chain rule for twice epi-differentiability)

Suppose all the assumptions of Theorem 3.6. Then ψ is twice epi-differentiable at \bar{x} for \bar{v} .

4. The second order optimality conditions for composite optimization problems

- 1 Classical optimality conditions
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4.1 The second order optimality conditions for composite optimization problems

We consider the following composite optimization problem:

$$(P) \quad \min_{x \in \mathbb{R}^n} \{f(x) + g(F(x))\},$$

where $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is twice differentiable on \mathbb{R}^n , $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a twice differentiable mapping, and $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is a proper lower semicontinuous extended real-valued function.

Clearly, if $g(\cdot) := \delta_K(\cdot)$ is the indicator function of a nonempty set $K \subset Y$, then the problem (P) reduces to (CP).

$$(CP) \quad \min_{x \in \mathbb{R}^n} f(x) \quad \text{subject to } F(x) \in K.$$

The Lagrangian function for (P) is defined by

$$L(x, \lambda) = f(x) + \langle \lambda, F(x) \rangle,$$

and the Lagrangian multiplier set is defined by

$$\Lambda(\bar{x}) = \{\lambda \in \mathbb{R}^m \mid \nabla f(\bar{x}) + DF(\bar{x})^*(\lambda) = 0, \lambda \in \partial g(F(\bar{x}))\}.$$

If \bar{x} is a locally optimal solution of (P), then

$$0 \in \partial(f + g \circ F)(\bar{x}) = \nabla f(\bar{x}) + \partial(g \circ F)(\bar{x}).$$

Hence $-\nabla f(\bar{x}) \in \partial(g \circ F)(\bar{x})$. If the assumption of Theorem 3.4 is satisfied, then $-\nabla f(\bar{x}) \in DF(\bar{x})^* \partial g(F(\bar{x}))$. It is clear in this case that

$$\Lambda(\bar{x}) = \Lambda(\bar{x}, -\nabla f(\bar{x})). \quad (4.1)$$

Proposition 4.1

Let $\phi: \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$ be an extended lower semicontinuous real-valued function.

(i) If \bar{x} is a locally minimizer of ϕ on \mathbb{R}^n , then $d^2\phi(\bar{x}, 0)(h) \geq 0$ for all $h \in \mathbb{R}^n$.

(ii) If the second order growth condition holds at \bar{x} iff

$$d^2\phi(\bar{x}, 0)(h) > 0, \quad \forall h \in \mathbb{R}^n \setminus \{0\}. \quad (4.2)$$

Proposition 4.2

Assume that the first order necessary condition $d\phi(\bar{x}) \geq 0$. Then

(i) If \bar{x} is a local minimizer of ϕ , then

$$\inf_{w \in \mathbb{R}^n} \phi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) \geq 0, \quad \forall h \in K_{\phi}(\bar{x}, 0). \quad (4.3)$$

(ii) If, in addition, f is parabolically regular at \bar{x} , for $\bar{v} = 0$, then the second order growth condition holds at \bar{x} iff

$$\inf_{w \in \mathbb{R}^n} \phi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) > 0, \quad \forall h \in K_{\phi}(\bar{x}, 0). \quad (4.4)$$

Example 4.1

Let S be a convex closed subset of X and consider the corresponding indicator function $g(\cdot) := \delta_S(\cdot)$, a point $x \in S$ and a direction $h \in T_S(x)$. Recall that $dg(x, h) = \delta_{T_S(x)}(h)$, that $g_{-}^{\downarrow}(x, h, \cdot) = \delta_{T_S^2(x, h)}(\cdot)$, and that the indicator function g is (outer) second order regular at x iff the set S is (outer) second order regular at x . Let $v \in N_S(x)$ be such that $\langle v, h \rangle = 0$. By proposition 3.2 we have that if S is outer second order regular at x in the direction h , then g is parabolically regular at x in the direction h , for v , and

$$d^2g(x, v)(h) = \inf_{w \in T_S^2(x, h)} (-\langle v, w \rangle) = -\sigma(v, T_S^2(x, h)).$$

It is interesting to note that if $v = 0$, then $d^2g(x, 0)(h) = 0$ whether S is second order regular or not. On the other hand, $\inf_w g_{-}^{\downarrow}(x, h, w)$ is equal to 0 iff $T_S^2(x, h)$ is nonempty, and is $+\infty$ otherwise. Therefore, g is parabolically regular at x in the direction h , for $v = 0$, iff $T_S^2(x, h)$ is nonempty.

Results presented in this section can be used to derive second order optimality conditions for constrained problems. Let S be a closed set of \mathbb{R}^n , let $f: S \rightarrow \mathbb{R}$ be twice continuously differentiable function, and consider the problem

$$\min_{x \in S} f(x). \quad (4.5)$$

Clearly, the above problem is equivalent to minimization of the extended real valued function $\phi(\cdot) = f(\cdot) + \delta_S(\cdot)$ over \mathbb{R}^n .

Let $\bar{x} \in S$. We have then that for any $h \in X$,

$$d\phi(\bar{x})(h) = \nabla f(\bar{x})h + \delta_{T_S(\bar{x})}(h).$$

It follows that if \bar{x} is a local minimizer of f over S , then $\nabla f(\bar{x})h \geq 0$ for all $h \in T_S(\bar{x})$. Moreover, we have that

$$\phi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) = \nabla f(\bar{x})w + \nabla^2 f(\bar{x})(h, h) + \delta_{T_S^2(\bar{x}, h)}(w),$$

for all $h \in T_S(\bar{x})$ and $w \in \mathbb{R}^n$, and hence

$$\inf_{w \in X} \phi_{-}^{\downarrow\downarrow}(\bar{x}, h, w) = \nabla^2 f(\bar{x})(h, h) - \sigma(-\nabla f(\bar{x}), T_S^2(\bar{x}, h)).$$

If \bar{x} is a local minimizer of f over S , then

$$\nabla^2 f(\bar{x})(h, h) - \sigma(-\nabla f(\bar{x}), T_S^2(\bar{x}, h)) \geq 0, \forall h \text{ s.t. } \nabla f(\bar{x})h \in T_S(\bar{x}). \quad (4.6)$$

Condition (4.6) hold irrespective of S being convex or not.

ϕ is out second order regular at \bar{x} iff the set S is outer second order regular at \bar{x} .

If S is outer second order regular at \bar{x} and \bar{x} satisfies the first order necessary optimality conditions, then the second order growth condition holds at \bar{x} iff

$$\nabla^2 f(\bar{x})(h, h) - \sigma(-\nabla f(\bar{x}), T_S^2(\bar{x}, h)) > 0, \forall h \neq 0 \text{ s.t. } \nabla f(\bar{x})h \in T_S(\bar{x}). \quad (4.7)$$

Suppose now that the set S is given in the form $S := F^{-1}(K)$, where $K \subset \mathbb{R}^m$ is a closed convex and $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a twice continuously differentiable mapping. Suppose further that Robinson's constraint qualification holds at \bar{x} . Then by the the chain rule (1.8) we have

$$T_S^2(\bar{x}, h) = \{w: DF(\bar{x})w + D^2F(\bar{x})(h, h) \in T_K^2(F(\bar{x}), DF(\bar{x})h)\}. \quad (4.8)$$

Combing (4.8), the second necessary condition (4.6) becomes for every $h \in C(\bar{x}) := \{h \in \mathbb{R}^n | DF(\bar{x})h \in T_K(F(\bar{x})), \nabla f(\bar{x})h = 0\}$,

$$\nabla^2 f(\bar{x})(h, h) + \nabla f(\bar{x})w \geq 0, \quad (4.9)$$

$$\text{subject to } DF(\bar{x})w + D^2F(\bar{x})(h, h) \in T_K^2(F(\bar{x}), DF(\bar{x})h). \quad (4.10)$$

By using duality approach, we can get the second order necessarily condition. For every $h \in C(x_0)$ and any convex set $\mathcal{T}(h) \subset T_K^2(G(x_0), DG(x_0)h)$, the following inequality holds:

$$\sup_{\lambda \in \Lambda(x_0)} \{D_{xx}^2 L(x_0, \lambda)(h, h) - \sigma(\lambda, \mathcal{T}(h))\} \geq 0. \quad (4.11)$$

We now consider the second order optimality condition for (P) for $\phi(x) = f(x) + g(F(x))$, where $g: \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is proper, l.s.c., convex and Lipschitz continuous around $\bar{y} = F(\bar{x})$ (with $g(\bar{y})$ finite) relative to its domain.

Theorem 4.1

Suppose that the metric subregularity constraint qualification (3.24) holds. Let $\bar{v} = -\nabla f(\bar{x}) \in \partial\psi(\bar{x})$. Suppose that for every $\lambda \in \Lambda(\bar{x}, \bar{v})$, g is parabolically epi-differentiable at \bar{y} in every direction $d \in K_g(\bar{y}, \lambda)$, and parabolically regular at $F(\bar{x})$ for λ .

(i) if \bar{x} is a local minimum of (P), then the second-order necessary condition

$$\max_{\lambda \in \Lambda(\bar{x}, \bar{v})} \{ \langle \nabla_{xx}^2 L(\bar{x}, \lambda) h, h \rangle + d^2 g(F(\bar{x}), \lambda)(DF(\bar{x})h) \} \geq 0 \quad (4.12)$$

holds for all $h \in C(\bar{x}) = \{h \in \mathbb{R}^n \mid \langle -\nabla f(\bar{x}), h \rangle = dg(F(\bar{x}))(DF(\bar{x})h)\}$;

(ii) the second-order condition

$$\max_{\lambda \in \Lambda(\bar{x})} \{ \langle \nabla_{xx}^2 L(\bar{x}, \lambda) h, h \rangle + d^2 g(F(\bar{x}), \lambda)(DF(\bar{x})h) \} > 0 \quad \text{for all } h \in C(\bar{x}) \setminus \{0\} \quad (4.13)$$

is equivalent to the second order growth condition for $\phi(x)$.

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4.2. The second order optimality condition for augmented Lagrangian function

4.2 The second order optimality condition for augmented Lagrangian function

We consider the augmented Lagrangian function for the following composite optimization problem:

$$(P) \quad \min_{x \in \mathbb{R}^n} \{f(x) + g(F(x))\}.$$

The Moreau envelope $e_\nu\phi(\cdot)$ of a convex function ϕ at $u \in \mathbb{R}^m$ for parameter $\nu > 0$, defined as

$$e_\nu\phi(u) = \inf_{w \in \mathbb{R}^m} \left\{ \phi(w) + \frac{1}{2\nu} \|u - w\|^2 \right\}, \quad (4.14)$$

is real-valued, convex and continuous, and the infimum in (4.14) is uniquely attained for every $u \in \mathbb{R}^m$. We denote its unique minimizer by $P_\nu\phi(u)$, i.e.,

$$P_\nu\phi(u) := \operatorname{argmin}_{w \in \mathbb{R}^m} \left\{ \phi(w) + \frac{1}{2\nu} \|u - w\|^2 \right\} = (I + \nu\partial\phi)^{-1}(u), \quad (4.15)$$

which is called the *proximal mapping* of ϕ . Moreover, it was well known that $e_\nu\phi(u)$ is differentiable, and its gradient

$$\nabla e_\nu\phi(u) = \nu^{-1}(u - P_\nu\phi(u)) \quad (4.16)$$

is ν^{-1} -Lipschitz continuous.

Following [14], the augmented Lagrangian for the problem (P) can be expressed by

$$\begin{aligned}
 \mathcal{L}(x, \lambda, \tau) &= f(x) + \inf_{y \in \mathbb{R}^m} \{g(F(x) + y) + \frac{\tau}{2} \|y\|^2 - \langle \lambda, y \rangle\} \\
 &= f(x) + \inf_{y \in \mathbb{R}^m} \{g(F(x) + y) + \frac{\tau}{2} \|\tau^{-1} \lambda - y\|^2\} - \frac{1}{2\tau} \|\lambda\|^2 \\
 &= f(x) + e_{\tau^{-1}} g(\tau^{-1} \lambda + F(x)) - \frac{1}{2\tau} \|\lambda\|^2. \quad (4.17)
 \end{aligned}$$

Proposition 4.3

The set $\Lambda(\bar{x})$ of the Lagrange multipliers of the problem (P) at \bar{x} can be expressed as

$$\Lambda(\bar{x}) := \{\lambda \in \mathbb{R}^m : D_x \mathcal{L}(\bar{x}, \lambda, \tau) = 0 \text{ and } D_\lambda \mathcal{L}(\bar{x}, \lambda, \tau) = 0\},$$

for any $\tau > 0$.

In the following we assume that the function $g : \mathbb{R}^m \rightarrow \bar{\mathbb{R}}$ is proper lower semicontinuous convex and twice epi-differentiable at $F(\bar{x})$.

Lemma 4.1

Suppose that the set $\Lambda(\bar{x})$ of the Lagrange multipliers is nonempty and $\lambda \in \Lambda(\bar{x})$. Then, for any $\xi \in \mathbb{R}^n$, one has

$$d^2 e_{\tau^{-1}} g(\tau^{-1} \lambda + F(\cdot))(\bar{x}, DF(\bar{x})^* \lambda)(h) = 2d_{\tau^{-1}}(DF(\bar{x})h) + \langle \lambda, D^2 F(\bar{x})(h, h) \rangle, \quad (4.18)$$

where

$$d_{\tau^{-1}}(DF(\bar{x})h) := \inf_{y \in \mathbb{R}^m} \left\{ \frac{1}{2} d^2 g(F(\bar{x}), \lambda)(y) + \frac{\tau}{2} \|DF(\bar{x})h - y\|^2 \right\} \quad (4.19)$$

denotes the Moreau envelope of $\frac{1}{2} d^2 g(F(\bar{x}), \lambda)(\cdot)$ at $DF(\bar{x})h$.

Denote by $\ell_\tau(x) = \mathcal{L}(x, \lambda, \tau)$. From the proof of Proposition 5.1, we have $D\ell_\tau(\bar{x}) = D_x L(\bar{x}, \lambda) = 0$. By Proposition 2.10 in [11], the second-order epi-derivative of $\ell_\tau(x)$ at \bar{x} relative to $w = D\ell_\tau(\bar{x}) = 0$ can be expressed as

$$d^2(\ell_\tau)(\bar{x}, 0)(h) = D_{xx}^2 L(\bar{x}, \lambda)(h, h) + 2d_{\tau-1}(DF(\bar{x})h). \quad (4.20)$$

The necessary and sufficient conditions are stated as follows:

Theorem 4.2

Let $\lambda \in \Lambda(\bar{x})$. Then, $0 \in \hat{\partial}\ell_\tau(\bar{x})$ for any $\tau > 0$, and the following assertions hold:

(a) (Necessary condition). If $\ell_\tau(\cdot)$ has a local minimum at \bar{x} , then

$$D_{xx}^2 L(\bar{x}, \lambda)(h, h) + 2d_{\tau-1}(DF(\bar{x})h) \geq 0, \text{ for all } h \in \mathbb{R}^n. \quad (4.21)$$

(b) (Sufficient condition). If the condition

$$D_{xx}^2 L(\bar{x}, \lambda)(h, h) + 2d_{\tau-1}(DF(\bar{x})h) > 0, \text{ for all } h \in \mathbb{R}^n \setminus \{0\}, \quad (4.22)$$

holds, then $\ell_\tau(\cdot)$ has a local minimum at \bar{x} in the sense of quadratic growth condition.

Proposition 4.4

Let $\lambda \in \Lambda(\bar{x})$ be a Lagrange multiplier of the problem (P) at \bar{x} . Then (4.22) is equivalent to the following condition

$$\nabla_{xx}^2 L(\bar{x}, \lambda)(h, h) + d^2 g(F(\bar{x}), \lambda)(DF(\bar{x})h) > 0, \quad \forall h \in C(\bar{x}) \setminus \{0\} \quad (4.23)$$

for τ sufficiently large.

Proof.

For any $h \in C(\bar{x})$, by taking $y = DF(\bar{x})h$, we have that

$$d_{\tau-1}(DF(\bar{x})h) \leq \frac{1}{2}d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h).$$

Hence, the condition (4.22) implies that

$$\nabla_{xx}^2 L(\bar{x}, \lambda)(h, h) + d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h) > 0, \quad \forall h \in C(\bar{x}) \setminus \{0\}. \quad (4.24)$$

Suppose that (4.22) doesn't hold. Then, $\exists \tau_k \rightarrow +\infty$ as $k \rightarrow \infty, \exists h_k \in \mathbb{R}^n$ with $\|h_k\| = 1$ such that

$$\nabla_{xx}^2 L(\bar{x}, \lambda)(h_k, h_k) + 2d_{\tau_k-1}(DF(\bar{x})h_k) \leq 0.$$

Observe that $d_{\tau_k}^{-1}(\cdot)$ defined in (4.19) is the Moreau envelope of $\frac{1}{2}d^2g(F(\bar{x}), \lambda)(\cdot)$. Since $\frac{1}{2}d^2g(F(\bar{x}), \lambda)(\cdot)$ is proper l.s.c. convex, the infimum of $d_{\tau_k}^{-1}(DF(\bar{x})h_k)$ can be attained uniquely at $\bar{z}_k := \text{Prox}_{\tau_k}^{-1}(\frac{1}{2}d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h_k)) \in \text{dom}\frac{1}{2}d^2g(F(\bar{x}), \lambda)(\cdot)$. It follows that

$$\nabla_{xx}^2 L(\bar{x}, \lambda)(h_k, h_k) + \tau_k \|DF(\bar{x})h_k - \bar{z}_k\|^2 + \frac{1}{2}d^2g(F(\bar{x}), \lambda)(\bar{z}_k) \leq 0. \quad (4.25)$$

Since $\{\nabla_{xx}^2 L(\bar{x}, \lambda)(h_k, h_k)\}$ is bounded, the term $d^2g(F(\bar{x}), \lambda)(\bar{z}_k)$ is nonnegative and $\tau_k \rightarrow +\infty$ as $k \rightarrow +\infty$, it follows from (4.25) that

$$\lim_{k \rightarrow +\infty} \|DF(\bar{x})h_k - \bar{z}_k\| = 0. \quad (4.26)$$

Because $\|h_k\| = 1$, by passing to a subsequence if necessary, we assume that $h_k \rightarrow h$. Hence, we have $\|h\| = 1$ and by (4.26), $\bar{z}_k \rightarrow DF(\bar{x})h$ as $k \rightarrow +\infty$. The lower semicontinuity of $\frac{1}{2}d^2g(F(\bar{x}), \lambda)(\cdot)$ implies that $DF(\bar{x})h \in \text{dom}d^2g(F(\bar{x}), \lambda)(\cdot)$, and so that

$$dg(F(\bar{x})(DF(\bar{x})h) = \langle \lambda, DF(\bar{x})h \rangle = \langle -\nabla f(\bar{x}), h \rangle.$$

This implies that $h \in C(\bar{x})$. Again, by the lower semicontinuity of $\frac{1}{2}d^2g(F(\bar{x}), \lambda)(\cdot)$, we have from (4.25) that





$$\nabla_{xx}^2 L(\bar{x}, \lambda)(h, h) + d^2g(F(\bar{x}), \lambda)(DF(\bar{x})h) \leq 0,$$

which contradicts (4.24). □





Papers for algorithm for composite optimization problems

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Thank you very much for your attention!