Bilevel programs: directional constraint qualifications and necessary optimality conditions

Jane Ye

University of Victoria, Canada

Lecture 3 at the Forum on Developments and Origins of Operations Research

November 26, 2021 Organizers: The Mathematical Programming Branch of OR Society of China & Southern University of Science and Technology Directional optimality conditions Application to bilevel programming

Outline







Bilevel program

$$\begin{array}{ll} (\mathsf{BP}) & \min_{x,y} & F(x,y) \\ & s.t. & y \in S(x), \end{array}$$

where S(x) denotes the solution set of the lower level program

$$(P_x) \quad \min_y f(x,y) \quad \text{s.t. } g(x,y) \leq 0,$$

where $F, f : \mathbb{R}^n \times \mathbb{R}^m \to \mathbb{R}$, $g : \mathbb{R}^n \times \mathbb{R}^m \to \mathbb{R}^p$ are continuously differentiable.

・ 同 ト ・ ヨ ト ・ ヨ ト

Recall the following reformulations using the value function $V(x) := \inf_{v} \{f(x, y) : g(x, y) \le 0\}$:

$$\begin{array}{ll} (VP) & \min_{x,y} & F(x,y) \\ & s.t. & f(x,y) - V(x) \leq 0, \\ & g(x,y) \leq 0, \end{array}$$

(CP)
$$\min_{\substack{x,y,u}} F(x,y)$$

s.t. $f(x,y) - V(x) \le 0$,
 $\nabla_y f(x,y) + \nabla_y g(x,y)^T u = 0$,
 $g(x,y) \le 0, \ u \ge 0, \ u^T g(x,y) = 0$.

æ

If the value function is Lipschitz continuous, then (VP) and (CP) can be written as the nonsmooth optimization problem:

 $(P) \quad \min \quad f(x)$ $s.t. \quad g(x) \leq 0,$ where $f : \mathbb{R}^n \to \mathbb{R}$ is smooth and $g : \mathbb{R}^n \to \mathbb{R}^m$ is Lipschitz continuous.

Regular and Limiting Subdifferentials

• For a function $\varphi : \mathbb{R}^n \to [-\infty, \infty]$ and a point $\bar{x} \in \mathbb{R}^n$ where $\varphi(\bar{x})$ is finite, the regular (or Fréchet) subdifferential of φ at \bar{x} :

$$\widehat{\partial}\varphi(\bar{x}) := \left\{ \xi : \varphi(x) \ge \varphi(\bar{x}) + \xi^{\mathsf{T}}(x - \bar{x}) + o(||x - \bar{x}||) \ \forall x \right\}.$$

• The regular subdifferential may be empty. The limiting/Mordukhovich subdifferential of φ at \bar{x} defined by

$$\partial \varphi(\bar{x}) := \left\{ \lim_{k} \xi^{k} | x^{k} \to \bar{x}, \varphi(x^{k}) \to \varphi(\bar{x}), \ \xi^{k} \in \widehat{\partial} \varphi(x^{k}) \right\}$$

is more robust.

• If φ is Lipschitz continuous around \bar{x} , the Clarke subdifferential

$$\partial^{\mathsf{c}}\varphi(\bar{x}) = \mathrm{co}\varphi(\bar{x}).$$

Theorem (Nonsmooth necessary optimality condition)

If \bar{x} is a local minimum for (P), then FJ condition holds: $\exists (\lambda_0, \lambda) \neq 0$ such that

 $0 \in \lambda_0 \nabla f(\bar{x}) + \partial g(\bar{x})^T \lambda, \\ \lambda \perp g(\bar{x}), \quad \lambda_0, \lambda \ge 0.$

If there is no nonzero abnormal multipliers (nonsmooth MFCQ):

$$\begin{cases} 0 \in \frac{\partial g(\bar{x})^T \lambda}{0 \le \lambda \perp g(\bar{x})}, \implies \lambda = 0 \end{cases}$$

then KKT condition holds: there exists $\lambda \ge 0$ such that

 $0 \in
abla f(ar{x}) + \partial g(ar{x})^T \lambda, \ \lambda \perp g(ar{x}).$

Jane Ye

7/41

・ 同 ト ・ ヨ ト ・ ヨ ト

э

But the nonsmooth MFCQ/NNAMCQ never hold for problem (VP) and (CP)! So we need weaker constraint qualifications.

Directional optimality conditions Application to bilevel programming

Clarke calmness condition

Definition (Clarke)

Suppose that \bar{x} be a local optimal solution of (P). We say (P) is Clarke calm at \bar{x} if \bar{x} also solves the penalized problem

> min $f(x) + \rho ||g_+(x)||$ s.t. $x \in \overline{x} + \mathbb{B}_{\epsilon}$

for some $\rho, \epsilon \geq 0$.

• Under the Clarke calmness condition, a local optimal solution must satisfy the KKT condition.

9/41

Definition (Metric subregularity/local error bound/calmness)

The set-valued map $G(x) := -g(x) + \mathbb{R}^m_-$ is metrically subregular at $(\bar{x}, 0) \in gphG$ if there are positive reals $\epsilon > 0$ and $\kappa > 0$ such that

$$\mathrm{dist}(x, G^{-1}(0)) \leq \kappa \|g_+(x)\| \quad orall x \in ar{x} + \mathbb{B}_\epsilon,$$

where $G^{-1}(0) := \{x | g(x) \le 0\}.$

• The set-valued map $G(x) := -g(x) + \mathbb{R}^m_-$ is metrically subregular at $(\bar{x}, 0)$ if and only if the perturbed feasible map $\mathcal{F}(v) := \{x | g(x) \le v\}$ is calm at $(0, \bar{x}) \in \mathrm{gph}\mathcal{F}$.

< 同 > < 三 > < 三 > -

Directional optimality conditions Application to bilevel programming

Sufficient conditions for metric subregularity

Definition (Quasi-normality; Guo, JY and Zhang, 2013)

Let $g(\bar{x}) \leq 0$. g is Lipschitz continuous. We say the quasi-normality holds at \bar{x} if

$$\begin{array}{ll} 0 \in \partial g(\bar{x})^T \lambda, & 0 \leq \lambda \perp g(\bar{x}), \\ \exists x^k \to \bar{x} \text{ s.t. } g_i(x^k) > 0 \text{ if } \lambda_i > 0 \end{array} \implies \quad \lambda = 0$$

Constraint qualifications and their relationships

Nonsmooth MFCQ/NNAMCQ

- \implies quasi-normality
- \implies metric subregularity/calmness
- \implies Clarke calmness
- \implies KKT condition for a local minimum.

< 1 →

Constraint qualifications for (VP) or (CP) Since Nonsmooth MFCQ/NNAMCQ does not hold for (VP) and (CP), we have

quasi-normality

- \implies metric subregularity/calmness
- \implies Clarke calmness
- \implies KKT condition for a local minimum.

KKT condition for (VP)

Theorem (JY and Zhu 1995)

Let (\bar{x}, \bar{y}) be a local minimizer of (BP). Suppose a constraint qualification for problem (VP) holds at (\bar{x}, \bar{y}) , MFCQ holds at each $y \in S(\bar{x})$ and the feasible region for lower level problem is uniformly bounded. Then there exist $\lambda^i \geq 0, \sum_{i=1}^{n+1} \lambda^i = 1, y^i \in S(\bar{x}), (\lambda, \mu, \mu_g^i) \in \mathbb{R}^{1+p+p}_+$ such that

$$0 = \nabla_{x} F(\bar{x}, \bar{y}) + \lambda \left(\nabla_{x} f(\bar{x}, \bar{y}) - \sum_{i=1}^{n+1} \lambda^{i} (\nabla_{x} f(\bar{x}, y^{i}) + \nabla_{x} g(\bar{x}, y^{i})^{T} \mu_{g}^{i}) \right) + \nabla_{x} g(\bar{x}, \bar{y})^{T} \mu, 0 = \nabla_{y} F(\bar{x}, \bar{y}) + \lambda \nabla_{y} f(\bar{x}, \bar{y}) + \nabla_{y} g(\bar{x}, \bar{y})^{T} \mu, 0 = \nabla_{y} f(\bar{x}, y^{i}) + \nabla_{y} g(\bar{x}, y^{i})^{T} \mu_{g}^{i}, \quad \mu_{g}^{i} \perp g(\bar{x}, y^{i}), \mu \perp g(\bar{x}, \bar{y}).$$

Proof

There exist $\lambda \ge 0$ and $\mu \ge 0$ such that

$$0 \in \nabla F(\bar{x}, \bar{y}) + \lambda (\nabla f(\bar{x}, \bar{y}) - \partial^{c} V(\bar{x}) \times \{0\}) + \nabla g(\bar{x}, \bar{y})^{T} \mu,$$

$$\mu \perp g(\bar{x}, \bar{y}).$$

Under the assumptions, V is Lipschitz continuous and

 $\partial^{c} V(\bar{x}) \subseteq co\{\nabla_{x} f(\bar{x}, y) + \nabla_{x} g(\bar{x}, y)^{T} \lambda : y \in S(\bar{x}), \lambda \in \mathrm{KT}(\bar{x}, y)\}.$ Hence there exist $\lambda^{i} \geq 0, \sum_{i=1}^{n+1} \lambda^{i} = 1, y^{i} \in S(\bar{x}), \mu^{i}_{g} \in KT(\bar{x}, y^{i})$ such that

$$\xi \in \partial^{c} V(\bar{x}) \Rightarrow \xi = \sum_{i=1}^{n+1} \lambda^{i} (\nabla_{x} f(\bar{x}, y^{i}) + \nabla_{x} g(\bar{x}, y^{i})^{T} \mu_{g}^{i}).$$

・ 同 ト ・ ヨ ト ・ ヨ ト …

3

Definition (Inner Semicontinuity)

Given $\bar{y} \in S(\bar{x})$, we say that the set-valued map $S(x) : \mathbb{R}^n \Longrightarrow \mathbb{R}^m$ is inner semicontinuous at (\bar{x}, \bar{y}) if for any sequences $x^k \to \bar{x}$, there exists a sequence $y^k \in S(x^k)$ converging to \bar{y} .

If S(x) is single valued, then the concept of inner semicontinuity reduced to the continuity of the function.

KKT condition for (VP), inner semicontinuous case

Theorem (cf. Theorem 6.21 of Mordukhovich 2018)

Let (\bar{x}, \bar{y}) be a local minimizer of (BP). Suppose that S(x) is inner semicontinuous at (\bar{x}, \bar{y}) , a constraint qualification for (VP) holds at (\bar{x}, \bar{y}) , and MFCQ for problem $(P_{\bar{x}})$ holds at \bar{y} . Then there exists a vector $(\lambda, \mu, \mu_g) \in \mathbb{R}^{1+p+p}_+$ satisfying

$$0 = \nabla_{x} F(\bar{x}, \bar{y}) + \nabla_{x} g(\bar{x}, \bar{y})^{T} (\mu - \lambda \mu_{g}),$$

$$0 = \nabla_{y} F(\bar{x}, \bar{y}) + \lambda \nabla_{y} f(\bar{x}, \bar{y}) + \nabla_{y} g(\bar{x}, \bar{y})^{T} \mu, \quad \mu \perp g(\bar{x}, \bar{y})$$

$$0 = \nabla_{y} f(\bar{x}, \bar{y}) + \nabla_{y} g(\bar{x}, \bar{y})^{T} \mu_{g}, \mu_{g} \perp g(\bar{x}, \bar{y}),$$

Proof. There exist $\lambda \ge 0, \mu \ge 0$ such that KKT condition holds:

$$0 \in \nabla F(\bar{x}, \bar{y}) + \lambda (\nabla f(\bar{x}, \bar{y}) - \partial^{c} V(\bar{x}) \times \{0\}) + \nabla g(\bar{x}, \bar{y})^{T} \mu,$$

$$\mu \perp g(\bar{x}, \bar{y}).$$

$$\partial^{c} V(\bar{x}) \subseteq \left\{ \nabla_{x} f(\bar{x}, \bar{y}) + \nabla_{x} g(\bar{x}, \bar{y})^{T} \mu_{g} : \mu_{g} \in \mathrm{KT}(\bar{x}, \bar{y}) \right\}.$$

Reall that if there is no nonzero abnormal multipliers:

$$\left\{ \begin{array}{ll} 0 \in \frac{\partial g(\bar{x})^T \lambda,}{0 \leq \lambda \perp g(\bar{x})}, & \Longrightarrow & \lambda = 0 \end{array} \right.$$

then there exists $\lambda \ge 0$ such that

 $0 \in \nabla f(\bar{x}) + \partial g(\bar{x})^T \lambda,$ $\lambda \perp g(\bar{x}).$

If $\partial g(\bar{x})$ can be replaced by a smaller set, then the NNAMCQ becomes weaker and KKT condition becomes sharper. This motivates us to consider directional constraint qualification and directional KKT condition.

Directional optimality conditions Application to bilevel programming

 Consider problem (P) : min f(x) s.t. g(x) ≤ 0, where f is smooth, g = (g₁,...,g_m) and g_i is directionally differentiable and Lipschitz continuous at x̄.

• Let
$$\bar{I}_g := \{i | g_i(\bar{x}) = 0\}.$$

• Define $g' := (g'_1, \ldots, g'_m)$ and $g'_i(\bar{x}; d) = 0$ if $i \notin \bar{I}_g$.

The linearized cone:

$$L(\bar{x}) := \{ d | g'(\bar{x}; d) \le 0 \}.$$

The critical cone:

$$C(\bar{x}) = \{ d \in L(\bar{x}) | \nabla f(\bar{x})^T d \leq 0 \}.$$

Note that if $C(\bar{x}) = \{0\}$, then \bar{x} is already a local minimum. Otherwise if $C(\bar{x}) \neq \{0\}$, then one can use a nonzero critical direction to obtain a sharper optimality condition.

Directional Clarke subdifferentials

• $x^k \xrightarrow{d} \bar{x} \iff$ there exist $t_k \downarrow 0, d^k \rightarrow d$ such that $x^k = \bar{x} + t_k d^k$.

Definition (Ginchev and Mordukhovich 2011)

Let $\varphi : \mathbb{R}^n \to \overline{\mathbb{R}}$ and $\varphi(\overline{x})$ is finite. The limiting subdifferential of φ at \overline{x} in direction $d \in \mathbb{R}^n$ is defined as

$$\partial \varphi(\bar{x}; d) := \left\{ \lim_{k} \xi^{k} | \exists x^{k} \xrightarrow{d} \bar{x}, \varphi(x^{k}) \to \varphi(\bar{x}), \xi^{k} \in \widehat{\partial} \varphi(x^{k}) \right\}$$

• Let $\varphi(x) = |x|$. The directional limiting subdifferential at 0 is

$$\partial \varphi(\mathbf{0}; \mathbf{1}) = \{\mathbf{1}\}, \quad \partial \varphi(\mathbf{0}; -\mathbf{1}) = \{-1\}.$$

• But the limiting subdifferential $\partial \varphi(0) = [-1, 1]$. When φ is Lipschitz continuous, we define the directional Clarke subdifferential as $\partial^c \varphi(\bar{x}; d) := \operatorname{co} \partial \varphi(\bar{x}; d)$. Directional optimality conditions Application to bilevel programming

Directional optimality conditions, Gfrerer 2013

 Suppose x̄ is a local optimal solution of (P). Let d ∈ C(x̄). Then there exists nonnegative numbers λ₀, λ not all equal to zero such that Fritz John type directional optimality condition holds at x̄ in direction d:

$$0 \in \lambda_0 \nabla f(\bar{x}) + \partial g(\bar{x}; d)^T \lambda, \\ 0 \le \lambda \perp g(\bar{x}), \quad \lambda \perp g'(\bar{x}; d).$$

If

$$\begin{cases} 0 \in \partial g(\bar{x}; d)^{\mathsf{T}} \lambda, \\ 0 \leq \lambda \perp g(\bar{x}), \quad \lambda \perp g'(\bar{x}; d) \end{cases} \implies \lambda = 0$$

then the KKT type directional optimality condition holds.

• We call the above condition the first order sufficient condition for metric subregularity (or the directional MFCQ).

Directional neighborhood

$$\mathcal{V}_{arepsilon,\delta}(d) = \left\{ egin{array}{c} \mathbb{B}_arepsilon(0), & d=0, \ \{0\}\cup\{z\in\mathbb{B}_arepsilon(0)|ig\|rac{z}{\|z\|}-rac{d}{\|d\|}ig\|\leq\delta\}, \ d
eq 0. \end{array}
ight.$$



▶ ★ 臣 ▶ 22 / 41

э

Directional optimality conditions Application to bilevel programming

Directional Lipschitz continuity

Definition

We say that a single-valued map $\varphi : \mathbb{R}^n \to \mathbb{R}^m$ is Lipschitz continuous at \bar{x} in direction d if there exists L > 0 and a directional neighborhood $\mathcal{V}_{\varepsilon,\delta}(d)$ of d such that

$$\|\varphi(x) - \varphi(x')\| \le L \|x - x'\| \quad \forall x, x' \in \bar{x} + \mathcal{V}_{\varepsilon,\delta}(d).$$

• When d = 0, the directional Lipschitz continuity reduces to the classical Lipschitz continuity.

Directional Clarke calmness condition

Definition

Suppose \bar{x} is a local optimal solution of (P). We say that (P) is (Clarke) calm at \bar{x} in direction d if \bar{x} also solves

min $f(x) + \rho ||g_+(x)||$ s.t. $x \in \overline{x} + \mathcal{V}_{\epsilon,\delta}(d)$,

for some $\rho, \epsilon, \delta \geq 0$.

• When d = 0, the directional calmness reduces to the calmness introduced by Clarke. Directional Clarke calmness with $d \neq 0$ is weaker than the Clarke calmness.

▲ □ ▶ ▲ □ ▶ ▲ □ ▶

24/41

Directional KKT condition under directional calmness

Theorem (Bai and JY 2021)

Let \bar{x} be a local minimizer of (P). Suppose f(x) is continuously differentiable at \bar{x} and g(x) is directionally Lipschitz and directionally differentiable at \bar{x} in direction $d \in C(\bar{x})$. Suppose that (P) is Clarke calm at \bar{x} in direction d. Then there exists a vector $\lambda \in \mathbb{R}^m$ such that the directional KKT condition holds at \bar{x} in direction d:

$$0 \in \nabla f(\bar{x}) + \partial g(\bar{x}; d)^{\mathsf{T}} \lambda$$
$$0 \leq \lambda \perp g(\bar{x}), \ \lambda \perp g'(\bar{x}; d)$$

・ 同 ト ・ ヨ ト ・ ヨ ト

25/41

• When $d \neq 0$, directional KKT condition is sharper than the (nondirectional) KKT condition.

Definition (Directional metric subregularity; Gfrerer, 2013)

Given $d \in \mathbb{R}^n$. The set-valued map $G(x) := -g(x) + \mathbb{R}^m_-$ is metrically subregular at $(\bar{x}, 0) \in gphG$ in direction d if there are positive reals $\epsilon > 0, \delta > 0$, and $\kappa > 0$ such that

 $\operatorname{dist}(x, G^{-1}(0)) \leq \kappa \|g_+(x)\| \quad \forall x \in \bar{x} + \mathcal{V}_{\epsilon,\delta}(d),$

where $G^{-1}(0) := \{x | g(x) \le 0\}.$

• When d = 0, the set-valued map $G(x) := -g(x) + \mathbb{R}^m_-$ is metrically subregular at $(\bar{x}, 0)$ or the metric subregularity constraint qualification (MSCQ) holds at \bar{x} or the perturbed set-valued map $M(v) := \{x | g(x) \le v\}$ is calm at $(0, \bar{x}) \in gphM$.

・ 同 ト ・ ヨ ト ・ ヨ ト …

Directional FOSCMS/MFCQ

Definition (Directional first order sufficient condition for metric subregularity (FOSCMS) by Gfrerer)

Let $g(\bar{x}) \leq 0$ and $0 \neq d \in L(\bar{x})$. Let g be directionally Lipschitz continuous and directionally differentiable at \bar{x} in direction d. We say FOSCMS/directional MFCQ at \bar{x} in direction d if

 $0 \in \partial g(\bar{x}; d)^{\mathsf{T}} \lambda, 0 \leq \lambda \perp g(\bar{x}), \lambda \perp g'(\bar{x}; d) \implies \lambda = 0$

Directional optimality conditions Application to bilevel programming

Directional guasi-normality

Definition (Directional quasi-normality; Bai, Ye and Zhang, 2019)

Let $g(\bar{x}) \leq 0$. Let g be directionally Lipschitz continuous and directionally differentiable. We say the directional quasi-normality holds at \bar{x} in direction $0 \neq d \in L(\bar{x})$ if

 $\begin{array}{ll} 0 \in \partial g(\bar{x};d)^T \lambda, & 0 \leq \lambda \perp g(\bar{x}), \quad \lambda \perp g'(\bar{x};d) \\ \exists x^k \to \bar{x} \text{ s.t. } g_i(x^k) > 0 \text{ if } \lambda_i > 0 \end{array} \implies \lambda = 0$

(日本) (日本)

28/41

Directional constraint qualifications and their relationships

FOSCMS/MFCQ in direction d

- \implies quasi-normality in direction d
- \implies metric subregularity in direction d
- \implies Clarke calmness in direction d
- \implies KKT condition in direction d for a local minimum.

Directional KKT condition for bilevel program

Theorem (Bai and JY 2021)

Let (\bar{x}, \bar{y}) be a local minimizer of (BP). Suppose the value function V(x) is Lipschitz continuous and directionally differentiable at \bar{x} in direction u and $(u, v) \in C(\bar{x}, \bar{y})$.

$$L(\bar{x},\bar{y}) := \{(u,v) | \nabla f(\bar{x},\bar{y})(u,v) \leq V'(\bar{x};u), \nabla g(\bar{x},\bar{y})(u,v) \leq 0\}$$

$$C(\bar{x},\bar{y}) := \{(u,v) \in L(\bar{x},\bar{y}) | F(\bar{x},\bar{y})(u,v) \leq 0\}$$

Suppose that (VP) is calm at (\bar{x}, \bar{y}) in direction (u, v). Then there exists $(\lambda, \mu) \ge 0$ such that

$$0 \in \nabla F(\bar{x}, \bar{y}) + \lambda \left(\nabla f(\bar{x}, \bar{y}) - \partial^{c} V(\bar{x}; u) \times \{0\} \right) \\ + \nabla g(\bar{x}, \bar{y})^{T} \mu, \\ \mu \perp g(\bar{x}, \bar{y}), \ \mu \perp \nabla g(\bar{x}, \bar{y})(u, v).$$

▲ □ ▶ ▲ □ ▶ ▲ □ ▶

Difficulties

Problem (VP) is a constrained optimization problem with the difficult constraint

$$\varphi(x,y) := f(x,y) - V(x) \leq 0.$$

There are two important issues.

- Which directional constraint qualification is applicable for bilevel programs?
- How to calculate or give the upper estimate for the directional Clarke subdifferential $\partial^c V(\bar{x}; u)$ and the directional derivative $V'(\bar{x}; u)$?

Recall that

MFCQ/NNAMCQ

- \implies MFCQ in direction d
- \implies quasi-normality in direction d
- \implies Metric subregularity in direction d
- \implies calmness in direction d.

Question We know that MFCQ/NNAMCQ fails for (VP). Directional MFCQ is weaker. Does directional MFCQ hold for (VP)?

We will show that the answer is "No"!

Directional optimality conditions Application to bilevel programming

Directional FOSCMS/MFCQ fails for VP

Proposition (Ye and Zhu, 1995)

The nonsmooth MFCQ fails at any feasible point of (VP).

Proposition (Bai and JY 2021)

FOSCMS/MFCQ fails at any feasible solution of (VP) in any critical direction.

・ 同 ト ・ ヨ ト ・ ヨ ト

Denote the feasible mapping by $\mathcal{F}(x) := \{y \in \mathbb{R}^m | g(x, y) \le 0\}$ and the active index set $I_g(x, y) := \{i = 1, \dots, p | g_i(x, y) = 0\}$.

Definition (RCR regularity; Minchenko and Stakhovski, 2011)

We say that the set-valued map $\mathcal{F}(x)$ is relaxed constant rank (RCR) regular at $(\bar{x}, \bar{y}) \in \operatorname{gph}\mathcal{F}$ if there exists $\delta > 0$ such that for any index subset $K \subseteq I_g(\bar{x}, \bar{y})$, the family of gradient vectors $\nabla_y g_j(x, y), j \in K$, has the same rank for all $(x, y) \in \mathbb{B}_{\delta}(\bar{x}, \bar{y})$.

• RCR regularity is weaker then MFCQ.

・ 同 ト ・ ヨ ト ・ ヨ ト …

Definition (Directional Inner Semicontinuity)

Given $\bar{y} \in S(\bar{x})$, we say that the set-valued map $S(x) : \mathbb{R}^n \Longrightarrow \mathbb{R}^m$ is inner semicontinuous at (\bar{x}, \bar{y}) in direction u, if for any sequences $x^k \xrightarrow{u} \bar{x}$, there exists a sequence $y^k \in S(x^k)$ converging to \bar{y} .

Formula for directional derivative, directional inner semicontinuous case

Theorem (Bai and JY 2021)

Suppose that there exists $\bar{y} \in S(\bar{x})$ such that S(x) is inner semicontinuous at (\bar{x}, \bar{y}) in direction u. Moreover assume that $\mathcal{F}(x)$ is RCR-regular at (\bar{x}, \bar{y}) . Then the value function V(x) is directionally differentiable at \bar{x} in direction u and

$$V'(\bar{x}; u) = \max_{\lambda \in \mathrm{KT}(\bar{x}, \bar{y})} \nabla_{x} \mathcal{L}(\bar{x}, \bar{y}; \lambda) u,$$

where $\mathcal{L}(\bar{x}, \bar{y}; \lambda) := f(\bar{x}, \bar{y}) + \langle g(\bar{x}, \bar{y}), \lambda \rangle$ is the Lagrange function.

Estimate for $\partial^c V(\bar{x}; u)$, directional inner semicontinuous

case

Theorem (Bai and JY 2021)

Suppose that there exists $\bar{y} \in S(\bar{x})$ such that S(x) is inner semicontinuous at (\bar{x}, \bar{y}) in direction u and the set-valued map $\mathcal{F}(x) := \{y | g(x, y) \leq 0\}$ is RCR-regular at (\bar{x}, \bar{y}) , then V(x) is Lipschitz at \bar{x} in direction u. Moreover

$$\partial^{c} V(\bar{x}; u) \subseteq \left\{ \nabla_{x} \mathcal{L}(\bar{x}, \bar{y}; \lambda) | \lambda \in \mathrm{KT}(\bar{x}, \bar{y}) \cap \{ \nabla g(\bar{x}, \bar{y})(u, v) \}^{\perp} \right\},\$$

where v satisfies $\nabla g(\bar{x}, \bar{y})(u, v) \leq 0, V'(\bar{x}; u) = \nabla f(\bar{x}, \bar{y})(u, v).$

▲ □ ▶ ▲ □ ▶ ▲ □ ▶

Directional KKT condition for (VP), directional IS case

Theorem (Bai and JY 2021)

Let (\bar{x}, \bar{y}) be a local minimizer of (BP). Suppose $\mathcal{F}(x)$ is RCR-regular at (\bar{x}, \bar{y}) and S(x) is inner semicontinuous at (\bar{x}, \bar{y}) in direction u. Suppose that there exists v such that the directional quasi-normality holds at (\bar{x}, \bar{y}) in direction $(u, v) \in C(\bar{x}, \bar{y})$. Then there exists a vector $(\lambda, \mu, \mu_g) \in \mathbb{R}^{1+p+p}_+$ satisfying

$$0 = \nabla_{x} F(\bar{x}, \bar{y}) + \nabla_{x} g(\bar{x}, \bar{y})^{T} (\mu - \lambda \mu_{g}),$$

$$0 = \nabla_{y} F(\bar{x}, \bar{y}) + \lambda \nabla_{y} f(\bar{x}, \bar{y}) + \nabla_{y} g(\bar{x}, \bar{y})^{T} \mu,$$

$$0 = \nabla_{y} f(\bar{x}, \bar{y}) + \nabla_{y} g(\bar{x}, \bar{y})^{T} \mu_{g}, \mu_{g} \perp g(\bar{x}, \bar{y}), \mu_{g} \perp \nabla g(\bar{x}, \bar{y})(u, v),$$

$$\mu \perp g(\bar{x}, \bar{y}), \ \mu \perp \nabla g(\bar{x}, \bar{y})(u, v).$$

・ 同 ト ・ ヨ ト ・ ヨ ト

Example

(BP) min
$$F(x, y) = (\sqrt{3}x - y - \sqrt{3})^2 + x + \sqrt{3}y + 3$$

s.t. y minimizes $f(x, y) = -(x - y)^2 + 1$,
s.t. $g_1(x, y) = (x - 1)^2 + y^2 - 4 \le 0$,
 $g_2(x, y) = -\sqrt{3}x - y - \sqrt{3} \le 0$.

In this example, we have shown that

- *S*(*x*) is not inner semi-continuous but it is inner semi-continuous in a critical direction.
- The quasi-normality holds in a critical direction but the classical quasi-normality fails.
- The directional KKT condition holds.

Reference

- JY AND D.L. ZHU 1995, *Optimality conditions for bilevel* programming problems, Optimization **33**, 9-27.
- B. S. MORDUKHOVICH 2018, Variational Analysis and Applications, Springer 2018.
- K. BAI, J. J. YE AND J. ZHANG 2019, Directional quasi-/pseudo-normality as sufficient conditions for metric subregularity, SIAM J. Optim., 29(2019), pp. 2625-2649.
- K. BAI AND J. J. YE 2021, *Directional necessary optimality* conditions for bilevel programs, to appear in Math. Oper. Res., arXiv (2007.07933).

- Thank You !-

・部・・モト・モン

æ