Quantitative stability analysis of stochastic generalized equations

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Introduction

We consider stochastic generalized equations (SGE) of the form

 $0 \in \mathbb{E}_P[\Gamma(x,\xi)] + \mathcal{G}(x),$

where $\Gamma :: \mathcal{X} \times \Xi \to 2^{\mathcal{Y}}$ and $\mathcal{G} : \mathcal{X} \to 2^{\mathcal{Y}}$ are closed set-valued mappings, \mathcal{X} and \mathcal{Y} are subsets of Banach spaces X and Y (with norm $\|\cdot\|_X$ and $\|\cdot\|_Y$) respectively, $\xi : \Omega \to \Xi$ is a random vector defined on a probability space $(\Omega, \mathbb{F}, \mathbb{P})$ with support set $\Xi \in \mathbb{R}^d$ and probability distribution P, and $\mathbb{E}_P[\cdot]$ denotes Aumann's setvalued integral with respect to P, i.e.,

$$\begin{split} \mathbb{E}_P[\Gamma(x,\xi)] &:= \int_{\Xi} \Gamma(x,\xi) P(d\xi) \\ &= \left\{ \int_{\Xi} \gamma_x(\xi) P(d\xi) : \gamma_x \text{ is integrable selection of } \Gamma(x,\cdot) \right\} \end{split}$$

If X is separable and P non-atomic, Γ closed-valued and integrably bounded, then $E_P[\Gamma(x,\xi)]$ is convex.

Stochastic generalized equations were first studied by Ralph-Xu in MOR 2011.

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Aim: Framework for stability analysis of stochastic variational problems

Contents:

- (1) Prerequisites about support functions
- (2) Quantitative stability analysis of SGEs
- (3) Stability of linear two-stage stochastic programs (revisited)
- (4) Stability of two-stage stochastic programs with complementarity constraints
- (5) Stability of stochastic programs with second order dominance constraints
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Prerequisites about support functions

Lemma:

Let C, D be nonempty compact and convex subsets of a Banach space E with support functions $\sigma(\cdot, C)$ and $\sigma(\cdot, D)$ given on the dual E^* , i.e., $\sigma(u, C) = \sup_{x \in C} \langle u, x \rangle$. Then it holds for the excess

$$\mathbb{D}(C,D) := \sup_{x \in C} d(x,D) = \max_{\|u\|_* \leq 1} (\sigma(u,C) - \sigma(u,D))$$

and for the Pompeiu-Hausdorff distance

$$\mathbb{H}(C,D) := \max\{\sup_{x \in C} d(x,D), \sup_{x \in D} d(x,C)\}$$
$$= \max_{\|u\|_* \le 1} |\sigma(u,C) - \sigma(u,D)|.$$

Lemma:

If $\Gamma:\Xi\to E$ is closed convex-valued and integrably bounded, then it holds for all $u\in E^*$

 $\mathbb{E}_P[\sigma(u,\Gamma(\xi))] = \sigma(u,\mathbb{E}_P[\Gamma(\xi)]).$

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Main quantitative stability result

We consider the stochastic generalized equation

(SGE(P)) $0 \in \mathbb{E}_P[\Gamma(x,\xi)] + \mathcal{G}(x)$

and its perturbation

(SGE(Q)) $0 \in \mathbb{E}_Q[\Gamma(x,\xi)] + \mathcal{G}(x).$

for probability measures P and Q on Ξ .

We consider the following "distance" of probability measures

$$\mathscr{D}(Q,P) := \sup_{g \in \mathscr{F}} \left(\mathbb{E}_Q[g(\xi)] - \mathbb{E}_P[g(\xi)] \right)$$

where $\mathscr{F} := \{g : g(\xi) := \sigma(u, \Gamma(x, \xi)), \text{ for } x \in \mathcal{X}, \|u\|_* \leq 1\}.$

Note that $\mathscr{D}(Q, P)$ may be bounded by the ζ -metric

$$\zeta_{\mathscr{F}}(Q,P) := \sup_{g \in \mathscr{F}} \left| \mathbb{E}_Q[g(\xi)] - \mathbb{E}_P[g(\xi)] \right|.$$



Theorem:

Let \mathcal{X} be a compact subset of X and S(P) and S(Q) denote the solution sets of (SGE(P)) and (SGE(Q)) restricted to \mathcal{X} . Assume (a) Γ is set-valued taking convex and compact values in Y, (b) Y is finite-dimensional or Γ is single-valued, (c) $\Gamma(\cdot, \xi)$ is upper semi-continuous for every $\xi \in \Xi$ and integrably bounded, i.e., $\sup_{y \in \Gamma(x,\xi)} ||y||$ integrable for all $x \in \mathcal{X}$, (d) \mathcal{G} is upper semi-continuous, (e) S(Q) is nonempty if $\mathscr{D}(Q, P)$ is small.

For any $\epsilon > 0$, let

$$R(\epsilon) := \inf_{x \in \mathcal{X}, \ d(x, S(P)) \ge \epsilon} d(0, \mathbb{E}_P[\Gamma(x, \xi)] + \mathcal{G}(x)).$$

Then $R(\epsilon) \rightarrow 0$ as $\epsilon \rightarrow 0$ and

 $\mathbb{D}(S(Q), S(P)) \le R^{-1}(2\mathscr{D}(Q, P)),$

where $R^{-1}(t) := \min\{\epsilon \in \mathbb{R}_+ : R(\epsilon) = t\}.$

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Stability of linear two-stage stochastic programs

We consider

 $\min_{x \in \mathbb{R}^n} c^\top x + \mathbb{E}_P[v(x,\xi)]$ s.t. $x \in X$,

where X is convex polyhedral and $v(x,\xi)$ is the second stage optimal value function

$$\begin{split} \min_{y \in \mathbb{R}^m} & q(\xi)^\top y \\ \text{s.t.} & T(\xi)x + Wy = h(\xi), \ y \geq 0, \end{split}$$

where $W \in \mathbb{R}^{r \times m}$ is a fixed recourse matrix, $T(\xi) \in \mathbb{R}^{r \times n}$ is a random matrix, and $h(\xi) \in \mathbb{R}^r$ and $q(\xi) \in \mathbb{R}^m$ are random vectors. We assume that $T(\cdot)$, $h(\cdot)$ and $q(\cdot)$ are affine functions of ξ and that Ξ is a polyhedral subset of \mathbb{R}^s (for example, $\Xi = \mathbb{R}^s$). Stochastic generalized equation

 $0 \in \mathbb{E}_P[c - T(\xi)^\top D(x,\xi)] + \mathbb{N}_X(x),$

where $D(x,\xi)$ is the solution set of the dual second stage problem

 $D(x,\xi) := \arg \max_{W^{\top} \zeta \le q(\xi)} \zeta^{\top} (h(\xi) - T(\xi)x).$

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Theorem: Assume that

(a) $h(\xi) - T(\xi)x \in W(\mathbb{R}^m_+)$ for all $(\xi, x) \in \Xi \times X$, (b) $\mathcal{M}(q(\xi)) = \{\pi : W^\top \pi \le q(\xi)\} \ne \emptyset$ is bounded for all $\xi \in \Xi$, (c) P has finite second order moments, i.e., $\mathbb{E}_P[||\xi||^2] < +\infty$ and (d) X is a nonempty and bounded polyhedron. Then it holds for any probability measure Q such that $\mathscr{D}(Q, P)$ is sufficiently small

 $\mathbb{D}(S(Q), S(P)) \le R^{-1}(2\mathscr{D}(Q, P)),$

where the function R is defined by

 $R(\epsilon) := \inf_{x \in X, d(x, S(P)) \ge \epsilon} d(0, \mathbb{E}_P[\Gamma(x, \xi)] + \mathbb{N}_X(x)).$

The class \mathscr{F} for defining \mathscr{D} is contained in

 $\{g: g(\xi) - g(\tilde{\xi}) \le C \max\{1, \|\xi\|, \|\tilde{\xi}\|\}^2 \|\xi - \tilde{\xi}\|, \forall \xi, \tilde{\xi} \in \Xi\}.$

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Stability of two-stage SMPCC

The theory applies to

$$\begin{split} \min_{\substack{x, \, y(\cdot) \in \mathscr{Y} \\ x, \, y(\cdot) \in \mathscr{Y}}} & \mathbb{E}_P[f(x, y(\omega), \xi(\omega))] \\ \text{subject to} & x \in X \text{ and for almost every } \omega \in \Omega : \\ & g(x, y(\omega), \xi(\omega)) \leq 0, \\ & h(x, y(\omega), \xi(\omega)) = 0, \\ & 0 \leq G(x, y(\omega), \xi(\omega)) \perp H(x, y(\omega), \xi(\omega)) \geq 0, \end{split}$$

where X is a nonempty closed convex subset of \mathbb{R}^n , f, g, h, G, Hare continuously differentiable functions from $\mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^q$ to $\mathbb{R}, \mathbb{R}^s, \mathbb{R}^r, \mathbb{R}^m, \mathbb{R}^m$, respectively, $\xi : \Omega \to \Xi$ is a vector of random variables defined on probability (Ω, \mathbb{F}, P) with compact support set $\Xi \subset \mathbb{R}^q$, and $\mathbb{E}_P[\cdot]$ denotes the expected value with respect to probability measure P, and ' \perp ' denotes the perpendicularity of two vectors, \mathscr{Y} is a space of functions $y(\cdot) : \Omega \to \mathbb{R}^m$ such that $\mathbb{E}_P[f(x, y(\omega), \xi(\omega))]$ is well defined.

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The two-stage SPCC may be reformulated as

$$\min_{x} \quad \theta(x) = \mathbb{E}_{P}[v(x,\xi)]$$

s.t. $x \in X$,

where $\boldsymbol{v}(\boldsymbol{x},\boldsymbol{\xi})$ denotes the optimal value function of the following second stage problem:

$$\begin{split} \operatorname{MPCC}(x,\xi) : & \min_{y} \quad f(x,y,\xi) \\ & \mathsf{s.t.} \quad g(x,y,\xi) \leq 0, \\ & \quad h(x,y,\xi) = 0, \\ & \quad 0 \leq G(x,y,\xi) \perp H(x,y,\xi) \geq 0 \end{split}$$

Under certain assumptions $v(\cdot, \xi)$ is locally Lipschitz continuous (with a *P*-integrable Lipschitz constant) and one may consider the necessary optimality conditions (using the Clarke subdifferential)

 $0 \in \mathbb{E}_P[\partial_x v(x,\xi)] + \mathbb{N}_X(x).$

as SGE that hopefully satisfies the assumptions of the stability result.

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Let us, in particular, consider the stochastic optimization model $\min\{c^{\top}x + \mathbb{E}_{P}[q(\xi)^{\top}y]: 0 \in Wy + T(\xi)x - h(\xi) + N_{R^{\overline{m}}_{+}}(y), x \in X\},\$ where similar conditions are imposed as in the (standard) two-stage model before. The linear generalized equation is equivalent to the linear complementarity problem

 $Wy + T(\xi)x \ge h(\xi), y \ge 0, y^{\top}(Wy + Tx - h(\xi)) = 0.$

Its solution set is a polyhedral multifunction (of $a = h(\xi) - T(\xi)x$) and, hence, is locally upper Lipschitz continuous at each a (with the same modulus L > 0). Hence, the reformulation reads

 $\min\{c^{\top}x + \mathbb{E}[v(x,\xi)] : x \in X\}$

and the function $v(\cdot, \xi)$ is locally Lipschitz continuous (with constant $L \|q(\xi)\| \|T(\xi)\|$). Then the general theory implies (local) upper Lipschitz continuity of the solution set mapping at P with respect to the ζ -distance $\zeta_{\mathscr{F}}$ and the function class

 $\mathscr{F} = \{ v^o(x,\cdot;u) : x \in X, \|u\| \le 1 \},$

where $v^o(x,\xi;u)$ denotes the Clarke directional derivative of $v(\cdot,\xi)$ at x.

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Stability of convex programs with second order dominance constraints

We consider convex programs with second order dominance constraints

$$\min_{x} f(x)$$

s.t. $\mathbb{E}_{P}[(\eta - G(x,\xi))_{+}] \leq \mathbb{E}_{P}[(\eta - Y(\xi))_{+}], \forall \eta \in [a,b],$
 $x \in X,$

where X is a closed convex subset of \mathbb{R}^n , $f : \mathbb{R}^n \to \mathbb{R}$ is convex and differentiable and $G : \mathbb{R}^n \times \Xi \to \mathbb{R}$ is concave in the first component and has linear growth in the second, ξ is a random vector with distribution P and support Ξ in \mathbb{R}^d .

The constraint satisfies the uniform dominance condition (udc) at P if $\bar{x} \in X$ exists such that

$$\min_{\eta \in [a,b]} \left(\mathbb{E}_P[(\eta - G(\bar{x},\xi))_+] - \mathbb{E}_P[(\eta - Y(\xi))_+] \right) > 0.$$

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Optimality condition:

Let udc be satisfied at P. If a feasible $x^* \in X$ is optimal, there exists $u^* \in \mathcal{U}_1$ satisfying

$$0 \in \mathbb{E}_P[\Gamma(x^*, u^*, \xi)] + \mathcal{G}(x^*)$$

i.e.,

$$0 \in f'(x^*) + \mathbb{E}_P[\partial_x(-u^*(G(x^*,\xi))] + N_X(x^*)]$$

$$0 = \mathbb{E}_P[u^*(G(x^*,\xi)) - u^*(Y(\xi)))],$$

where $\mathcal{U}_1 = \{ u \in C^1(\mathbb{R}) : \exists \varphi : I \to \mathbb{R}_+ \text{ nonincreasing,} \\$ left-continuous and bounded such that $u'(t) = \varphi(t), t \in [a, b], \\ u'(t) = \varphi(a), t < a, u(t) = 0, t \ge b \}.$

Theorem:

Let udc be satisfied at P and X be compact. Then it holds

 $\mathbb{D}(S(Q),S(P)) \leq R^{-1}(2\mathscr{D}(Q,P)),$

where R and R^{-1} are defined in the stability theorem. Characterization of the class \mathscr{F} ?

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Another stability result for such models (Dentcheva/Römisch 12):

Let $v(\xi, Y)$ denote the optimal value and $S(\xi, Y)$ the solution set and $\mathcal{X}(\xi, Y)$ the feasible set.

We consider the growth function

$$\psi_{(\xi,Y)}(\tau) := \inf\{f(x) - v(\xi,Y) : d(x,S(\xi,Y)) \ge \tau, \, x \in \mathcal{X}(\xi,Y)$$

and

$$\Psi_{(\xi,Y)}(\theta) := \theta + \psi_{(\xi,Y)}^{-1}(2\theta) \quad (\theta \in \mathbb{R}_+),$$

where we set $\psi_{(\xi,Y)}^{-1}(t) = \sup\{\tau \in \mathbb{R}_+ : \psi_{(\xi,Y)}(\tau) \le t\}$. Note that $\Psi_{(\xi,Y)}$ is inreasing and vanishes at $\theta = 0$.

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Theorem:

Let X be compact and assume that the function G satisfies

$$|G(x,u) - G(x,\tilde{u})| \le L_G ||u - \tilde{u}||$$

for all $x \in D$, $u, \tilde{u} \in \Xi$ and some constant $L_G > 0$. Assume that udc is satisfied at (ξ, Y) .

Then there exist positive constants L and δ such that

$$\begin{split} |v(\xi,Y) - v(\tilde{\xi},\tilde{Y})| &\leq L \, d_2((\xi,Y),(\tilde{\xi},\tilde{Y})) \\ \mathbb{D}(S(\tilde{\xi},\tilde{Y}),S(\xi,Y)) &\leq \Psi_{(\xi,Y)}(L \, d_2((\xi,Y),(\tilde{\xi},\tilde{Y}))) \\ \end{split}$$
 whenever $d_2((\xi,Y),(\tilde{\xi},\tilde{Y})) < \delta.$

The metric d_2 is defined by

$$d_2((\xi, Y), (\tilde{\xi}, \tilde{Y})) = \ell_1(\xi, \tilde{\xi}) + \sup_{t \in \mathbb{R}} |F_Y^{(2)}(t) - F_{\tilde{Y}}^{(2)}(t)|$$

with the L_1 -minimal metric ℓ_1 and $F_Y^{(2)}(t) = \int_{-\infty}^t F_Y(x) dx$ ($t \in \mathbb{R}$).

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Conclusions

- The stability analysis of SGEs allows to extend the stability theory to more general stochastic variational problems.
- In particular, quantitative stability results for two-stage SPCCs and programs with stochastic dominance constraints were obtained.
- A characterization of the distances D and the function classes
 F might improve the understanding of scenario generation for such models.

Thank you !



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