

# Confidence Regions for Stochastic Variational Inequalities

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Talk at University of Wisconsin-Madison, Feb 6, 2012

Based on joint work with Amarjit Budhiraja

SVI introduction Confidence region method A numerical example Summary and plan

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## Outline

- 1 An introduction to stochastic variational inequalities
- 2 A method to build confidence regions
- 3 A numerical example
- 4 Summary and future research plan

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## An overview of related problems

Variational Inequalities

Stochastic Variational Inequalities

Stochastic Equilibria

Stochastic Optimization

Economic equilibria Nash equilibria userequilibria Frictional contactproblems Elastoplastic structure analysis

Nonlinear obstacle problems

Data Uncertainty

## Example: an SVI formulation of the lasso

$X \in \mathbb{R}^p$ : a random input vector;  $Y \in \mathbb{R}$ : a random output variable

$\beta = (\beta_0$

$, \beta_1$

$, \dots, \beta_p$

)

); the parameter to be determined

Let  $\lambda$  be a positive constant. A formulation of the lasso is

$\min_{\beta} \sum_{j=1}^p | \beta_j | + \lambda \sum_{j=1}^p \beta_j^2$

## Example: an SVI formulation of the lasso

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Let  $\lambda$  be a positive constant. A formulation of the lasso is

minimize  $\sum_{j=1}^p |\beta_j|$

subject to  $E[Y - \beta_0 - \sum_{j=1}^p \beta_j X_j] = 0$

min  $\sum_{j=1}^p |\beta_j|$

$E[Y - \beta_0 - \sum_{j=1}^p \beta_j X_j] = 0$

$-$

$\sum_{j=1}^p \beta_j X_j$

$\beta_0$

$\beta_j$

$X_j$

$]$

$^2 + \lambda$

$\sum_{j=1}^p |\beta_j|$

$|\beta_j|$

$j=1$

$j=1$

$]$

$\geq 0, t$

$+$

Define  $S = \{(\beta, t) \in \mathbb{R}^{p+1} \times \mathbb{R}^p \mid t_j \geq 0, j = 1, \dots, p\}$

$- \beta_j$

$\geq 0, t_j$

$+$

$\beta_j$

$\geq 0, j = 1, \dots, p\}$

$\geq 0, j = 1, \dots, p\}$

$\geq 0, j = 1, \dots, p\}$

$\geq 0, j = 1, \dots, p\}$

Under mild conditions, the lasso can be reformulated as

$$0 \in E$$

$$Y - X\beta$$

$$\beta$$

$$0$$

$$X^T \beta$$

$$\lambda \|\beta\|_1$$

$$\beta_1, \dots, \beta_p$$

$$p$$

$$X$$

$$+ N$$

$$S$$

$$(\beta, t)$$

where  $e \in \mathbb{R}^p$  is all ones, and  $N$

$$S$$

$(\beta, t)$  is the normal cone to  $S$  at  $(\beta, t)$

## Example: an energy market problem

Players of the game:  $m$  producers/exporters of a type of energy

$x_i$

: decision variable for player  $i$  (how much to invest, produce and ship ...), which takes values in a set  $K_i \subset \mathbb{R}^n$

$\theta$

: the profit of player  $i$ , a random function of  $x = (x_1, \dots, x_m)$

$x_1$

, ...,  $x_m$

)

)

Each player selects  $x_i$

to maximize the expected profit  $E(\theta_i)$

)

An equilibrium is attained, when no player can increase his expected profit by unilaterally changing his decision

Under mild conditions, the equilibrium problem can be formulated as

$0 \in -E$

$\nabla$

$x$

$1$

$\theta \dots$

$1$

$(x)$

$+ N$

$K$

$1 \nabla$

$x$

$m$

$x_1 \dots x_m$

$N$

$(x) \theta$

$m$



(x)

## Definition of stochastic variational inequalities

$(\Omega, \mathcal{F}, P)$ : a probability space  $\xi$ : a random vector from  $\Omega$  to a closed set  $\Xi \subset \mathbb{R}^d$   $O$ : an open subset of  $\mathbb{R}^n$ ;  $X$ : a nonempty compact subset of  $O$   $F : O \times \Xi \rightarrow \mathbb{R}^n$ , with  $E \|F(x, \xi)\| < \infty$  for each  $x \in O$  Let  $f$

0

$$f(x) = E[F(x, \xi)] \text{ for each } x \in O \quad S = \{x \in \mathbb{R}^n \mid Ax \leq b\} = \{x \in \mathbb{R}^n \mid \langle a_i,$$

$$x \rangle \leq b_i$$

$i$

$$, i = 1, \dots, m\}$$

The SVI problem is to find  $x \in S \cap O$  such that

$$0 \in f$$

0

$$f(x) + N$$

S

$$f(x) \text{ (SVI)}$$

where  $N$

S

$N(x)$  is the normal cone to S at x and is defined as

N

S

$$N(x) = \{v \in \mathbb{R}^n \mid \langle v, s - x \rangle \leq 0 \text{ for each } s \in S\}$$

## Example: a linear complementarity problem

Let  $F : \mathbb{R}^2 \times \mathbb{R}^6 \rightarrow \mathbb{R}^2$  be defined by

$$F(x, \xi) =$$

$$\begin{bmatrix} \xi_1 \\ \xi_2 \\ \xi_3 \\ \xi_4 \\ \xi_5 \\ \xi_6 \end{bmatrix} +$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} +$$

$$\begin{bmatrix} \xi_1 \\ \xi_2 \\ \xi_3 \\ \xi_4 \\ \xi_5 \\ \xi_6 \end{bmatrix},$$

where  $\xi$  is a random vector uniformly distributed on

$$\{\xi \in \mathbb{R}^6 \mid (0, 0, 0, 0, -1, -1) \leq \xi \leq (2, 1, 2, 4, 1, 1)\}$$

Let  $S = \mathbb{R}^2$

$$+$$

$$\xi$$

$$2 \xi$$

$$3$$

$$\xi$$

$$4$$

. The SVI problem becomes the following LCP:

$$0 \in$$

$$\begin{bmatrix} 1 \\ 1/2 \end{bmatrix} +$$

$$\begin{bmatrix} 1/2 \\ 1 \end{bmatrix} +$$

$$\begin{bmatrix} 1 \\ 1/2 \end{bmatrix} +$$

$$x + N$$

$$\mathbb{R}^2$$

$$+$$

(x),

which has a unique solution x

0

= 0

## The sample average approximation problem

In most problems of interest,  $f$

$0$

does not have a closed form expression and requires a numerical approximation

Let  $\xi_1, \dots, \xi_N$  be i.i.d. r.v.s with distribution same as  $\xi$

Define  $f$

$N$

$: O \times \Omega \rightarrow R^n$  by  $f$

$N$

$(x, \omega) = N^{-1}$

$\sum$

$N \sum_{i=1}^N$

$F(x, \xi_i(\omega))$

The SAA problem is to find  $x \in S \cap O$  such that

$0 \in f$

$N$

$(x, \omega) + N$

$S$

**(x) (SAA)**

For the LCP example, an SAA problem with  $N = 10$  is given by

$0 \in$

[

0.9292 0.7536 0.5400 2.1111

]

$x +$

[

-0.1319 -0.2906

]

$+ N$

$R^2$

$+$

$(x)$ ,

which has a unique solution  $x$

10

= (0.0782, 0.1097)