

# Analysis of a two-stage probabilistic programming model

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# A simple two-stage probabilistic program

$$x_1 \leadsto \xi_1 \leadsto x_2 \leadsto \xi_2$$

$x$  = decisions;  $\xi$  = random observations

Probabilistic constraint:  $\varphi(x_1, x_2) := \mathbb{P}(\xi_1 \leq x_1, \xi_2 \leq x_2(\xi_1)) \geq p$

Probabilistic program:  $\min \{f(x_1, x_2) \mid \varphi(x_1, x_2) \geq p, (x_1, x_2) \in \mathbb{R} \times L^2(\mathbb{R})\}$

For a basic structural analysis of general multistage probabilistic programs with  $L^2$  and  $W^{1,2}$  decision spaces we refer to:

*T. Gonzalez Grandon, R. Henrion u. P. Pérez-Aros, Dynamic probabilistic constraints under continuous random distributions, to appear in: Mathematical Programming.*



# Properties of the probability function

For simplicity:  $\xi = (\xi_1, \xi_2) \sim \mathcal{N}(\mu, \Sigma)$  ( bivariate Gaussian)

Then,  $\varphi(x_1, x_2) := \mathbb{P}(\xi_1 \leq x_1, \xi_2 \leq x_2(\xi_1))$  is Lipschitz continuous and

$$\nabla_{x_2} \varphi(x_1, x_2)(r) = K \exp\left(-\frac{1}{2} \begin{pmatrix} r - \mu_1 \\ x_2(r) - \mu_2 \end{pmatrix}^T \Sigma^{-1} \begin{pmatrix} r - \mu_1 \\ x_2(r) - \mu_2 \end{pmatrix}\right) \chi_{(-\infty, x_1]}$$

If  $x_2(\cdot)$  is continuous, then

$$\frac{\partial \varphi}{\partial x_1}(x_1, x_2) = K \exp\left(-\frac{1}{2\Sigma_{11}}(r - \mu_1)^2\right) \Phi\left(\frac{x_2(x_1) - \mu_2 - \Sigma_{11}^{-1}\Sigma_{12}(x_1 - \mu_1)}{\sqrt{\Sigma_{22} - \Sigma_{11}^{-1}\Sigma_{12}^2}}\right)$$



# Necessary optimality condition in $L^2$

Probabilistic program:  $\min\{c_1 x_1 + c_2 \mathbb{E}x_2(\xi_1) |_{\xi_1 \leq x_1} \text{ s.t. } \varphi(x_1, x_2) \geq p\}$

For any fixed  $x_1 \in \mathbb{R}$ , an optimal policy  $x_2(\cdot)$  in the conditional problem

$$\min\{c_2 \mathbb{E}x_2(\xi_1) |_{\xi_1 \leq x_1} \text{ s.t. } \varphi(x_1, x_2) \geq p\}$$

has the form  $x_2(r) = \frac{\Sigma_{12}}{\Sigma_{11}} r + C$  ( $\xi \sim \mathcal{N}(\mu, \Sigma)$ )    affine linear  $\notin L^2(\mathbb{R})$

No solution in  $L^2$ !



# Identification of a global solution

Probabilistic program:  $\min\{c_1 x_1 + c_2 \mathbb{E}x_2(\xi_1) |_{\xi_1 \leq x_1} \text{ s. t. } \varphi(x_1, x_2) \geq p\}$

We allow  $x_2(\cdot)$  to be an arbitrary Borel measurable function. Global solution:

$$x_1^* = \arg \min_{t \in \mathbb{R}} \frac{\Sigma_{12}}{\Sigma_{11}} \beta(t) + \alpha(t) \Phi \left( \frac{t - \mu_1}{\sqrt{\Sigma_{11}}} \right); \quad x_2^*(r) = \frac{\Sigma_{12}}{\Sigma_{11}} r + \alpha(x_1^*)$$

$$\alpha(t) = \Phi^{-1} \left( \frac{p}{\Phi \left( \frac{t - \mu_1}{\sqrt{\Sigma_{11}}} \right)} \right) \sqrt{\Sigma_{22} - \frac{\Sigma_{12}^2}{\Sigma_{11}}} + \mu_2 - \frac{\Sigma_{12}}{\Sigma_{11}} \mu_1; \quad \beta(t) = \mu_1 - \sqrt{\Sigma_{11}} \frac{\varphi \left( \frac{t - \mu_1}{\sqrt{\Sigma_{11}}} \right)}{\Phi \left( \frac{t - \mu_1}{\sqrt{\Sigma_{11}}} \right)}; \quad \varphi, \Phi - \text{density and CDF of } \mathcal{N}(0,1)$$

