# Bernstein approximation of chance constrained problems: an example

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#### Abstract

We study the example in [2] in more detail and repeat the computation using the new conic program solver.

### The model

We describe the chance constrained problem in detail. As in [2], consider the following chance constrained program

$$\max_{\substack{\tau \in \mathbb{R} \\ x_0, x_1 \cdots, x_n \ge 0}} (\tau - 1) \quad \text{s.t.} \quad \mathbb{P}\left(\tau > \sum_{j=0}^n r_j x_j\right) \le \alpha, \ \sum_{j=0}^n x_j \le 1$$
 (1)

where  $\alpha \in [0,1]$  is a given constant. The assumptions are

- 1. The returns  $r_0, r_1, \dots, r_n$  satisfy  $r_0 = 1$  and  $\mathbb{E}(r_i) = 1 + \rho_i$  with  $0 \le \rho_1 \le \dots \le \rho_n$ .
- 2. For  $1 \leq j \leq n$  and  $1 \leq l \leq q$ , one has  $r_j = \eta_j + \sum_{l=1}^q \gamma_{jl} \zeta_l$  where  $\eta_j \sim \mathcal{LN}(\mu_j, \sigma_j^2)$  (the individual noises) and  $\zeta_l \sim \mathcal{LN}(\nu_l, \theta_l^2)$ . All  $\eta_j$  and  $\zeta_l$  are independent of each other.
- 3. One has  $\nu_l = 0$ ,  $\theta_l = 0.1$  for all l,  $\mu_j = \sigma_j$  for all j,  $\sum_{l=1}^q \gamma_{jl} \exp\left(\nu_l + \frac{\theta_l^2}{2}\right) = \frac{\rho_j}{2}$  for all j and  $\sum_{j=1}^n \exp\left(\mu_j + \frac{\sigma_j^2}{2}\right) = 1 + \frac{\rho_j}{2}$ .

We see that the problem can be rewritten into (1.1) in [2] with m=1. Denote

$$\tilde{x} = (\tau, x_0, x_1, \cdots, x_n)^T.$$

The objective function is simply  $f(\tilde{x}) = -\tau$ , and the chance constraint is

$$\mathbb{P}\left(F(\tilde{x},\xi) \le 0\right) \ge 1 - \alpha$$

where

$$F(\tilde{x},\xi) = g_0(\tilde{x}) + \sum_{j=1}^{d} \xi_j g_j(\tilde{x}), \ d = n + q, \ g_0(\tilde{x}) = \tau - x_0,$$

$$\xi_j = \eta_j, \ g_j(\tilde{x}) = -x_j, \ 1 \le j \le n,$$

$$\xi_{n+l} = \zeta_l, \ g_{n+l}(\tilde{x}) = -\sum_{j=1}^n \gamma_{jl} x_j, \ 1 \le l \le q.$$

# The Bernstein approximation and standard form formulation

Here we construct the Bernstein approximation to (1), which is a convex optimization problem and reformulate it into a standard form conic program involving exponential cone constraints.

Note that the discretization scheme described in [2] has been adopted and all random variables  $\xi_j$ ,  $1 \leq j \leq d$  are now discrete with finite support. For each j, denote the support and the associated probability masses as  $\left\{(v_k^j, p_k^j)\right\}_{1 \leq k \leq N_j}$ . In other words, for each j,  $k = 1, \dots, N_j$ , one has  $\mathbb{P}\left(\xi_j = v_k^j\right)$  and the moment generating function of  $\xi_j$  is  $M_j: z \to \sum_{k=1}^{N_j} p_k^j \exp\left(v_k^j z\right)$ .

The Bernstein approximation to (1) is therefore the following convex maximization problem

$$\max_{\substack{\tau \in \mathbb{R} \\ x_0, x_1 \cdots, x_n \ge 0}} (\tau - 1) \quad \text{s.t.} \quad \sum_{j=0}^n x_j \le 1, \quad \inf_{t>0} \left( g_0(\tilde{x}) + \sum_{j=1}^d t \Lambda_j \left( t^{-1} g_j(\tilde{x}) \right) - t \log \alpha \right) \le 0$$

$$(2)$$

Note that problem (2) is equivalent to

$$\max_{\tau \in \mathbb{R}} (\tau - 1)$$

$$\tau \in \mathbb{R}$$

$$x_0, x_1, \dots, x_n \ge 0$$

$$g_0, g_1, \dots, g_d \in \mathbb{R}$$

$$s_1, \dots, s_d \in \mathbb{R}$$

$$s.t. \quad \sum_{j=0}^n x_j \le 1$$

$$g_0 + \sum_{j=1}^d s_j - t \log \alpha = 0$$

$$(*)_j: \quad s_j \ge t\Lambda_j \left(\frac{g_j}{t}\right), \quad j = 1, \dots, d$$

$$g_0 = \tau - x_0$$

$$g_j = -x_j, \quad j = 1, \dots, n$$

$$g_{n+l} = -\sum_{j=1}^n \gamma_{jl} x_j, \quad l = 1, \dots, q$$

$$(3)$$

Since  $\Lambda_{j}(\cdot) = \log M_{j}(\cdot)$ , for  $j = 1, \dots, d$ , constraint  $(*)_{j}$  in (3) is equivalent to

$$\sum_{k=1}^{N_j} p_k^j \exp\left(v_k^j \cdot \frac{g_j}{t}\right) \le \exp\left(\frac{s_j}{t}\right)$$

$$\Leftrightarrow \sum_{k=1}^{N_j} p_k^j \exp\left(\frac{v_k^j g_j - s_j}{t}\right) \le 1$$

$$\Leftrightarrow \sum_{k=1}^{N_j} p_k^j \cdot t \exp\left(\frac{v_k^j g_j - s_j}{t}\right) \le t$$

$$\Leftrightarrow \sum_{k=1}^{N_j} p_k^j u_k^j = t, \quad t \exp\left(\frac{w_k^j}{t}\right) \le u_k^j, \ w_k^j = v_k^j g_j - s_j, \ k = 1, \dots, N_j$$

$$\Leftrightarrow \sum_{k=1}^{N_j} p_k^j u_k^j = t, \quad \left[w_k^j; u_k^j; t\right] \in \mathcal{K}_{\exp}, \ w_k^j = v_k^j g_j - s_j, \ k = 1, \dots, N_j.$$

Eventually, problem (2) can be reformulated into the standard form (PD') in [1], namely (note that d = n + q and the constant term in the objective has been dropped)

s.t. 
$$x_0 + x_1 + \dots + x_n + s_x = 1$$

$$g_0 + \left(\sum_{j=1}^d s_j\right) - (\log \alpha) t_0 = 0$$

$$g_0 - \tau + x_0 = 0$$

$$g_j + x_j = 0, \ j = 1, \dots, n$$

$$g_{n+l} + \sum_{j=1}^n \gamma_{jl} x_j = 0, \ l = 1, \dots, q$$

$$w_k^j - v_k^j g_j + s_j = 0, \ j = 1, \dots, d, \ k = 1, \dots, N_j$$

$$\sum_{k=1}^{N_j} p_k^j u_k^j - t_0 = 0, \ j = 1, \dots, d$$

$$t_0 - t_k^j = 0, \ j = 1, \dots, d, \ k = 1, \dots, N_j$$

with decision variables

$$\begin{aligned} & \tau \in \mathbb{R} \\ & x_0, x_1, \cdots, x_n, s_x \geq 0 \\ & g_0, g_1, \cdots, g_d \in \mathbb{R} \\ & t_0 \geq 0 \\ & s_1, \cdots, s_d \in \mathbb{R} \\ & \left[ w_k^j; u_k^j; t_k^j \right] \in \mathcal{K}_{\text{exp}}, \ j = 1, \cdots, d, \ k = 1, \cdots, N_j. \end{aligned}$$

Note that we keep the slack variable  $s_x \ge 0$  in the first constraint, although it can be shown that there is always an optimal solution  $(x_0^*, x_1^*, \dots, x_n^*)$  with  $\sum_{j=0}^n x_j^* = 1$ .

# References

- [1] Y. Gao. Design and implementation of a homogeneous interior-point method for conic programming involving exponential cone constraints, 2006.
- [2] A. Nemirovsky and A. Shapiro. Convex approximation of chance constrained programs. *SIAM Journal on Optimization*, 17(4):969–996, 2006.