
Multistage Stochastic Programming Problems; Stability and Approximation

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1 Introduction

A multistage stochastic programming problem can be introduced as a finite system of parametric one-stage optimization problems with an inner type of dependence and mathematical (mostly conditional) expectation in objective functions of the individual problems (for more details see e.g. [1], [3], [8]). The constraints sets can depend on the “underlying” probability measure.

To recall the multistage (M -stage, $M \geq 2$) stochastic programming problem let $\xi^j := \xi^j(\omega)$, $j = 1, \dots, M$ denote an s -dimensional random vector defined on a probability space (Ω, \mathcal{S}, P) ; $F^{\xi^j}(z^j)$, $z^j \in R^s$, $j = 1, 2, \dots, M$ denote the distribution function of the ξ^j and $F^{\xi^k|\bar{\xi}^{k-1}}(z^k|\bar{z}^{k-1})$, $z^k \in R^s$, $\bar{z}^{k-1} \in R^{(k-1)s}$, $k = 2, \dots, M$ denote the conditional distribution function (ξ^k conditioned by $\bar{\xi}^{k-1}$); $Z_{F^{\xi^j}} \subset R^s$, $j = 1, 2, \dots, M$ denote the support of the probability measure corresponding to F^{ξ^j} . Let, moreover, $g_0^M(\bar{x}^M, \bar{z}^M)$ be a continuous function defined on $R^{nM} \times R^{sM}$; $X^k \subset R^n$, $k = 1, 2, \dots, M$ be a nonempty set; $\mathcal{K}_{\mathcal{F}}^{k+1}(\bar{x}^k, \bar{z}^k) := \mathcal{K}_{F^{\xi^{k+1}|\bar{\xi}^k}}^{k+1}(\bar{x}^k, \bar{z}^k)$, $k = 1, \dots, M-1$ be a multifunction mapping $R^{nk} \times R^{sk}$ into the space of subsets of R^n . $\bar{\xi}^k(= \bar{\xi}^k(\omega)) = [\xi^1, \dots, \xi^k]$; $\bar{z}^k = [z^1, \dots, z^k]$, $z^j \in R^s$; $\bar{x}^k = [x^1, \dots, x^k]$, $x^j \in R^n$; $\bar{X}^k = X^1 \times X^2 \times \dots \times X^k$; $\bar{Z}_{\mathcal{F}}^k = Z_{F^{\xi^1}} \times Z_{F^{\xi^2}} \times \dots \times Z_{F^{\xi^k}}$, $j = 1, \dots, k$, $k = 1, 2, \dots, M$. See that (generally) the multifunctions $\mathcal{K}_{F^{\xi^{k+1}|\bar{\xi}^k}}^{k+1}(\bar{x}^k, \bar{z}^k)$, $k = 1, \dots, M-1$ can depend on the system of the conditional probability measures.

We introduce the M -stage stochastic programming problem ($M \geq 2$) in a rather general form as the problem.

$$\text{Find} \quad \varphi_{\mathcal{F}}(M) = \inf \{E_{F^{\xi^1}} g_{\mathcal{F}}^1(x^1, \xi^1) \mid x^1 \in \mathcal{K}^1\}, \quad (1)$$

where the function $g_{\mathcal{F}}^1(x^1, z^1)$ is defined recursively

$$g_{\mathcal{F}}^k(\bar{x}^k, \bar{z}^k) = \inf\{E_{F^{\xi^{k+1}}|\bar{\xi}^k=\bar{z}^k} g_{\mathcal{F}}^{k+1}(\bar{x}^{k+1}, \bar{\xi}^{k+1}) | x^{k+1} \in \mathcal{K}_{\mathcal{F}}^{k+1}(\bar{x}^k, \bar{z}^k)\},$$

$$k = 1, \dots, M-1,$$

$$g_{\mathcal{F}}^M(\bar{x}^M, \bar{z}^M) := g_0^M(\bar{x}^M, \bar{z}^M), \quad \mathcal{K}^1 = X^1. \quad (2)$$

$E_{F^{\xi^1}}, E_{F^{\xi^{k+1}}|\bar{\xi}^k=\bar{z}^k}, k = 1, \dots, M-1$ denote the operators of mathematical expectation corresponding to $F^{\xi^1}, F^{\xi^{k+1}}|\bar{\xi}^k=\bar{z}^k$.

Evidently, generally, a complete information on the system

$$\mathcal{F} = \{F^{\xi^1}(z^1), \quad F^{\xi^k|\bar{\xi}^{k-1}}(z^k|\bar{z}^{k-1}), k = 2, \dots, M\} \quad (3)$$

is a necessary condition to solve the above defined problem. However in applications very often one of the following cases happen: The system (3) must be replaced by its statistical estimates; the system (3) must be (for numerical difficulties) replaced by some simpler one; the actual system (3) is a little modified (e.g. by a contamination). Consequently, a stability (considered with respect to a probability measures space) and empirical estimates in the multistage problems have been (in the stochastic programming literature) investigated (see e.g. [3], [4], [9]). Approximation solution schemes (in the case of “deterministic” constraints set) have been suggested e.g. in [13] (see also [4]). We focus to a special case when the random element follows an autoregressive random sequence and the constraints sets correspond to a system of the individual probability constraints. In details we assume:

A.1 $\{\xi^k\}_{k=-\infty}^{\infty}$ follows a nonlinear autoregressive sequence

$$\xi^k = H(\xi^{k-1}) + \varepsilon^k, \quad (4)$$

where $\xi^0, \varepsilon^k, k = 1, 2, \dots$ are stochastically independent ($\varepsilon^k, k = 1, \dots$ identically distributed), $H(z)$ is a Lipschitz function on R^s (we denote the distribution function corresponding to ε^1 by the symbol F^ε and suppose the realization ξ^0 to be known),

A.2 there exist $g_i^{k+1}(\bar{x}^{k+1}), k = 1, 2 \dots, M-1$ defined on $R^{n(k+1)}$ and $\alpha_i \in (0, 1), i = 1, \dots, s, \bar{\alpha} = (\alpha_1, \dots, \alpha_s)$ such that

$$\mathcal{K}_{\mathcal{F}}^{k+1}(\bar{x}^k, \bar{z}^k) \quad (:= \mathcal{K}_{\mathcal{F}}^{k+1}(\bar{x}^k, \bar{z}^k; \bar{\alpha})) =$$

$$\bigcap_{i=1}^s \{x^{k+1} \in X^{k+1} : P_{F^{\xi^{k+1}}|\bar{\xi}^k=\bar{z}^k} \{g_i^{k+1}(\bar{x}^{k+1}) \leq \xi_i^{k+1}\} \geq \alpha_i\}. \quad (5)$$

This special case has been already investigated in [5]. In this contribution we go deeper. Our stability results will be based on the \mathcal{L}_1 norm and, moreover we intend to investigate (separately) the stability of the constraints sets; the last is important from applications point of view. To this end we employ the results achieved for one-stage problems. To justify this approach see e.g. [3].