

# Very Short Introduction to Stochastic Programming

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- 1 Dealing with Uncertainty
- 2 Stochastic Programming Modelling

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# An optimization problem

A standard optimization problem

$$\begin{array}{ll} \min_{u_0} & L(u_0) \\ \text{s.t.} & g(u_0) \leq 0 \end{array}$$

# An optimization problem with uncertainty

Adding uncertainty  $\xi$  in the mix

$$\begin{aligned} \min_{u_0} \quad & L(u_0, \xi) \\ \text{s.t.} \quad & g(u_0, \xi) \leq 0 \end{aligned}$$

Remarks:

- $\xi$  is unknown. Two main way of modelling it:
  - $\xi \in \Xi$  with a known uncertainty set  $\Xi$ , and a pessimistic approach. This is the robust optimization approach (RO).
  - $\xi$  is a random variable with known probability law. This is the Stochastic Programming approach (SP).
- Cost is not well defined.
  - RO :  $\max_{\xi \in \Xi} L(u, \xi)$ .
  - SP :  $\mathbb{E}[L(u, \xi)]$ .
- Constraints are not well defined.
  - RO :  $g(u, \xi) \leq 0, \quad \forall \xi \in \Xi$ .
  - SP :  $L(u, \xi) \leq 0, \quad \mathbb{P} - a.s.$

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# One-Stage Problem

Assume that  $\Xi$  as a discrete distribution, with  $\mathbb{P}(\xi = \xi_i) = p_i$  for  $i \in \llbracket 1, n \rrbracket$ . Then, the one-stage problem

$$\begin{aligned} \min_{u_0} \quad & \mathbb{E} [L(u_0, \xi)] \\ \text{s.t.} \quad & g(u_0, \xi) \leq 0, \quad \mathbb{P} - a.s \end{aligned}$$

can be written

$$\begin{aligned} \min_{u_0} \quad & \sum_{i=1}^n p_i L(u_0, \xi_i) \\ \text{s.t.} \quad & g(u_0, \xi_i) \leq 0, \quad \forall i \in \llbracket 1, n \rrbracket. \end{aligned}$$



# Recourse Variable

In most problem we can make a correction  $u_1$  once the uncertainty is known:

$$u_0 \rightsquigarrow \xi_1 \rightsquigarrow u_1.$$

As the **recourse** control  $u_1$  is a function of  $\xi$  it is a random variable. The **two-stage** optimization problem then reads

$$\begin{aligned} \min_{u_0} \quad & \mathbb{E} \left[ L(u_0, \xi, \mathbf{u}_1) \right] \\ \text{s.t.} \quad & g(u_0, \xi, \mathbf{u}_1) \leq 0, \quad \mathbb{P} - a.s \end{aligned}$$

# Two-stage Problem

The **extensive formulation** of

$$\begin{aligned} \min_{u_0} \quad & \mathbb{E} \left[ L(u_0, \boldsymbol{\xi}, \mathbf{u}_1) \right] \\ \text{s.t.} \quad & g(u_0, \boldsymbol{\xi}, \mathbf{u}_1) \leq 0, \quad \mathbb{P} - a.s \end{aligned}$$

is

$$\begin{aligned} \min_{u_0, \{u_1^i\}_{i \in [1, n]}} \quad & \sum_{i=1}^n p_i L(u_0, \xi_i, u_1^i) \\ \text{s.t.} \quad & g(u_0, \xi_i, u_1^i) \leq 0, \quad \forall i \in [1, n]. \end{aligned}$$

# Multi-stage Problem

- We can consider a multi-stage problem with successive decision and aleas

$$u_0 \rightsquigarrow \xi_1 \rightsquigarrow u_1 \rightsquigarrow \xi_2 \rightsquigarrow \cdots u_T.$$

- If each each alea  $\xi_i$  has 10 possible realizations, then there are
  - 1 control  $u_0$
  - 10 control  $u_1^i$
  - 100 control  $u_2^i$
  - ...
- In practice only two or three-stage problem can be solved by Stochastic Programming approaches.
- Remark : a stage is not necessarily a time-step.