

Consider the **deterministic-equivalent LP** derived from the 2-stage stochastic LP:

$$Z = \min cx + \sum_{k=1}^{K} p_k q_k y_k$$
(0.1)
subject to
$$T_k x + W y_k = h_k, k = 1, \dots K;$$
(0.2)
$$x \in X$$
(0.3)
$$y_k \ge 0, k = 1, \dots K \dots$$
(0.4)

where, for example, the feasible set of first-stage decisions is defined by

$$X = \left\{ x \in \mathbb{R}^{n} : Ax = b, x \ge 0 \right\}$$
(0.5)

Note that in general, the parameters q, T, h (and sometimes W) may vary by scenario!

SLPwR

Here k indexes the finitely-many possible *realizations* of a random vector ξ , with p_k the probability of realization k.

The set of **first-stage decision variables** x are to be selected *before* ξ is observed.

Then the set of **second-stage decision variables** y_k are to be selected once x has been selected and the k^{th} realization of ξ is observed.

Note that in general, the coefficient matrices T and W, the right-hand-side vector h, and the second-stage cost vector q are all *random*.

We assume here that for any choice of x and realization ξ , the constraints (0.2) are feasible in y, a condition known as *complete* recourse. (This may require the introduction of *artificial* variables with large costs.)

The objective is to minimize the **expected total costs** of first and second stages

Minimize
$$cx + \sum_{k=1}^{K} p_k Q_k(x)$$
 (0.6)

subject to $x \in X$

where the cost of the second stage is

$$Q_k(x) = \text{Minimum} \{ q_k y : W_k y = h_k - T_k x, \ y \ge 0 \}$$
(0.7)

The function $Q_k(x)$ is nonlinear and costly to evaluate for any x, but has some nice properties (e.g., convexity and continuity).