Online Appendix

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This is the online appendix of our submitted manuscript "Feasibility in Multistage Robust Dispatch with Renewables: A Recursive Characterization and Scalable Approximation". Reviewers/researchers are requested to read the full text of both the manuscript and this appendix for better understanding.

In a backward recursion, given a polyhedron \mathcal{F}_t and a polyhedral uncertainty set Ξ_t , we need to compute \mathcal{F}_{t-1} where

$$\mathcal{F}_{t-1} = \left\{ x_{t-1} \middle| \begin{matrix} \forall \xi_t \in \Xi_t, \exists (x_t, y_t) \colon x_t \in \mathcal{F}_t \\ Ax_t + By_t + Cx_{t-1} \leq b + D\xi_t \end{matrix} \right\}$$

The submitted manuscript has provided the solution when the uncertainty set Ξ_t is low-dimensional, whose vertices can be enumerated. This appendix focuses on the case where vertex enumeration is intractable.

To this end, define

$$\mathcal{F}_{t-1}(\xi_t) = \left\{ x_{t-1} \middle| \begin{array}{l} \exists (x_t, y_t) : x_t \in \mathcal{F}_t \\ Ax_t + By_t + Cx_{t-1} \le b + D\xi_t \end{array} \right\}$$
 (1)

Obviously, we have

$$\mathcal{F}_{t-1} = \bigcap_{\forall \xi_t \in \Xi_t} \mathcal{F}_{t-1}(\xi_t) \quad (2)$$

 $\mathcal{F}_{t-1}(\xi_t)$ is the orthogonal projection of $\mathcal{Q}_{t-1}(\xi_t)$ where

$$Q_{t-1}(\xi_t) = \left\{ (x_{t-1}, x_t, y_t) \middle| \begin{array}{c} x_t \in \mathcal{F}_t \\ Ax_t + By_t + Cx_{t-1} \le b + D\xi_t \end{array} \right\}$$
(3)

By denoting $\mathcal{F}_{t-1}(\xi_t) = \text{Proj}_{x_{t-1}}[\mathcal{Q}_{t-1}(\xi_t)]$, condition (2) is equivalent to

$$\mathcal{F}_{t-1} = \bigcap_{\forall \xi_t \in \Xi_t} \text{Proj}_{x_{t-1}} [\mathcal{Q}_{t-1}(\xi_t)] \quad (4)$$

Due to the difficulty in computing large-scale polyhedral projection, the idea is to approximate \mathcal{F}_{t-1} from within by an affine transformation $\Lambda \mathbb{B} + \mu$. Polyhedron \mathbb{B} is a template, diagonal matrix Λ is for stretch, and column vector μ is for translation. Please refer to the submitted manuscript for details.

To find the maximal inner approximation of \mathcal{F}_{t-1} , we resort to the following optimization problem:

$$\max_{\Lambda \geq 0, \mu} \det(\Lambda)$$
s. t. $\Lambda \mathbb{B} + \mu \subseteq \mathcal{F}_{t-1} = \bigcap_{\forall \xi_t \in \Xi_t} \operatorname{Proj}_{x_{t-1}} [\mathcal{Q}_{t-1}(\xi_t)]$ (5)

Problem (5) is equivalent to

$$\max_{\Lambda \geq 0, \mu} \det(\Lambda)$$
s. t. $\forall \xi_t \in \Xi_t : \Lambda \mathbb{B} + \mu \subseteq \operatorname{Proj}_{x_{t-1}}[Q_{t-1}(\xi_t)]$ (6)

By Claim 1 in the submitted manuscript, problem (6) is further equivalent to $\max_{\Lambda \geq 0, \mu} \det(\Lambda)$

s. t.
$$\forall \xi_t \in \Xi_t, \forall u \in \mathbb{B}, \exists \hat{u}: \begin{bmatrix} u \\ \hat{u} \end{bmatrix} \in \Phi \mathcal{Q}_{t-1}(\xi_t) + \eta$$
 (7)

where Φ is a diagonal matrix built with Λ , and η is a vector that depends on μ . Please refer to the submitted manuscript for details.

Apparently, problem (7) is a two-stage robust optimization and shares the same mathematical structure with the problem (25) in the submitted manuscript. Hence, the proposed solution procedure in **Algorithm 1**, which consists of a master problem and a sub-problem, is applicable to problem (7).

In conclusion, this online appendix reveals that when the vertices of uncertainty set Ξ_t cannot be enumerated, the proposed method still works with minor amendments to the model.