

## Multistage Stochastic Mixed Integer Programming

Let  $\alpha_k(n)$  denote the  $k^{th}$  ancestor of node  $n$ ,  $t(n)$  the time period of node  $n$  in the scenario tree, and  $p_n$  the probability associated with node  $n$ .

$$z := \min_{\mathbf{y}_{gn}} \sum_{n=1}^N p_n \sum_{g=1}^G \mathbf{d}_{gn}^\top \mathbf{y}_{gn} \quad (1a)$$

$$\text{s.t. } \mathbf{W}_{g0} \mathbf{y}_{g0} = \mathbf{h}_{g0} \quad \forall g \in \{1, \dots, G\}, \quad (1b)$$

$$\mathbf{T}_{g, \alpha_k(n)} \mathbf{y}_{g, \alpha_k(n)} + \mathbf{W}_{gn} \mathbf{y}_{gn} = \mathbf{h}_{gn} \quad \forall g \in \{1, \dots, G\}, \quad (1c)$$

$$\mathbf{y}_{gn} \in \mathbf{Y} \quad \forall g \in \{1, \dots, G\}, n \in \{1, \dots, N\}. \quad (1d)$$

**Challenges:** Complexity: NP-hard (generally); Curse of dimensionality; etc.

## Applications

- Unit Commitment & Economic Dispatch:

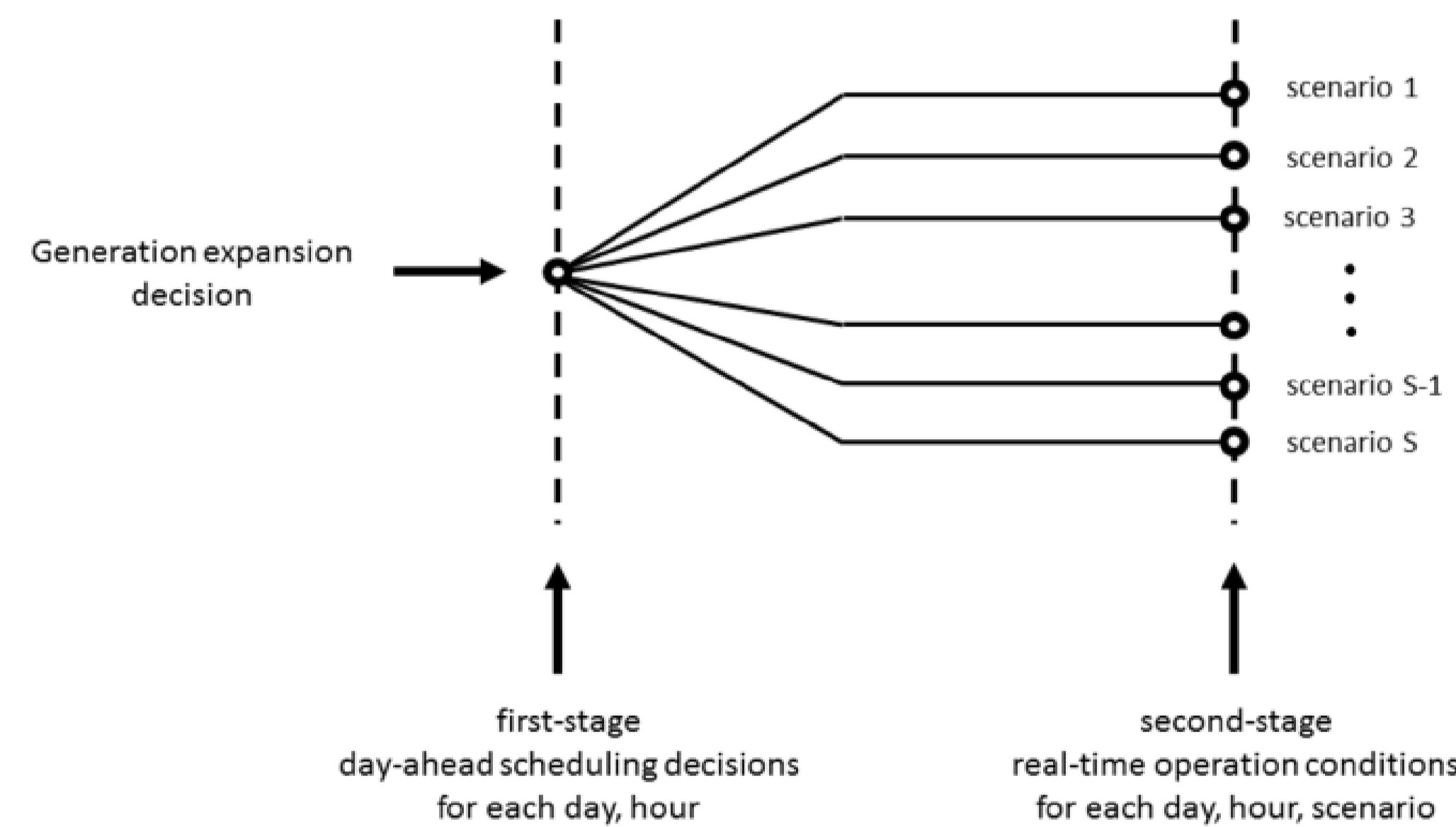


Figure 1:A. Schewe, et. al. (2020). *Do UC constr.'s affect gen. exp. planning? Scal. stoch. model*

- Other applications: portfolio management, resource allocation, disaster mitigation.

## Motivation: When & How?

### When does a multistage extension make sense?

- Sequential decision making; handling uncertainty dynamically; minimizing costs over multiple time periods.

**Aim: implement a parallel algorithm to solve multistage SMIPs.**

- Using Julia and a parallel decomposition algorithm speeds up computations.
- Using a multistage model rather than a two-stage model improves solution quality and allows for uncertainty to be handled dynamically.
- Motivated by increase in computing power & Julia modeling languages.

## Decomposition for Scenario Trees/Lattices

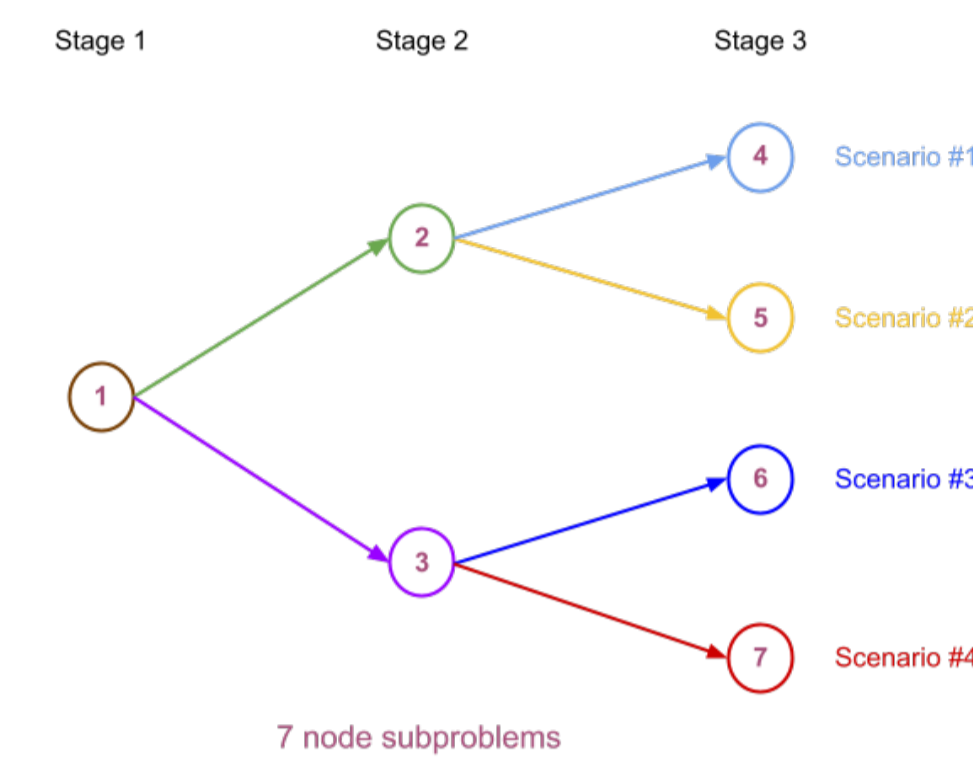


Figure 2:3-stage Scenario Tree

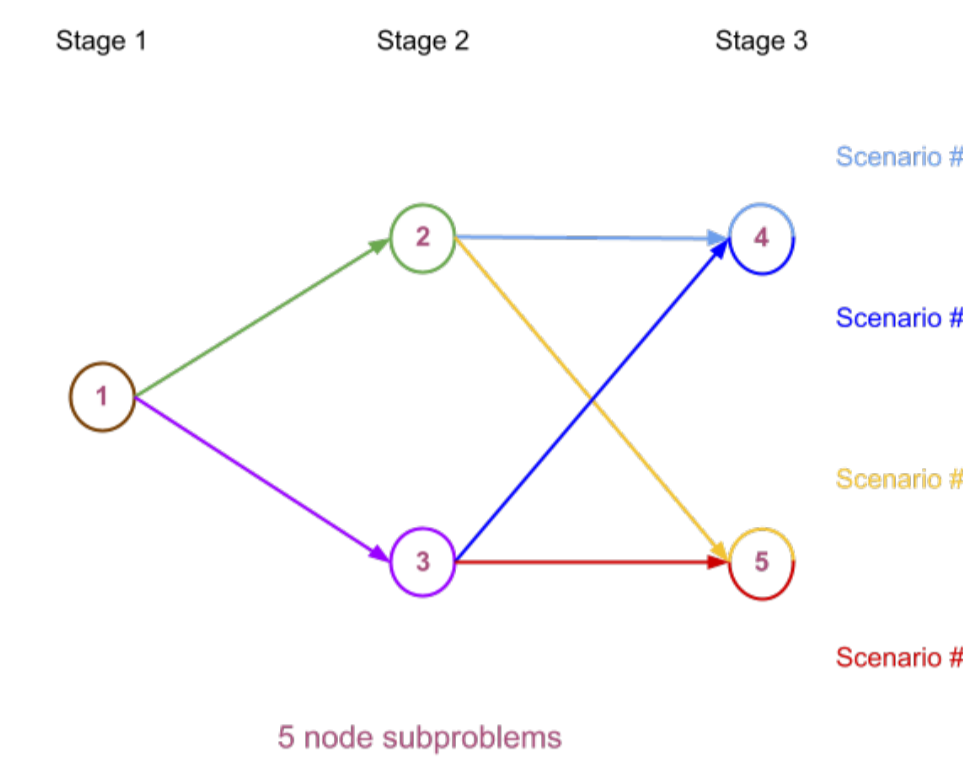


Figure 3:3-stage Scenario Lattice

- Discrete-time stochastic process.
- Markovian (memoryless) process.

We can decompose the problem by *scenarios* or by *nodes*.

## Decomposition of Structured Programs (DSP)

- DSP is a **parallel decomposition MIP solver**.
- Can implement decomposition methods for structured MIPs.
- Multistage SMIP has a block-diagonal structure that makes it easily parallelizable.**

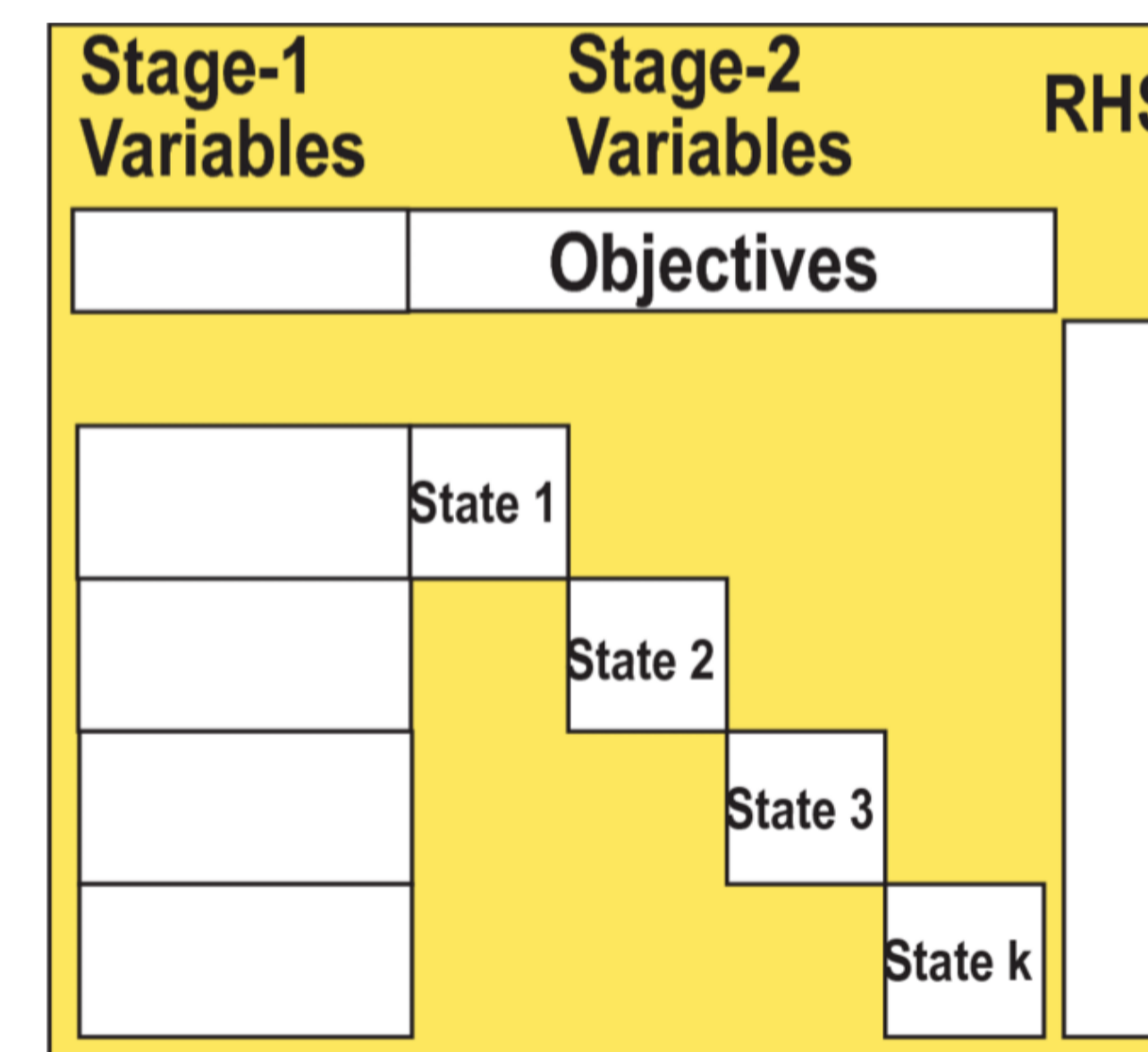
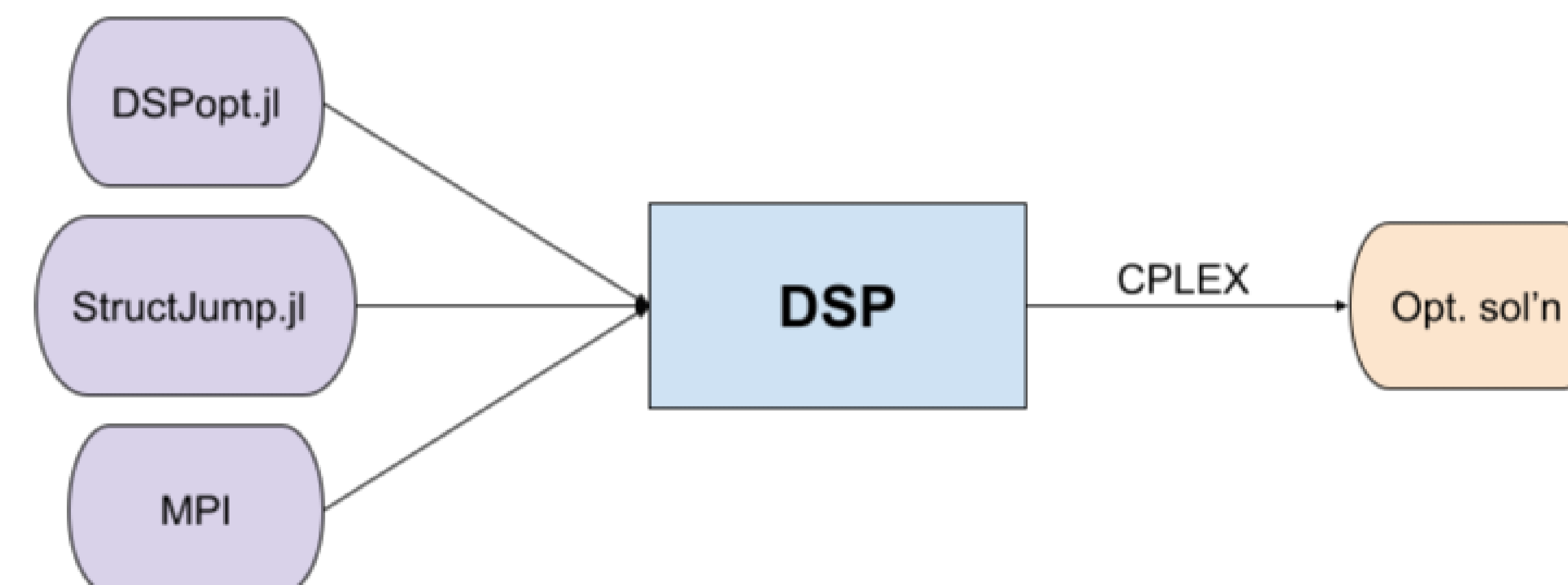
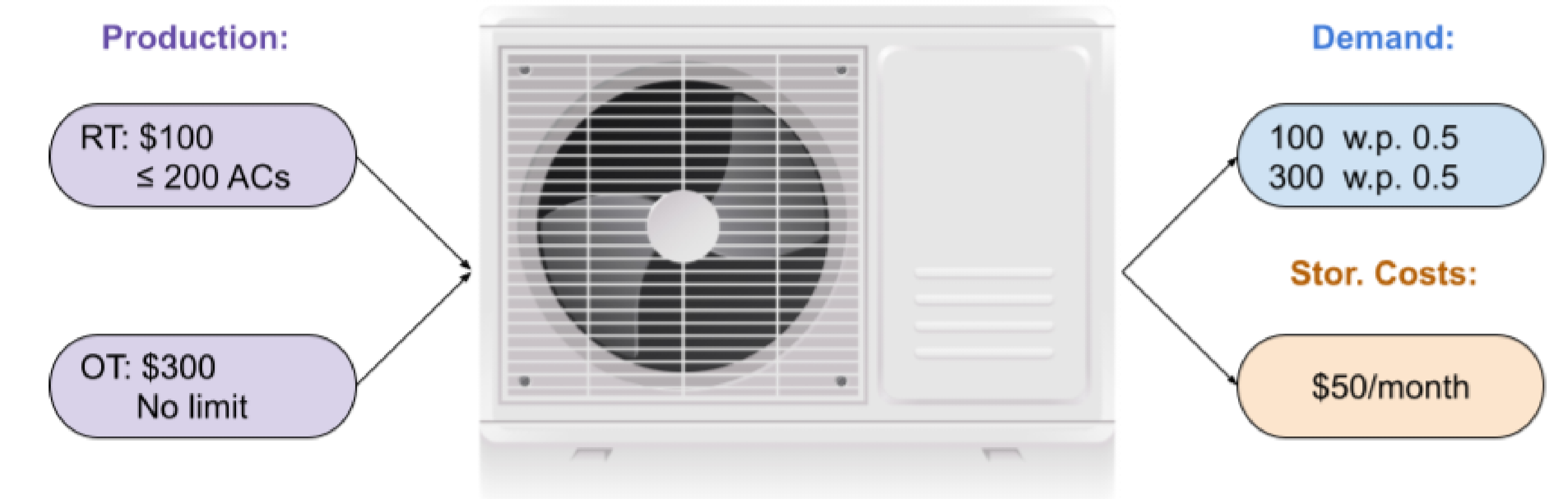


Figure 4:R.M. Freund (2004). *Benders' Decomposition Methods for Struct. Opt., incl. Stoch. Opt.*



**Goal: Encode optimization problem so that it can be read by StructJuMP and fed into DSPopt to solve in parallel.**

## Example: T-stage AC production problem



**How many ACs do we produce, with which type of labor, in each time period, to meet demand while minimizing production cost?**

Code is available here: <https://github.com/kibaekkim/DSPopt.jl/tree/ra/multistage/examples/multistage>.

## Computational Results

The following computations were run on an Argonne compute node with CPU: 1X AMD EPYC 7453 28-Core CPU, 256GB Memory, running Ubuntu 20.04 as its OS.

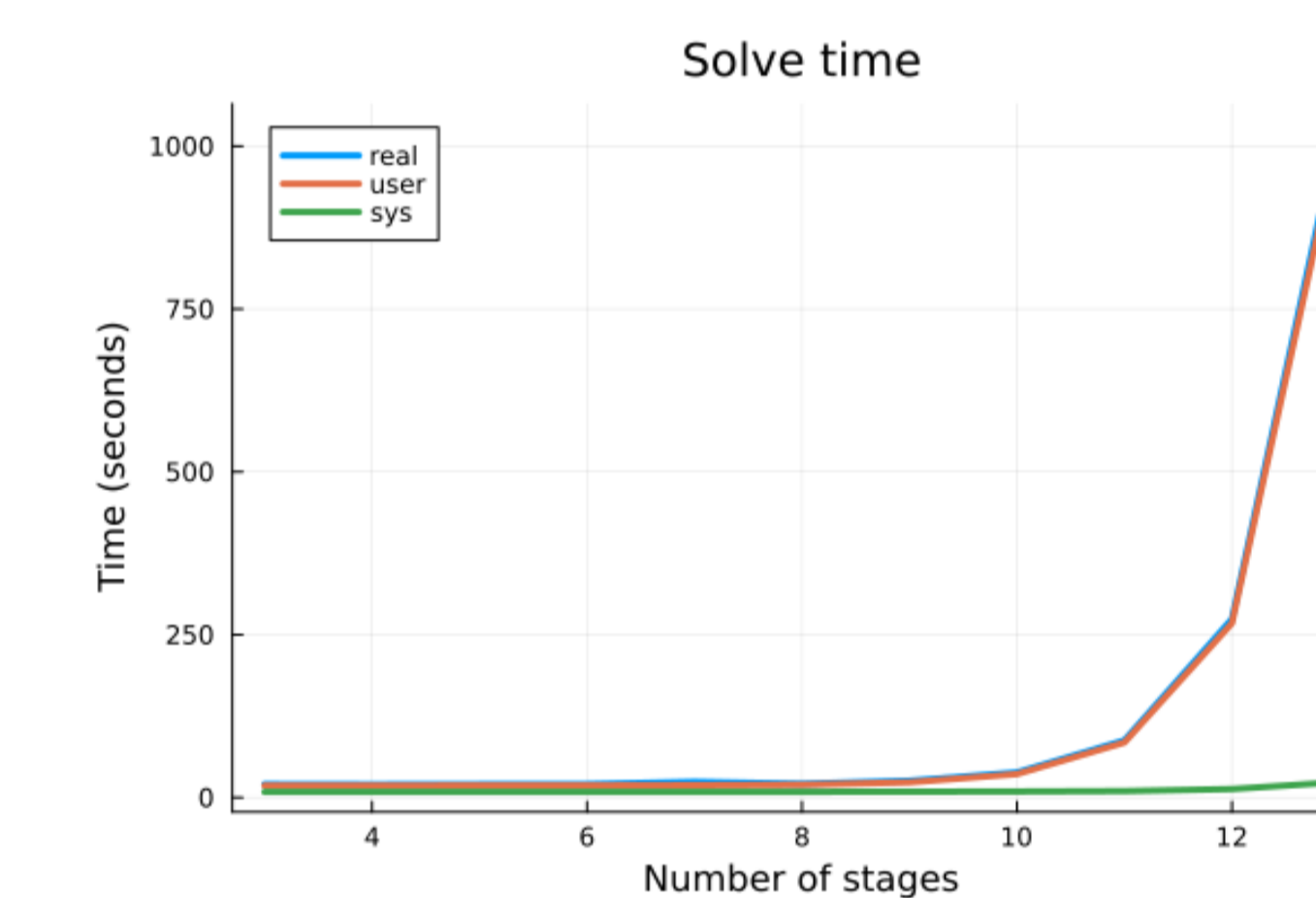


Figure 5:Solve time for Scenario Tree

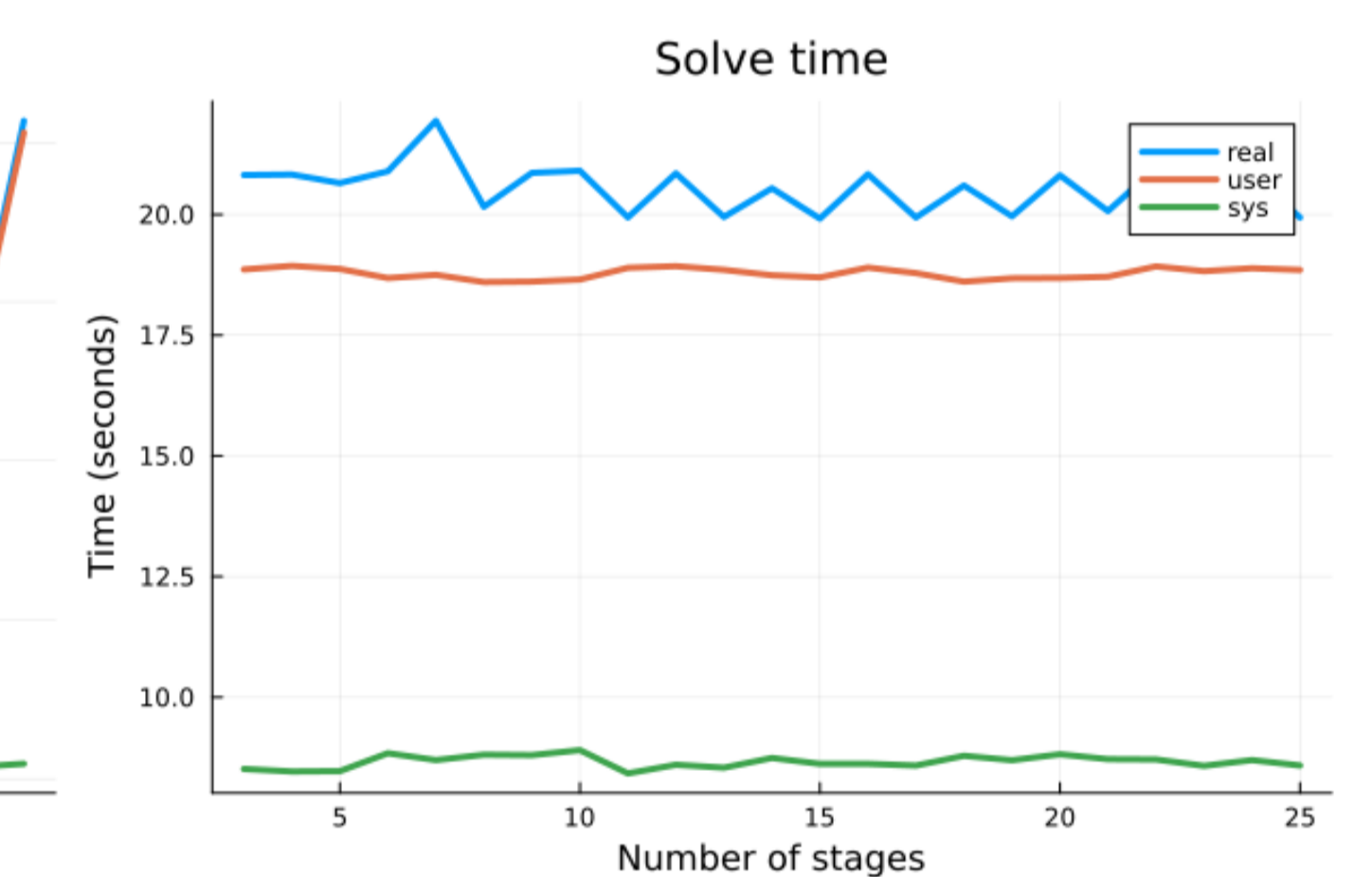


Figure 6:Solve time for Scenario Lattice

### Future directions:

- Generate cuts to decrease the feasible region & speed up computations.
- Apply breakstages (Zou, J. et al. (2018)).

## References

- Birge, J. R., & Louveaux, F. (2011). *Introduction to Stochastic Programming*. Springer Science & Business Media.
- Kim, K. & Zavala, V.M. (2018). *Algorithmic innovations and software for the dual decomposition method applied to stochastic mixed-integer programs*. Math. Prog. Comp. 10, 225-266.
- Zou, J., Ahmed, S., & Sun, X. A. (2018). *Multistage Stochastic Unit Commitment using SDDiP*. IEEE Transactions on Power Systems, 34(3), 1814-1823.

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