

A Discrete-Time Scheduling Model for Continuous Power-intensive Process Networks with Various Electricity Contracts

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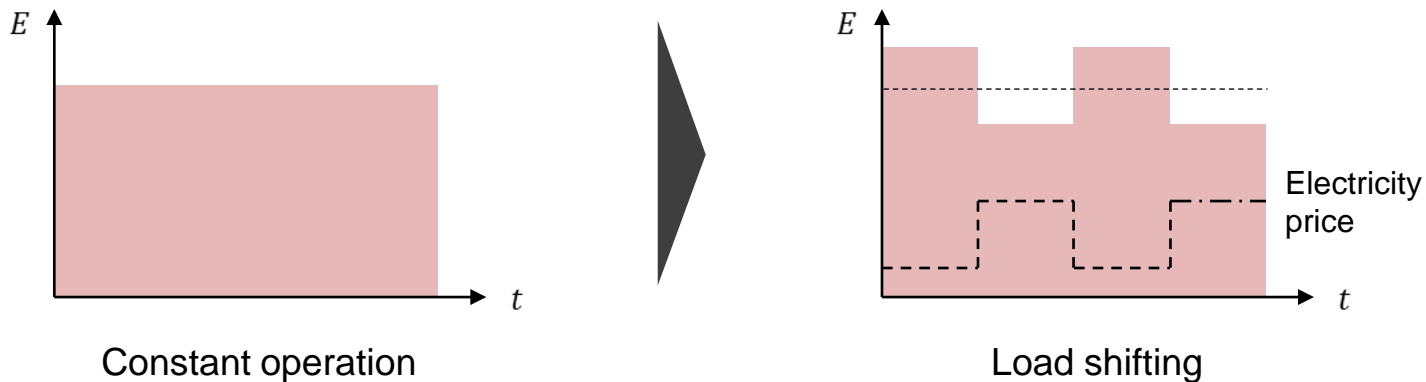
Arul Sundaramoorthy, Jose M. Pinto

Praxair, Inc., Business and Supply Chain Optimization R&D

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In the era of Smart Grid, efficient scheduling models for industrial power-intensive processes are crucial.

- In the context of Demand Side Management, industrial power-intensive processes can achieve large benefits by optimizing their power consumption
- The plant's flexibility can be used to shift load depending on the time-sensitive prices and the conditions of the electricity contracts

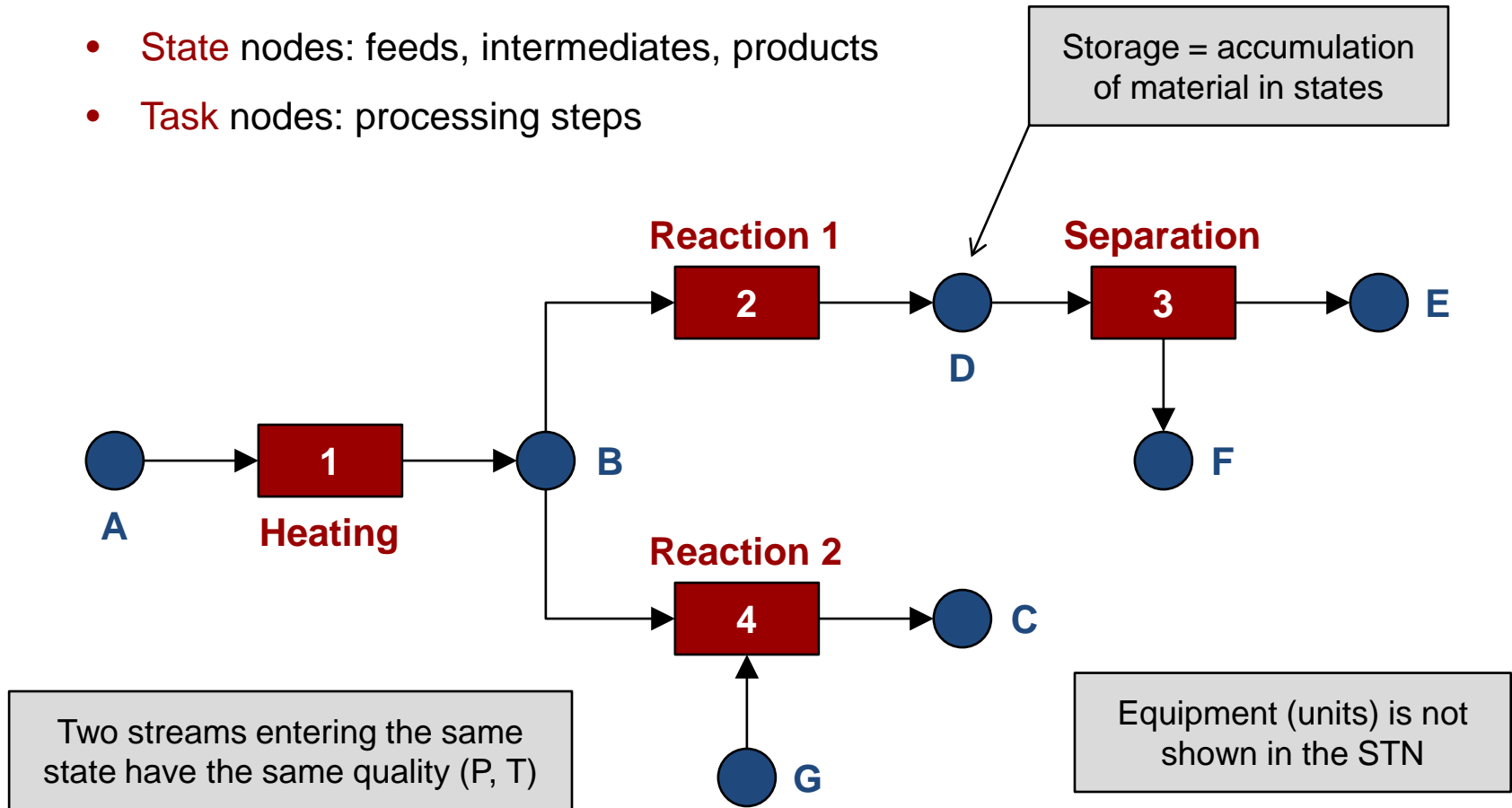


We develop a discrete-time scheduling model for general continuous power-intensive process networks considering various electricity contracts.

To model the process network, we apply the concept of the state-task network (STN).

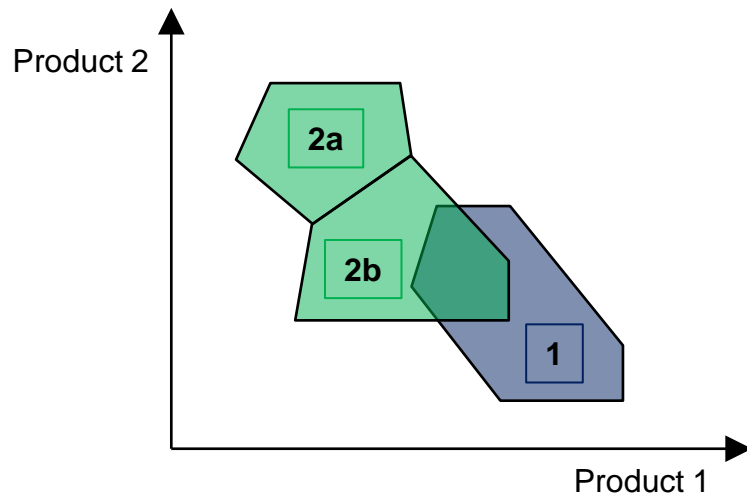
- Two types of nodes:

- State nodes: feeds, intermediates, products
- Task nodes: processing steps



Each equipment unit is modeled as a set of Convex Region Surrogate (CRS) models.

- Assume that a piece of equipment can run in different **operating modes**
- The feasible region of each operating mode is expressed as a union of convex regions in the **state space**
- For each region, the required **power consumption** is a linear function of the states
- Can be stated as a nested disjunction, which can be reformulated as a set of MILP constraints¹

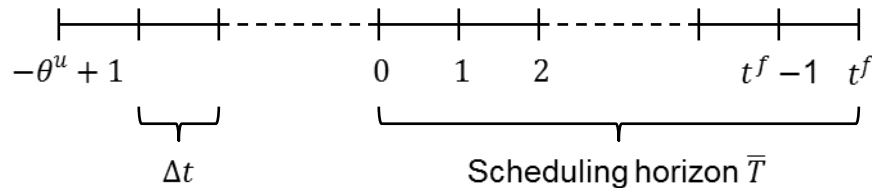


$$\begin{aligned}
 & \bigvee_{m \in M_{ij}} \left[\bigvee_{r \in R_{ijm}} \left(\begin{array}{l} Y_{ijmt} \\ \bar{Y}_{ijmrt} \\ P_{ijst} = \sum_{l \in L_{ijmr}} \lambda_{ijmrlt} \phi_{ijmrls} \quad \forall s \in S_i \\ \sum_{l \in L_{ijmr}} \lambda_{ijmrlt} = 1 \\ 0 \leq \lambda_{ijmrlt} \leq 1 \quad \forall l \in L_{ijmr} \\ U_{ijt} = \delta_{ijmr} + \sum_{s \in S_i} \gamma_{ijmrs} P_{ijst} \end{array} \right) \right] \quad \forall i, j \in J_i, t \in \bar{T} \\
 & \bigvee_{m \in M_{ij}} Y_{ijmt} \quad \forall i, j \in J_i, t \in \bar{T} \\
 & Y_{ijmt} \Leftrightarrow \bigvee_{r \in R_{ijm}} \bar{Y}_{ijmrt} \quad \forall i, j \in J_i, m \in M_{ij}, t \in \bar{T} \\
 & Y_{ijmt} \in \{true, false\} \quad \forall i, j \in J_i, m \in M_{ij}, t \in \bar{T} \\
 & \bar{Y}_{ijmrt} \in \{true, false\} \quad \forall i, j \in J_i, m \in M_{ij}, r \in R_{ijm}, t \in \bar{T}
 \end{aligned}$$

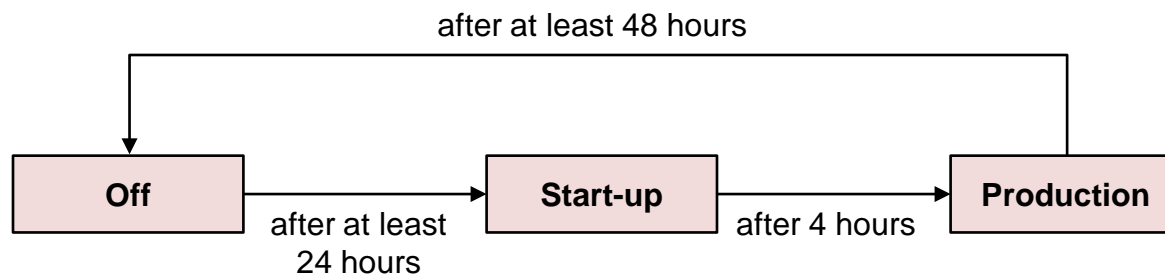
1. Grossmann and Trespalacios (2013). *AIChE Journal*.

We apply a discrete-time formulation to model transitions between operating modes¹.

- The time horizon is divided into time periods of equal length (typically one hour)



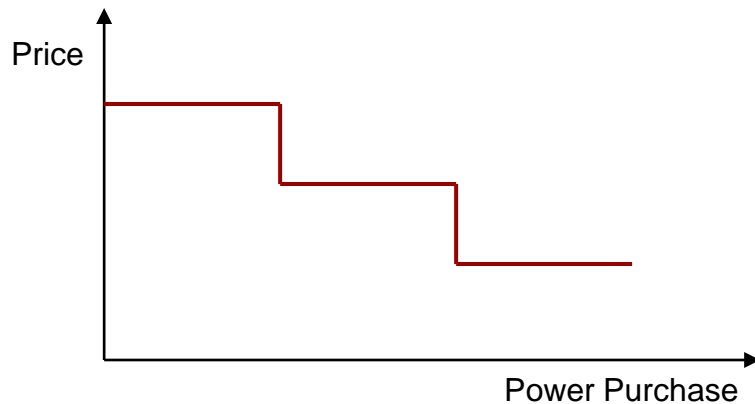
- For every time period, **operating mode** and **production rates** are determined
- Constraints on **mode transitions** can be imposed



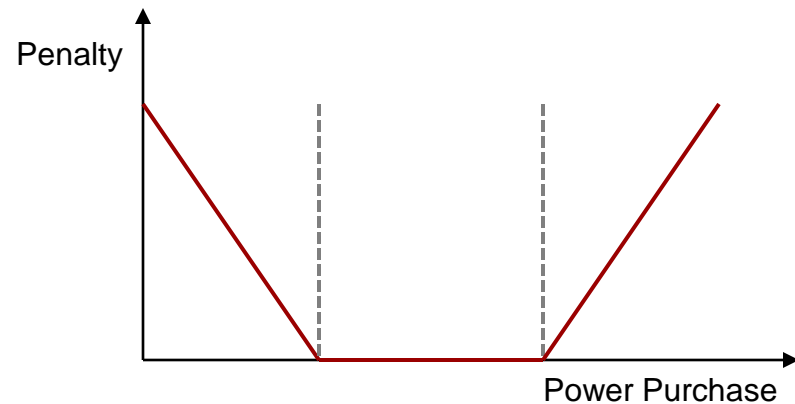
1. Mitra et al. (2012). *Computers and Chemical Engineering*.

A large variety of electricity contracts can be modeled by using a block contract formulation.

- Assigns prices to every time period
- Tracks cumulative power purchases for predefined meter reading times
- Can accommodate discount prices that depend on the amount of power purchased
- Can accommodate penalties for under- and overconsumption

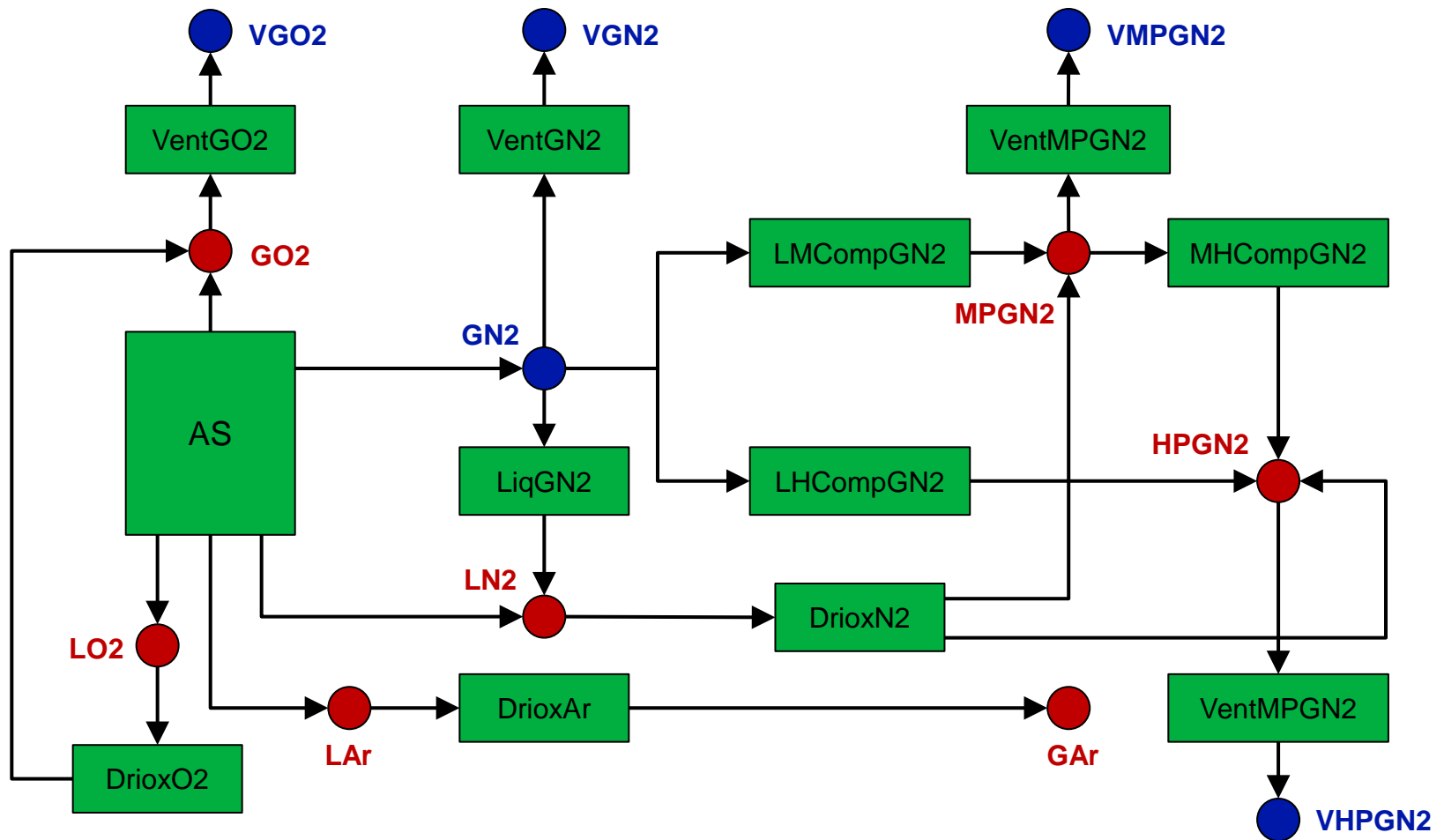


Discount Contract



Penalty Contract

Air Separation Case Study: State-Task Network



Air Separation Case Study: Results

