



Carnegie Mellon



Decomposition Approaches for Optimal Production – Distribution Coordination of Industrial Gases Supply Chains

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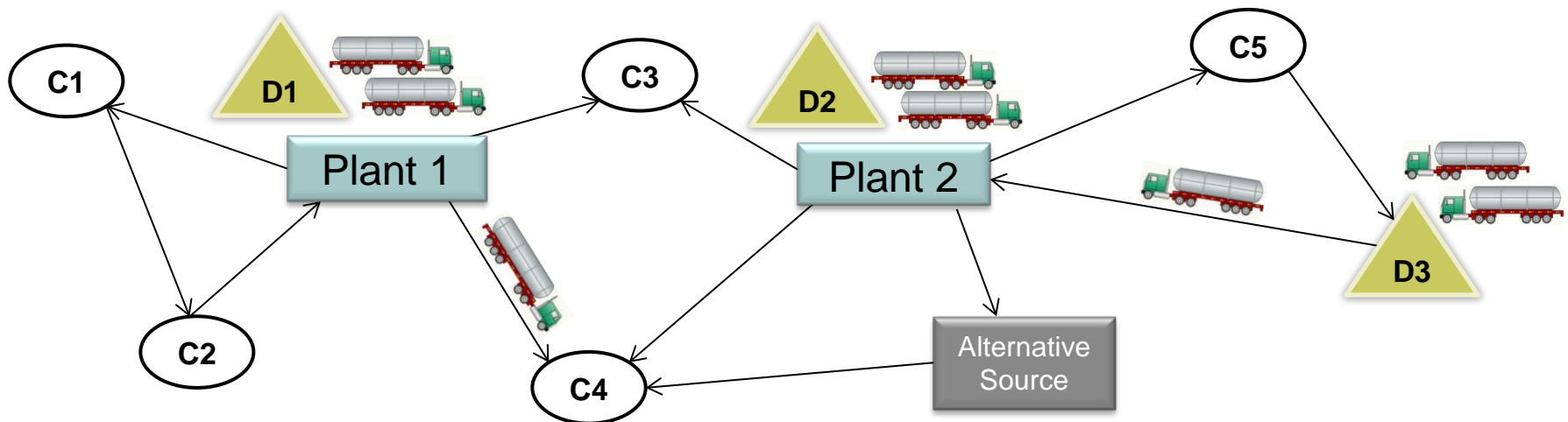
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Center for Advanced Process Decision-making
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Background and Motivation

Industrial Gases Supply-Chain

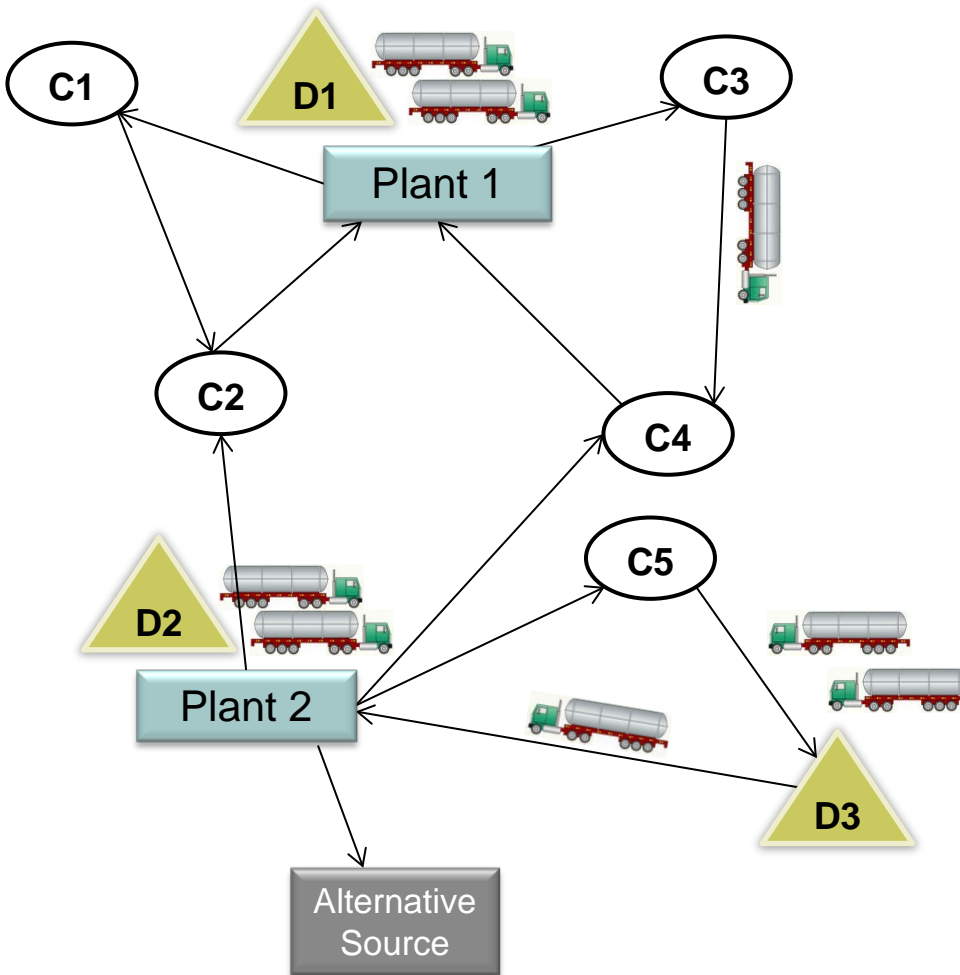
- ❑ Multiple Plants and Depots (located or not at plants)
- ❑ Multiple Products (*LIN, LOX etc.*) and Product Grades
- ❑ *Coordination among plants, distribution options and customers*
- ❑ Large computational effort (> 12 hrs. and > 4% gap)
- ❑ Large combinatorial problem



Goal

*Apply decomposition strategy in order to reduce the computational effort and improve the **optimal decision making** in complex production distribution problems*

Problem Statement and Main Assumptions



Given

- ❑ Plants, Products, Operating Modes and Production Limits
- ❑ Daily Electricity Prices (off-peak and peak)
- ❑ Customers and their demand/consumption profiles
- ❑ Max/Min inventory at production sites and customer locations
- ❑ Alternative sources and product availabilities
- ❑ Depots, Truck availabilities and capacities, Distances
- ❑ Fixed Planning Horizon (usually 1-2 weeks)

Decisions in each time period t

- ❑ Modes and production rates at each plant
- ❑ Inventory level at customer location and plants
- ❑ How much product to be delivered to each customer through which route

Objective Function

- ❑ Minimize total production and distribution cost over planning horizon

Main Assumptions – Distribution Side

- ❑ Two time periods per day (peak and off-peak) are considered
- ❑ Trucks do not visit more than 4 customers in a single delivery

Mathematical Model (MILP)

Objective

Minimize total Production and Distribution Costs

Constraints on Production Side

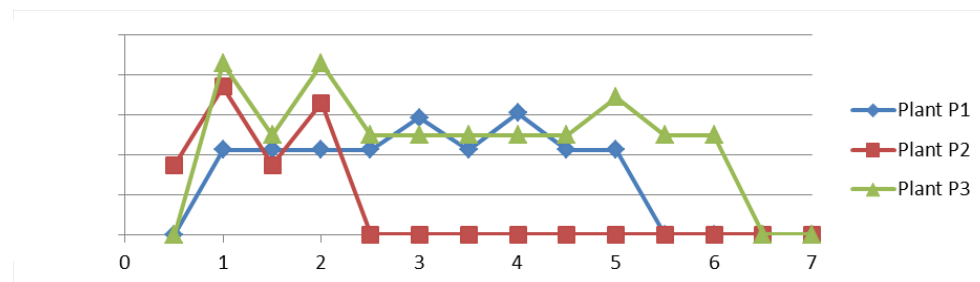
Production Cost = Fixed Start-up cost
+ Variable production cost

- Min/Max Production Capacity Constraints** in each mode of operation
- Logic Constraints** for switching between various modes of operation
- Max/Min Inventory limits** at the production sites
- Plant Inventory Balance Constraints**
- Demand satisfaction for pick-up customers**
- Ad Hoc Production Models**
 - To account for specific equipment configurations and production modes.

Main binary variables

mode selection

B_{pmt}



Mathematical Model (MILP)

Constraints on Distribution Side

Distribution Cost = Cost of deliveries by trucks
+ purchases from competitors

- ❑ Max/Min Inventory limits at the customer locations
- ❑ Customer Inventory Balance Constraints
- ❑ Truck Capacity constraints
- ❑ Material balance constraints for product pick-up and delivery points
- ❑ Max product purchase limit from competitor sources

Main binary variables

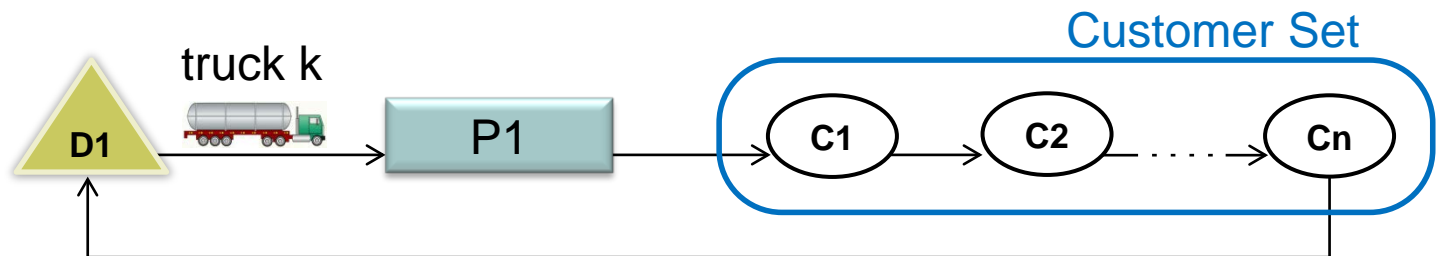
truck-customers
association

$$y_{kst}$$

truck-source
association

$$Y_{kpt}$$

Route = Depot + Plant + Customers



Route Generation Algorithm

- ❑ Parameters: maximum distance, number of customers visited, etc.
- ❑ Main ordering criteria: traveling distance

Industrial Size Test Case (previous work)

- ❑ 4 Plants / Depots
 - ❑ 2 products (LIN, LOX)
 - ❑ 2-4 production modes for each plant
 - ❑ 15 alternative sources
 - ❑ 229 customers
 - ❑ 14 time periods (peak and off-peak)
 - ❑ 46 trucks (25 for LIN, 12 for LOX)
 - ❑ Min/max inventory, distances, electricity prices, truck deliveries, etc.
- 19 sources
- 10,032 possible routes **reduced** to 1,900 by the route generation algorithm

12 h gap > 4% !

Model Size	Binary variables	26,486
	Continuous variables	69,902
	Constraints	35,233

CPU results	Time	7978 s
	Nodes	11,537
	Relative gap	10 %

*Objective: propose a **decomposition** strategy in order to reduce the computational effort*

Models implemented with GAMS
CPU results obtained with solver CPLEX 12.3

Rolling Horizon approach

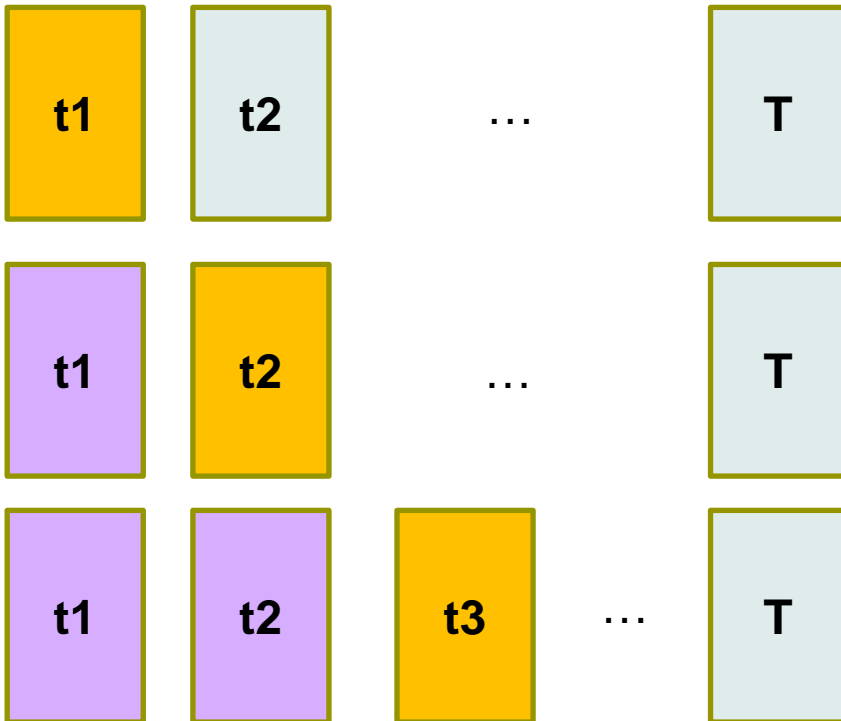
The detailed time block:

$X \in \{0, 1\}$ (binary variable)

The aggregated time block:

$X' \in [0, 1]$ (continuous variable)

Algorithm:



Full space

$$\min TotalCost(x, y)$$

s.t.

$$Ax_s^t = Bx_s^{t-1} \quad \forall s \in S, t \in T$$

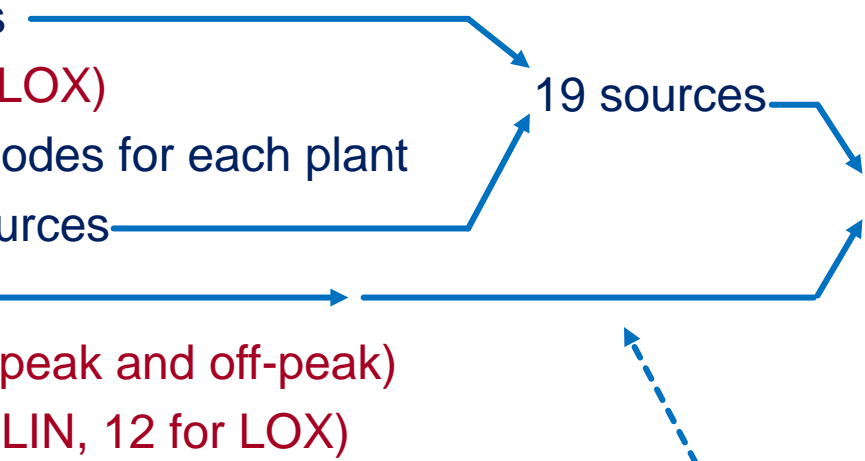
$$Cx_s^t = Dy_s^t \quad \forall s \in S, m \in M, t \in T$$

Note:

- The planning horizon in detail (“the detailed time block”),
- Rest of the horizon is represented in an aggregated manner (“the aggregate time block”).

$\forall t \in T \rightarrow$ Sub-problems

Industrial Size Test Case – Results

- ❑ 4 Plants / Depots
 - ❑ 2 products (LIN, LOX)
 - ❑ 2-3 production modes for each plant
 - ❑ 15 alternative sources
 - ❑ 229 customers
 - ❑ 14 time periods (peak and off-peak)
 - ❑ 46 trucks (25 for LIN, 12 for LOX)
 - ❑ Min/max inventory, distances, electricity prices, truck deliveries, etc.
- 19 sources → 10,032 possible routes reduced to 1,900 by the route generation algorithm
- 

		Original	Rolling H
Model Size	Binary variables	26,486	-
	Continuous variables	69,902	-
	Constraints	35,233	-
	Normalized total cost	100	102

CPU results	Time	7,978 s	1,147 s
	Nodes	11,537	1,278
	Relative gap	10 %	1.7 %

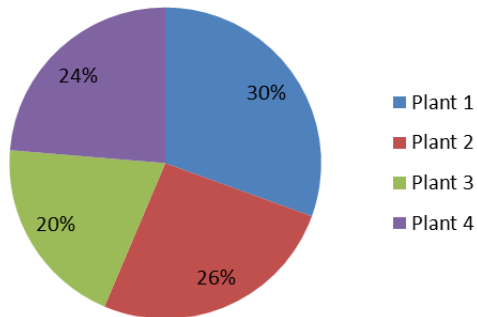
Models implemented with GAMS CPU results obtained with solver CPLEX 12.3

RH: 7 times faster!
1.8% worse

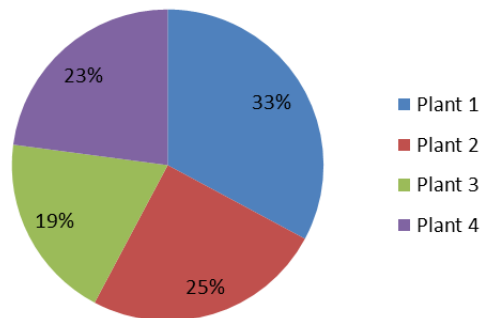
Industrial Size Test Case – Results

Production level

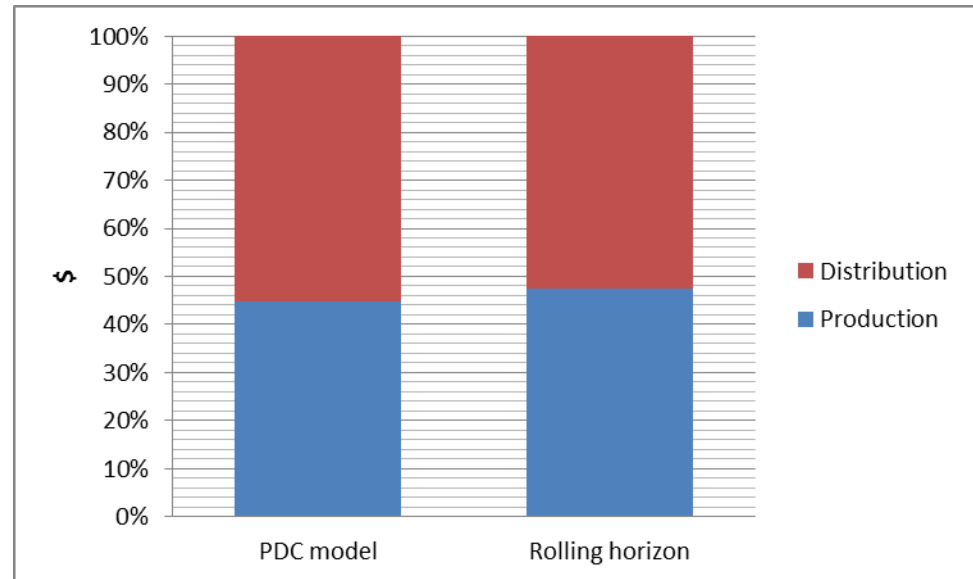
Original PDC model



Rolling Horizon



Total Cost (original model 1.83 % lower)



*A **Novel** Framework for **coordinated** production-distribution decision making has been developed*

Remarks and Conclusions

The proposed framework provides optimal production and distribution coordination reducing the computational effort

- ❑ The proposed rolling horizon is computationally faster than simultaneous approach with small penalty for suboptimality
- ❑ Lagrangean Decomposition might be also promising but for problems with strong linking no feasible solutions have been encountered.

Work under development

- ❑ Investigate extension of Lagrangean Decomposition to handle strong linking among the time periods (start up detection, high inventory levels, and high final inventory or safety stock considered).
- ❑ Comprehensive comparison between Rolling horizon vs Lagrangean Decomposition