



Day 1

Model and Process Uncertainty: Quantification and Usage

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FIPSE 2018



Summary



Design vs. Operations/control problems

- Design: make decisions *once* – *have to get it right*
- Operations/control: make decisions iteratively – *we have recourse*

Operations/control problems

- Open-loop and closed-loop problems are very different problems
- Which factors impact (the most) the quality of implemented (online) solutions?
Frequency of reoptimization, quality of model, quality of uncertainty representation, optimality gap, horizon?
- What can we do to obtain better solutions?
Model uncertainty and solve multi-stage stochastic programming problem?
Solve robust, adjustable robust optimization?
What is a *good enough* planning horizon?
Solve to optimality?
Solve to suboptimal solutions but faster?



Design Problem



Example

- A manufacturing facility is used for the production of chemicals A and B
- The capacity of the facility is 10 tons/year;
Can be expanded up to 12 tons/year at \$100,000
- The demand for product A is expected to be constant: $D_A = 4$ ton/year
- The demand for product B is uncertain:
Three scenarios (analysts): $D_B^{Low} = 4$, $D_B^{Med} = 6$ and $D_B^{High} = 8$ ton/year
- Profit from sales: Drug A = \$100,000/ton & Drug B = \$200,000/ton

Should we expand the facility?

Calculations using expected demand

✓ Expected demand: $D_A = 4$, $D_B = 6$ ton/year \Rightarrow **No expansion**

✓ Expected profit:

- If $D_A = 4$, $D_B = 4$ ton/year $\Rightarrow P_A = 4$, $P_B = 4$ ton/year \Rightarrow Profit = \$1,200,000
- If $D_A = 4$, $D_B = 6$ ton/year $\Rightarrow P_A = 4$, $P_B = 6$ ton/year \Rightarrow Profit = \$1,600,000
- If $D_A = 4$, $D_B = 8$ ton/year $\Rightarrow P_A = 4$, $P_B = 6$ ton/year \Rightarrow Profit = \$1,600,000

✓ Objective = E[Profit] – Expansion Cost = **\$1,466,667**



Design Problem



Example

- A manufacturing facility is used for the production of chemicals A and B
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- The demand for product B is uncertain:
Three scenarios (analysts): $D_B^{\text{Low}} = 4$, $D_B^{\text{Med}} = 6$ and $D_B^{\text{High}} = 8$ ton/year
- Profit from sales: Drug A = \$100,000/ton & Drug B = \$200,000/ton

Should we expand the facility?

Calculations using scenarios

- ✓ Consider expansion because demand can be for B can be as high as 8 ton/year
- ✓ Scenarios with expansion:
 - If $D_A = 4$, $D_B = 4$ ton/year $\Rightarrow P_A = 4$, $P_B = 4$ ton/year \Rightarrow Profit = \$1,200,000
 - If $D_A = 4$, $D_B = 6$ ton/year $\Rightarrow P_A = 4$, $P_B = 6$ ton/year \Rightarrow Profit = \$1,600,000
 - If $D_A = 4$, $D_B = 8$ ton/year $\Rightarrow P_A = 4$, $P_B = 8$ ton/year \Rightarrow Profit = **\$2,000,000**
- ✓ Objective = E[Profit] – Expansion Cost = 1,600,000 – 100,000 = **\$1,500,000**



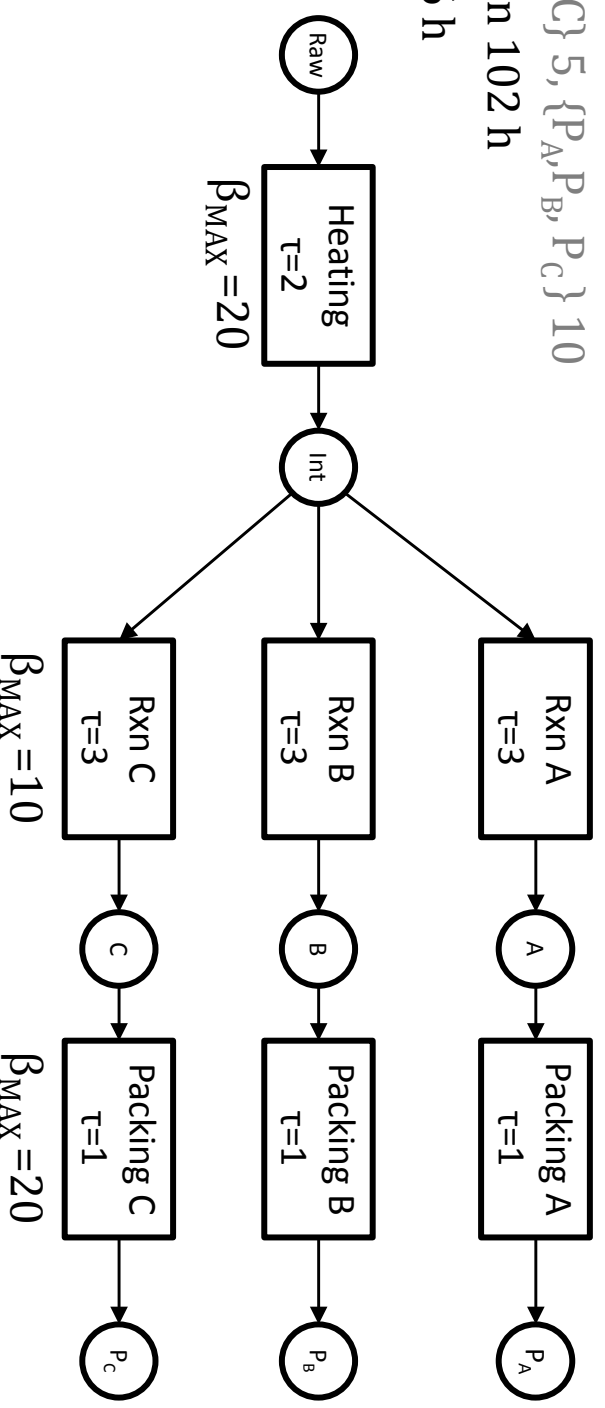
Scheduling Problem



Example solved with (30 repetitions):

- Offline deterministic (nominal)
- Offline robust
- Offline stochastic programming (16 scenarios)
- Perfect information
- Online deterministic
- Online robust
- Online stochastic

- Facility: 2 heaters, 3 reactors, 2 packing lines; no storage
- Demand: nominal = 30 kg for P_A, P_B, P_C , simultaneously, at every 18 time steps; uniformly distributed with 50% uncertainty in demand magnitude; true demand known 12 time points ahead
- Prices: Raw 0, Int 1, $\{A, B, C\}$ 5, $\{P_A, P_B, P_C\}$ 10
- Total (simulation) horizon 102 h
- Rolling horizon length 36 h





Method Comparison



Closed-loop average costs

- Offline deterministic: \$160K
 - Offline robust: \$350K
 - Offline stochastic: \$110K
 - Perfect information: \$30K
 - Online deterministic: \$85K
 - Online robust: \$80K
 - Online stochastic: \$70K
- **Adding feedback brings the most benefit**
 - **The choice of model/method is secondary**



What Else Can We Do?



- **Stochastic optimization is (typically) computationally more expensive**
- **Can/should we solve a more *useful*, and equally expensive, problem?**
 - e.g, Deterministic with longer planning horizon?
 - Deterministic over a larger system?
- **Can/should we be doing anything else?**
 - e.g, Solve a deterministic problem more frequently?

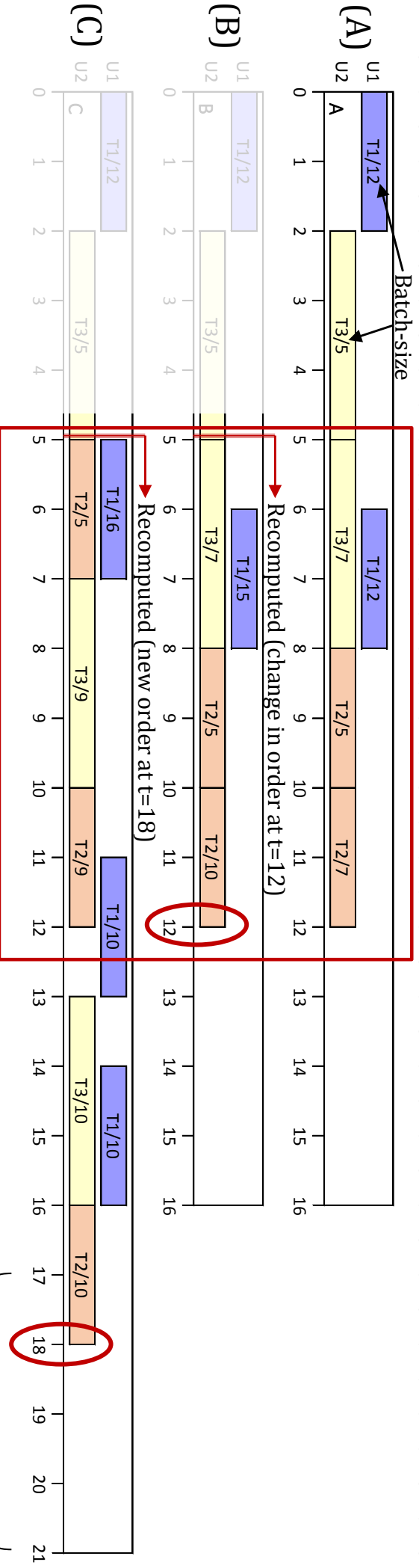
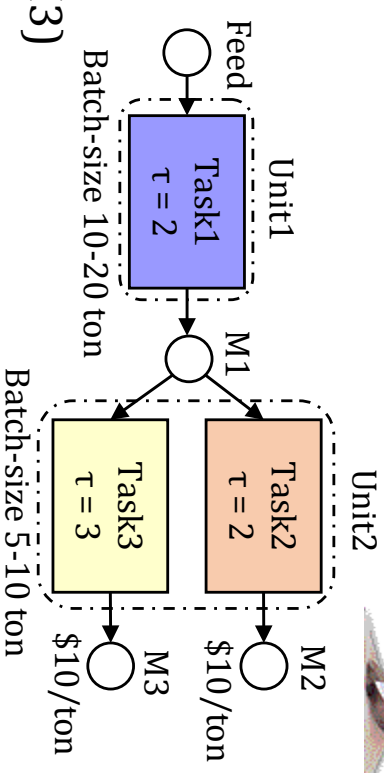


Uncertainty vs. New Information



Experiment #1

- Orders (12 tons M2, M3) due $t=12, 18, 24, \dots$
- Use horizon $H = 16$ h
- (A) Original schedule computed at $t=0$.
- (B) Observation at $t=5 \rightarrow$ order due at $t=12$ changes (+3 tons M3)
Re-compute schedule - shrinking horizon
- (C) At $t=5 \rightarrow$ order at $t = 12$ unchanged
Re-compute schedule - moving horizon



- (A, C) much more different than (A, B)
- Accounting for new information can be more important than uncertainty*
- \Rightarrow Use moving horizon long planning horizon*



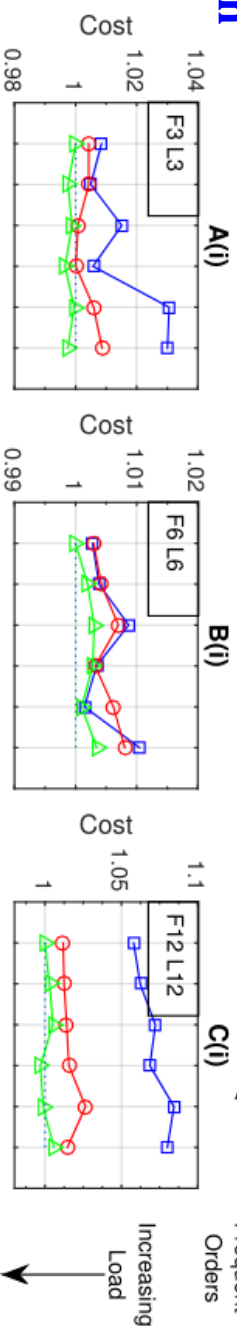
Rescheduling Frequency

Orders: every 3 hours

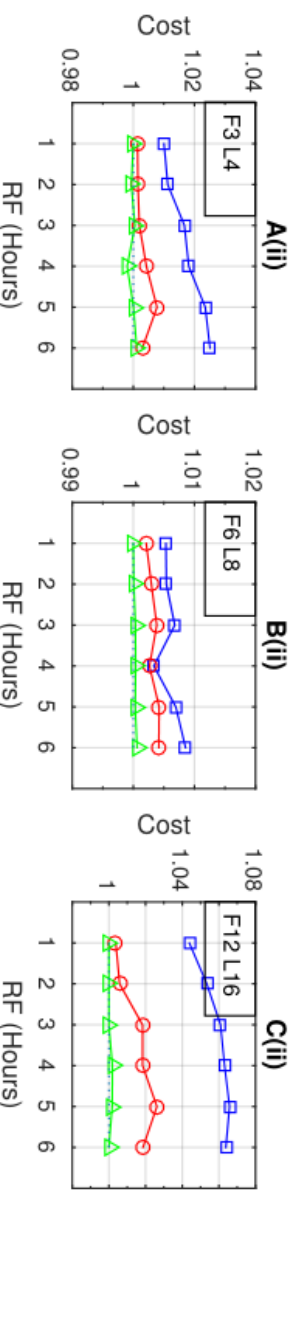
Orders: every 6 hours

Orders: every 12 hours

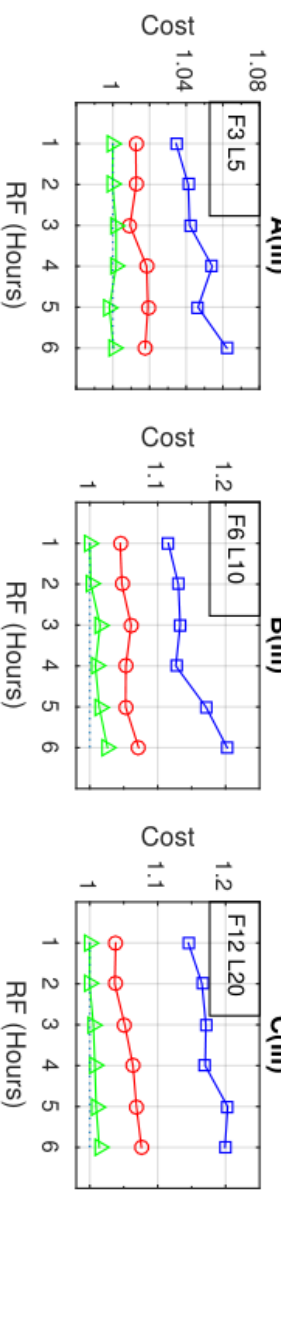
Cost minimization



Load: 50% of capacity

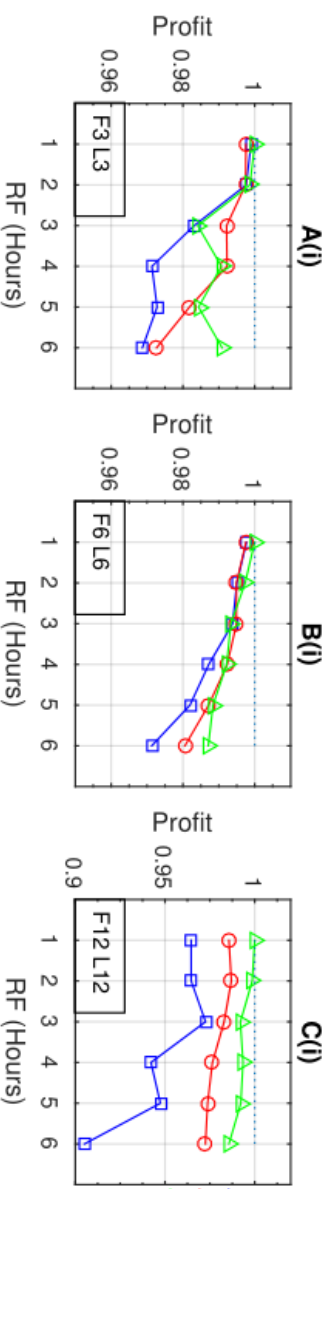


Load: 75% of capacity



Load: 100% of capacity

Profit maximization



Load: 50% of capacity to meet demand, remaining capacity for extra sales

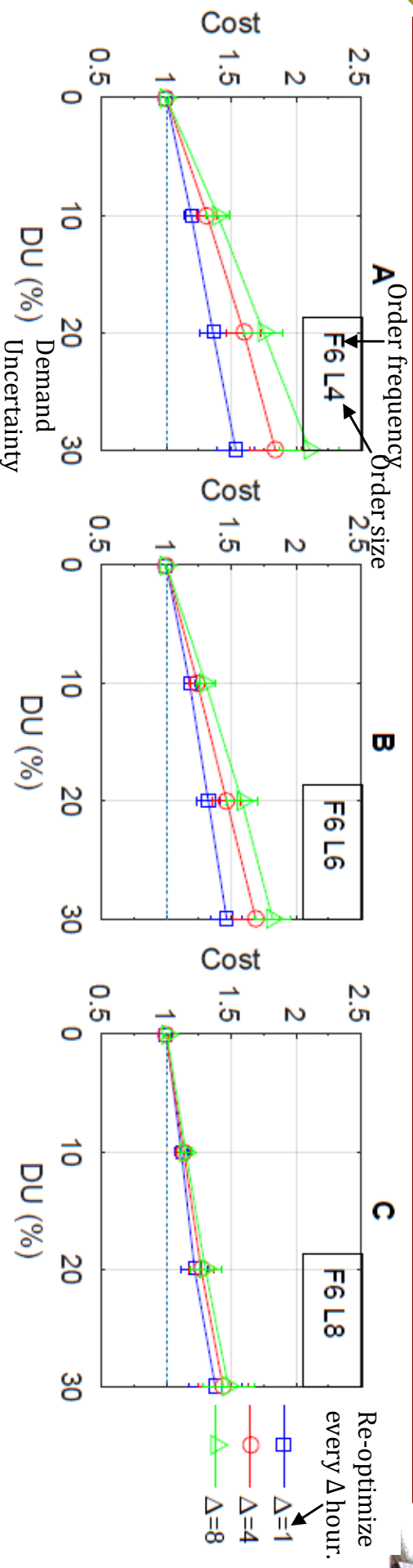
¹All open-loop solved to optimality.

²Each data point mean of 10 closed-loop runs (1 week each).





Demand Uncertainty & Effect of Rescheduling Frequency



- ❑ Larger Demand Uncertainty (DU) increases closed-loop cost
- ❑ Frequent re-optimization is beneficial for “tackling” uncertainty (nominal model)



Design Framework for Online Scheduling Methods



- ❑ Operations are always reactive (or have a reactive element)
- ❑ Deterministic models can “handle” uncertainty through feedback & re-computation
- ❑ Re-optimize early and often (even if uncertainty is modeled)
- ❑ Question: what constraints and objective function weighs can we add to *obtain* good closed-loop solution?

