

Hierarchical Routing Control in Discrete Manufacturing Plants Via Model Predictive Path Allocation and Greedy Path Following

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Motivation

Research in advanced manufacturing solutions is motivated by several trends:

- higher product customization
- more agile supply chains
- higher environmental sustainability



The problem of **routing control in discrete manufacturing plants** is considered in this paper.



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Problem description Control problem with large number of integer variables and temporal logic constraints





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Literature Optimality vs. Scalability

Approaches in the literature include:

- Rule-based techniques (Gupta et al., 1989, Byrne et al., 1997, Saygin et al., 2001, Bucki et al., 2015, Souier et al., 2010);
- Integer programming (Das et al., 1997);
- Multi-agent architectures (Kouiss et al., 1997);
- Heuristic search combined with Petri nets (Moro et al., 2002);
- MPC (Cataldo and Scattolini, 2016)

Tradeoff between optimality and scalability



Contribution Scalable predictive approach with hierarchical problem decomposition





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"Lagrangian" model of the plant A change of perspective

- "Eulerian" model (most common framework):
- Each binary state corresponds to a node (1=part is present)
- Each binary input is a transition (1=transition occurs from time k to time k + 1)



 Results in rather large-scale MILP or MIQP "Lagrangian" model (introduced in this paper):

- Each state is linked to a part on the plant
- States for each part *i* :
 - Current sequence s_i,
 - Position p_i along the sequence;
 - Elapsed time t_i since the part entered the plant.



Greedy path following strategy

The Lagrangian state $X_{N_{p(k)}}(k)$ is conveniently used by a low-level strategy that ensures satisfaction of all constraints:

- 1. Try to propagate forward all parts according to their current sequences;
- 2. Detect and resolve any conflicts
 - Parts held in place have highest priority
 - Parts that are more advanced in their sequence have 2nd highest priority;
 - Parts with higher t_i values have 3^{rd} highest priority
- 3. Compute accordingly the plant inputs







PC

Model Predictive Path Allocation Close loop strategy

- At time k, solve the FHOCP (unconstrained integer program of small-medium size)
- 2. Change the Lagrangian state from the current one to the one obtained ^k as solution to the FHOCP
- 3. Apply the greedy path-following strategy
- 4. Set $k \leftarrow k + 1$, go to 1.





Numerical example Plant description and employed sequence

- Parts must visit node 12, then 11, then exit
- Lockout may occur due to working time in the machine nodes and conflicts in nodes 6 and 7
- Employ one longer sequence obtained by merging shorter ones





Numerical example Simulation results



Computational time: approx 0.5 s per time step, with prediction horizon of 50 time steps (impractical with non-hierarchical approach)



Conclusions and next steps

- Novel "Lagrangian" modeling approach and hierarchical control structure shows optimal performance with scalable computation
- Next steps:
 - Apply to a real plant (undergoing);
 - Improve the solution to the predictive path allocation problem (undergoing);
 - Investigate the sequence generation problem;
 - Develop fault tolerant and robust extensions.



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