WIDE End-User Panel Meeting WP3: Distributed MPC

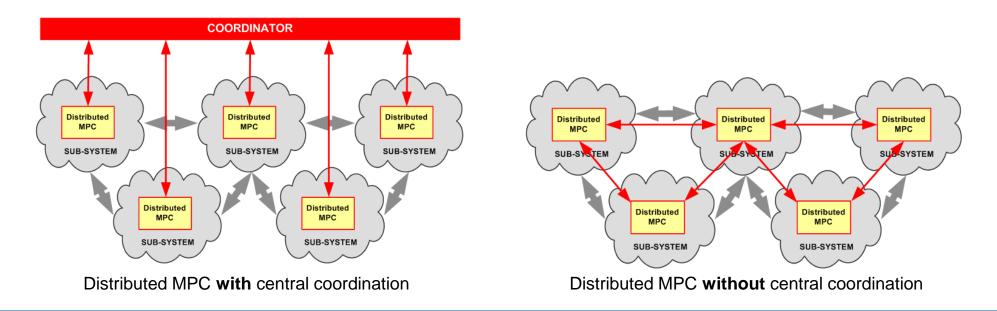
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Distributed MPC for large-scale systems

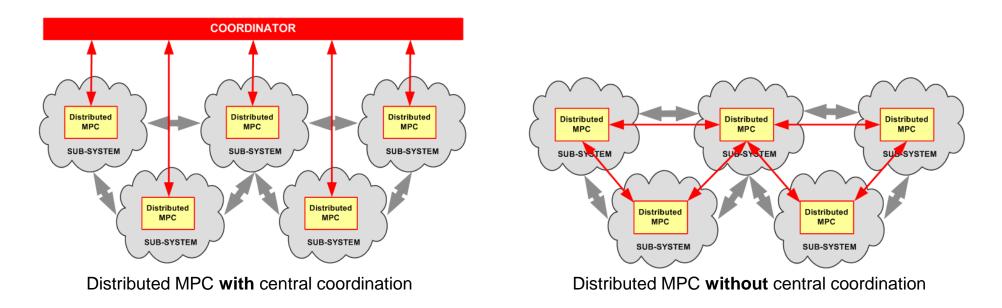
- The complexity of MPC increases quickly with system size (~O(n³))
- Application of MPC to large-scale systems requires distributed or decentralized solution
- Our effort was directed to distributed MPC:
 - + Minimum performance degradation to centralized MPC
 - Requires consensus iterations in each sampling period (more then one information exchange in each sampling period for every distributed MPC)
- Target systems with 10x seconds sampling period and hundreds of MV's



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Distributed MPC for large-scale systems

- Our solutions are based on a well known dual decomposition of optimization problems
- Original problem is distributed to multiple optimization problems, which are iteratively coordinated together until consensus is reached (price coordination)
- Efficient coordination algorithm is crucial
- Coordination algorithm types:
 - Centralized coordination: Novel algorithm with Parametric coordination
 - Distributed coordination: Nesterov accelerated gradient method

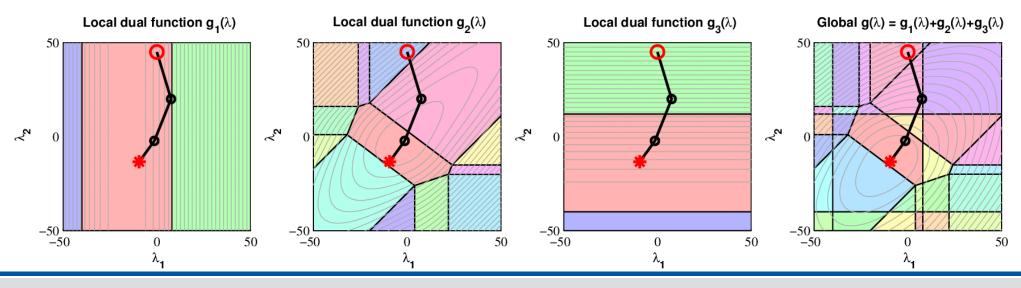


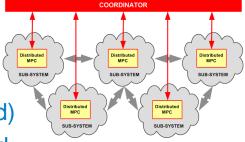
Parametric coordination

- For linear MPC only
- Properties:

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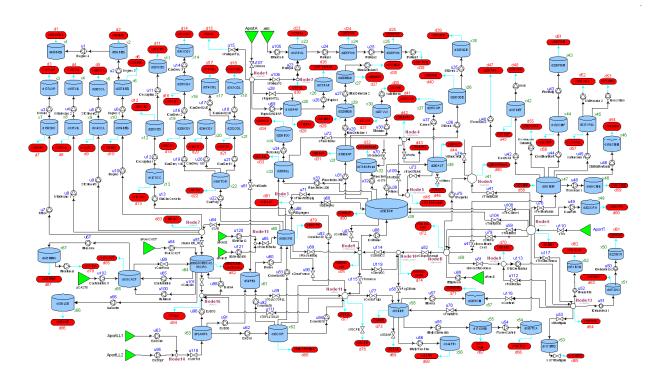
- Consensus reached in a finite number of iterations
- Fast algorithm with low number of iterations (Newton type method)
- Each iteration require only some local problems to be recomputed
- Consensus can be reached in a single iteration for warm start
- Coordination is based on multi-parametric programming
- Local optimizers (local MPC's) compute and return their solution in a requested point and its polytopic neighborhood
- Only required polytopes are computed (explicit solution of local MPC problems is too big)





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Parametric Coordination: Results on Barcelona Network



Water Tank Water Demand Water Source →⊕ Pump Valve

Overview:

- 70 tanks
- 15 main water nodes,
- 110 pumps/valves
- 80 demands (disturbances)
- 9 water sources

Control Objectives:

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- fulfill demands tank levels above safety level on prediction horizon
- minimize prices for:
 - fresh water
 - pumping (time varying electricity price)
- minimize MV changes, minimize tank levels
- constraints (pumps, valves, tanks, water sources)

MPC Optimization Problem Size

Quadratic programming problem imposed by MPC: (~120 MV's, ~80 DV's, ~70 CV's)

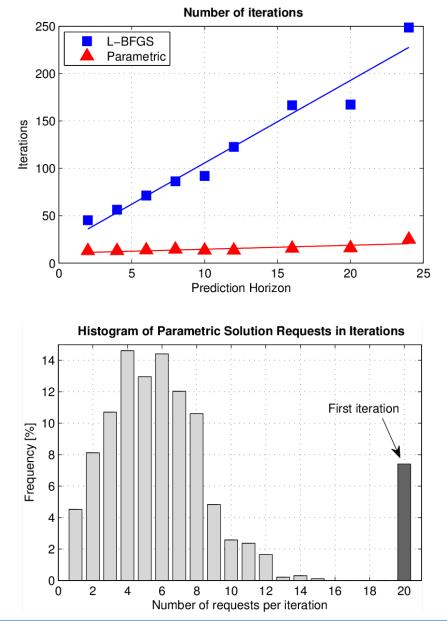
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Prediction horizon	24 hours
Sampling period	1 hour
Number of variables	~5.000
Number of constraints	~10.000

Parametric Coordination: Results on Barcelona Network

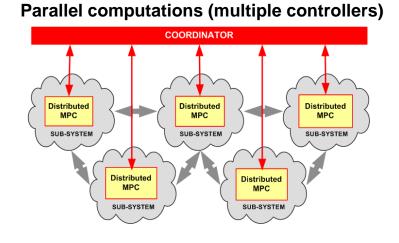
- Network was separated to 20 groups
- Each group has local MPC controller
- Same result with centralized MPC up to ~0.1% consensus error threshold
- Compared methods

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- Centralized MPC
- Distributed MPC with L-BFGS coordination
- Distributed MPC with Parametric coordination
- Results:
 - ~10x reduction in a number of iterations (compared to quasi-Newton L-BFGS)
 - Only ~25% of local problems are recomputed in each iteration
 - ~100x faster than centralized MPC and ~10x faster than L-BFGS



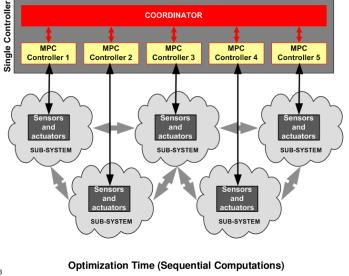
Parametric Coordination: Results on Barcelona Network

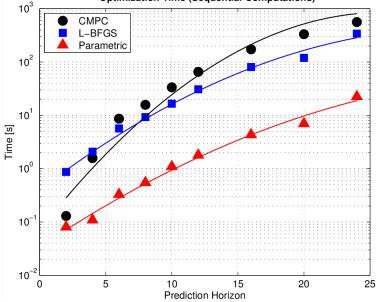


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Optimization Time (Paralel Computations) 10^{3} CMPC L-BFGS Parametric 10^{2} 10 Time [s] 10⁰ 10 10^{-2} 0 5 10 15 20 25 Prediction Horizon

Sequential computations (single controller)





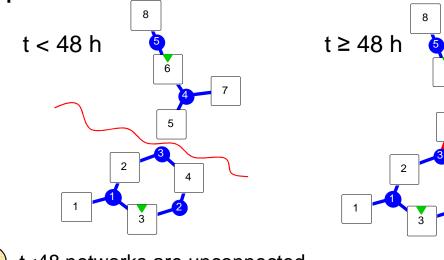
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Distributed coordination: Nesterov Accelerated Gradient Method

- Communication exchange with neighbors only
- Highly modular solution changes in system configuration without global control strategy adjustment
- 100x of iterations for Barcelona network (24 hrs horizon)

Example:

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- t<48 networks are unconnected
-) t≥48 networks are connected
- 3 t≥96 connecting pump is switched to MAN
- Blue line optimal control of connected networks Green line – optimal control of unconnected networks Red line – control of network with configuration changes

