

MULTI-AGENT CONTROL OF FLUID SYSTEMS

COMPARISON OF APPROACHES



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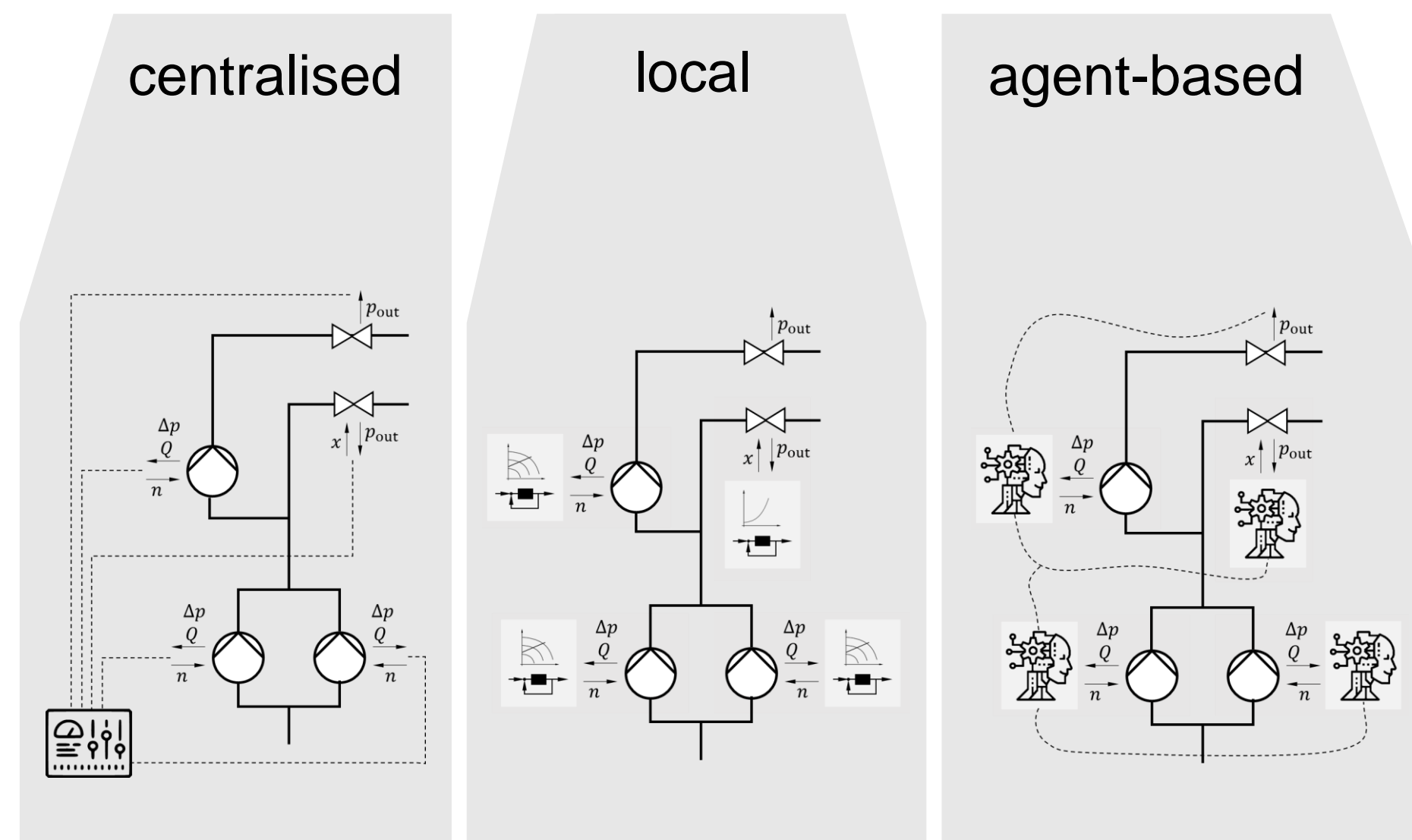
Kevin T. Logan, J. Marius Stürmer, Tim M. Müller, Tobias C. Meck, Peter F. Pelz
Chair of Fluid Systems, TU Darmstadt

Multi-agent systems allow system-wide, energy efficient control of fluid systems, fulfilling volume flow demands even in the face of disruptions within the communication network.

MOTIVATION

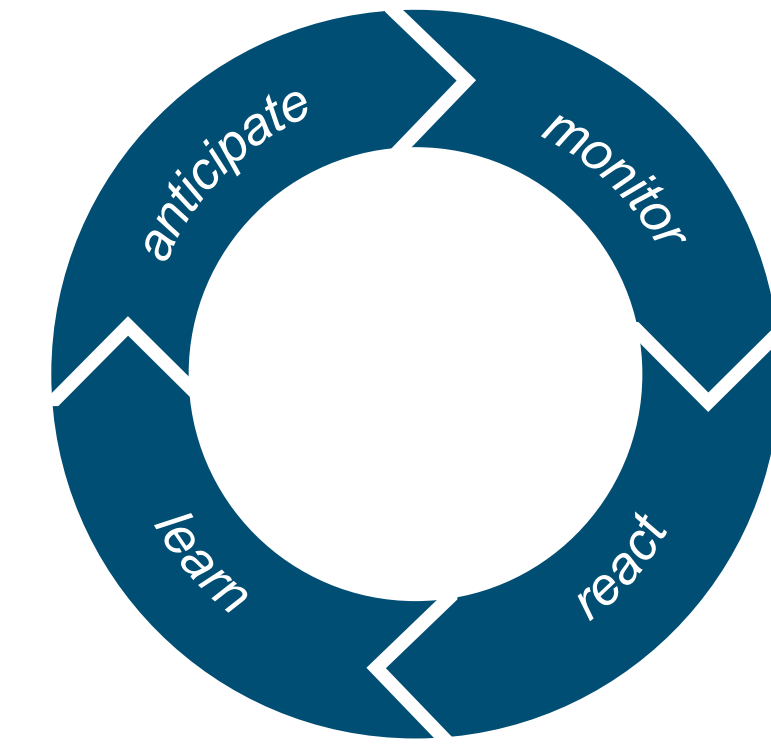
ensuring fluid systems fulfil their function while

- (i) minimising effort
 - set-up and start-up
 - energy consumption
- (ii) maximising acceptability
 - transparency
 - comprehensibility
 - traceability
- (iii) maximising availability in the face of disruptions



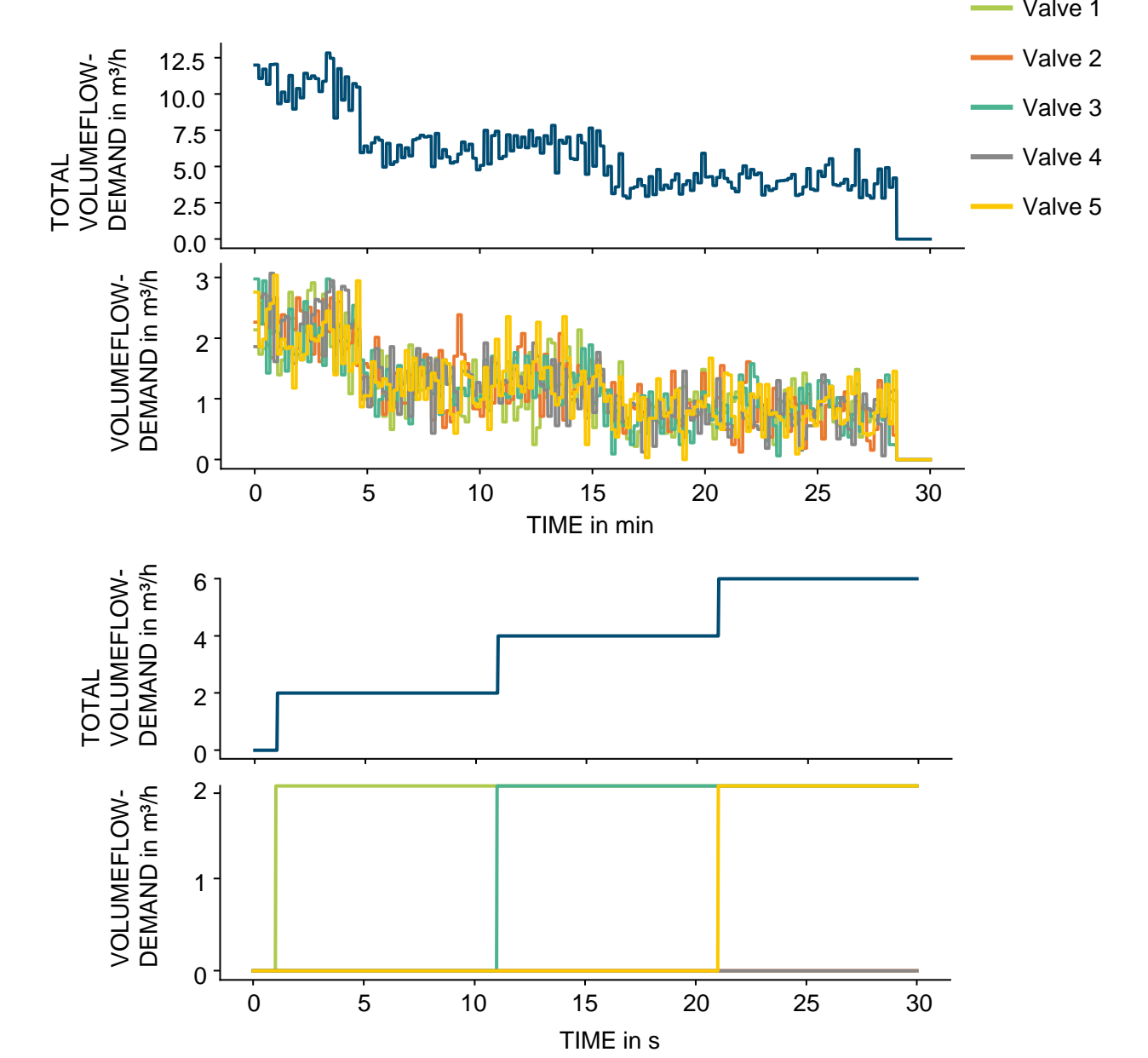
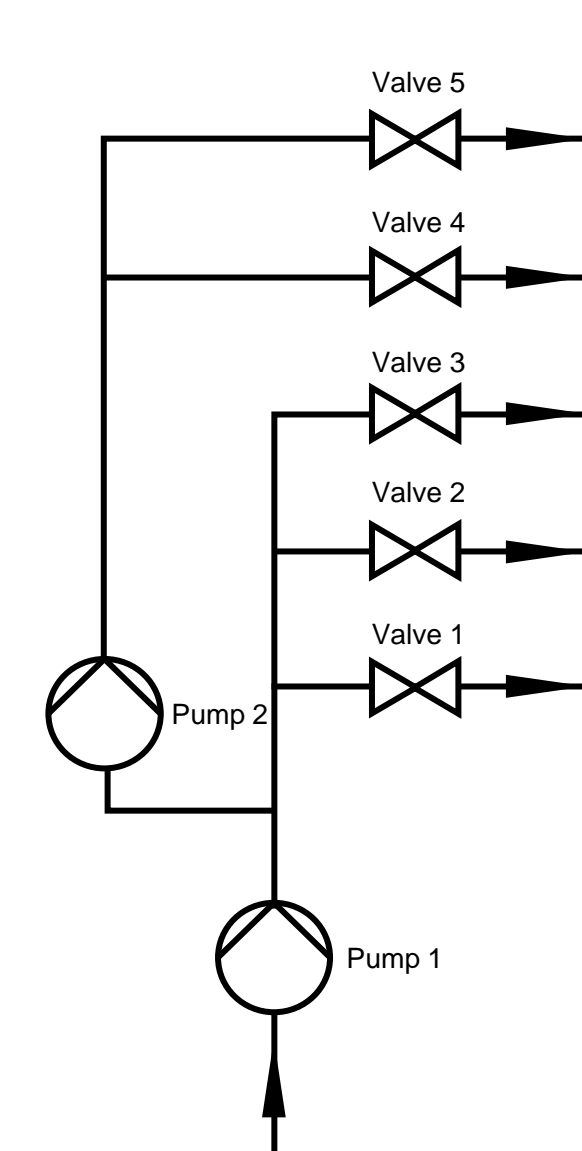
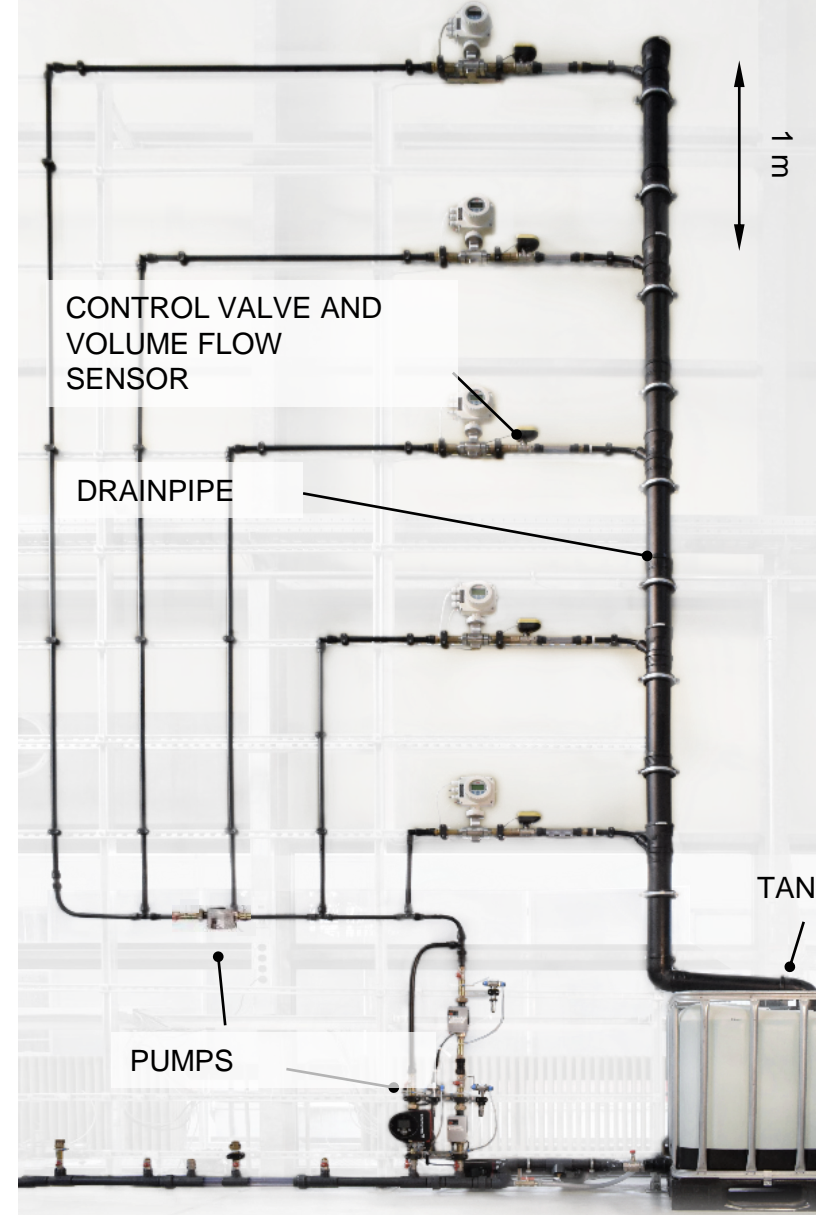
"Hollnagel outlines [...] four resilience functions [...]. All four functions (learning, responding, anticipating, monitoring) are seen as cross-sectional claims. This means that people are also required to learn, anticipate, monitor and respond in order to create a resilient socio-technical system."

Pelz et al., 2021. "Mastering Uncertainty in Mechanical Engineering." Springer, Cham



USE-CASE

- model of a high-rise building scaled to 5 m with central and decentral pumping station and five consumers on five floors
- representative load profile of 30 minutes duration with phases of high, medium, low and zero demand
- simplified load profile of 30 seconds duration for tracing and comprehending decisions made by the multi-agent system



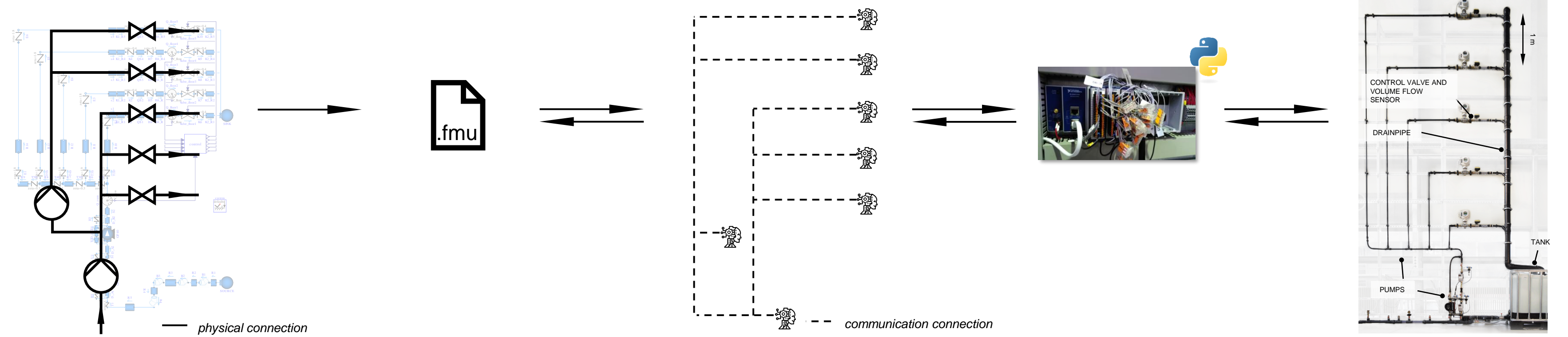
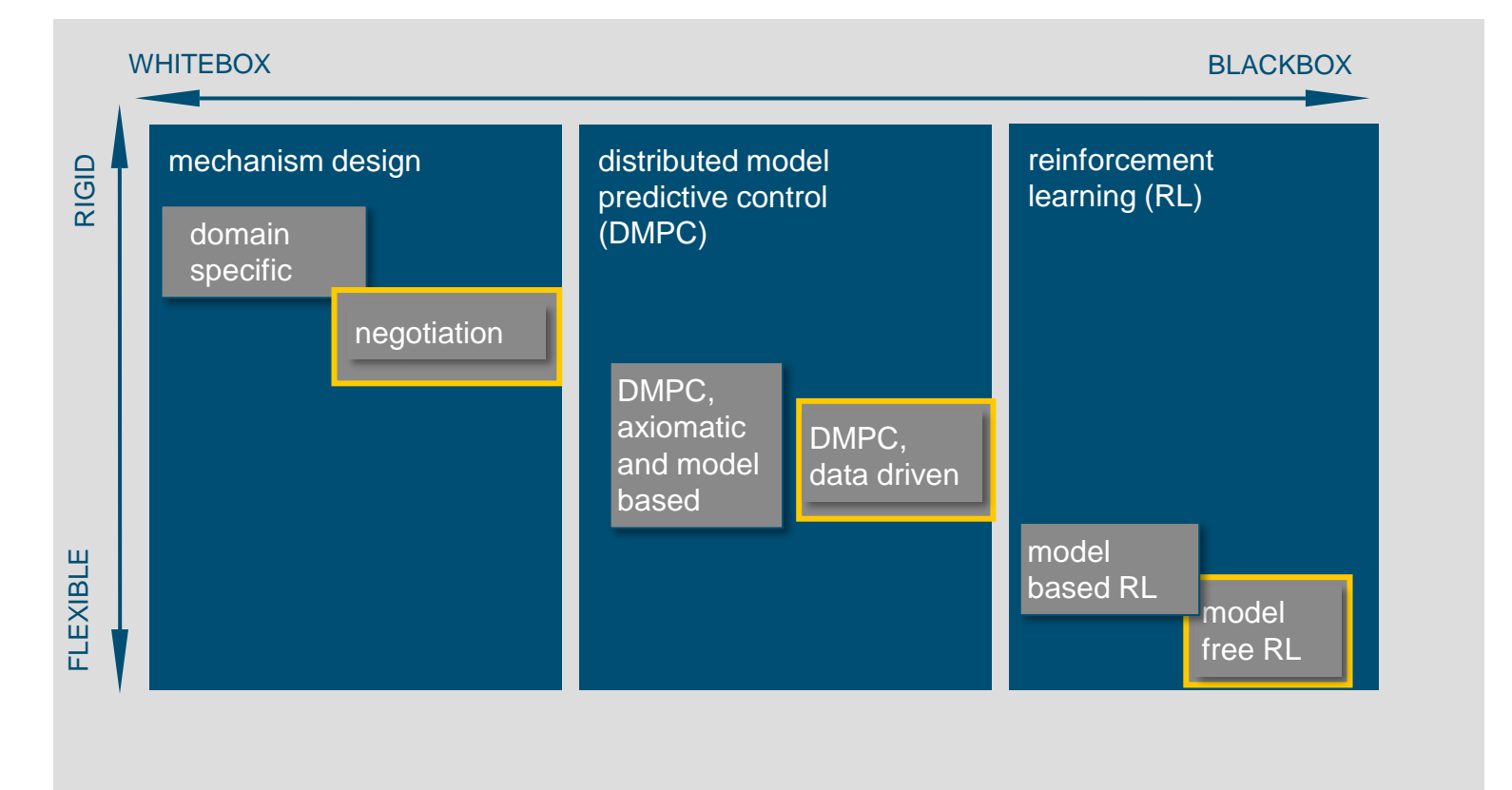
METHODS

combining advantages of centralised system-wide control and local component control in distributed multi-agent system (MAS)

comparison of methods for designing the MAS:

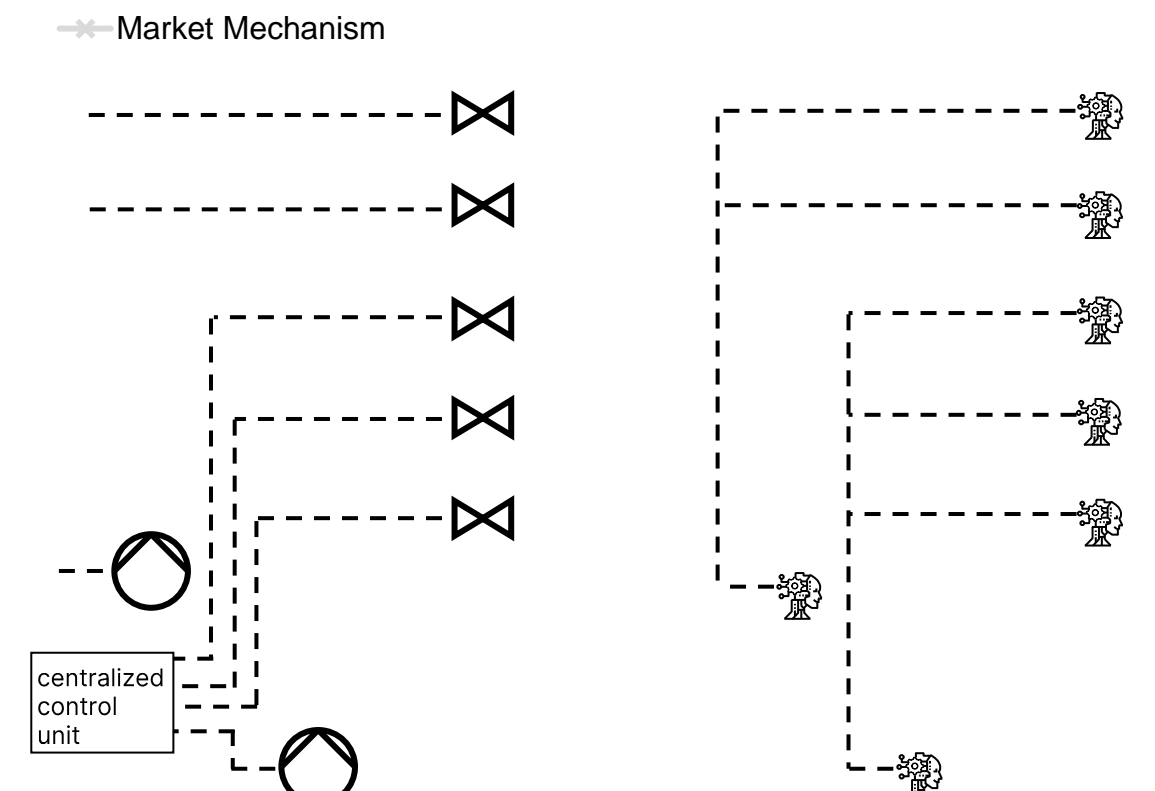
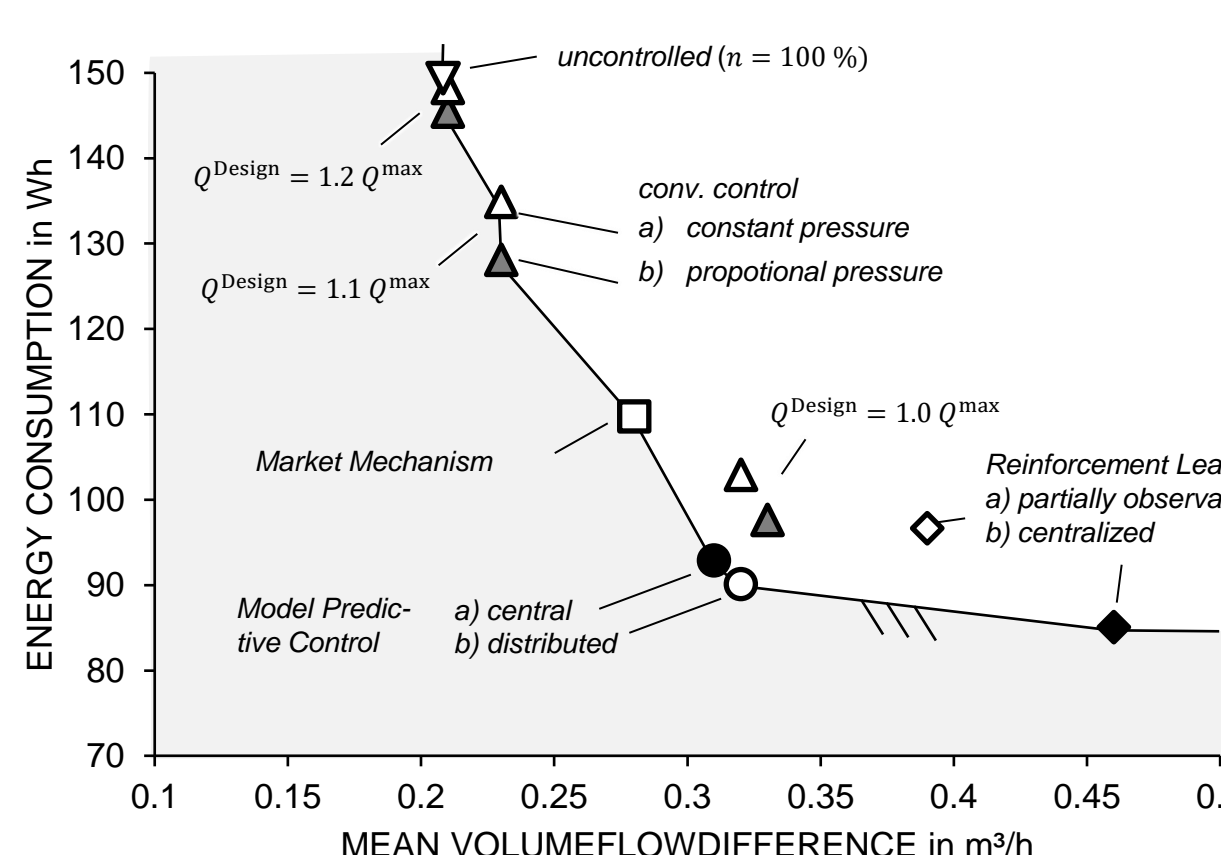
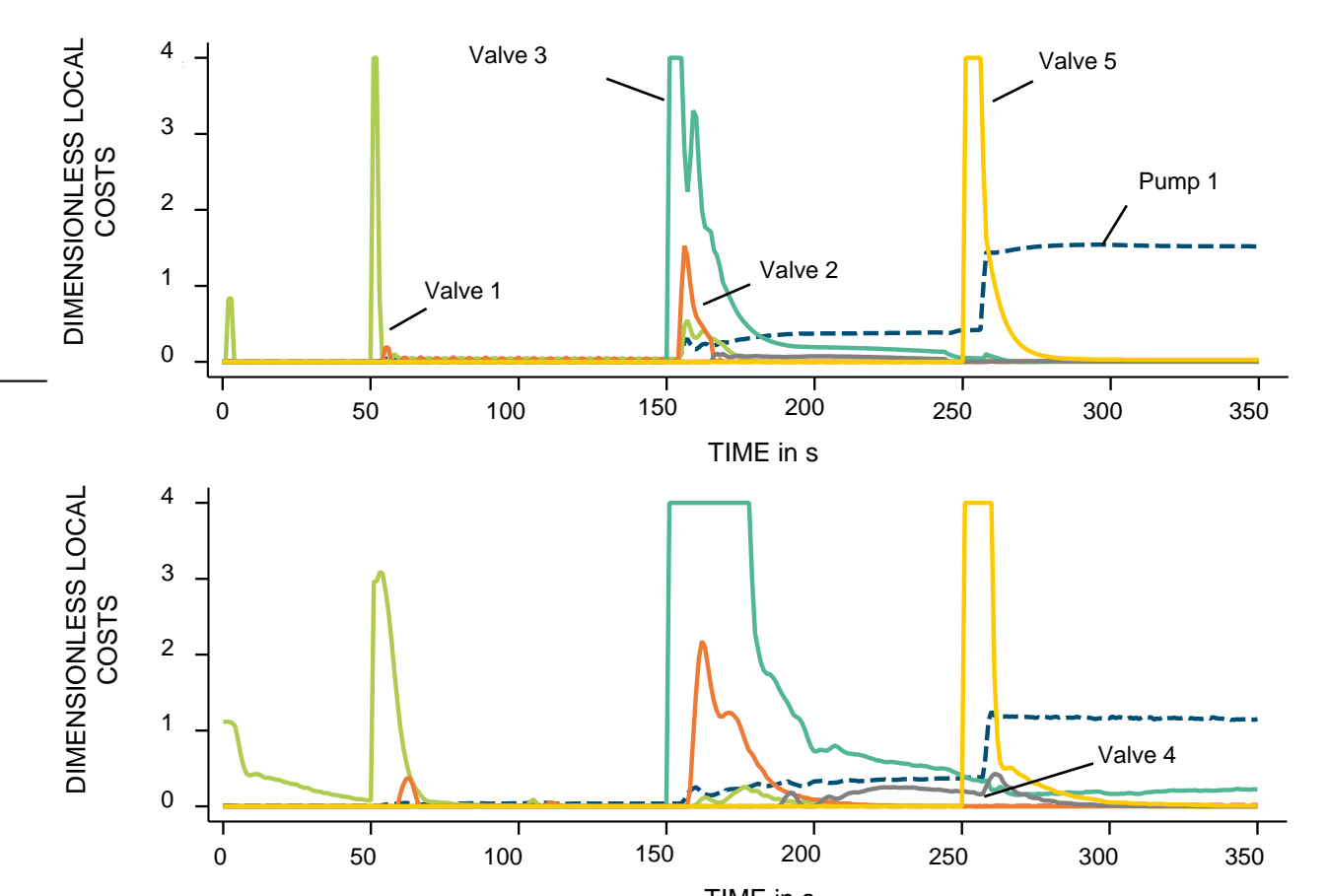
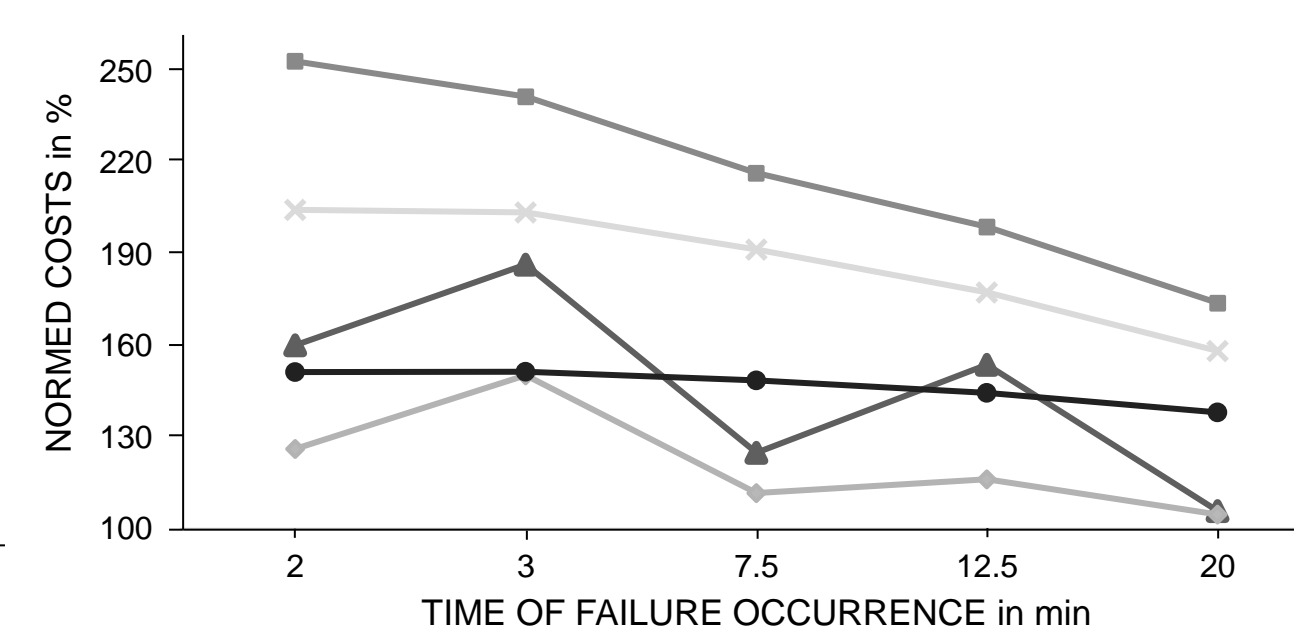
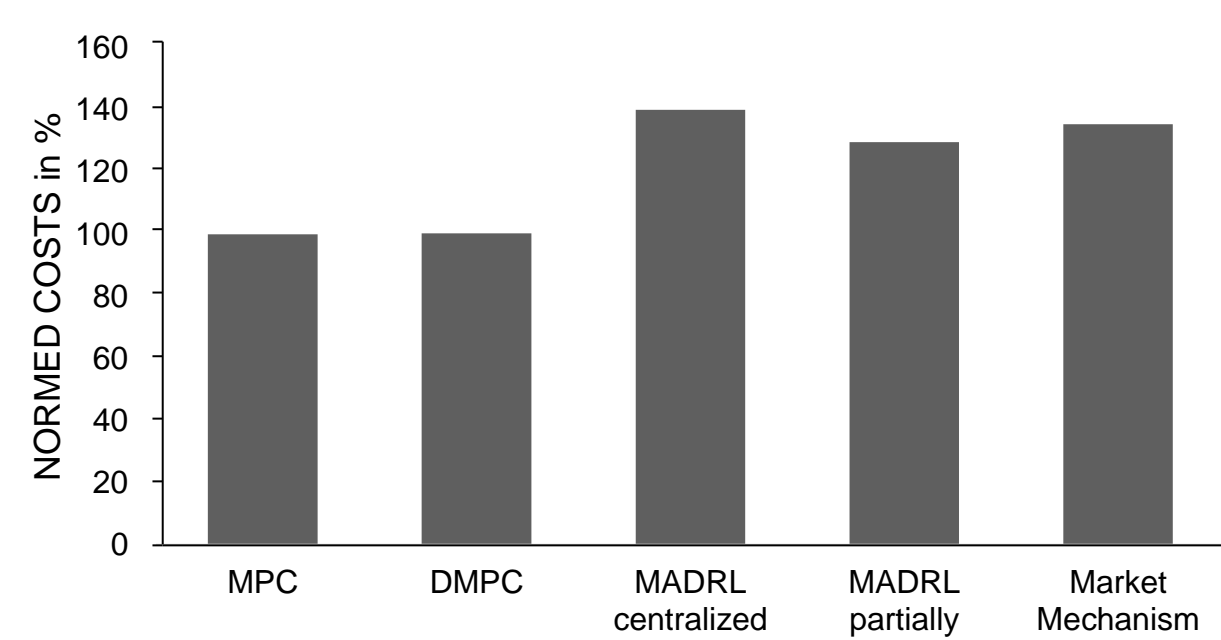
- (i) control technology
 - (distributed) model-predictive control (D)MPC
- (ii) machine learning
 - multi-agent deep reinforcement learning (MADRL)
- (iii) game theory
 - market mechanism for trading guarantees of volume flow

- What is my purpose?
 - What can I monitor?
 - What can I influence?
- Purpose:** fulfill volume flow demand
Monitor: target and actual volume flow
Variable: valve opening
Costs: $c_{valve,t} = \lambda_i (Q_{i,t,demand} - Q_{i,t,actual})^2, \forall t \in T, i \in V$
- Purpose:** minimise power consumption
Monitor: power consumption, volume flow
Variable: on/off, rotational velocity
Costs: $c_{pump,t} = \lambda_i \left(\frac{P_{i,t} - P_{i,t,max}}{P_{i,t,max}} \right)^2, \forall t \in T, i \in P$



RESULTS

- (i) DMPC performance close to optimum of centralised system-wide control
- (ii) costs of MADRL and market mechanism 30-40% higher
- (iii) costs of the different approaches pareto-optimal with regard to conflicting optimisation goals energy efficiency and control accuracy
- (iv) DMPC and one MADRL approach more robust with regard to a disruption within the communication network compared to centralised control



- ▶ The costs of the physical system exceed those of the simulation by $\approx 51\%$.
- ▶ The simulated model does not consider system dynamics.



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