

Interrelation between agent technology and distributed model predictive control

Control in Steel,

Distributed Autonomous Control Solutions

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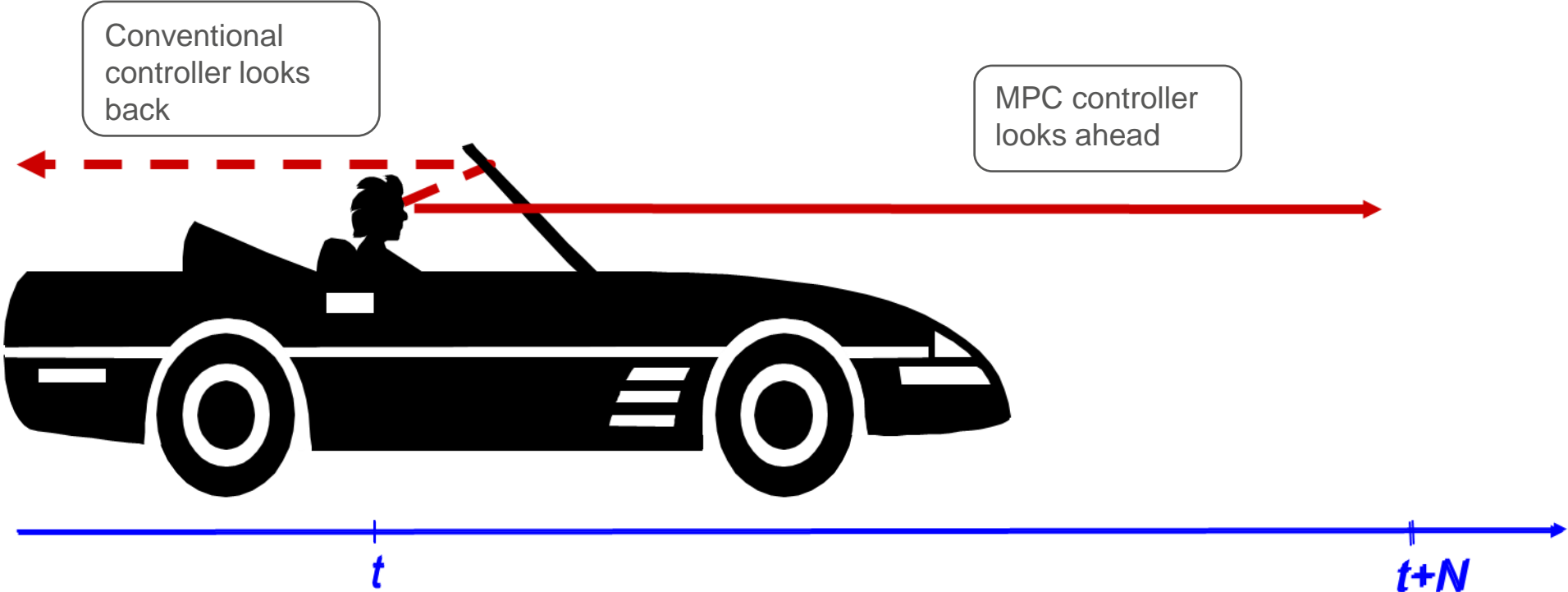
What is an agent?

- › *An agent can be defined as an entity that behaves **autonomously** within an arbitrary system in the **pursuit** of some **goals**¹*
- › *A multi-agent system (MAS) is a system where multiple agents interact with each other directly, via communication acts , or indirectly, by acting on the shared environment.²*
- › In addition, agents may decide to cooperate, to achieve a common goal, or to compete, to serve their own interests at the expense of other agents.²
- › In particular, agents may form cooperative teams, which can in turn compete against other teams of agents.²

[1] M. Wooldridge: An introduction to multiagent systems. (2009)

[2] F. Fioretto, E. Pontelli, W. Yeoh: Distributed Constraint Optimization Problems and Applications A Survey (2018)

Background of MPC control



Background of MPC control

- › minimise:
 - › An objective function in terms of action over the prediction horizon
- › subject to:
 - › The dynamics of the system over the prediction horizon
 - › The constraints on the system
 - › The initial state of the system at the beginning of the current control cycle

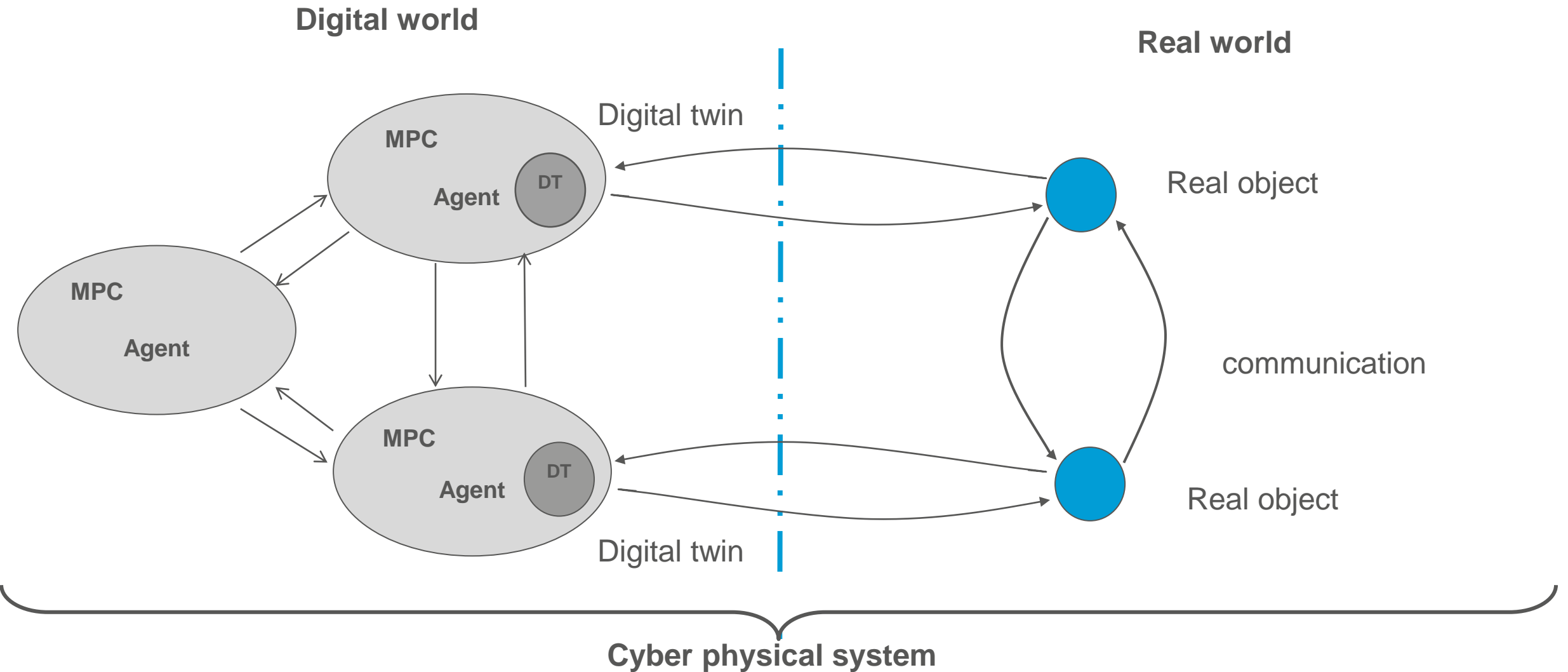
$$\min_{x,u,y} J(x, u, y)$$

$$\begin{aligned} x_{k+1} &= Ax_k + Bu_k \\ y_k &= Cx_k + Du_k \end{aligned}$$

$$\begin{aligned} x_L &\leq x_k \leq x_U \\ u_L &\leq u_k \leq u_U \\ y_L &\leq y_k \leq y_U \end{aligned}$$

$$x_{k=0} = x_0$$

Relation between digital agents and digital twins



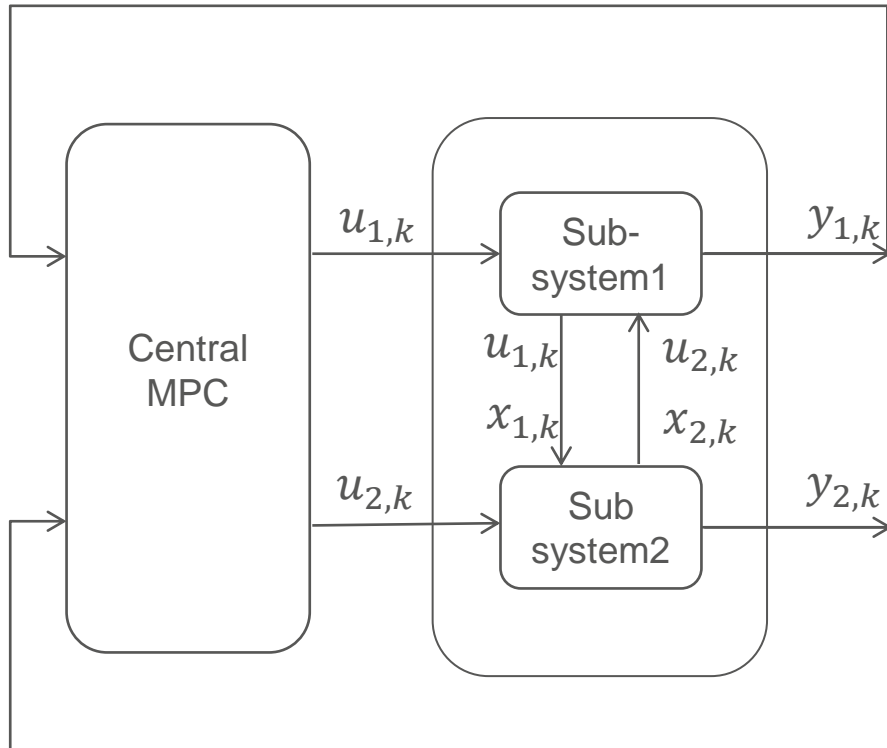
Different set up of the MPC system

- › **Centralized:** Subsystem communicates with a centralized MPC that performs computation and sends new commands
- › **Decentralized:** purely local MPC, i.e. no communication between local MPC controller (agents)
- › **Hierarchical:** MPC where computations are done by MPCs that communicate with other MPCs at a higher level in a hierarchical structure, eventually leading to a centralized controller
- › **Distributed:** local MPC controls local subprocess where each MPC communicates with its neighbours, but there is not a central controller

Daniel K. Molzahn, Florian Dörfler, Henrik Sandberg, Steven H. Low, Sambuddha Chakrabarti, Ross Baldick, Javad Lavaei
A Survey of Distributed Optimization and Control Algorithms for Electric Power Systems,
IEEE TRANSACTIONS ON SMART GRID, VOL 8, NO. 11, NOVEMBER 2017

Central MPC architecture

Centralized: Subsystem communicates with a centralized MPC that performs computation and sends new commands



> Model: Interacting subsystems

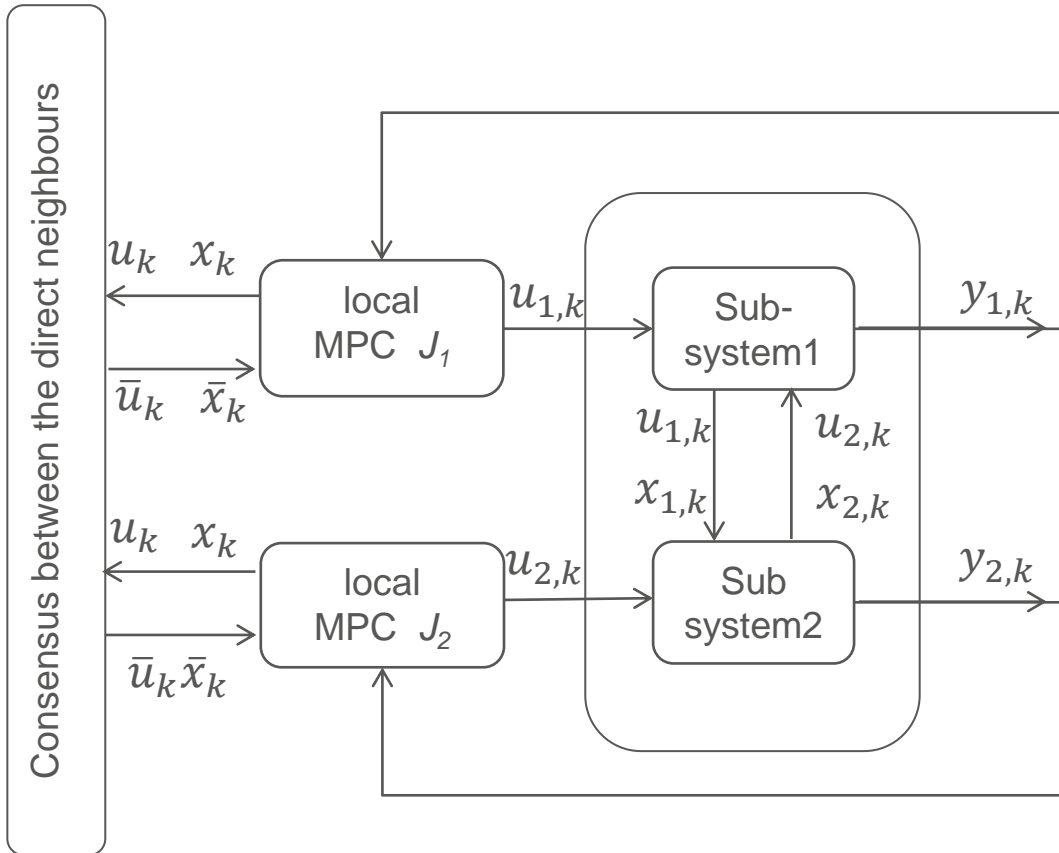
$$G_1 := \begin{cases} x_{1,k+1} = A_{11}x_{1,k} + B_{11}u_{1,k} + A_{12}x_{2,k} + B_{12}u_{2,k} \\ y_{1,k} = C_1x_{1,k} \end{cases}$$

$$G_2 := \begin{cases} x_{2,k+1} = A_{22}x_{2,k} + B_{22}u_{2,k} + A_{21}x_{1,k} + B_{21}u_{1,k} \\ y_{2,k} = C_2x_{2,k} \end{cases}$$

> Cost function

$$J = \sum_{k=t}^{t+N-1} \frac{1}{2} \left(\|x_k\|_Q^2 + \|u_k\|_R^2 \right) + \frac{1}{2} [\|x_{t+N}\|_R^2]$$

Distributed: cooperative consensus based DMPC



- › Model: Interacting subsystems

$$G_1 := \begin{cases} x_{1,k+1} = A_{11}x_{1,k} + B_{11}u_{1,k} + A_{12}x_{2,k} + B_{12}u_{2,k} \\ y_{1,k} = C_1x_{1,k} \end{cases}$$

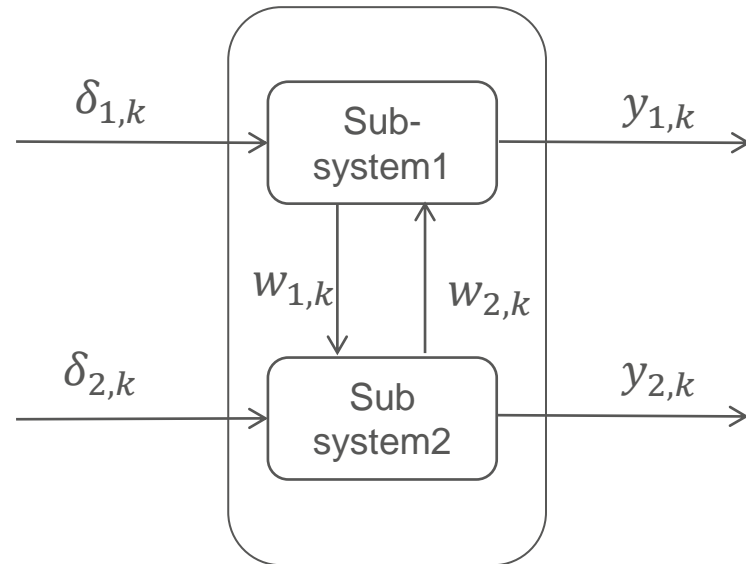
$$G_2 := \begin{cases} x_{2,k+1} = A_{22}x_{2,k} + B_{22}u_{2,k} + A_{21}x_{1,k} + B_{21}u_{1,k} \\ y_{2,k} = C_2x_{2,k} \end{cases}$$

- › Assuming the cost function is separable $J = \rho_1 J_1 + \rho_2 J_2$
- › Extend the cost function

$$J_1 = \sum_{k=t}^{t+N-1} \frac{1}{2} (\|x_k\|_Q^2 + \|u_k\|_R^2) + \frac{1}{2} (\|x_k\|_R^2) +$$

$$+ \sum_{k=t}^{t+N-1} \gamma_k^T \left(\begin{bmatrix} x_k \\ u_k \end{bmatrix} - \begin{bmatrix} \bar{x}_k \\ \bar{u}_k \end{bmatrix} \right) + \frac{\rho}{2} \left(\left\| \begin{bmatrix} x_k \\ u_k \end{bmatrix} - \begin{bmatrix} \bar{x}_k \\ \bar{u}_k \end{bmatrix} \right\|^2 \right)$$

Distributed MPC architecture



› Model: Interacting subsystems (optimal flow problem)

$$G_1 := \begin{cases} x_{1,k+1} = A_{11}x_{1,k} + B_{11}\delta_{1,k} + B_{12}w_{2,k} + E_1v_{1,k} \\ y_{1,k} = C_1x_{1,k} \end{cases}$$

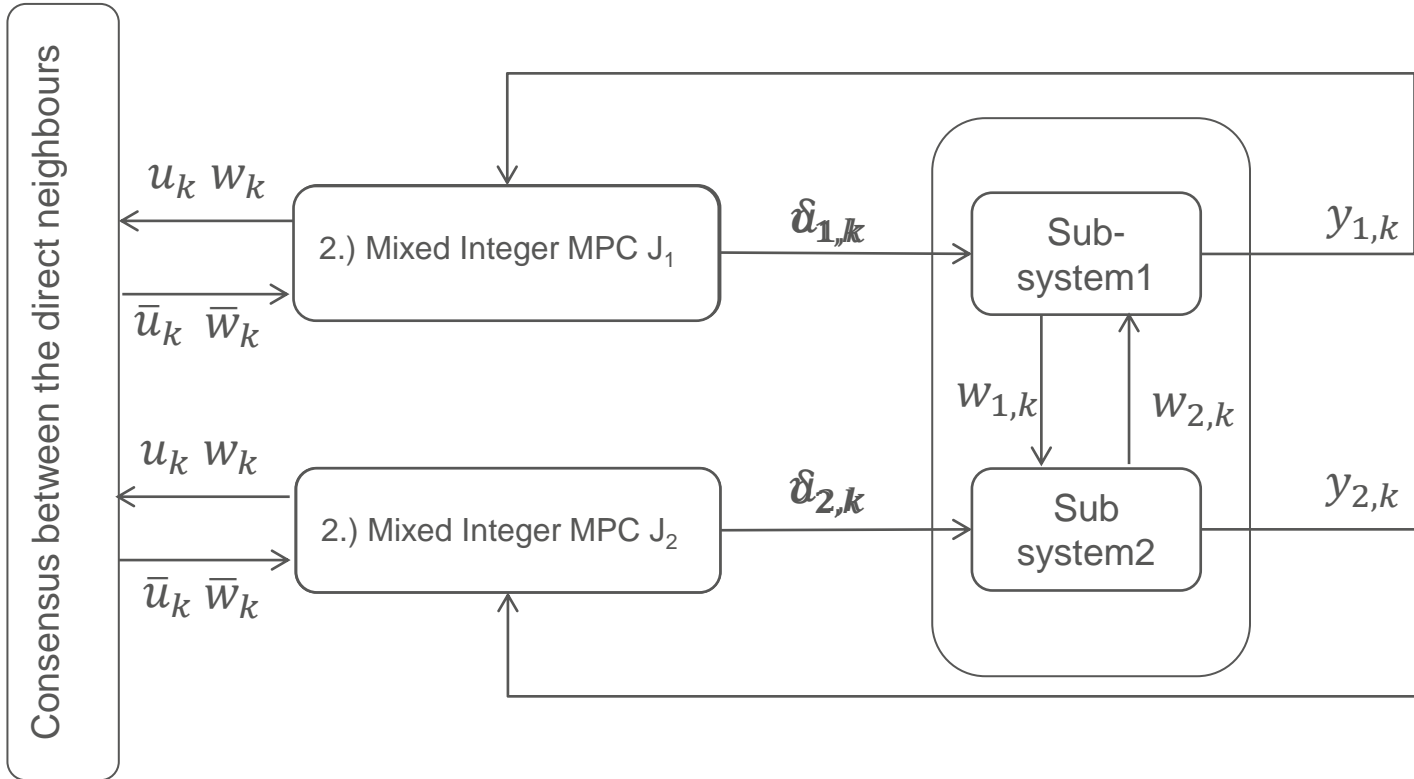
$$G_2 := \begin{cases} x_{2,k+1} = A_{22}x_{2,k} + B_{22}\delta_{2,k} + B_{21}w_{1,k} + E_2v_{2,k} \\ y_{2,k} = C_2x_{2,k} \end{cases}$$

Coupling boundary conditions

$$F_2w_{2,k} - F_1w_{1,k} = 0$$

Distributed MPC architecture

w_k



> Main loop:

1. Solve the local relaxed MPC's J_1 and J_2 in parallel where $\delta_{1,k}$ is continuous
2. Send u_k and w_k to the central unit and find a consensus
3. Send \bar{u}_k and \bar{w}_k to the local MPC's
4. Repeat until converged

> Mixed integer update:

Solve the local mixed integer MPC J_1 and J_2 by using the solution of loop 1-4 as a starting point and where $\delta_{1,k}$ is now discrete

Applications

- › Interconnected micro grids
- › Drinking water networks
- › Steam networks
- › Natural gas networks
- › Off gas networks in steelworks

All these applications belong to the class of optimal flow problems.

Conclusion

Distributed algorithms have advantages over centralized approaches:

- › Agents share a limited amount of information => increase of cybersecurity
- › Flexible => can be extended
- › Robustness with respect to failure
- › Parallel computation => faster
- › suitable for very big problems
- › potential to respect privacy of data

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Thank you for your attention!

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