Abstract: This paper considers networked discrete-event systems. Local state-feedback controllers enable each subsystem to reach local target states. Due to the physical restrictions between the subsystems, the local target states might not be reachable autonomously, but cooperatively with the help of other subsystems. Therefore, each subsystem is extended by a network unit, which detects and resolves possible physical restrictions. If cooperation is necessary, the network units temporarily modify their local target states while applying situation-dependent communication. The proposed method is applied to the flexible manufacturing system HANS and its applicability to real-life systems is demonstrated by experimental results.

Keywords: Networked discrete-event systems, cooperative tracking control

1. INTRODUCTION

1.1 Networked discrete-event systems

Networked discrete-event systems (Fig. 1(a)) consist of controlled subsystems $\Sigma_i$, which are able to use a digital communication network $\Sigma_N$. Each subsystem has a local state-feedback controller to reach a target state $z_{F_i}$. The physical couplings $\Sigma_K$ are represented as synchronous state transitions of neighbouring subsystems. If they become active, two subsystems have to execute these synchronous state transitions together before the subsystems are able to reach their local target states. The task to follow the synchronous state transitions necessitates a cooperation among the controllers.

![Fig. 1. Networked system](image1)

![Fig. 2. Flexible manufacturing system HANS](image2)

This paper experimentally evaluates a cooperative control method for networked discrete-event systems. If a subsystem has to execute a cooperative task to satisfy its local task, then the subsystems apply the communication network to organise the execution of this cooperation. That means, each subsystem is steered into target states $z_{F_i}$ to solve its local tasks. If a subsystem has to execute a synchronous state transition to reach $z_{F_i}$, then the subsystems temporarily modify the local target states $z_{F_i}$ to ensure the execution of this synchronous state transition.

To this aim, the network units $\Sigma'_i$ in Fig. 1(a) are introduced to enable each controlled subsystem $\Sigma_i$ to exchange situation-dependent information among each other via the digital communication network $\Sigma_N$. Based on a reduced overall model, which mainly describes the physical restrictions among the subsystems and abstracts from the autonomous behaviour, the network units cooperatively determine the modified target states $\hat{z}_{F_i}$ for their subsystems.

As an example, the flexible manufacturing system HANS in Fig. 2 is considered (Zgorzelski and Lunze, 2018a) (Modular production system MPS® of the company FESTO), which is a pick-and-place system for handling workpieces. It consists of four subsystems: The area-gantry robot, the line-gantry robot, the 3-axis robot and the conveyor belt. Each subsystem can transport the workpieces autonomously in the direction indicated by the solid red arrows (autonomous mode). However, in the networked system, the subsystem can cooperatively transfer the workpieces autonomously in the direction indicated by the solid red arrows (autonomous mode). However, in the networked system, the subsystem can cooperatively transfer the workpieces between the subsystems, which is shown by the red double arrows (Fig. 2). It is clear to see that for the transfer the subsystem have to cooperate temporarily (cooperative mode), which is solved by the proposed method.

The first contribution of this paper is an extension of the cooperative control method of Zgorzelski and Lunze (2019) from two subsystems to an arbitrary number of coupled subsystems. In order to do that, the method from Zgorzelski

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and Lunze (2019) is generalised through the application of extended Input/Output (I/O)-automata (Drüppel et al., 2008), which introduces a new form of model describing the system structure. The second contribution is the experimental application of the proposed method to the cooperative pick-and-place process HANS with four subsystems (Fig. 2), which will demonstrate the applicability of the presented method to real-life systems.

1.2 Literature

The supervisory control of discrete-event system proposed by Ramadge and Wonham (1989) has been investigated to connect decentralised supervisors via a digital communication network (Barrett and Lafortune, 2000), (Rudie et al., 2003) to restrict the plant behaviour to a given specification. Moreover, in the multi-agent supervisory control (Su and Lemmatron, 2017) and distributed supervisory control (Seow et al., 2009), (Cai and Wonham, 2010) the communication network is used for the synchronous event passing between the local supervisors. Other works on networked supervisory control (Wang et al., 2016) consider a communication network between the plant and the supervisor, which causes a delayed transmission of events and control actions.

In contrast, our control objective is the state-feedback control of I/O automata (Żgorzelski and Lunze, 2017) to steer physically coupled subsystems into local target states. Although the state attraction problem (Brave and Heymann, 1989) of the supervisory control is related to this control problem, Brave and Heymann (1989) did not discuss how to cope with more than one subsystem and a control input that flexibly allows the subsystems to adapt to its environment for reaching different target states given at runtime. Moreover, our communication depends on the plant behaviour but also on the control input, which requires more flexibility.

The cooperative tracking control proposed in this paper has similarities to the distributed optimal controller proposed by Stursberg and Hillmann (2017), which has the control objective to minimize the state trajectory cost of subsystems into their goal states within a distributed architecture with a priority structure. The difference is that we do not want to minimize the costs and we do not restrict our system to a certain priority structure. Moreover, the physical restrictions in this paper have a more generalised description and the solution presented in this paper is given for arbitrary couplings and any number of subsystems. Hence, in the literature, the networked discrete-event systems appear with a different structure or a different control objective.

1.3 Organisation of the paper and notation

In Section 2, the physically coupled controlled subsystems are presented and the cooperative control problem is formulated. The main result is presented in Section 3 that gives the solution of the cooperative control problem for networked discrete-event systems. Finally, Section 4 presents the modelling and the cooperative control of the flexible manufacturing system HANS.

In the following, sets are denoted by calligraphic letters (e.g. \( \mathcal{Z} \)) and scalars by lower case italics (e.g. \( z_0 \)). To denote sequences of scalars, upper case italics are used (e.g. \( Z(0 \ldots k_s) = (z(0), z(1), \ldots z(k_s)) \)) with \( k_s \) being the time horizon. State transitions are denoted by \( (z_i \rightarrow z'_i) \) and the empty symbol is denoted by \( \varepsilon \). \( Z^* \) denotes the kleene closure of the set \( Z \) and \( \prod_i I \) denotes the cartesian product.

2. PHYSICALLY COUPLED CONTROLLED SUBSYSTEMS

2.1 Physically coupled plants

Each controlled subsystem \( \Sigma_i \) consists of a plant \( \Sigma_{P_i} \) and a state-feedback controller \( \Sigma_{C_i} \) (Fig. 1(b)). The plant is modelled by a deterministic extended Input/Output (I/O)-automaton (Lunze, 2017), \( \Sigma_{P_i} \) is the state transition function \( G_i : Z_i \times V_i \times S_i \rightarrow Z_i \), \( z_i(0) = z_{i0} \).

\[
\Sigma_{P_i} : \begin{cases} z_i(k+1) = G_i(z_i(k), v_i(k), s_i(k)) \quad z_i(0) = z_{i0} \\
r_i(k) = F_i(z_i(k), v_i(k), s_i(k)) \end{cases}
\]

(1)

with the state set \( Z_i \), the input set \( V_i \), the coupling input set \( S_i \) and the coupling output set \( R_i \). The state transition function \( G_i : Z_i \times V_i \times S_i \rightarrow Z_i \) maps the current state \( z_i(k) \), the current input \( v_i(k) \) and the coupling input \( s_i(k) \) at time step \( k \) to the next state \( z_i(k+1) \), whereas the coupling function \( F_i : Z_i \times V_i \times S_i \rightarrow R_i \) maps the same arguments to the current value of the coupling output \( r_i(k) \).

In the following, the current state \( z_i(k) \) is assumed to be measurable and the next state \( z_i(k+1) \) is denoted by \( z'_i \).

Physically coupled state transitions are represented as *synchronous state transitions* having a coupling input \( s_i \neq \varepsilon \), whereas physically uncoupled state transitions are denoted as *autonomous state transitions* having the empty symbol \( \varepsilon \) as the coupling input \( s_i = \varepsilon \). The description of the synchronous state transitions is similar to the notion of \( \gamma \) introduced by Żgorzelski and Lunze (2019, 2018b), however, in this paper, the generalised description with extended I/O-automata is applied.

The couplings between the plants (1) are described by the coupling function \( K : \prod_{i=1}^t R_i \rightarrow \prod_{i=1}^t S_i \)

\[
\Sigma_K : (s_1(k), \ldots, s_t(k))^T = K((r_1(k), \ldots, r_t(k))^T), \quad (2)
\]

which maps all coupling outputs \( r_i(k) \) to the coupling inputs \( s_i(k) \), whereas the component \( K_i : \prod_{i=1}^t R_i \rightarrow S_i \) of (2) \( \Sigma_K : s_i(k) = K_i((r_1(k), \ldots, r_t(k))^T) \) does the same for the \( i \)-th subsystem.

2.2 Local state-feedback controller

The plants \( \Sigma_{P_i} \) are connected to a local state-feedback controller

\[
\Sigma_{C_i} : v_i(k) = C_i(z_i(k), e_i(k)) \quad (3)
\]

with the controller function \( C_i : Z_i \times V_i \rightarrow V_i \) that maps the current state \( z_i(k) \) and the control input \( e_i(k) \) to an input symbol \( v_i(k) \) for the plant (1). The controller (3) has the control aim to steer the plant (1) into the desired target state \( z_{f_{P_i}} \in Z_i \) given to the control input \( e_i(k) = z_{f_{P_i}} \).

The controller function \( C_i \) is designed by the Algorithm 1 with the active input set \( V_i(z_i, z'_i) = \{ v_i \mid \exists s_i \in S_i : z'_i = G_i(z_i, v_i, s_i) \} \).

**Algorithm 1** Local state-feedback controller design

**Input:** Plant \( \Sigma_{P_i} \)

**For** \( \forall z_i, z_{f_{P_i}} \in Z_i \) **do**

(1) Search shortest state sequence: \( Z_i(0 \ldots k_s) = (z_i(0) = z_i, z_i(1) = z'_i, \ldots, z_i(k_s) = z_{f_{P_i}}) \)

(2) Define: \( C_i(z_i, z_{f_{P_i}}) \in V_i(z_i, z'_i) \)

**Result:** Controller function \( C_i \)
The controlled subsystem $\Sigma_i = (Z_i, V_i, S_i, R_i, G_i, F_i, z_{i0})$

$$\Sigma_i : \begin{cases} z_i(k+1) = G_i(z_i(k), e_i(k), s_i(k)), z_i(0) = z_{i0} \\ r_i(k) = F_i(z_i(k), e_i(k), s_i(k)) \end{cases}$$

(4)

with $G_i(z_i, e_i, s_i) = G_i(z_i, G_i(z_i, e_i, s_i))$ and $F_i(z_i, e_i, s_i) = F_i(z_i, G_i(z_i, e_i, s_i))$ obtained by the connection of the plant $(1)$ with the state-feedback controller $(3)$.

The set of all local state transitions is given by

$$G_i = \{ (z_i \rightarrow e_i) \mid \exists s_i \in S_i : e_i = G_i(z_i, e_i, s_i) \}$$

and the set of all coupling output signals $r_i$ of a local state transition $(z_i \rightarrow z'_i)$ is described by the active coupling output set

$$\mathcal{R}_i(z_i, z'_i) = \{ r_i \mid \exists s_i \in S_i : r_i = F_i(z_i, z'_i, s_i) \land z'_i = G_i(z_i, e_i, s_i) \}.$$

The overall state transition function $G$ and coupling function $F$ of all controlled subsystems are denoted by

$$\bar{G}(z, e, s) = \begin{cases} G_i(z_i, e_i, s_i) \\ \bar{G}_l(z_l, e_l, s_l) \end{cases}, \mathcal{F}(z, e, s) = \begin{cases} F_i(z_i, e_i, s_i) \\ \bar{F}_l(z_l, e_l, s_l) \end{cases}$$

with $z = (z_1, \ldots, z_m)^T$, $e = (e_1, \ldots, e_l)^T$, $s = (s_1, \ldots, s_l)^T$ and the overall state set is given by $Z = \prod_{i=1}^{m} Z_i$.

The active coupling input set is defined by

$$\mathcal{S}_i(z_i, z'_i) = \{ s_i \mid z'_i = G_i(z_i, e_i, s_i) \}$$

(5)

and the following behaviour is assumed that the plant remains in the current state for the wrong coupling input:

**Assumption 1.** $G_i(z_i, e_i, s_i) = z_i$ if $s_i \notin \mathcal{S}_i(z_i, e_i)$

In the proof of the following Lemma 2, it is shown that the local control objective is met by the controlled subsystem $(4)$, if the coupling input $s_i(k)$ is an element of the active coupling input set $(5)$.

**Lemma 2.** (Local target state control). In the controlled subsystem $(4)$ with the control input $u_i(k)$, $k \geq 0$, the relation

$$\forall z_i, z'_i \in Z_i, \exists k_0 : z_i(k_0) = z'_i$$

holds if $\exists r_i(k) \geq k_0 : s_i(k) \in \mathcal{S}_i(z_i(k), z'_i(k + 1))$.

**Proof.** The proof is given by induction. For all state pairs $z_i, z'_i \in Z_i$ and its connecting shortest state sequences $Z_i(0 \ldots k_0)$, it holds $z_i(k_0 + 1) = G_i(z_i(k_0), e_i(k_0), s_i(k_0))$ in the controlled subsystem $\Sigma_i$ with $e_i(k) = z_{Fi}$ if $s_i(k) \in \mathcal{S}_i(z_i(k), z_i(k + 1))$ holds. Base case $k = 0$: $z_i(1) = G_i(z_i(0), z_{Fi}, s_i(0))$ if $s_i(0) \in \mathcal{S}_i(z_i(0), z_i(1))$. Induction case $k \rightarrow k + 1$: From the optimality principle (Bellman, 1957) it follows that each shortest path includes subpaths that are the shortest subpaths. Hence, it holds that $z_i(k + 2) = G_i(z_i(k + 1), z_{Fi}, s_i(k + 1))$ if $s_i(k + 1) \in \mathcal{S}_i(z_i(k + 1), z_i(k + 2))$, which concludes the proof. \(\square\)

2.3 Cooperative control problem

As shown in Lemma 2, the local control aim is satisfied only if the physical coupling input $s_i(k)$ of each local subsystem $(4)$ is an element of the active coupling input set $(5)$, $s_i(k)$ depends on the synchronous state transitions with other subsystems as explained below.

The coupling input $s_i(k)$ of each subsystem depends on the coupling output $r_i(k)$ of their neighbouring subsystems according to the physical coupling network (2). Furthermore, the current value $r_i(k)$ of each subsystem is determined by the coupling output function $F_i$, which, in turn, depends on the execution of synchronous state transitions with other subsystems. These synchronous state transitions have to be executed by the subsystems to solve physical restrictions before reaching their local target states $z_{Fi}$.

Hence, the controlled subsystems cooperatively have to change their local state trajectories $Z_i(0 \ldots k_0)$ from the current state into their target state to include all necessary synchronous state transitions to reach the required state $z_{Fi}$, respectively.

**Problem 3.** (Cooperative control problem). In the networked discrete-event system, each controlled subsystem $(4)$ has to reach its individual local target state:

$$\forall z_i(0), z_{Fi} \in Z_i, \exists \mathcal{Z}_i^* : z_i(k_0) = z_{Fi}.$$ 

To solve the cooperative control problem, this paper applies the network unit $\Sigma_i$ introduced by Zgorzelski and Lunze (2019) and extends $\Sigma_i$ for its application to an arbitrary number of subsystems. The network units $\Sigma_i$ with the control input $u_i(k)$ are connected to each controlled subsystem (4) and they are able to exchange information via the communication network $\mathcal{N}_i$ (Fig. 1(a)). Depending on the current states $z_i(k)$ and the target states $z_{Fi}$, being set to the control inputs $u_i(k) = z_{Fi}$, the network units $\Sigma_i$ temporarily modify the local target states $e_i(k) = z_{Fi}$ for their controlled subsystems $\Sigma_i$ to cooperatively change the local state trajectories $Z_i(0 \ldots k_0)$ such that all necessary synchronous state transitions are executed to satisfy the Problem 3.

3. COOPERATIVE STATE-FEEDBACK CONTROL

3.1 Model abstraction

The challenge to solve Problem 3 is to ensure the execution of the synchronous state transitions by modifying the local target states. The main idea is to describe the cooperative behaviour of the controlled subsystems by a reduced overall model $\Sigma$. The modified target states $z_{Fi}$ that lead to the new state trajectories solving Problem 3 are selected from a state sequence $\mathcal{Z}(0 \ldots k_0)$ in the abstracted model $\Sigma$.

To obtain the simplified overall model, each subsystem is reduced to an abstracted model based on an equivalence relation and all abstracted subsystem models are combined to the abstracted overall model afterwards.

The equivalence relation

$$z_i \sim z_i' \text{ if } P_i(z) = P_i(z_i')$$

(7)

partition the local state sets $Z_i$ into disjoint subsets $Z_i \subseteq Z_i(m = 1, 2, \ldots, m)$. The states of the abstracted local models are defined to be the abstracted state set $\mathcal{Z} = \{ z_i \} \cup \{ z_{Ei} \}$ forming the equivalence classes (8).

The local abstracted model is defined as an extended I/O-automaton $\hat{\Sigma}_i = (\hat{Z}_i, \hat{V}_i, \hat{S}_i, \hat{R}_i, \hat{G}_i, \hat{F}_i, z_{i0})$.
\[ \begin{aligned}
\Sigma_i : & \begin{cases}
\tilde{z}_i(k+1) = \tilde{G}_i(\tilde{z}_i(k), \tilde{e}_i(k), \tilde{s}_i(k)), \quad \tilde{z}_i(0) = \tilde{z}_{i0} \\
\tilde{r}_i(k) = \tilde{F}_i(\tilde{z}_i(k), \tilde{e}_i(k), \tilde{s}_i(k))
\end{cases} \\
\text{with } \tilde{z}_i = \tilde{V}_i = [\tilde{z}_i], \tilde{S}_i = \tilde{S}_i \text{ and } \tilde{R}_i = \tilde{R}_i. \text{ The state transition function and the coupling function are defined by } \\
\tilde{G}_i(\tilde{z}_i, \tilde{e}_i, \tilde{s}_i) = \tilde{G}_i(z_i, e_i, s_i) \text{ and } \\
\tilde{F}_i(\tilde{z}_i, \tilde{e}_i, \tilde{s}_i) = \tilde{F}_i(z_i, e_i, s_i)
\end{aligned} \tag{9}
\]

with \( r_i \in \tilde{R}_i(z_i, e_i). \) \([\tilde{G}_i(\ldots)]\) denotes the equivalence class of the codomain of the function \( \tilde{G}_i. \) The set of all local abstracted state transitions is given by

\[ \tilde{G}_i = \{ ([z_i] \to [e_i]) \mid \exists s_i \in \tilde{S}_i : e_i = \tilde{G}_i(z_i, e_i, s_i) \} \]

and the state transition function and coupling output function of all abstracted models are denoted by

\[ \tilde{G}(\tilde{z}, \tilde{e}, \tilde{s}) = \left( \tilde{G}_i(\tilde{z}_i, \tilde{e}_i, \tilde{s}_i) \right) \quad \tilde{F}(\tilde{z}, \tilde{e}, \tilde{s}) = \left( \tilde{F}_i(\tilde{z}_i, \tilde{e}_i, \tilde{s}_i) \right). \]

with \( \tilde{z} = (\tilde{z}_1, \ldots, \tilde{z}_I)^T, \tilde{e} = (\tilde{e}_1, \ldots, \tilde{e}_I)^T \) and \( \tilde{s} = (\tilde{s}_1, \ldots, \tilde{s}_I)^T. \)

### 3.2 Properties of the abstracted model

This section investigates the properties of the proposed model abstraction. The question to be answered is under what condition is the abstracted model suitable for the selection of the modified target states to include all necessary synchronous state transitions.

The proof of Lemma 4 will show that the composition of the abstracted subsystem models (9) with the coupling network (2) is equal to the abstraction of the composition of all original controlled subsystems (4), if the local coupling signals of all synchronous state transitions connecting the same abstracted states are equal according to condition (11).

#### Lemma 4. (Compositional abstraction)

\[ \forall ([z] \to [e]) \in \prod_{i=1}^I \tilde{G}_i, \forall (z \to e) \in \prod_{i=1}^I G_i \exists r \in \prod_{i=1}^I R_i : \]

\[ \tilde{G}(\tilde{z}, \tilde{e}, \tilde{s}) = \left( \tilde{G}_i(\tilde{z}_i, \tilde{e}_i, \tilde{s}_i) \right) \quad \tilde{F}(\tilde{z}, \tilde{e}, \tilde{s}) = \left( \tilde{F}_i(\tilde{z}_i, \tilde{e}_i, \tilde{s}_i) \right). \tag{10} \]

if \( \bigwedge_{i=1}^I (R_i, z_i, e_i) = \{ r_i \} \wedge S_i(z_i, e_i) = \{ K_i(r) \}. \tag{11} \]

with \( r = (r_1, \ldots, r_I)^T. \)

**Proof.**

\[ \forall ([z] \to [e]) \in \prod_{i=1}^I \tilde{G}_i, \forall (z \to e) \in \prod_{i=1}^I G_i \exists r \in \prod_{i=1}^I R_i : \]

\[ \bigwedge_{i=1}^I (R_i, z_i, e_i) = \{ r_i \} \wedge S_i(z_i, e_i) = \{ K_i(r) \} \]

\[ \Rightarrow \bigwedge_{i=1}^I \left( \tilde{G}_i(\tilde{z}_i, \tilde{e}_i, K_i(r)) \right) \quad \left( \tilde{F}_i(\tilde{z}_i, \tilde{e}_i, K_i(r)) \right) \]

\[ \Rightarrow \text{eqn. (10)}. \]

In the following, the state transitions of the composed abstracted subsystem models (9) are described by the set

\[ \tilde{G} = \{ (\tilde{z} \to \tilde{e}) \mid r \in R, \tilde{z} = \tilde{G}(\tilde{z}, \tilde{e}, K(r)) \} \quad \text{and } \tilde{r} = \tilde{F}(\tilde{z}, \tilde{e}, K(r)). \tag{12} \]

In the proof of Lemma 5 it is shown, that under the condition (11), all original state transitions \( \prod_{i=1}^I G_i \) that correspond to an abstracted state transition (12) are feasible in the local controlled subsystems (4), which means that the abstracted subsystems (4) can follow each abstracted state transition because it leads to state transitions satisfying condition (6) in Lemma 2.

#### Lemma 5. (Feasibility of abstracted state transitions)

\[ \forall ([z] \to [e]) \in \prod_{i=1}^I \tilde{G}_i, \forall (z \to e) \in \prod_{i=1}^I G_i \exists r \in \prod_{i=1}^I R_i : \]

\[ e = \tilde{G}(z, e, K(r)) \quad \text{and } \tilde{r} = \tilde{F}(z, e, K(r)) \tag{13} \]

if condition (11) holds.

**Proof.**

\[ \forall ([z] \to [e]) \in \prod_{i=1}^I \tilde{G}_i, \forall (z \to e) \in \prod_{i=1}^I G_i \exists r \in \prod_{i=1}^I R_i : \]

\[ \bigwedge_{i=1}^I (R_i, z_i, e_i) = \{ r_i \} \wedge S_i(z_i, e_i) = \{ K_i(r) \} \]

\[ \Rightarrow \bigwedge_{i=1}^I (\tilde{G}_i(\tilde{z}_i, \tilde{e}_i, K_i(r)) \wedge r_i = \tilde{F}_i(\tilde{z}_i, \tilde{e}_i, K_i(r)) \]

\[ \Rightarrow \text{eqn. (13)}. \]

In the next subsection, an abstracted state-feedback controller is designed to determine state sequences in the abstracted model for selecting the temporary modified target states.

### 3.3 Abstracted state-feedback controller

The abstract model \( \Sigma \) is obtained from the composition of the abstracted subsystem models (9) with the coupling network (2). \( \Sigma = (\tilde{Z}, \tilde{V}, \tilde{G}_i, \tilde{z}_0) \) is a deterministic I/O-automaton

\[ \Sigma : \tilde{z}(k+1) = \tilde{G}(\tilde{z}(k), \tilde{e}(k)), \quad \tilde{z}(0) = \tilde{z}_0 \tag{14} \]

with \( \tilde{z} = \tilde{V} = \prod_{i=1}^I \tilde{z}_i \) and the measurable abstracted current state \( \tilde{z}(k). \)

To obtain a state sequence from the current abstracted state \( \tilde{z}(k) = [z(k)] \) into the desired abstracted target state \( \tilde{z}_F = [\tilde{z}_F] \), a state-feedback controller

\[ \Sigma_C : \tilde{c}(k) = \tilde{C}(\tilde{z}(k), \tilde{u}(k)) \tag{15} \]

is designed by applying the method from Zgorzelski and Lunze (2017) with the abstracted controller function \( \tilde{C} : \tilde{Z} \times \tilde{Z} \to \tilde{Z} \). The controller function \( \tilde{C} \) maps the current abstracted state \( \tilde{z}(k) = \tilde{z} \) and the abstracted target state \( \tilde{u}(k) = \tilde{z}_F \) to the successor state \( \tilde{c}(k) = \tilde{z}'(k) \) in the shortest state sequence from \( \tilde{z}(0) \ldots k \) to \( \tilde{z}(k) \) using the abstracted model (14).

### 3.4 Network unit

The state-feedback controller (15) is decomposed into components.
\[ \tilde{\Sigma}_{Ci} : \tilde{e}_i(k) = \tilde{C}_i(\tilde{z}(k), \tilde{u}(k)) \]

with the local controller functions \( \tilde{C}_i : \tilde{Z} \times \tilde{Z} \to \tilde{Z}_i \). Each controller (16) is applied by the network unit \( \Sigma'_i \) belonging to the subsystem as shown in Fig. 1(a).

The network units

\[ \Sigma'_i : \begin{cases} c_i(k) = u_i(k), & \text{if } [z(k)] = [u(k)] \\ e_i(k) \in \tilde{C}_i(\tilde{z}(k), \tilde{u}(k)), & \text{else} \end{cases} \]

are defined to have an autonomous and a cooperative mode. In the autonomous mode \( (c_i(k) = u_i(k)) \), any controlled subsystem has to execute synchronous state transitions \( ([z(k)] \neq [u(k)]) \) to reach its target state and no cooperation is needed between the subsystems. However, in the cooperative mode \( (c_i(k) \in \tilde{C}_i(\tilde{z}(k), \tilde{u}(k))) \), the controlled subsystems have to execute synchronous state transitions \((z(k)) \) and the network unit (17) applies the controller (16) and connects all controlled subsystems \( (4) \) to the cooperating subsystems

\[ \Sigma_i : \begin{cases} z_i(k+1) = G_i(z_i(k), e_i(k), s_i(k)), z_i(0) = z_{i0} \\ r_i(k) = F_i(z_i(k), e_i(k), s_i(k)) \\ e_i(k) \in \tilde{C}_i(\tilde{z}(k), \tilde{u}(k)) \end{cases} \]

by selecting the corresponding local target states \( e_i(k) = z_{Fi} \in e_i(k) \) from its output by and by exchanging their abstracted current states \( \tilde{z}(k) = [z(k)] \) and their target states \( \tilde{u}(k) = [u(k)] \) via the communication network \( \Sigma_N \) (Fig. 1(a)).

Following from Lemma 5, Lemma 6 states that in the cooperative subsystems (18), the abstracted target states \( \tilde{u}_i(k) = z_{Fi} \) are reached after a finite time step \( k_{ci} \).

Lemma 6. (Abstracted state-feedback control). For the digitally connected cooperating subsystem (18) with the control input \( u_i(k) = z_{Fi} \), \( k \geq 0 \), it holds that

\[
\begin{bmatrix}
(z_{i0}^1) \\
(z_{i0}^2) \\
\vdots \\
(z_{i0}^k) \\
\vdots \\
(z_{i0}^{k_{ci}})
\end{bmatrix}
\in \mathbb{Z}, \exists k_{ci} \begin{bmatrix}
[0] \\
[k_{ci}]
\end{bmatrix}
:\begin{bmatrix}
(z_i(k_{ci})) \\
\vdots \\
(z_i(k_{ci}))
\end{bmatrix} = \begin{bmatrix}
[z_{Fi}^1] \\
\vdots \\
[z_{Fi}^c]
\end{bmatrix}
\]

if condition (11) holds and the abstracted state set \( \tilde{Z} = \tilde{Z} \) in the abstracted model (14) is strongly connected.

Proof. Similar to the proof of Lemma 2. \( \square \)

Finally, the Theorem 7 follows from Lemma 2 and Lemma 6 and states the main result:

Theorem 7. (Cooperative target state control). In the networked discrete-event system

\[ \hat{\Sigma} = (\hat{\Sigma}_1, ..., \hat{\Sigma}_i, \Sigma_K, \Sigma'_1, ..., \Sigma'_i, \Sigma_N) \]

in Fig. 1(a), the Problem 3 is solved if condition (11) is satisfied and if the abstracted system set \( \hat{\tilde{Z}} = \hat{\tilde{Z}} \) in the abstracted model (14) is strongly connected.

4. COOPERATIVE CONTROL OF THE FLEXIBLE MANUFACTURING SYSTEM

4.1 Model of the flexible manufacturing system

In this experimental setup, we consider the flexible manufacturing system (see Section 1.1) with three workpieces highlighted in Fig. 2 in grey, red and green. For simplification we reduce each subsystem to handle only one workpiece at the same time. Each subsystem \( \Sigma_{Pi} \)

is modelled by three extended I/O-automata \( \Sigma_{Pi} = (\zeta_{ij}, v_{ij}, s_{ij}, R_{ij}, g_{ij}, f_{ij}, z_{ij0}) \) \( (j = 1, 2, 3) \)

\[ \begin{cases} z_{ij} = G_{ij}(z_{ij}, v_{ij}, s_{ij}, t_{ij}) \quad z_{ij0} = z_{ij0} \\ r_{ij} = F^i_{ij}(z_{ij}, v_{ij}, s_{ij}, t_{ij}) \end{cases} \]

that are connected to the subsystems (1) by physical couplings

\[ \Sigma_{K_i} : (s^i_1, ..., s^i_J)^T = K^i_{L}((r^i_1, ..., r^i_L)^T) \]

with the coupling function \( K^i_{L} : \prod_{j=1}^3 \mathbb{R}_{ij} \to \prod_{j=1}^3 \mathbb{R}_{ij} \) (Fig. 3). In the following, all sets are modelled by natural numbers \( \zeta_{ij}, v_{ij}, R_{ij}, s_{ij} \subseteq \mathbb{N}_0 \), which are noted as vectors in the composed subsystem (1). Hence, the local coupling functions (20) are modelled by coupling matrices \( K^i_{L} \) and the local coupling relation is written as

\[ (s^i_1, ..., s^i_J)^T = K^i_{L}((r^i_1, ..., r^i_L)^T) \]

for each subsystem.

The area gantry robot \( \Sigma_{P1} \), the line gantry robot \( \Sigma_{P2} \) and the 3-axis robot \( \Sigma_{P3} \) are composed of the extended I/O-automata (19) shown in Figs. 4-6 (abbreviations: gripper \( i \) (GI), handler \( i \) (HI), workpiece \( i \) (Wi), magazine (M), conveyor belt (C) and position \( i \) (Pi)).

Gripper model \( \Sigma_{P1i} \): Pneumatic gripper for gripping workpieces with \( \Sigma_{P1i} = \{i \in \text{GI up without W1, } 2 \in \text{GI up with W1, } 3 \in \text{GI down without W1, } 4 \in \text{GI down with W1} \} \), \( \Sigma_{P2i} = \{i \in \text{GI up without W2, } 2 \in \text{GI up with W2, } 3 \in \text{GI down without W2, } 4 \in \text{GI down with W2} \} \), \( \Sigma_{P3i} = \{i \in \text{GI up without W3, } 2 \in \text{GI up with W3, } 3 \in \text{GI down without W3, } 4 \in \text{GI down with W3} \} \).

Handler model \( \Sigma_{P2i} \): Handler for the movement of the gripper with \( \Sigma_{P2i} = \{i \in \text{GI over the M, } 2 \in \text{GI above the C, } 3 \in \text{GI above storage position 1, } 4 \in \text{GI above storage position 2, } 5 \in \text{GI above waiting position} \} \), \( \Sigma_{P2i} = \{i \in \text{GI up with W2, } 2 \in \text{GI up without W2, } 3 \in \text{GI down without W2, } 4 \in \text{GI down with W2} \} \).

Workpiece model \( \Sigma_{P3i} \): Position of the workpiece with \( \Sigma_{P3i} = \{i \in \text{GI up without W3, } 2 \in \text{GI above the C, } 3 \in \text{GI above storage position 1, } 4 \in \text{GI above storage position 2, } 5 \in \text{GI above storage position 1, } 6 \in \text{GI at the G1} \} \),
\[ Z_{P23} = \{ 1 = \text{W2 at G2}, 2 = \text{W2 at handover position}, 3 = \text{No W2 in subsystem}, 4 = \text{W2 at storage position 1}, 5 = \text{W2 at storage position 2} \} \]

\[ Z_{P32} = \{ 1 = \text{W in M}, 2 = \text{W3 at G3}, 3 = \text{W3 at HP of G2}, 4 = \text{W3 at P1}, 5 = \text{W3 at HP of C}, 6 = \text{W3 at P2} \} \]

The conveyor belt is composed of the motor model \( \Sigma_{P41} \), the sensor model of the belt \( \Sigma_{P42} \) and the workpiece model \( \Sigma_{P43} \) (Fig. 7).

Conveyor belt models: Motor, belt and workpieces with \( \Sigma_{P41} = \{ 1 = \text{Motor on}, 2 = \text{Motor off} \} \), \( \Sigma_{P42} = \{ 1 = \text{C empty}, 2 = \text{W4 at HP1}, 3 = \text{W4 at position 2}, 4 = \text{W4 at position 3}, 5 = \text{W4 at HP2} \} \), \( \Sigma_{P43} = \{ 1 = \text{WP4 in subsystem}, 2 = \text{W3 at G1 at HP1}, 3 = \text{W3 at position 2}, 4 = \text{W3 at HP} \} \).

Due to the space limitation, the description of the signals cannot be given.

The global coupling input vectors \( v_{ij} \) with the entry \( v_{ij} \in \mathbb{Z} \). The local coupling input vectors \( s^L_{ij} = (s^L_{i1}, s^L_{i2}, s^L_{i3})^T \) and output vectors \( r^L_{ij} = (r^L_{i1}, r^L_{i2}, r^L_{i3})^T \) are connected according to eqn. (21) with the local coupling matrices

\[ K^L_{1} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}, \quad K^L_{2} = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}, \quad K^L_{3} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \]

In the following section, the states of each subsystem (1) are encoded from of the local states of the subsystems (19) and decoded with

\[ z_i = \sum_{j=1}^{3} z_{ij} \cdot 10^{j-1} \quad \text{and} \quad z_{ij} = \left\lfloor \frac{z_i \mod 10^j}{10^{j-1}} \right\rfloor \]  \hspace{1cm} (22)

The global coupling input vectors \( s = (s_1, s_2, s_3, s_4)^T \) and coupling output vectors \( r = (r_1, r_2, r_3, r_4)^T \) of the subsystems (1) are connected by the physical coupling relation \( s = K \cdot r \) with

\[ K = \begin{pmatrix} 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 \end{pmatrix} \]
4.2 Model abstraction

The composed subsystems (1) (Figs. 4-7) are extended to controlled subsystems (4) by local controllers (3), which are designed by applying Algorithm 1. Finally, the controlled subsystems (4) are abstracted with the equivalence relation (7) to the abstracted subsystems (9), whose extended I/O-automata are shown in Fig. 8 with the following disjoint state sets, which are encoded and decoded by applying eqn. (22):

\[ Z_1 = \{323\} \cup \{212, 222, 232, 242, 252, 412, 422, 432, 442, 452\} \cup \{123\} \cup \{111, 121, 131, 141, 151, 311, 321, 331, 341, 351\} \cup \{114, 115, 124, 125, 134, 135, 144, 145, 154, 155, 216, 226, 236, 246, 256, 314, 315, 324, 325, 334, 335, 344, 345, 354, 355, 416, 436, 456\}, \]

\[ Z_2 = \{451\} \cup \{113, 114, 123, 124, 133, 134, 143, 144, 153, 154, 215, 225, 235, 245, 255, 313, 314, 323, 324, 333, 334, 335, 344, 354, 415, 425, 435, 445, 455\} \cup \{112, 122, 132, 142, 152, 121, 124, 221, 224, 242, 222, 314, 317, 334, 337, 347, 324, 327, 412, 442, 422\}, \]

and

\[ Z_3 = \{216\} \cup \{212, 222, 232, 242, 252, 412, 422, 432\} \cup \{114, 117, 134, 137, 144, 147, 124, 127, 212, 232, 242, 222, 314, 317, 334, 337, 347, 324, 327, 412, 442, 422\}. \]

Fig. 8. Abstracted controlled subsystems \( \hat{\Sigma}_i \)

Note that only the states reachable from the initial states \( z_{i0} \) are shown in the disjoint sets above. Moreover, condition (11) is satisfied because only one local subsystem \( \Sigma_{P3} \) (Workpiece model) connects the controlled subsystems (4) by physical couplings. Thus, the composed abstracted subsystem models (9) in Fig. 8 satisfy condition (11). In the following experiment, we apply only the strongly connected and reachable abstracted state set \( \hat{\mathcal{Z}} \) of the model (14) satisfying the condition of Theorem 7.

4.3 Experimental results of the cooperative control

This section presents the experimental results of the cooperative control of the flexible manufacturing system HANS (Fig. 2). Each controlled subsystem (4) can perform an autonomous transportation task with a workpiece, however, each subsystem is able to handle only one workpiece at the same time. The area gantry robot \( \Sigma_{P2} \) is able to add new workpieces to the process from a storage and the line gantry robot \( \Sigma_{P4} \) is able to remove workpieces from the process, autonomously.

Moreover, the subsystems can transfer the workpieces from the area gantry robot \( \Sigma_{P1} \) to the conveyor belt \( \Sigma_{P4} \), from the conveyor belt to the 3-axis robot \( \Sigma_{P3} \) and, finally, from the 3-axis robot to the line gantry robot \( \Sigma_{P2} \) (double arrows in Fig. 2). The cooperative mode of the network unit (17) is necessary whenever any workpiece has to be transferred between two subsystems.

Figure 9(a) presents the graphs of the experiment showing the state output \( z_i \) and control input \( u_i \) of each controlled subsystem (4) and the control input \( u_i \) of the network units (17). Figure 9(b) depicts the active physical couplings between the subsystems \( \Sigma_{P1} \) as black bars showing when the transfer of the workpieces between the subsystems is performed.

In the time interval \( t = [0, t_1) \), the network units (17) are in the autonomous mode \( e_i = u_i \) (Fig. 9(a)) because each subsystem receives a target state that is reachable without synchronous state transitions and it holds \( [z_{i0}] = [u_i] \).

At time \( t = t_1 \), the area gantry robot receives a new target state to transfer its workpiece to the conveyor belt \( \Sigma_{P4} \), which necessitates cooperation \( [z_i(t_1)] \neq [u_i(t_1)] \) and the network units \( \Sigma_i \) switch into the cooperative mode according to eqn. (17). As the conveyor belt \( \Sigma_{P4} \) has the local task to remove the workpiece from the belt given by the local target state \( u_4 \), it has to transfer the workpiece to the 3-axis robot \( \Sigma_{P3} \). However, as the 3-axis robot \( \Sigma_{P3} \) handles a workpiece (Fig. 2), \( \Sigma_{P3} \) has to transfer this workpiece to the line gantry robot \( \Sigma_{P2} \), before \( \Sigma_{P3} \) is ready for the cooperation with \( \Sigma_{P4} \).

In the physical coupling diagram in Fig. 2(b), it can be seen that the area gantry robot \( \Sigma_{P1} \) transfers the workpiece to the conveyor belt \( \Sigma_{P4} \) in the time interval \( t = [t_1, t_2) \). This transfer results from the target states \( e_i(t) \neq u_i(t) \) determined by the distributed controller (16) being applied by the network unit (17). Moreover, in the time interval \( t = [t_2, t_3) \), the line gantry robot \( \Sigma_{P3} \) brings its workpiece into the storage to be ready for the next cooperation with the 3-axis robot \( \Sigma_{P3} \), which also results from the dependent target states \( e_3(t) \) determined by the abstracted state-feedback controller (Fig. 9(a)). In the time interval \( t = [t_3, t_4) \), the application of the abstracted state-feedback controller leads to the transfer of the green workpiece (Fig. 2) from the 3-axis robot \( \Sigma_{P3} \) to the line gantry robot \( \Sigma_{P2} \) (Fig. 9(b)).

In the time interval \( t = [t_4, t_5) \), the conveyor belt \( \Sigma_{P4} \) transfers its workpiece to the 3-axis robot \( \Sigma_{P3} \) (Fig. 9(b)). Finally, at time \( t_5 \), all cooperating subsystem switch into the autonomous mode according to eqn. (17) and all subsystems reach their target states \( z_i(t_5) = z_i(t_5) \) at time \( t = t_6 \).

5. CONCLUSION

This paper has presented a cooperative control method for networked discrete-event system. The control aim of each subsystem is to reach a given target state, which necessitates...
cooperation in certain situation due to physical restriction. Each controlled subsystem is extended by a network unit implementing a cooperative state-feedback controller that temporarily organises the necessary cooperation between them with the help of the digital communication network. The proposed method has been applied to a flexible manufacturing system and it has been shown that the cooperative control method for networked discrete-event system is applicable to real-life systems.

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