

Safety control with performance guarantees of cooperative systems using compositional abstractions

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- 1 Cooperative control system
- 2 Abstraction-based synthesis
- 3 Compositional synthesis
- 4 Experimental validation

Nonlinear control system:

$$\dot{x} = f(x, u, w)$$

- x : state
- u : control input
- w : disturbance input

Trajectories:

$$\Phi(\cdot, x_0, \mathbf{u}, \mathbf{w})$$

- x_0 : initial state
- \mathbf{u}, \mathbf{w} : control and disturbance functions
- $\Phi(t, x_0, \mathbf{u}, \mathbf{w})$: state reached at time t

Bounded inputs:

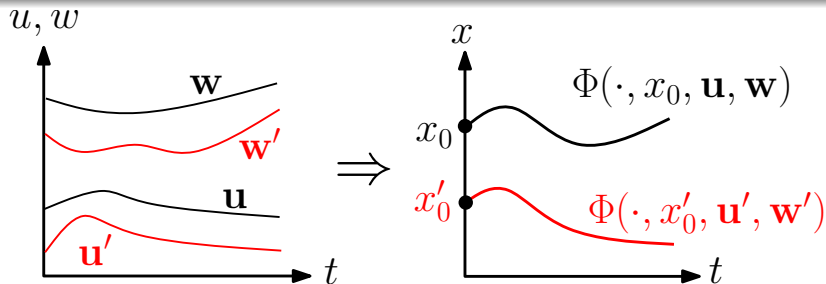
$$u \in [\underline{u}, \bar{u}], \quad w \in [\underline{w}, \bar{w}]$$

Cooperative system

Definition (Cooperativeness)

The system is cooperative if Φ preserves the componentwise inequality:

$$\mathbf{u} \geq \mathbf{u}', \mathbf{w} \geq \mathbf{w}', x_0 \geq x'_0 \Rightarrow \forall t \geq 0, \Phi(t, x, \mathbf{u}, \mathbf{w}) \geq \Phi(t, x', \mathbf{u}', \mathbf{w}')$$



Cooperative systems can be characterized using partial derivatives of f (Angeli and Sontag, 2003).

Outline

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Problem formulation

Sampled dynamics (constant period τ)

→ non-deterministic transition system S :

$$x \xrightarrow{u} x' \Leftrightarrow \exists \mathbf{w} : [0, \tau] \rightarrow [\underline{w}, \overline{w}] \mid x' = \Phi(\tau, x, u, \mathbf{w})$$

Safety specification for S :

$$\forall k \in \mathbb{N}, x(k\tau) \in [\underline{x}, \overline{x}]$$

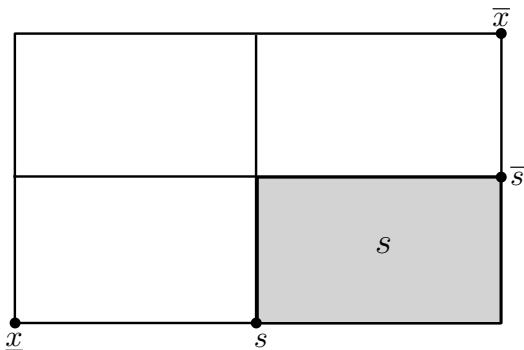
Goal: synthesize a safety controller of S using an abstraction of the system

Symbolic abstraction

Goal: get a **finite** abstraction S_a of the sampled system S

All transitions of S must have an equivalence in S_a

- Discretization of the control space $[\underline{u}, \bar{u}]$
- Partition \mathcal{P}^0 of the safe interval $[\underline{x}, \bar{x}]$ into **symbols**

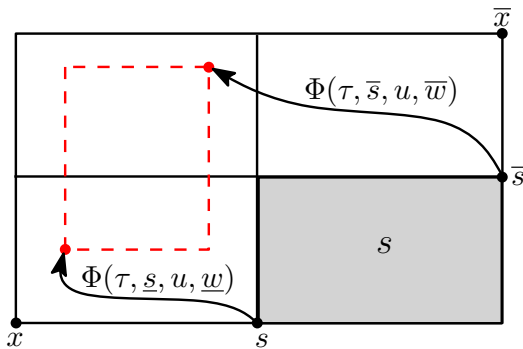


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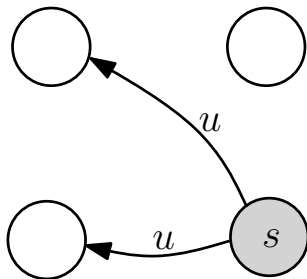


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- Over-approximation of the reachable set
- Intersection with the partition



Safety of S_a in the partition \mathcal{P}^0

Fixed-point algorithm on the operator:

$$F_{\mathcal{P}^0}(Z) = \{s \in Z \cap \mathcal{P}^0 \mid \exists u, \forall s \xrightarrow[u]{a} s', s' \in Z\}$$

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Fixed-point Z_a reached in **finite time**

Z_a is the **maximal safe set** for S_a , associated with the safety controller :

$$C_a(s) = \{u \mid \forall s \xrightarrow{u}_a s', s' \in Z_a\}$$

Theorem

C_a is a safety controller for S in Z_a .

Performance criteria

Performance criterion on a trajectory $(x^0, u^0, x^1, u^1, \dots)$ of S :

$$\sum_{k=0}^{+\infty} \lambda^k g(x^k, u^k)$$

with a cost function g and a **discount factor** $\lambda \in (0, 1)$

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Cost function on S_a : $g_a(s, u) = \max_{x \in s} g(x, u)$

Finite horizon of N sampling periods

Approximation of the performance criterion, if $\lambda^{N+1} \ll 1$:

$$\sum_{k=0}^N \lambda^k g_a(s^k, u^k)$$

Dynamic programming algorithm:

$$J_a^N(s) = \min_{u \in C_a(s)} g_a(s, u)$$

$$J_a^k(s) = \min_{u \in C_a(s)} \left(g_a(s, u) + \lambda \max_{s \xrightarrow[u]{a} s'} J_a^{k+1}(s') \right), \quad \forall k < N$$

$J_a^0(s)$ is the **worst-case minimization** of $\sum_{k=0}^N \lambda^k g_a(s^k, u^k)$

Performance optimization

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Receding horizon controller:

$$C_a^*(s) = \arg \min_{u \in C_a(s)} \left(g_a(s, u) + \lambda \max_{s \xrightarrow{u}_a s'} J_a^1(s') \right)$$

Performance guarantee

$$\text{Let } M_a = \max_{s \in Z_a} \min_{u \in C_a(s)} g_a(s, u)$$

Theorem

Let $(x^0, u^0, x^1, u^1, \dots)$ be a trajectory of S controlled with C_a^* .
Let s^0, s^1, \dots such that $x^k \in s^k$, for all $k \in \mathbb{N}$. Then for all $k \in \mathbb{N}$,

$$\sum_{j=0}^{+\infty} \lambda^j g(x^{k+j}, u^{k+j}) \leq J_a^0(s^k) + \frac{\lambda^{N+1}}{1-\lambda} M_a.$$

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Worst-case optimization on finite horizon:

$$\sum_{j=0}^N \lambda^j g_a(s^{k+j}, u^{k+j}) \leq J_a^0(s^k)$$

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Worst-case of remaining cost, with receding horizon
we minimize at least the current step:

$$\sum_{j=N+1}^{+\infty} \lambda^j g_a(s^{k+j}, u^{k+j}) \leq \sum_{j=N+1}^{+\infty} \lambda^j \max_{s \in Z_a} \min_{u \in C_a(s)} g_a(s, u) = \frac{\lambda^{N+1}}{1-\lambda} M_a.$$

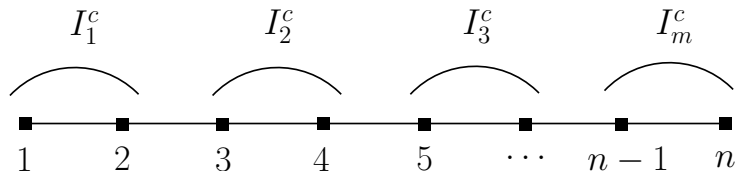
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Decomposition: states

Decomposition into m subsystems:

Partition (I_1^c, \dots, I_m^c) of the state dimensions $\{1, \dots, n\}$



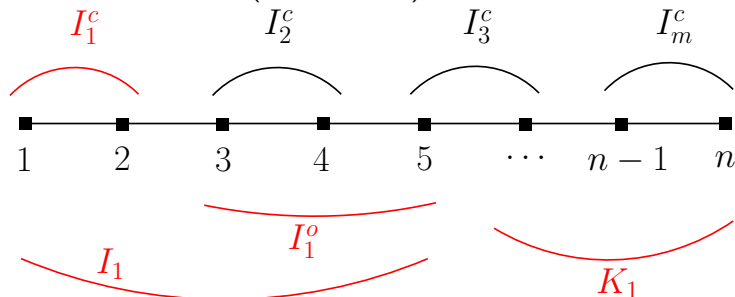
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Subsystem $i \in \{1, \dots, m\}$:

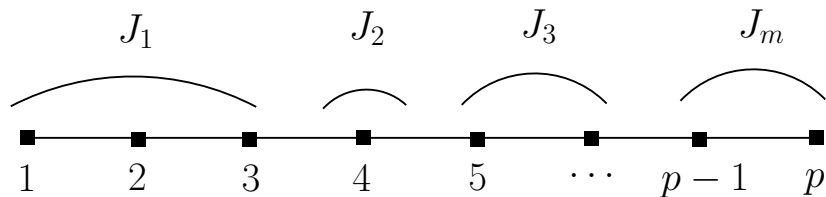
- I_i^c : controlled states
- I_i^o : observed but uncontrolled states
- $I_i = I_i^c \cup I_i^o$: all modeled states
- K_i : unobserved states (disturbances)



Decomposition: inputs

Decomposition into m subsystems:

Partition (J_1, \dots, J_m) of the input dimensions $\{1, \dots, p\}$



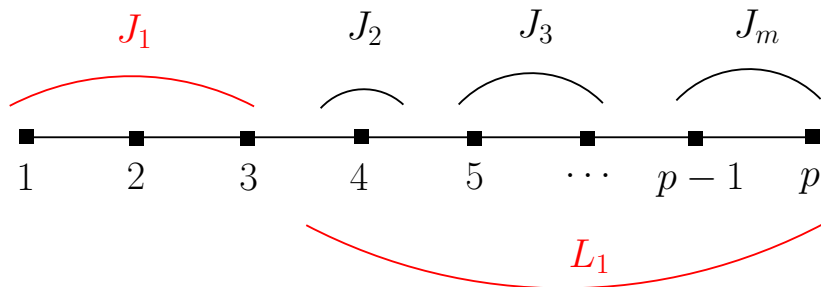
Decomposition: inputs

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Subsystem $i \in \{1, \dots, m\}$:

- J_i : controlled inputs
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Subsystem abstraction

Subsystem $i \in \{1, \dots, m\}$: create a symbolic abstraction S_i
Classical method, but with 2 **assume-guarantee obligations**:

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A/G Obligation (K_i)

Unobserved states: $x_{K_i} \in [\underline{x}_{K_i}, \bar{x}_{K_i}]$

Need **bounded disturbances**: w, x_{K_i}, u_{L_i}

For cooperative system \Rightarrow larger over-approximation of the reachable set

$$\begin{cases} \Phi(\tau, (\underline{s}_i, \underline{x}_{K_i}), (u_{J_i}, \underline{u}_{L_i}), \underline{w}) \leq \Phi(\tau, \underline{s}, u, \underline{w}), \\ \Phi(\tau, (\bar{s}_i, \bar{x}_{K_i}), (u_{J_i}, \bar{u}_{L_i}), \bar{w}) \geq \Phi(\tau, \bar{s}, u, \bar{w}). \end{cases}$$

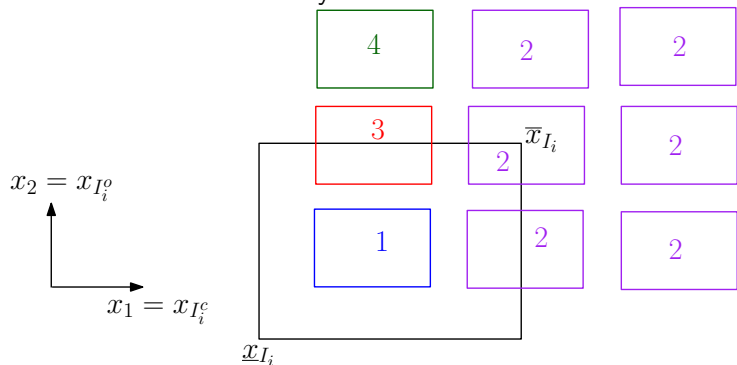
Subsystem abstraction

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Classical method, but with 2 **assume-guarantee obligations**:

A/G Obligation (I_i^o)

Observed but uncontrolled states: $x_{I_i^o} \in [\underline{x}_{I_i^o}, \bar{x}_{I_i^o}]$

Remove transitions where only uncontrolled states violate the safety



Same methods than the centralized approach
with a focus only on controlled states: $x_{I_i^c}$

Safety in the partition of $[\underline{x}_{I_i}, \bar{x}_{I_i}]$

- (safety already realized for uncontrolled states $x_{I_i^o}$)
- maximal safe set: Z_i
- safety controller: C_i

Performances

- cost function $g_i(s_{I_i}, u_{J_i})$
- g_i only depends on the controlled components $s_{I_i^c}$
- deterministic controller: C_i^*

S_c : composition of all subsystems S_i

- Safe transitions: $s \xrightarrow[c]{u} s' \iff \forall i \in \{1, \dots, m\}, s_{l_i} \xrightarrow[i]{u_{j_i}} s'_{l_i}$
- Unsafe transition: $s \xrightarrow[c]{u} Out \iff \exists i \in \{1, \dots, m\} \mid s_{l_i} \xrightarrow[i]{u_{j_i}} Out_i$

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Proposition (Alternating simulation)

Let a symbol s , a state $x \in s$ and a discrete input u .

For any successor x' in S ($x \xrightarrow{u} x'$), the corresponding symbol ($s' \mid x' \in s'$) is a successor in S_c ($s \xrightarrow[c]{u} s'$).

Composition of safe sets and safety controllers:

- $Z_c = Z_1 \cap \dots \cap Z_m$
- $\forall s \in Z_c, C_c(s) = C_1(s_{I_1}) \times \dots \times C_m(s_{I_m})$

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Proposition (Safety comparison)

$Z_c \subseteq Z_a$.

Composition of the deterministic controllers:

- $\forall s \in Z_c, C_c^*(s) = (C_1^*(s_{l_1}), \dots, C_m^*(s_{l_m}))$
- Let $M_i = \max_{s_i \in Z_i} \min_{u_i \in C_i(s_i)} g_i(s_i, u_i)$

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Theorem (Performance guarantee)

Let $(x^0, u^0, x^1, u^1, \dots)$ be a trajectory of S controlled with C_c^* .
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Performances

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Proposition (Guarantees comparison)

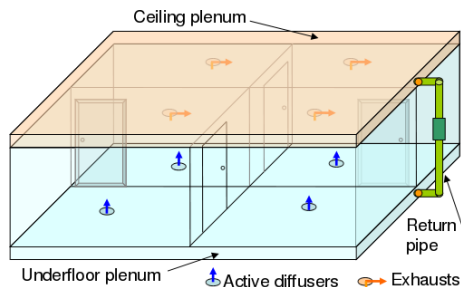
$$\forall s \in Z_c, \quad J_a^0(s) + \frac{\lambda^{N+1}}{1-\lambda} M_a \leq \sum_{i=1}^m J_i^0(s_{l_i}) + \frac{\lambda^{N+1}}{1-\lambda} \sum_{i=1}^m M_i$$

- n : state space dimension
- p : control space dimension
- $\alpha_x \in \mathbb{N}$: precision of the state partition (per dimension)
- $\alpha_u \in \mathbb{N}$: precision of the input discretization (per dimension)
- $|\cdot|$: cardinality of a set

	Method	
	Centralized	Compositional
Abstraction (successors computed)	$2\alpha_x^n \alpha_u^p$	$\sum_{i=1}^m 2\alpha_x^{ I_i } \alpha_u^{ J_i }$
Dynamic programming (max iterations)	$N\alpha_x^{2n} \alpha_u^p$	$\sum_{i=1}^m N\alpha_x^{2 I_i } \alpha_u^{ J_i }$

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UnderFloor Air Distribution



- Underfloor air cooled down
- Sent into the rooms by fans
- Air excess pushed through the ceiling exhausts
- Returned to the underfloor
- Disturbances: heat sources; opening of doors

Assume a **uniform temperature** in each room
Temperature variations obtained from **energy and mass conservation** equations in each room

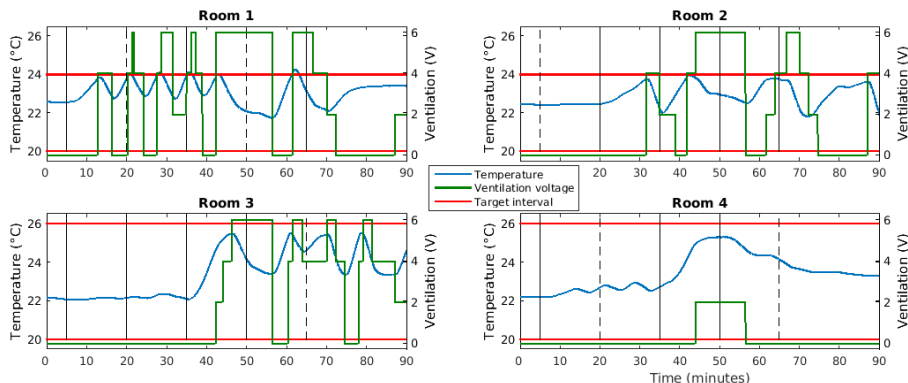
Our model: $\dot{T} = f(T, u, w, \delta)$

- $T \in \mathbb{R}^4$: state (temperature)
- $u \in \mathbb{R}^4$: controlled input (fan air flow)
- w : exogenous input (other temperatures)
- δ : discrete disturbance

The system is **cooperative**

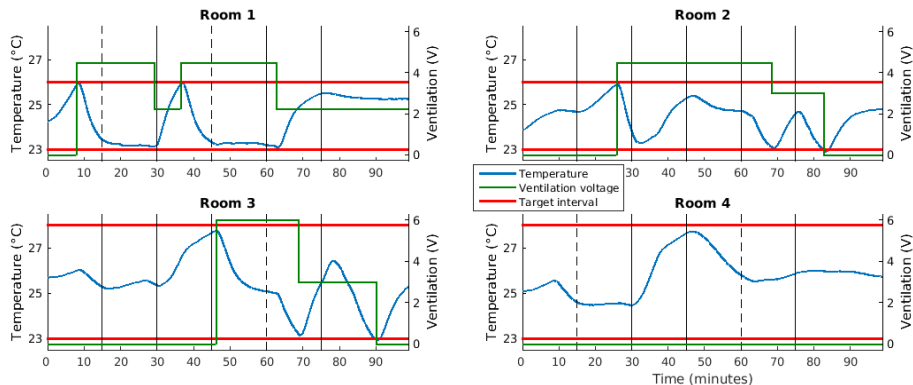
Centralized symbolic control

- $\alpha_x = 10$, $\alpha_u = 4$, $\tau = 34$ s
- $g_a(s^k, u^k, u^{k-1}) = \|u^k\| + \|u^k - u^{k-1}\| + \|s_*^k - T_*\|$
- $N = 5$, $\lambda = 0.5$: $\lambda^{N+1} \approx 3\%$
- Computation time: more than 2 days



Compositional symbolic control

- 1D subsystems: $I_i = I_i^c = J_i = i$ and $I_i^o = \emptyset$
- $\alpha_x = 20$, $\alpha_u = 9$, $\tau = 10$ s
- Computation time: 1.1 s



Compositional approach to addresses scalability issue of symbolic control:

- Decomposition into partial description of the system
- Tradeoff between model accuracy and complexity reduction
- Classical symbolic methods applied to each subsystem under assume-guarantee obligations
- Composition preserves safety and performance guarantees

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Perspectives

- Extension to observed but uncontrolled inputs
- Adaptive symbolic control framework:
 - measure the disturbance; tight estimation of its future bounds
 - synthesize compositional controller on the more accurate abstraction
 - apply controller until the next measure

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