

Invariance and symbolic control of cooperative systems for temperature regulation in intelligent buildings

Pierre-Jean Meyer

Université Grenoble-Alpes

PhD Defense, September 24th 2015



gipsa-lab

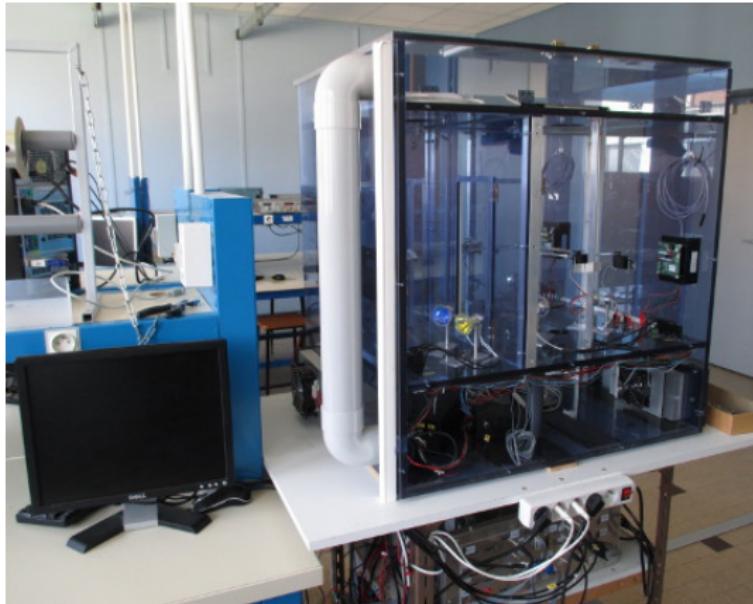


Motivations



Develop new control methods
for intelligent buildings

Focus on temperature control
in a small-scale experimental
building



Outline

- 1 Monotone control system
- 2 Robust controlled invariance
- 3 Symbolic control
- 4 Compositional approach
- 5 Control in intelligent buildings

System description

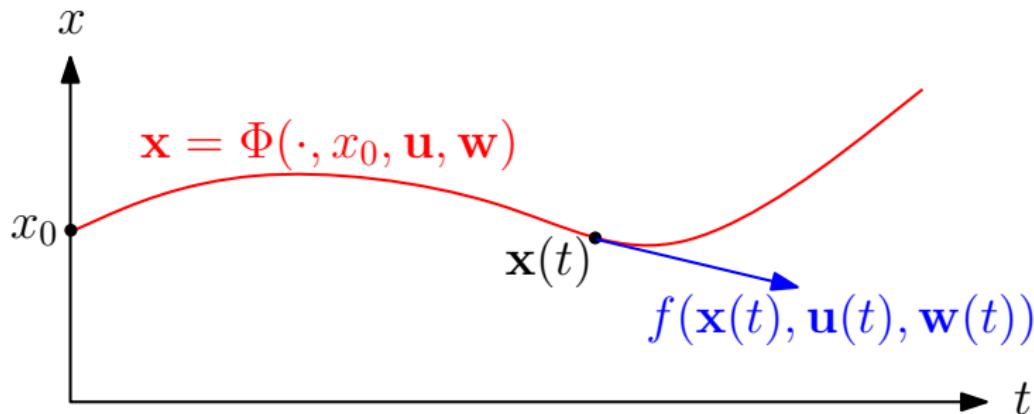
Nonlinear control system:

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

Trajectories:

$$x = \Phi(\cdot, x_0, u, w)$$

- x : state
- u : control input
- w : disturbance input
- x, u, w : time functions

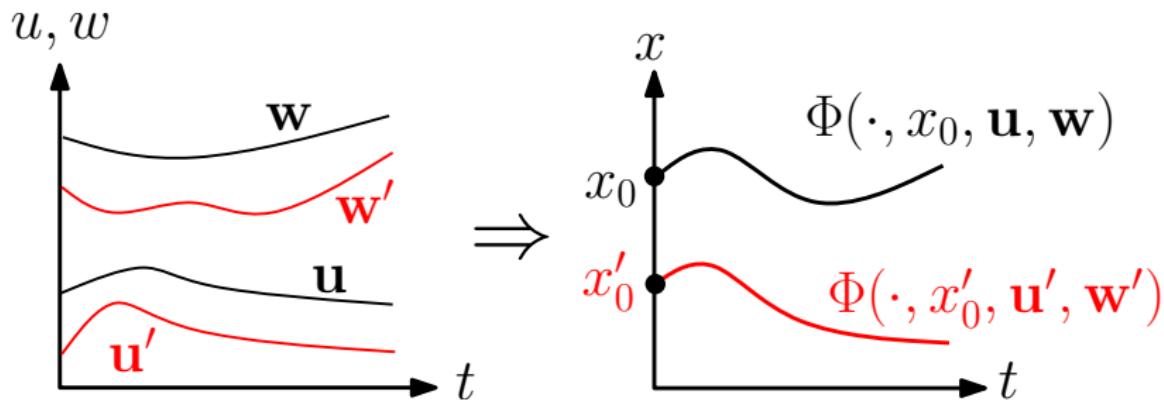


Monotone system

Definition (Monotonicity)

The system is monotone if Φ preserves the componentwise inequality:

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



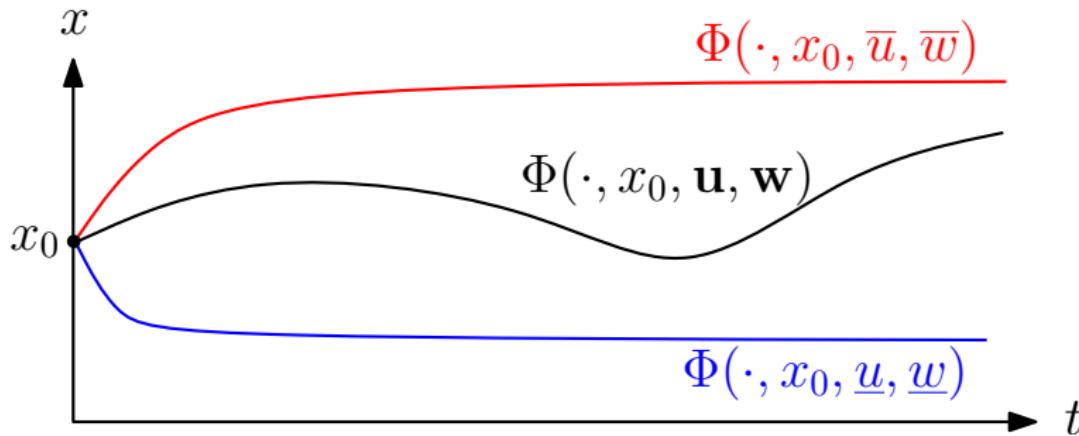
Bounded inputs

Control and disturbance inputs bounded in intervals:

$$\forall t \geq 0, \mathbf{u}(t) \in [\underline{u}, \bar{u}], \mathbf{w}(t) \in [\underline{w}, \bar{w}]$$

$$\implies$$

$$\forall t \geq 0, \Phi(t, x_0, \mathbf{u}, \mathbf{w}) \in [\Phi(t, x_0, \underline{u}, \underline{w}), \Phi(t, x_0, \bar{u}, \bar{w})]$$



Characterization

Proposition (Kamke-Müller)

The system $\dot{x} = f(x, u, w)$ is monotone if and only if the following implication holds for all i :

$$u \geq u', \quad w \geq w', \quad x \geq x', \quad x_i = x'_i \Rightarrow f_i(x, u, w) \geq f_i(x', u', w')$$

Proposition (Partial derivatives)

The system $\dot{x} = f(x, u, w)$ with continuously differentiable vector field f is monotone if and only if:

$$\forall i, j \neq i, k, l, \frac{\partial f_i}{\partial x_j} \geq 0, \quad \frac{\partial f_i}{\partial u_k} \geq 0, \quad \frac{\partial f_i}{\partial w_l} \geq 0$$

Outline

- 1 Monotone control system
- 2 Robust controlled invariance
- 3 Symbolic control
- 4 Compositional approach
- 5 Control in intelligent buildings

Definitions

Definition (Robust Controlled Invariance)

A set \mathcal{S} is a *robust controlled invariant* if there exists a controller such that the closed-loop system stays in \mathcal{S} for any initial state and disturbance:

$$\exists \mathbf{u} : \mathcal{S} \rightarrow [\underline{u}, \bar{u}] \mid \forall x_0 \in \mathcal{S}, \forall \mathbf{w} \in [\underline{w}, \bar{w}], \forall t \geq 0, \Phi_{\mathbf{u}}(t, x_0, \mathbf{w}) \in \mathcal{S}$$

Definitions

Definition (Robust Controlled Invariance)

A set \mathcal{S} is a *robust controlled invariant* if there exists a controller such that the closed-loop system stays in \mathcal{S} for any initial state and disturbance:

$$\exists u : \mathcal{S} \rightarrow [\underline{u}, \bar{u}] \mid \forall x_0 \in \mathcal{S}, \forall w \in [\underline{w}, \bar{w}], \forall t \geq 0, \Phi_u(t, x_0, w) \in \mathcal{S}$$

Definition (Local control)

Each control input affects a single state variable:

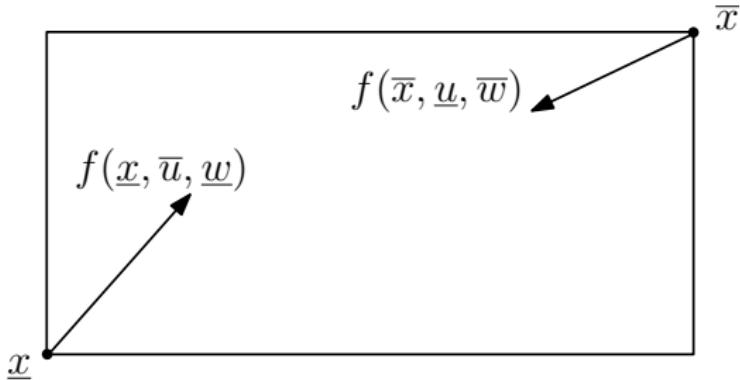
$$\forall k, \exists i \mid \frac{\partial f_i}{\partial u_k} \neq 0$$

Robust controlled invariance

Theorem (Meyer, Girard, Witrant, CDC 2013)

With the monotonicity and local control properties, the interval $[\underline{x}, \bar{x}] \subseteq \mathbb{R}^n$ is robust controlled invariant if and only if

$$\begin{cases} f(\bar{x}, \underline{u}, \bar{w}) \leq 0 \\ f(\underline{x}, \bar{u}, \underline{w}) \geq 0 \end{cases}$$

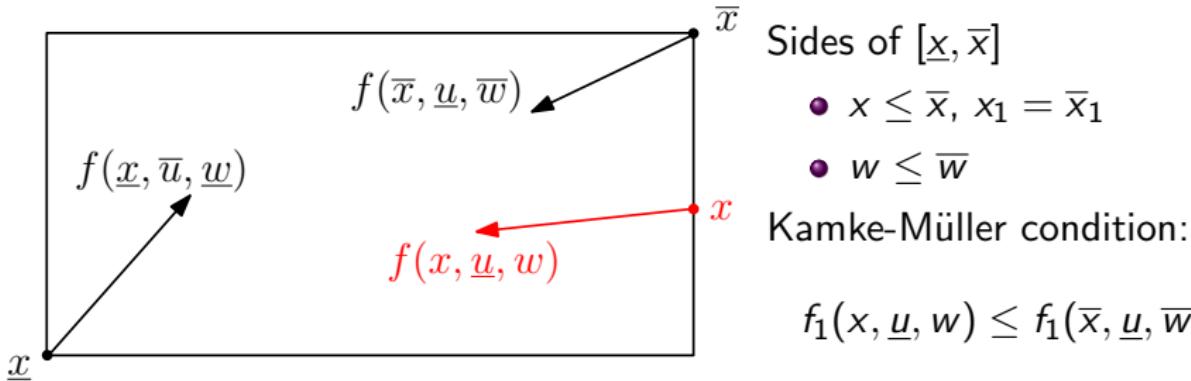


Robust controlled invariance

Theorem (Meyer, Girard, Witrant, CDC 2013)

With the monotonicity and local control properties, the interval $[\underline{x}, \bar{x}] \subseteq \mathbb{R}^n$ is robust controlled invariant if and only if

$$\begin{cases} f(\bar{x}, \underline{u}, \bar{w}) \leq 0 \\ f(\underline{x}, \bar{u}, \underline{w}) \geq 0 \end{cases}$$

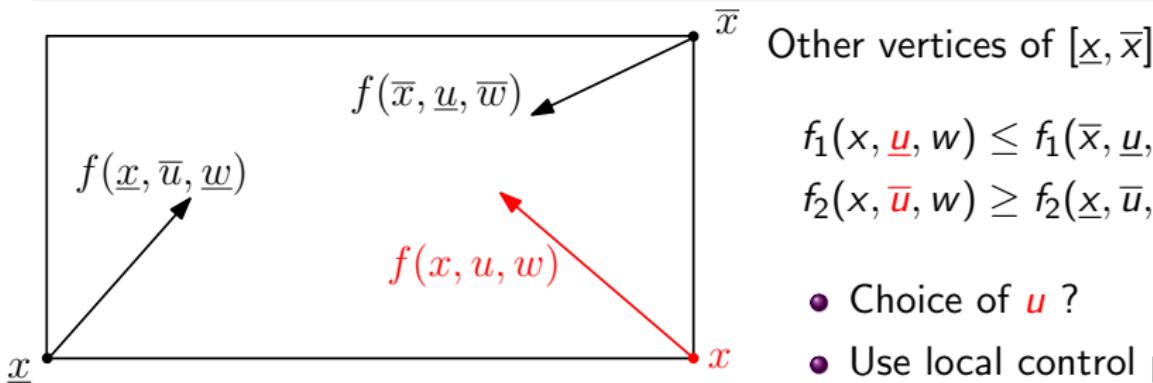


Robust controlled invariance

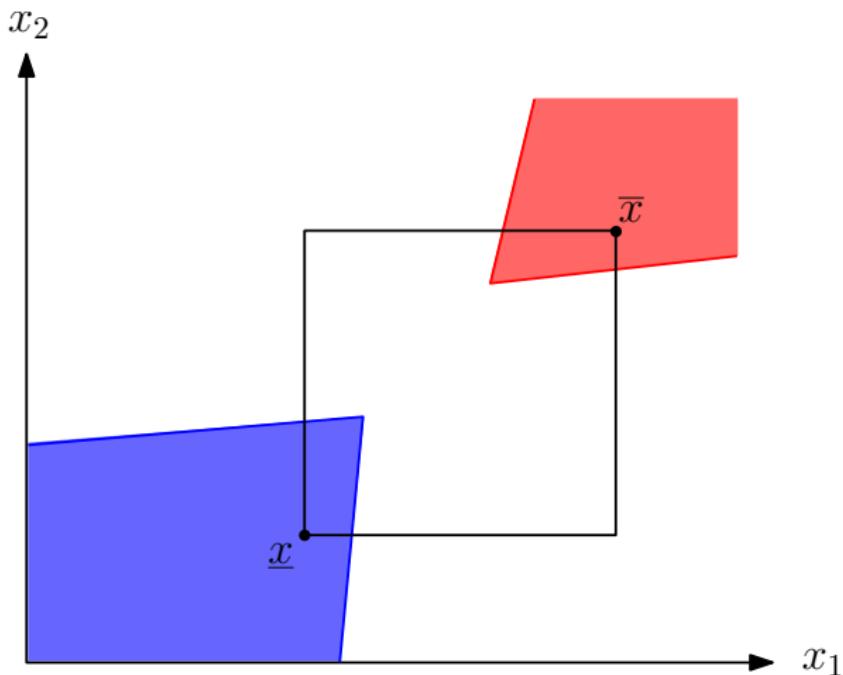
Theorem (Meyer, Girard, Witrant, CDC 2013)

With the monotonicity and local control properties, the interval $[\underline{x}, \bar{x}] \subseteq \mathbb{R}^n$ is robust controlled invariant if and only if

$$\begin{cases} f(\bar{x}, \underline{u}, \bar{w}) \leq 0 \\ f(\underline{x}, \bar{u}, \underline{w}) \geq 0 \end{cases}$$



Choice of the interval



$$x \in \mathbb{R}^2$$

2 conditions on \bar{x}

- $f_1(\bar{x}, \underline{u}, \bar{w}) \leq 0$
- $f_2(\bar{x}, \underline{u}, \bar{w}) \leq 0$

2 conditions on \underline{x}

- $f_1(\underline{x}, \bar{u}, \underline{w}) \geq 0$
- $f_2(\underline{x}, \bar{u}, \underline{w}) \geq 0$

Robust set stabilization

Definition (Stabilizing controller)

A controller $u : \mathcal{S}_0 \rightarrow [\underline{u}, \bar{u}]$ is a stabilizing controller from \mathcal{S}_0 to \mathcal{S} if

$$\forall x_0 \in \mathcal{S}_0, \quad \forall w \in [\underline{w}, \bar{w}], \quad \exists T \geq 0 \mid \forall t \geq T, \quad \Phi_u(t, x_0, w) \in \mathcal{S}$$

Robust set stabilization

Definition (Stabilizing controller)

A controller $u : \mathcal{S}_0 \rightarrow [\underline{u}, \bar{u}]$ is a stabilizing controller from \mathcal{S}_0 to \mathcal{S} if

$$\forall x_0 \in \mathcal{S}_0, \quad \forall w \in [\underline{w}, \bar{w}], \quad \exists T \geq 0 \mid \forall t \geq T, \quad \Phi_u(t, x_0, w) \in \mathcal{S}$$

Let $\mathcal{S}_0 = [\underline{x}_0, \bar{x}_0]$ and $\mathcal{S} = [\underline{x}, \bar{x}] \subseteq [\underline{x}_0, \bar{x}_0]$

Assumption

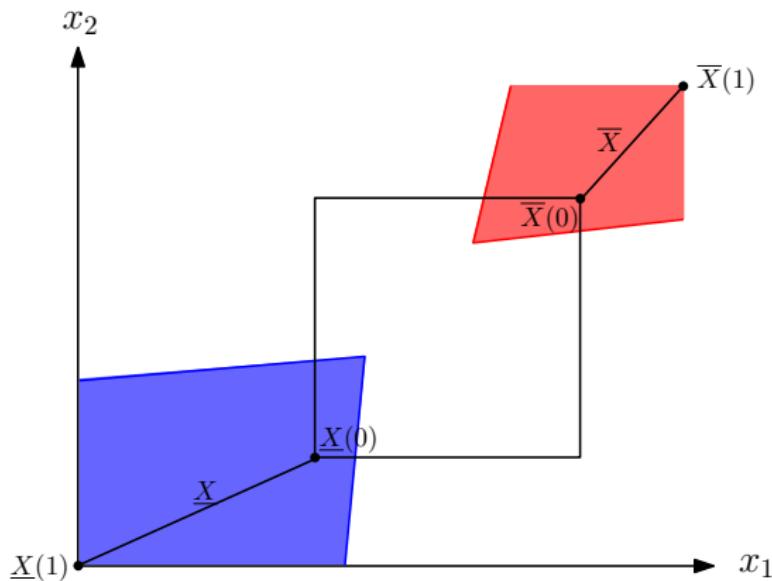
$\exists \underline{X}, \bar{X} : [0, 1] \rightarrow \mathbb{R}^n$, respectively **strictly decreasing and increasing**, such that $[\underline{X}(1), \bar{X}(1)] = [\underline{x}_0, \bar{x}_0]$, $[\underline{X}(0), \bar{X}(0)] = [\underline{x}, \bar{x}]$ and satisfying

$$f(\underline{X}(\lambda), \bar{u}, \underline{w}) > 0, \quad f(\bar{X}(\lambda), u, \bar{w}) < 0, \quad \forall \lambda \in [0, 1]$$

Robust set stabilization

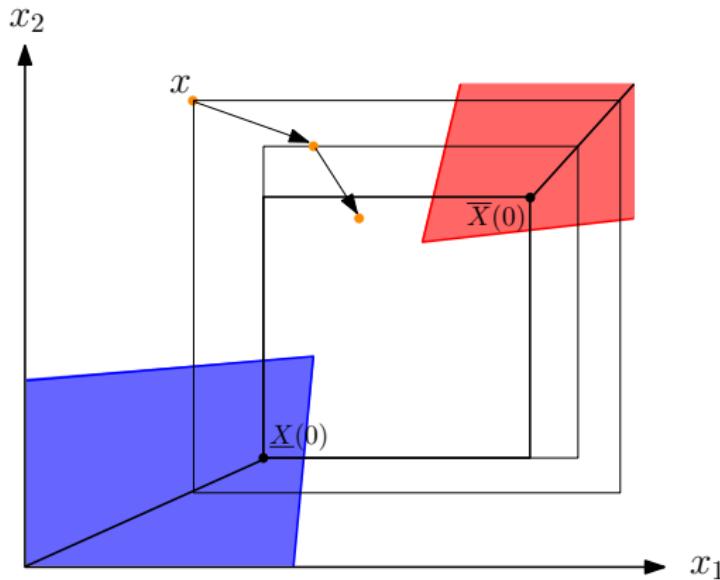
$$f(\underline{X}(\lambda), \bar{u}, \underline{w}) > 0, \quad f(\bar{X}(\lambda), \underline{u}, \bar{w}) < 0, \quad \forall \lambda \in [0, 1]$$

$\forall \lambda, \lambda' \in [0, 1], [\underline{X}(\lambda), \bar{X}(\lambda')] \text{ is a robust controlled invariant interval}$



Robust set stabilization

Take the smallest interval of this family containing the current state x
Apply any invariance controller in this interval



Robust set stabilization

$$\begin{cases} \bar{\lambda}(x) = \min\{\lambda \in [0, 1] \mid \bar{X}(\lambda) \geq x\} \\ \underline{\lambda}(x) = \min\{\lambda \in [0, 1] \mid \underline{X}(\lambda) \leq x\} \end{cases}$$

$[\underline{X}(\underline{\lambda}(x)), \bar{X}(\bar{\lambda}(x))]$ is the smallest interval containing the current state x

Robust set stabilization

$$\begin{cases} \bar{\lambda}(x) = \min\{\lambda \in [0, 1] \mid \bar{X}(\lambda) \geq x\} \\ \underline{\lambda}(x) = \min\{\lambda \in [0, 1] \mid \underline{X}(\lambda) \leq x\} \end{cases}$$

$[\underline{X}(\underline{\lambda}(x)), \bar{X}(\bar{\lambda}(x))]$ is the smallest interval containing the current state x

Invariance controller $u_i(x) = \underline{u}_i + (\bar{u}_i - \underline{u}_i) \frac{\bar{x}_i - x_i}{\bar{x}_i - \underline{x}_i}$	Candidate stabilizing controller $u_i(x) = \underline{u}_i + (\bar{u}_i - \underline{u}_i) \frac{\bar{X}_i(\bar{\lambda}(x)) - x_i}{\bar{X}_i(\bar{\lambda}(x)) - \underline{X}_i(\underline{\lambda}(x))}$ (1)
---	---

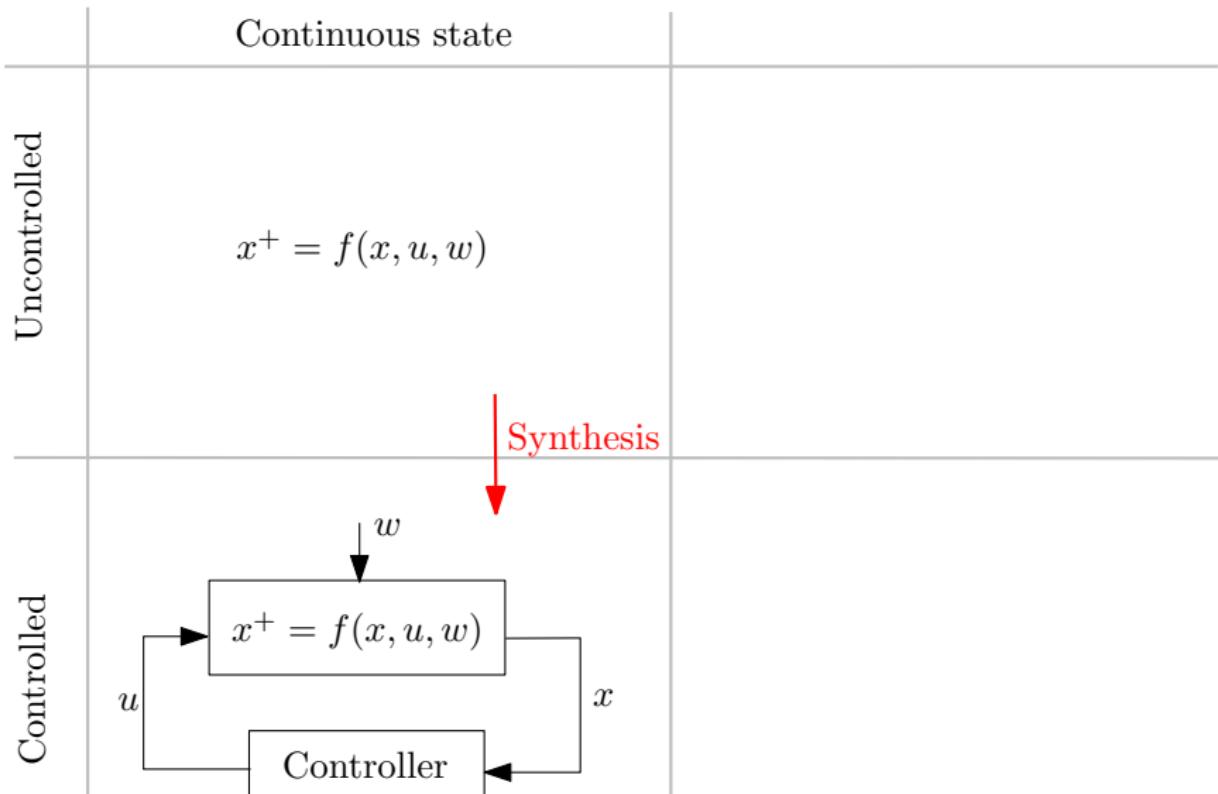
Theorem (Meyer, Girard, Witrant, prov. accepted in Automatica)

(1) is a stabilizing controller from $[\underline{X}(1), \bar{X}(1)]$ to $[\underline{X}(0), \bar{X}(0)]$.

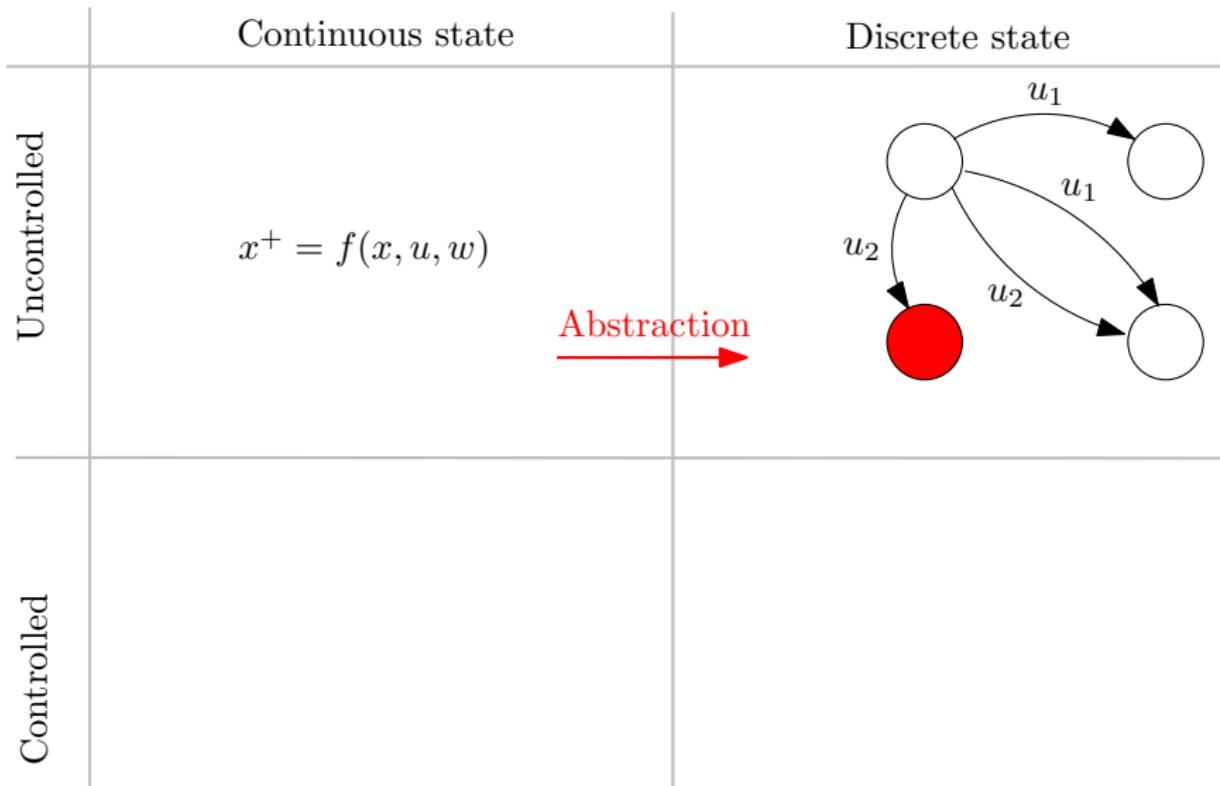
Outline

- 1 Monotone control system
- 2 Robust controlled invariance
- 3 Symbolic control
- 4 Compositional approach
- 5 Control in intelligent buildings

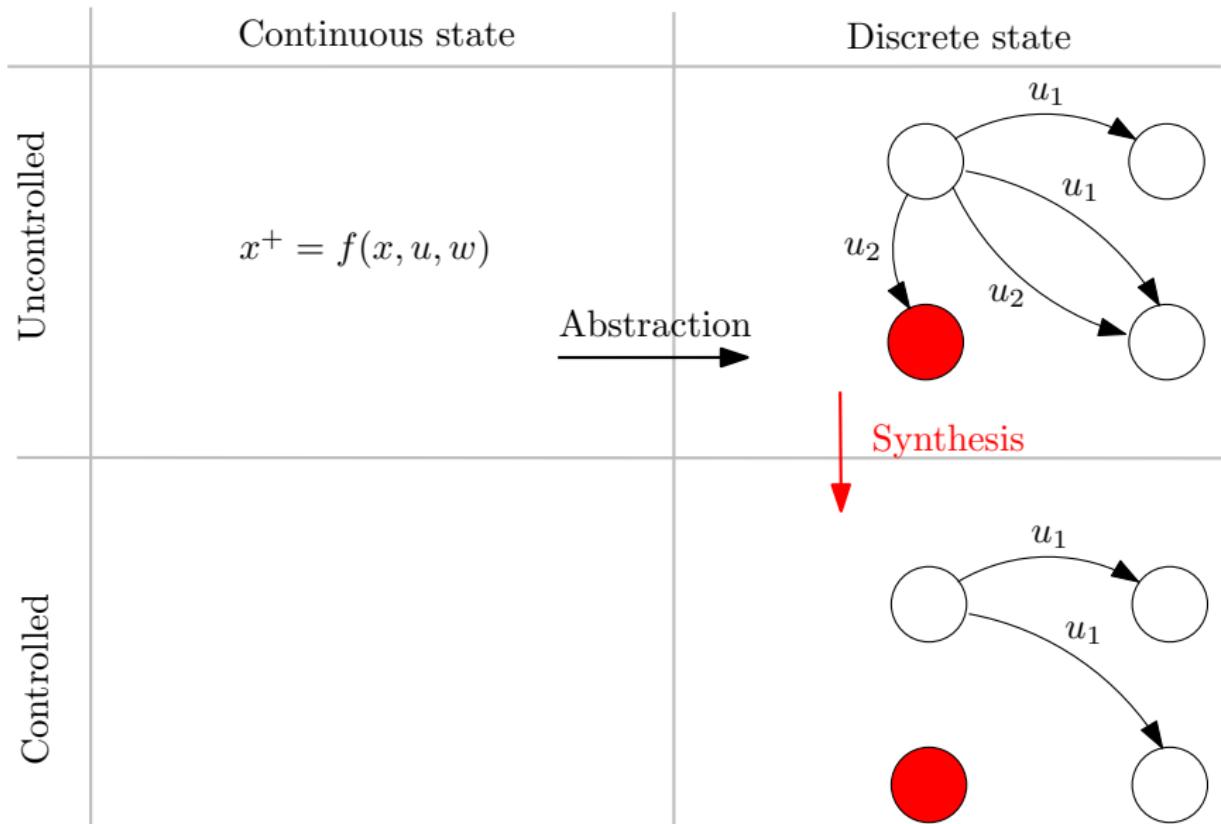
Abstraction-based synthesis



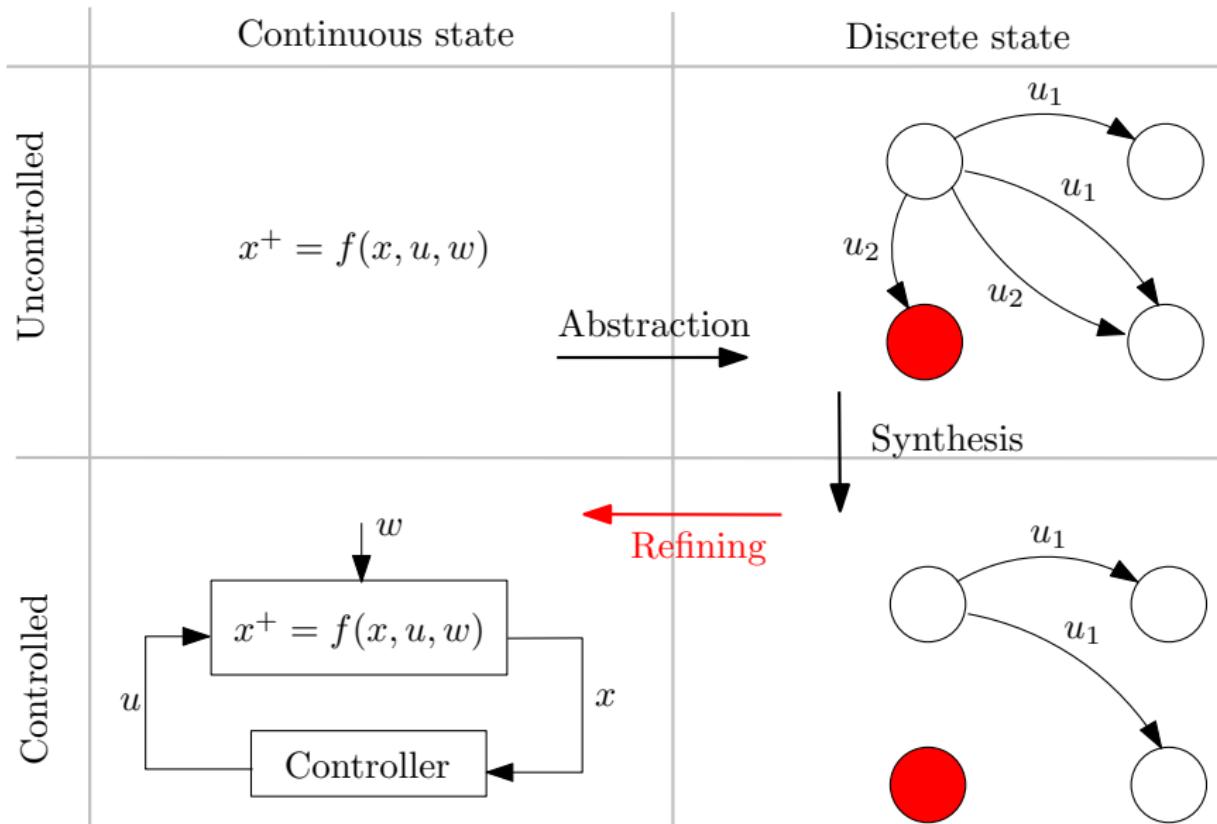
Abstraction-based synthesis



Abstraction-based synthesis



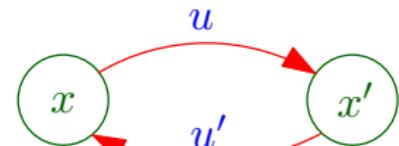
Abstraction-based synthesis



Transition systems

$$S = (X, U, \rightarrow)$$

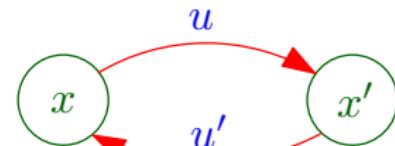
- Set of states X
- Set of inputs U
- Transition relation \rightarrow
- Trajectories: $x_1 \xrightarrow{u_1} x_2 \xrightarrow{u_2} x_3 \xrightarrow{u_3} \dots$



Transition systems

$$S = (X, U, \rightarrow)$$

- Set of states X
- Set of inputs U
- Transition relation \rightarrow
- Trajectories: $x_1 \xrightarrow{u_1} x_2 \xrightarrow{u_2} x_3 \xrightarrow{u_3} \dots$

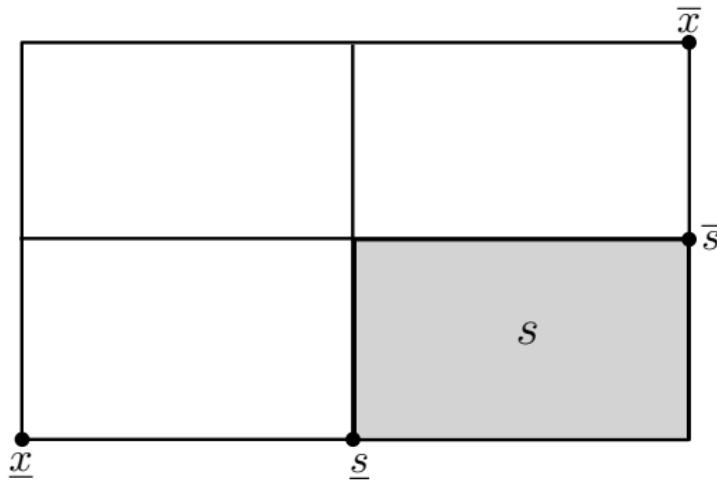


Sampled dynamics (sampling τ)

- $X = \mathbb{R}^n$
- $U = [\underline{u}, \bar{u}]$
- $x \xrightarrow{u} x' \iff \exists \mathbf{w} : [0, \tau] \rightarrow [\underline{w}, \bar{w}] \mid x' = \Phi(\tau, x, u, \mathbf{w})$
- Safety specification in $[\underline{x}, \bar{x}] \subseteq \mathbb{R}^n$

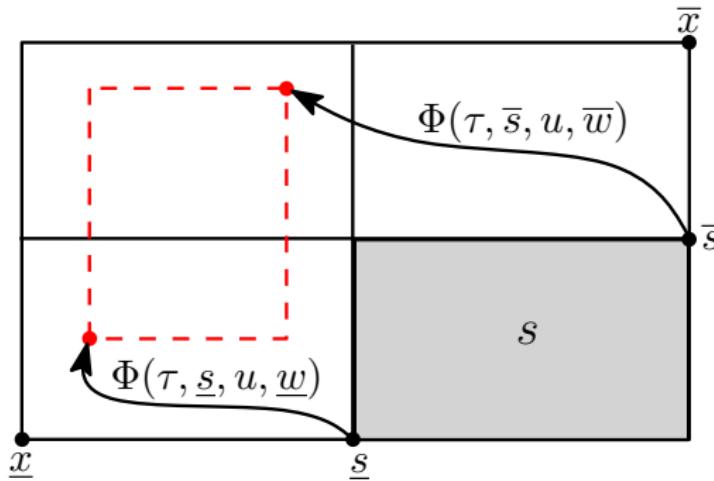
Abstraction

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



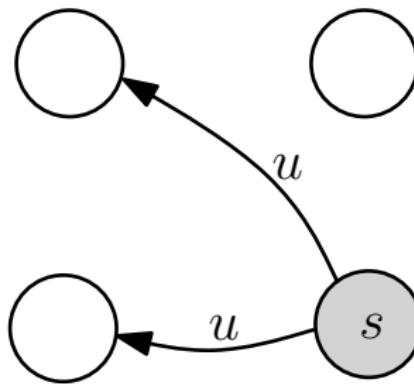
Abstraction

- Discretization of the control space $[\underline{u}, \bar{u}]$
- Partition \mathcal{P} of the interval $[\underline{x}, \bar{x}]$ into **symbols**
- Over-approximation of the reachable set (monotonicity)



Abstraction

- Discretization of the control space $[\underline{u}, \bar{u}]$
- Partition \mathcal{P} of the interval $[\underline{x}, \bar{x}]$ into **symbols**
- Over-approximation of the reachable set (monotonicity)
- Intersection with the partition



Obtain a finite abstraction $S_a = (X_a, U_a, \xrightarrow{a})$

Alternating simulation

Definition (Alternating simulation relation)

$H : X \rightarrow X_a$ is an alternating simulation relation from S_a to S if:

$$\forall u_a \in U_a, \exists u \in U \mid x \xrightarrow{u} x' \text{ in } S \implies H(x) \xrightarrow[a]{u_a} H(x') \text{ in } S_a$$

Alternating simulation

Definition (Alternating simulation relation)

$H : X \rightarrow X_a$ is an alternating simulation relation from S_a to S if:

$$\forall u_a \in U_a, \exists u \in U \mid x \xrightarrow{u} x' \text{ in } S \implies H(x) \xrightarrow[a]{u_a} H(x') \text{ in } S_a$$

Proposition

The map $H : X \rightarrow X_a$ defined by

$$H(x) = s \iff x \in s$$

is an alternating simulation relation from S_a to S :

$$\forall u_a \in U_a \subseteq U \mid x \xrightarrow{u_a} x' \text{ in } S \implies H(x) \xrightarrow[a]{u_a} H(x') \text{ in } S_a$$

Safety synthesis

Specification: safety of S_a in \mathcal{P} (the partition of the interval $[\underline{x}, \bar{x}]$)

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

Safety synthesis

Specification: safety of S_a in \mathcal{P} (the partition of the interval $[\underline{x}, \bar{x}]$)

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

Fixed-point Z_a of $F_{\mathcal{P}}$ reached in **finite time**

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

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

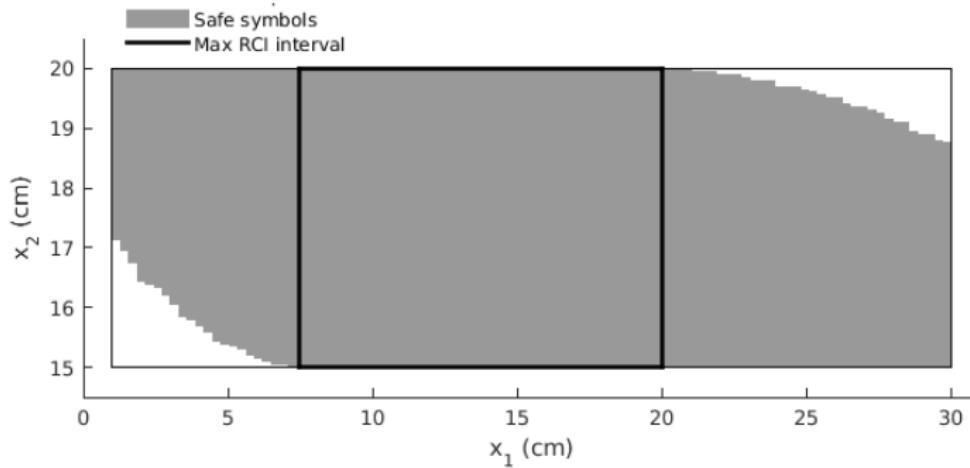
Theorem

C_a is a safety controller for S in Z_a .

Safety vs invariance

2D example with a partition of 100×100 symbols

- Chosen interval is **not** robust controlled invariant
- Compare the safe set Z_a with the largest robust controlled invariant sub-interval



Performance criterion

Minimize 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** $\lambda \in (0, 1)$

Performance criterion

Minimize 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** $\lambda \in (0, 1)$

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

Focus the optimization on a **finite horizon** of N sampling periods

Accurate approximation if $\lambda^{N+1} \ll 1$

Optimization

Dynamic programming algorithm:

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

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

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

Optimization

Dynamic programming algorithm:

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

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

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

Receding horizon controller:

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

Performance guarantees

Theorem (Meyer, Girard, Witrant, HSCC 2015)

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,

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

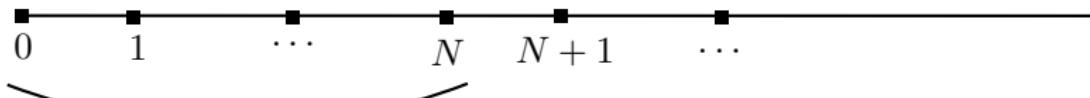
Performance guarantees

Theorem (Meyer, Girard, Witrant, HSCC 2015)

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,

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

Worst-case minimization on finite horizon:



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

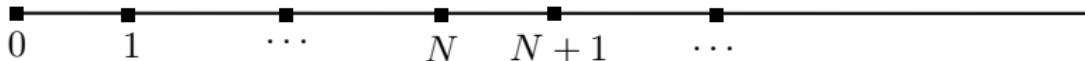
Performance guarantees

Theorem (Meyer, Girard, Witrant, HSCC 2015)

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,

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

Worst-case minimization of each remaining steps (receding horizon):

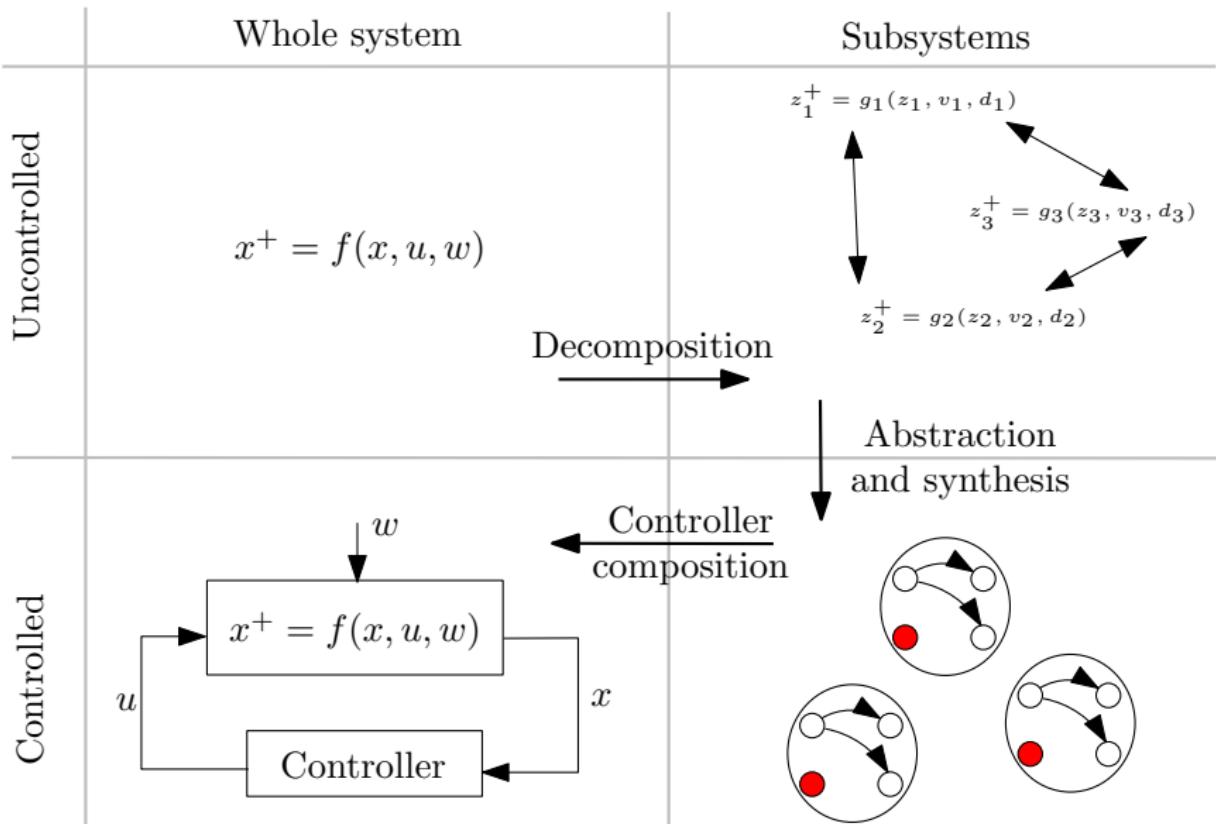


$$g_a(s^k, u^k) \leq \max_{s \in Z_a} \min_{u \in C_a(s)} g_a(s, u) = M_a$$

Outline

- 1 Monotone control system
- 2 Robust controlled invariance
- 3 Symbolic control
- 4 Compositional approach
- 5 Control in intelligent buildings

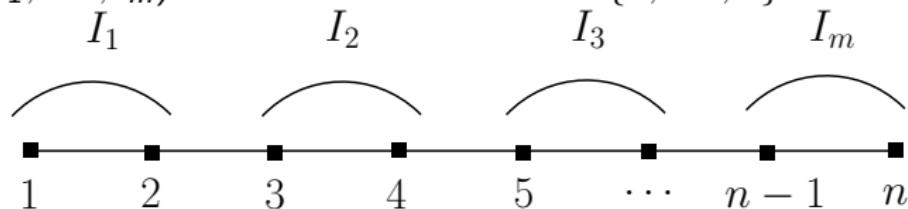
Compositional synthesis



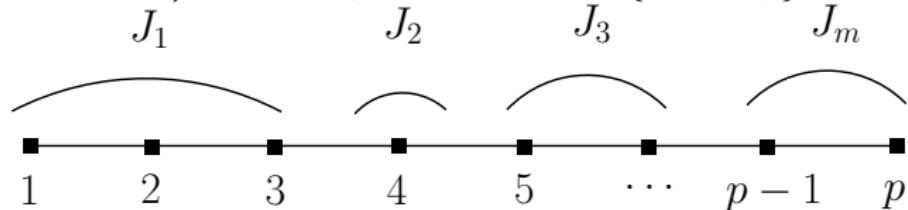
Decomposition

Decomposition into m subsystems:

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



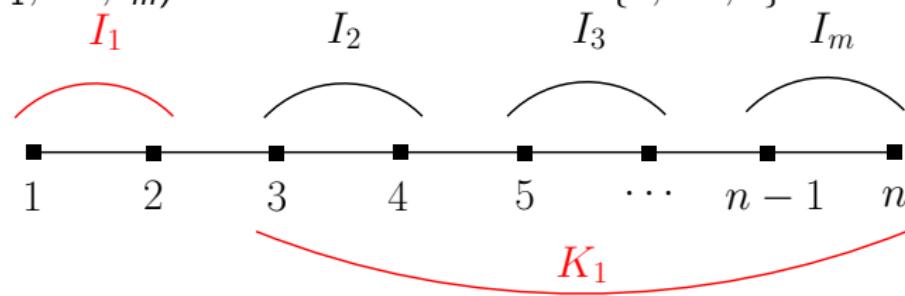
Partition (J_1, \dots, J_m) of the **input** dimensions $\{1, \dots, p\}$



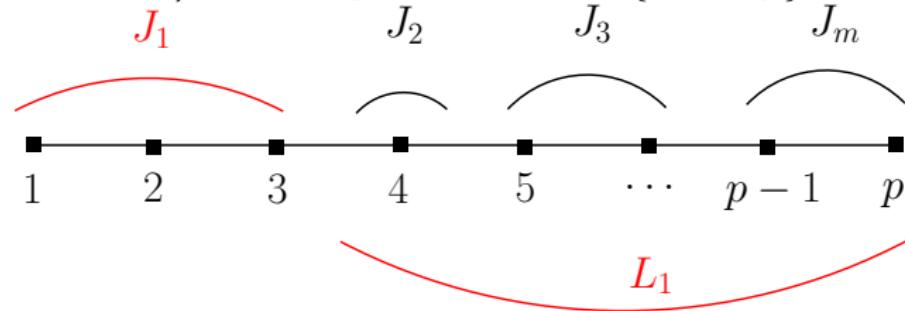
Decomposition

Decomposition into m subsystems:

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



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



Control the **states** x_{I_1} using the **inputs** u_{J_1} with **disturbances** x_{K_1} and u_{L_1}

Abstraction

Symbolic abstraction $S_i = (X_i, U_i, \xrightarrow{i})$ of subsystem $i \in \{1, \dots, m\}$:

Classical method, but with an **assume-guarantee obligation**:

A/G Obligation (K_i)

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

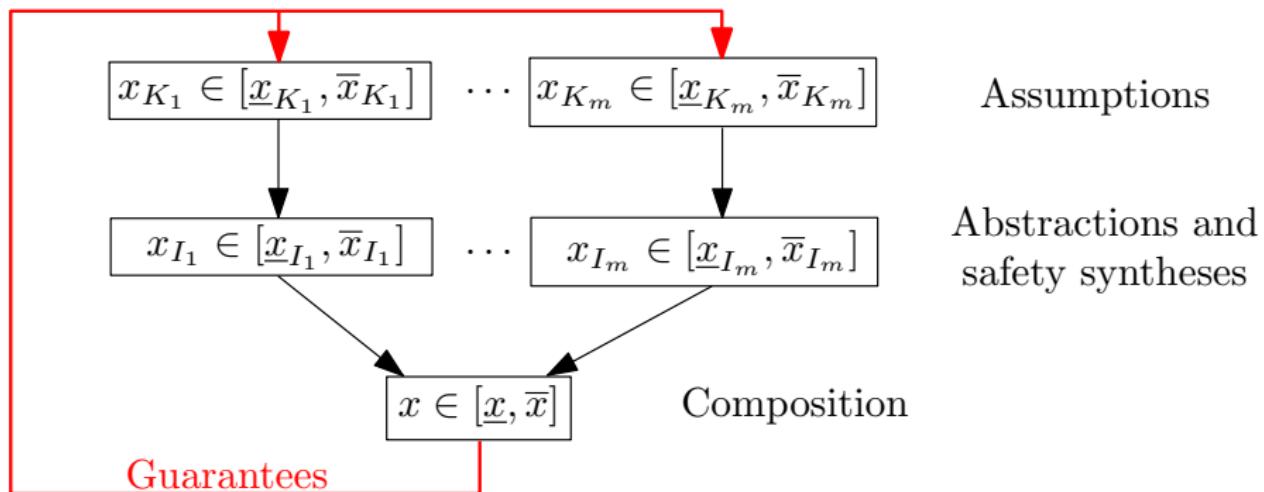
Abstraction

Symbolic abstraction $S_i = (X_i, U_i, \xrightarrow{i})$ of subsystem $i \in \{1, \dots, m\}$:

Classical method, but with an **assume-guarantee obligation**:

A/G Obligation (K_i)

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



Synthesis

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

- maximal safe set: $Z_i \subseteq X_i$
- safety controller: $C_i : Z_i \rightarrow 2^{U_i}$

Performances optimization:

- cost function $g_i(s_{I_i}, u_{J_i})$, with $g_a(s, u) \leq \sum_{i=1}^m g_i(s_{I_i}, u_{J_i})$
- deterministic controller: $C_i^* : Z_i \rightarrow U_i$

Safety

Composition of safe sets and safety controllers:

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

Theorem (Meyer, Girard, Witrant, ADHS 2015)

C_c is a safety controller for S in Z_c .

Safety

Composition of safe sets and safety controllers:

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

Theorem (Meyer, Girard, Witrant, ADHS 2015)

C_c is a safety controller for S in Z_c .

Proposition (Safety comparison)

$Z_c \subseteq Z_a$.

Performance guarantees

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

Theorem (Meyer, Girard, Witrant, ADHS 2015)

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

Let s^0, s^1, \dots such that $x^k \in s^k$, for all $k \in \mathbb{N}$. Then,

$$\sum_{k=0}^{+\infty} \lambda^k g(x^k, u^k) \leq \sum_{i=1}^m J_i^0(s_{l_i}^0) + \frac{\lambda^{N+1}}{1-\lambda} \sum_{i=1}^m M_i$$

Performance guarantees

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

Theorem (Meyer, Girard, Witrant, ADHS 2015)

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

Let s^0, s^1, \dots such that $x^k \in s^k$, for all $k \in \mathbb{N}$. Then,

$$\sum_{k=0}^{+\infty} \lambda^k g(x^k, u^k) \leq \sum_{i=1}^m J_i^0(s_{l_i}^0) + \frac{\lambda^{N+1}}{1-\lambda} \sum_{i=1}^m M_i$$

Proposition (Guarantees comparison)

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

Complexity

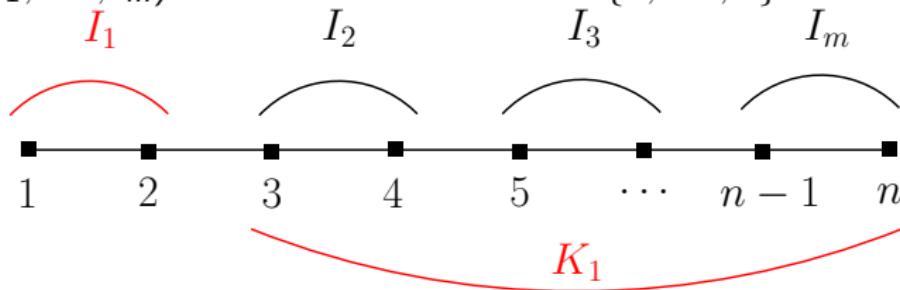
- n : state space dimension
- p : control space dimension
- $\alpha_x \in \mathbb{N}$: number of symbols **per dimension** in the state partition
- $\alpha_u \in \mathbb{N}$: number of controls **per dimension** in the input discretization
- $|\cdot|$: cardinality of a set

	Method	
	Centralized	Compositional
Complexity	$\alpha_x^n \alpha_u^p$	$\sum_{i=1}^m \alpha_x^{ I_i } \alpha_u^{ J_i }$

Generalization

Decomposition into m subsystems:

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



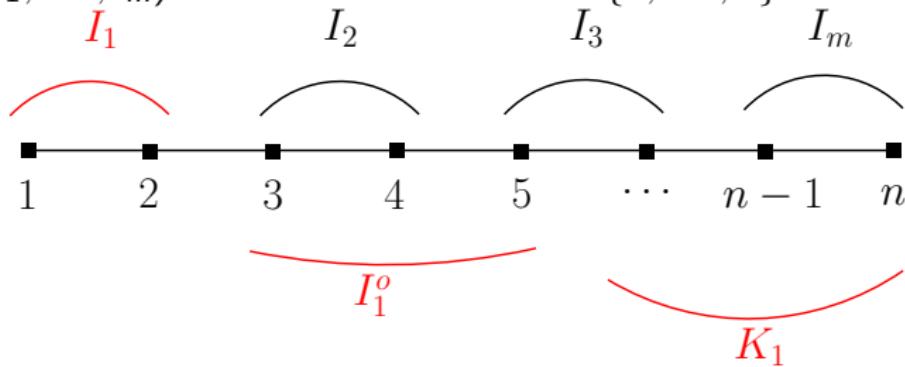
Subsystem $i \in \{1, \dots, m\}$:

- I_i : controlled states
- K_i : unobserved states (disturbances)

Generalization

Decomposition into m subsystems:

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



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

- I_i : controlled states
- I_i^o : observed but uncontrolled states
- K_i : unobserved states (disturbances)

Consequences

- Symbolic abstraction: needs two **assume-guarantee obligations**

A/G Obligation (K_i)

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

A/G Obligation (I_i^o)

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

Consequences

- Symbolic abstraction: needs two **assume-guarantee obligations**

A/G Obligation (K_i)

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

A/G Obligation (I_i^o)

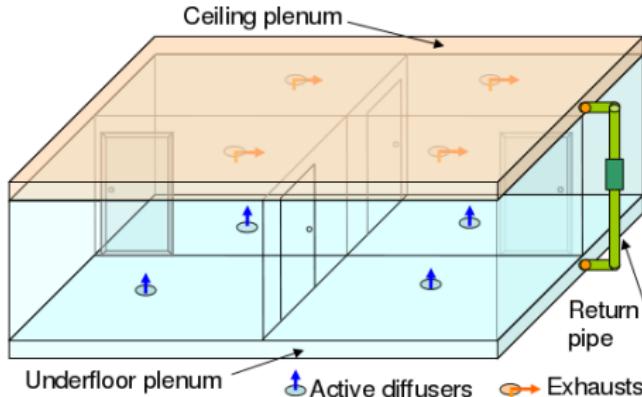
Observed but uncontrolled states: $x_{I_i^o} \in [\underline{x}_{I_i^o}, \bar{x}_{I_i^o}]$

- State composition with **overlaps**: $Z_c = Z_1 \cap \cdots \cap Z_m$
- Complexity depends on all **modeled states**: $\alpha_x^{|I_i \cup I_i^o|}$

Outline

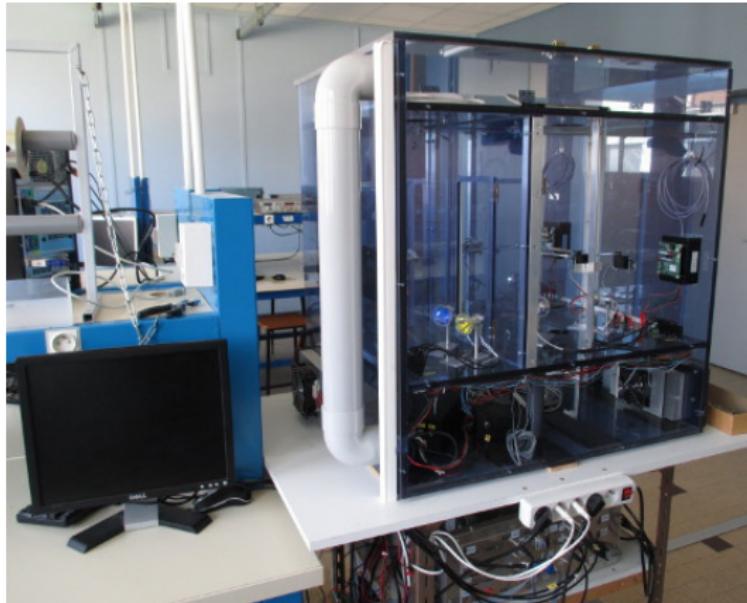
- 1 Monotone control system
- 2 Robust controlled invariance
- 3 Symbolic control
- 4 Compositional approach
- 5 Control in intelligent buildings

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

Experimental building



- $\approx 1m^3$
- 4 rooms with 4 doors
- 3 Peltier coolers
- Heat sources: lamps
- CompactRIO
- LabVIEW

Temperature model

Assume a **uniform temperature** in each room

Combine **energy and mass conservation** equations

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 model:

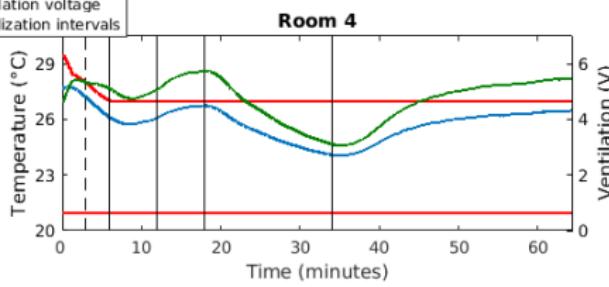
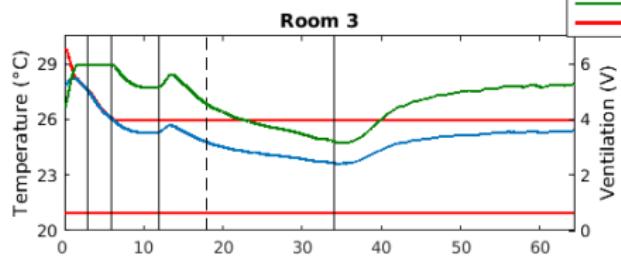
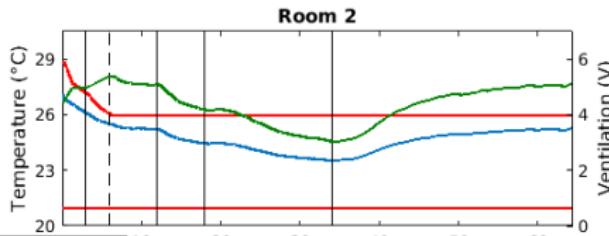
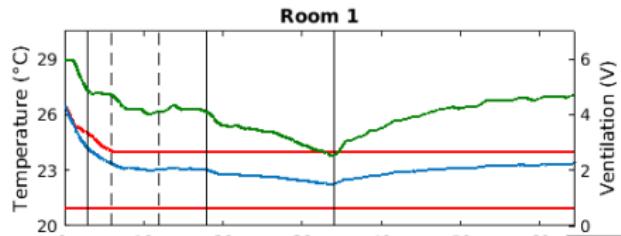
- is **monotone**
- has been **validated** by experimental data¹
- its parameters are **identified** to match the experiment¹

¹ Meyer, Nazarpour, Girard, Witrant, BuildSys 2013 & ECC 2014

Robust controlled invariance

Invariance controller: $u_i(T) = \overline{u_i} \frac{T_i - \overline{T}_i}{\underline{T}_i - \underline{T}_i}$

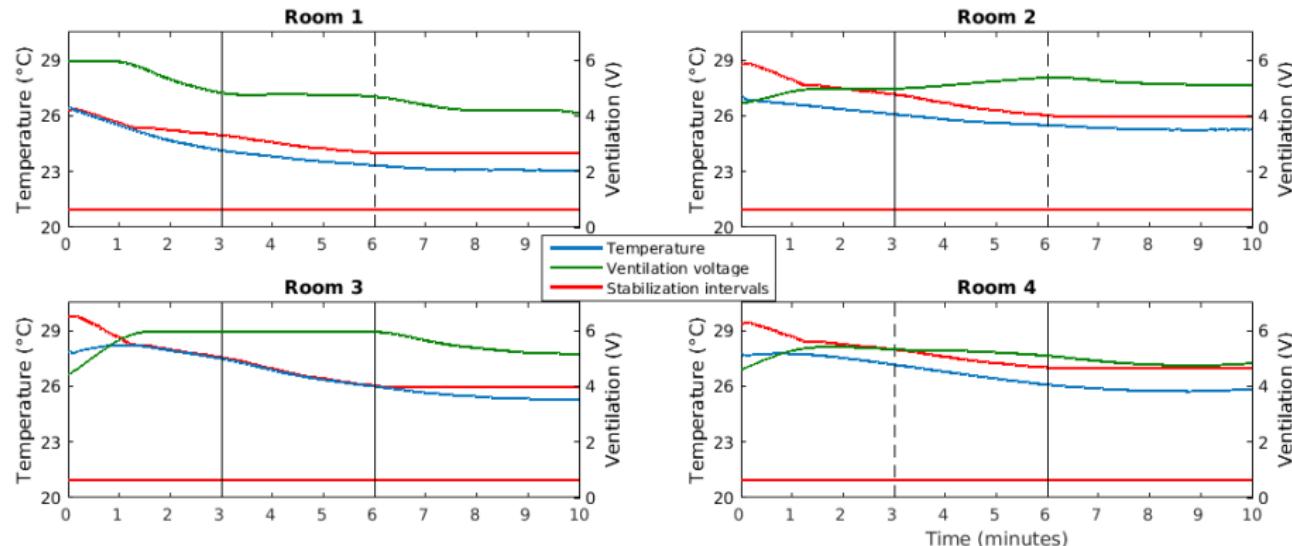
- $T_i = \overline{T}_i$: max ventilation
- $T_i = \underline{T}_i$: no ventilation



Robust set stabilization

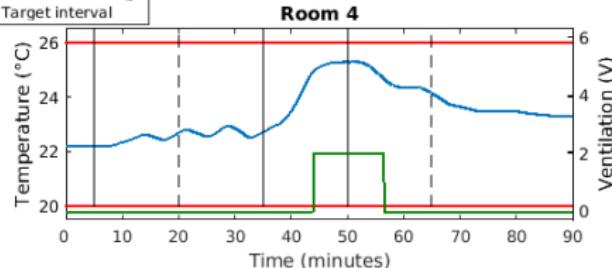
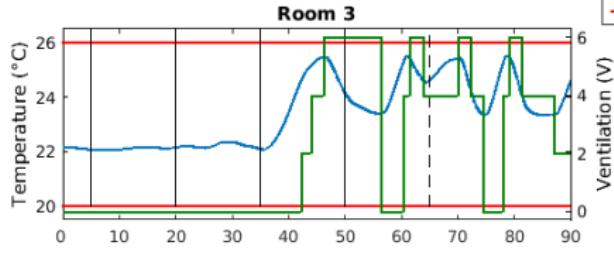
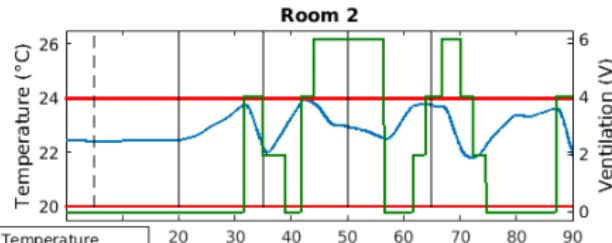
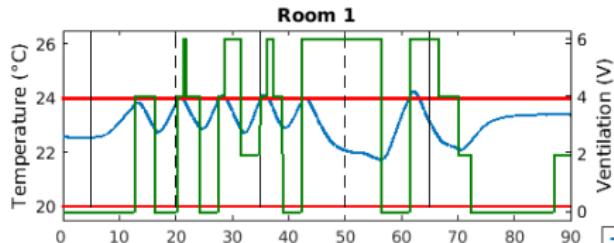
Stabilizing controller: $u_i(T) = \bar{u}_i \frac{T_i - \underline{X}_i(\underline{\lambda}(T))}{\bar{X}_i(\bar{\lambda}(T)) - \underline{X}_i(\underline{\lambda}(T))}$

$[\underline{X}_i(\underline{\lambda}(T)), \bar{X}_i(\bar{\lambda}(T))]$: smallest RCI interval containing T



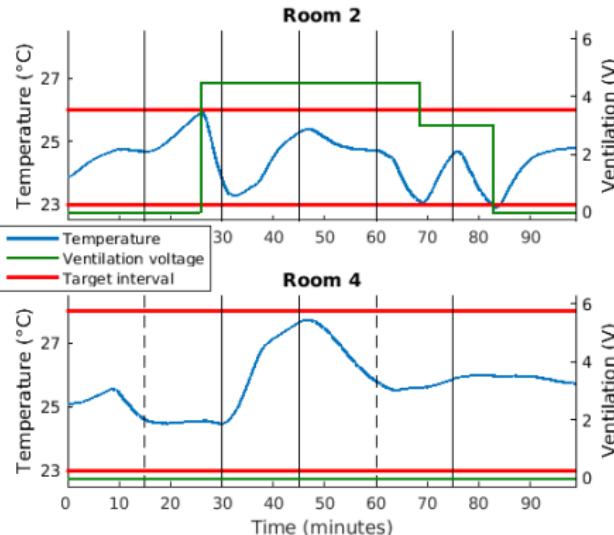
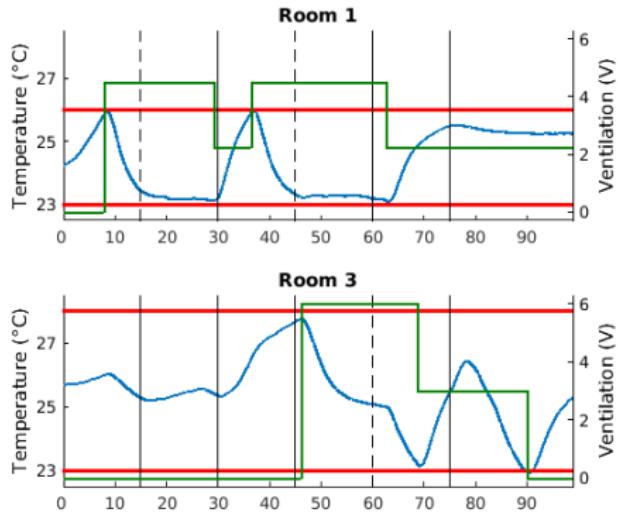
Centralized symbolic control

- $\alpha_x = 10, \alpha_u = 4, \tau = 34 \text{ s}$
- $g_a(s^k, u^k, u^{k-1}) = \|u^k\| + \|u^k - u^{k-1}\| + \|s_*^k - T_*\|$
- $N = 5, \lambda = 0.5: \lambda^{N+1} \approx 1.6\%$
- **Computation time: more than 2 days**



Compositional symbolic control

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

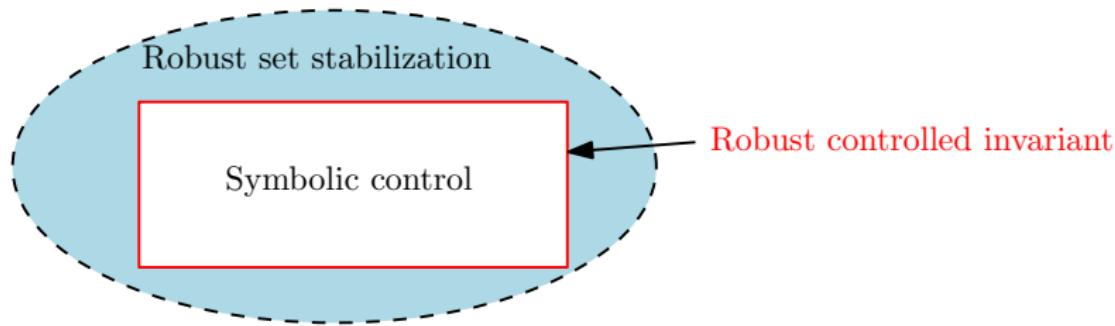


Conclusion

- Developed two approaches for the **robust control of monotone systems** with **safety specification**, based on:
 - Invariance, with an extension to **set stabilization**
 - Symbolic abstraction: both **centralized** and **compositional**
- Applied to the **temperature control** in a small-scale **experimental building**

Conclusion

- Developed two approaches for the **robust control of monotone systems** with **safety specification**, based on:
 - Invariance, with an extension to **set stabilization**
 - Symbolic abstraction: both **centralized** and **compositional**
- Applied to the **temperature control** in a small-scale **experimental building**



Perspectives

- Monotonicity appears in various other fields:
 - biology, chemistry, economy, population dynamics, ...
- Extension of the symbolic compositional approach
 - to non-monotone systems
 - to other specifications than safety
- 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

⇒ increased precision and robustness, local monotonicity

Publications

Authors: P.-J. Meyer, H. Nazarpour, A. Girard and E. Witrant

Journal paper:

- Robust controlled invariance for monotone systems: application to ventilation regulation in buildings. **Automatica**, provisionally accepted.

International conference:

- Safety control with performance guarantees of cooperative systems using compositional abstractions. **ADHS**, 2015.
- Experimental Implementation of UFAD Regulation based on Robust Controlled Invariance. **ECC**, 2014.
- Controllability and invariance of monotone systems for robust ventilation automation in buildings. **CDC**, 2013.

Conference poster:

- Symbolic Control of Monotone Systems, Application to Ventilation Regulation in Buildings. **HSCC**, 2015.
- Robust Controlled Invariance for UFAD Regulation. **BuildSys**, 2013.

Stabilization functions

Robust set stabilization from $[\underline{x}_0, \overline{x}_0]$ to $[\underline{x}_f, \overline{x}_f]$

Linear function	$\overline{X}(\lambda) = \lambda \overline{x_0} + (1 - \lambda) \overline{x_f}$
Static input-state characteristic $k_x : \mathcal{U} \times \mathcal{W} \rightarrow \mathcal{X}$	$\exists \overline{u_f}, \overline{u_0} \mid \underline{u} < \overline{u_f} < \overline{u_0}, \quad \overline{x_0} = k_x(\overline{u_0}, \overline{w}), \quad \overline{x_f} = k_x(\overline{u_f}, \overline{w})$
Family of equilibria	$\overline{U}(\lambda) = \lambda \overline{u_0} + (1 - \lambda) \overline{u_f}$ $\overline{X}(\lambda) = k_x(\overline{U}(\lambda), \overline{w})$
Trajectory $\overline{x_f} \rightarrow \overline{x_0}$	$\overline{X}(\lambda) = \Phi\left(\frac{\lambda}{1 - \lambda}, \overline{x_f}, \overline{u_0}, \overline{w}\right)$
Trajectory $\overline{x_0} \rightarrow \overline{x_f}$	$\overline{X}(\lambda) = \Phi\left(\frac{1 - \lambda}{\lambda}, \overline{x_0}, \overline{u_f}, \overline{w}\right)$

Symbolic abstraction

State partition \mathcal{P} of $[\underline{x}, \bar{x}] \subseteq \mathbb{R}^n$ into α_x identical intervals per dimension

$$\mathcal{P} = \left\{ \left[\underline{s}, \underline{s} + \frac{\bar{x} - \underline{x}}{\alpha_x} \right] \mid \underline{s} \in \left(\underline{x} + \frac{\bar{x} - \underline{x}}{\alpha_x} * \mathbb{Z}^n \right) \cap [\underline{x}, \bar{x}] \right\}$$

Input discretization U_a of $[\underline{u}, \bar{u}] \subseteq \mathbb{R}^p$ into $\alpha_u \geq 2$ values per dimension

$$U_a = \left(\underline{u} + \frac{\bar{u} - \underline{u}}{\alpha_u - 1} * \mathbb{Z}^p \right) \cap [\underline{u}, \bar{u}]$$

Sampling period

Guidelines for the **viability kernel**² (maximal invariant set):

$$2L\tau^2 \sup_{x \in [\underline{x}, \bar{x}]} \|f(x, \bar{u}, \bar{w})\| \geq \frac{\|\bar{x} - \underline{x}\|}{\alpha_x}$$

- τ : sampling period
- $\frac{\|\bar{x} - \underline{x}\|}{\alpha_x}$: step of the state partition
- L : Lipschitz constant
- $\sup_{x \in [\underline{x}, \bar{x}]} \|f(x, \bar{u}, \bar{w})\|$: supremum of the vector field

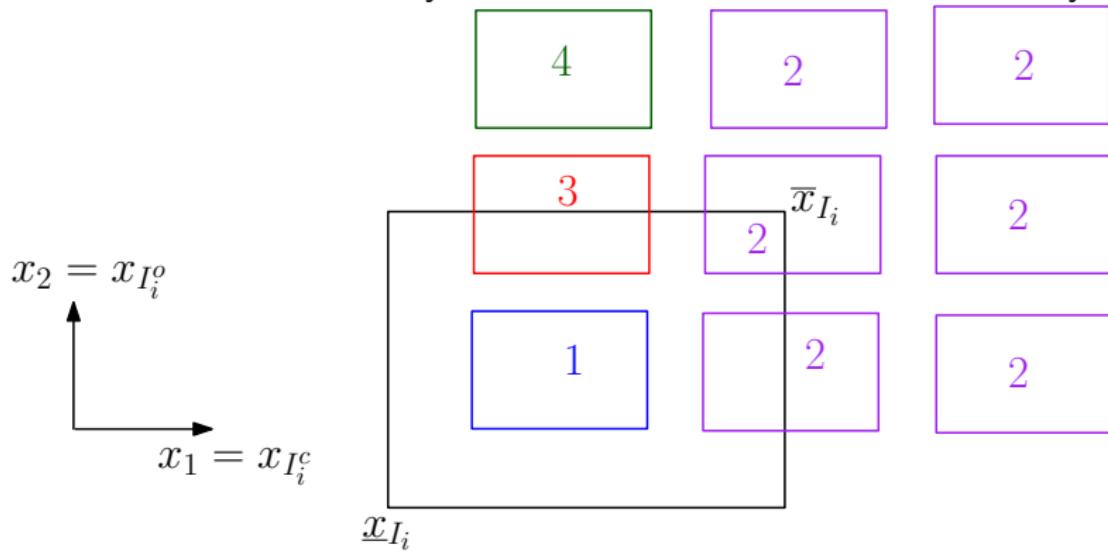
²P. Saint-Pierre. Approximation of the viability kernel. *Applied Mathematics and Optimization*, 29(2):187–209, 1994.

2^{nd} A/G obligation

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



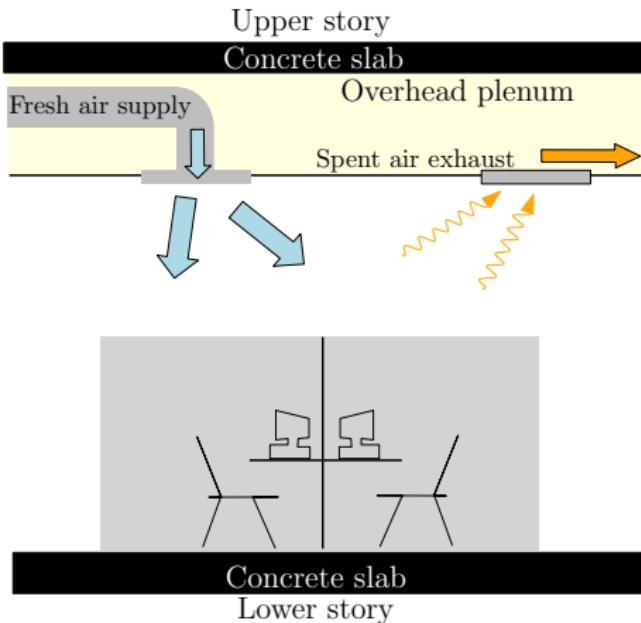
Complexity

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

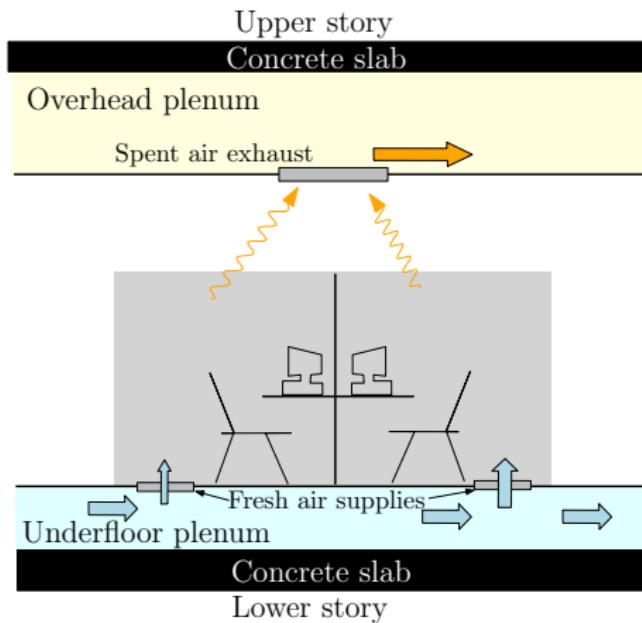
	Method	
	Centralized	Compositional
Abstraction (successors computed)	$2\alpha_x^n \alpha_u^p$	$\sum_{i=1}^m 2\alpha_x^{ I_i } \alpha_u^{ J_i }$
Dynamic programming (max iterations)	$N\alpha_x^{2n} \alpha_u^p$	$\sum_{i=1}^m N\alpha_x^{2 I_i } \alpha_u^{ J_i }$

UnderFloor Air Distribution

Overhead ventilation



UnderFloor Air Distribution



UFAD model

- \mathcal{N}_i : indices of neighbor rooms of room i
- \mathcal{N}_i^* : \mathcal{N}_i , underfloor, ceiling, outside
- $\delta_{s_i}, \delta_{d_{ij}}$: discrete state of heat source and doors
- $\dot{m}_{u \rightarrow i}$: mass flow rate forced by the underfloor fan (control input)

$$\rho V_i C_v \frac{dT_i}{dt} = \sum_{j \in \mathcal{N}_i^*} \frac{k_{ij} A_{ij}}{\Delta_{ij}} (T_j - T_i) \quad (\text{Conduction})$$

$$+ \delta_{s_i} \varepsilon_{s_i} \sigma A_{s_i} (T_{s_i}^4 - T_i^4) \quad (\text{Radiation})$$

$$+ C_p \dot{m}_{u \rightarrow i} (T_u - T_i) \quad (\text{Ventilation})$$

$$+ \sum_{j \in \mathcal{N}_i} \delta_{d_{ij}} C_p \rho A_d \sqrt{2R} \max(0, T_j - T_i)^{3/2} \quad (\text{Open doors})$$