

# Controller synthesis for linear systems and safe linear-time temporal logic

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# Motivation

- Linear discrete-time system:

$$\Sigma : x^+ = Ax + Bu$$

with state space  $\mathbb{R}^n$  and input space  $\mathbb{R}^m$ .

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- Safe linear temporal logic specification (LTL)  $\varphi$  over atomic propositions:

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with each  $p_i \in \mathcal{P}$  a polytope in  $\mathbb{R}^n$ .

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with each  $p_i \in \mathcal{P}$  a polytope in  $\mathbb{R}^n$ .

- Expressive specifications:

- sequencing of actions, “if then else” requirements, fault recovery, ...
- guaranteed to only define safety properties;
- negation is only allowed on atomic propositions, *until* is replaced with *wait*.

# Cruise Control Example

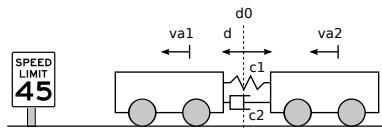
## Compliance with Speed Limits

### ■ Dynamics:

$$\dot{x} = \begin{bmatrix} 0 & -1 & 1 \\ \frac{k_s}{m} & -\frac{k_d}{m} & \frac{k_d}{m} \\ 0 & 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u$$

with  $x = (d, v_1, v_2) \in \mathbb{R}^3$  and  $u \in \mathbb{R}$ .

- $d$  distance between the truck and the trailer
- $v_1$  velocity of the truck
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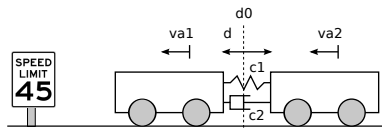
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### ■ Specification:

- compliance with speed limits  $v_a, v_b$  after at most  $T \in \mathbb{N}$  time-steps
- acceleration constraints  $u \in [\underline{u}, \bar{u}]$
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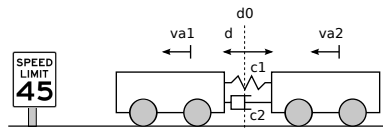
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### ■ Safe LTL formula

$$\Box(D \wedge U \wedge \varphi_a \wedge \varphi_b)$$

with  $\varphi_a$  and  $\varphi_b$  given by

$$m_a \implies \Diamond_{\leq T}(t_a \mathbf{W} m_b)$$

$$m_b \implies \Diamond_{\leq T}(t_b \mathbf{W} m_a)$$

- $m_i$ :  $v_i$  is active  $i \in \{a, b\}$
- $t_i$ :  $v_1 \leq v_i$

# Reducing Controller Synthesis to a Safety Game

[Kupferman and Vardi, 2001]

- Construct the bad-prefix automaton  $A_{\neg\varphi}$  from the safe LTL formula  $\varphi$ :

$$A_{\neg\varphi} = (Q, F, \delta, g, 2^{\mathcal{P}});$$



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- Given the **safe set**  $K = (Q \setminus F) \times H \subseteq (Q \setminus F) \times \mathbb{R}^n$  compute its largest controlled invariant subset:

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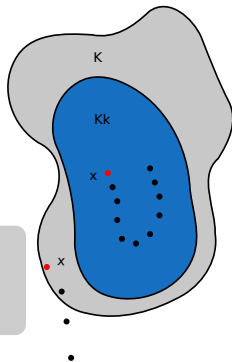
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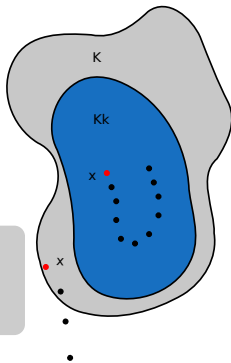
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- We know that:

$(q, x) \in \mathcal{K}(K) \Leftrightarrow$  existence of a control strategy enforcing  $\varphi$  from  $x$ .



# Computation of Controlled Invariant Subsets

## Basic Algorithm

- Fixed point computation [Bertsekas, 1972]:

$$K_{j+1} = \text{pre}(K_j) \cap K_j, \quad K_0 = K$$

with  $\text{pre}(K_j)$  being the set of states  $(q, x) \in Q \times \mathbb{R}^n$  for which there exists an input  $u \in \mathbb{R}^m$  forcing a transition to some state in  $K_j$ .

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- Safe set  $K = (Q \setminus F) \times H \subseteq (Q \setminus F) \times \mathbb{R}^n$  given by:

$$H = \bigcup_{i=1}^p H_i, \quad \text{each } H_i \text{ is a polytope}$$

$\Rightarrow$  each  $K_j$  is computable and the iteration is known to asymptotically converge:

$$\mathcal{K}(K) = \lim_{j \rightarrow \infty} K_j.$$

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  - [De Santis et al., 2004]: iteration is initialized with a controlled invariant set  $K_0 \subset K$ ;
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- For  $H$  given as union of polytopes, the iterative computation introduces combinatorial complexity!

In this work we approximate  $K$  by sets adapted to the dynamics.  
(Finite termination and symbolic implementation)

- Any *controllable* linear system can be transformed to the special Brunovsky normal form by an invertible linear change of coordinates and feedback:

$$A = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}, \quad b = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix};$$

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- We can *under-approximate* the safe set  $K = (Q \setminus F) \times H$  by  $\check{K} = (Q \setminus F) \times \check{H}$ :

$$\check{H} = \bigcup_{i=1}^p B_i \subseteq H$$

with each box  $B_i$  defined by  $B_i = [a_1^i, b_1^i] \times \dots \times [a_n^i, b_n^i]$ .

# Main Results

## Termination

### Theorem (Finite termination)

*Consider the composition  $A_{\neg\varphi} \parallel \Sigma$  where  $\Sigma$  is in special Brunovsky normal form and  $K = (Q \setminus F) \times H$  with  $H$  being a finite union of boxes. Then the largest controlled invariant subset of  $\check{K}$  can be computed in finitely many steps.*

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This result was first proved in [Tabuada and Pappas, 2003] and was used in [Tabuada and Pappas, 2006] to show, for the first time, that controllers can be synthesized to enforce LTL properties on control systems.

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How can we make use of this result when  $H$  is not a union of boxes?



# Main Results

## Completeness

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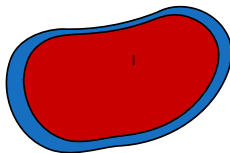
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We say that a set  $I \subseteq \mathbb{R}^n$  is strictly inside a set  $J \subseteq \mathbb{R}^n$  if there exists  $\gamma > 0$  for which:

$$I + \gamma \mathcal{B}_\gamma(0) \subseteq J.$$



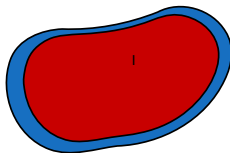
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### Theorem (Completeness)

*If there exists a controlled invariant set  $I \subseteq \mathcal{K}(K)$  for which  $I_q$  is strictly inside  $\mathcal{K}_q(K)$ , then there exists an under-approximation  $\check{K} = (Q \setminus F) \times \check{H}$  of  $K$ , with  $\check{H}$  being a finite union of boxes, such that  $I \subseteq \mathcal{K}(\check{K})$ .*

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## Symbolic Implementation

- Use binary decision diagrams (BDDs) to implement the iteration:

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$$\text{pre}(B_i) = \mathbb{R} \times [a_1^i, b_1^i] \times \dots \times [a_{n-1}^i, b_{n-1}^i];$$

- Symbolical computation of  $\text{pre}(K_j)$  can be done by shifting and variable reordering.

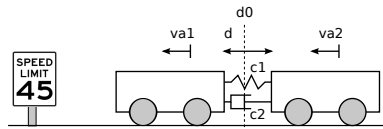
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Problem description:

- $\Sigma$  : 3 states, 1 input;
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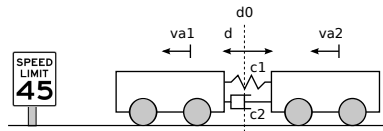
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- $N \in \{10, \dots, 13\}$  number of bits ( $2^N$  boxes) used in each dimension.



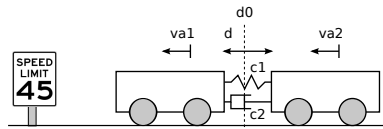
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Error bound:

$$\hat{\epsilon} = \frac{\text{vol } \mathcal{K}(\hat{K}) - \text{vol } \mathcal{K}(\check{K})}{\text{vol } \mathcal{K}(\check{K})} \geq \frac{\text{vol } \mathcal{K}(K) - \text{vol } \mathcal{K}(\check{K})}{\text{vol } \mathcal{K}(K)}$$

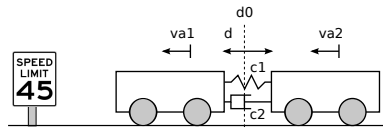
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$N \backslash T$	2		10	
	$t_r$	$\hat{\epsilon}$	$t_r$	$\hat{\epsilon}$
10	1m39s	2.31	2m40s	2.38
11	4m09s	1.01	4m31s	1.04
12	6m48s	0.58	7m52s	0.62
13	10m38s	0.43	16m01s	0.46

# Comparison with the Polyhedral Approach

- Example 5.1 in [Pérez et al., 2011]:

3 states + 2 inputs

- Workspace:

$$X = [0, 30]^3 \text{ and } U = [0, 2]^2$$

- Obstacles in the state space:

- $O_1 = [-5, 15]^3$
- $O_2 = [-5, 5]^3$
- $O_3 = [-15, 10]^3$

- Obstacles in the input space:

- $V_1 = [-3/2, 1/2]^2$
- $V_2 = [-1/4, 1/4]^2$
- $V_3 = [2/5, 1/5]^2$

- Specification with increasing complexity:

$$\varphi_0 = \Box(X \times U)$$

$$\varphi_1 = \Box((X \wedge \neg O_1) \times U)$$

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$$\varphi_3 = \Box((X \wedge \neg O_1) \times (U \wedge \neg V_1))$$

$$\varphi_4 = \Box((X \wedge_{i=1}^2 \neg O_i) \times (U \wedge_{i=1}^2 \neg V_i))$$

$$\varphi_5 = \Box((X \wedge_{i=1}^3 \neg O_i) \times (U \wedge_{i=1}^2 \neg V_i))$$

$$\varphi_6 = \Box((X \wedge_{i=1}^3 \neg O_i) \times (U \wedge_{i=1}^3 \neg V_i))$$

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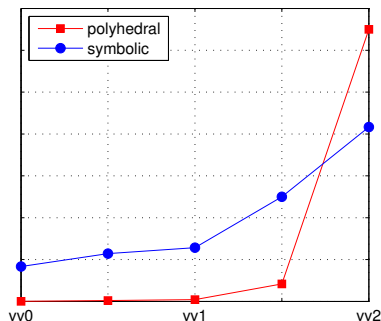
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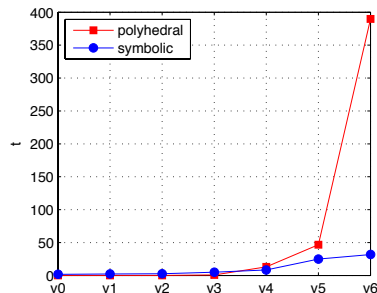
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# Summary and Remarks

Algorithm to synthesize controllers enforcing safe LTL specification on controllable linear systems:

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What is next?


- Boxes are not good enough  
(too many required to obtain reasonable approximations, CoD);
- Find more general polyhedra for which termination is guaranteed.





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