

Generating Longest Counterexample: At the Cross-roads of MILP and SMT

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American Control Conference (ACC), July 2020



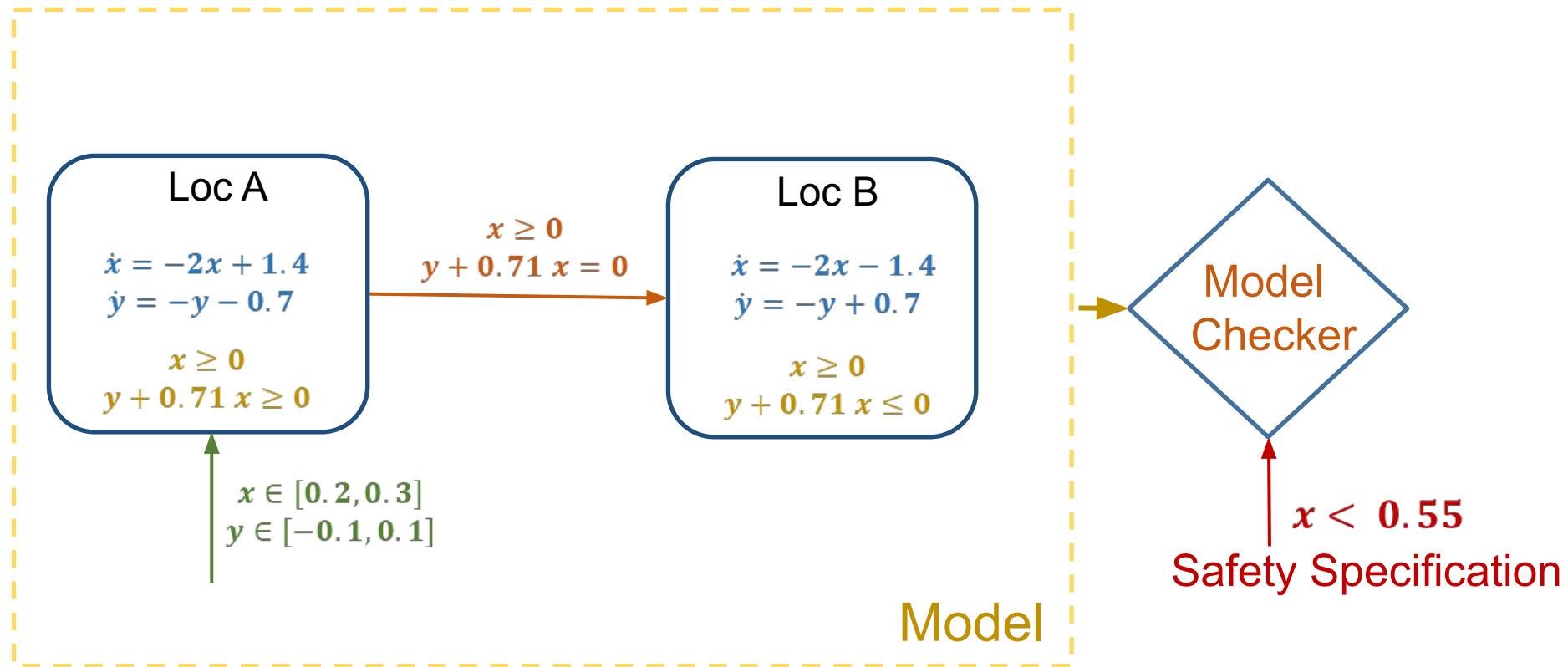
Outline

- [Background](#)
Verification, falsification and control
- [Introduction](#)
Longest counterexample
- [Preliminaries](#)
Simulation-equivalent analysis, reachable set computation
- [Methodology](#)
Constraint Propagation
- [Frameworks](#)
MILP-based and SMT-based



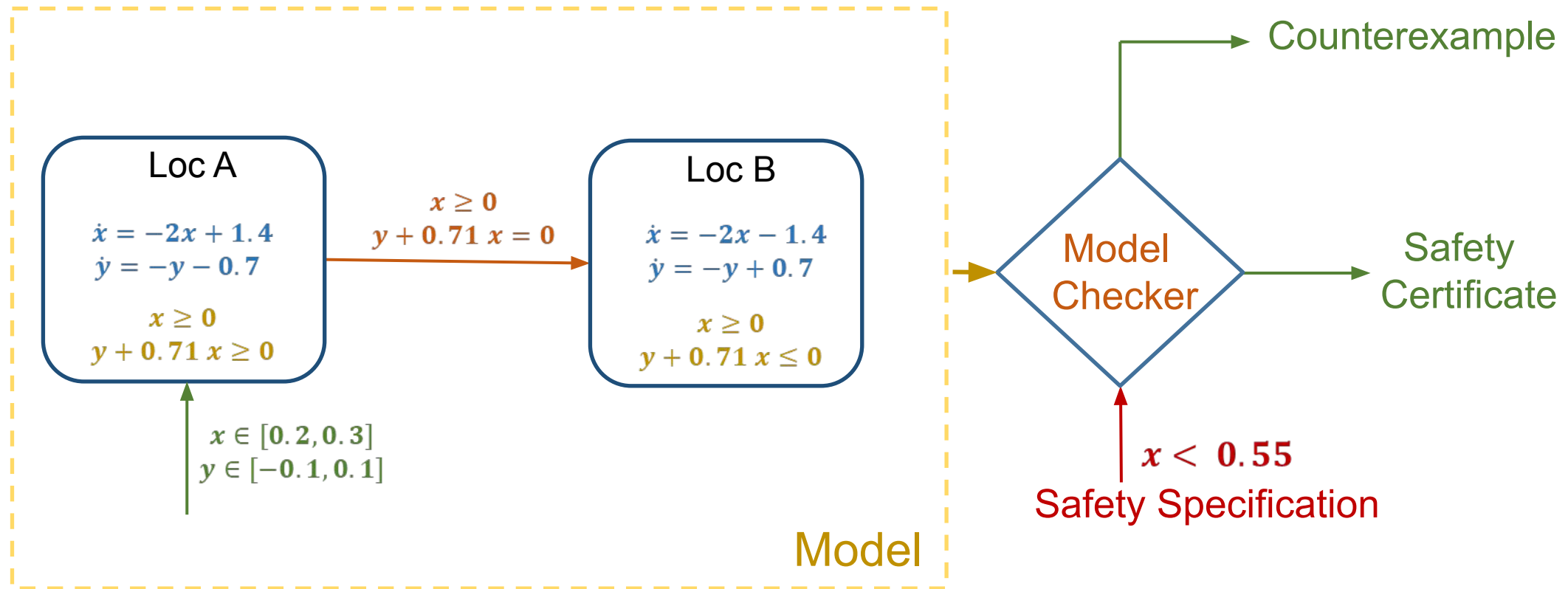
Verification

Analogous to Reachability



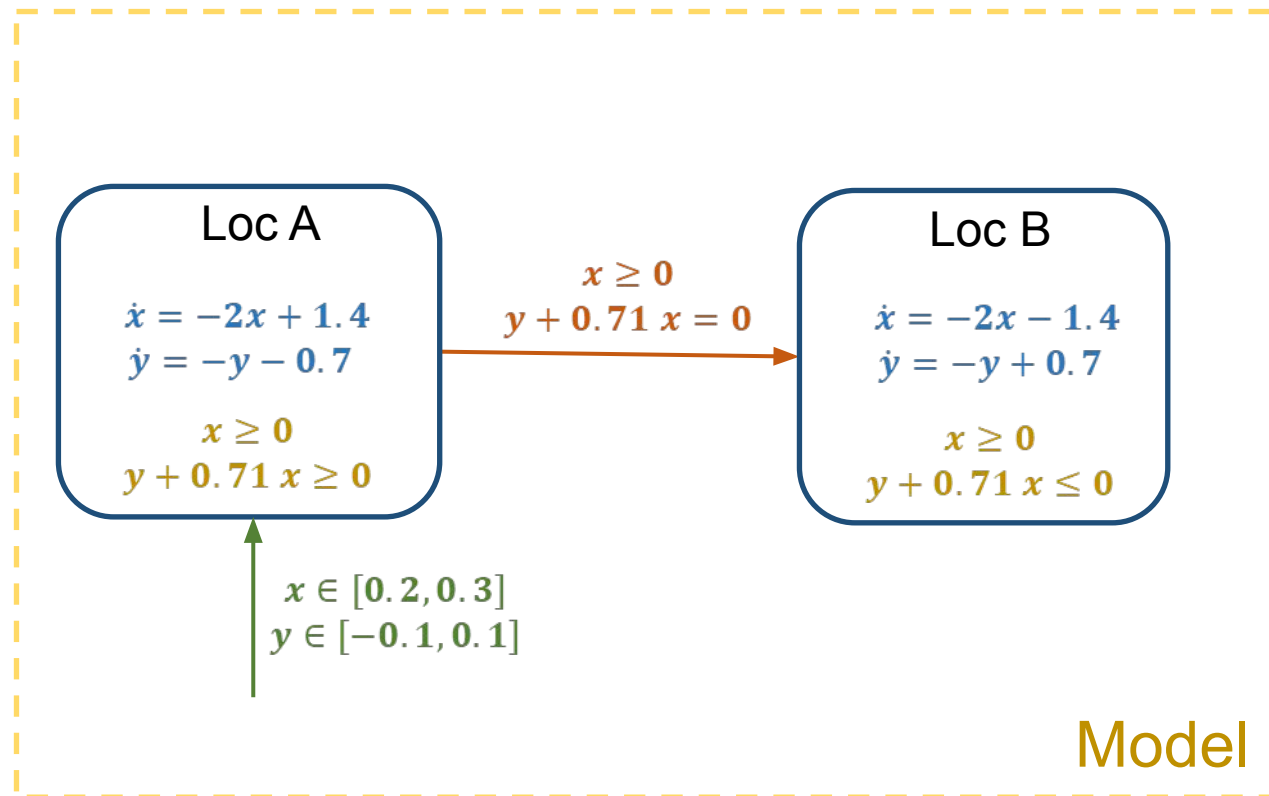
Verification

Analogous to Reachability



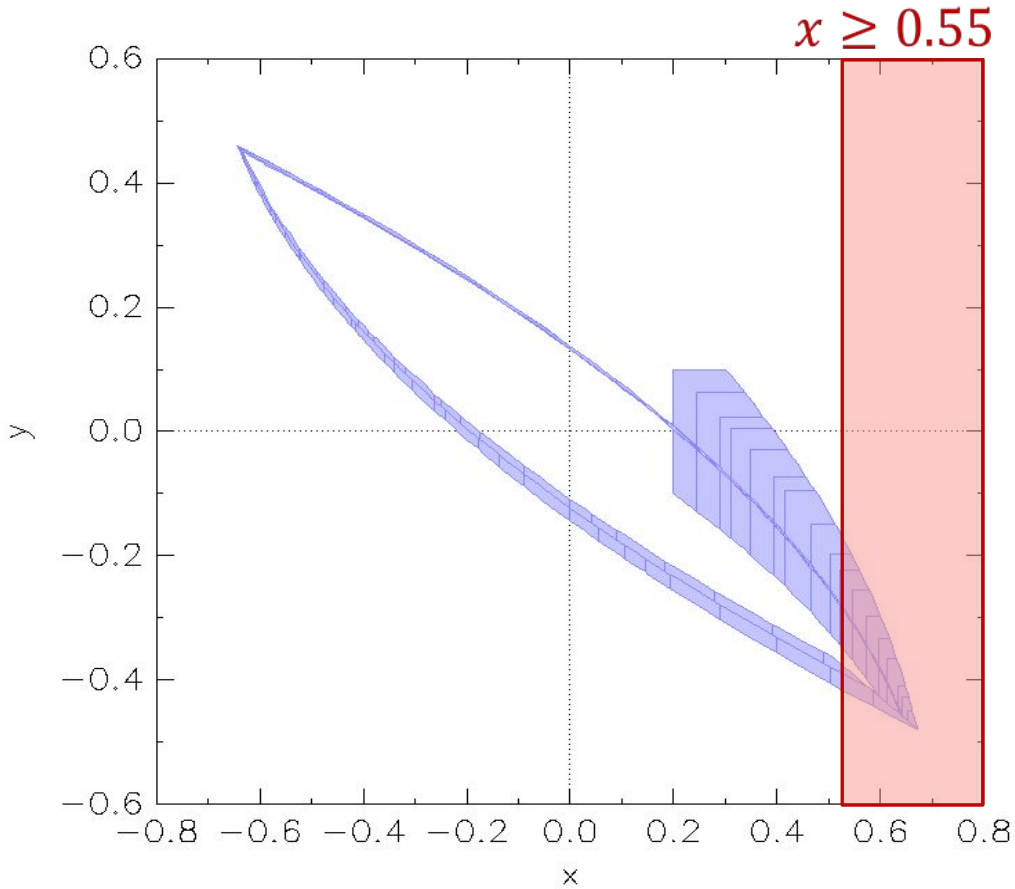
Verification

Hybrid Automaton

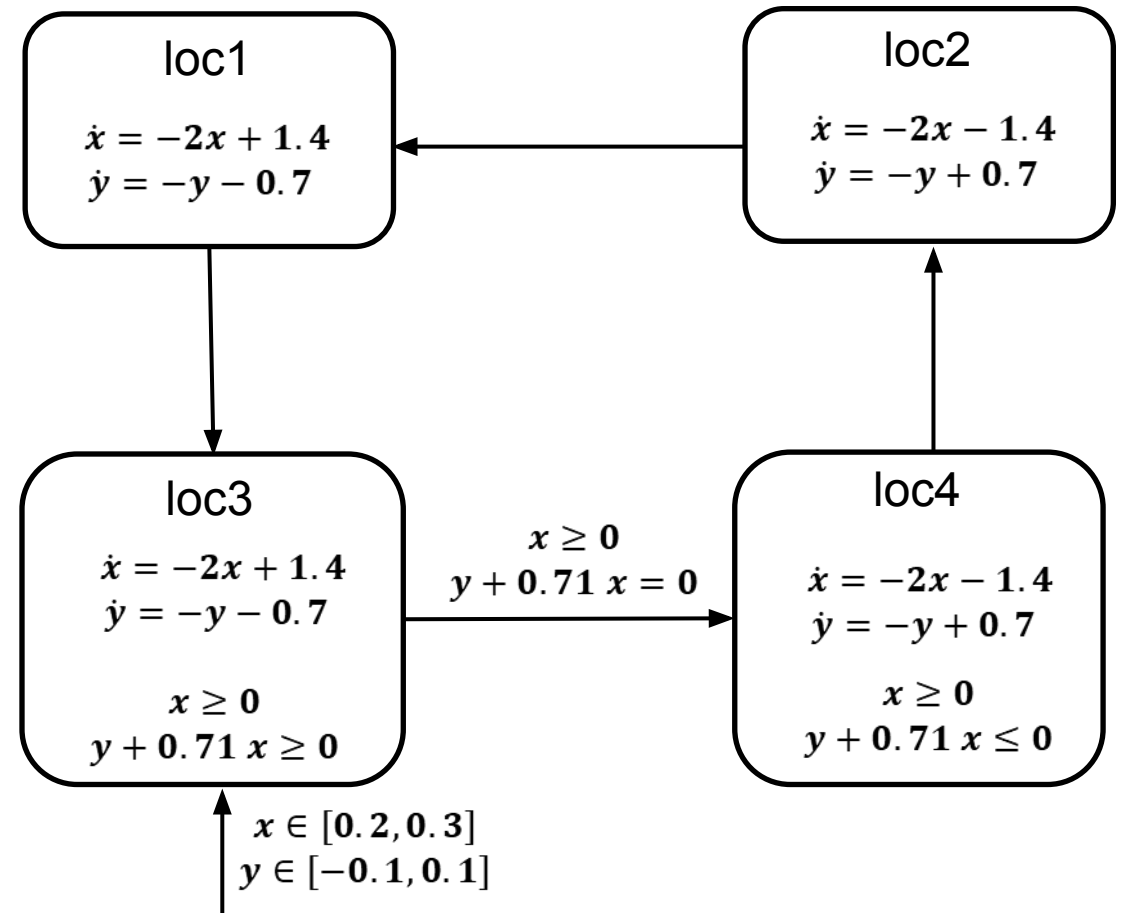


- Locations
- Flow
- Invariant
- Transitions
- Guard Condition
- Initial Condition

Verification



SpaceEx*: Filtered Oscillator**

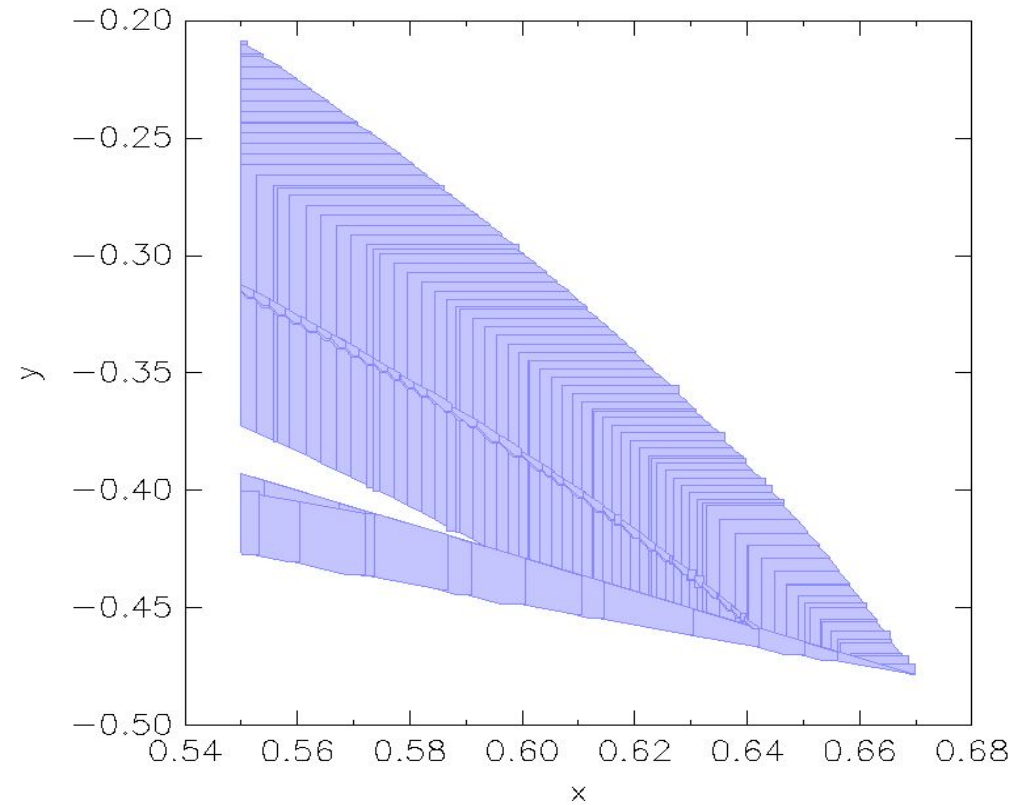
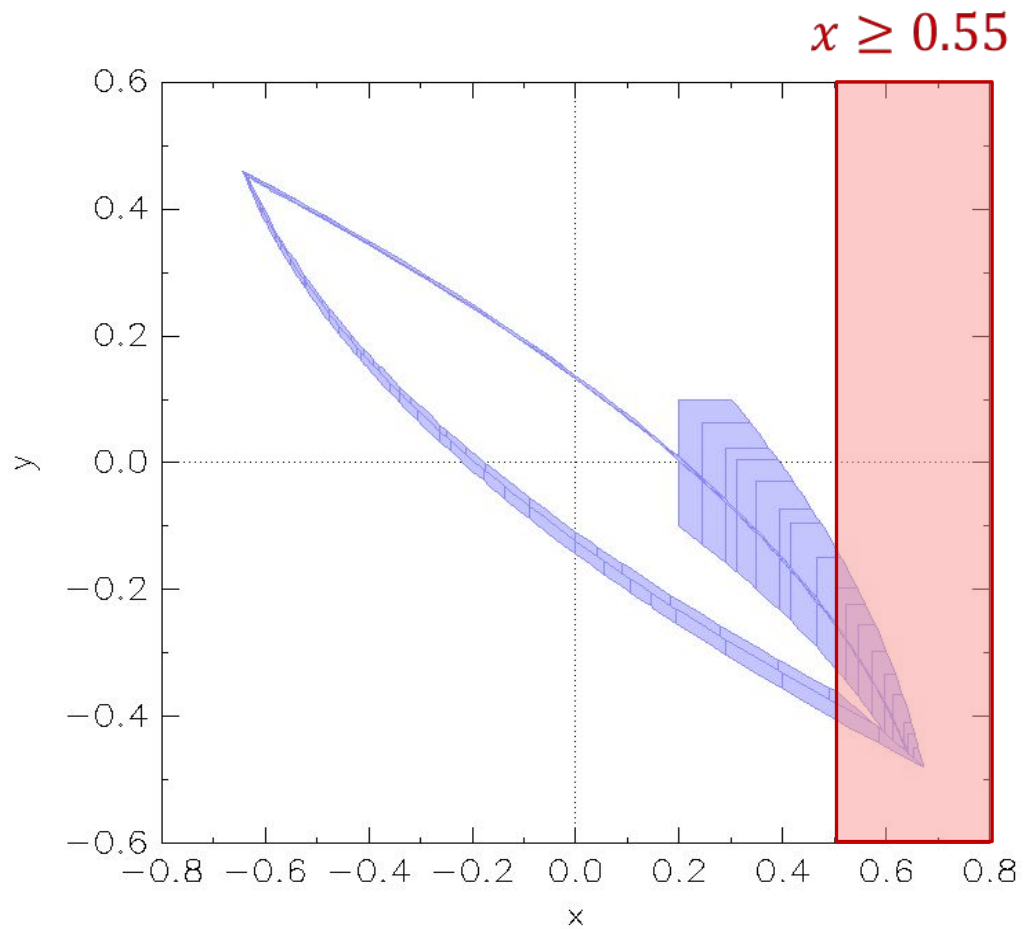


*G. Frehse et al, Spaceex: Scalable verification of hybrid systems. CAV 2011.

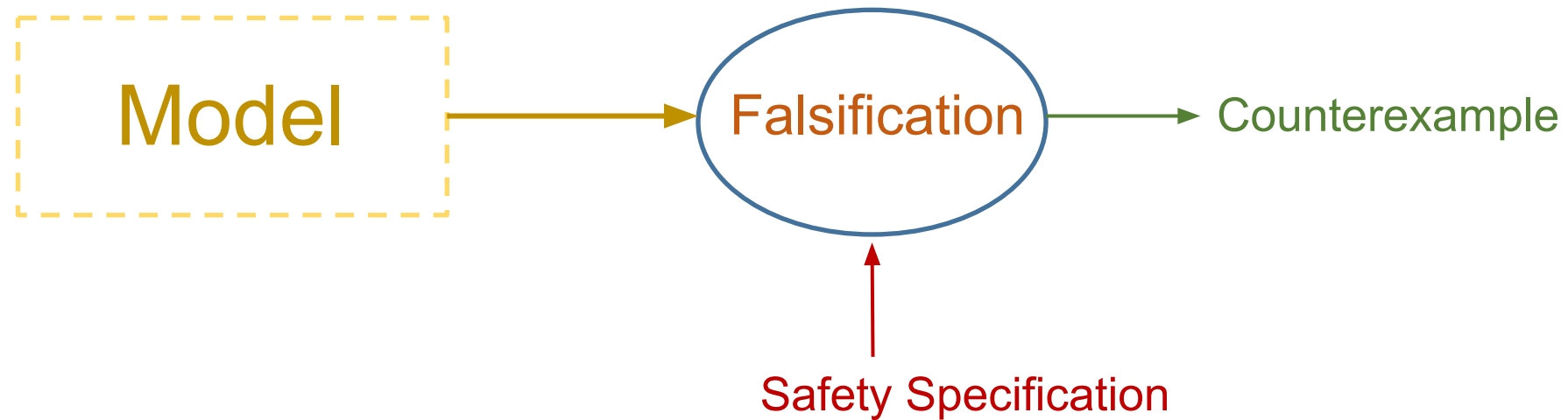
**<https://ths.rwth-aachen.de/research/projects/hypro/filtered-oscillator/>

Verification

SpaceEx: Filtered Oscillator

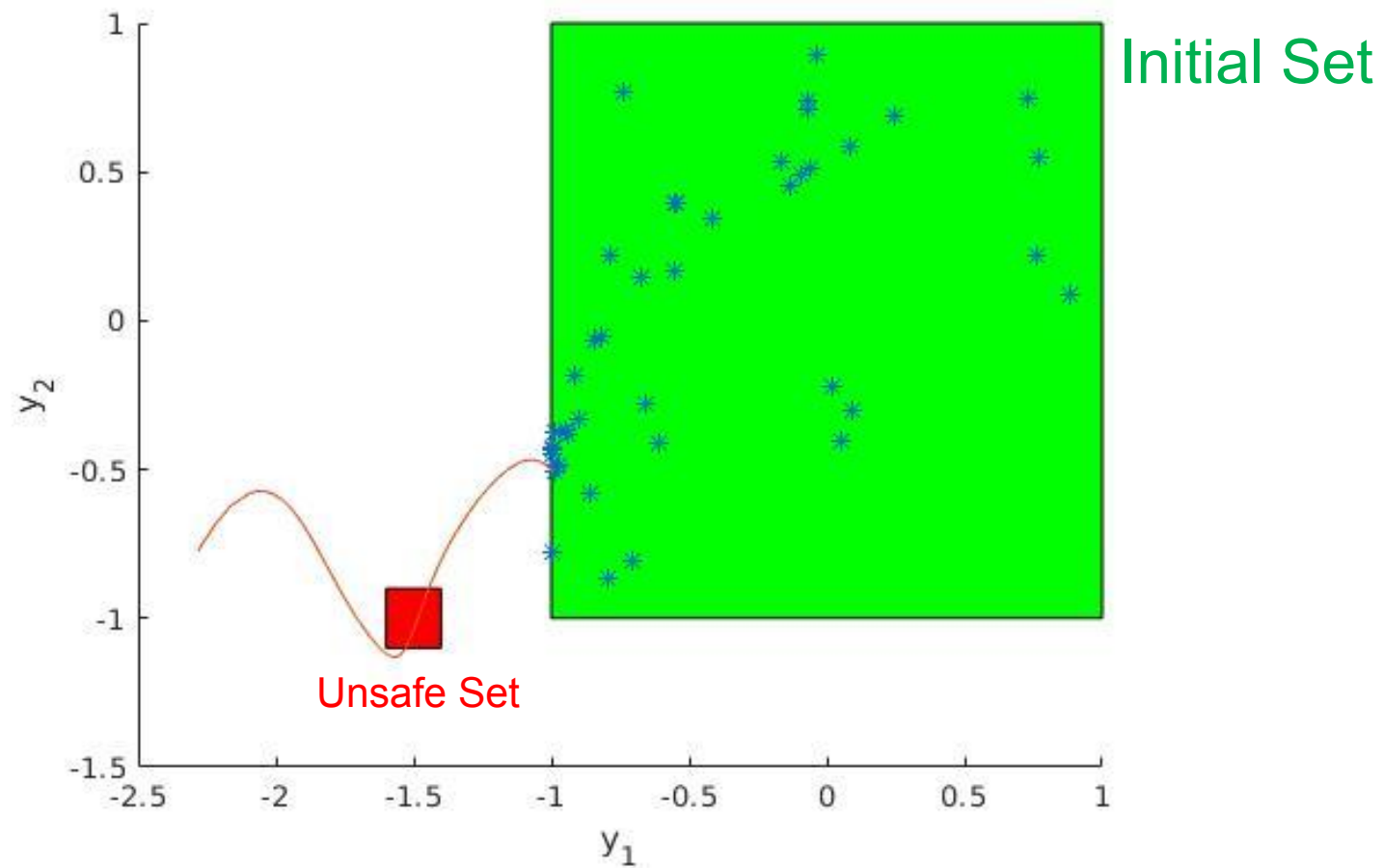


Falsification



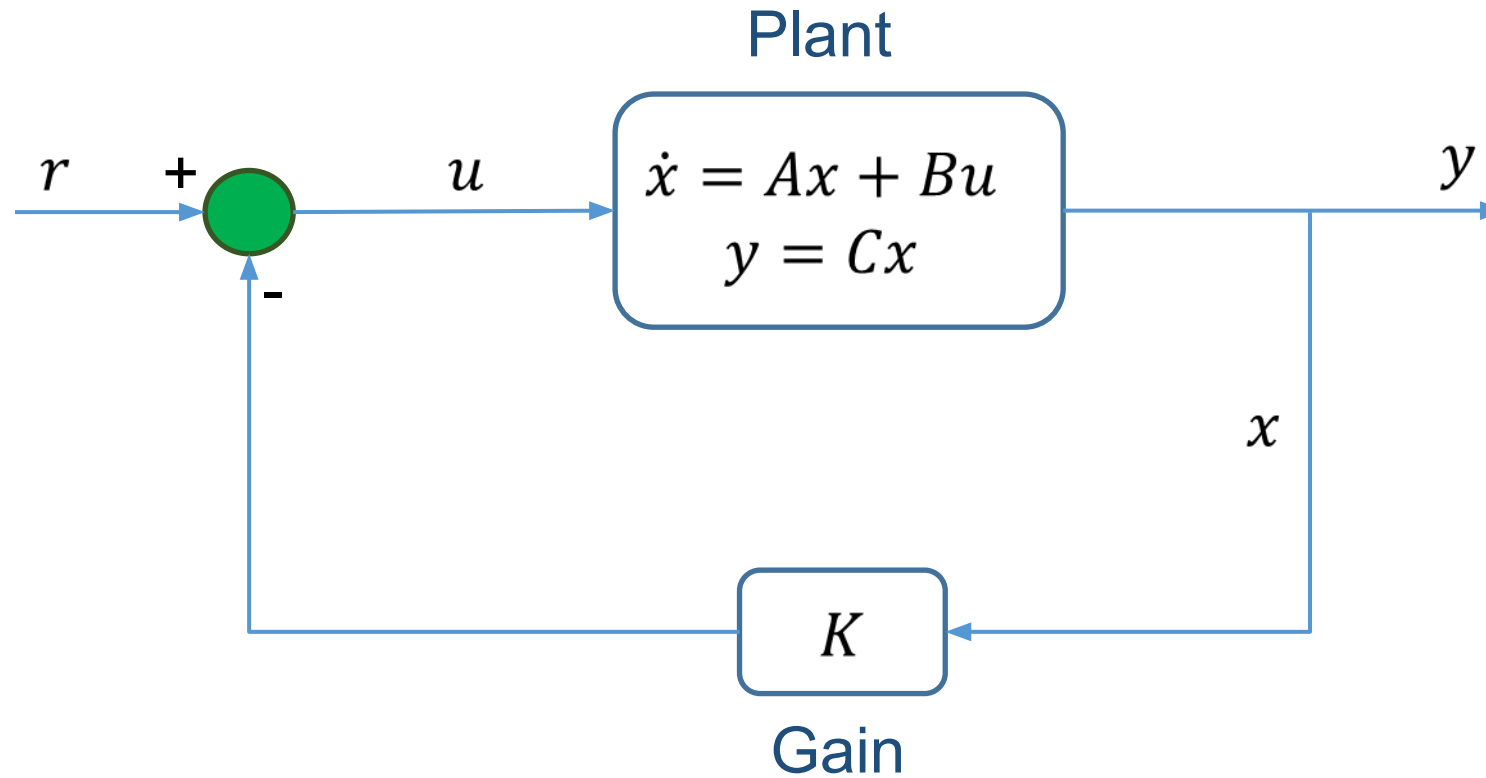
Falsification

S-TaLiRo*

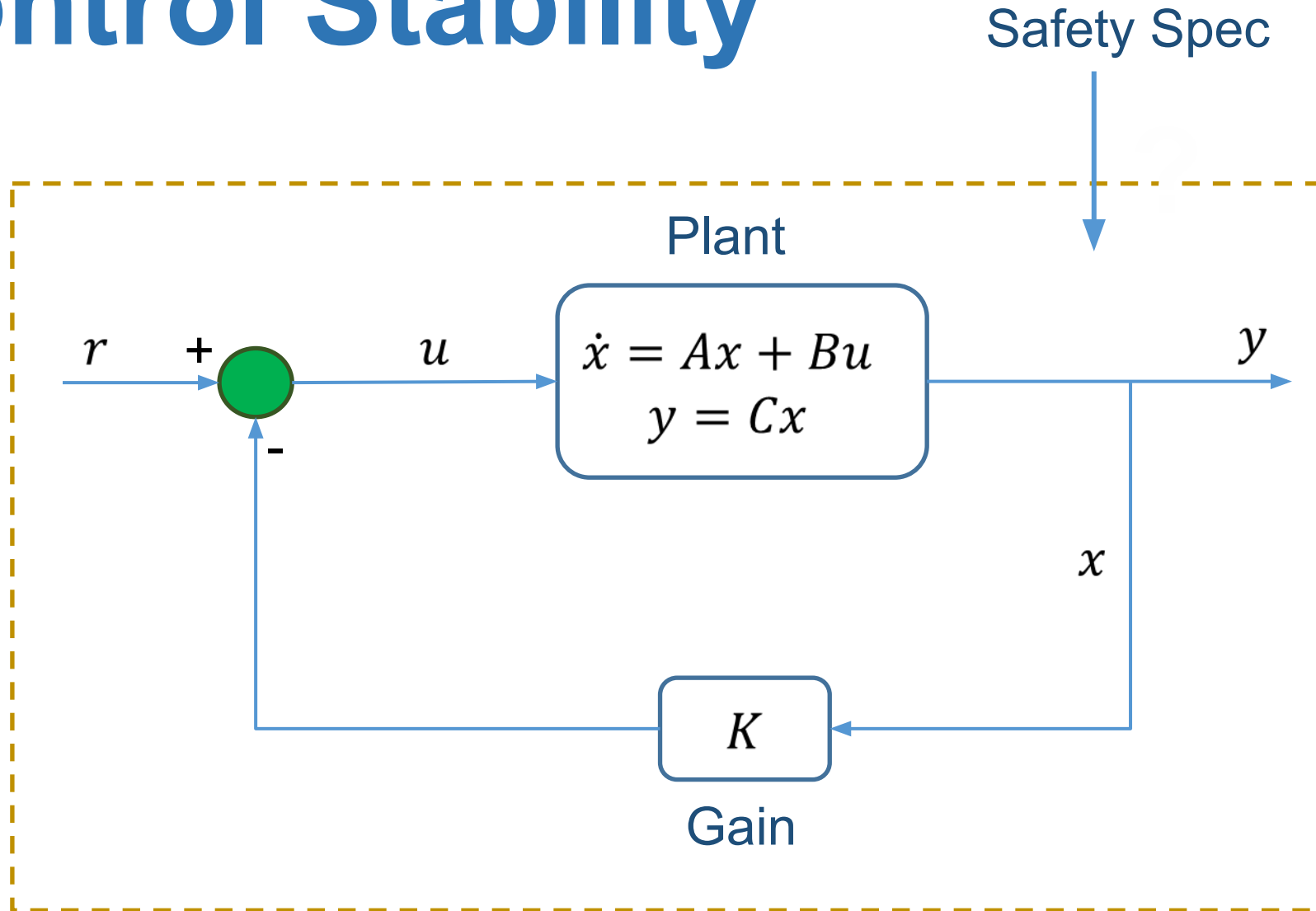


*Y. Annapureddy et al, S-TaLiRo: A Tool for Temporal Logic Falsification for Hybrid Systems. TACAS 2011.

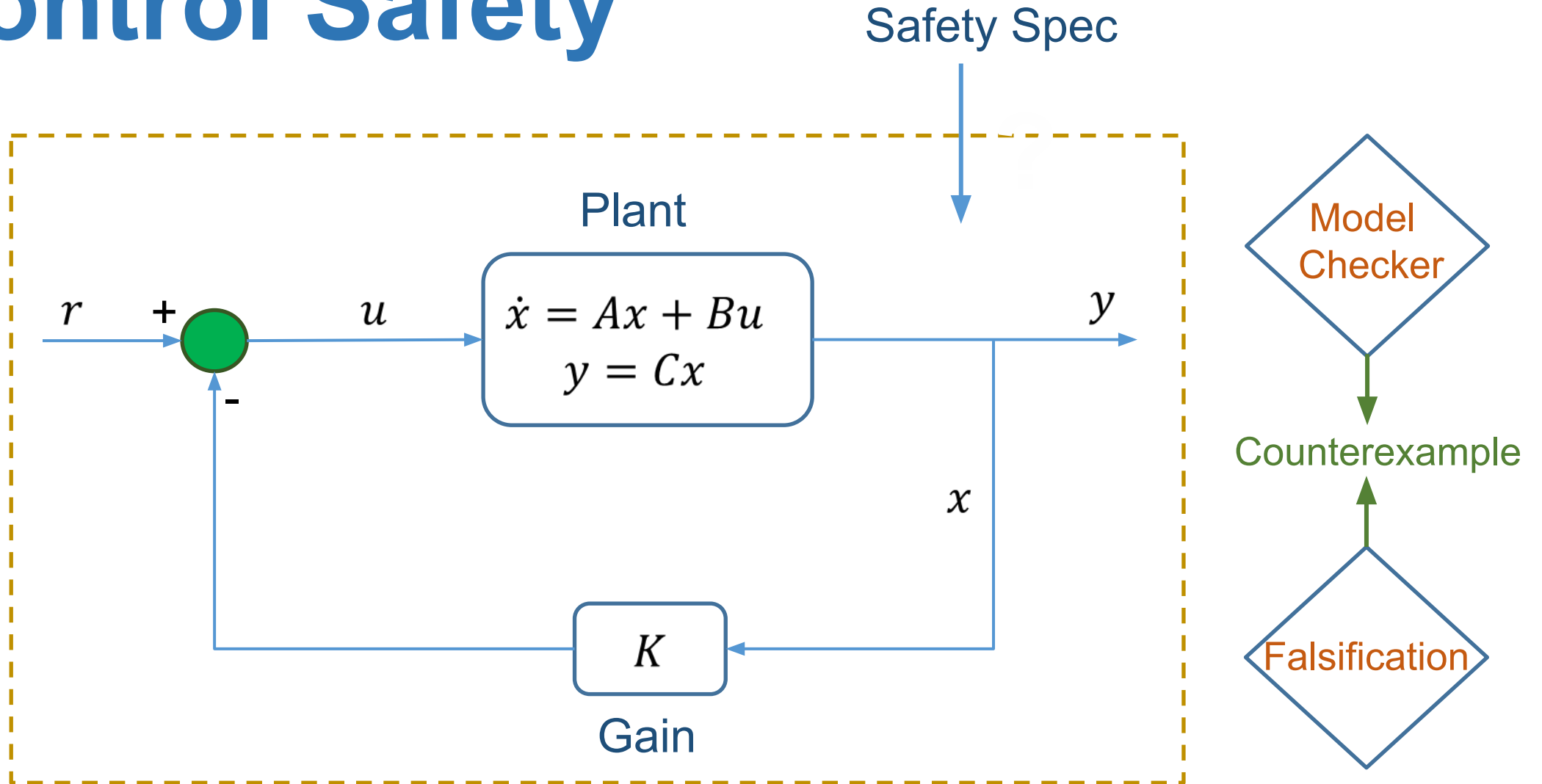
Control Stability



Control Stability



Control Safety



Motivation

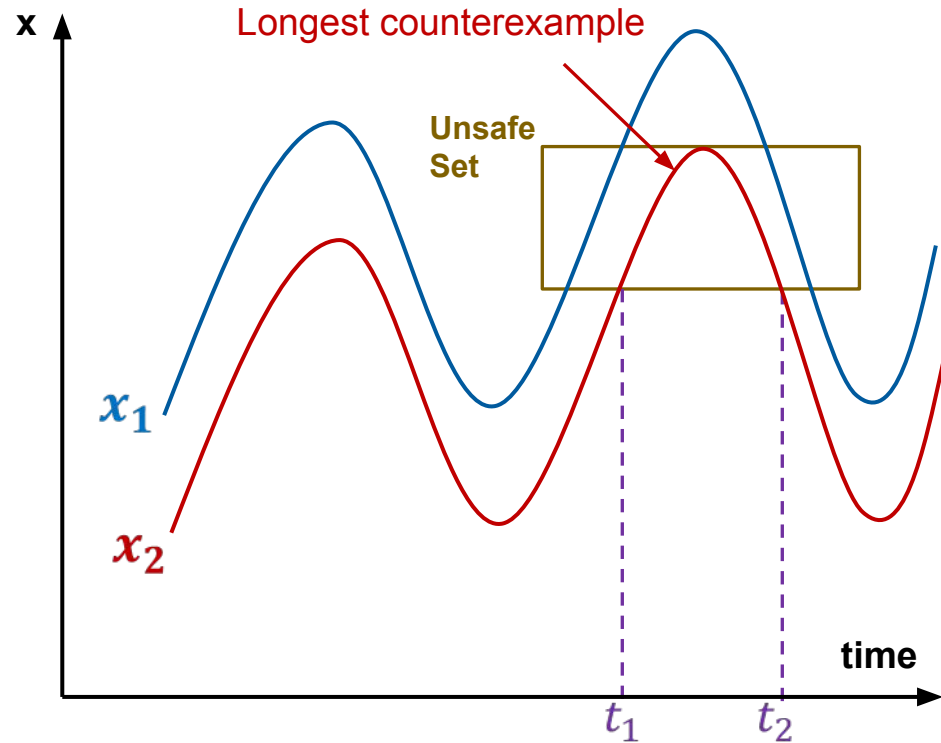


Outline

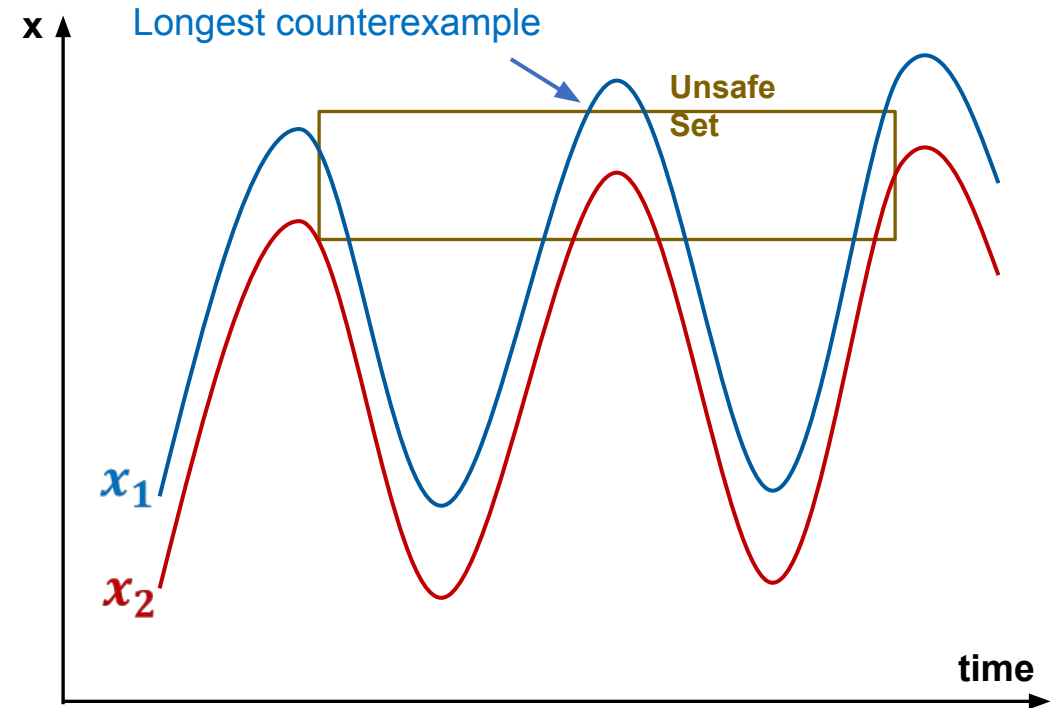
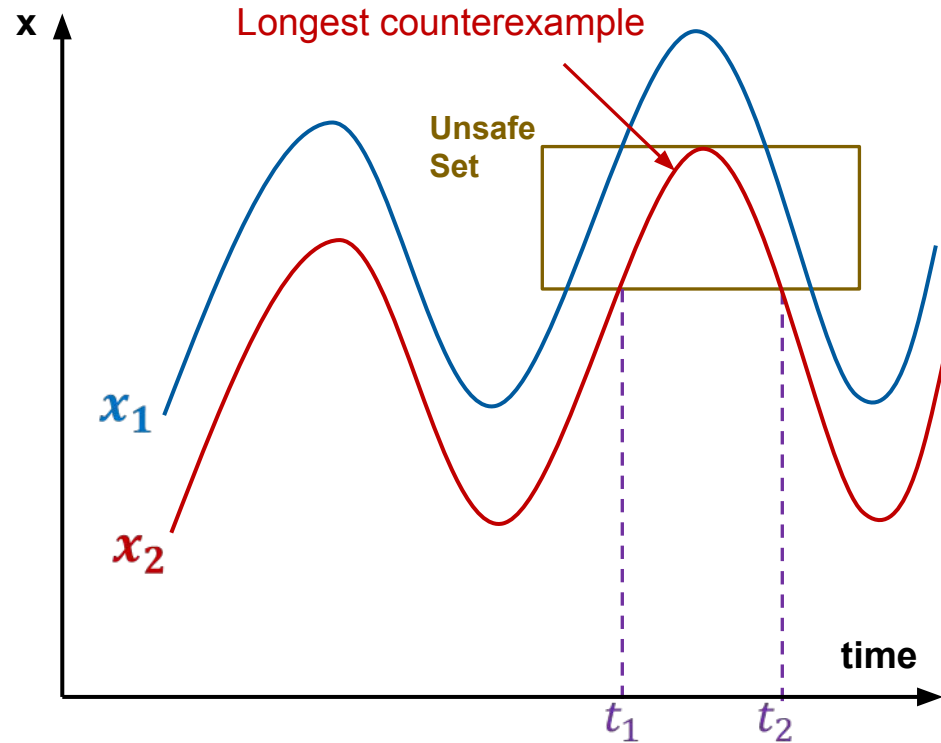
- ✓ Background
 - Introduction
 - Preliminaries
 - Methodology
 - Frameworks
 - Results



Introduction



Introduction



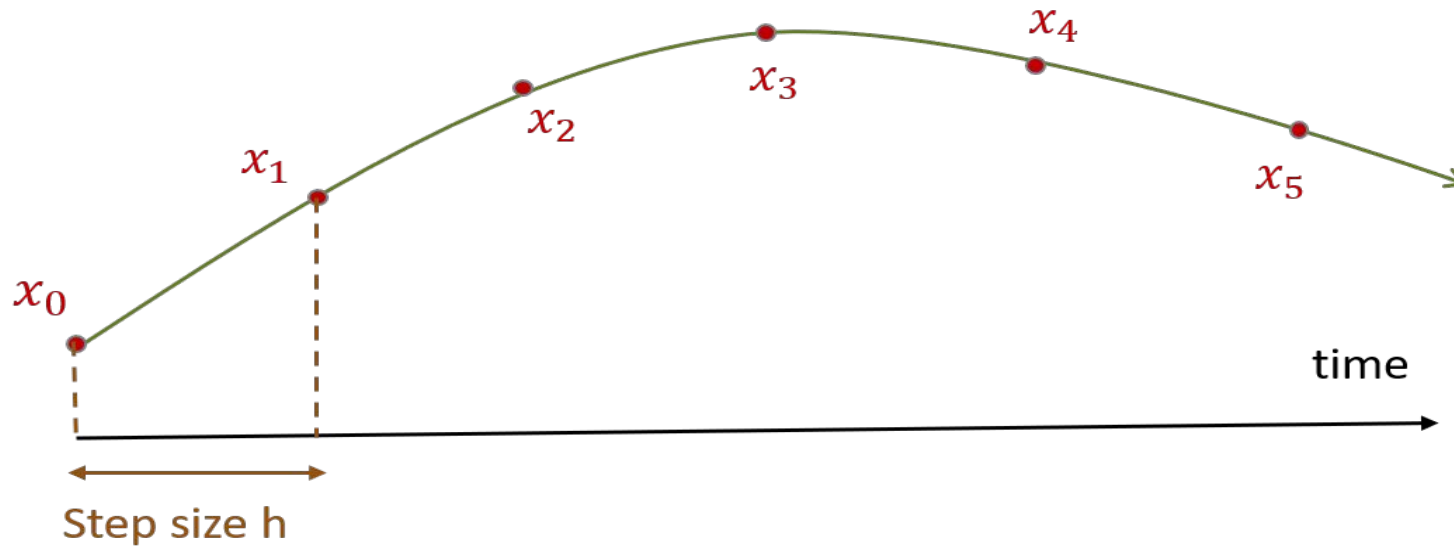
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Simulation-equivalent Analysis

For a dynamical system H with affine dynamics $\dot{x} = Ax + B$, the simulation starting from a state x_0 is computed as a sequence $\tau_H(x_0, h)$ of states at discrete time steps with step size h .

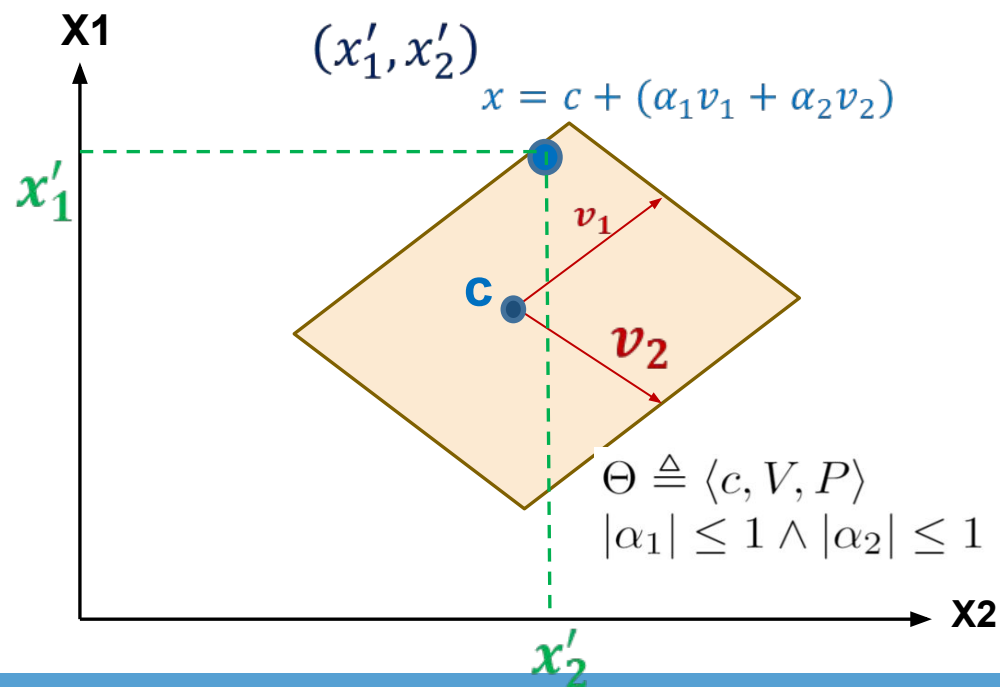


In the sequence $\tau_H(x_0, h) = x_0, x_1, x_2, \dots$, each pair (x_i, x_{i+1}) corresponds to a continuous trajectory starting at x_i and reaching x_{i+1} after h time units.

Star Representation

A *generalized star* Θ is a tuple $\langle c, V, P \rangle$ where $c \in \mathbb{R}^n$ is called the *center*, $V = \{v_1, v_2, \dots, v_m\}$ is a set of m ($\leq n$) vectors in \mathbb{R}^n called the *basis vectors*, and $P : \mathbb{R}^n \rightarrow \{\top, \perp\}$ is a predicate, defined as

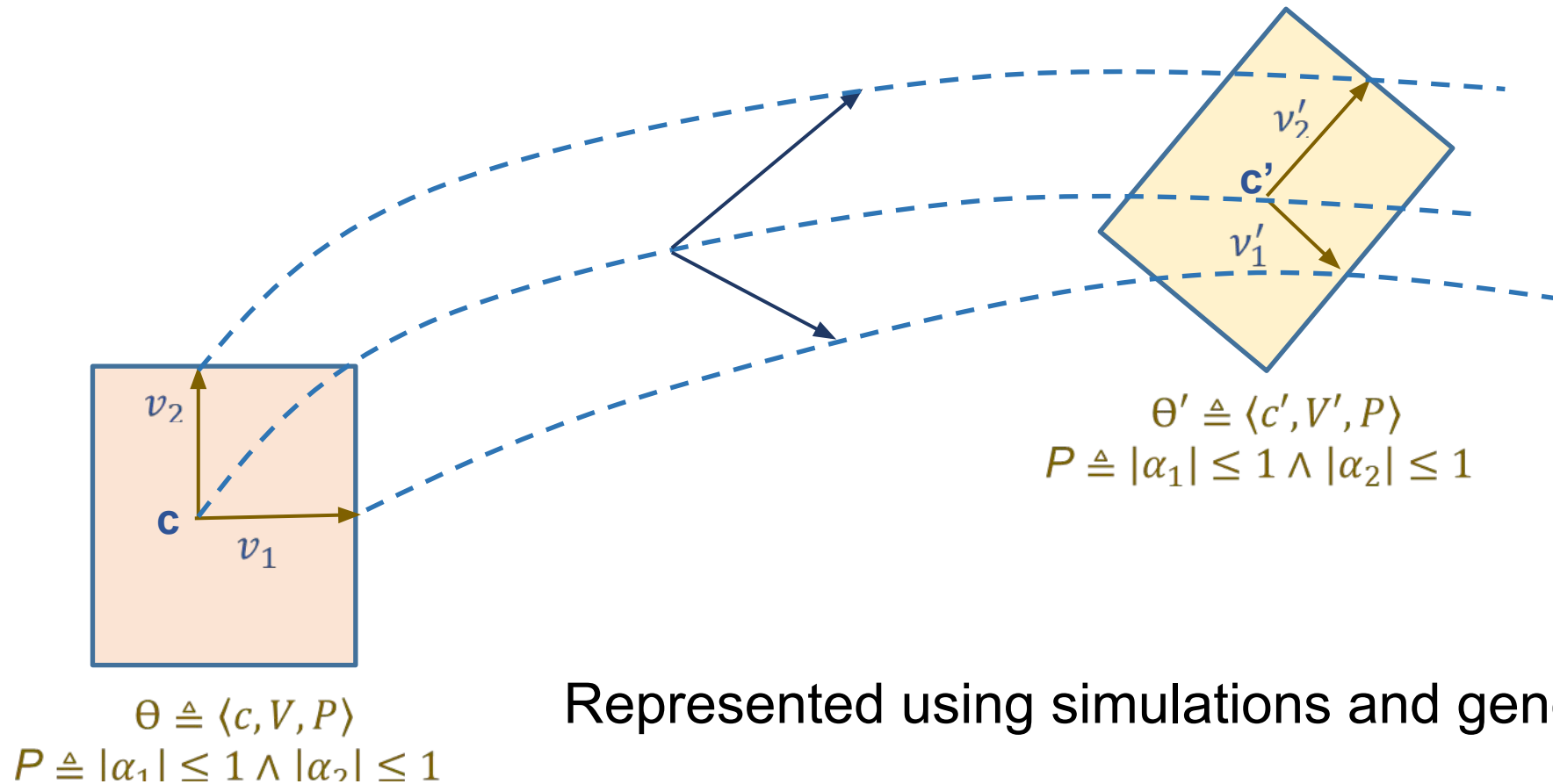
$$[\Theta] = \{x \mid \exists \bar{\alpha} = [\alpha_1, \dots, \alpha_m]^T \text{ such that } x = c + \sum_{i=1}^m \alpha_i v_i \text{ and } P(\bar{\alpha}) = \top\}$$



Variables

- Orthonormal: x'_1 and x'_2
- Basis: α_1 and α_2

Reachable Set Computation

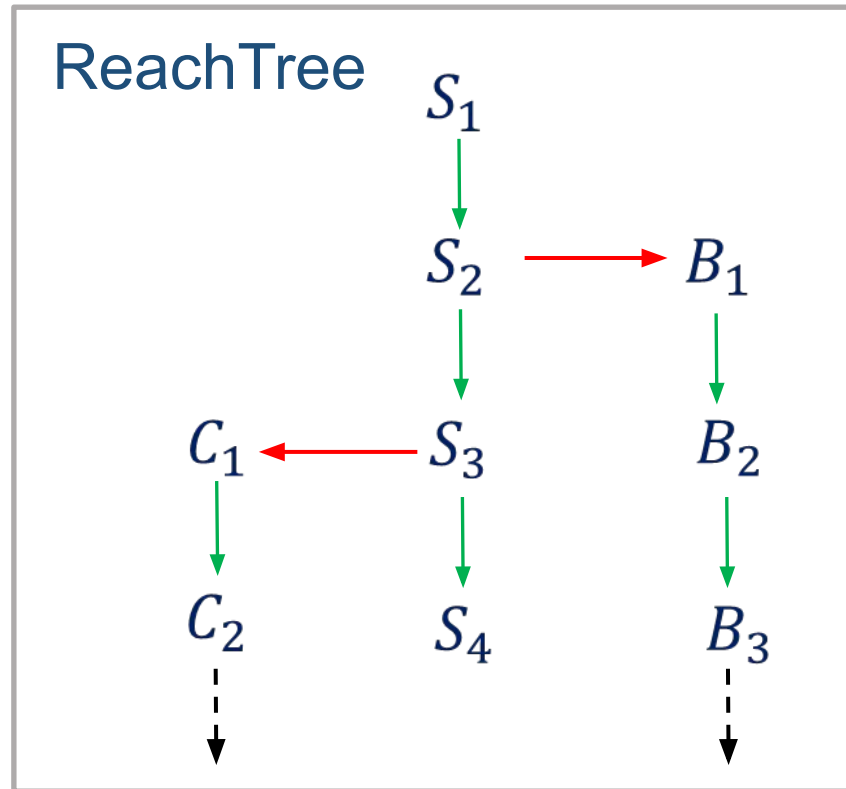


Represented using simulations and generalized star

The predicate P remains the same

Simulation-equivalent Analysis

Linear Hybrid Systems



→ Continuous transition

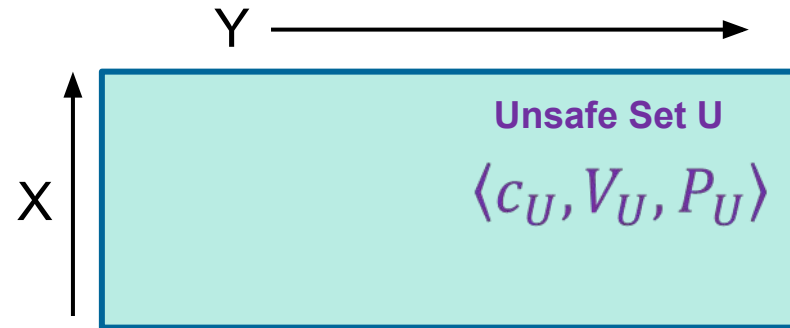
→ Discrete transition

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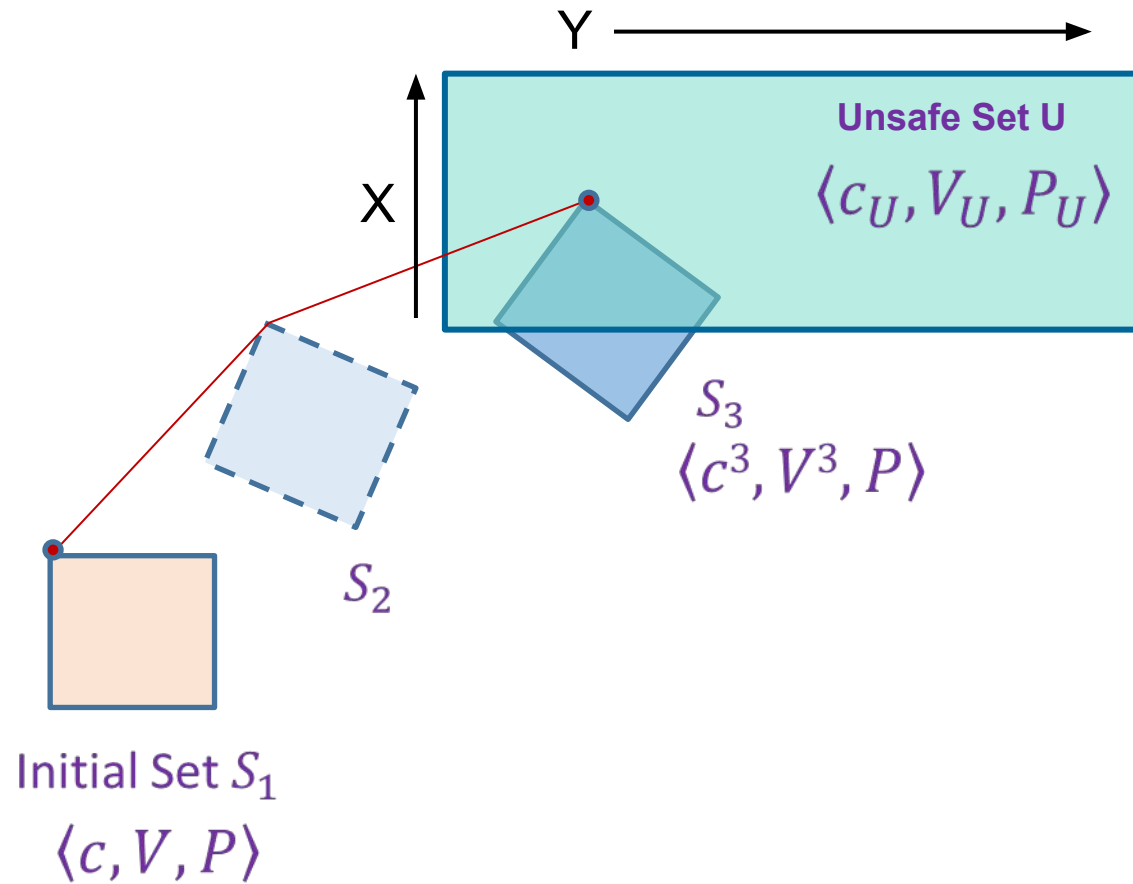


Constraint Propagation

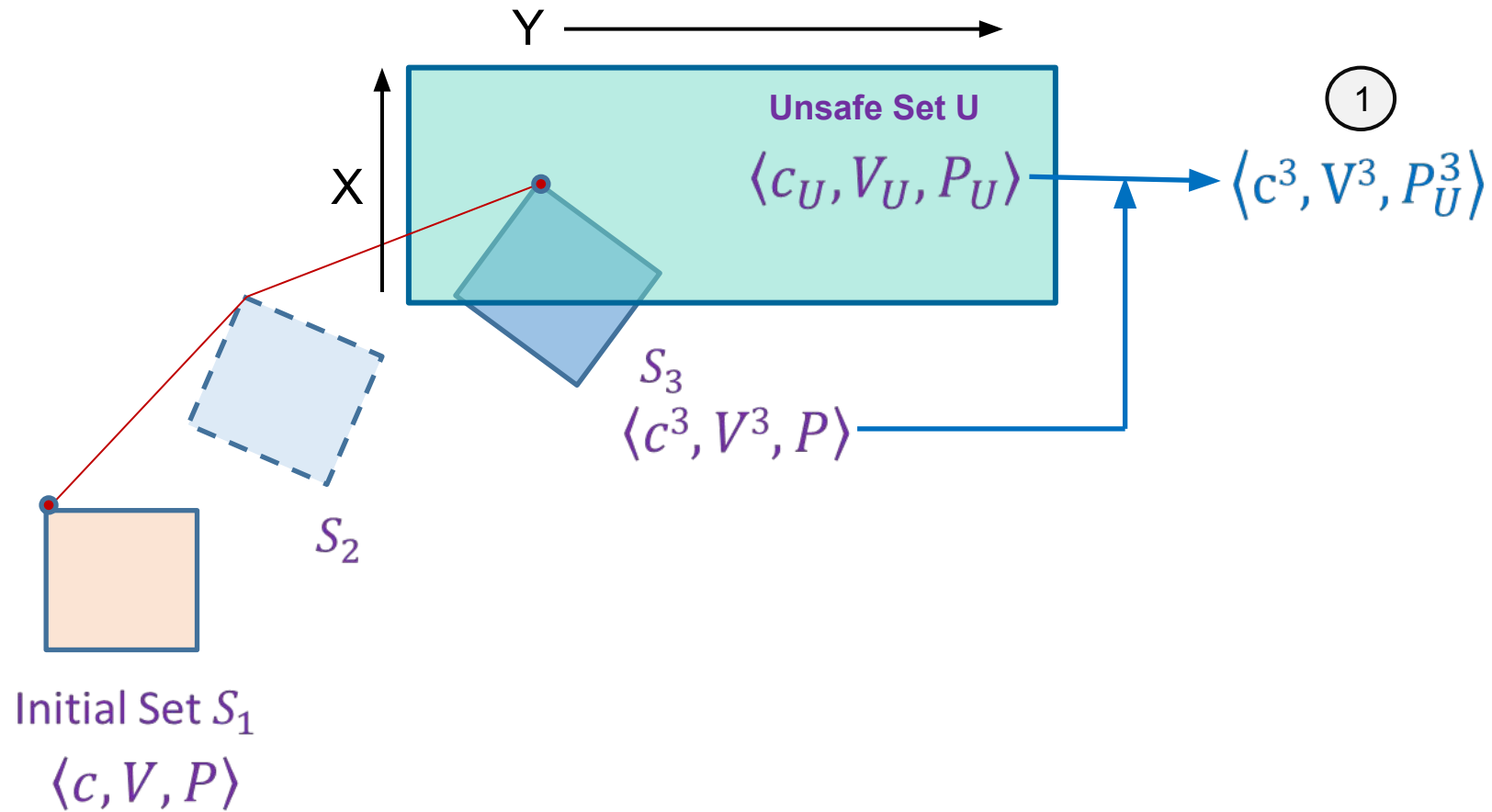


Initial Set S_1
 $\langle c, V, P \rangle$

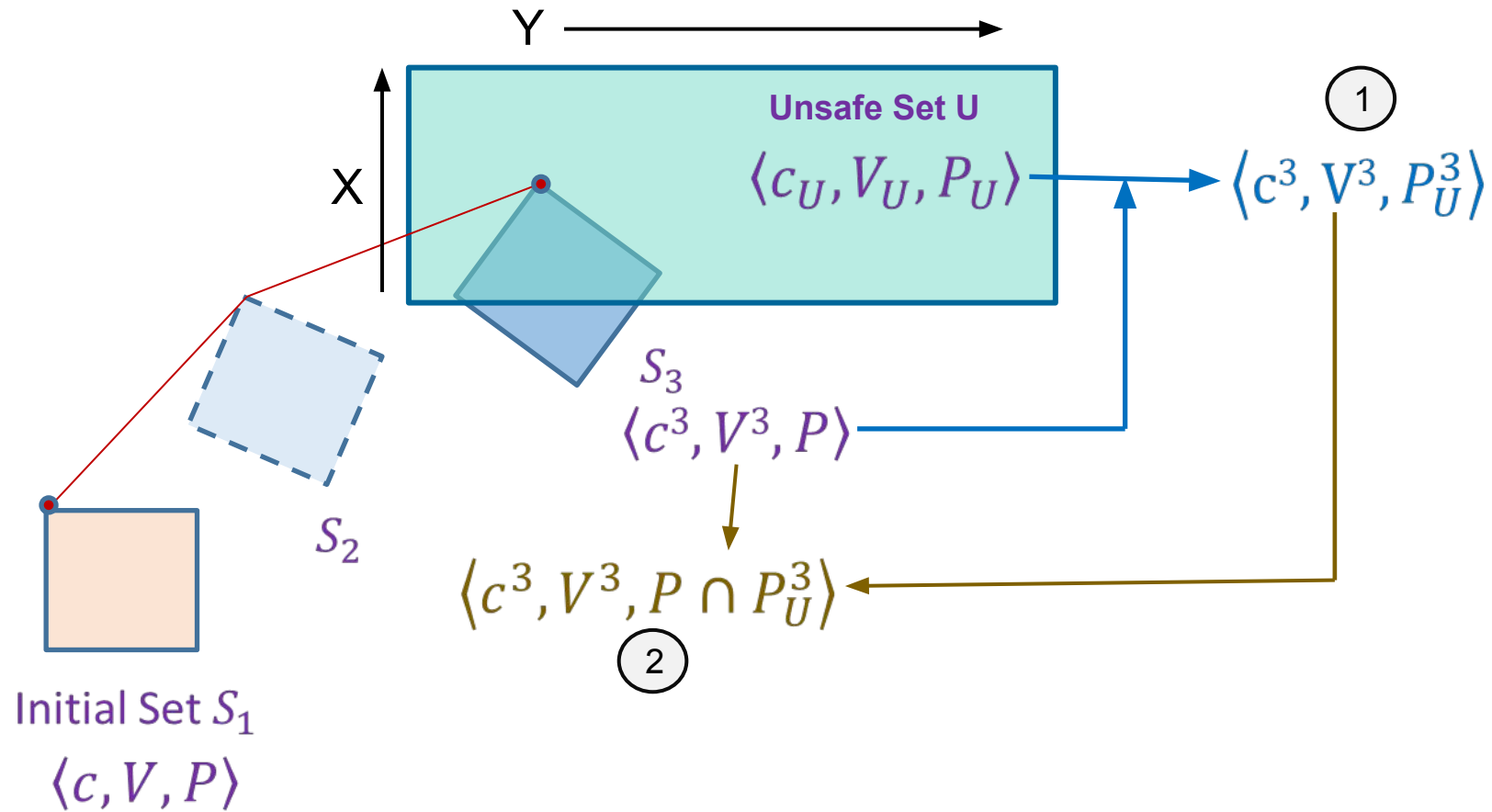
Constraint Propagation



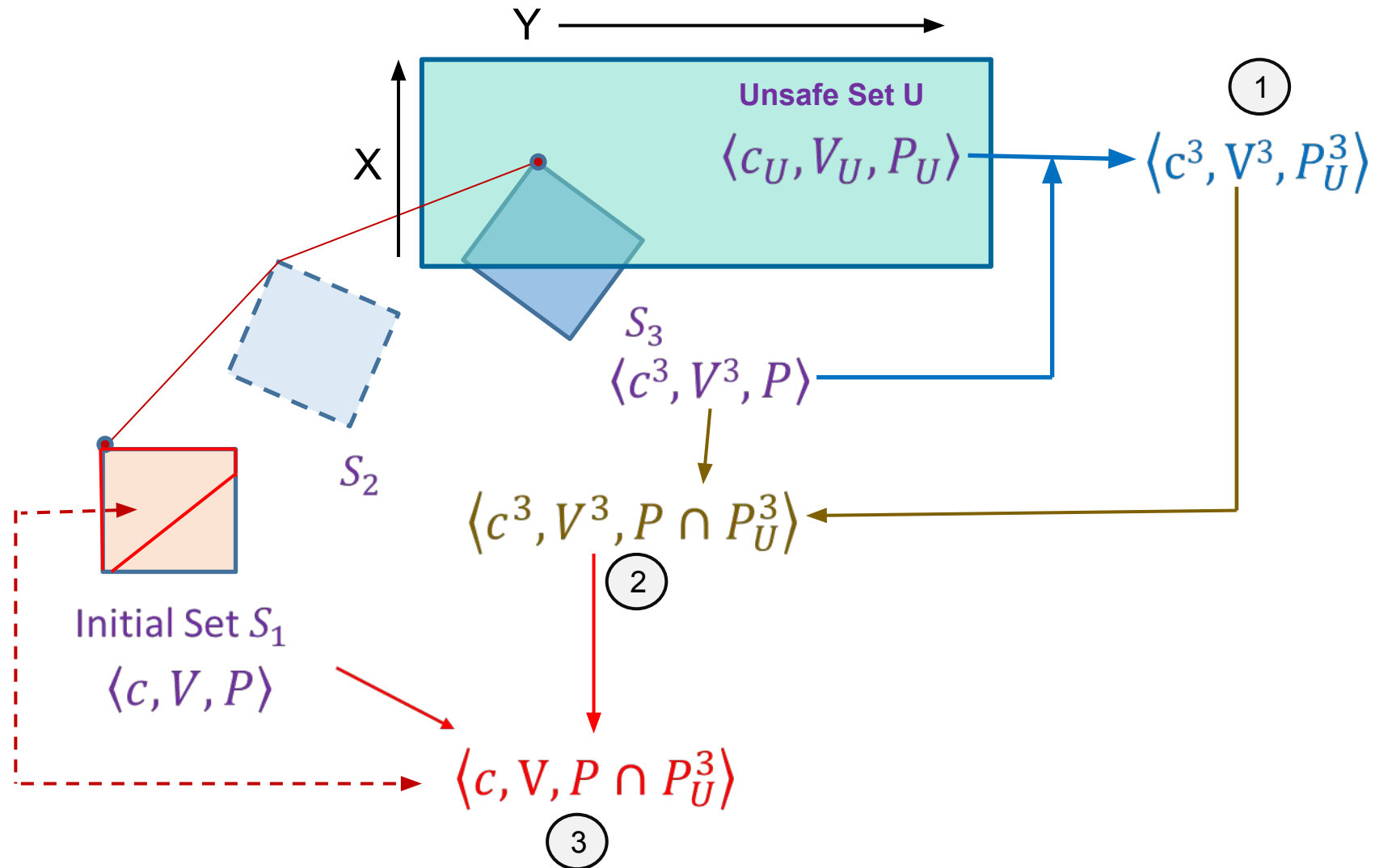
Constraint Propagation



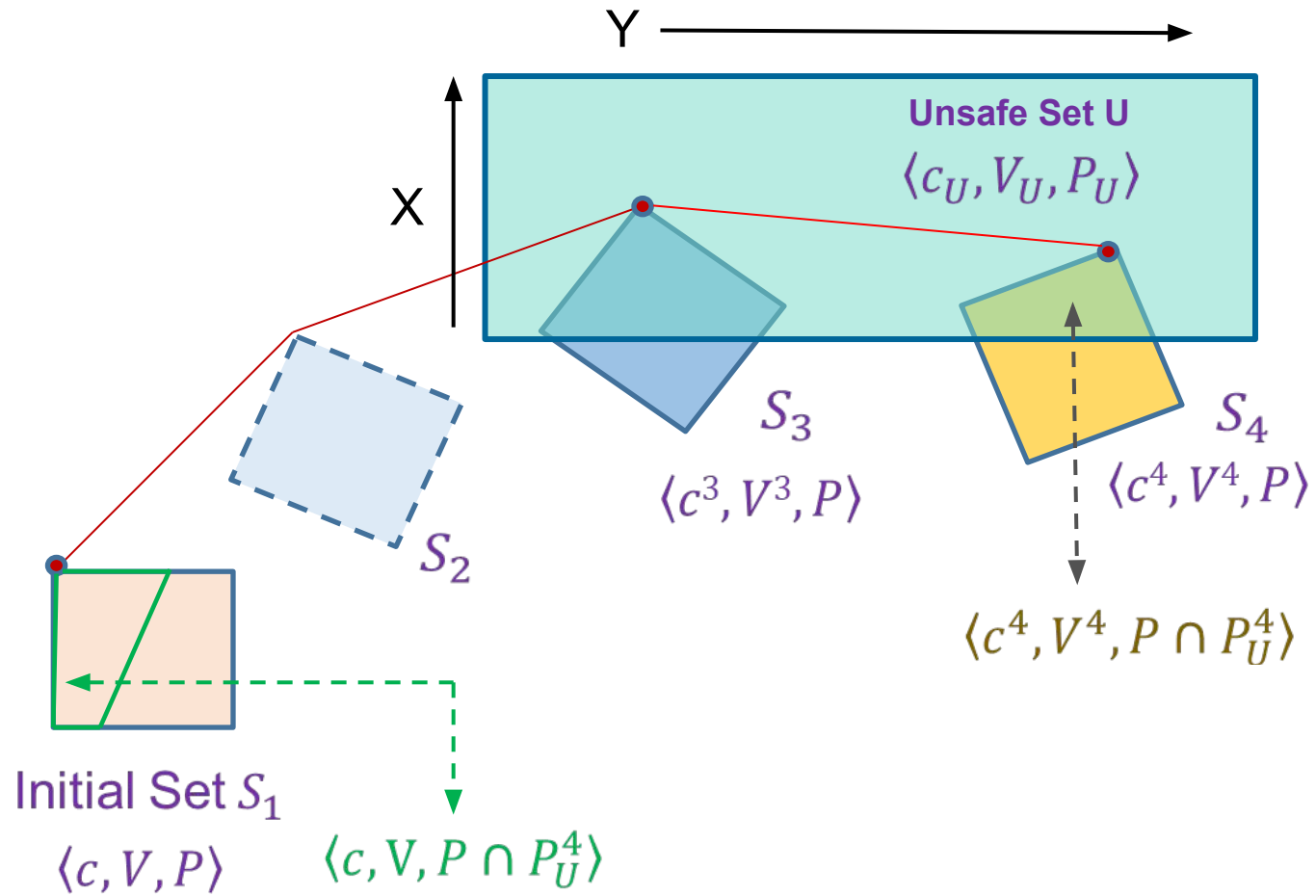
Constraint Propagation



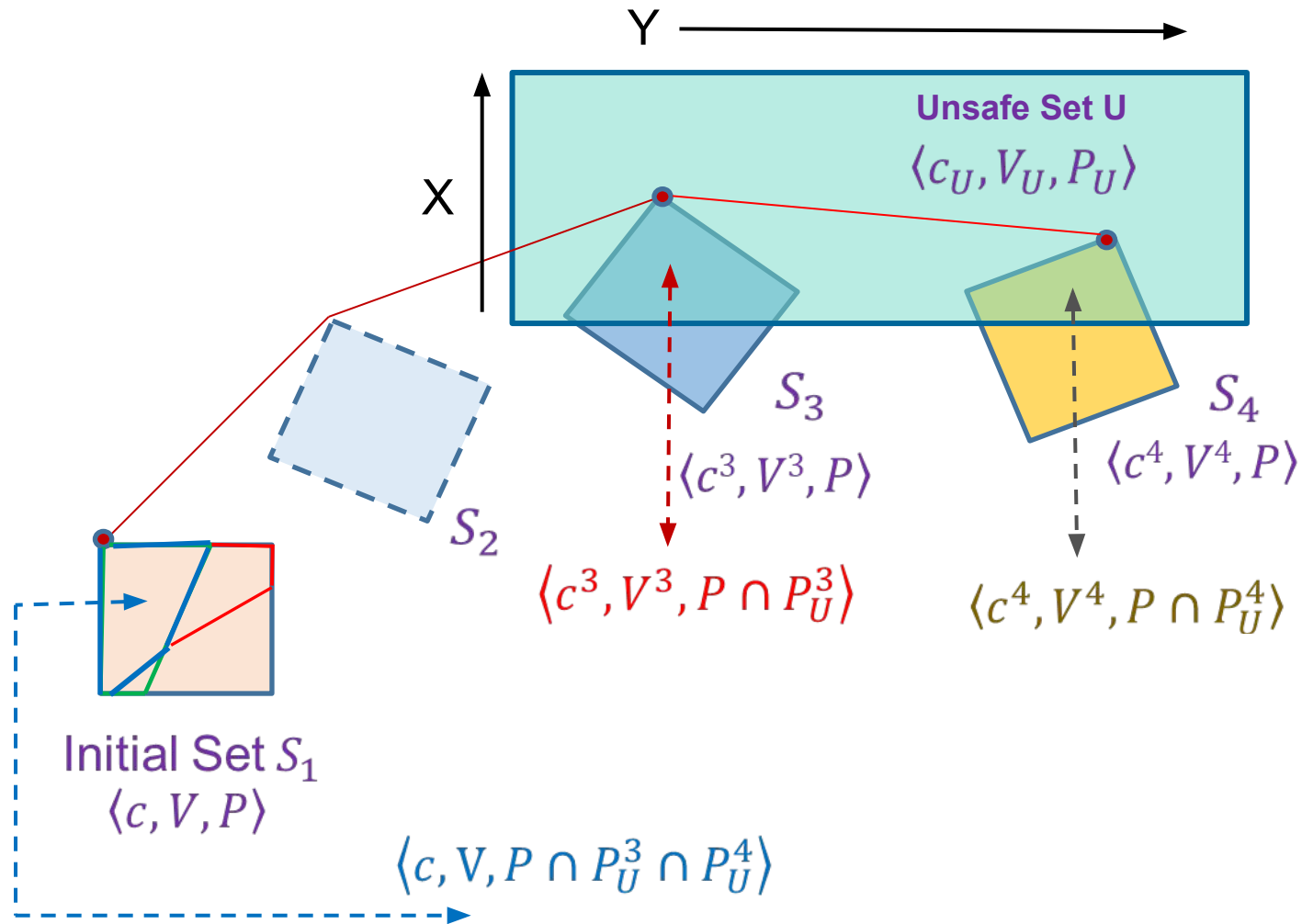
Constraint Propagation



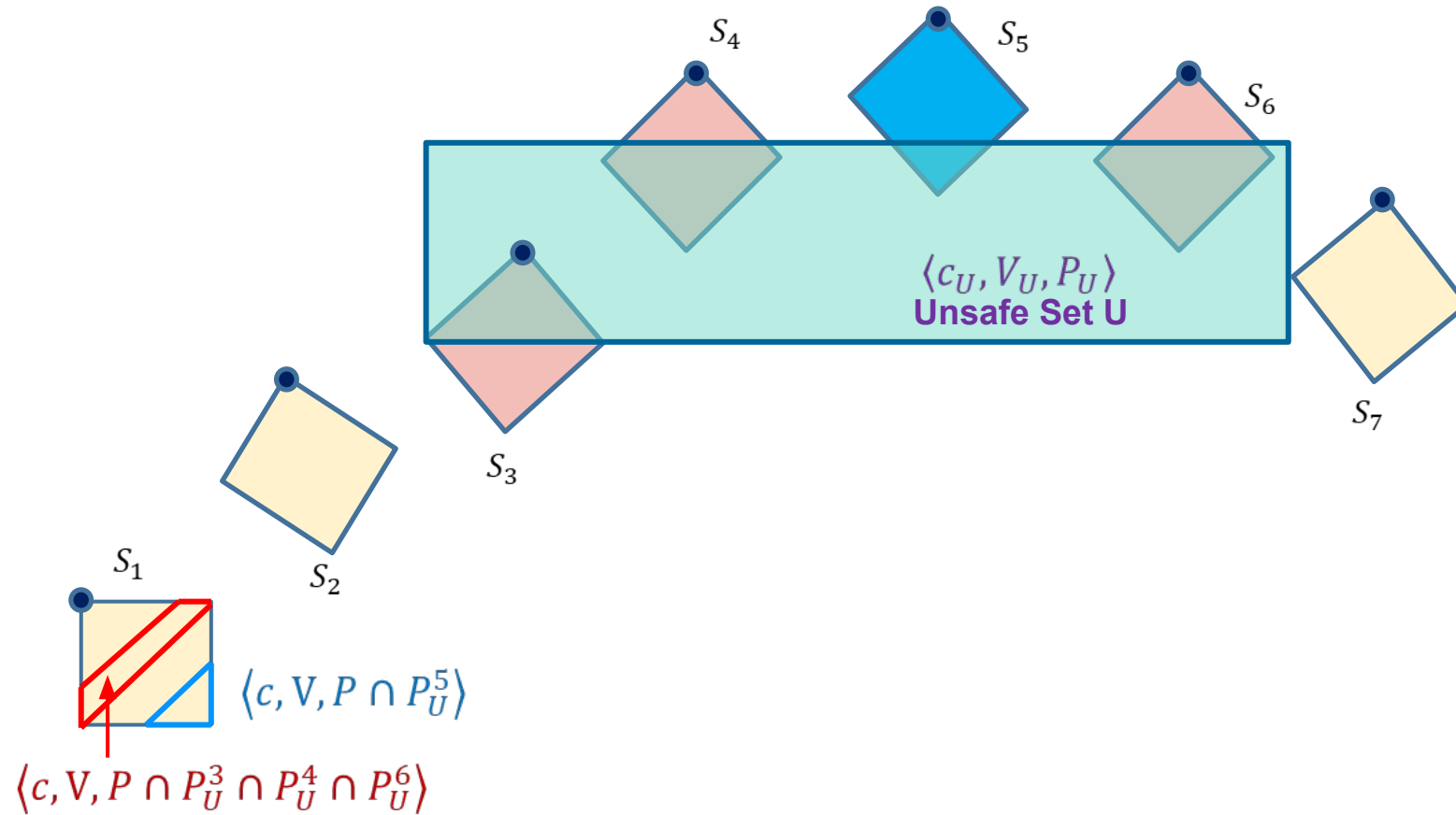
Constraint Propagation



Constraint Propagation

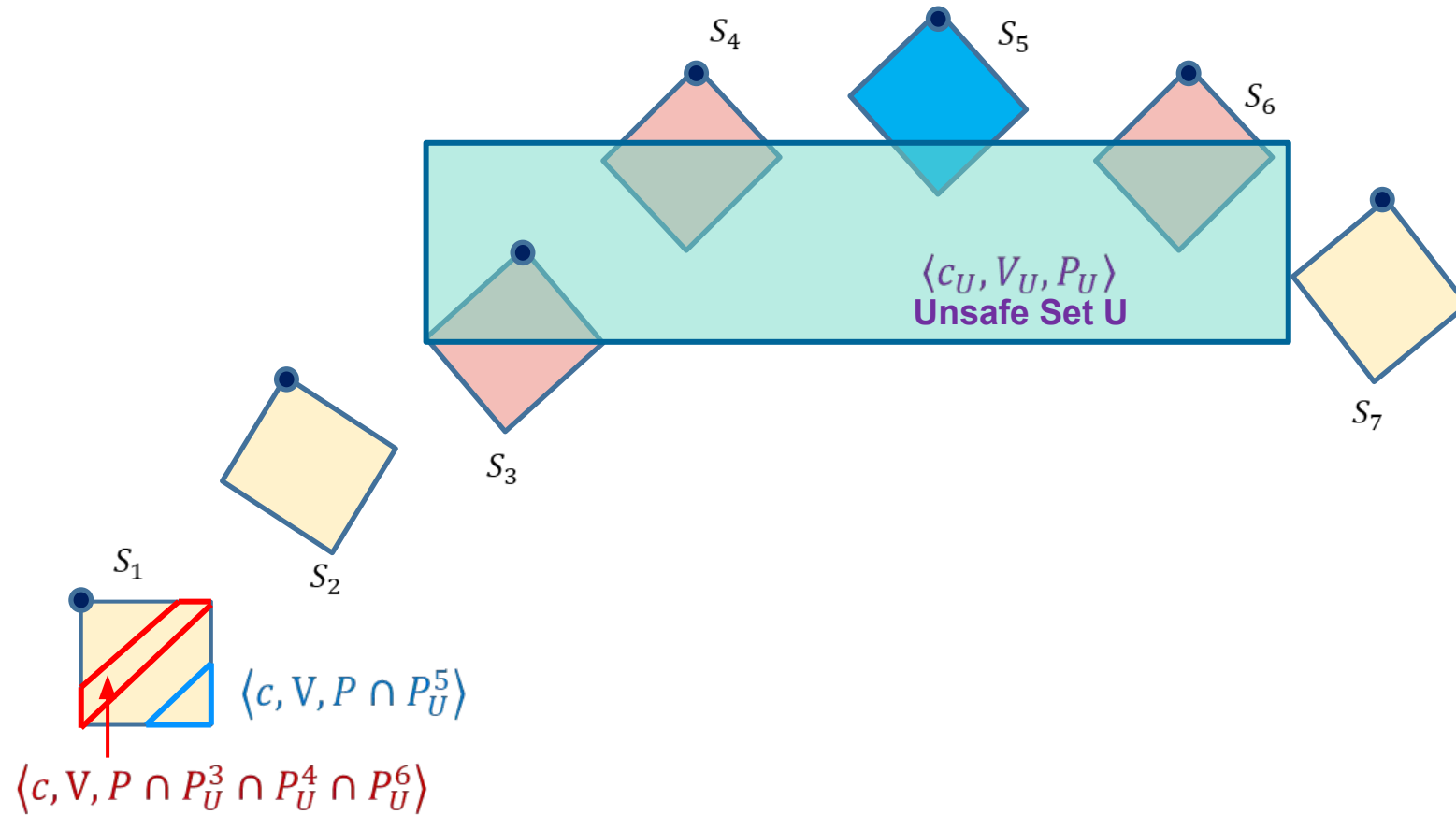


Constraint Propagation



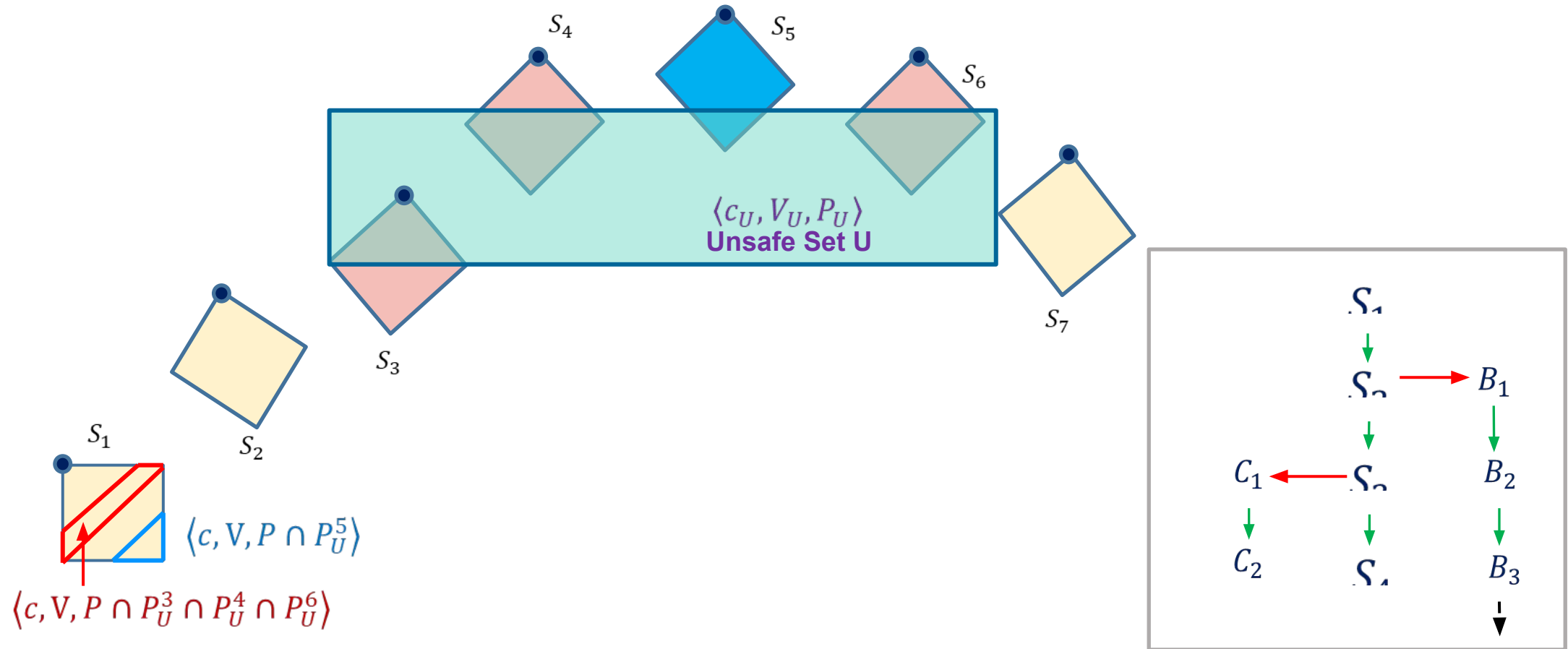
Constraint Propagation

Computationally hard!



Constraint Propagation

Computationally hard!



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MILP-based Framework

```
for each path  $\Gamma$  in ReachTree do  
   $\Pi \triangleq \{S_i | S_i \in \Gamma, S_i \cap U \neq \emptyset\}$ ;  
  Introduce  $|\Pi|$  decision variables  $z_1, z_2 \dots z_{|\Pi|}$ ;  
   $C_\Pi \leftarrow \emptyset$ ;  
  Transform  $U$  into  $\langle c^i, V^i, P_U^i \rangle$  where  
     $\Pi[i] \triangleq \langle c^i, V^i, P^i \rangle$ ;  
   $C_\Pi \leftarrow \bigwedge_{i=1}^{|\Pi|} \bigwedge_{c \in P_U^i \wedge P^i} c + M(1 - z_i)$ ;  
   $length_\Pi \leftarrow \max \sum_i z_i$  while  $C_\Pi$  is feasible;  
  if  $length_\Pi > length_{max}$  then  
     $length_{max} \leftarrow length_\Pi$ ;  
     $\bar{\alpha}_{len} \leftarrow feasible(C_\Pi)$ ;  
  end  
end
```



SMT-based Framework

for *each path* Γ *in* *ReachTree* **do**

$\Pi \triangleq \{S_i | S_i \in \Gamma, S_i \cap U \neq \emptyset\};$

Introduce $|\Pi|$ binary variables $b_1, b_2 \dots b_{|\Pi|};$

$C_\Pi \leftarrow \bigwedge_{i=1}^{|\Pi|} b_i;$ **Soft constraints**

Transform U into $\langle c^i, V^i, P_U^i \rangle$ where

$\Pi[i] \triangleq \langle c^i, V^i, P^i \rangle;$

$C_\Pi \leftarrow C_\Pi \bigwedge_{i=1}^{|\Pi|} (b_i == (P_U^i \wedge P^i));$ **Hard constraints**

$length_\Pi \leftarrow Optimize_{SMT}(C_\Pi)$ while C_Π is
feasible;

if $length_\Pi > length_{max}$ **then**

$length_{max} \leftarrow length_\Pi;$

$\bar{\alpha}_{len} \leftarrow feasible(C_\Pi);$

end

end



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Evaluation

Adaptive Cruise Control*

$$\dot{s} = (v_f - v)$$

$$\dot{v} = a$$

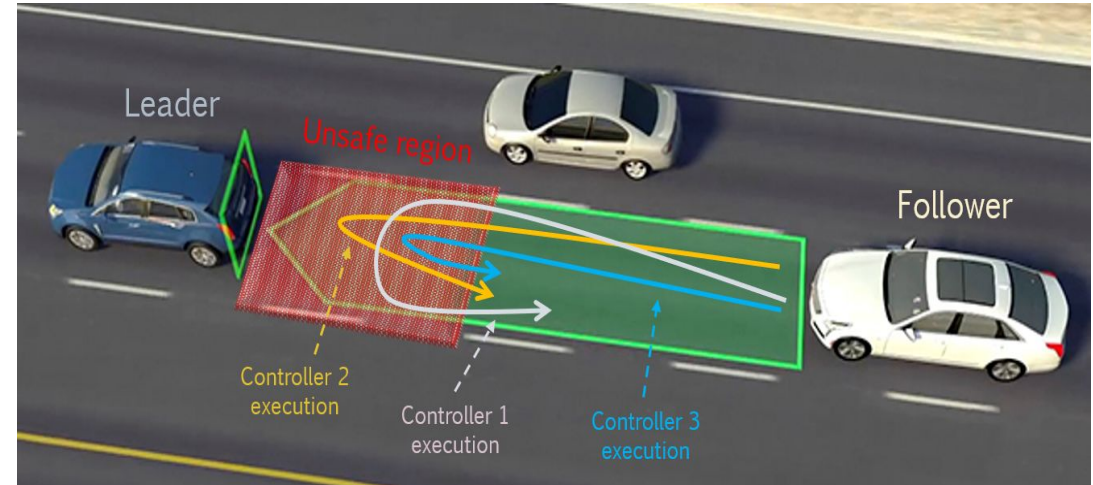
$$\dot{a} = g_1 a + g_2 (v - v_f) + g_3 (s - (v + 10))$$

v_f : leading car's velocity

v : follower's speed

a : follower's acceleration

s : distance



*A. Tiwari, Approximate reachability for linear systems. HSCC, 2003

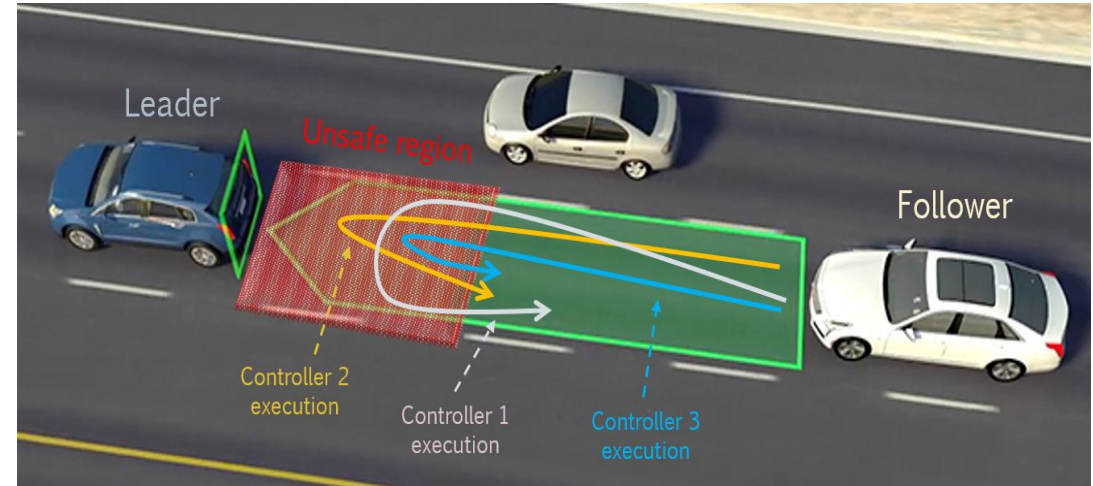
Evaluation

Adaptive Cruise Control*

$$\dot{s} = (v_f - v)$$

$$\dot{v} = a$$

$$\dot{a} = g_1 a + g_2 (v - v_f) + g_3 (s - (v + 10))$$



v_f : leading car's velocity

v : follower's speed

a : follower's acceleration

s : distance

Initial Values

$$s \in [2, 5]$$

$$v \in [18, 22]$$

$$v_f = 20$$

$$a \in [-1, 1]$$

Controller 1

$$g_1 = -3$$

$$g_2 = -3$$

$$g_3 = 1$$

Controller 2

$$g_1 = -1$$

$$g_2 = -3$$

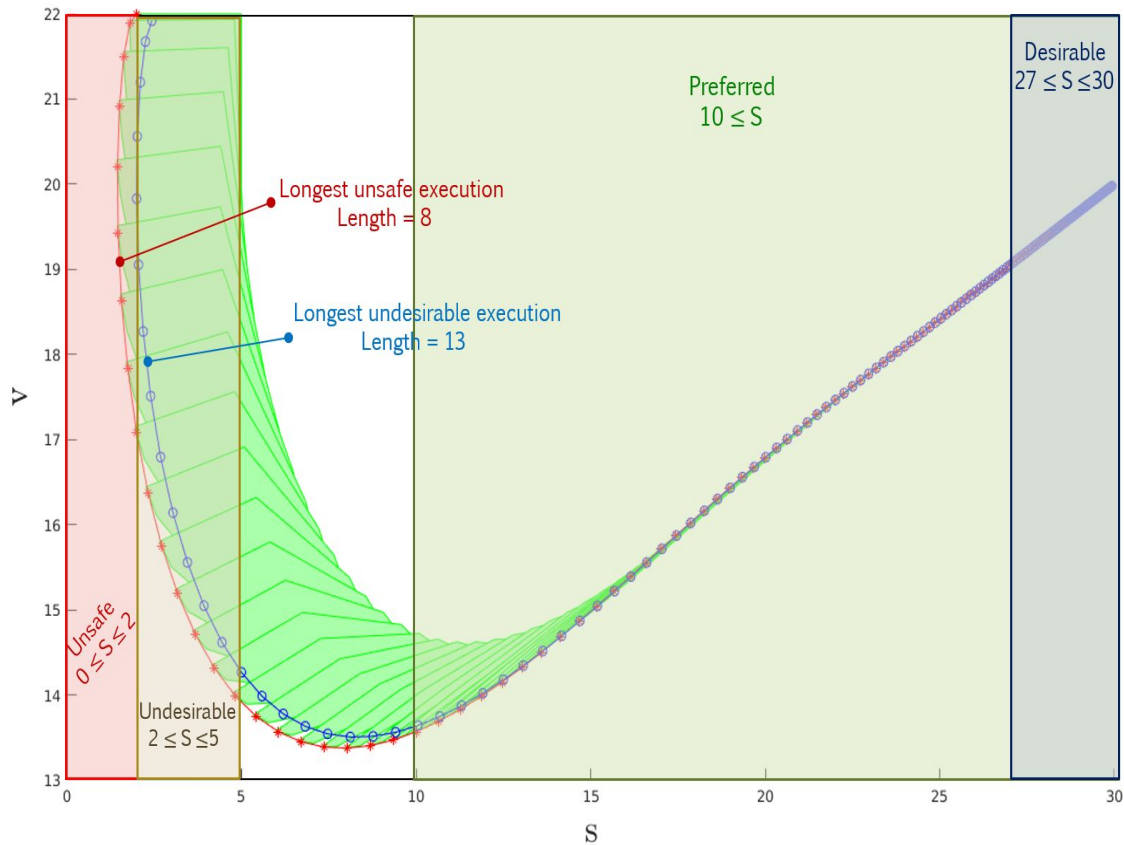
$$g_3 = 1$$

*A. Tiwari, Approximate reachability for linear systems. HSCC, 2003

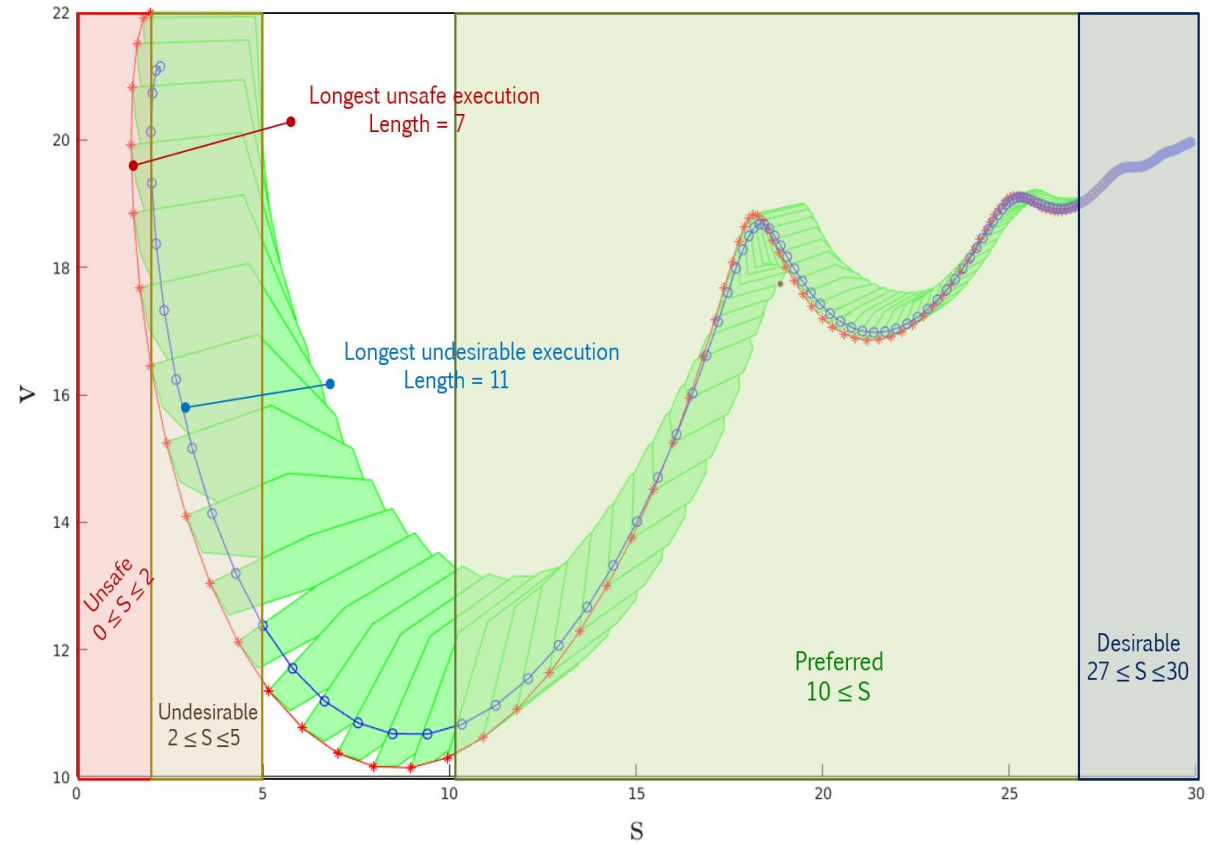
Evaluation

$$\begin{aligned} \dot{s} &= (v_f - v) \\ \dot{v} &= a \\ \dot{a} &= g_1 a + g_2 (v - v_f) + g_3 (s - (v + 10)) \end{aligned}$$

ACC Controller I



ACC Controller II



*M. Goyal, P. S. Duggirala, Extracting counterexamples induced by safety violation in linear hybrid systems, Automatica, 07/2020

Evaluation

Model	Dims, Modes	Actual Inter. Duration	LCE Duration		Counterexample		Verification Time (sec)	LCE Gen. Time	
			MILP	SMT	MILP	SMT		MILP	SMT
Damped Oscillator 1	2, 1	[5 10] [34 44] [66 74]	[5 9] [35 44] [66 73]	[5 9] [35 44] [66 73]	[-5.28 0.764]	[-5.321 0.865]	0.44	0.04	0.51
Damped Oscillator 2	2, 1	[3 10] [29 49] [59 100]	[3 10] [30 49] [59 100]	[3 9] [29 49] [59 100]	[-5.0 0.398]	[-5.0 0.606]	0.59	0.04	0.55
Buck Converter	4, 6	cl1: [13 21] op1: [22 50] cl2: [51]	[13 21] [22 50] [51]	[13 21] [22 50] [51]	il = 1.0 vc = 0 t = 0, gt = 0	il = 0.6892 vc = 0 t = 0, gt = 0	0.66	0.04	0.60
Filtered Oscillator	34, 4	loc3: [3 5] loc3: [7 21] loc4: [26]	[5] [7 21] [26]	[5] [7 21] [26]	[0.2069 0.07 0...]	[0.205 0.07 0...]	37	2.14	49

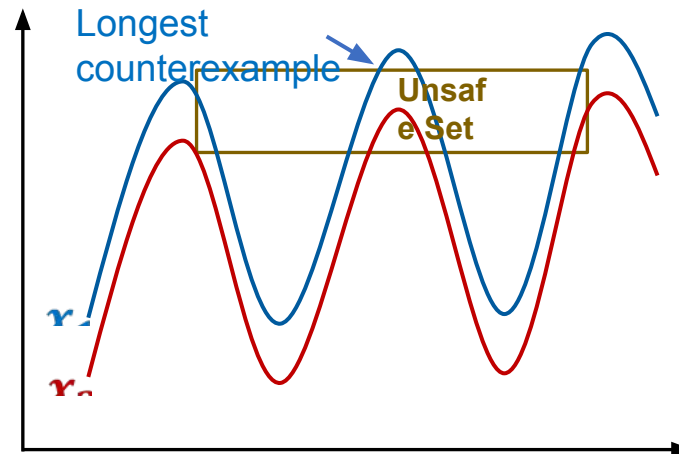
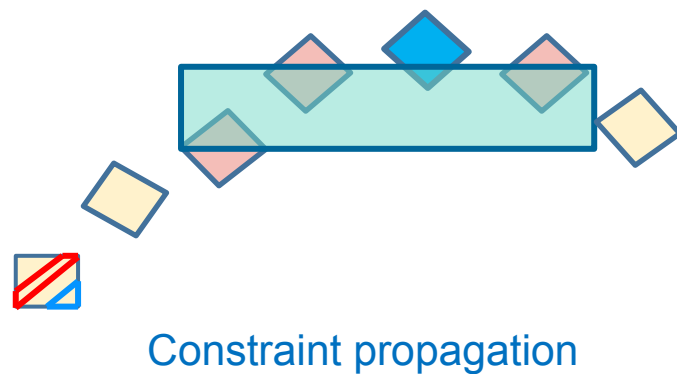
Takeaways

- Use the artifacts from verification
- Search in the space of basis variables defining the initial set
- Longest counterexample may not be unique
- MILP-based framework is faster
- SMT-based formalism provides guarantees

Future work

- Explore BDD-based optimization techniques
- Use various counterexamples in controller synthesis

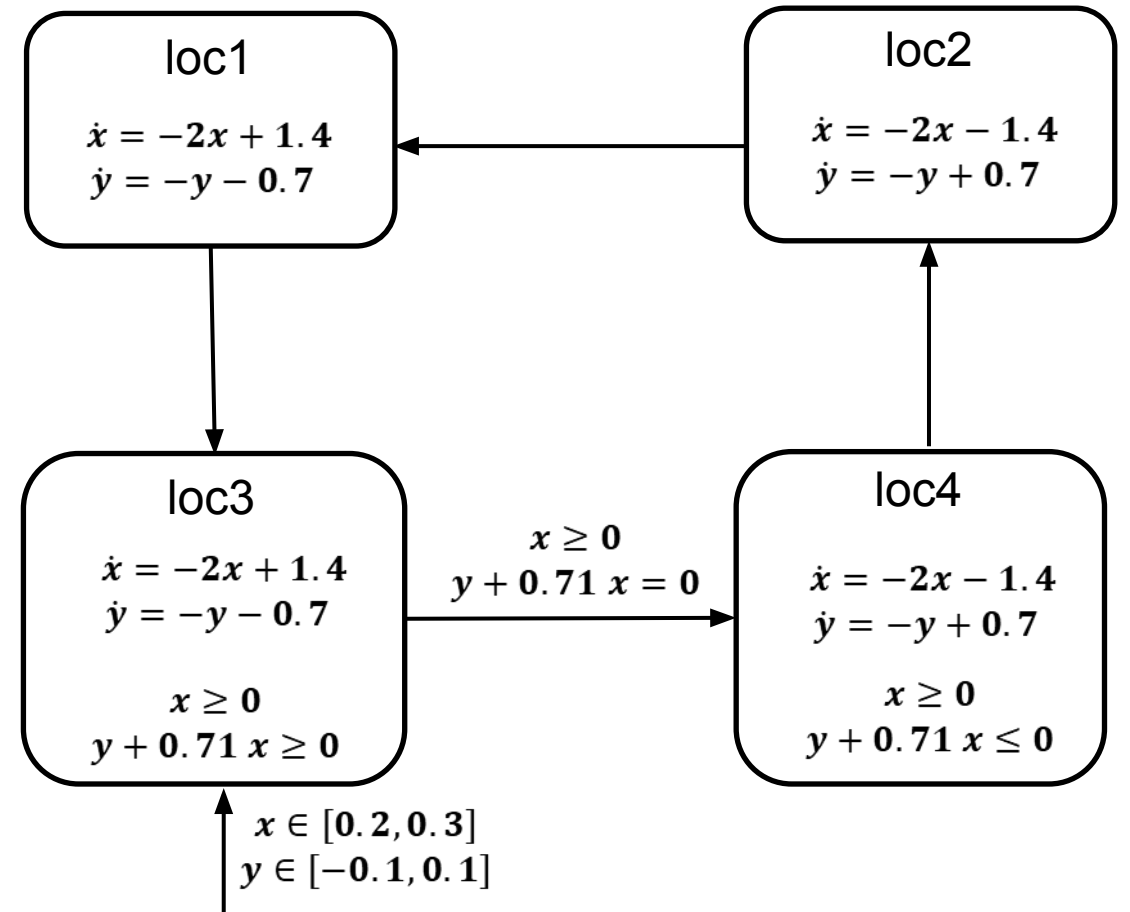
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Verification



Filtered Oscillator

What to do!

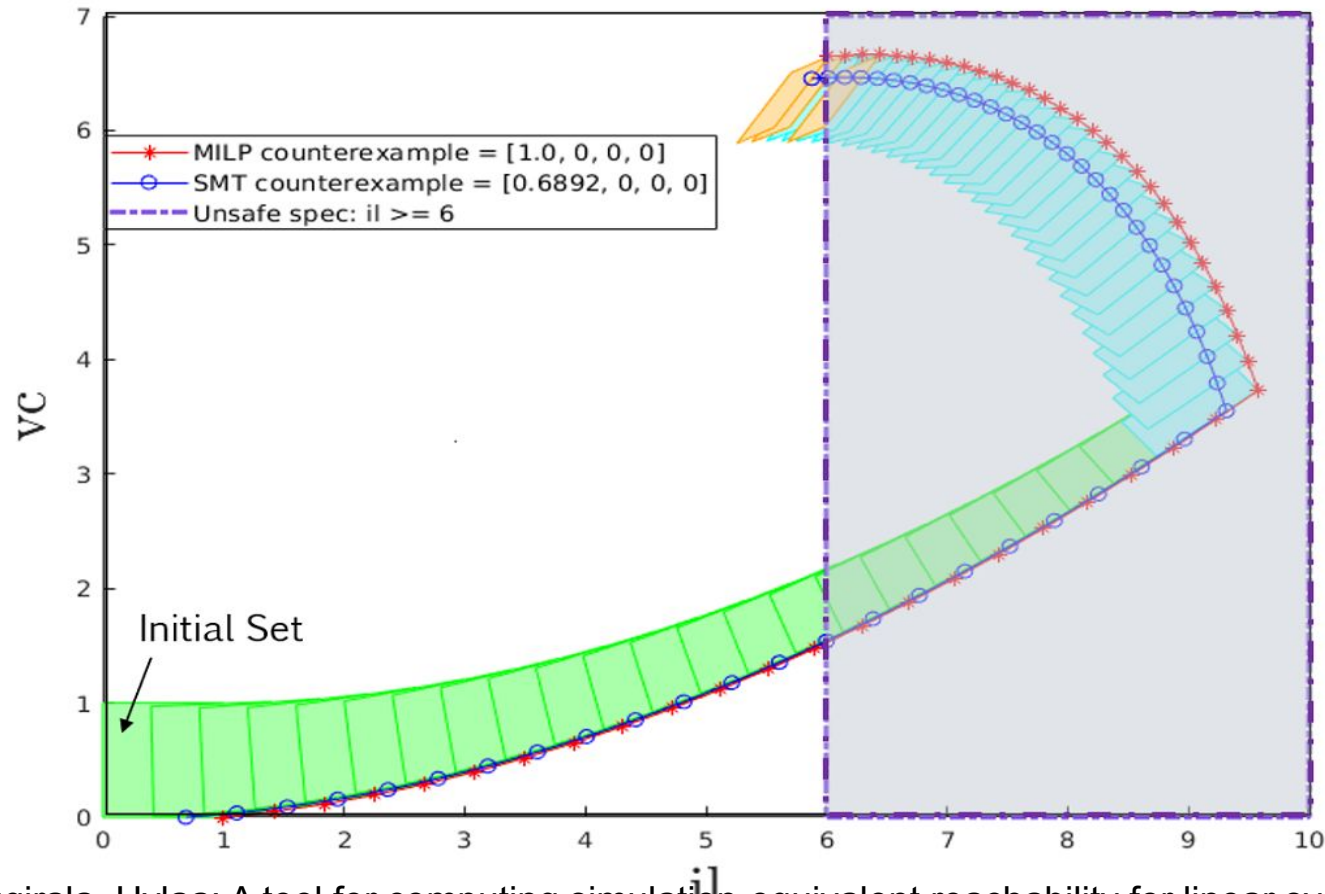
The entire 'unsafe' reachable set



Randomly generated counterexample

Evaluation*: HyLaa

Buck Converter



*S. Bak and P. S. Duggirala, Hylaa: A tool for computing simulation-equivalent reachability for linear systems. HSCC 2017