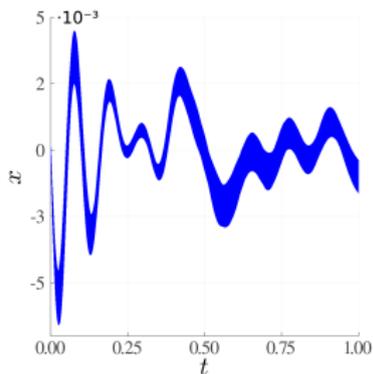
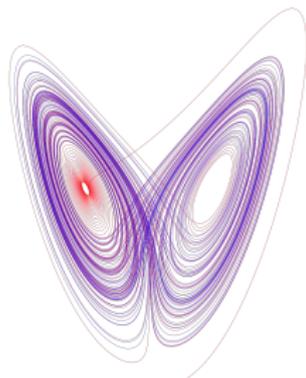

A gentle introduction to reachability analysis for dynamical systems



Christian Schilling
Aalborg University, Denmark
July 13, 2022



AALBORG UNIVERSITET



Dynamical systems

- **Continuous-time systems** modeled by **ordinary differential equations**

$$\dot{x}(t) = f(x(t)) \quad (x \in \mathbb{R}^n)$$

- **Initial-value problem**: Given an initial state $x_0 \in \mathbb{R}^n$, determine the **solution/trajectory** following f

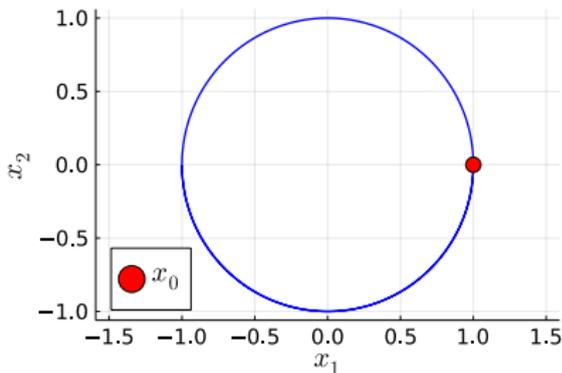
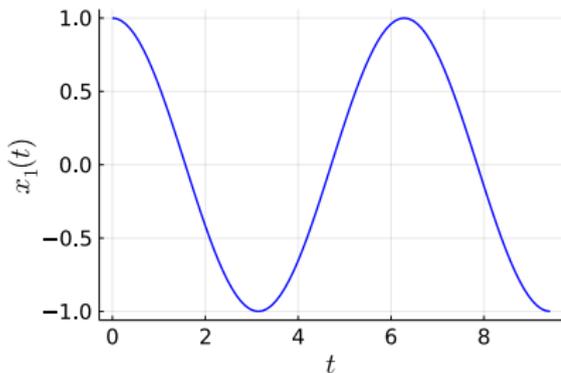
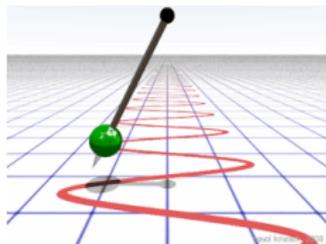
Simulation

- Traditionally we use **simulations** to understand such systems

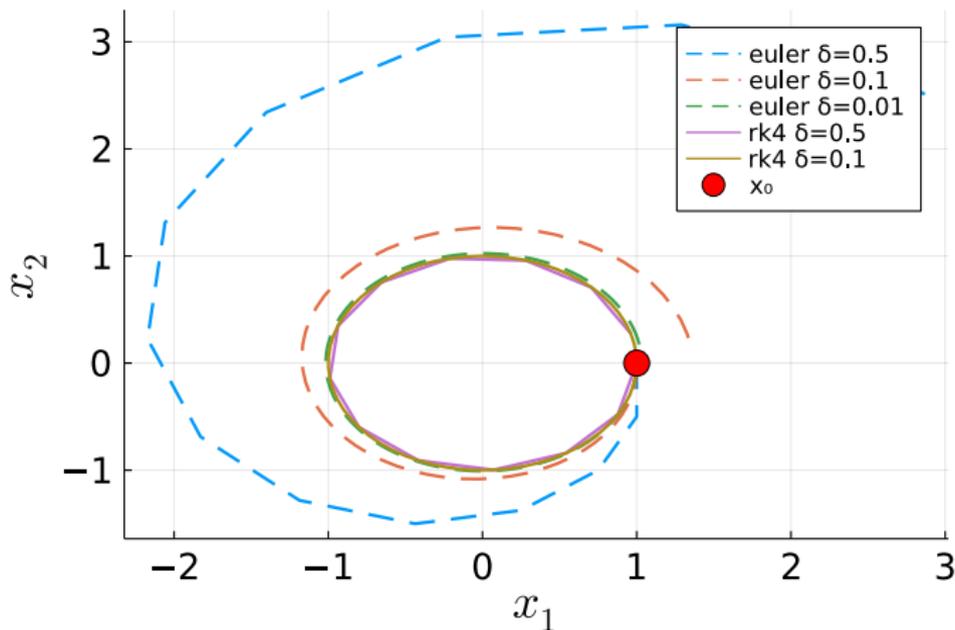


Example: **Harmonic oscillator**

$$\dot{x} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} x, \quad x_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

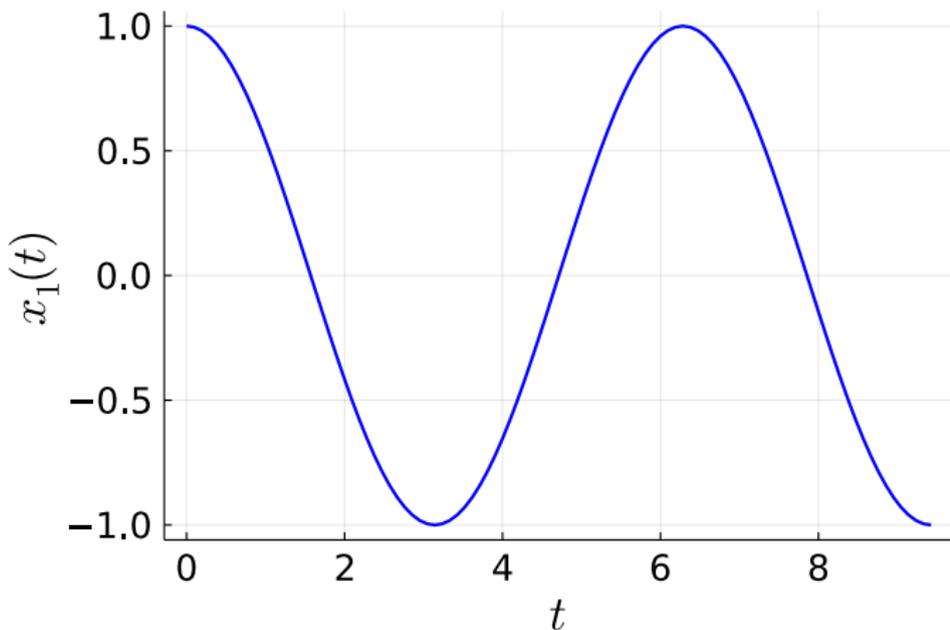


Problems with simulation (1): Precision



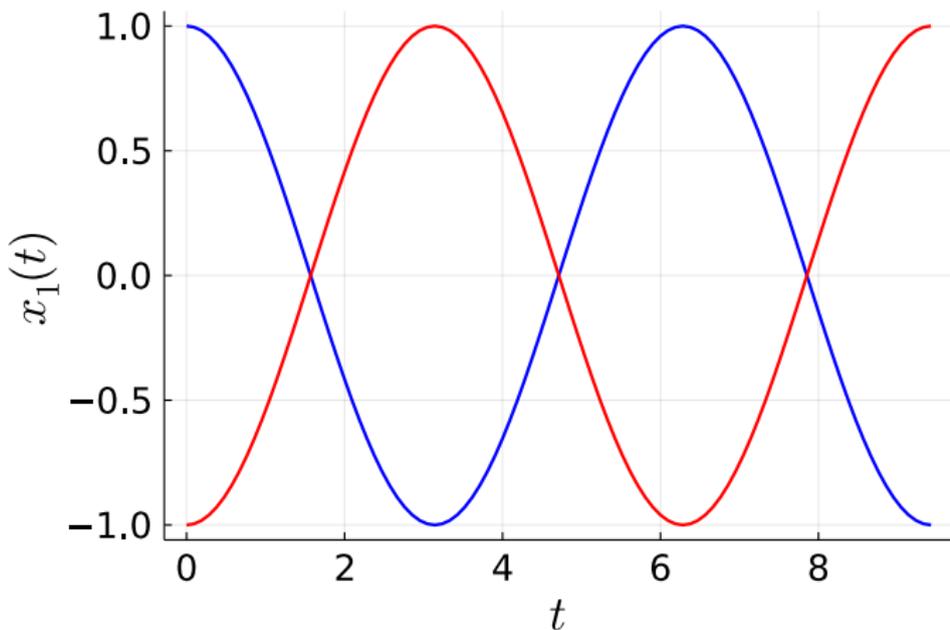
- Approaches: **reduce the time step**, **adaptive solvers**, ...

Problems with simulation (2): Coverage



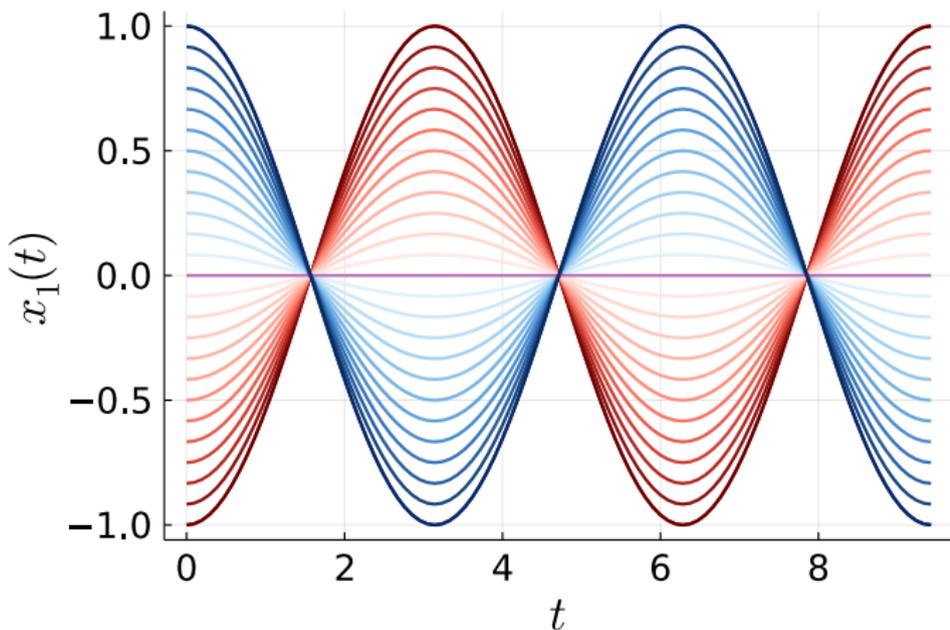
- Consider a **set of initial states** $x_1(0) \in [-1, 1]$

Problems with simulation (2): Coverage



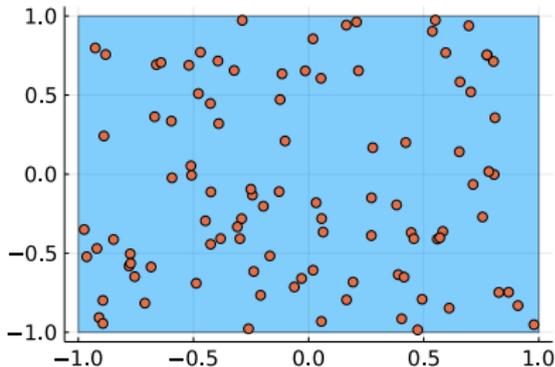
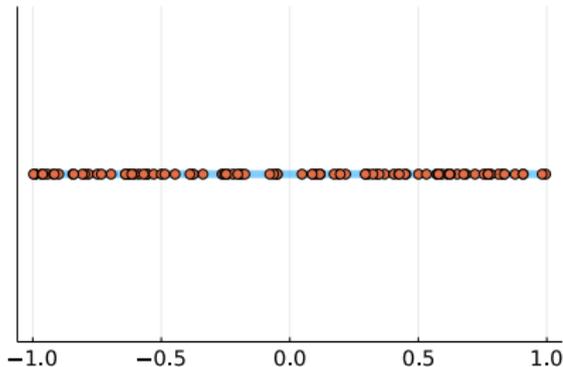
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Problems with simulation (2): Coverage



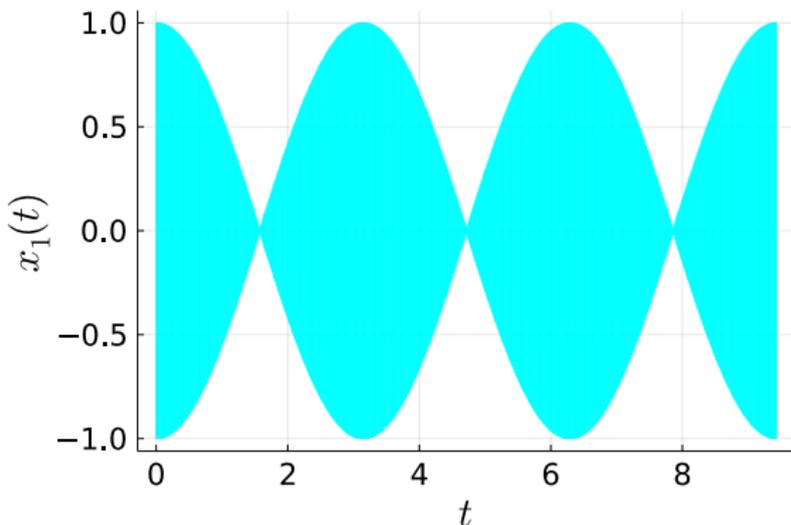
- Approach: **sample the corners** (if the set has corners)
(sufficient for **linear systems** only)

Problems with simulation (3): Dimensionality



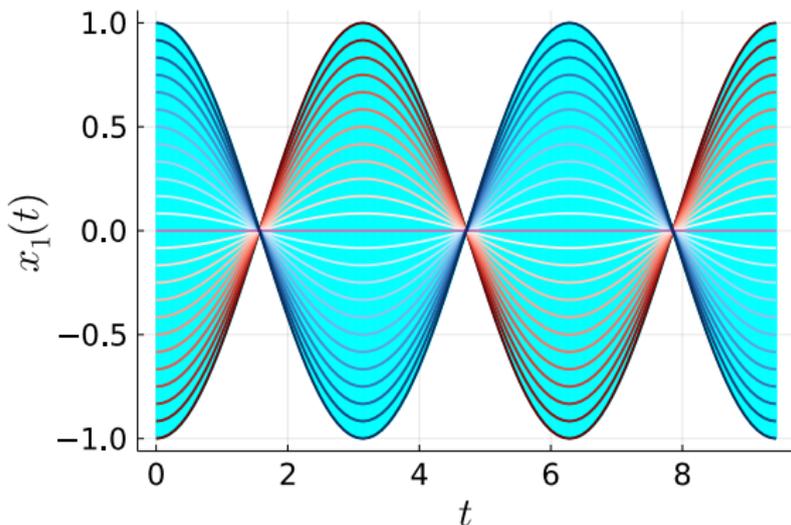
- Sampling coverage is low for higher dimensions
- Vertex sampling: n -dimensional hyperrectangle has 2^n vertices

Reachability analysis



- **Enclose** the **reachable states** $\{x(t) : x(0) \in \mathcal{X}_0, t \geq 0\}$
- **Set-based simulations** (same intuition)
- **Rigorous proof method** (captures **all solutions**)
- **Fast** (linear systems with thousands of dimensions)

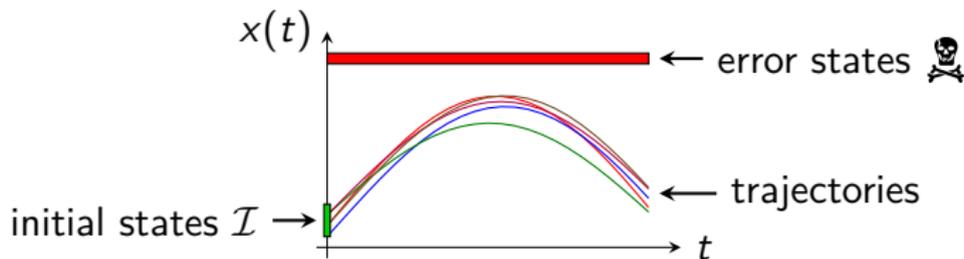
Reachability analysis



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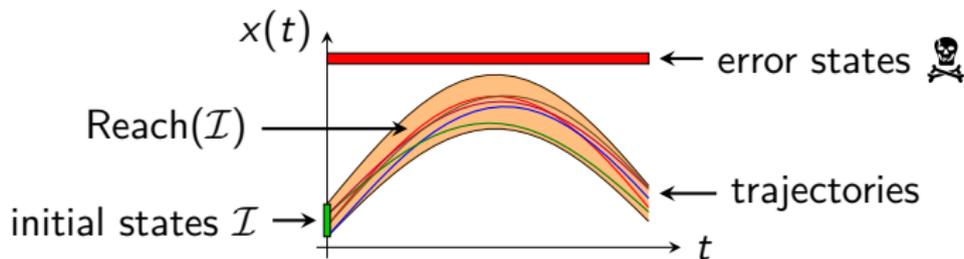
Safety verification

- Task: Verify that **no trajectory** leads to an **error state**



Safety verification

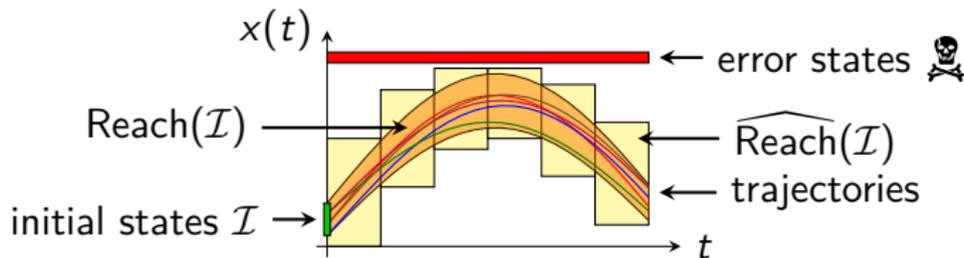
- Task: Verify that **no trajectory** leads to an **error state**
- Equivalent to showing $\text{Reach}(\mathcal{I}) \cap \text{skull} = \emptyset$
- Only **decidable** under strong restrictions



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- Showing $\widehat{\text{Reach}}(\mathcal{I}) \cap \text{skull} = \emptyset$ is sufficient

↑
overapproximation of $\text{Reach}(\mathcal{I})$

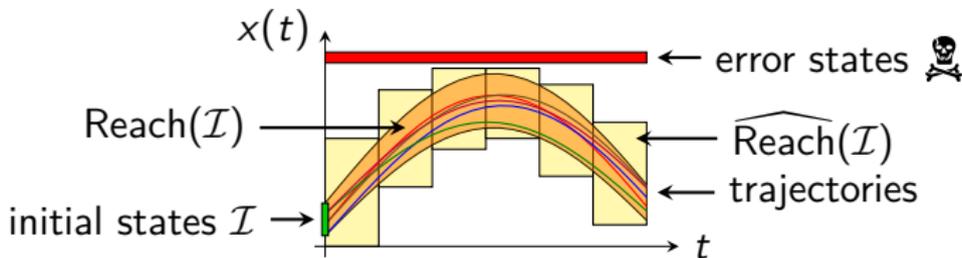


Safety verification

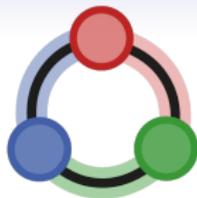
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↑
overapproximation of $\text{Reach}(\mathcal{I})$

restriction:
bounded time



JuliaReach¹



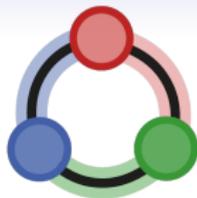
- Open-source **reachability toolbox**
<https://github.com/JuliaReach>
- Joint work with **Marcelo Forets** and many others
- Won ARCH-COMP friendly competition 2018 and 2020

Linear systems (times in seconds)

tool	BLDC01	CBF01	PLAD04-42	BRKDC01
dimension	48	200	9	4
CORA	2.9	30	1.4	12
HyDRA	0.426	—	1.83	—
JuliaReach	0.0096	12	0.031	0.82
SpaceX	1.6	319	0.36	21

¹S. Bogomolov, M. Forets, G. Frehse, K. Potomkin, and C. Schilling. *HSCC*. 2019.

JuliaReach¹



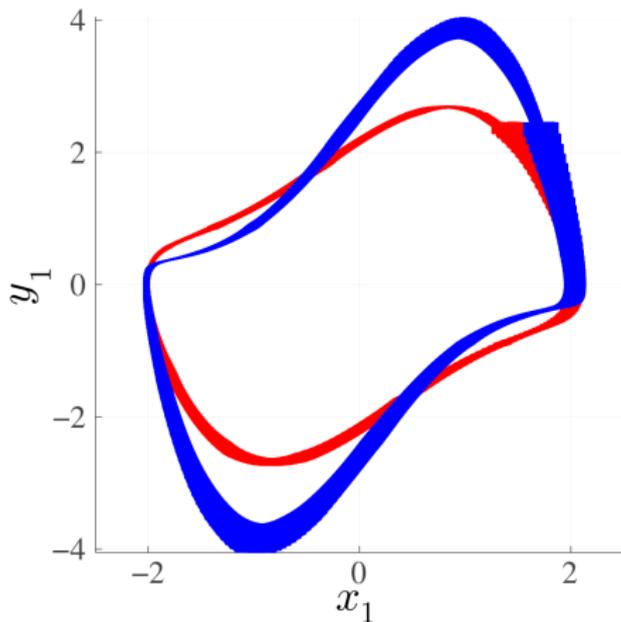
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Nonlinear systems (times in seconds)

tool	CVDP20	LALO20-W0.1	LOVO21	SPRE21
dimension	4	7	2	4
Ariadne	11	31	8	—
CORA	7.7	38	23	26
Dynlbex	510	1,851	75	144
JuliaReach	1.5	6.4	3.4	24

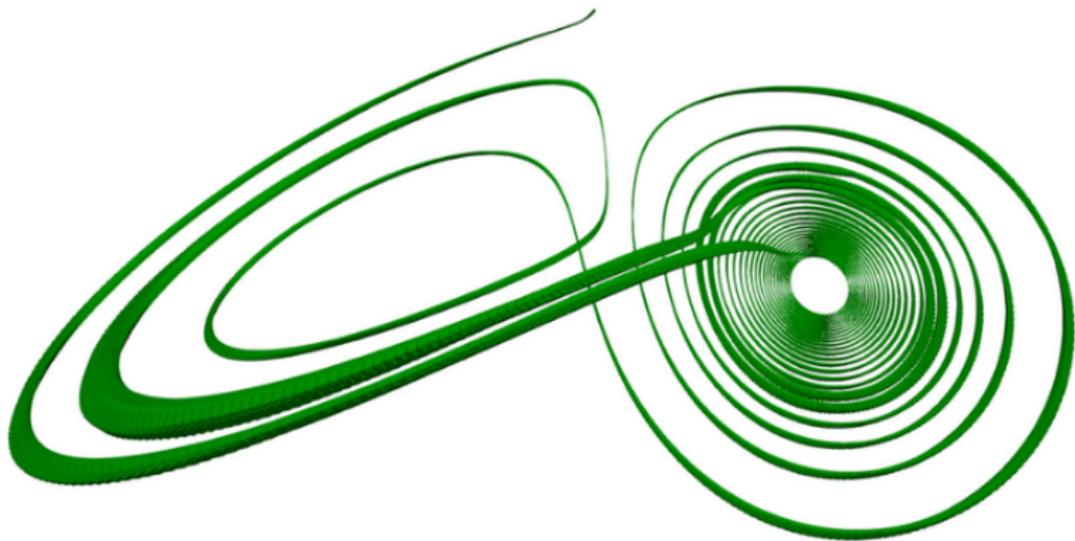
¹S. Bogomolov, M. Forets, G. Frehse, K. Potomkin, and C. Schilling. *HSCC*. 2019.

Examples



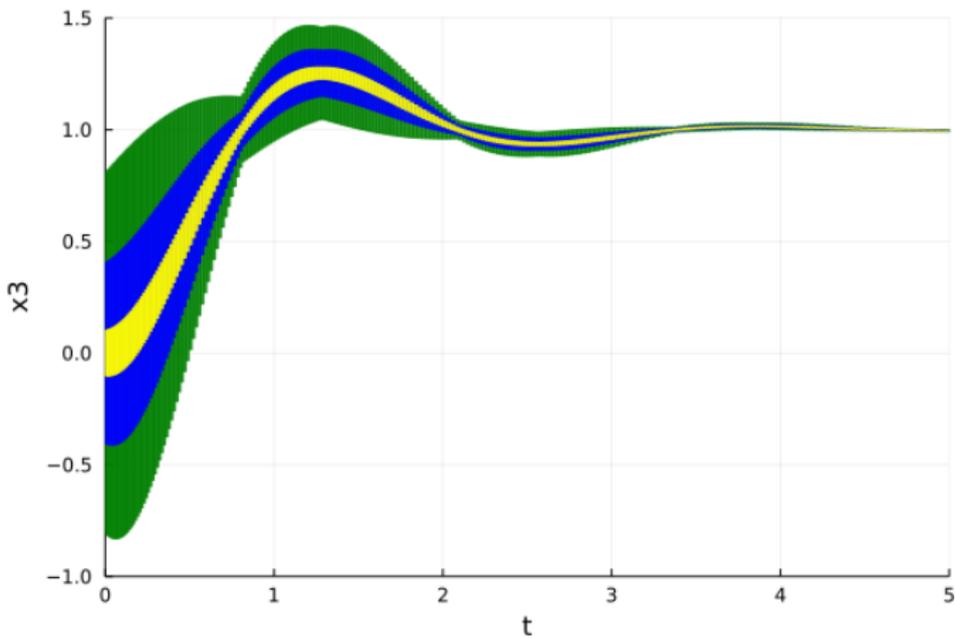
Van der Pol oscillator (limit cycle)

Examples



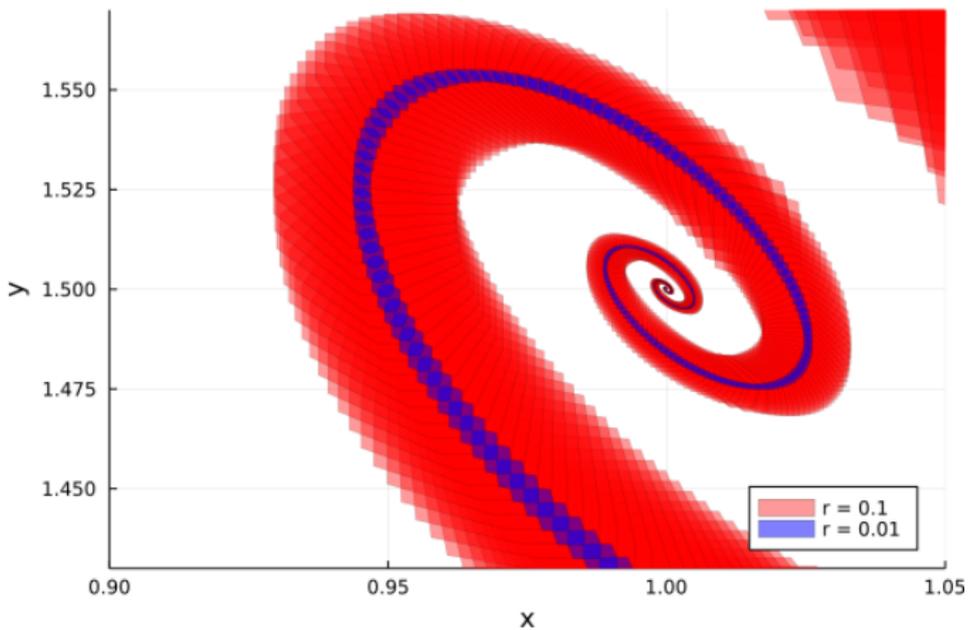
Lorenz system (chaotic behavior)

Examples



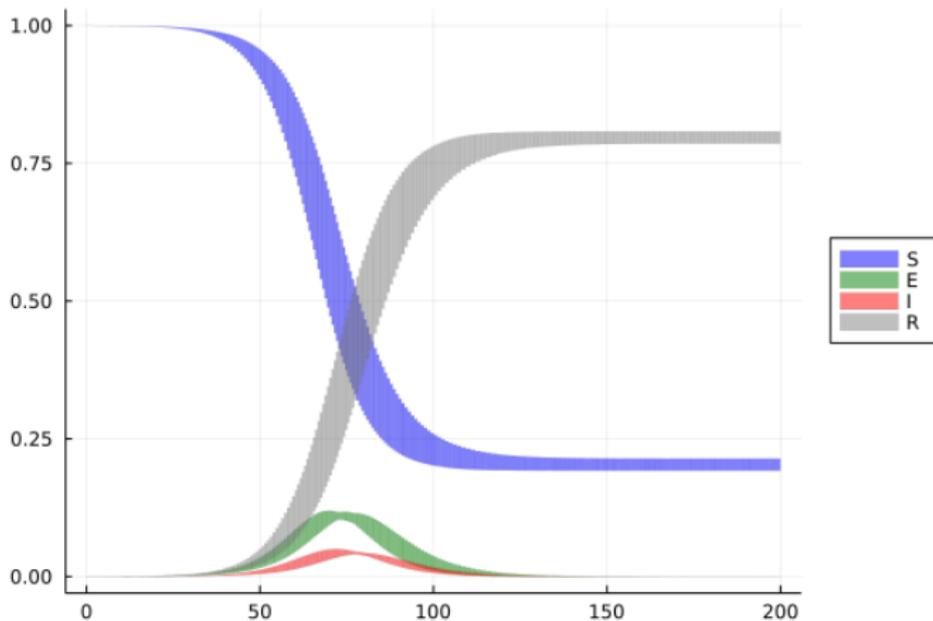
Quadrotor (robotics)

Examples



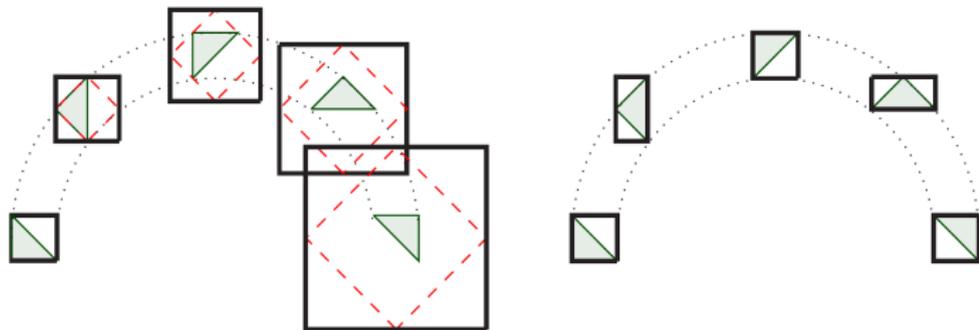
Brusselator (chemical reaction)

Examples



SEIR model (epidemiology)

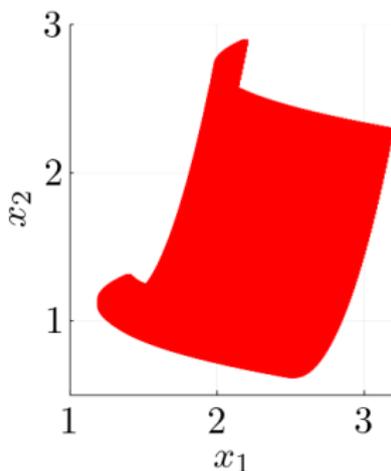
Reachability for nonlinear systems



- **Reachtube construction** computes a **sequence of sets** until a time horizon
- Checking whether a state is reachable is **undecidable**
Hence the true reachable states are **not computable**
- **Overapproximation** or **underapproximation**
- **Wrapping effect**
- Alternative approaches: **invariant generation**, **abstraction**

Taylor models

- **Truncated polynomials with interval remainder**
- **Rigorous arithmetic**



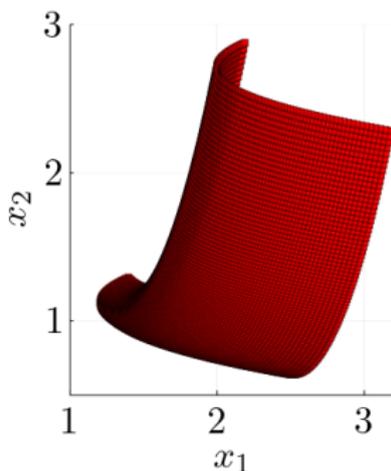
$$p_1(x) = 1.7 - 0.5x_1 + 0.4x_2 + 0.6x_1^2 + [-0.001, 0.001]$$

$$p_2(x) = 1.2 + 0.3x_1 + 0.8x_2 + 0.6x_2^2 + [-0.001, 0.001]$$

$$x \in [-1, 1]^2$$

Taylor models

- **Truncated polynomials with interval remainder**
- **Rigorous arithmetic**

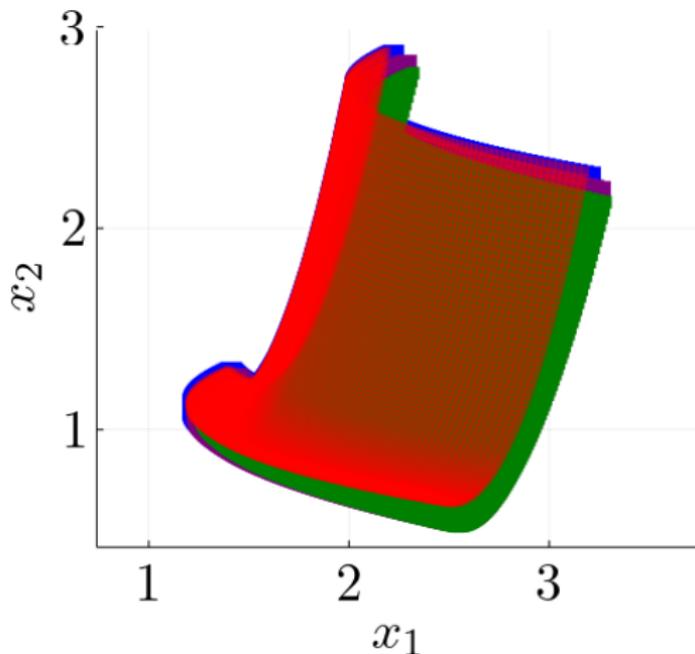


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$$x \in [-1, 1]^2$$

Taylor models for reachability



- Wrap in another **Taylor model over time t**
- **Forward computation** (here: two steps)

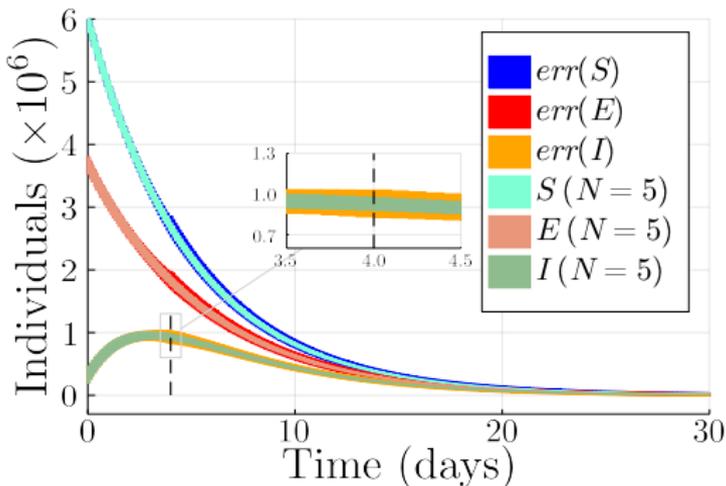
Taylor models for reachability

$$\begin{aligned}
 & 1.7 - 0.5 x_1 + 0.4 x_2 + 0.6 x_1^2 + (1.2 + 0.3 x_1 + 0.8 x_2 + 0.6 x_2^2) t + (\\
 & - 0.85 + 0.25 x_1 - 0.2 x_2 - 0.3 x_1^2) t^2 + (- 0.19999999999999998 - \\
 & 0.049999999999999996 x_1 - 0.13333333333333333 x_2 - 0.09999999999999999 x_2^2) \\
 & t^3 + (0.07083333333333333 - 0.020833333333333332 x_1 + 0.016666666666666666 \\
 & x_2 + 0.024999999999999998 x_1^2) t^4 + (0.009999999999999998 + \\
 & 0.0024999999999999996 x_1 + 0.0066666666666666666 x_2 + 0.004999999999999999 \\
 & x_2^2) t^5 + (- 0.0023611111111111111 + 0.00069444444444444445 x_1 - \\
 & 0.00055555555555555556 x_2 - 0.0008333333333333332 x_1^2) t^6 + (- \\
 & 0.00023809523809523804 - 5.952380952380951e-5 x_1 - 0.00015873015873 \\
 & x_2 - 0.00011904761904761902 x_2^2) t^7 + (4.2162698412698416e-5 - \\
 & 1.240079365079365e-5 x_1 + 9.92063492063492e-6 x_2 + 1.4880952380952378e-5 \\
 & x_1^2) t^8 + [-1.00001e-10, 1.00001e-10]
 \end{aligned}$$

$$\begin{aligned}
 & 1.2 + 0.3 x_1 + 0.8 x_2 + 0.6 x_2^2 + (- 1.7 + 0.5 x_1 - 0.4 x_2 - 0.6 x_1^2) t + \\
 & (- 0.6 - 0.15 x_1 - 0.4 x_2 - 0.3 x_2^2) t^2 + (0.28333333333333333 - \\
 & 0.083333333333333333 x_1 + 0.06666666666666667 x_2 + 0.09999999999999999 x_1^2) \\
 & t^3 + (0.049999999999999996 + 0.012499999999999999 x_1 + 0.03333333333333333 \\
 & x_2 + 0.024999999999999998 x_2^2) t^4 + (- 0.014166666666666666 + \\
 & 0.0041666666666666666 x_1 - 0.0033333333333333333 x_2 - 0.004999999999999999 \\
 & x_1^2) t^5 + (- 0.0016666666666666663 - 0.0004166666666666666 x_1 - \\
 & 0.0011111111111111111 x_2 - 0.0008333333333333332 x_2^2) t^6 + (\\
 & 0.00033730158730158733 - 9.92063492063492e-5 x_1 + 7.936507936507937e-5 x_2 + \\
 & 0.00011904761904761902 x_1^2) t^7 + (2.9761904761904755e-5 + \\
 & 7.440476190476189e-6 x_1 + 1.984126984126984e-5 x_2 + 1.4880952380952378e-5 \\
 & x_2^2) t^8 + [-1.00001e-10, 1.00001e-10]
 \end{aligned}$$

- The first set (blue) in time interval **[0, 0.225497]**

Linearization

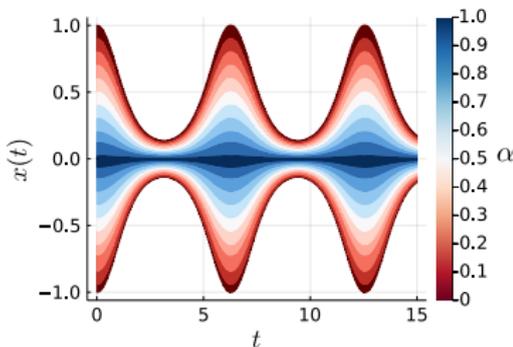


- **Carleman linearization** turns a **polynomial system** into an **infinite-dimensional linear system**
- **Truncation** leads to an **approximate system**
- Can **bound the approximation error** for **dissipative, weakly-nonlinear** systems \rightsquigarrow **reachability algorithm**¹

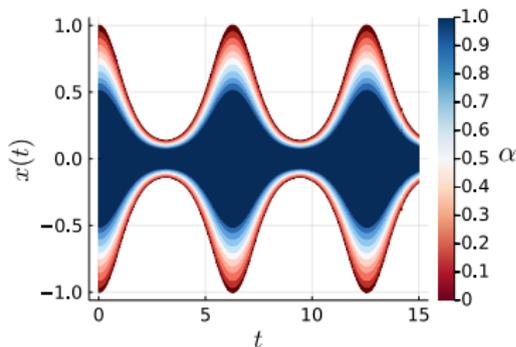
¹M. Forets and C. Schilling. *RP*. 2021.

Work in progress: Probabilistic initial conditions

Harmonic oscillator



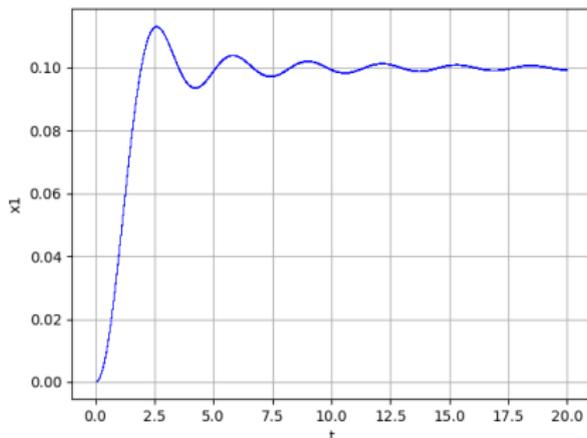
$$x_0 \sim U(-1, 1)$$



$$x_0 \sim U([-1, 0], [0, 1])$$

- Propagate **p-boxes** through **Taylor models**

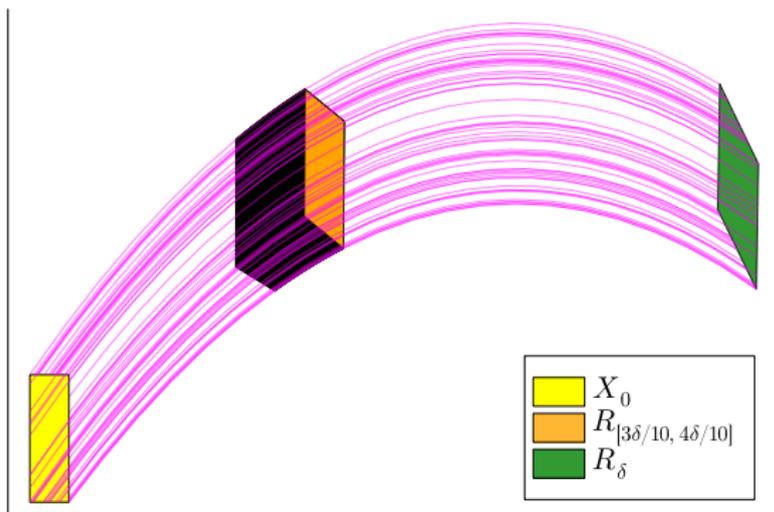
Linear systems



MNA5 (10,913 dimensions)

- **Linear systems:** $\dot{x}(t) = Ax(t)$
- **Reachability problem** still **not decidable**
- **Arbitrary precision** and **fast** (solution above: **90 sec**)

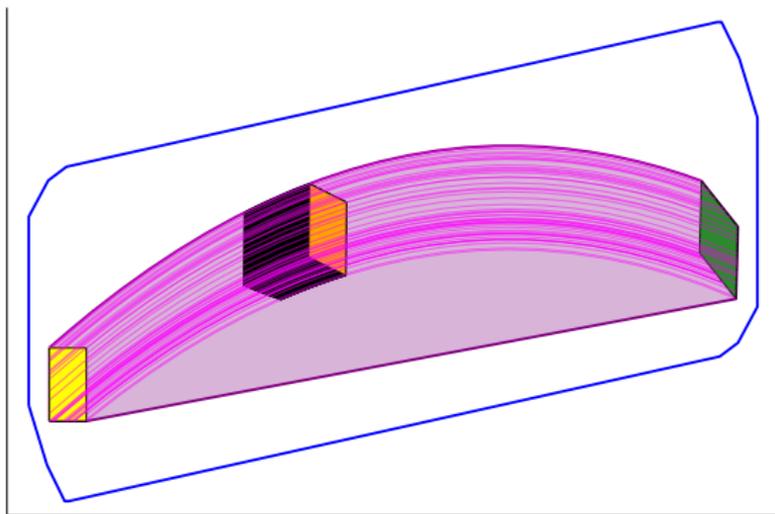
Time discretization



- **Linear systems** allow for **wrapping-free** algorithms based on **efficient set representations**¹

¹M. Forets and C. Schilling. *iFM*. 2022.

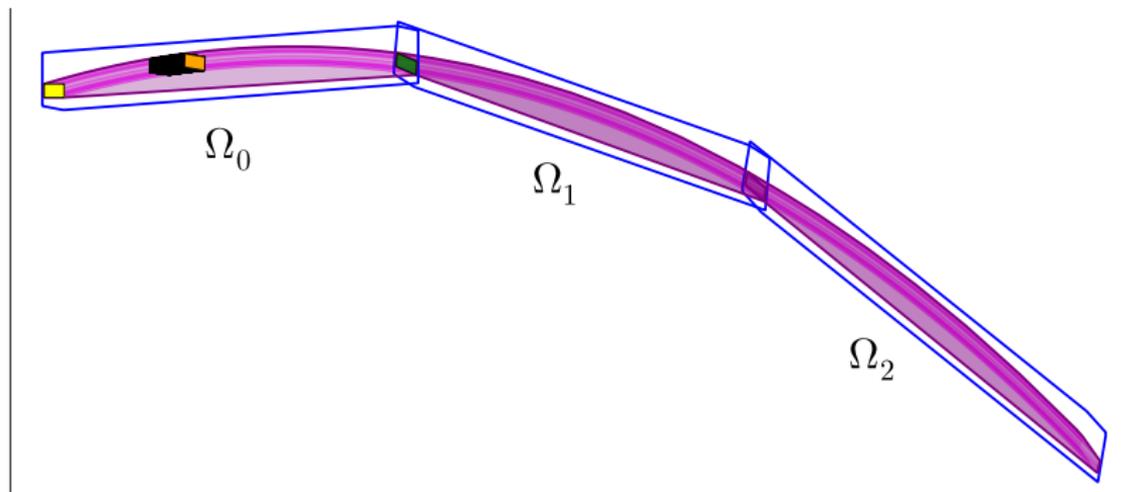
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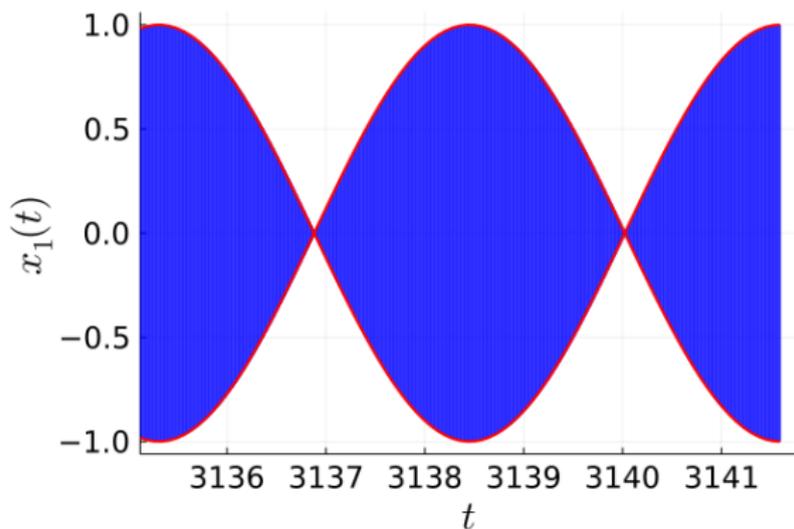
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Wrapping-free computation



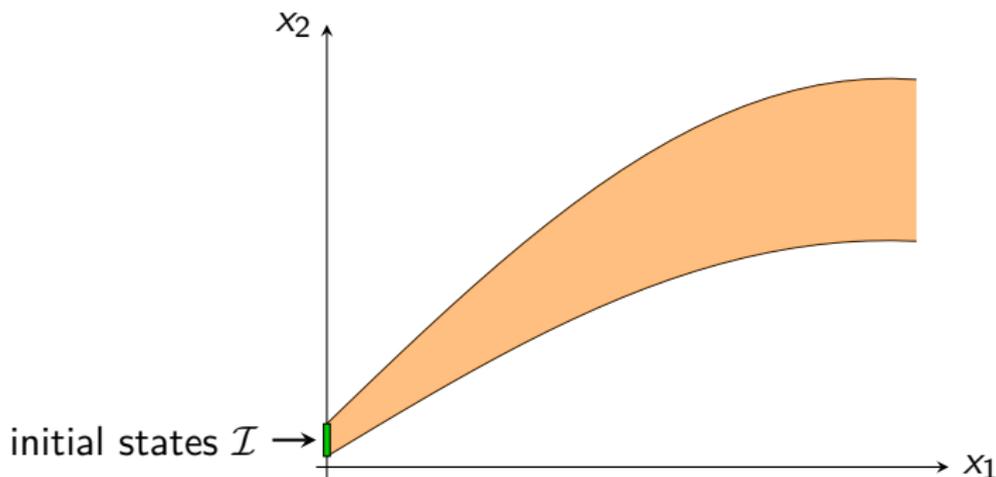
Harmonic oscillator after 500 periods

Time step: **0.01** \rightsquigarrow **314,160** steps

Computation time: **0.33 seconds**

Decomposition approach^{1,2}

Standard algorithm

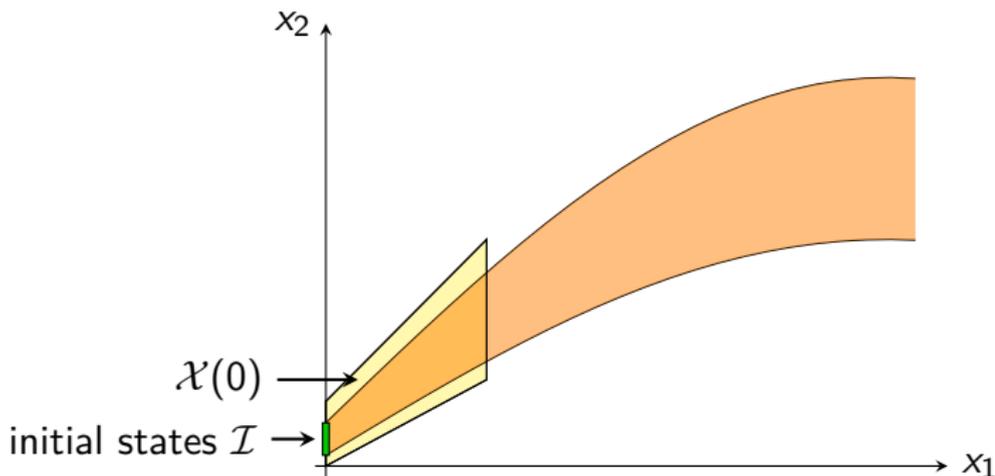


¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

Decomposition approach^{1,2}

Standard algorithm



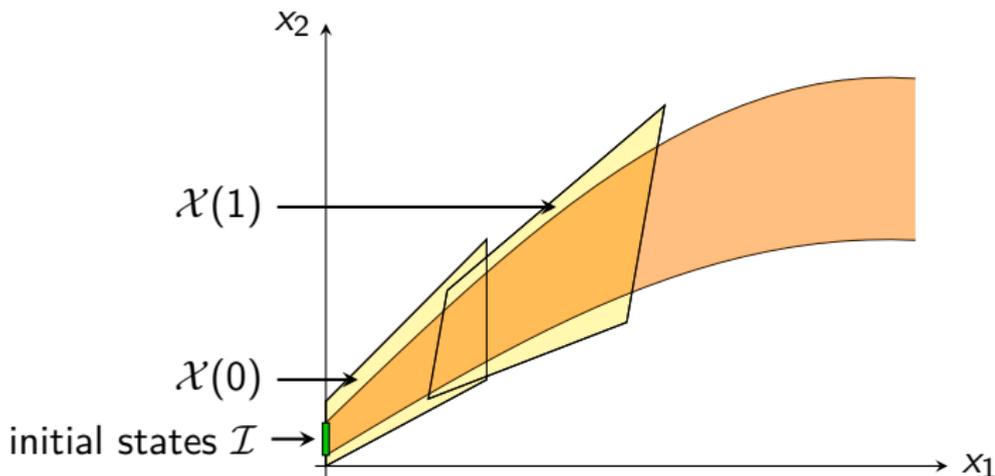
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Decomposition approach^{1,2}

Standard algorithm

$$\mathcal{X}(1) = \Phi \cdot \mathcal{X}(0)$$



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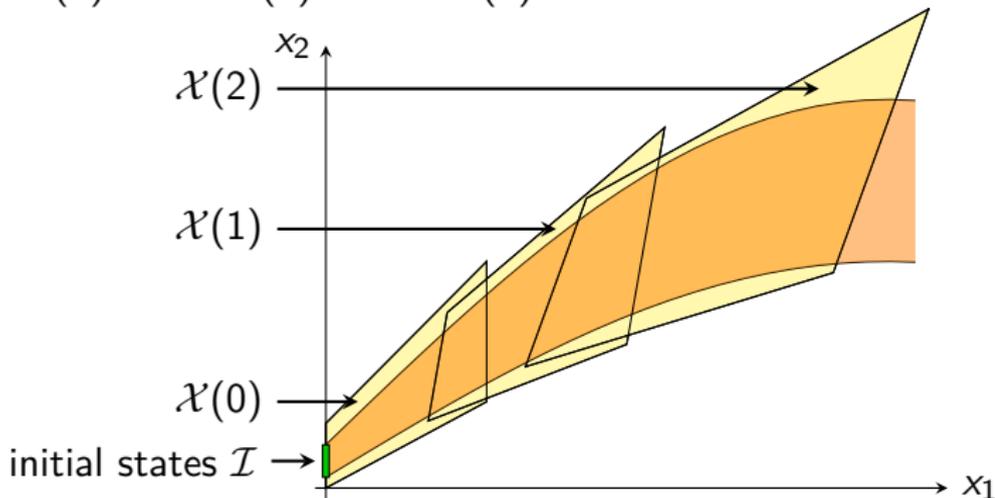
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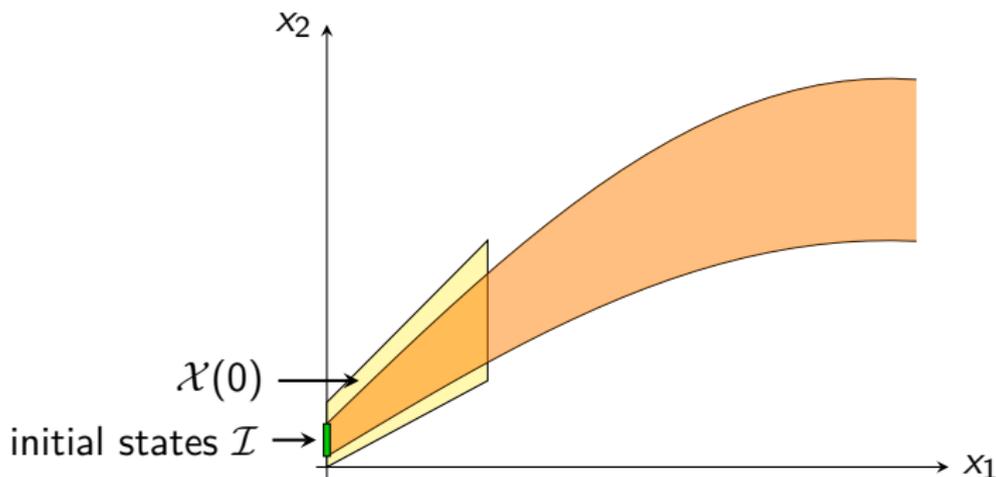
$$\mathcal{X}(2) = \Phi \cdot \mathcal{X}(1) = \Phi^2 \cdot \mathcal{X}(0)$$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

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Decomposition approach^{1,2}



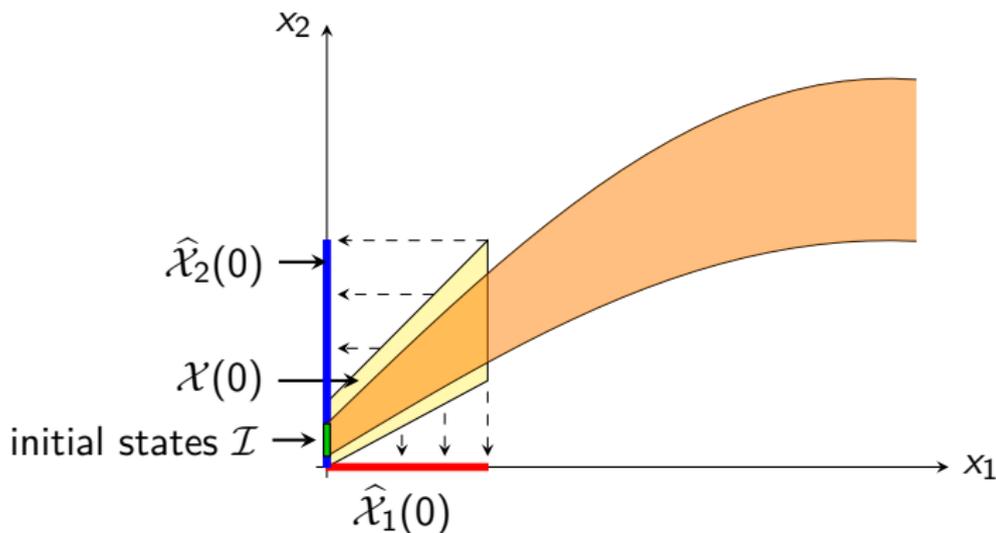
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Decomposition approach^{1,2}

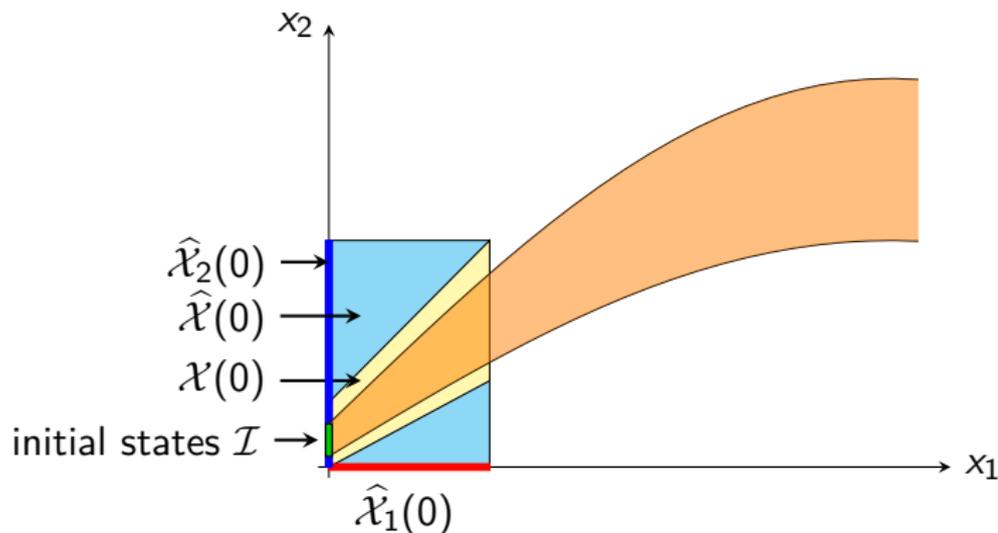
Decompose $\mathcal{X}(0)$ into **low-dimensional** sets $\hat{\mathcal{X}}_1(0)$ and $\hat{\mathcal{X}}_2(0)$

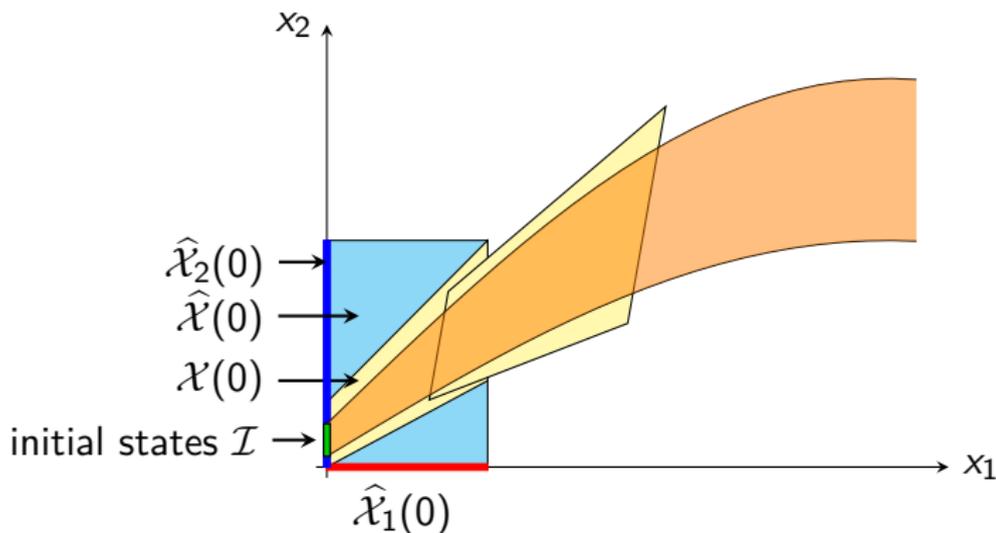
(Note: In general, we do not need to go down to 1D)



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

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Decomposition approach^{1,2}Define $\hat{\mathcal{X}}(k) := \hat{\mathcal{X}}_1(k) \times \hat{\mathcal{X}}_2(k)$ ¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

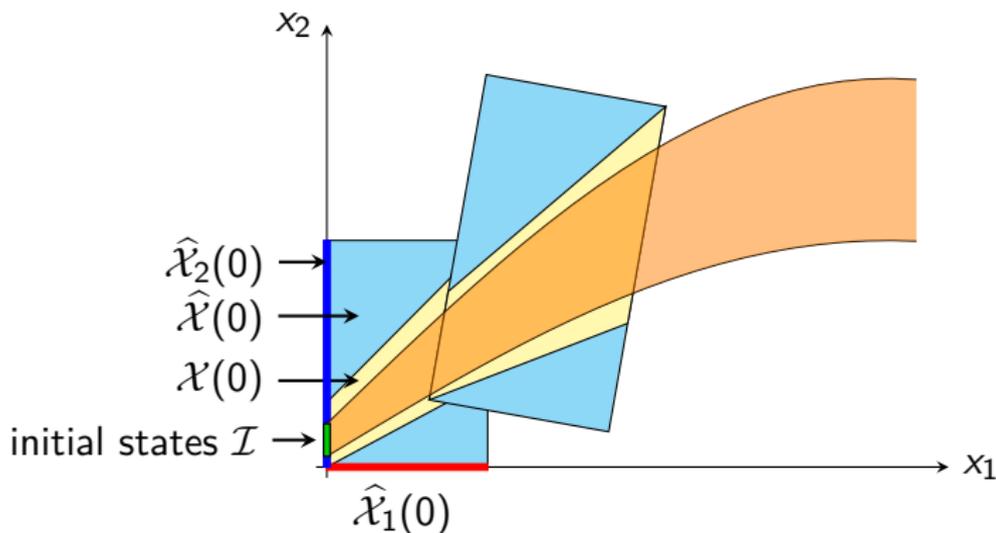
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Decomposition approach^{1,2}

Define $\hat{\mathcal{X}}(k) := \hat{\mathcal{X}}_1(k) \times \hat{\mathcal{X}}_2(k)$

Standard: $\mathcal{X}(k) = \Phi^k \cdot \mathcal{X}(0)$

Decomposed: $\hat{\mathcal{X}}(k) = \Phi^k \cdot \hat{\mathcal{X}}(0) ?$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

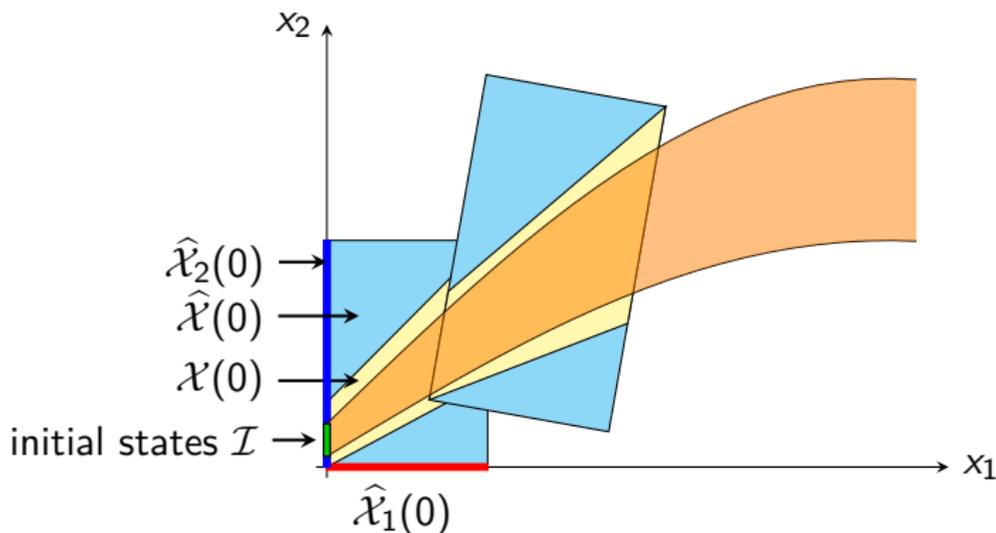
²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

Decomposition approach^{1,2}

Define $\hat{\mathcal{X}}(k) := \hat{\mathcal{X}}_1(k) \times \hat{\mathcal{X}}_2(k)$

Standard: $\mathcal{X}(k) = \Phi^k \cdot \mathcal{X}(0)$

Decomposed: $\hat{\mathcal{X}}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{\mathcal{X}}_j(0)$



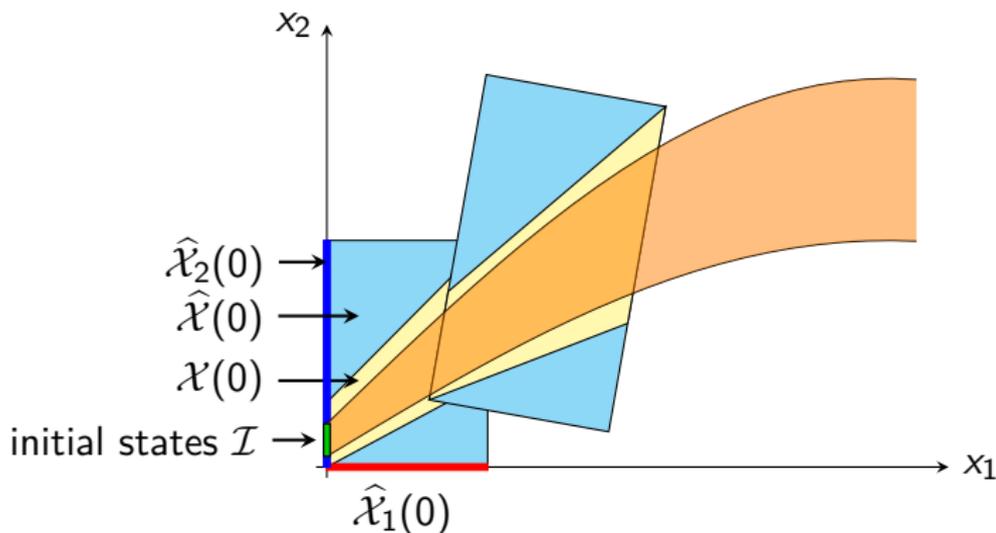
¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

Decomposition approach^{1,2}

$$\hat{\mathcal{X}}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{\mathcal{X}}_j(0)$$

$$\Phi = \left(\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right)$$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

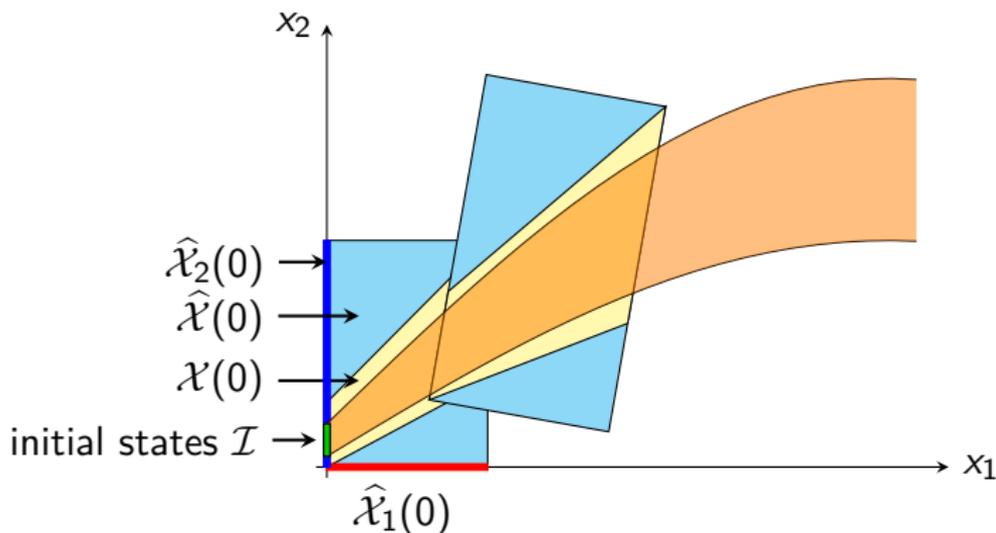
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Decomposition approach^{1,2}

$$\hat{\mathcal{X}}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{\mathcal{X}}_j(0)$$

$$\hat{\mathcal{X}}_1(1) = A \cdot \hat{\mathcal{X}}_1(0)$$

$$\Phi = \left(\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right)$$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

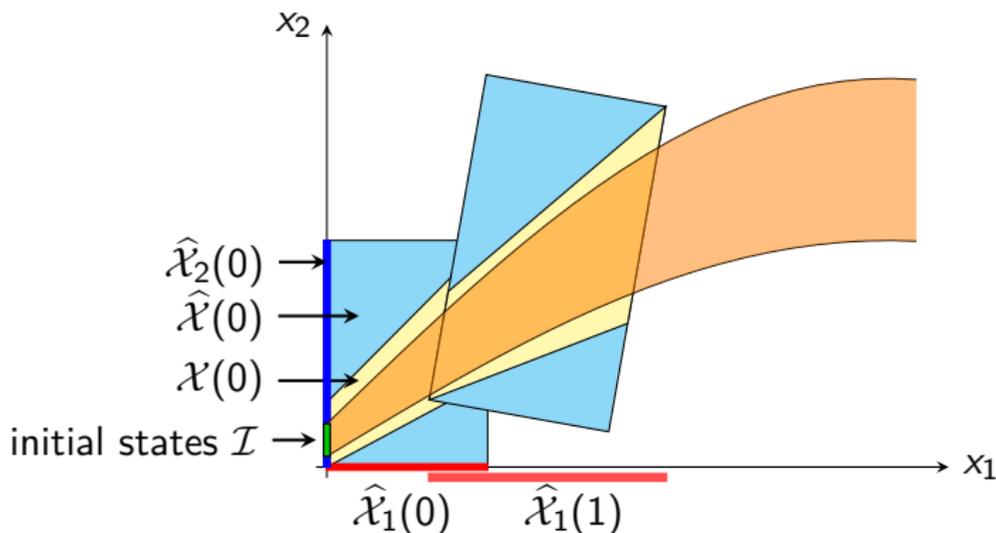
²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

Decomposition approach^{1,2}

$$\hat{\mathcal{X}}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{\mathcal{X}}_j(0)$$

$$\hat{\mathcal{X}}_1(1) = A \cdot \hat{\mathcal{X}}_1(0) \oplus B \cdot \hat{\mathcal{X}}_2(0)$$

$$\Phi = \left(\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right)$$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

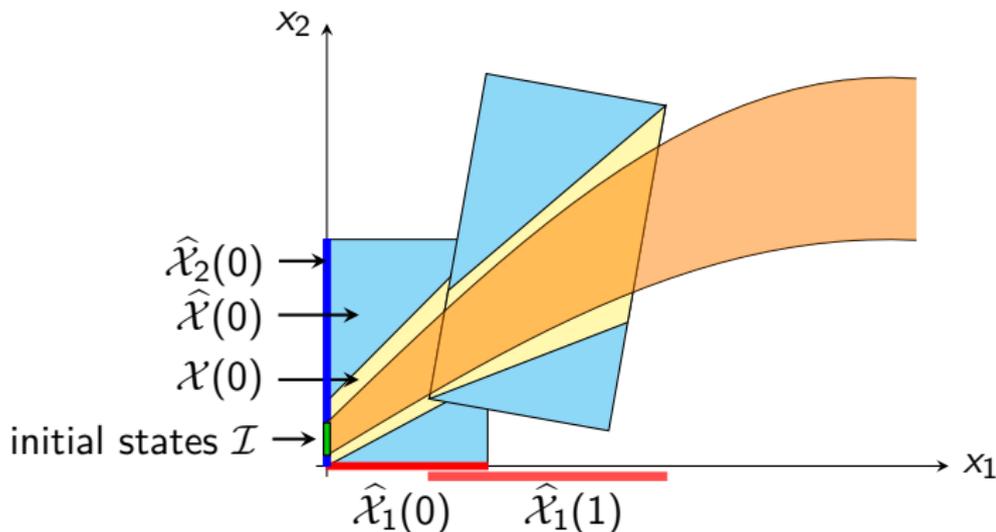
Decomposition approach^{1,2}

$$\hat{\mathcal{X}}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{\mathcal{X}}_j(0)$$

$$\hat{\mathcal{X}}_1(1) = A \cdot \hat{\mathcal{X}}_1(0) \oplus B \cdot \hat{\mathcal{X}}_2(0)$$

$$\hat{\mathcal{X}}_2(1) = C \cdot \hat{\mathcal{X}}_1(0)$$

$$\Phi = \left(\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right)$$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

Decomposition approach^{1,2}

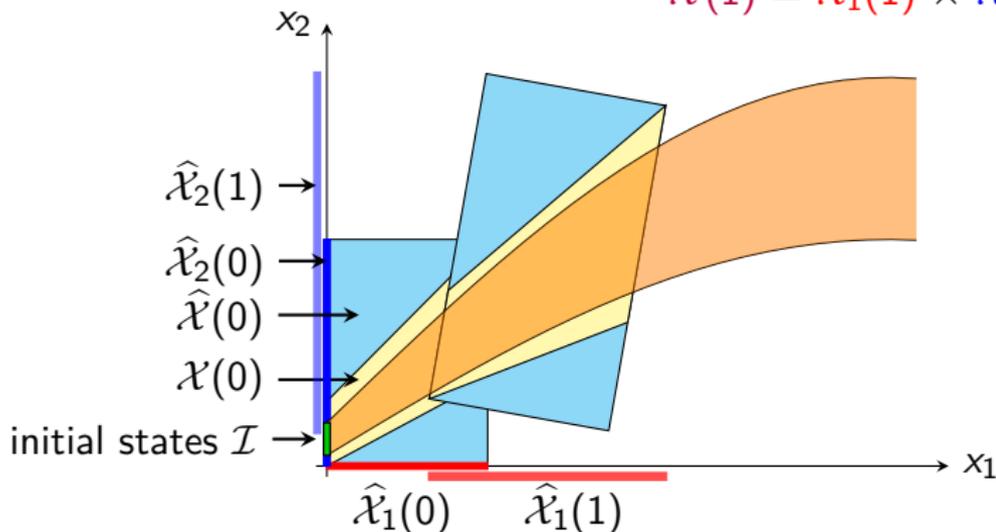
$$\hat{\mathcal{X}}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{\mathcal{X}}_j(0)$$

$$\hat{\mathcal{X}}_1(1) = A \cdot \hat{\mathcal{X}}_1(0) \oplus B \cdot \hat{\mathcal{X}}_2(0)$$

$$\hat{\mathcal{X}}_2(1) = C \cdot \hat{\mathcal{X}}_1(0) \oplus D \cdot \hat{\mathcal{X}}_2(0)$$

$$\Phi = \left(\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right)$$

$$\hat{\mathcal{X}}(1) = \hat{\mathcal{X}}_1(1) \times \hat{\mathcal{X}}_2(1)$$



¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

Decomposition approach^{1,2}

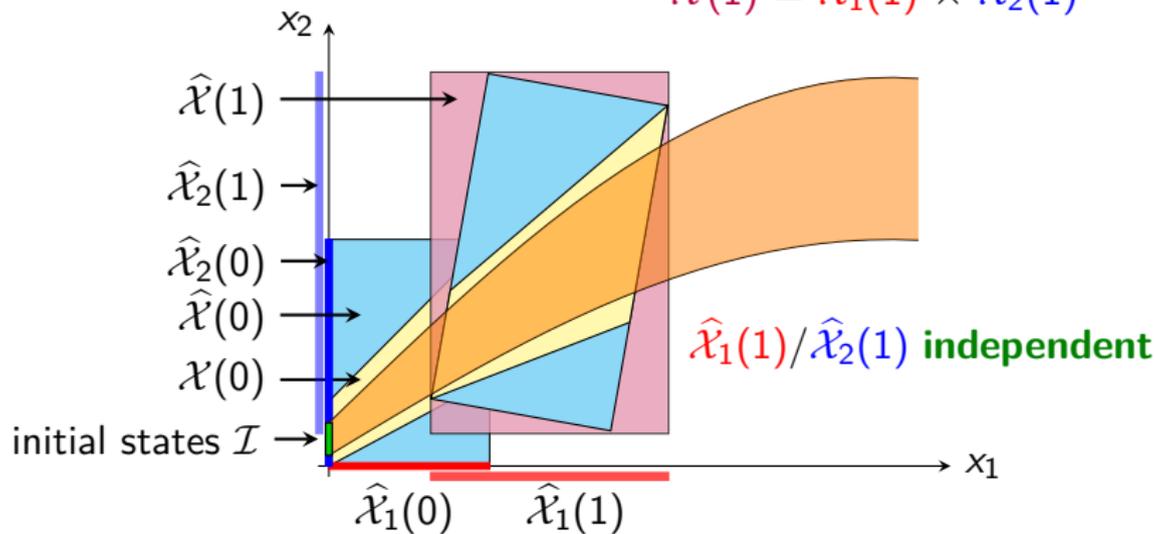
$$\hat{x}_i(k) = \bigoplus_j \Phi_{i,j}^k \cdot \hat{x}_j(0)$$

$$\hat{x}_1(1) = A \cdot \hat{x}_1(0) \oplus B \cdot \hat{x}_2(0)$$

$$\hat{x}_2(1) = C \cdot \hat{x}_1(0) \oplus D \cdot \hat{x}_2(0)$$

$$\Phi = \left(\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right)$$

$$\hat{x}(1) = \hat{x}_1(1) \times \hat{x}_2(1)$$



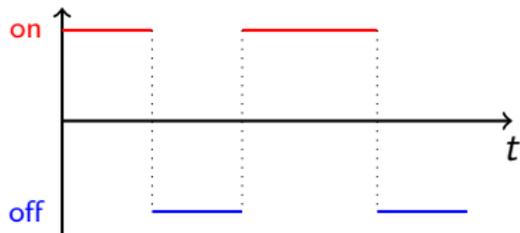
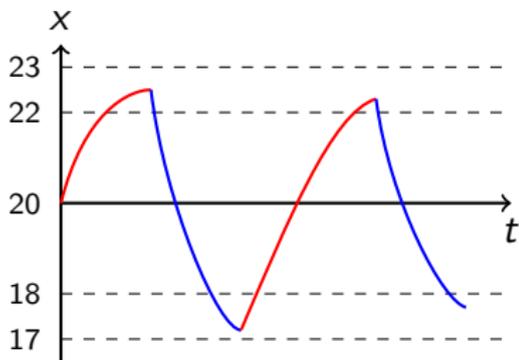
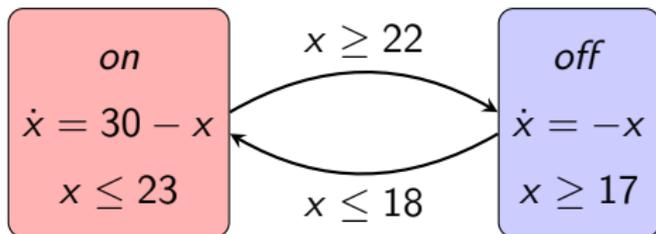
¹S. Bogomolov, M. Forets, G. Frehse, F. Viry, A. Podelski, and C. Schilling. *HSCC*. 2018.

²S. Bogomolov, M. Forets, G. Frehse, A. Podelski, and C. Schilling. *Inf. Comput.* (2022).

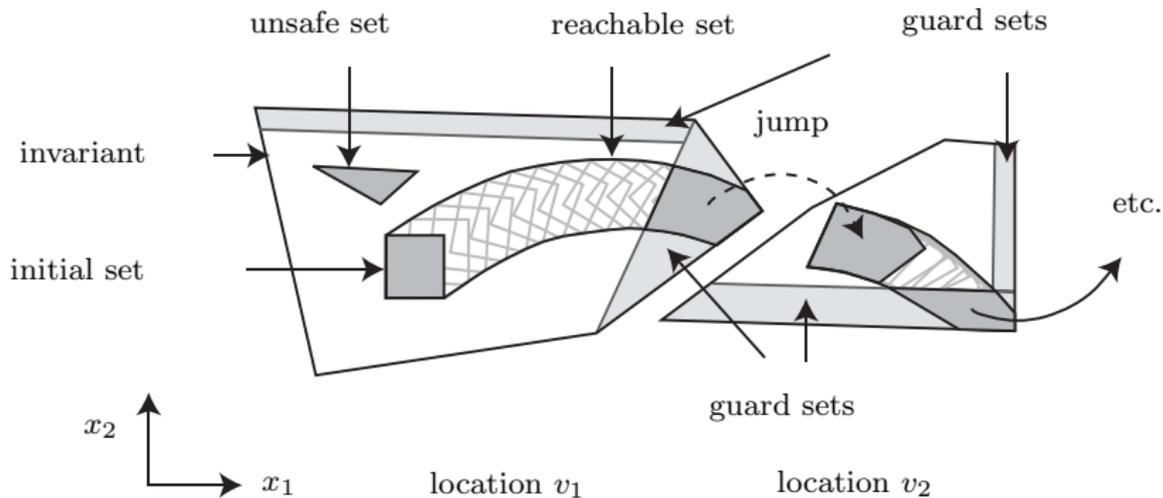
Hybrid and controlled systems

- Combine **continuous** and **discrete** behavior

Example: **Thermostat controller**



Reachability analysis for hybrid systems

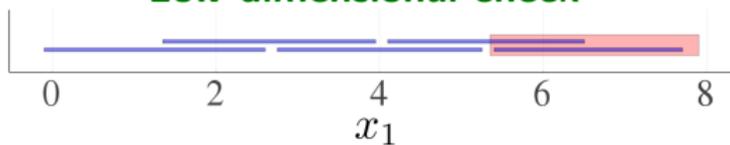


picture taken from¹

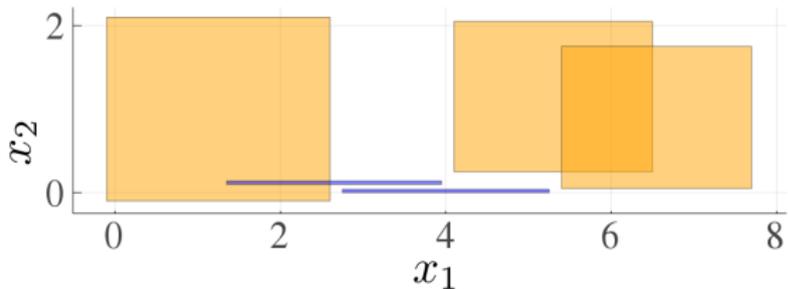
¹M. Althoff. *An Introduction to CORA 2015*. ARCH. 2015.

Decomposition approach¹

Low-dimensional check



High-dimensional sets



- **Check condition** for discrete transition in **low dimensions**
- Only compute **high-dimensional set** when necessary
- Allows to analyze a **1027-dimensional model** in 509 sec

¹S. Bogomolov, M. Forets, G. Frehse, K. Potomkin, and C. Schilling. *IEEE Trans. Comput. Aided Des. Integr. Circuits Syst.* (2020).

Periodic controllers with clock jitter

Electro-mechanical brake

$$\dot{i} = \frac{1}{L} \cdot (K_P \cdot x_e + K_I \cdot x_c) - \frac{1}{L} \left(R + \frac{K^2}{d_{rot}} \right)$$

$$\dot{x} = \frac{K}{i \cdot d_{rot}} \cdot i$$

$$\dot{x}_e = 0$$

$$\dot{x}_c = 0$$

$$i = 1$$

$$t \leq \tau + \iota$$

$$t \geq \tau - \iota$$

$$x'_e := x_0 - x$$

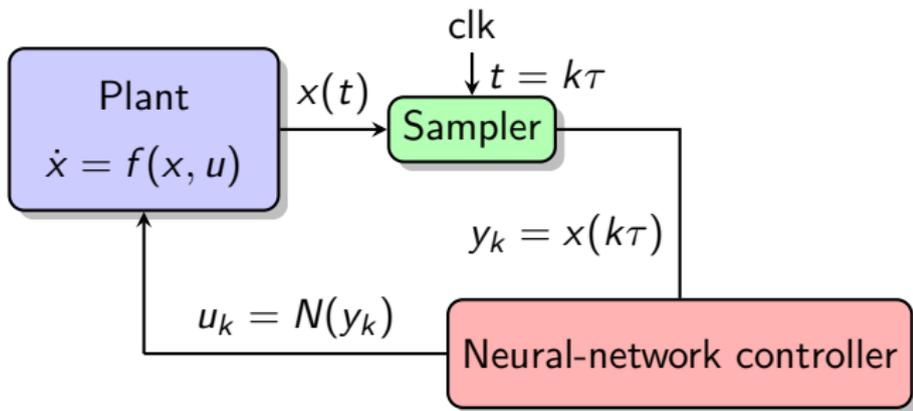
$$x'_c := x_c + \tau \cdot (x_0 - x)$$

$$t' := t - \tau$$

- Analysis for **1,001 transitions: 9 sec**¹ (previous work: **13 h**)

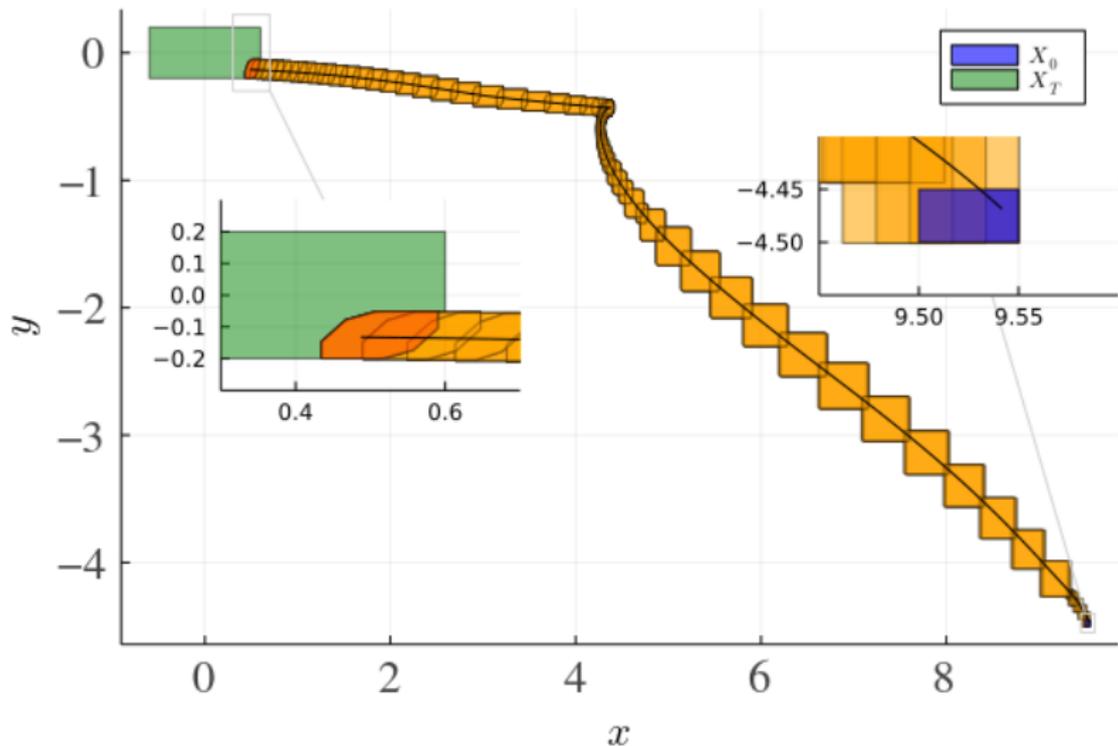
¹M. Forets, D. Freire, and C. Schilling. *MEMOCODE*. 2020.

Neural-network controllers¹



¹C. Schilling, M. Forets, and S. Guadalupe. *AAAI*. 2022.

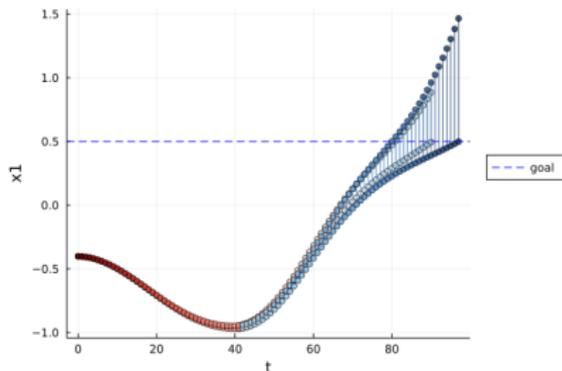
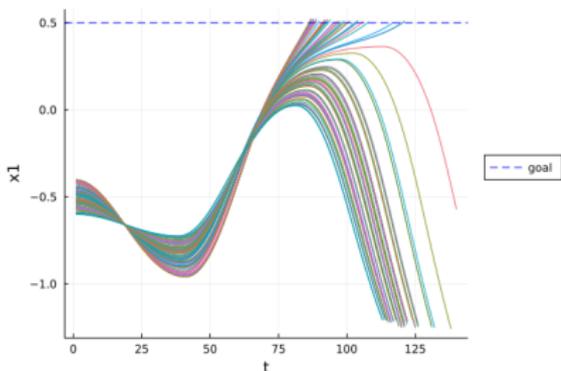
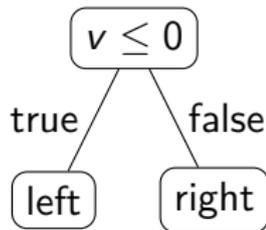
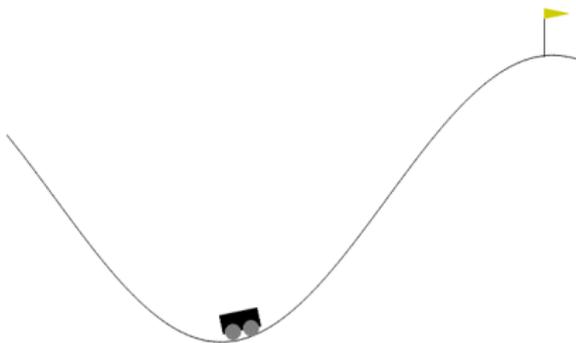
Neural-network controllers¹



¹C. Schilling, M. Forets, and S. Guadalupe. *AAAI*. 2022.

Work in progress: Decision-tree controllers

Mountain car



Summary

- **Reachability analysis** allows to reason about **sets of behaviors**
- Mature for **linear systems** (thousands of dimensions within seconds)
- Still hard for **nonlinear and hybrid systems**
- New challenges in **systems with learned controllers**
- **Exploiting structure** is key
- <https://github.com/JuliaReach>