
Shielding for Hybrid Systems

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joint work with Asger Horn Brorholt,
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AALBORG UNIVERSITET

Overview

Motivation

Approach

Experiments

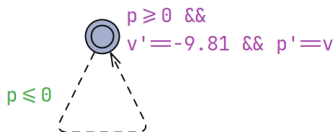
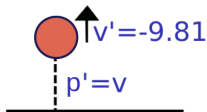
State-space transformation

Multi-agent systems

Conclusion

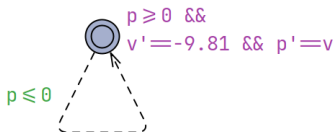
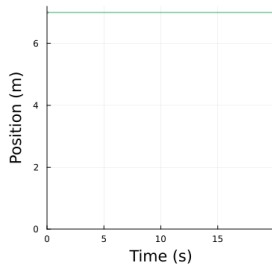
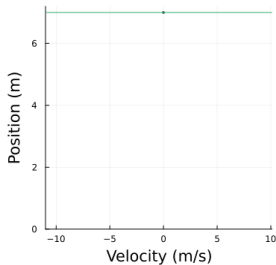
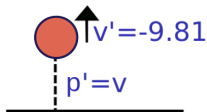
Control of physical systems

- **Physical systems have complex dynamics**
 - Continuous evolution
 - Discrete events
 - Stochastic uncertainty



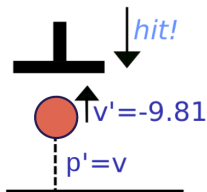
$$v = -(0.85 + \text{random}(0.12)) * v$$

Control of physical systems

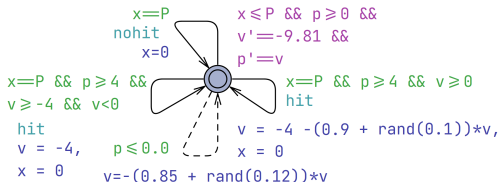


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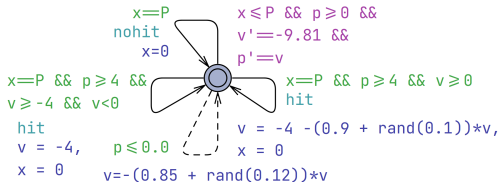
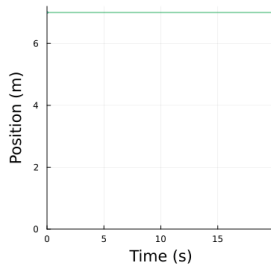
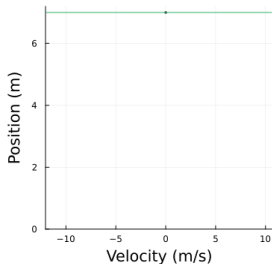
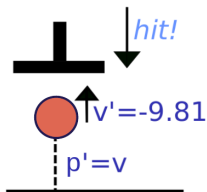
Control of physical systems



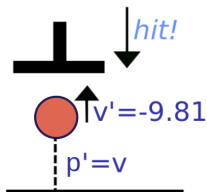
- Physical systems have complex dynamics
 - Continuous evolution
 - Discrete events
 - Stochastic uncertainty
- Goal: control subject to some optimality criterion, e.g., minimize number of hits**



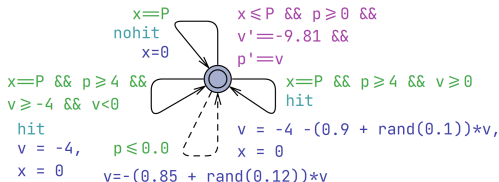
Control of physical systems



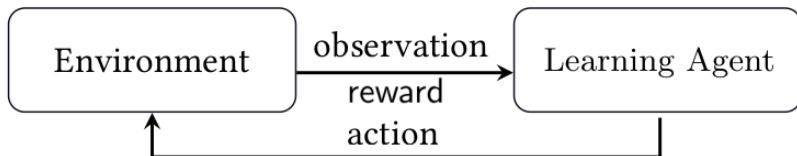
Control of physical systems



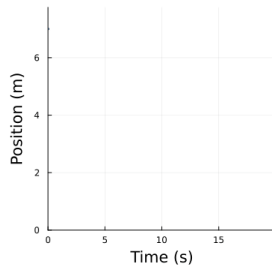
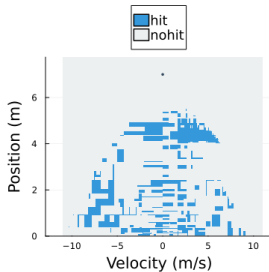
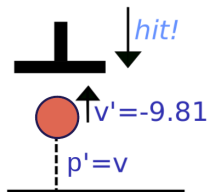
- Physical systems have complex dynamics
 - Continuous evolution
 - Discrete events
 - Stochastic uncertainty
- Goal: control subject to some optimality criterion, e.g., minimize number of hits
- We can reinforcement-learn a controller, e.g., with Uppaal Stratego**



Reinforcement learning

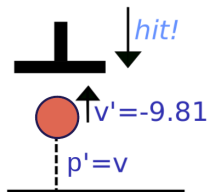


Control of physical systems



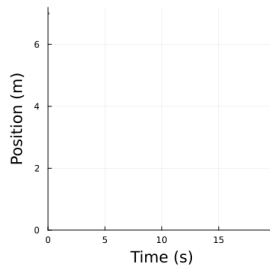
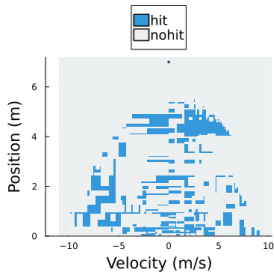
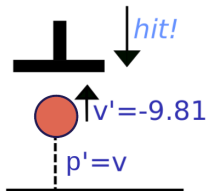
Trained for 12,000 episodes

Control of physical systems



- Physical systems have complex dynamics
 - Continuous evolution
 - Discrete events
 - Stochastic uncertainty
- Goal: control subject to some optimality criterion, e.g., minimize number of hits
- We can reinforcement-learn a controller, e.g., with Uppaal Stratego
- **We also have safety constraints,**
e.g., $p = 0 \implies |v| > 1$

Control of physical systems



2 % of executions unsafe

Overview

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Approach

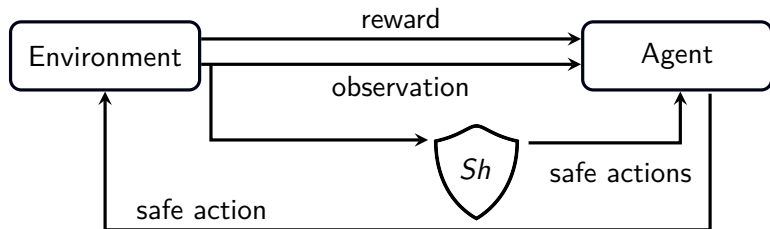
Experiments

State-space transformation

Multi-agent systems

Conclusion

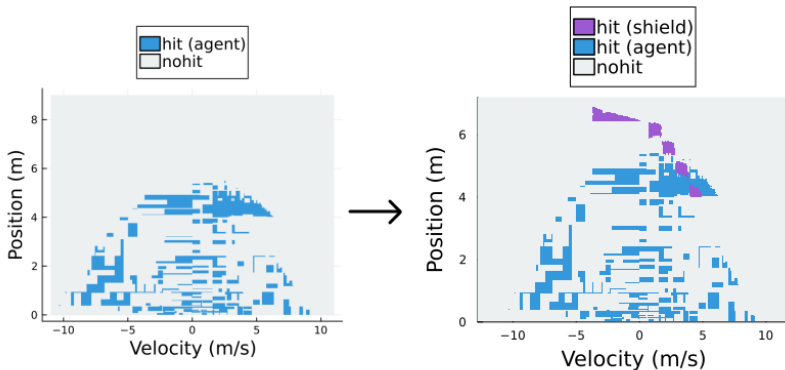
Shielding¹



- Shield = nondeterministic safe policy
- Under a shield, *any* controller/agent is safe

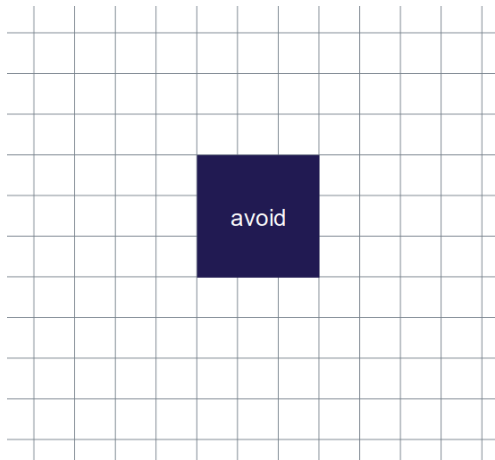
¹Alshiekh et al. *AAAI*. 2018.

Shielding via state-space partitioning¹



¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

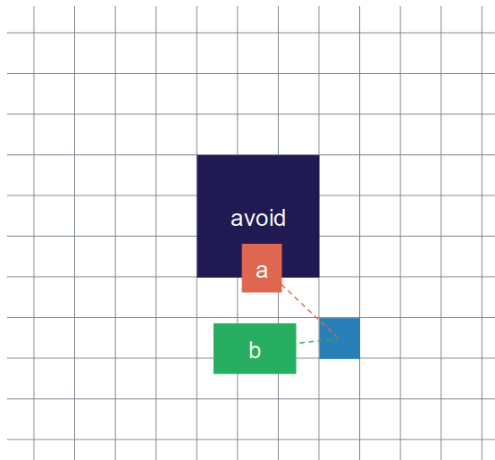
State-space partitioning and two-player game¹



- Avoid set = blocks with unsafe states
- For each block, compute successor states under each action

¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

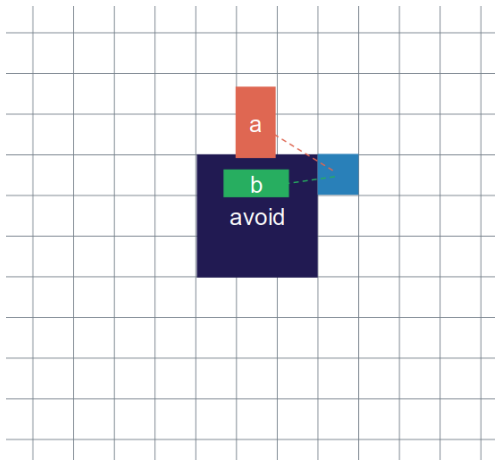
State-space partitioning and two-player game¹



- Avoid set = blocks with unsafe states
- For each block, compute successor states under each action
- Forbid actions that reach the avoid set

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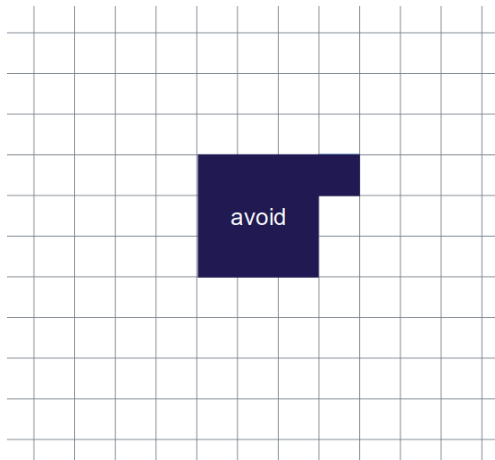
State-space partitioning and two-player game¹



- Avoid set = blocks with unsafe states
- For each block, compute successor states under each action
- Forbid actions that reach the avoid set
- If all actions reach the avoid set,

¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

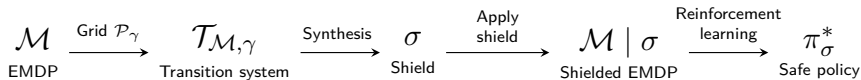
State-space partitioning and two-player game¹



- Avoid set = blocks with unsafe states
- For each block, compute successor states under each action
- Forbid actions that reach the avoid set
- If all actions reach the avoid set, add the block to the avoid set

¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

State-space partitioning and two-player game¹

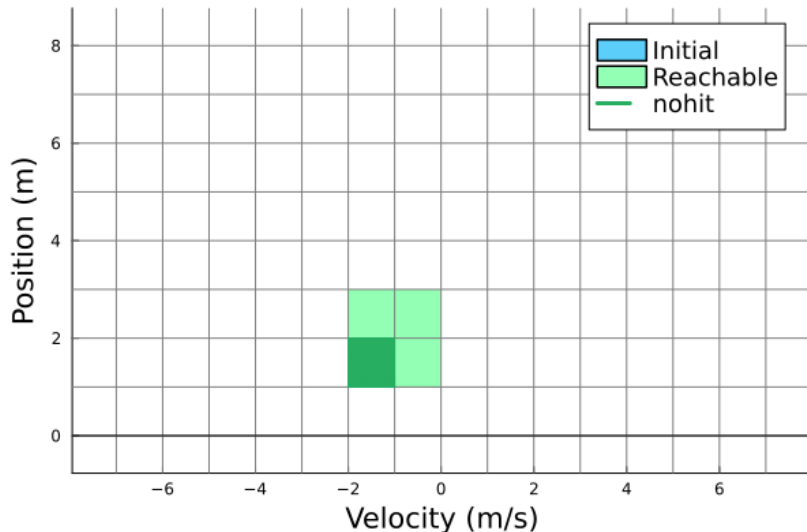


Theorem

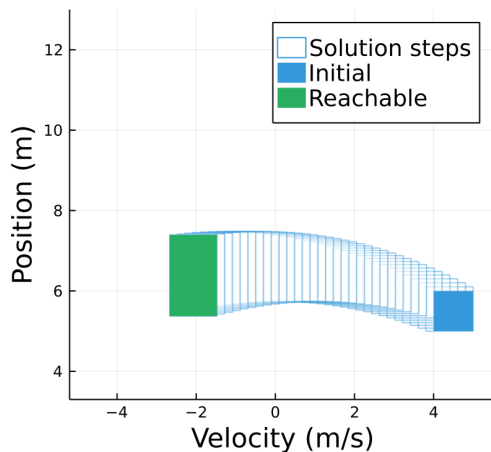
Any shield σ for $\mathcal{T}_{\mathcal{M},\gamma}$ is safe for $\mathcal{M}$

¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

How to compute reachable blocks for complex systems?



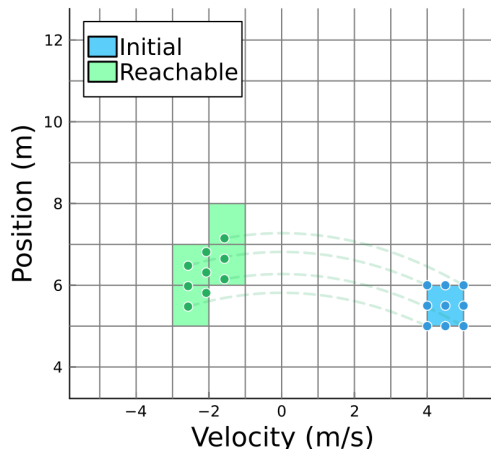
Computation via reachability tool (here: JuliaReach)



- Sound result
- Expensive

Computation via simulation-based method¹

Samples: 9

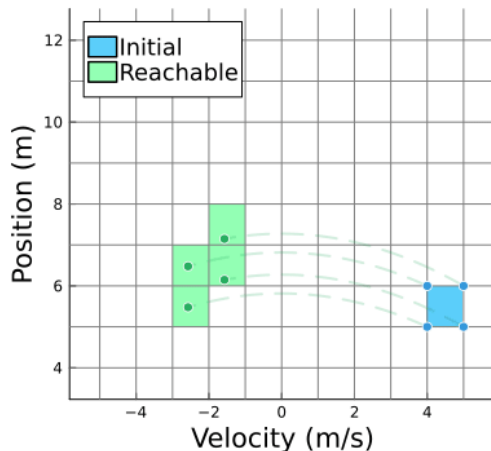


- Cheap
- May miss behavior
- Accuracy improves with more samples

¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

Computation via simulation-based method¹

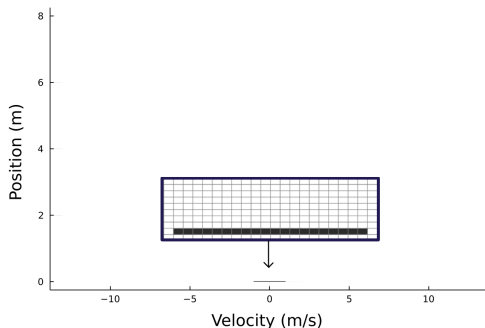
Samples: 4



- Cheap
- May miss behavior
- Accuracy improves with more samples

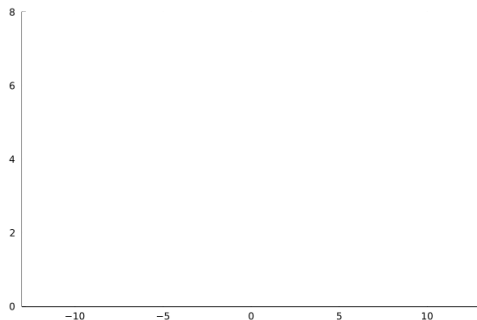
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Synthesis algorithm in action



- 16 samples per block
- Cell diameter 0.02
- 520,000 cells
- Time: 134 sec

Synthesis algorithm in action



- 16 samples per block
- Cell diameter 0.02
- 520,000 cells
- Time: 134 sec

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Scalability

Grid size	Samples	Time
0.02	4	2m 14s
0.02	8	4m 02s
0.02	16	11m 03s
0.01	4	19m 00s
0.01	8	27m 21s
0.01	16	56m 32s

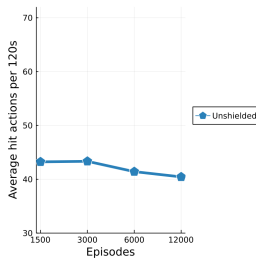
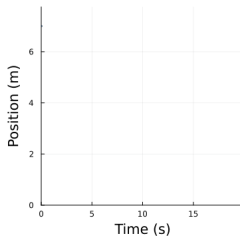
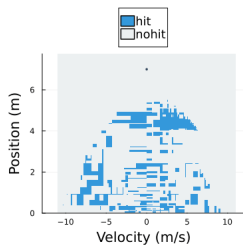
← Sampling-based

- Statistically safe ($\geq 99.99\%$ with confidence 99 %)

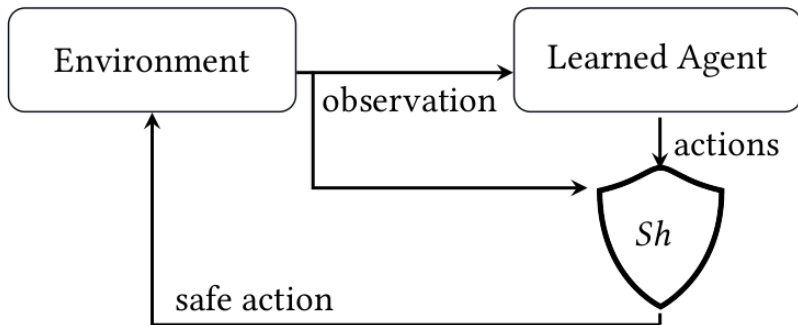
Grid size	Time
0.01	41h 05m

← Reachability-based

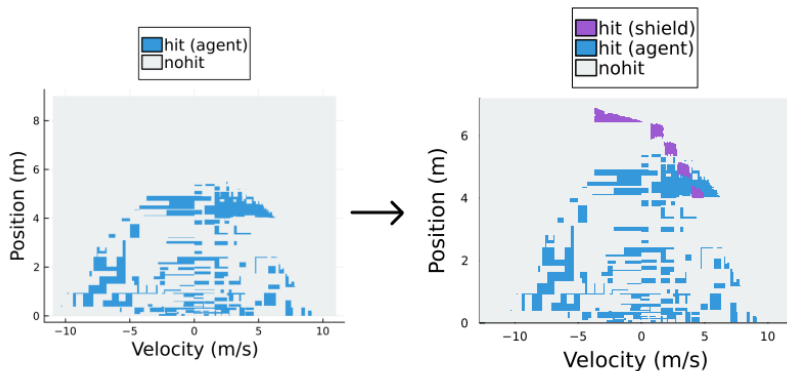
Unshielded agent



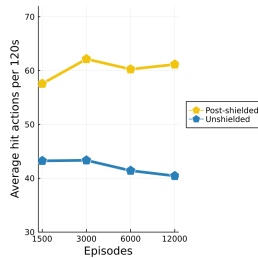
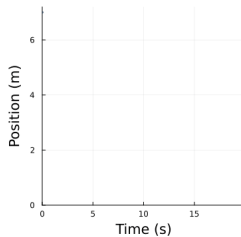
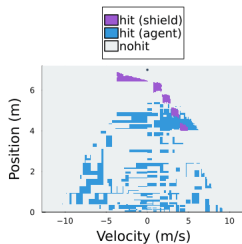
Post-shielded agent



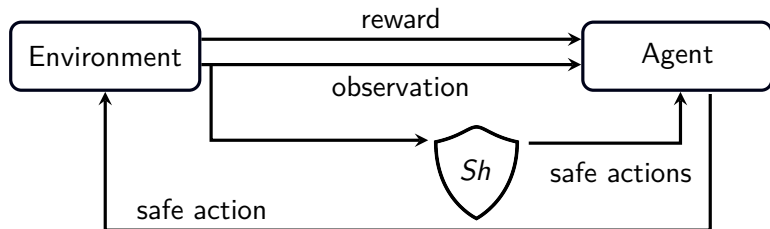
Post-shielded agent



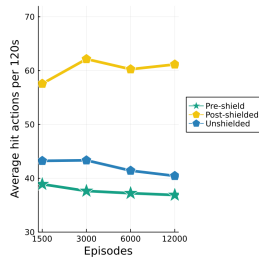
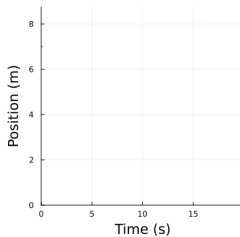
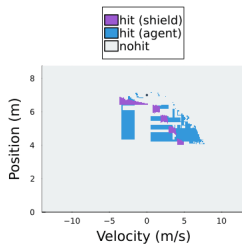
Post-shielded agent



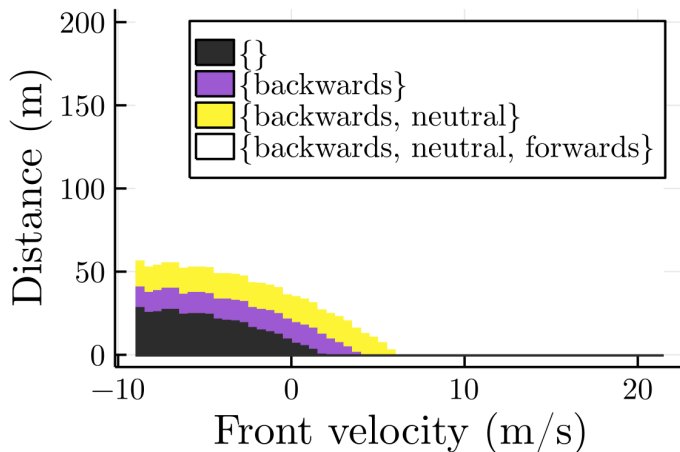
Pre-shielded agent



Pre-shielded agent

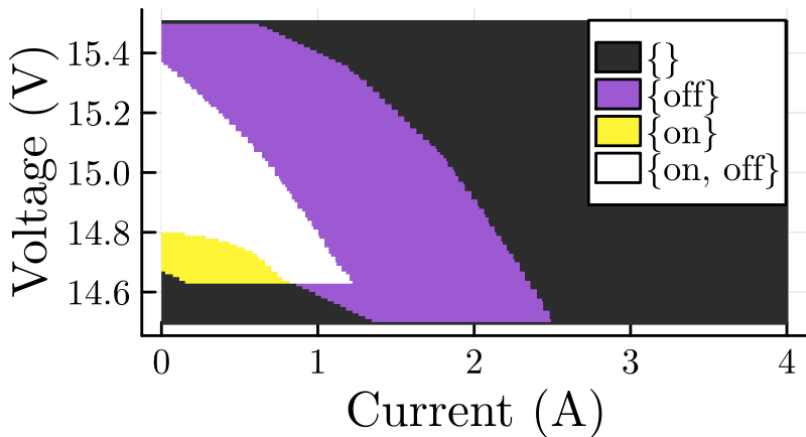


Cruise control (car behind another car)

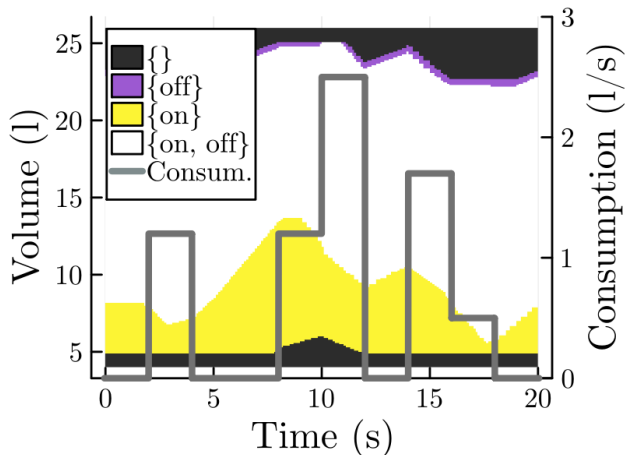


Synthesis time: 36 min

DC-to-DC boost converter



Synthesis time: 1 h 19 min

Oil pump (plot: pump is *on*)

Synthesis time: 5 h 23 min

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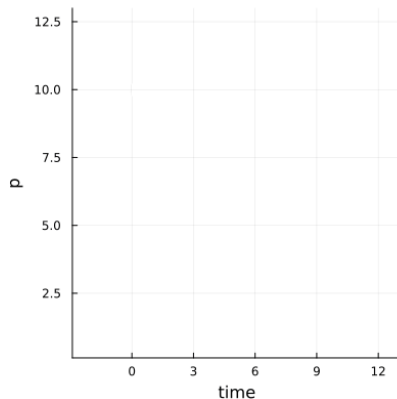
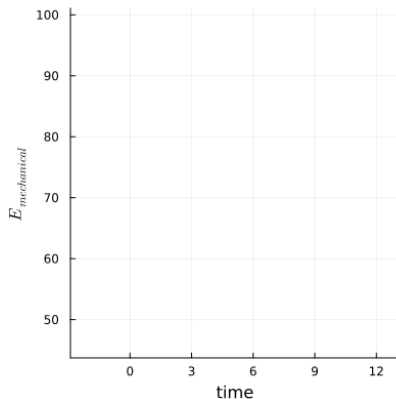
Multi-agent systems

Conclusion

A different perspective: Mechanical energy

$$E_{\text{mechanical}} = mgp + 0.5mv^2$$

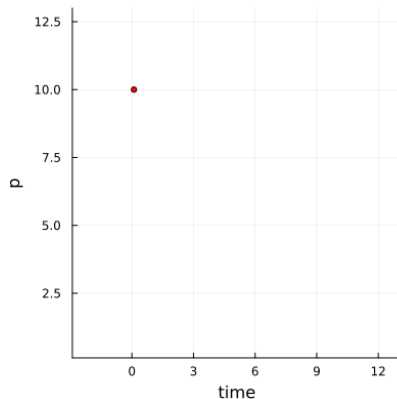
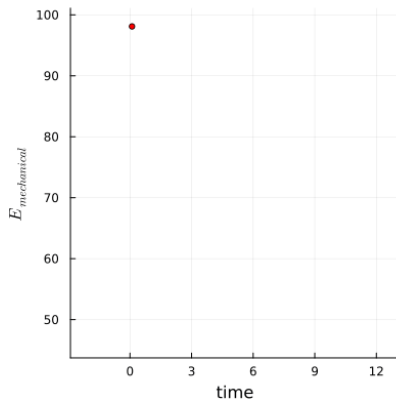
with mass m (1 kg), gravity g (9.81 m/s²), position p , velocity v



A different perspective: Mechanical energy

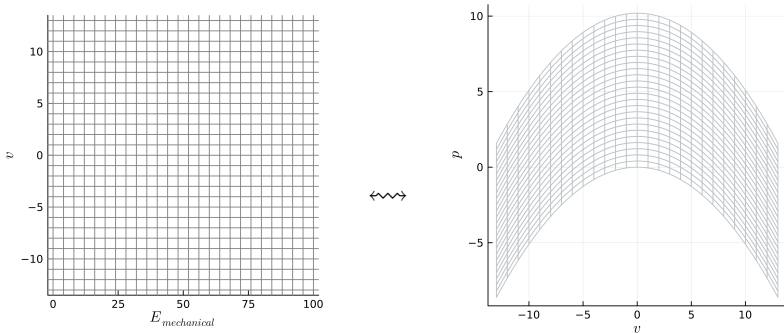
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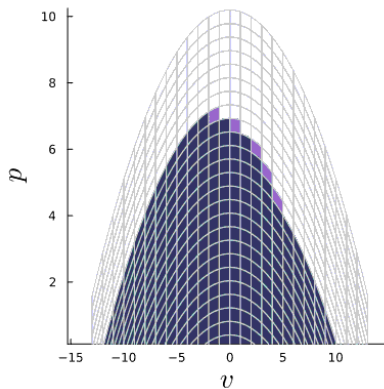
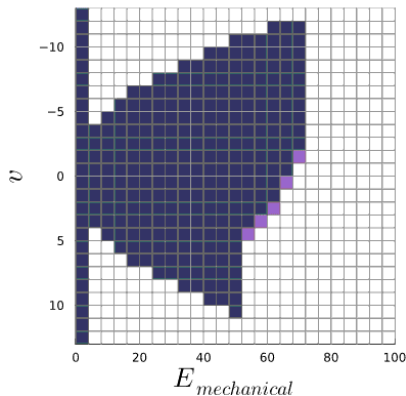


Transformed state space

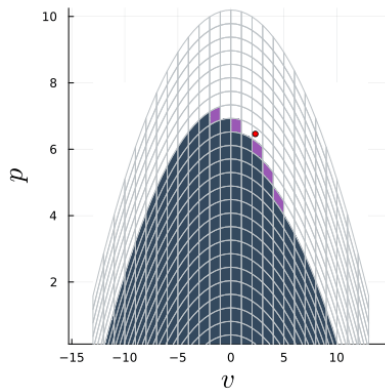
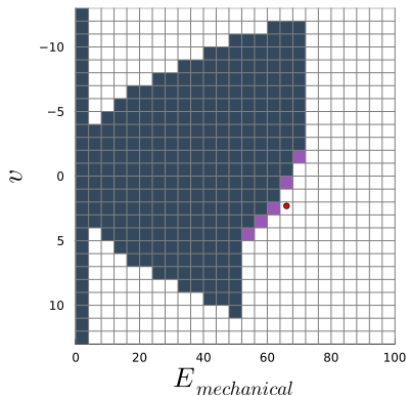
- New state dimensions ($E_{mechanical}, v$)
- We construct a regular grid again
- Much fewer cells (650) are sufficient
- Can go back and forth wrt. original state space (v, p)



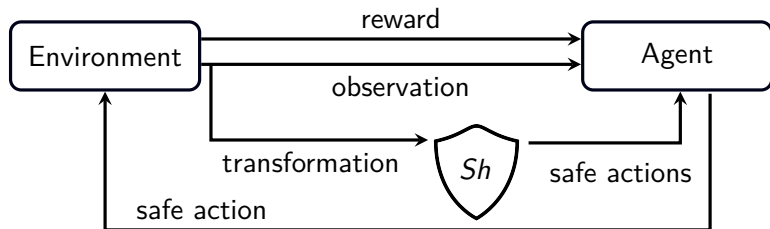
Shield in transformed state space



Shield in transformed state space

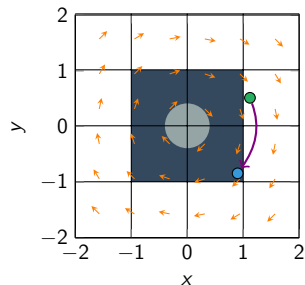


Shielding with transformation¹

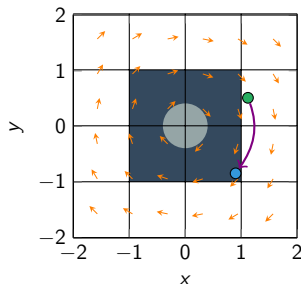


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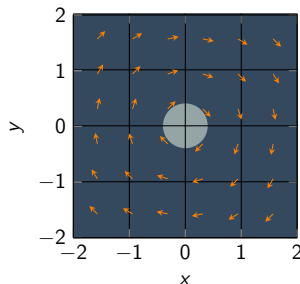
Another example: Bivariate harmonic oscillator



Another example: Bivariate harmonic oscillator

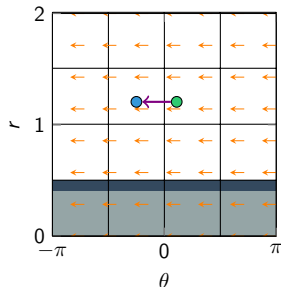
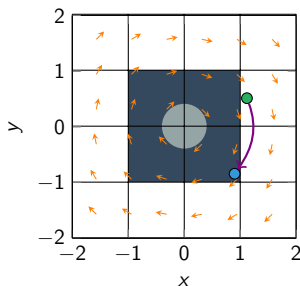


$\rightsquigarrow$



- All cells get into the avoid set $\rightarrow$ useless shield
- Would require a very fine-grained grid

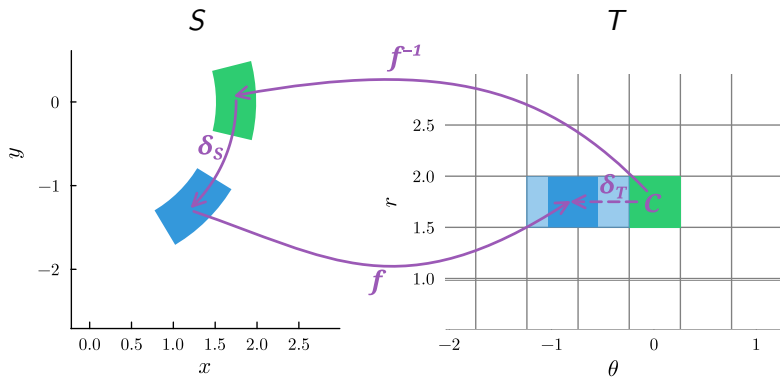
Another example: Bivariate harmonic oscillator



- Transformation to polar coordinates

$$f(x, y) = (\text{atan2}(y, x), \sqrt{x^2 + y^2}) =: (\theta, r)$$

Shield computation with a state-space transformation

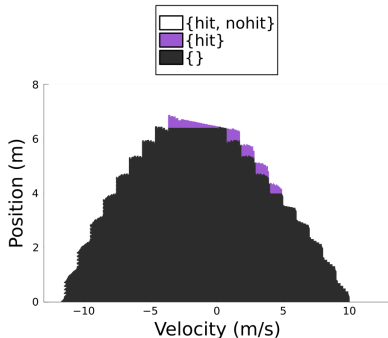
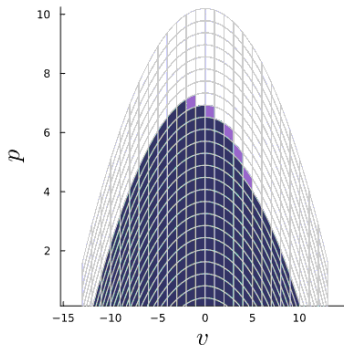


Compute successors δ_T from C (green) in transformed space T

1. Map to original state space S via f^{-1}
2. Compute successors via δ_S
3. Map back to transformed state space T via f (dark blue)

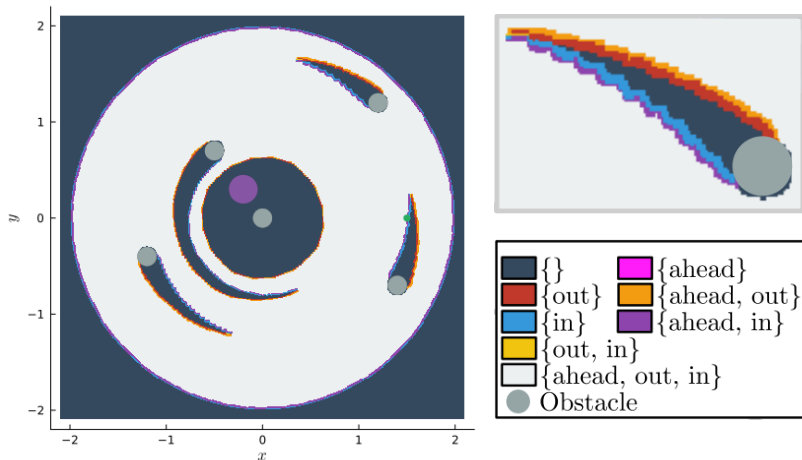
As before, identify all cells intersecting with this set (light blue)

Bouncing ball



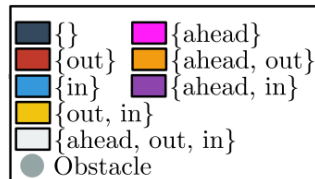
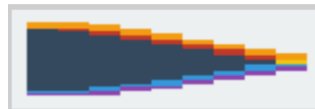
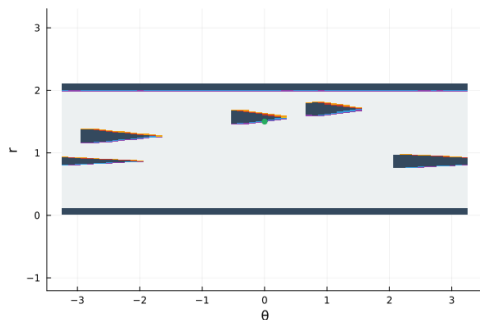
- 650 cells
- Computation: 1.3 sec
- Decision tree: 49 nodes
- 520,000 cells
- Computation: 134 sec
- Decision tree: 940 nodes

Satellite: Harmonic oscillator + {ahead, out, in}

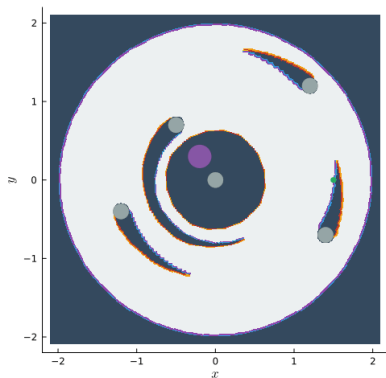


- Goal: reach a randomly spawning purple area
- Here this is impossible because the shield forbids it

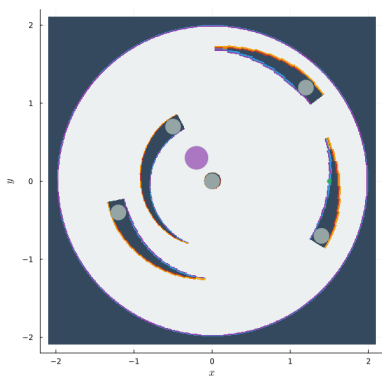
Satellite: Transformed state space



Satellite: Comparison

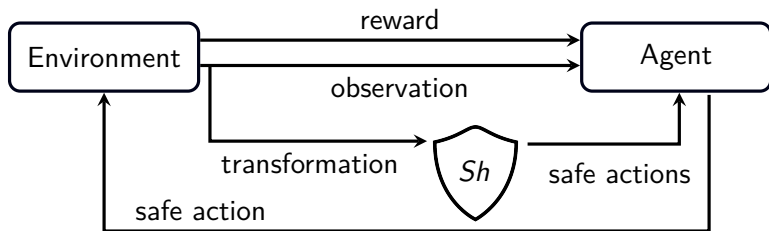


- 176,400 cells
- Computation: 161 sec
- Decision tree: 4,913 nodes



- 27,300 cells
- Computation: 10 sec
- Decision tree: 544 nodes

Impact on learning



- We evaluate the cumulative return in six scenarios
- Three shield variants: no shield, shield in S , shield in T
- Two learning variants: learn in S , learn in T
- (The diagram shows the variant *shield in T , learn in S*)

Impact on learning

Learning	Satellite (↗)			Bouncing ball (↘)		
	None	<i>S</i>	<i>T</i>	None	<i>S</i>	<i>T</i>
<i>S</i>	1.123	0.786	<u>1.499</u>	39.897	37.607	<u>36.593</u>
<i>T</i>	0.917	0.889	1.176	39.128	40.024	39.099

Learning	Cart-pole (↘)		
	None	<i>S</i>	<i>T</i>
<i>S</i>	0.007	0.019	0.001
<i>T</i>	<u>0.000</u>	<u>0.000</u>	<u>0.000</u>

Overview

Motivation

Approach

Experiments

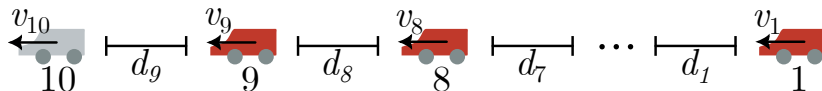
State-space transformation

Multi-agent systems

Conclusion

Multi-agent shields¹

- Shield synthesis scales exponentially in number of agents
- Compositional shielding based on assume-guarantee reasoning
- Shielding even enables more efficient reinforcement learning



¹Brorholt, Larsen, and Schilling. *AAMAS*. 2025.

Overview

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Conclusion

Conclusion

- Shield synthesis for hybrid systems
 - Replace hard steps by simulation¹
 - Now fully integrated in Uppaal
 - State-space transformation: more precise and efficient²
 - Multi-agent shielding: assume-guarantee structures³

Future work

- Dynamic partitioning
- Combination with symbolic approach
- Learning of suitable transformations and assumptions

¹Brorholt, Jensen, Larsen, Lorber, and Schilling. *AISoLA*. 2023.

²Brorholt, Høeg-Petersen, Larsen, and Schilling. *AISoLA*. 2024.

³Brorholt, Larsen, and Schilling. *AAMAS*. 2025.