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April 30, 2026

# Paper Review: Tolerance of RL Controllers in CPS

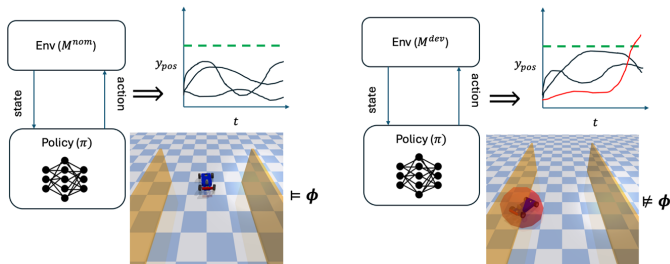
*Zhang, Kapoor et al.*

- 1 Real-World Deployment Issues
- 2 Contribution
- 3 System Model
- 4 Proposed Problem
- 5 Solution
- 6 Implementation
- 7 Evaluation
- 8 Future Works

- CPS with RL controllers face challenges in real-world deployment due to "reality gaps."
- Training methods like robust RL lack formal guarantees of tolerance.
- No existing frameworks for post-training analysis of tolerance under deviations.

- Formal definition of tolerance using Signal Temporal Logic (STL).
- Introduced the tolerance falsification problem.
- Proposed a two-layer analysis framework for finding tolerance violations.

- CPS modeled as Markov Decision Processes (MDPs) with RL-based controllers.
- STL used to describe system requirements and deviations.
- Deviations modeled as parameter changes in system dynamics.



**Figure:** STL is used to describe system requirements and deviations in the CarRun system.

- **Tolerance Falsification Problem:** - Identify system deviations causing violations of STL-defined requirements.
- Focus: Find minimal deviations likely to occur in practice.

- Two-layer framework:
  - **Upper layer:** Explores deviations in system parameters.
  - **Lower layer:** Finds violating trajectories for a given deviation.
- Heuristic leveraging trajectory similarity to guide the search.

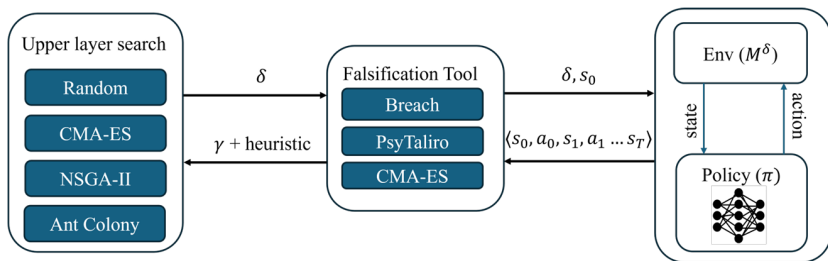


Figure: Two-layer framework

- › Python-based framework integrating:
  - › CPS falsifiers (e.g., Breach, PsyTaLiRo).
  - › Simulation platforms (OpenAI-Gym, PyBullet, Simulink).
  - › CMA-ES optimization algorithm for both layers.

- Benchmarked on diverse CPS scenarios:
  - Lunar Lander, Adaptive Cruise Control, etc.
- Outperformed one-layer falsifiers in finding smaller, realistic deviations.

	One-layer search			CMA-ES				CMA-ES w/ Heuristic		
	Viol.	Min Dst.	Avg. Dst.	Viol.	Min Dst.	Avg. Dst.	Viol.	Min Dst.	Avg. Dst.	
Cartpole	<b>90</b>	0.300	<b>0.399</b>	69	0.285	0.449	79	<b>0.256</b>	0.417	
LunarLander	-	-	-	74	0.026	<b>0.222</b>	<b>84</b>	<b>0.020</b>	0.293	
CarCircle	11	0.143	0.255	22	0.102	<b>0.219</b>	<b>57</b>	<b>0.068</b>	0.454	
CarRun	25	0.191	<b>0.249</b>	68	0.161	0.449	<b>109</b>	<b>0.156</b>	0.399	
ACC	N/A	N/A	N/A	43	<b>0.110</b>	<b>0.323</b>	<b>110</b>	0.138	0.415	
WTK	<b>300</b>	0.299	<b>0.443</b>	54	<b>0.296</b>	0.454	45	0.319	0.533	

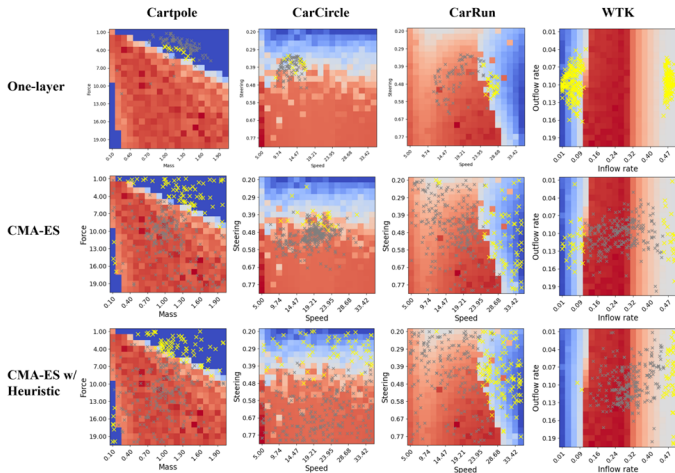


Figure: Visualise deviations with heatmaps

- Extend STL evaluation to probabilistic metrics.
- Explore alternative distance metrics (e.g., Wasserstein distance).
- Integrate STL decomposition for complex specifications.



Questions?