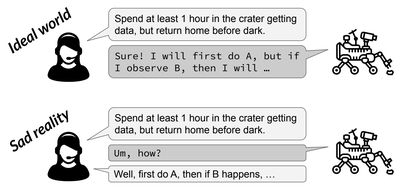
Robust Counterexample-guided Optimization for Planning from Differentiable Temporal Logic

PRESENTER:

Charles Dawson

BACKGROUND: There is a gap between how users would ***like*** to program our robots and what robots will accept in reality.

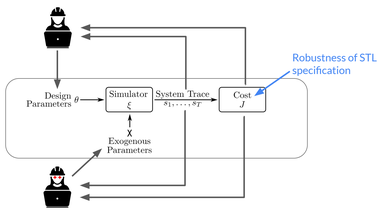


Contributions:

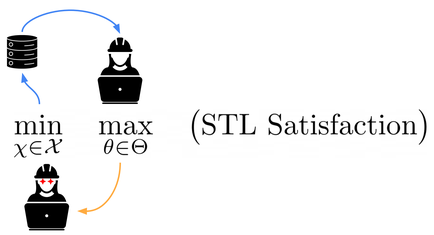
- Efficient, robust planner for STL tasks.
- Robustness from:
  - 2-player game formulation.
  - Counterexample-guided optimization.
- Efficiency from:
  - Differentiable simulation.
  - Differentiable temporal logic.
- Open-source implementation.

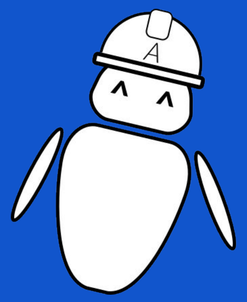
APPROACH:

The planner plays a zero-sum game with an adversary to maximize the robustness of its plan.



The planner caches counterexamples found by the adversary to guide optimization. Automatic differentiation of the simulator and the temporal logic specification enables efficient planning.





Architect

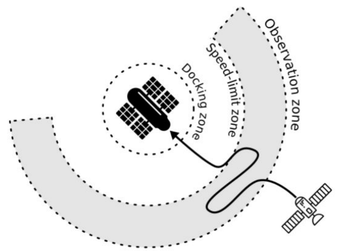
Automated, Robust Co-Design

Automatic differentiation

\+ Counterexample-guided optimization

= Fast, safe planning for temporal logic.

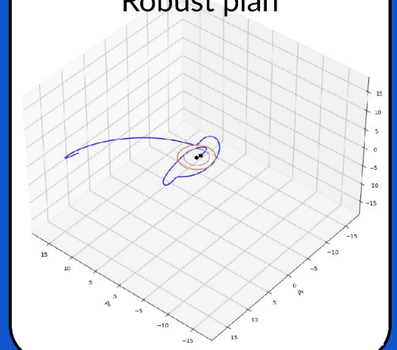
Robot task specification



STL formula

$$\psi_{\text{speed limit}} = (r \geq 2.0)\,\mathcal{U}\,\square\,(v \leq 0.1)$$
$$\psi_{\text{loiter}} = \lozenge\square_{[0,T_{obs}]}\,(2.0 \leq r \wedge r \leq 3.0)$$
$$\psi_{\text{reach target}} = \lozenge\,(r \leq 0.1)$$

Robust plan



Take a picture to **download** the **full paper and code**

Signal Temporal Logic (STL)

Formal task specification:

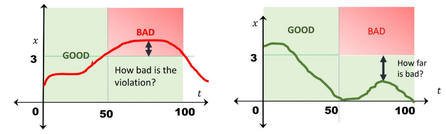
When close to target, **always** limit speed

Eventually be in grey region **for 2 minutes**

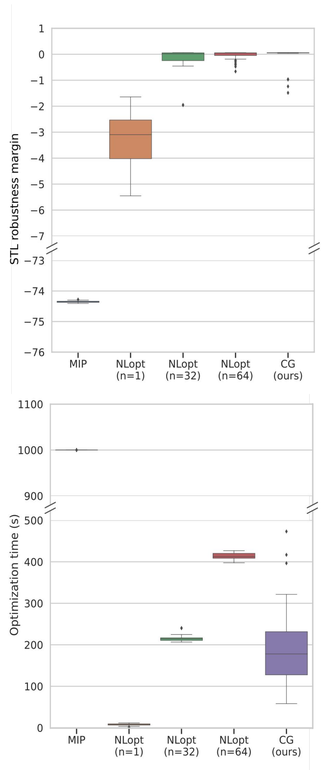
Eventually reach the target

$$\psi_{\text{speed limit}} = (r \geq 2.0)\,\mathcal{U}\,\square\,(v \leq 0.1)$$
$$\psi_{\text{loiter}} = \lozenge\square_{[0,T_{obs}]}\,(2.0 \leq r \wedge r \leq 3.0)$$
$$\psi_{\text{reach target}} = \lozenge\,(r \leq 0.1)$$

(Differentiable) STL Robustness Metric



Experimental Results



Authors

Charles Dawson, Chuchu Fan

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