

CRII: CPS: FAICYS: Model-Based Verification for AI-Enabled Cyber-Physical Systems Through Guided Falsification of Temporal Logic Properties

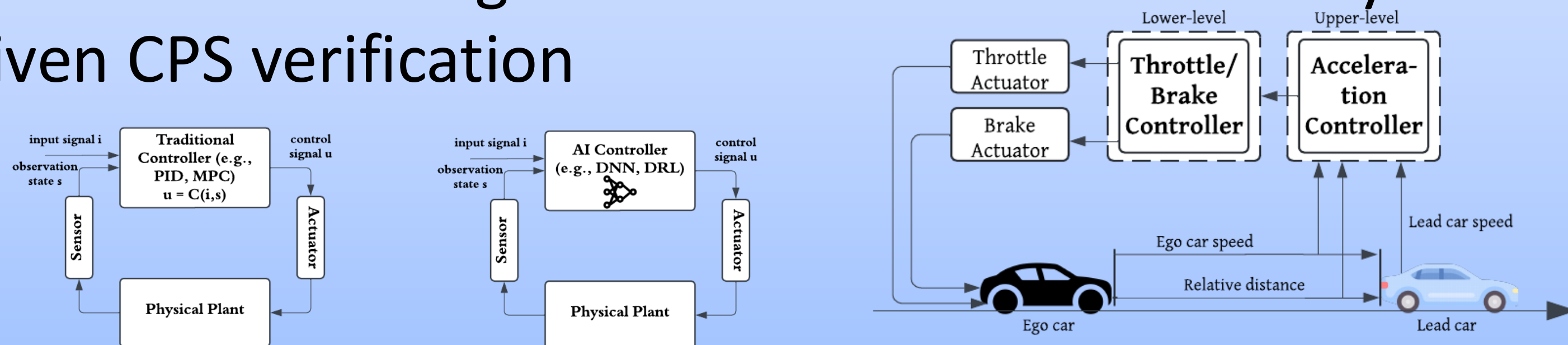
Khoulood Gaaloul, University of Michigan-Dearborn

https://www.nsf.gov/awardsearch/showAward?AWD_ID=2347294&HistoricalAwards=false



Challenge:

- Verification bias and lack of Benchmarks in AI-Enabled CPS
- Limitations of temporal Logic falsification for AI-enabled CPS
- Exploration challenges and curse of dimensionality in AI-Driven CPS verification

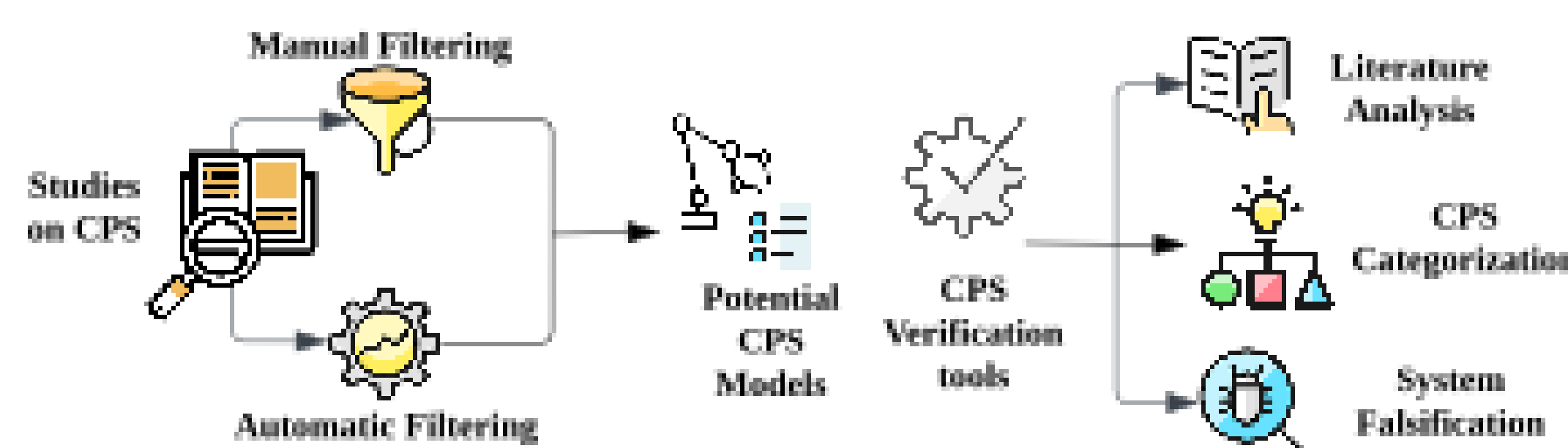


Scientific Impact:

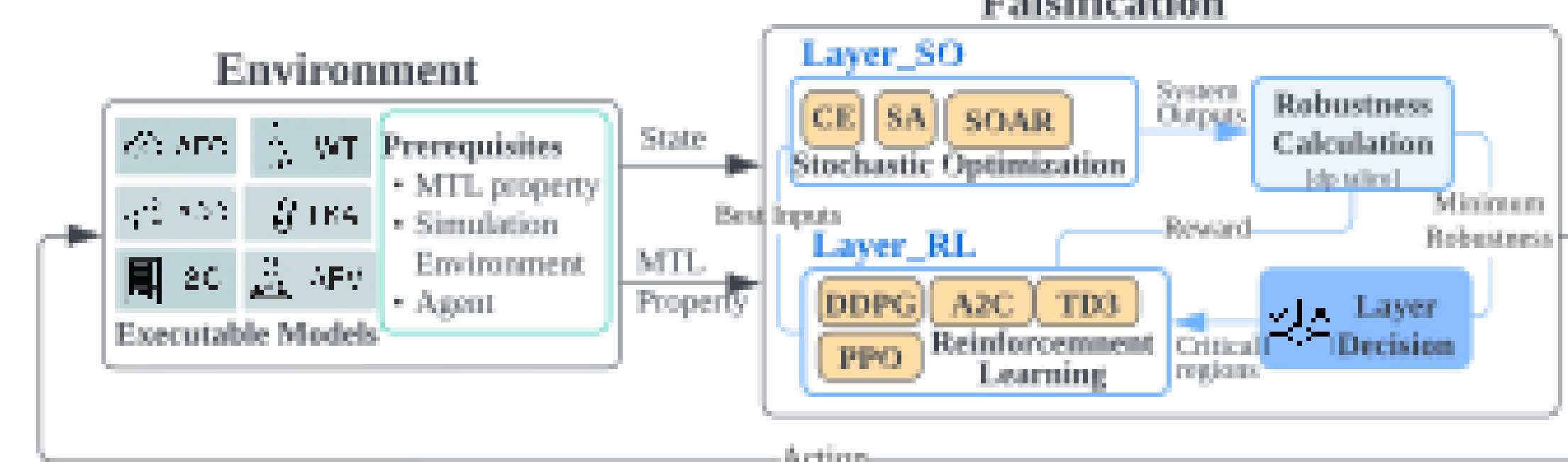
- Deep understanding of neural network interactions in AI-Enabled CPS through empirical analysis
- Optimized falsification through stochastic optimization and reinforcement learning in a novel framework that improves the efficiency and accuracy in detecting property violations in AI-enabled CPS

Solution: Thrust-I: Empirical analysis and benchmarking for

AI-enabled CPS verification



Thrust-II: FAICYS



- * Fully Automated
- * Accurate Detection
- * Efficient & Scalable

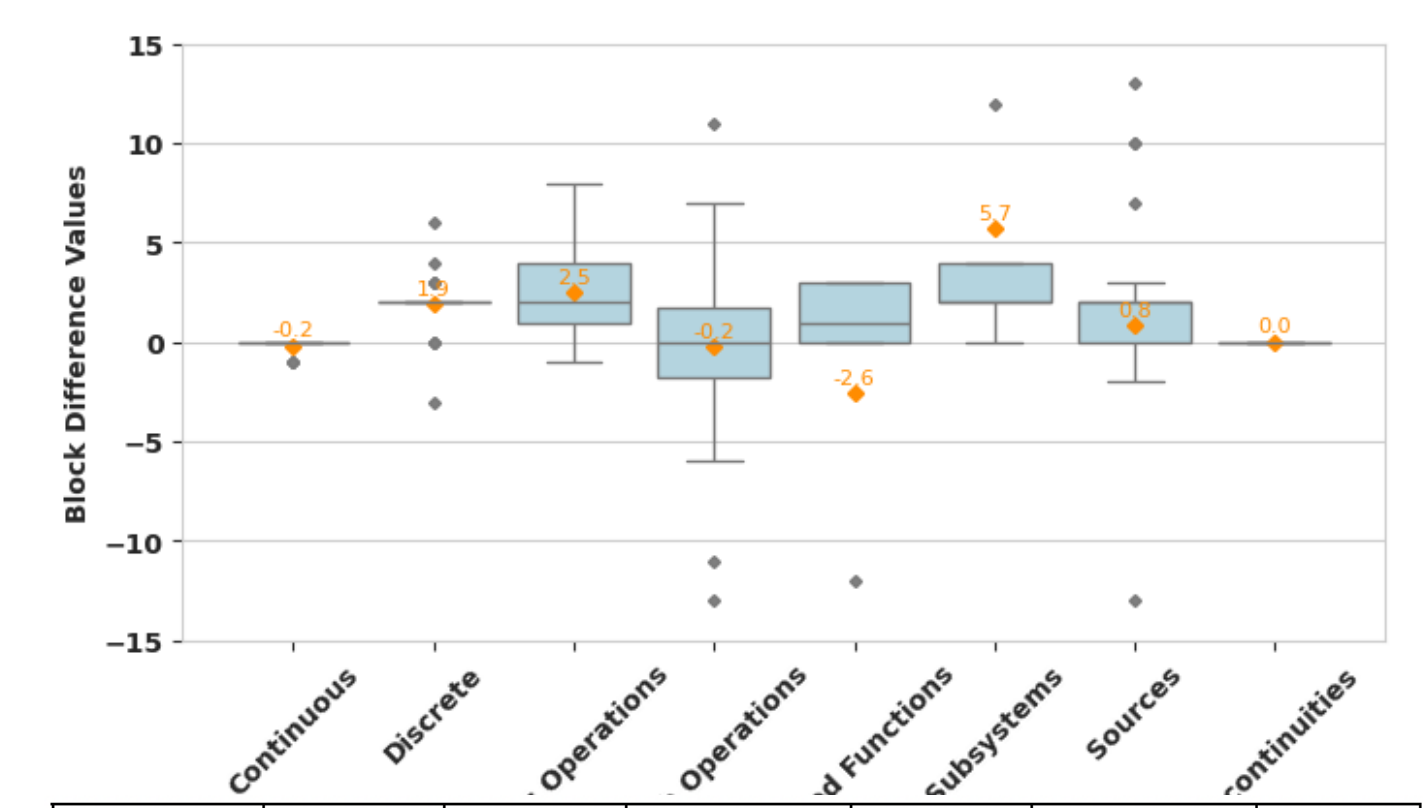
Broader Impact:

- Enhance the reliability of AI systems by reducing safety-critical failure rates by 20-30% across broad CPS applications
- Increase female representation in AI and CPS research by 10-20% through targeted mentorship and outreach



Accomplishments:

- Category-Wise Atomic Block Differences between AI-Driven and Traditional CPS models
- Flow Dynamics Metrics for Traditional (T) and AI-Driven (AI) CPS Models
- Fault Detection Results



	Model	Total BC	Relevant BC	Total CC	Relevant CC	HD
System Average	T AI	288.75 349.13	116.5 141.5	306 350.13	274.25 312	6.38 7.75
% Diff	-	+29.2	+25.7	+21.1	+20.5	+21.0

Model/Policy	#Violated Exec.	# Fals. Requirements	Avg. time
Traditional	30	8/8	0.2
AI Avg	26	6/8	73.9