CPS: Small: Neuro-Symbolic Learning and Control with High-Level Knowledge Inference

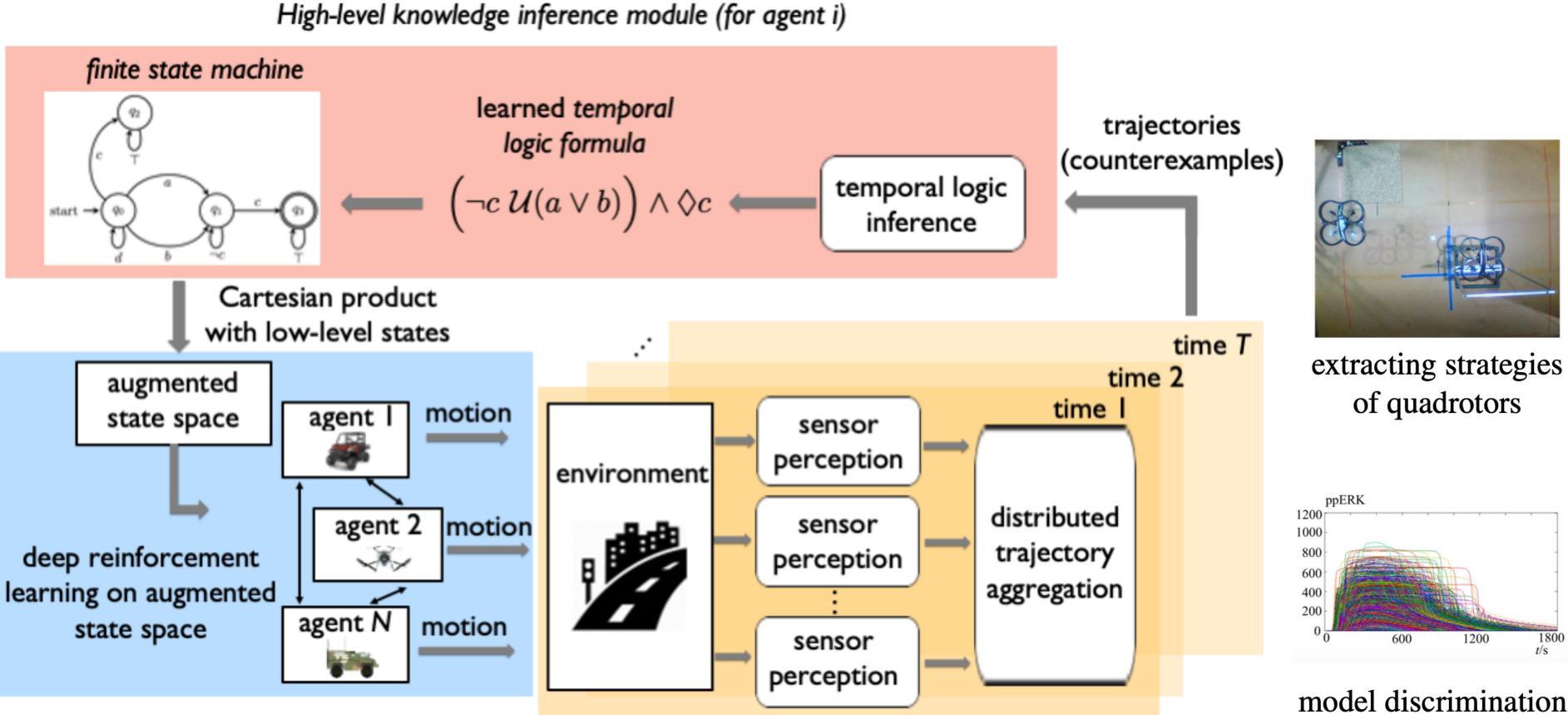
Zhe Xu (Arizona State University) https://sites.google.com/site/zhexudavid00710/home

Challenge: Application of AI to CPS:

- Extensive need for data and online interactions
- Lack of interpretability and transparency
- Lack of robustness and transferability

Solution:

- Data-efficient neuro-symbolic reinforcement learning with high-level knowledge inference
- Scalable and resilient neuro-symbolic multi-agent reinforcement learning with high-level knowledge inference
- Provably correct neuro-symbolic multi-agent learning-based adaptive control with unknown dynamics and high-level specifications



Stochastic game-based deep reinforcement learning module

Distributed perception module



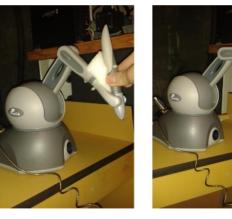
Scientific Impact:

neuro-symbolic reinforcement learning

high-level knowledge

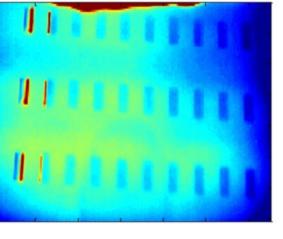
> neuro-symbolic learning-based adaptive control

model discrimination for biological systems









extracting robot

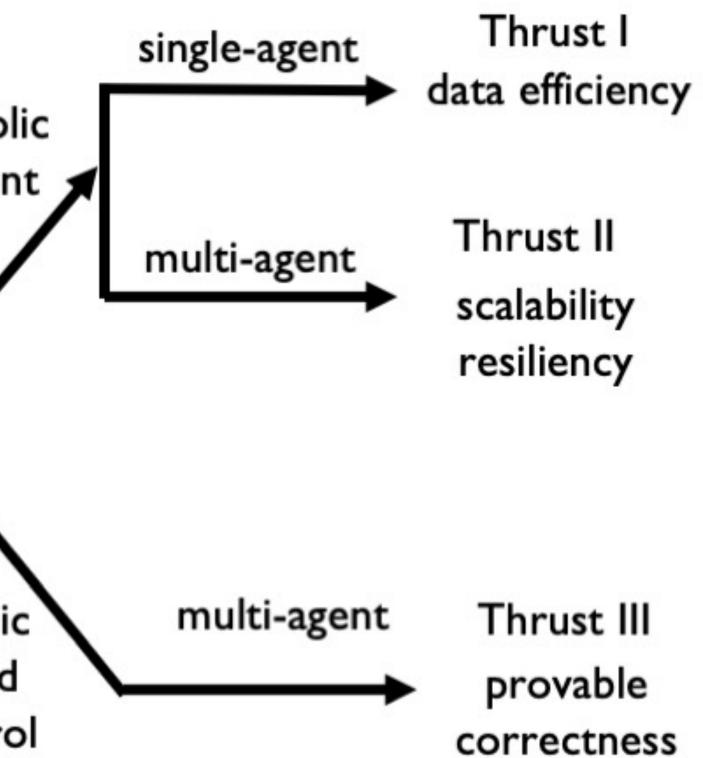
strategies

explaining additive manufacturing patterns



explaining privacy-preserving fault detection

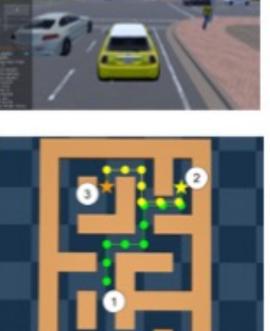
• To address the cross-cutting issues of applying the neuro-symbolic approaches to CPS so that data efficiency, interpretability, scalability, and resiliency be significantly improved.



Thrust III provable

Evaluation Tier I: high-fidelity simulations





Tier II: physical experiments



extracting soccer strategies

Broader Impact:

- Proposed effort has a special focus on CPS, including but not limited to robotic systems, autonomous systems, and networked CPS.
- Undergraduate and graduate courses and research workshops
- K-12 outreach by participating in the campus-wide activities at ASU









