



CRII: CPS: Towards a Model-Based Reinforcement Learning Approach for Safe Operation of Distributed Energy Systems

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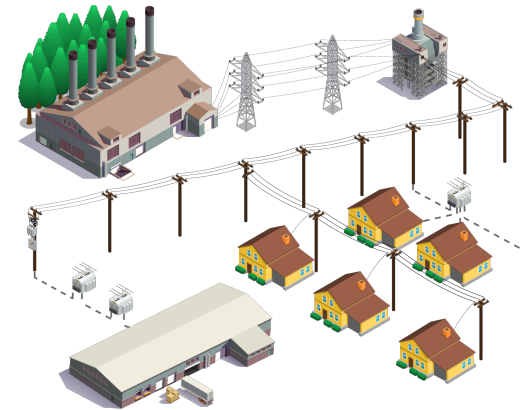
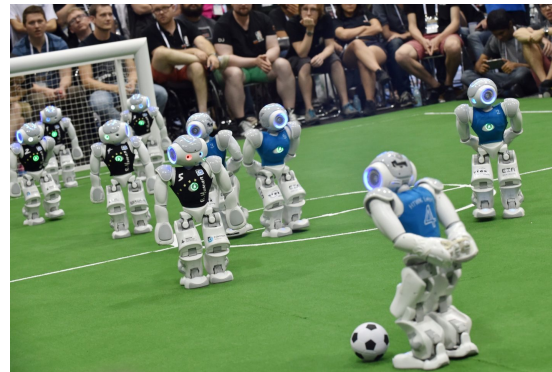
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Reinforcement Learning and Safety

- **Reinforcement Learning** (RL) addresses the problem of learning to control *unknown* systems by explicitly considering their inherent *dynamical* structure
- Control policy for any real-world systems should maintain some necessary safety criteria to avoid undesirable outcomes
 - Avoid collisions, avoid falling down, avoid blackouts

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- Standard RL algorithms typically focus on a single objective of maximizing the value function (expected cumulative reward)
- Standard RL algorithms may violate the necessary safety constraints and can lead to disastrous consequences
- **How do we learn RL algorithms that maximize the objective while satisfying the safety requirements?**

Safe RL Algorithms with Provable Guarantess

- **Objective:** Learn a safe policy with minimum number of online samples, with provable guarantees on performance
- **Contributions:**
 - RL algorithms that minimize the safety constraint violations while learning the optimal safe policy
 - Provable guarantees on the performance, characterized in terms of the sample complexity of the algorithms

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Theorem 1 (Sample complexity of Generative Model-Based CRL).

[HasanzadeZonuzy et al, 2020a]

Let π_{safe} be the policy obtained from the Generative Model-Based CRL Algorithm with

$$n_o \geq \frac{256}{\epsilon^2} |S| H^3 \log \frac{24|S||A|H}{\delta}$$

Then, $\mathbb{P}(V_{\pi_{safe}} \geq V_{\pi^*} - \epsilon \text{ and } C_{\pi_{safe}} \leq \bar{C} + \epsilon) \geq 1 - \delta.$

Publications

• Safe Reinforcement Learning

1. A. HasanzadeZonuzy, A., Bura, D. Kalathil, S. Shakkottai, ``Learning with safety constraints: Sample complexity of reinforcement learning for constrained MDPS'', AAAI Conference on Artificial Intelligence, February, 2021
2. A. HasanzadeZonuzy, A., Bura, D. Kalathil, S. Shakkottai, ``Model-Based Reinforcement Learning for Infinite-Horizon Discounted Constrained Markov Decision Processes'', International Joint Conference on Artificial Intelligence (IJCAI), August, 2021

• Reinforcement Learning for Power Systems

1. Dongqi Wu, Xiangtian Zheng, Dileep Kalathil, Le Xie, ``Nested Reinforcement Learning Based Control for Protective Relays in Power Distribution Systems'', IEEE Conference on Decision and Control (CDC), December, 2019
2. Rayan El Helou, Dileep Kalathil, Le Xie, ``Communication-free Voltage Regulation in Distribution Networks with Deep PV Penetration'', Hawaii International Conference on System Sciences (HICSS), January, 2020
3. Dongqi Wu, Dileep Kalathil, Miroslav Begovic, Le Xie, ``Deep Reinforcement Learning-Based Robust Protection in DER-Rich Distribution Grids'', submitted to IEEE Transactions of Smart Grid, May, 2021
4. Rayan El Helou, Dileep Kalathil, Le Xie, ``Fully Decentralized Reinforcement Learning-based Control of Photovoltaics in Distribution Grids for Joint Provision of Real and Reactive Power'', IEEE Open Access Journal on Power and Energy, May, 2021