

Action Masking for Safer Model-Free Building Energy Management

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Action Masking for Safer Model-Free Building Energy Management

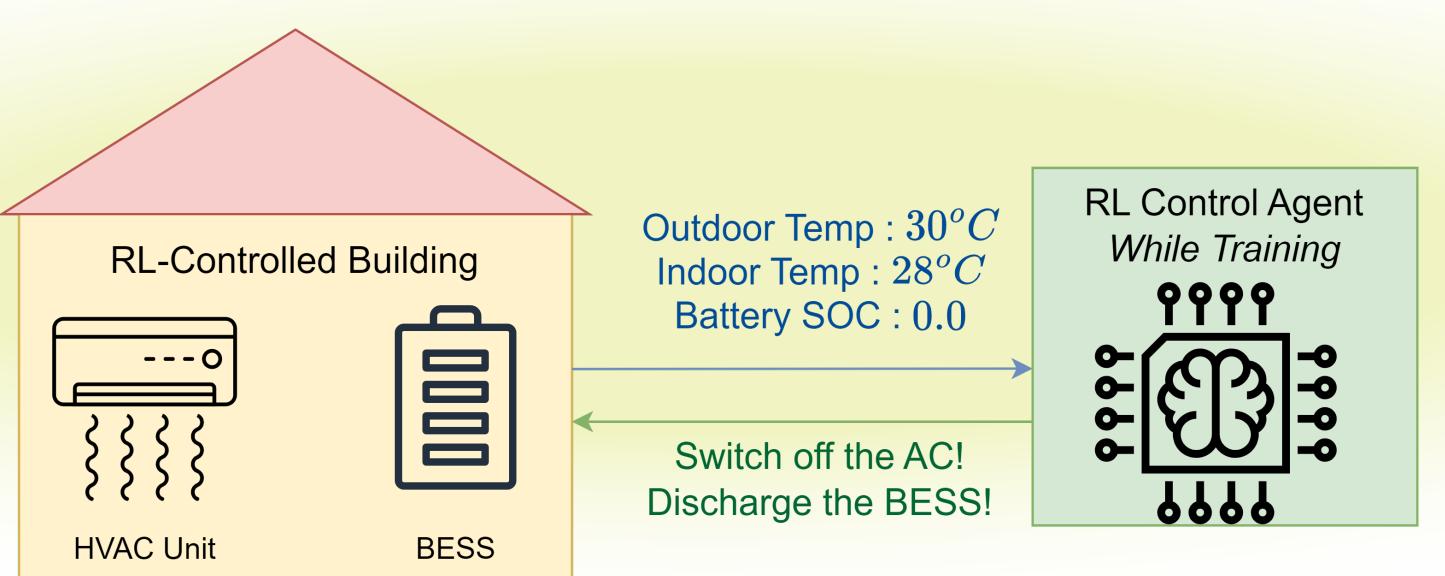
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REINFORCEMENT LEARNING METHODS ARE NOT PRACTICAL FOR CRITICAL APPLICATIONS

Most RL Algorithms require agents to explore the environment through a trial-and-error approach

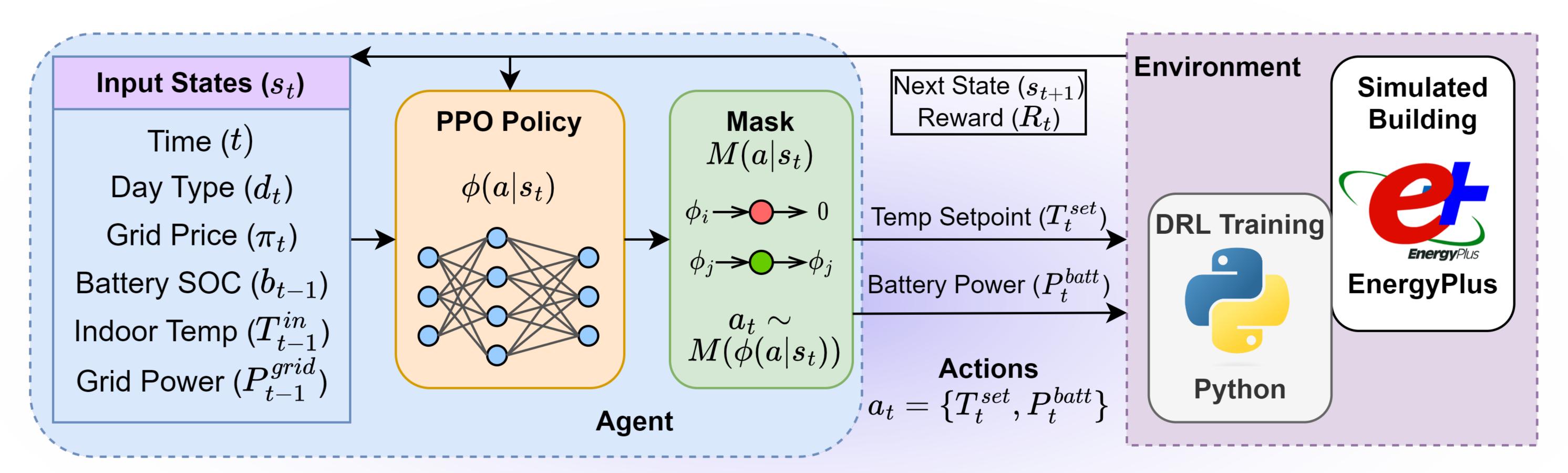
This is not feasible in many realworld control tasks, as unsafe actions may be taken in the learning phase



Problem Statement

To control the cooling HVAC and battery storage systems of a grid-connected building in Singapore, with a trade-off between the electricity bill and the thermal comfort.

ACTION MASKING TO ENFORCE RULES ON THE AGENT



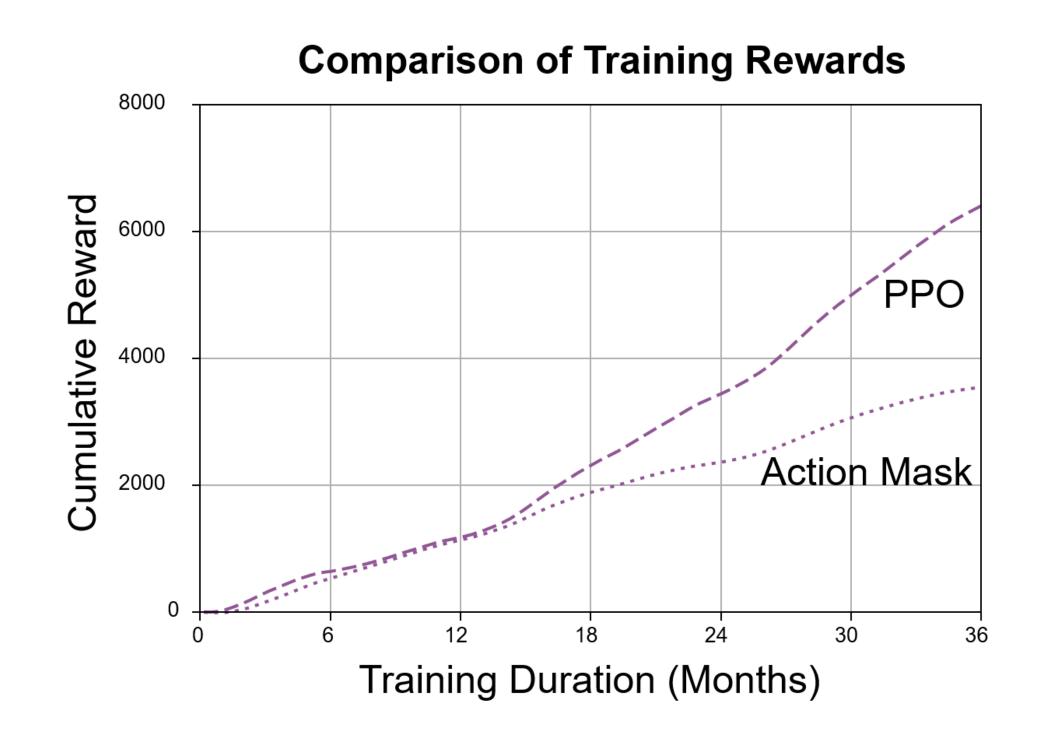
- * Cannot charge (discharge) a full (empty) battery
- * Cooling system switched off from 10 PM to 5 AM
- * Cooling system must stay ON if Tindoor > 26.5°C

The agent is trained using PPO, a popular DRL algorithm

The action mask constrains the exploration space by dynamically limiting the actions the agent can take.

MASKED AGENTS CAN OUTPERFORM DIRECT RL AGENTS

Table 1: Summary of Results



Agent	Cost (\$)	Comfort Score
RBC	468.43	0.43
Direct RL	428.89	0.09
Masked RL	439.27	0.39

Key Results

- 1. Both DRL controllers achieved a lower cost compared to the baseline RBC.
- 2. The direct RL controller led to a significantly worse comfort score.
- 3. Action masking achieved a similar comfort score to the baseline while reducing costs.

Conclusions

- 1. The Direct RL controller prioritized a lower energy bill over thermal comfort (local optima) due to the lack of constraints.
- 2. The use of Action Masking resulted in a policy that reduced the energy bill while respecting thermal comfort rules, without any modifications to the reward function or hyperparameters.

References and Acknowledgements

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