# Causally Aligned Curriculum Learning ICLR 2024, Paper #6200, Halle B #60



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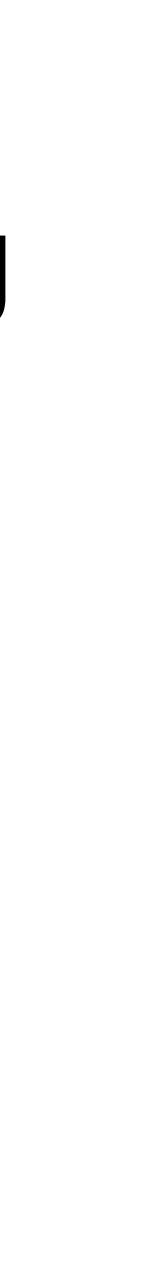
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### **Motivations**

- target environments.
- confounders.
- of optimal decision rules holds in curriculum learning and to generate such curricula to help agents learn more efficiently.

 In complex environments, humans may be guided to learn through a modality called curriculum learning. The expectation in these settings is that when some optimal decision rules are shared across source and target environments, the agent could more quickly pick up the necessary skills to behave optimally in the

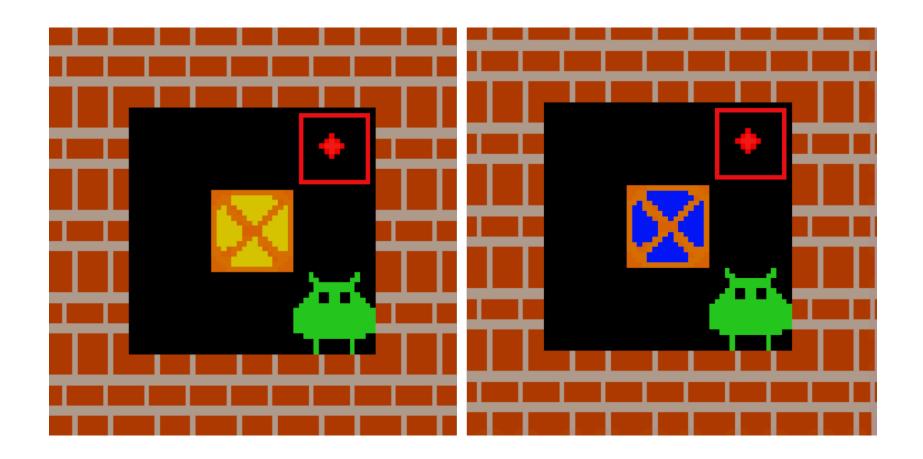
 In practice, however, the critical assumption of an invariance of the optimal policy does not necessarily hold when the underlying environment contains unobserved

• Our goal in this task is to characterize the conditions under which the invariance



# **Colored Sokoban - A Motivating Example**

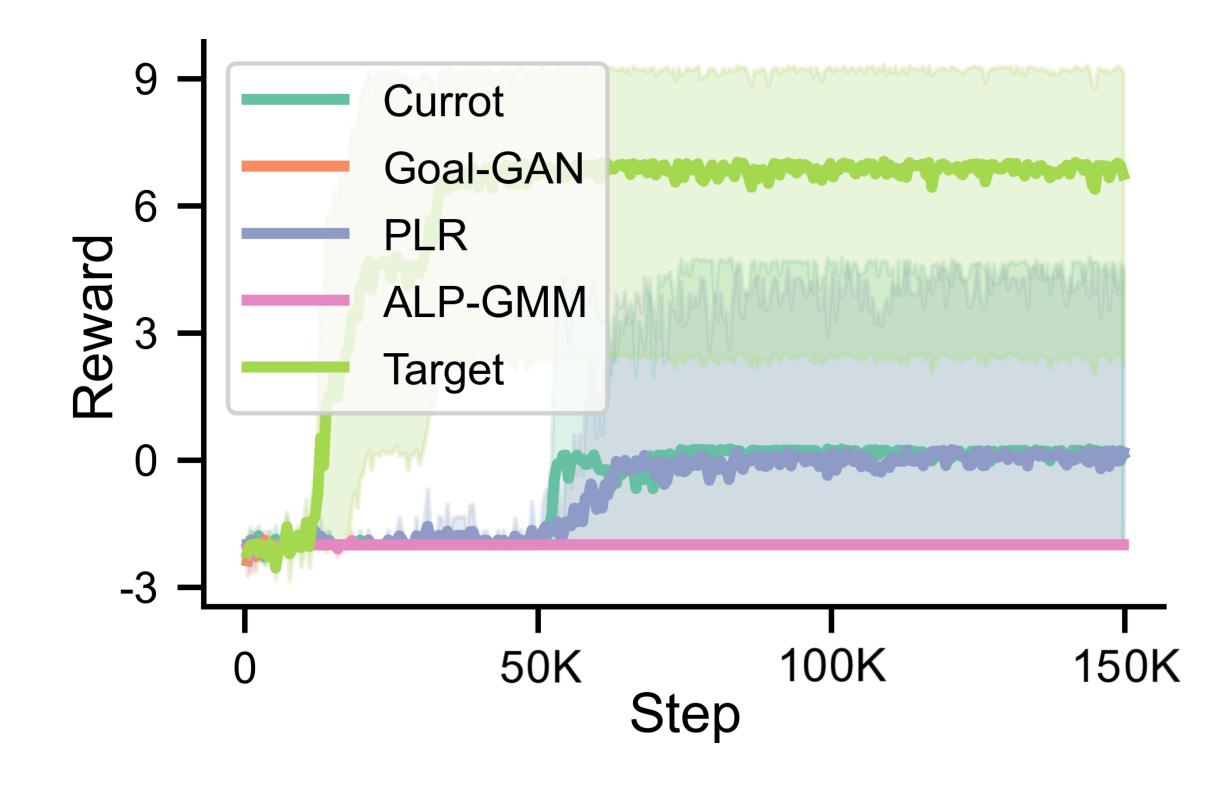
**Goal:** Find a policy ( $\pi$ ) so as to maximize cumulative rewards in the target task.



The box color changes at each time step. The target task is to push a yellow box into the red target position. If the agent pushes a blue box to the target location, it will receive a large penalty.

## **Colored Sokoban - A Harmful Source Task**

directly without curriculum perform?





If we were to design a curriculum to help the agent learn, should we modify every possible variable in the causal diagram?

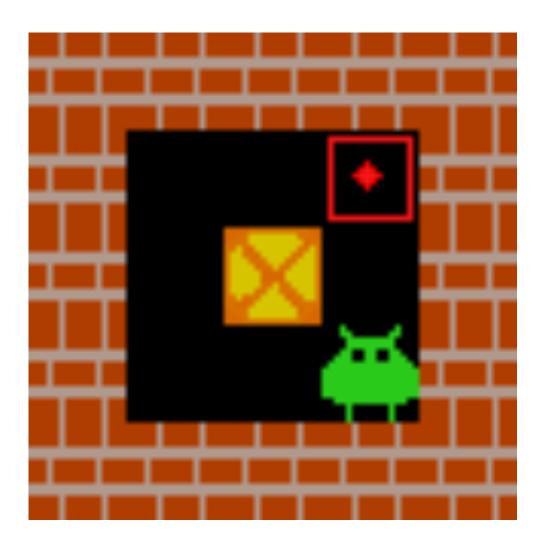
• How do the agents trained by causally unaware curriculum generators and by the target task



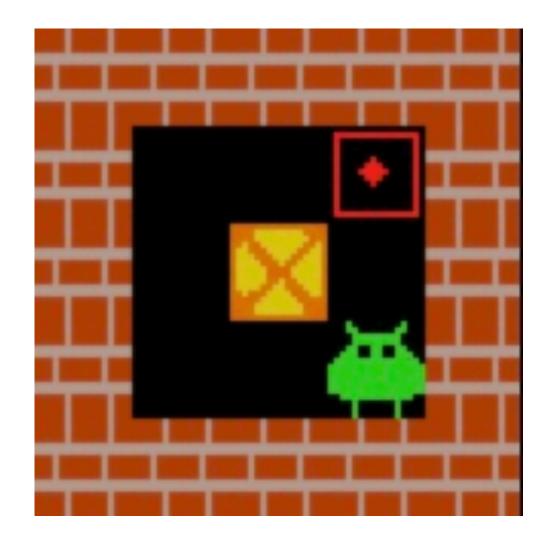
## **Contributions & Video Demonstration**

In this paper, we made the following contributions:

- We derive a sufficient graphical condition determining potentially misaligned source tasks;
- We develop efficient algorithms for detecting misaligned source tasks and constructing source tasks that are guaranteed to align with the target task;
- We introduce a novel augmentation procedure that enables state-of-the-art curriculum learning algorithms to generate aligned curricula to accelerate the agent's learning.



(a) Agent trained by original Currot: avoid pushing the box



(b) Agent trained by causally augmented Currot: push the yellow box to the target

### Paper Link

#### Please scan the following for more details on our paper, "Causally Aligned Curriculum Learning", ID 6200

