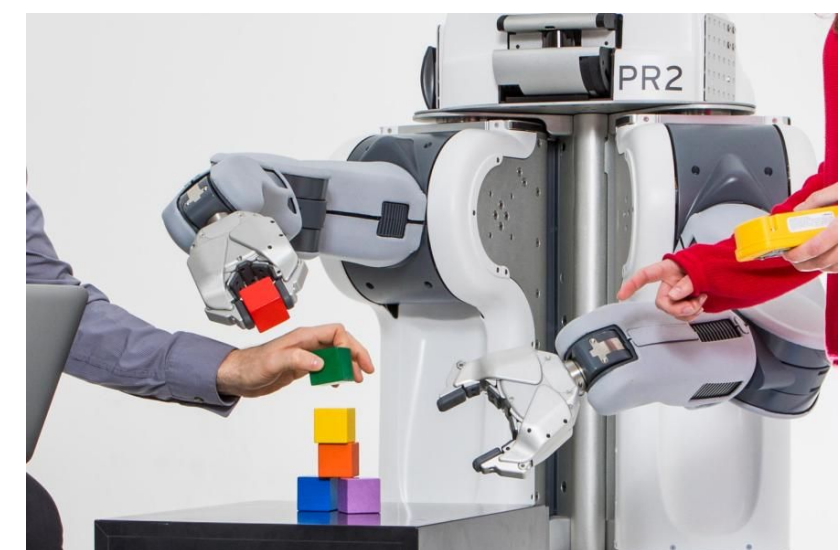


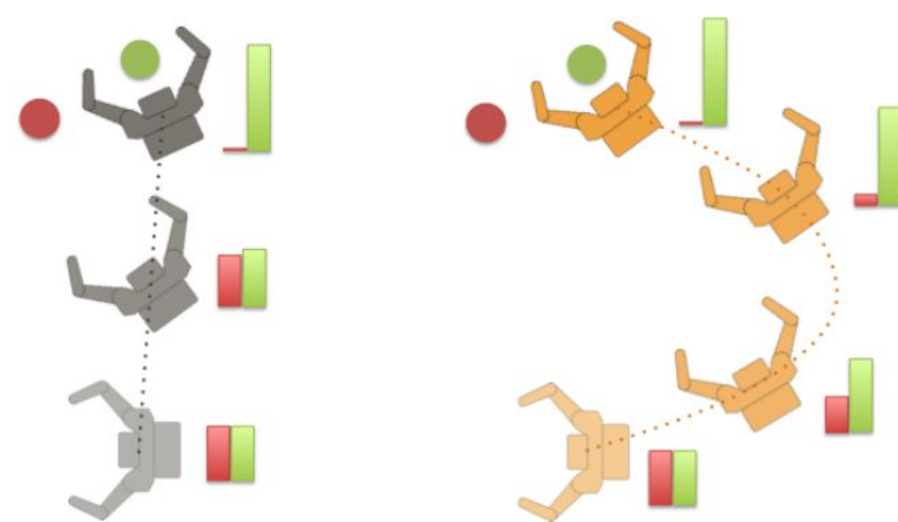


## Legible motion planning: the key to effective human-robot collaboration

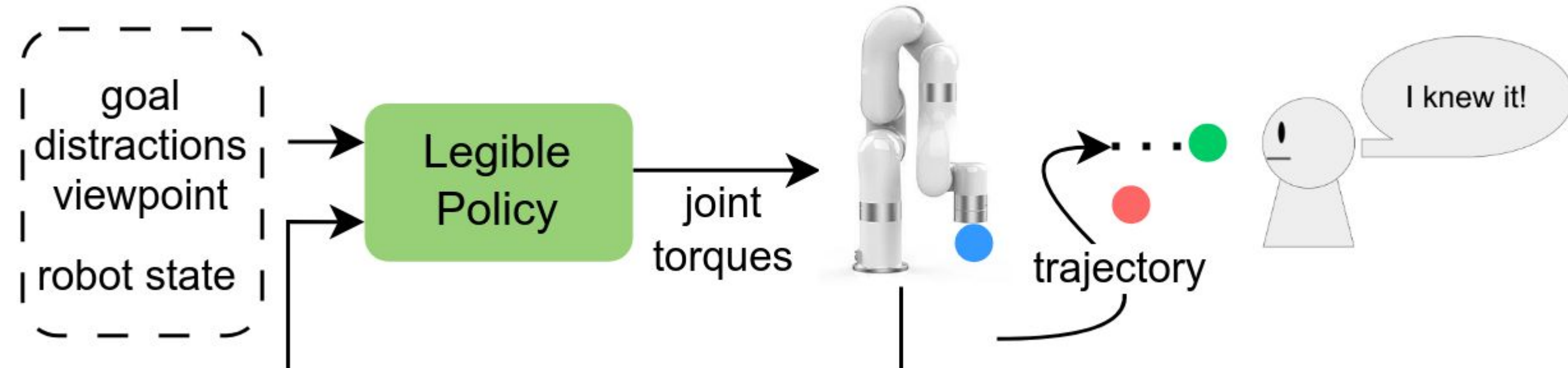


Legible motion is crucial for efficiency and trust in assistive robotics and cooperative tasks

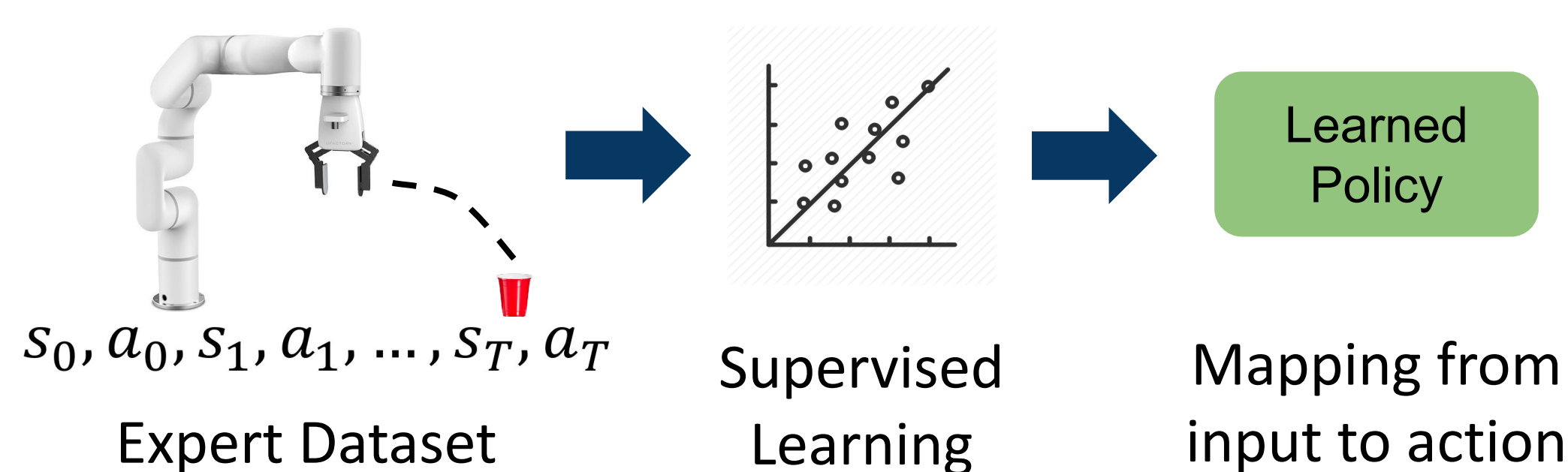
Robots that behave in an understandable manner boost human experience



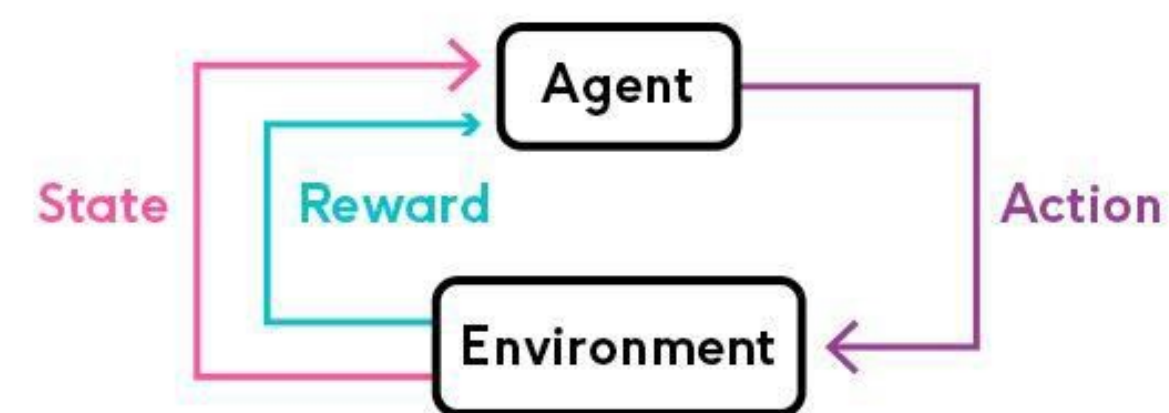
## A unified system for learning-based legible motion planning



## IL performs supervised learning on legible expert demonstrations



## RL maximizes a legibility reward by interacting with the environment



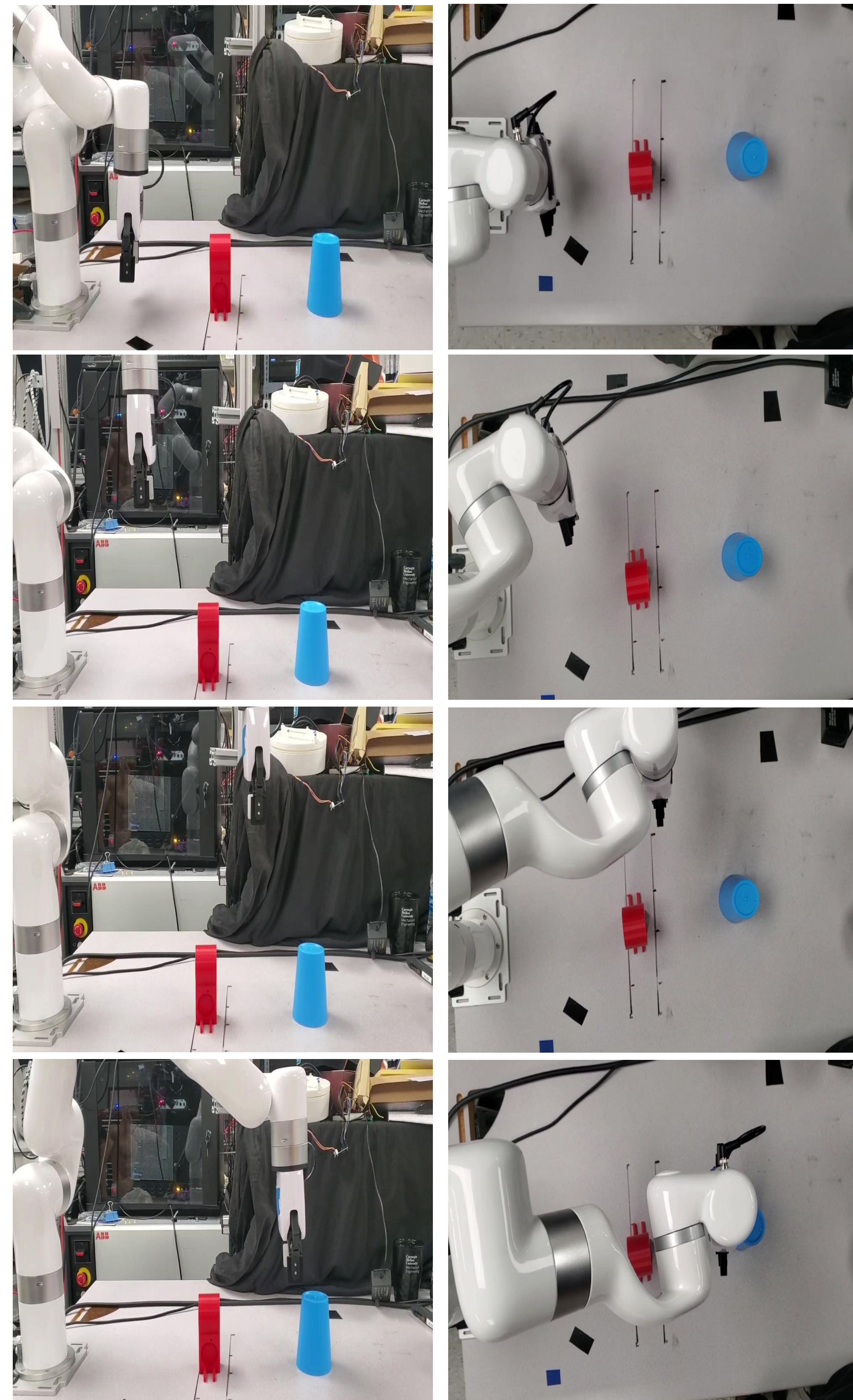
**Algorithm:**  
soft actor-critic +  
hindsight exp. replay

**Reward:**

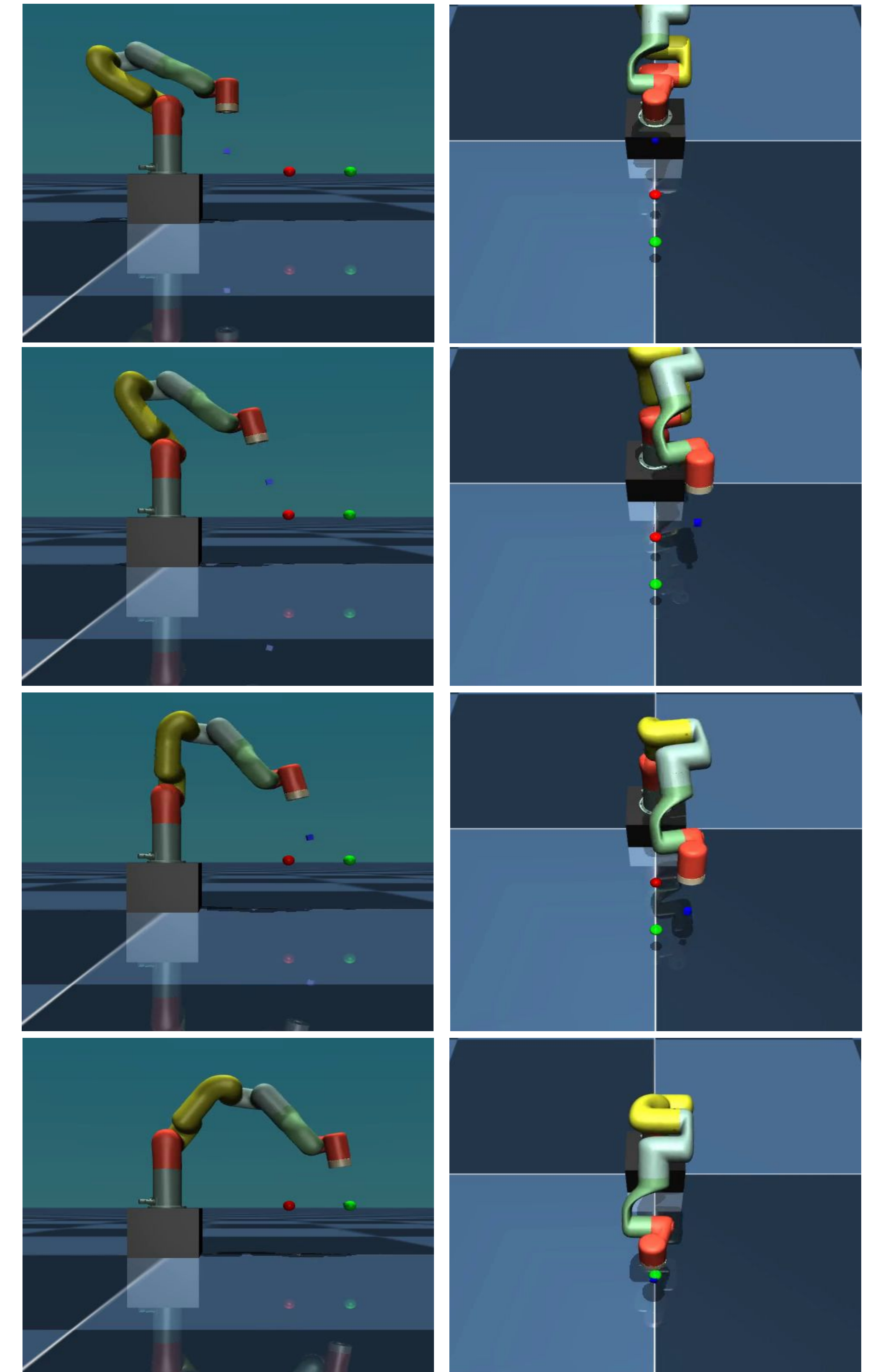
$$r = -\alpha_1 d_{\text{goal}} + \alpha_2 g(d_{\text{distract}}, \beta_1) - \alpha_3 e_{\text{rot}} + \alpha_4 g(z, \beta_2)$$

$$g(x, y) = -1 \text{ if } x \leq y, \text{ otherwise } 0$$

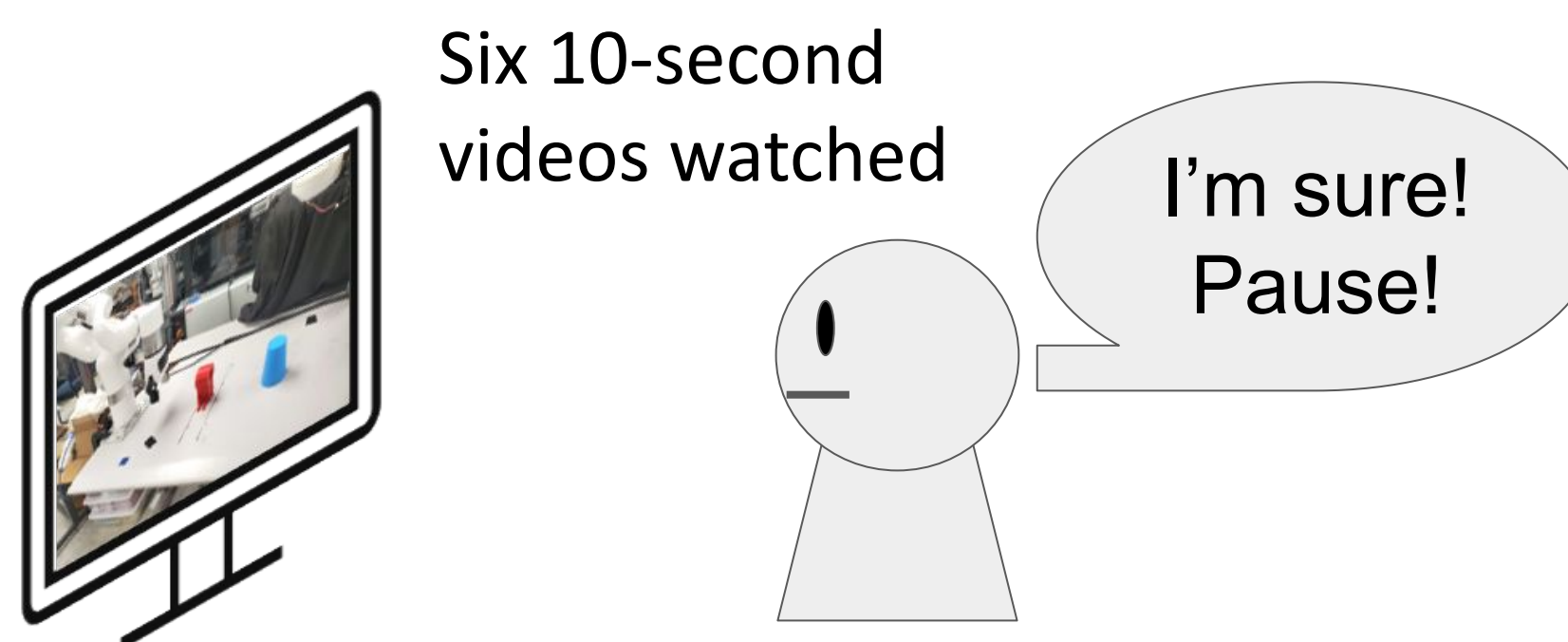
## Results of Imitation Learning Policy



## Results of Reinforcement Learning Policy



## We asked our friends for evaluation



Six 10-second videos watched

Expert	0.65
IM policy	0.64
RL policy	0.75

$$\text{correct} \cdot \left(1 - \frac{t_{\text{pause}}}{t_{\text{total}}}\right)$$

## Conclusions and Future Work

- ✓ Accomplished goal-oriented tasks with legible motion plans considering observer's viewpoint
- ✓ Evaluated two learning-based methods with humans and compared with expert baseline

**Future Work:** Integrate multi-modal sensory inputs for adaptive and dynamic legible motion