

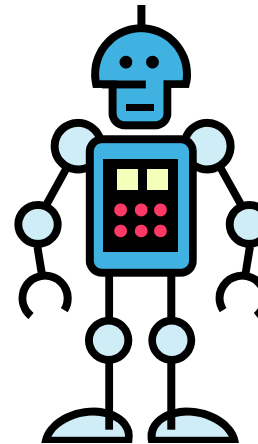
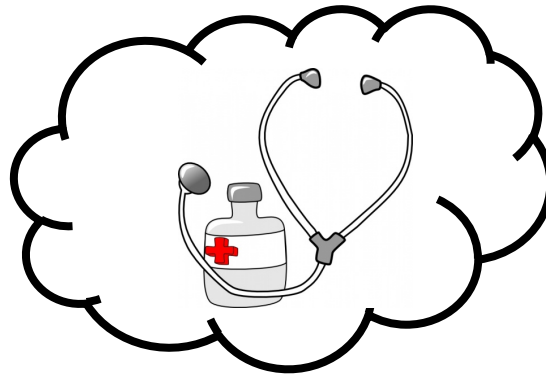
# Toward Robust Summarization of Agent Policies

Isaac Lage<sup>1</sup>, Daphna Lifschitz<sup>2</sup>,  
Finale Doshi-Velez<sup>1</sup> & Ofra Amir<sup>2</sup>

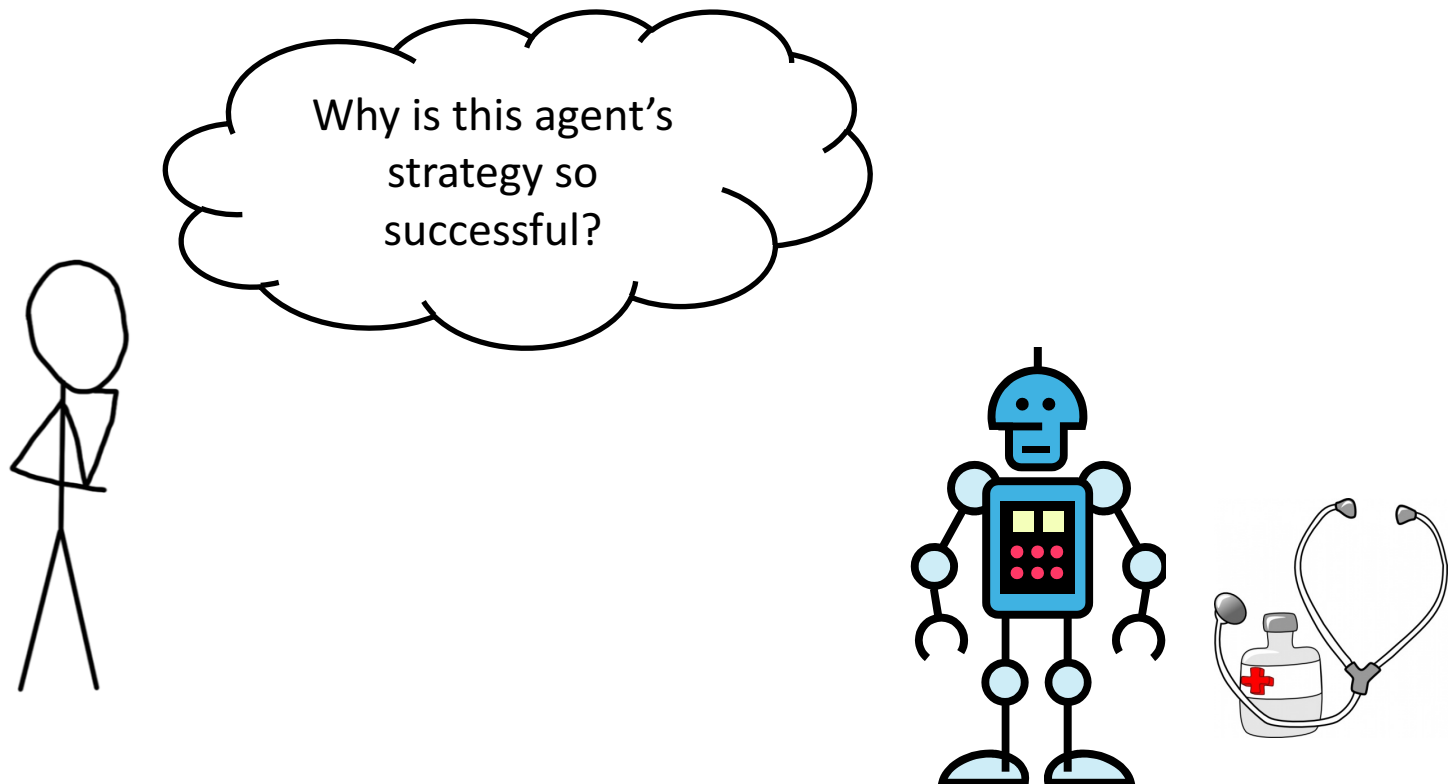
<sup>1</sup>Harvard University

<sup>2</sup>Technion – Israel Institute of Technology

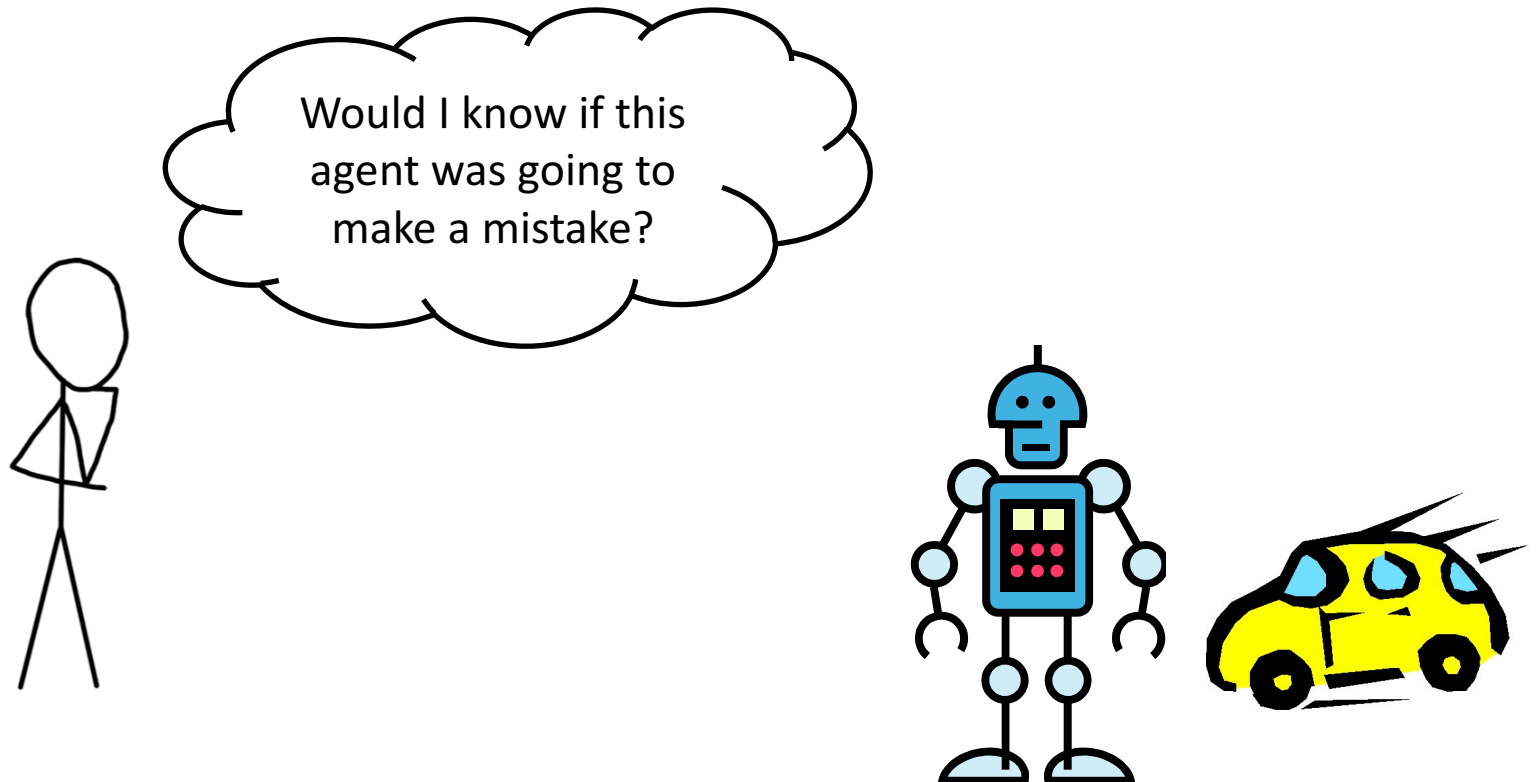
# Autonomous Agents



# Explaining Agent Behavior



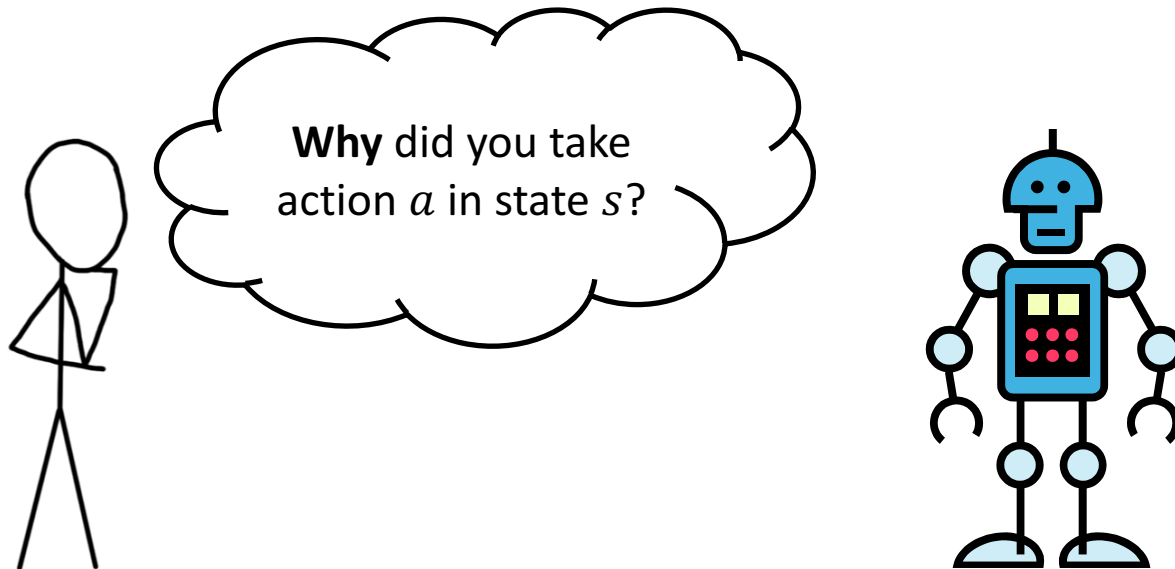
# Explaining Agent Behavior



# Related Work:

## *Explain Specific Agent Decisions*

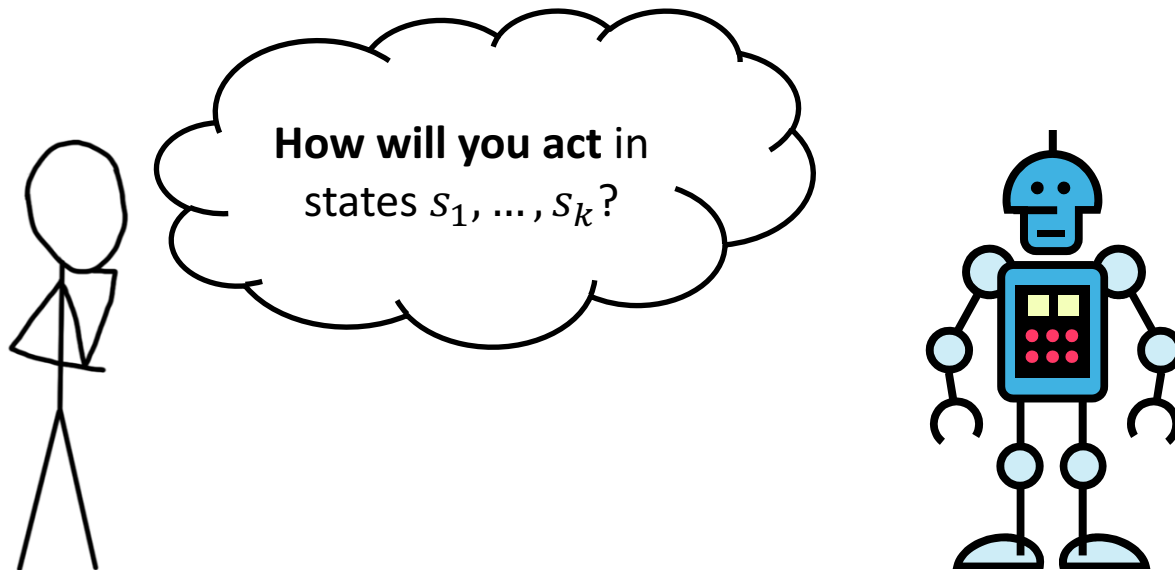
- “Explaining robot actions.” Lomas et al., 2012.
- “A natural language argumentation interface for explanation generation in Markov decision processes.” Dodson et al., 2011
- “Do you get it? user-evaluated explainable bdi agents.” Broekens et al., 2010.
- ...



# Related Work:

## *Describe Global Agent Behavior*

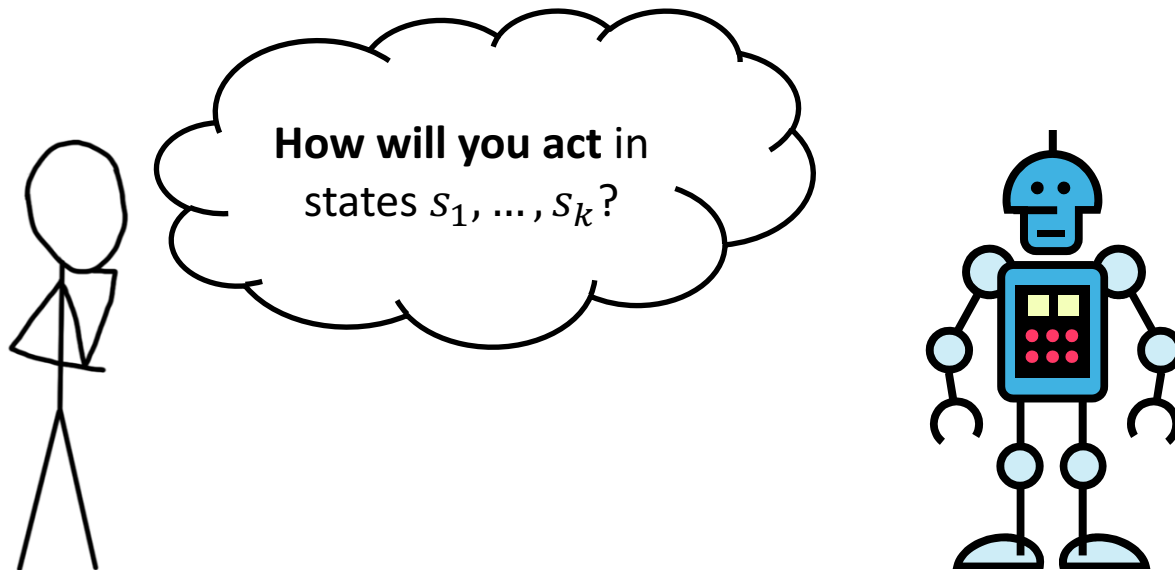
- Heuristics:
  - “Highlights: Summarizing agent behavior to people.” Amir and Amir, 2018
- Models of Humans:
  - “Enabling robots to communicate their objectives.” Huang et al. 2017.



# Related Work:

## *Describe Global Agent Behavior*

- Heuristics:
  - “Highlights: Summarizing agent behavior to people.” Amir and Amir, 2018
- **Models of Humans:**
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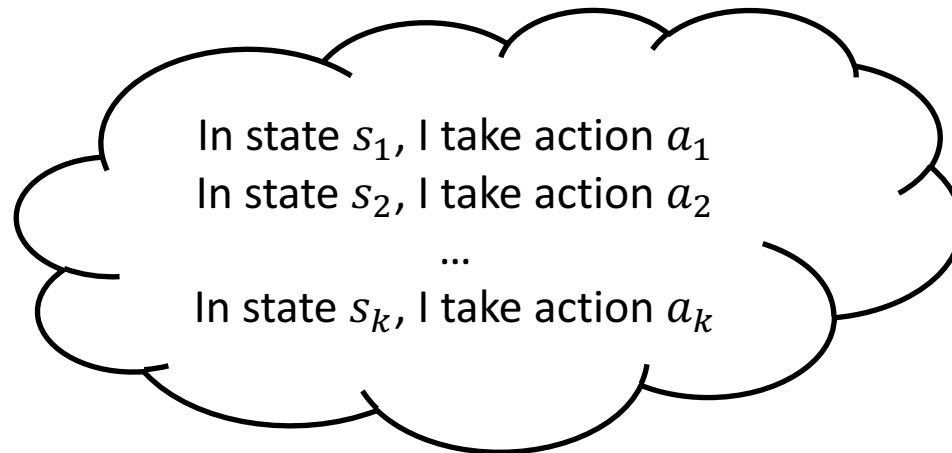
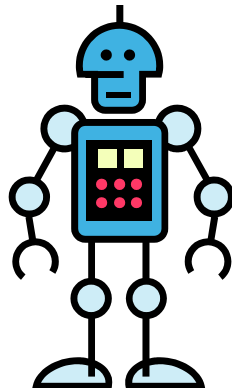
# Outline

- Policy Summarization
- Models of Humans
- Contributions
- Simulation Experiment
- Human Subject Experiment
- Open Challenges



# Policy Summary

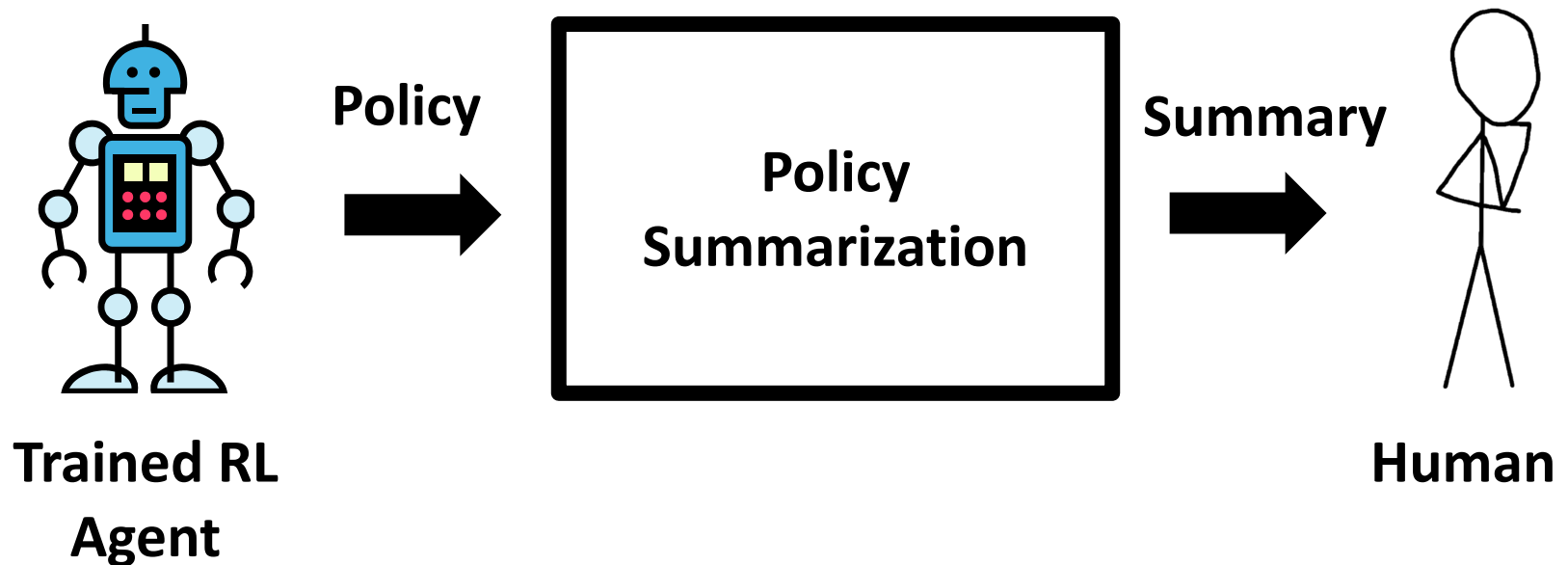
- A set of state-action pairs demonstrating the agent's behavior



$k$  is small to limit human effort

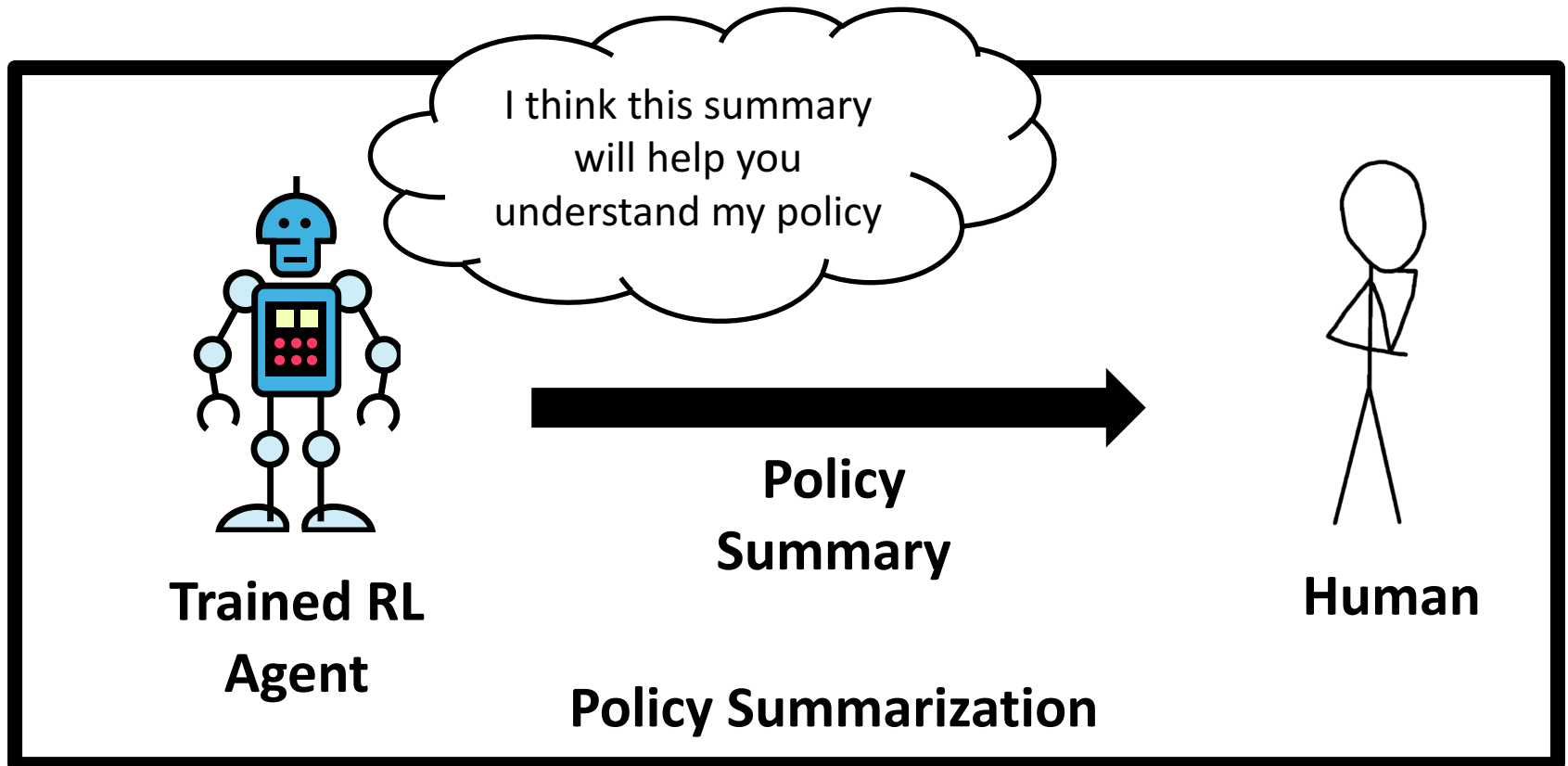
# Policy Summarization

- Goal: Describe the global behavior of an agent to a human



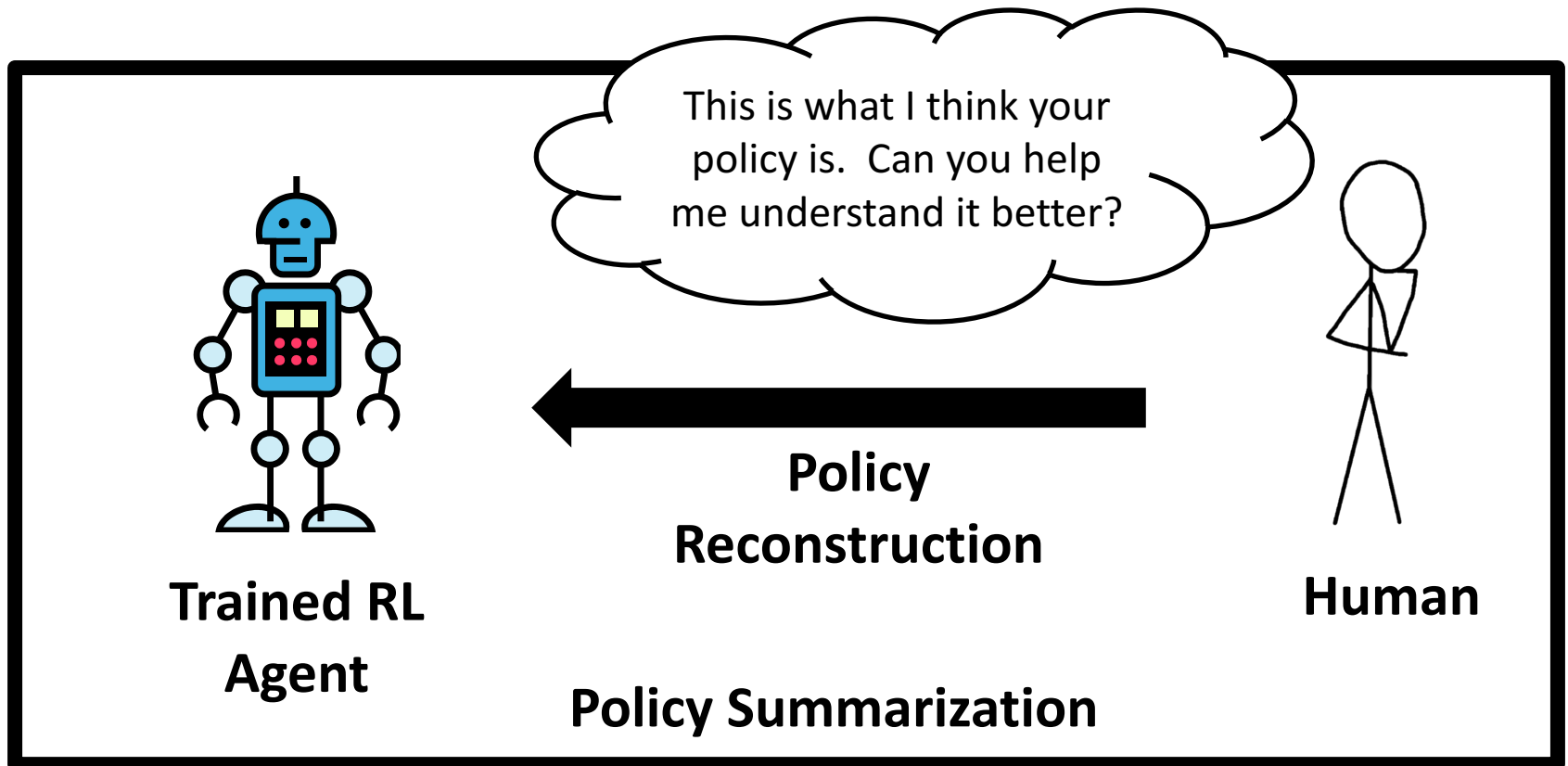
# Policy Summarization

- Goal: *Optimize* summary to describe agent behavior to *human*



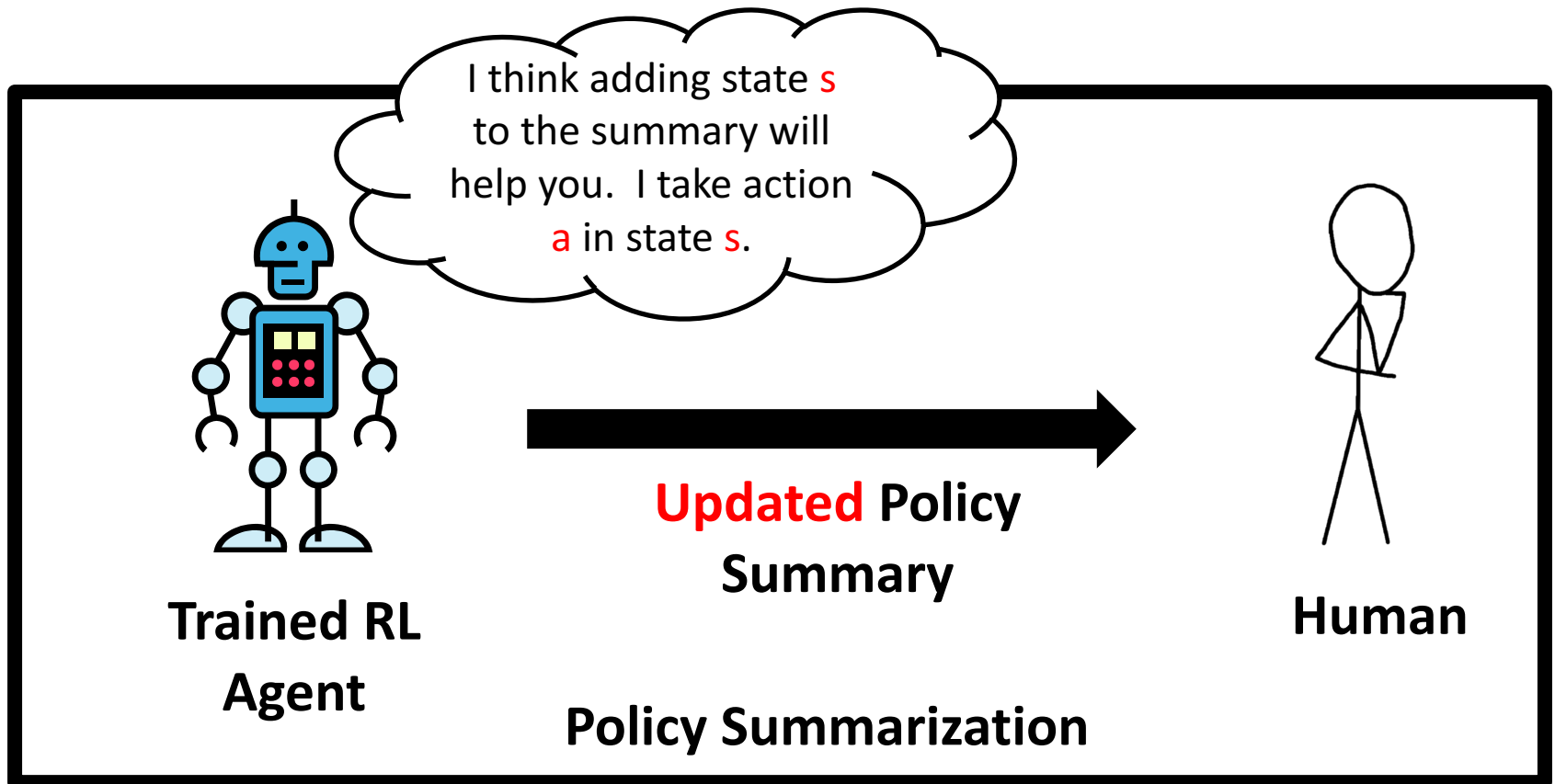
# Policy Summarization

- Goal: *Optimize* summary to describe agent behavior to *human*



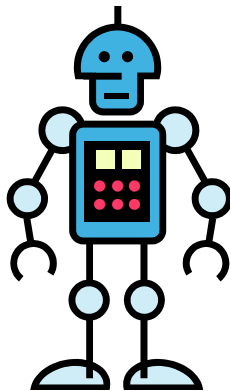
# Policy Summarization

- Goal: *Optimize* summary to describe agent behavior to *human*



# Policy Summarization

- Goal: *Optimize* summary to describe agent behavior to *human*



Trained RL  
Agent

**Problem: Hard to query  
humans at each step**

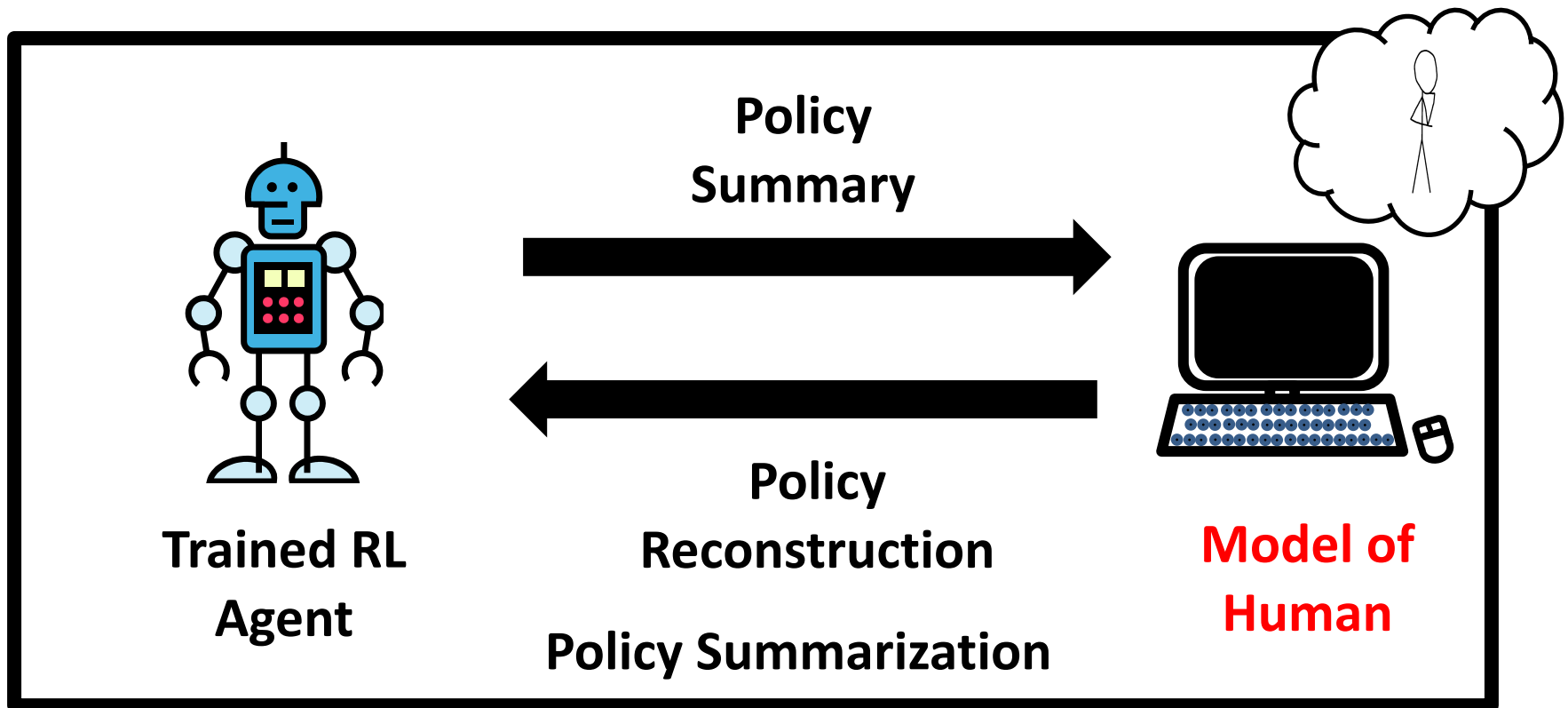


Human

Policy Summarization

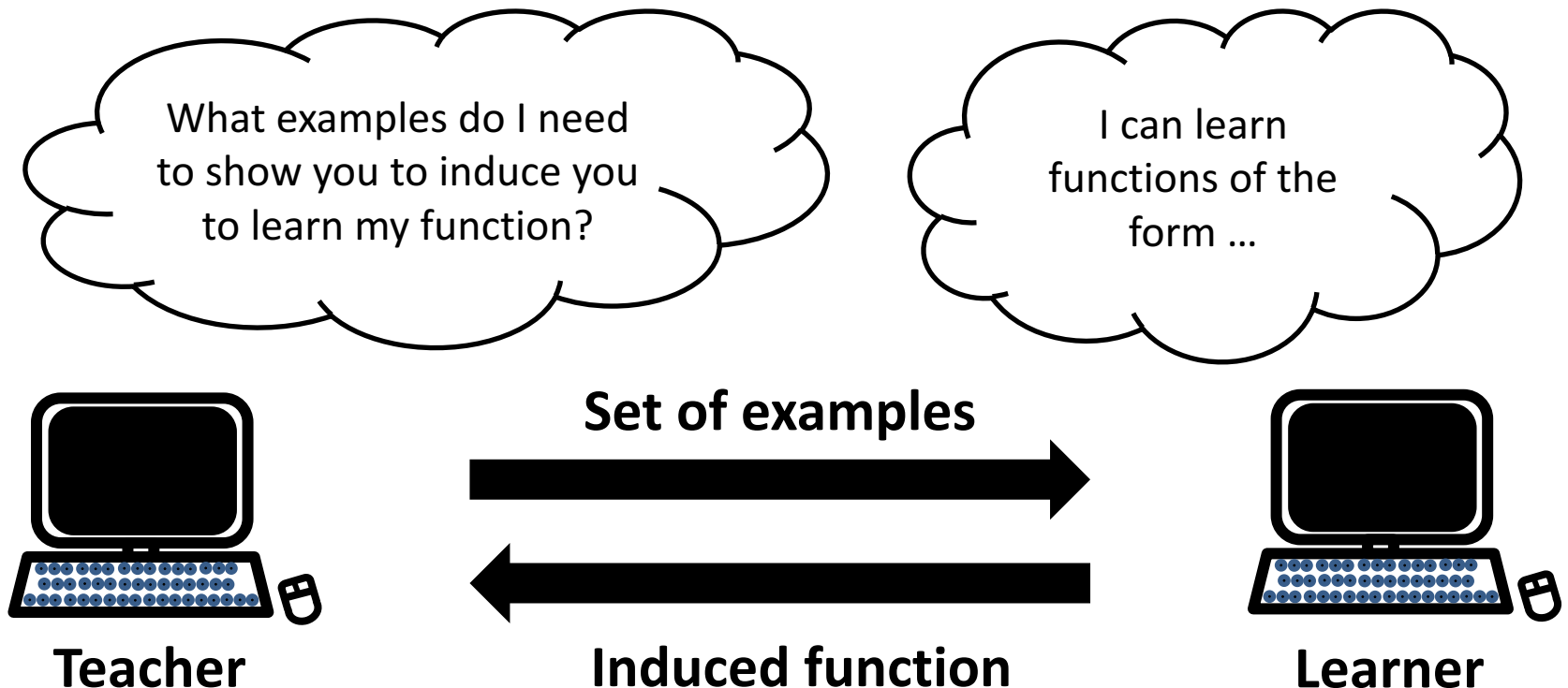
# Policy Summarization

- Goal: *Optimize* summary to describe agent behavior to *model of human*



# Mechanism: Machine Teaching

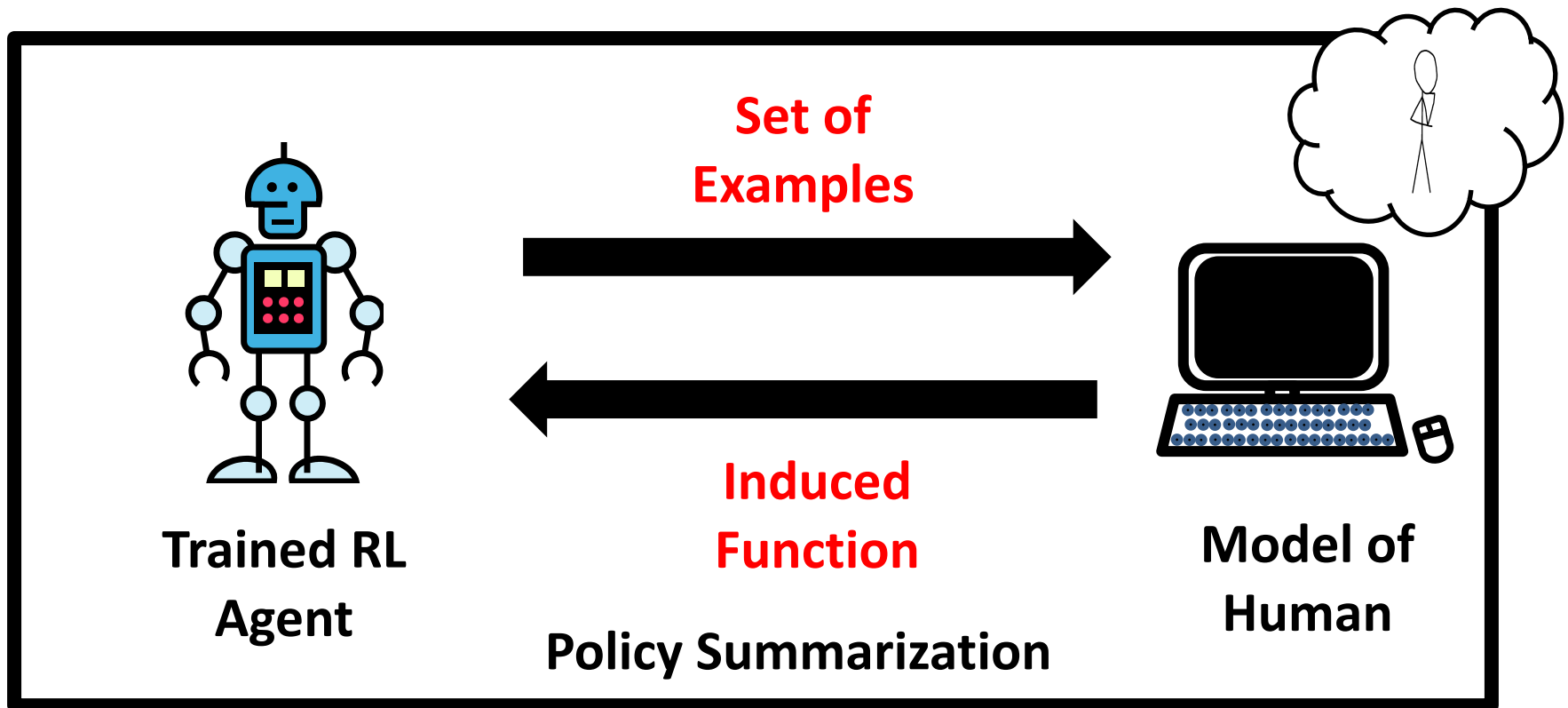
- Goal: Optimize a set of examples to induce a known model in a known class of learners.





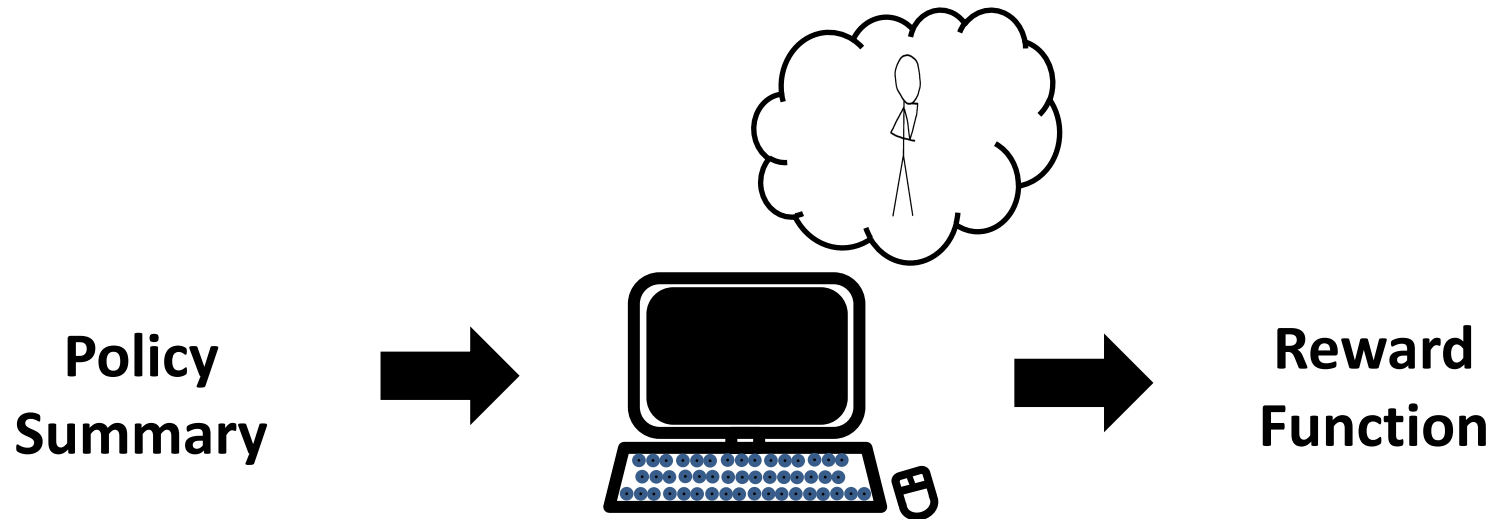
# Machine Teaching-Based Policy Summarization

- Goal: *Optimize* summary to describe agent behavior to *model of human*



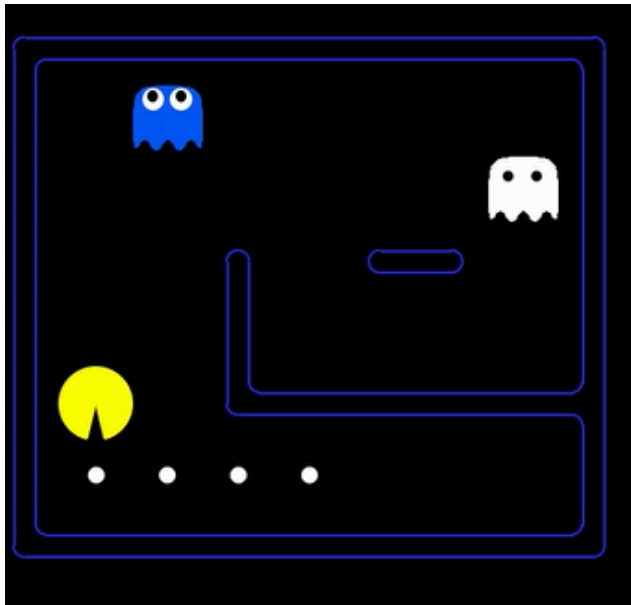
# Models of Humans – Inverse RL

- Goal: Extrapolate policy by first learning reward function



“Enabling robots to communicate their objectives.” Huang et al. 2017.

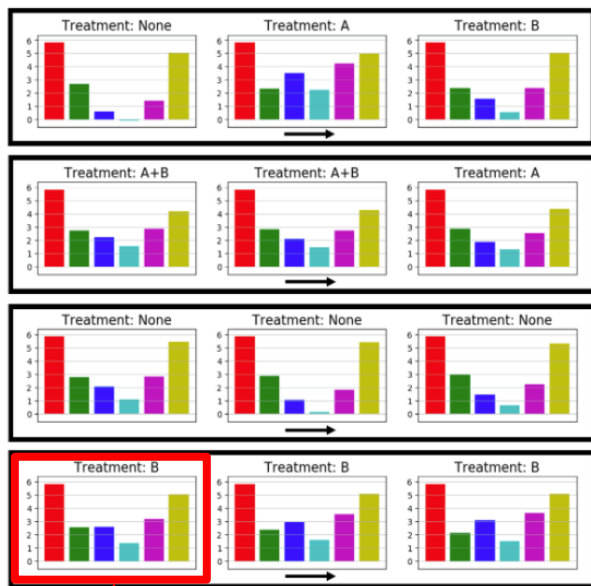
# IRL for PAC-MAN



Reward =  pellets +  ghosts



# IRL for HIV Simulator



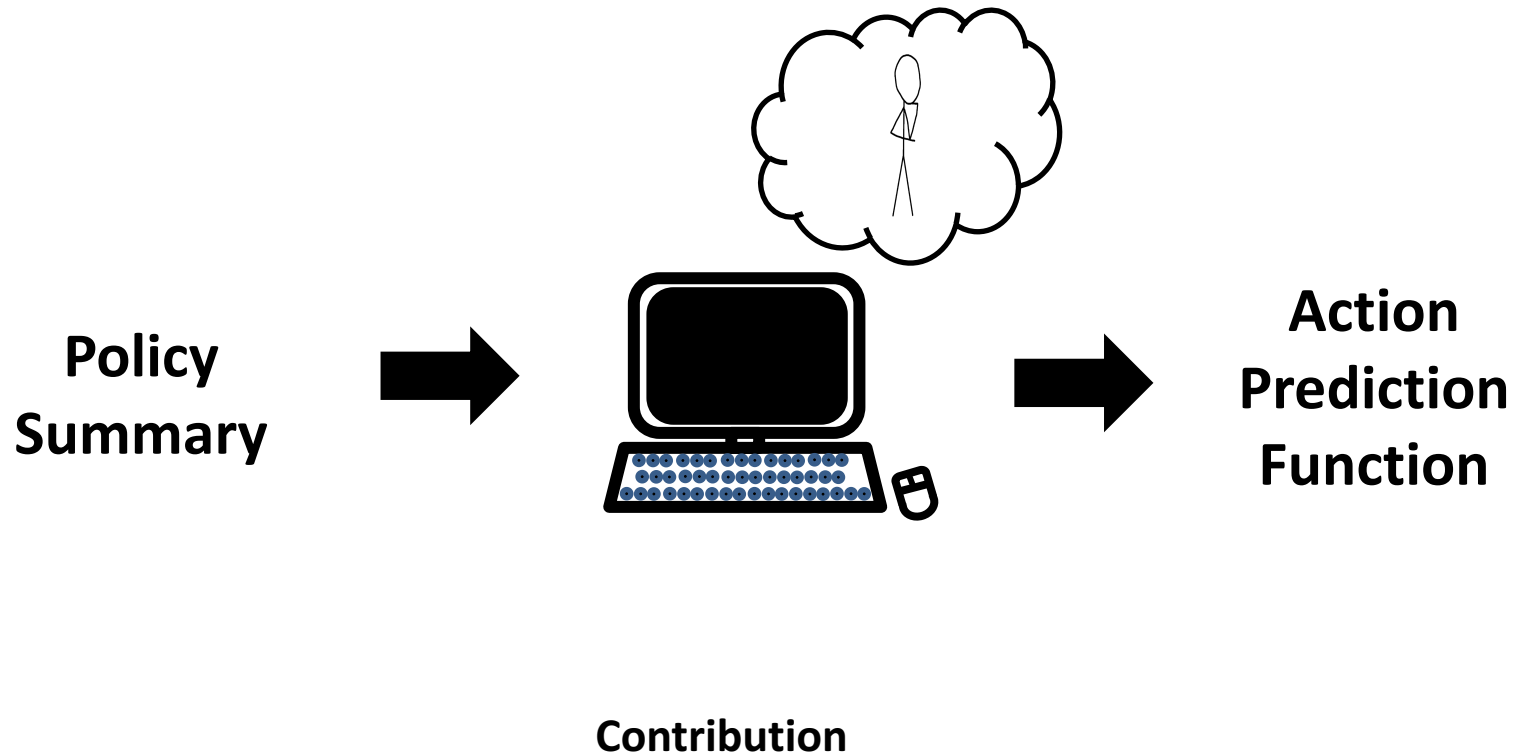
State space

Reward = ???

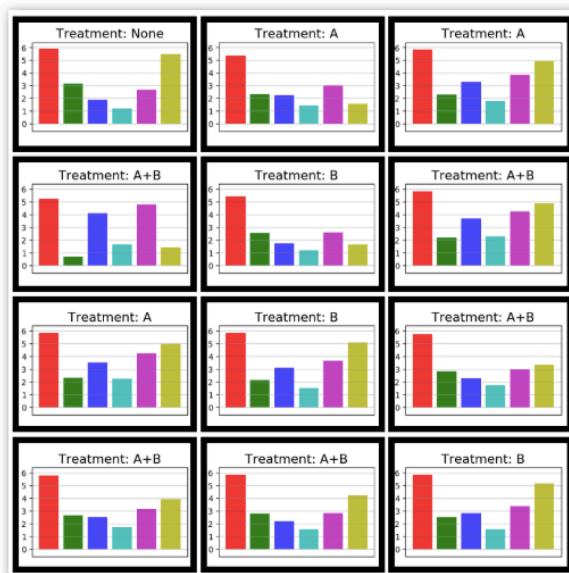


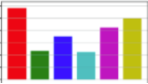
# Models of Humans – Imitation Learning

- Goal: Extrapolate policy by predicting the action taken in similar states



# IL for HIV Simulator

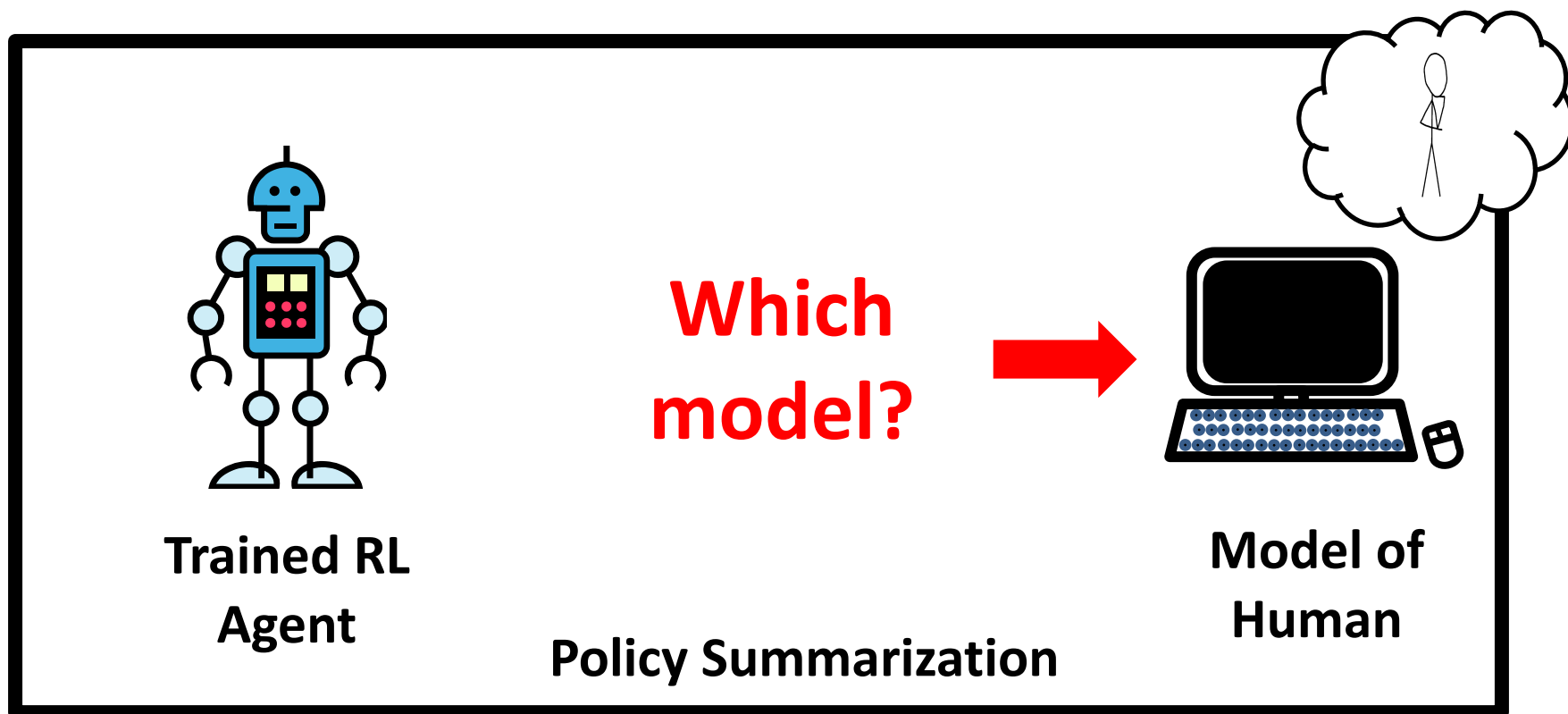


In states like , it gives  
**Treatment A**



# Contributions

- How should we model humans during policy summarization?

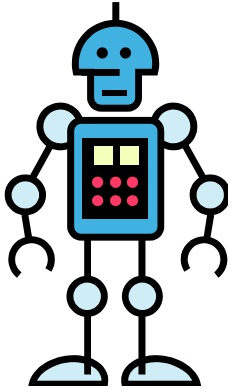


# Contributions

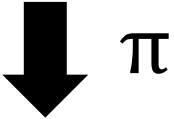
- Which reconstruction models do people use in different domains?
- Is it necessary to match the summarization model to peoples' reconstruction model to produce effective summaries?



# Evaluation



**Trained RL Agent**



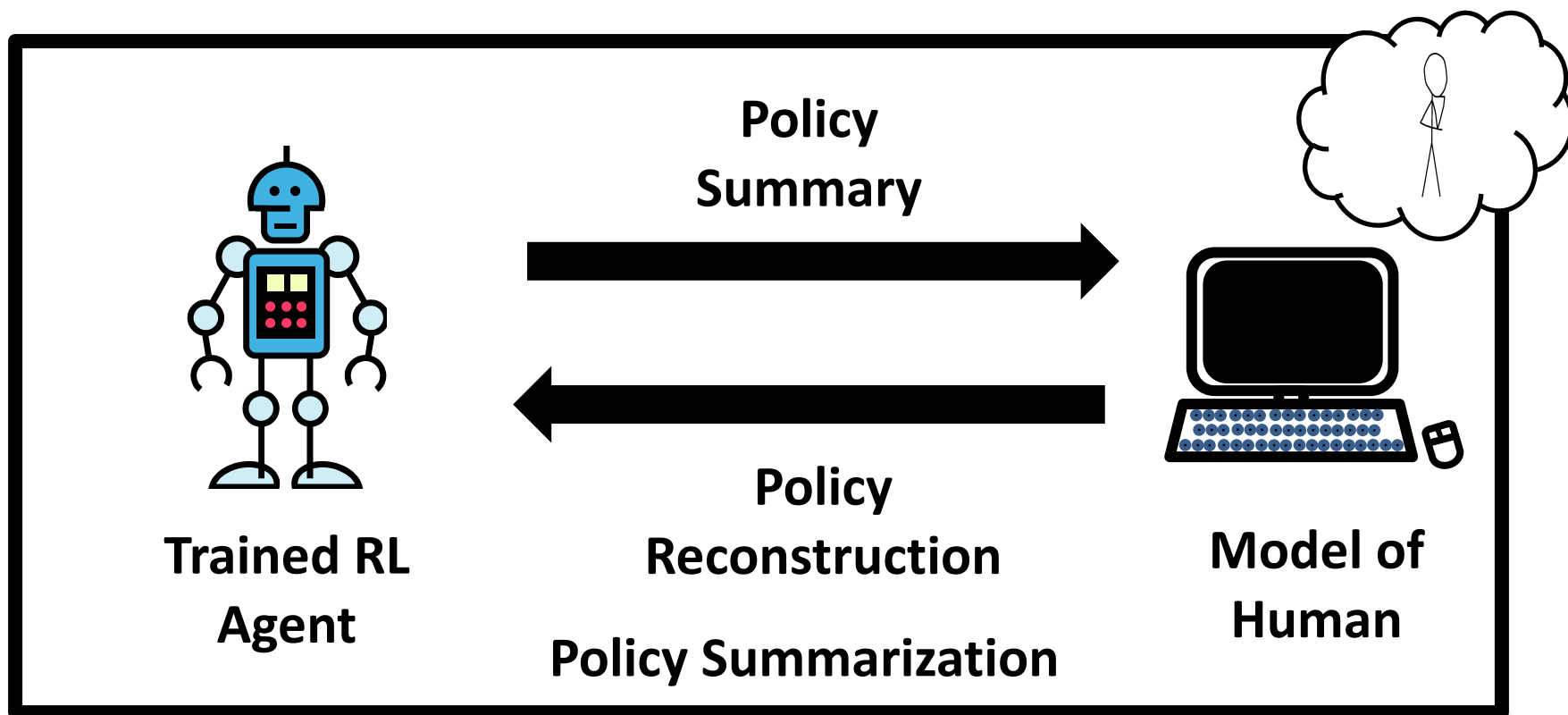
**Summary**



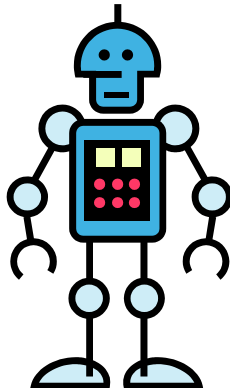
**Summarization**

# Recap: Policy Summarization

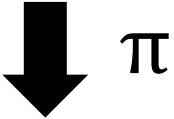
- Goal: *Optimize* summary to describe agent behavior to *model of human*



# Evaluation



Trained RL Agent



Summarization

Summary



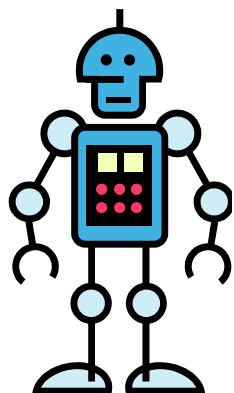
Human

$\hat{\pi}$

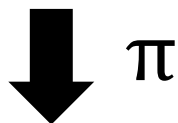


Reconstruction

# Evaluation



Trained RL Agent



Summarization

Summary



Human

Reconstruction

$\hat{\pi}$



Quality of Summary:

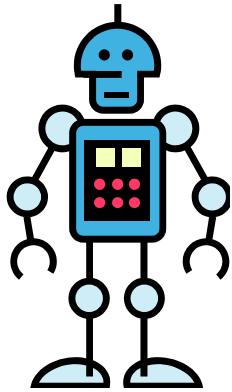
- $\text{Accuracy}(\hat{\pi}, \pi)$
- $\text{Value}(\pi) - \text{Value}(\hat{\pi})$

Evaluation

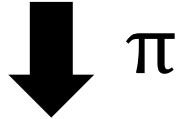
# Simulation Experiment

- Is it necessary to match the summarization model to the reconstruction model to produce high-quality reconstructions?

# Simulation Experiment



Trained RL Agent



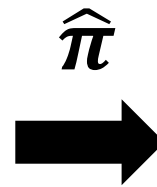
Summarization

Summary



Inverse  
RL  
Model

Reconstruction



Quality of Summary:

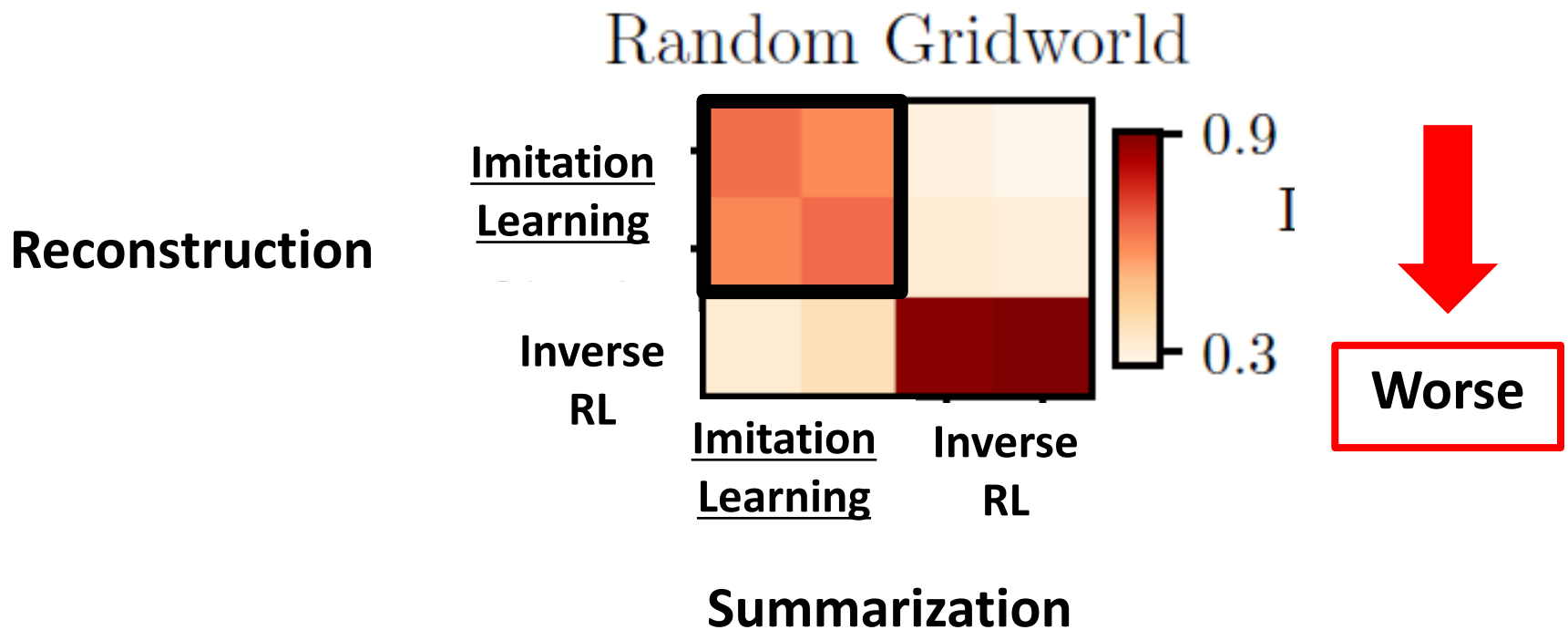
- Accuracy( $\hat{\pi}$ ,  $\pi$ )
- Value( $\pi$ ) - Value( $\hat{\pi}$ )

Evaluation

For example:

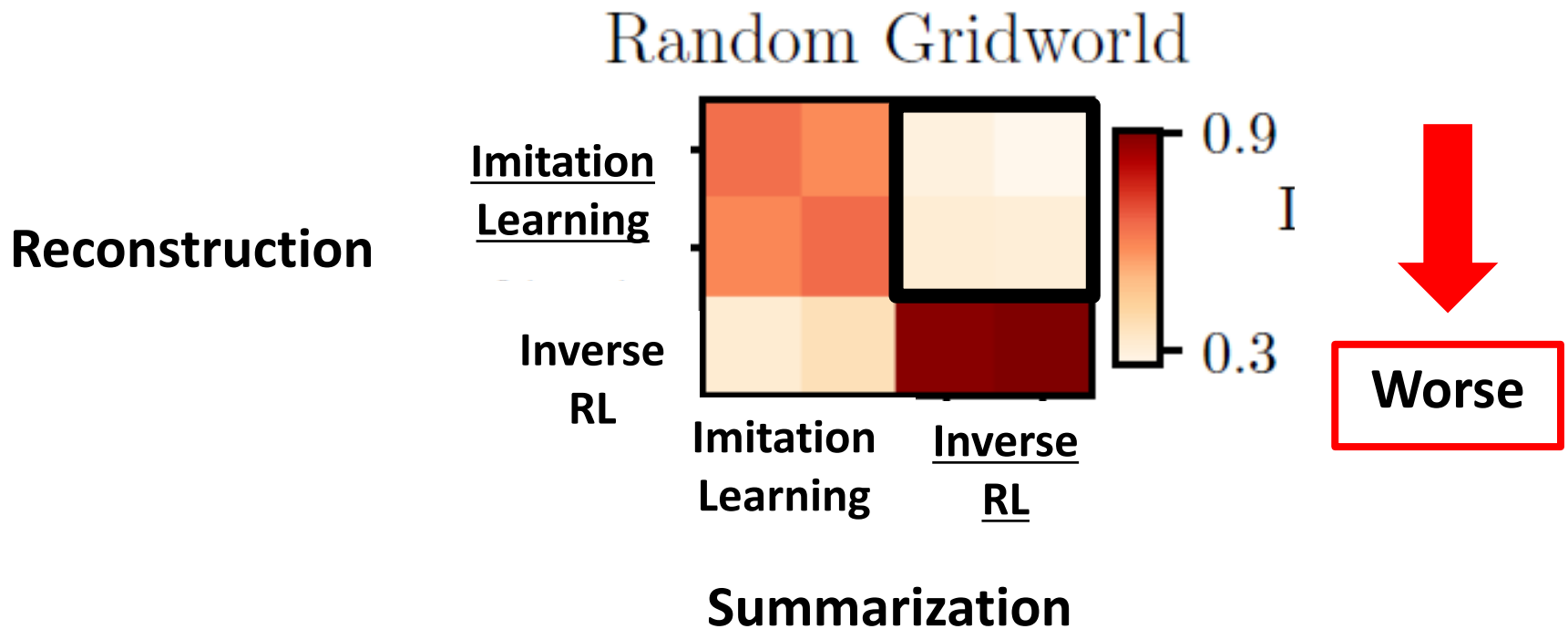
# Simulation Experiment

- Matched models: high policy-reconstruction quality



# Simulation Experiment

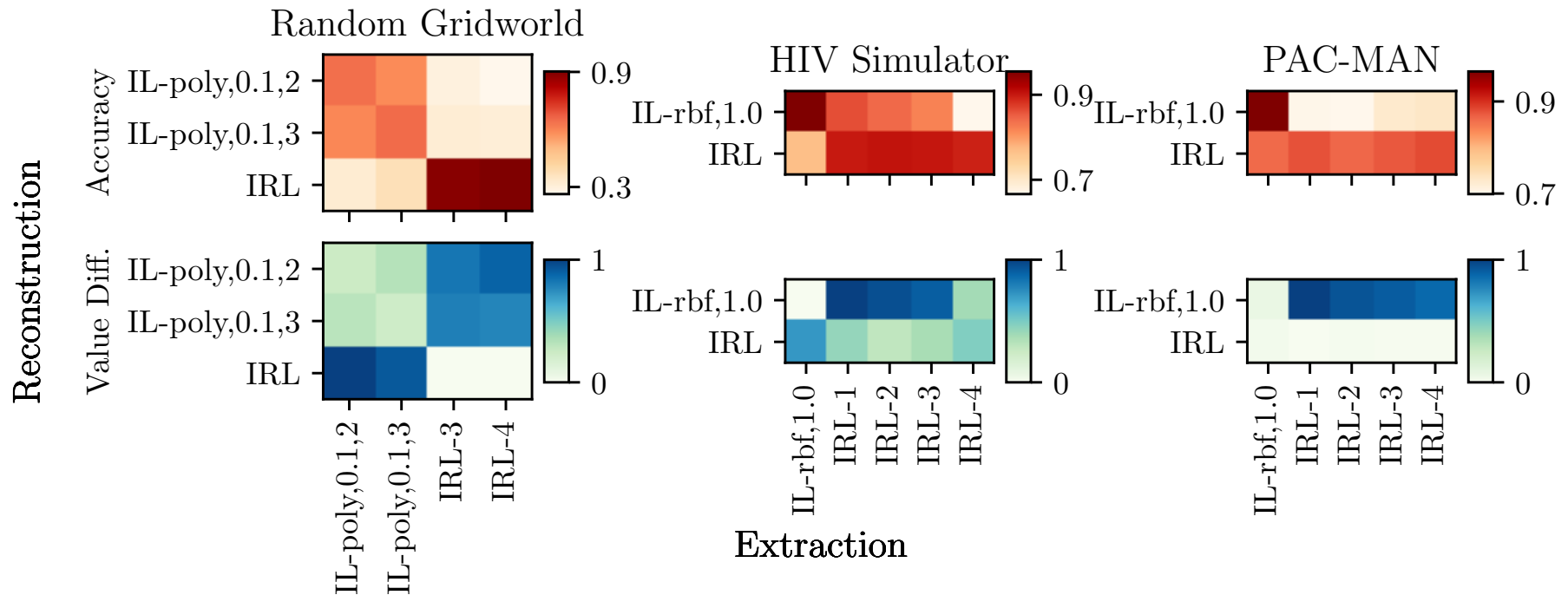
- Mismatched models: low policy-reconstruction quality





# Simulation Experiment

- Policy-reconstruction quality is worse when models don't match



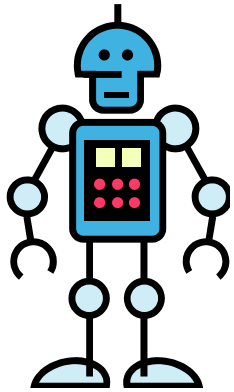
# Simulation Experiment

- Assuming a different model than the reconstruction model during summarization results in worse policy reconstructions.
- We shouldn't always use IRL in summarization if people sometimes do IL.

# Human Subject Experiment

- Which reconstruction models do people use?
- Does our finding about the importance of matching summarization models to reconstruction models hold for humans?

# Evaluation



Trained RL Agent



Summarization

Summary



Human

Reconstruction

$\hat{\pi}$



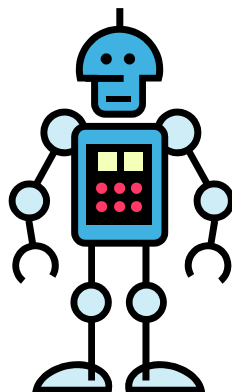
For example:

Quality of Summary:

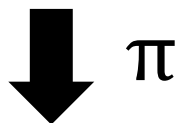
- Accuracy( $\hat{\pi}$ ,  $\pi$ )

Evaluation

# Evaluation



Trained RL Agent



Summarization

Reconstruction  
model = ???



Human

Summary



$\hat{\pi}$



Quality of Summary:

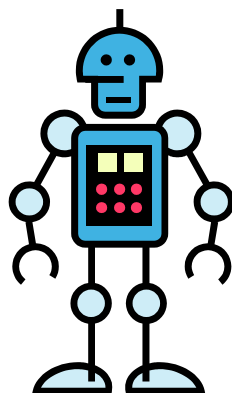
- Accuracy( $\hat{\pi}$ ,  $\pi$ )

Evaluation

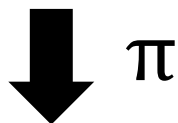
For example:

Reconstruction

# Evaluation



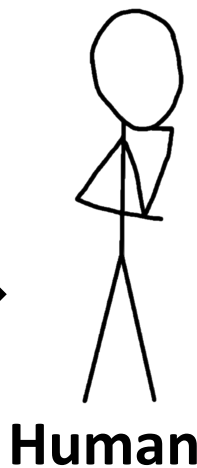
Trained RL Agent



**Imitation  
Learning**

Summarization

Summary



Human

Reconstruction

$\hat{\pi}$



For example:

**Quality of Summary:**

- Accuracy( $\hat{\pi}$ ,  $\pi$ )  
**for a sample of  $s$**

Evaluation

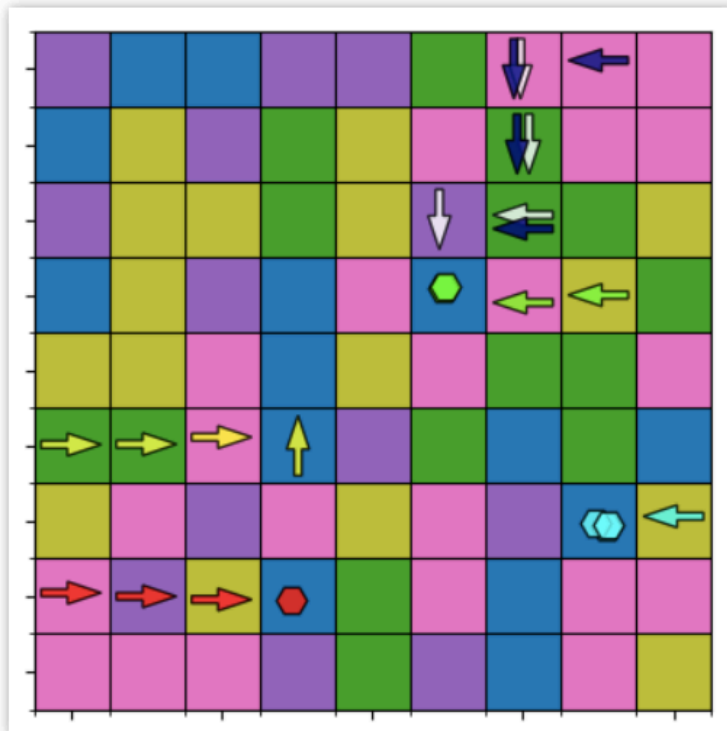
# Human Subject Experiment

Summarization Model			
		IL	IRL
Domain	HIV		
	Gridworld		

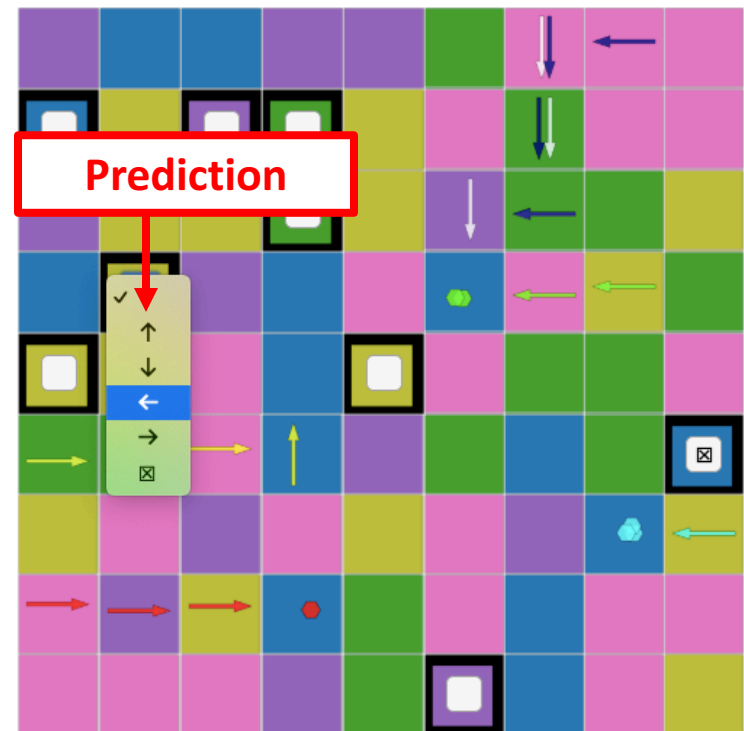
# Human Subject Experiment

- Task: predict the agent's behavior in a selection of states

Summary



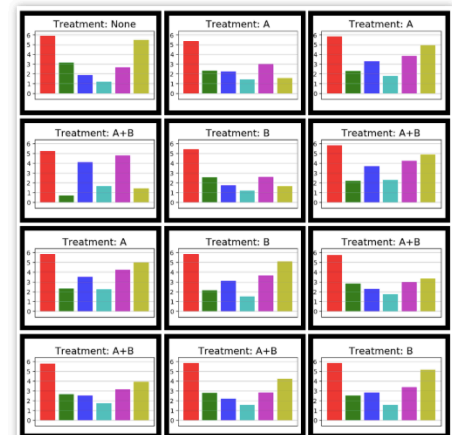
Task





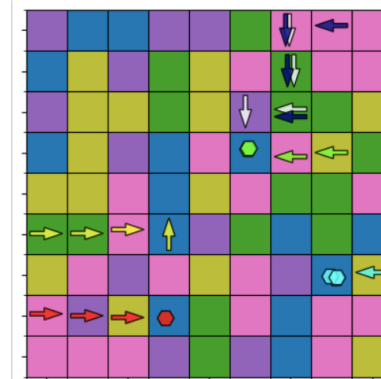
# Qualitative Reconstruction – HIV

- People reconstruct with IL in the HIV domain
  - **IL-based reconstruction: 78%**
    - E.g. *“I chose based on the similarity of the blood tests levels from the scenarios on the left.”*
  - **IRL-based reconstruction: 1 person**
    - E.g. *“Treatment A is used to decrease middle ones(blue, light blue and purple).”*



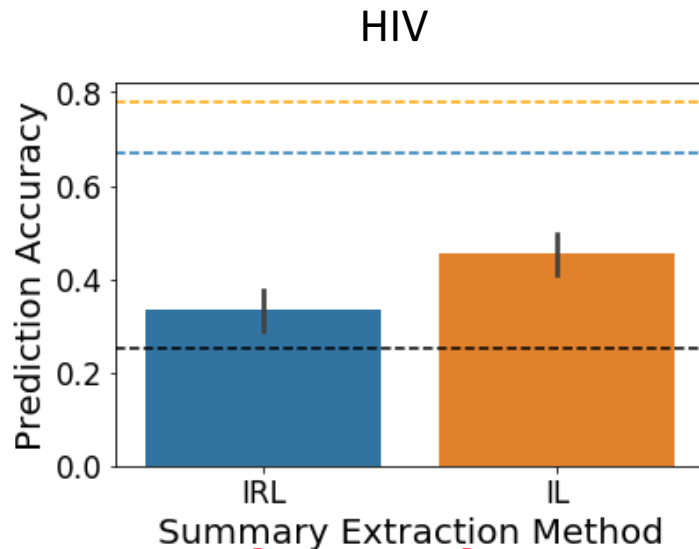
# Qualitative Reconstruction – Gridworld

- People reconstruct with IRL in the Gridworld
  - **IL-based reconstruction: 15%**
    - E.g. *“I tried comparing the colors and deciding which action was more frequent for a color.”*
  - **IRL-based reconstruction: 27%**
    - E.g. *“I decided that the computer seems to be always working towards a blue square. I chose the simplest path to get to a blue square”*

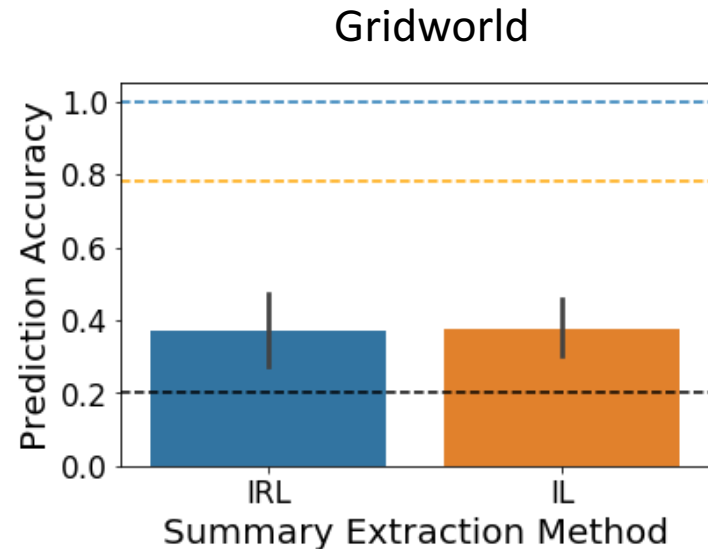


# Quantitative Reconstruction

- In the HIV domain, people reconstructed the policy more accurately with the matched IL summary
- In the gridworld domain, it made no difference



**Significant**

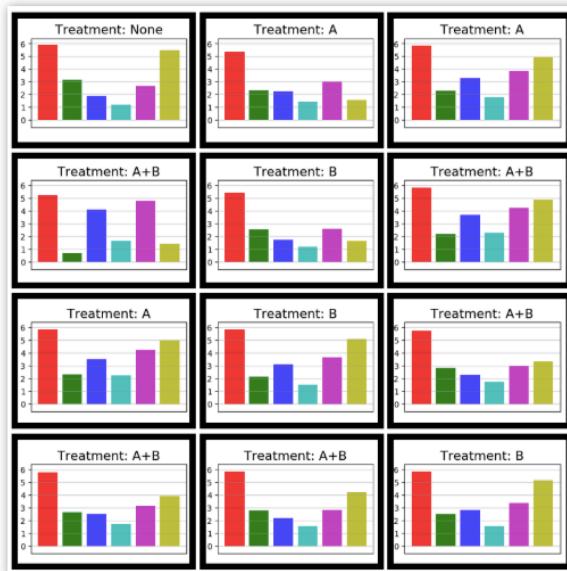



# Human Subject Experiment

- People use different models in different domains; it is important to consider IL as a possible model.
- Mismatch between the summarization model and peoples' reconstruction model can result in worse policy reconstructions.

# Our Contributions

- Introduced an imitation learning model for policy summarization



In states like , it gives  
**Treatment A**

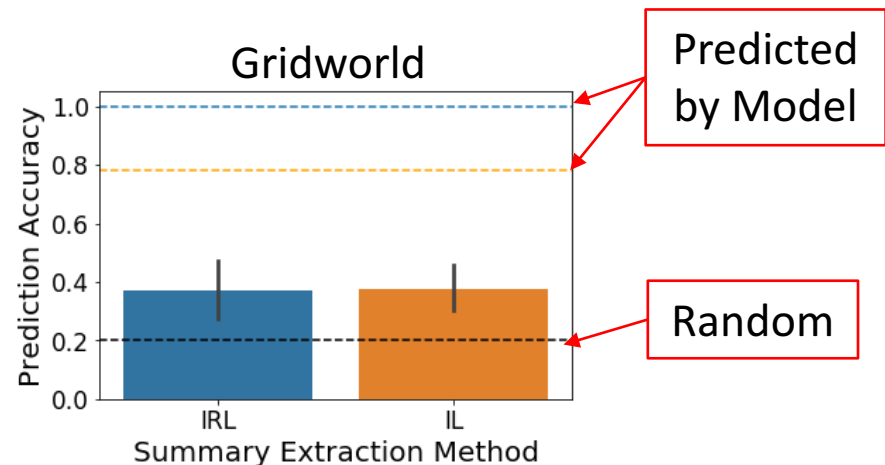
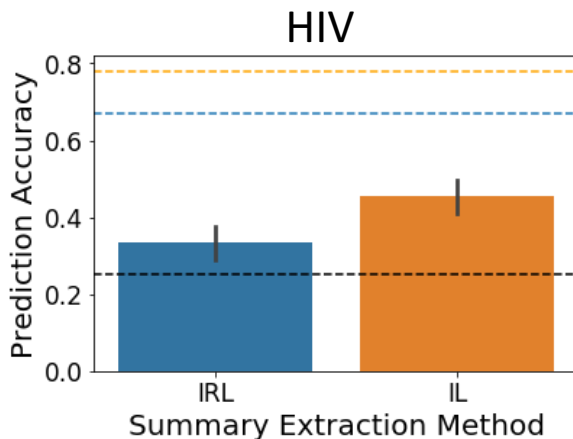


# Our Contributions

- Results:
  - Assuming a different model than the reconstruction model during summarization results in worse quality policy-reconstructions
  - People use different models in different domains; it is important to consider IL as a possible model

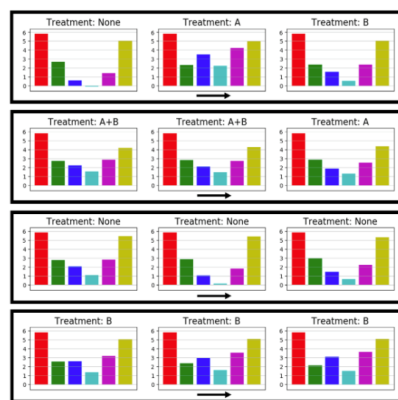
# Open Challenges

- Can we develop better models of how humans reconstruct policies?
  - Reconstruction was better than random but worse than predicted
    - Better similarity metrics?
    - Sparsity assumptions in IRL?



# Open Challenges

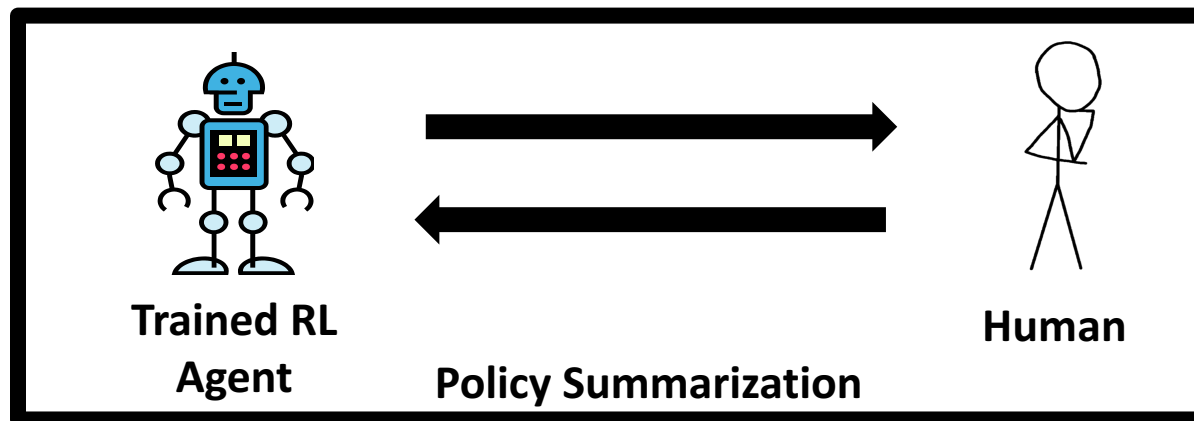
- How to scale these approaches to high dimensions?
  - People can only view a few dimensions
    - Can we develop better visualizations for high-dimensional data?
    - Can we build which dimensions to show into the models?





# Open Challenges

- Can we elicit peoples' reconstruction model?
  - Knowing reconstruction models is important
    - Can we discover properties of which domains correspond to which models?
    - Can we use human-in-the-loop approaches to learn peoples' reconstruction model during summarization?



# Open Challenges – Recap

- More nuanced models of humans
- Visualizing high-dimensional states
- Eliciting domain-specific reconstruction models through human-in-the-loop approaches