ICAPS 2021 Workshop on Knowledge Engineering for Planning and Scheduling

Learning User-Interpretable Descriptions of Black-Box AI System Capabilities





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How Would End Users Assess Their AI Systems?

- How would a lay user determine whether an Al agent will be safe/reliable for a certain task?
- More challenging in settings where agent's internal code is not available (black-box).
- Unproductive usage or safety risks in working with imperfect systems.
- Can we get insights from how we assess humans in such situations?



Do we learn Al system's action descriptions?

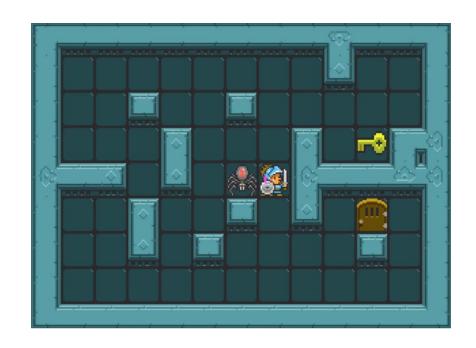


Agent Actions (Keystrokes)	Learned Capabilities
W	kill_monster
А	goto_door
S	goto_key goto monster
D	pick_key
E	open_door



Knowledge of primitive actions might be insufficient to understand the agent's capabilities

User-vocabulary may be limited



Agent's State Representation

pixel_1_1(#42A8B3) pixel_1_2(#42A8B3)

.

pixel n m(#203A3D)

Interpretable State Representation

monster_at(5,3)

player_at(6,3)

key_at(9,4)

door_at(9,2)



Might be more expressive than what the user understands

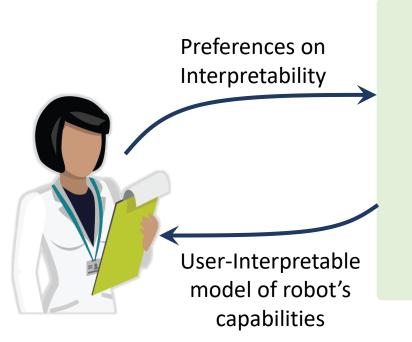
Related Work

Learning high-level symbolic models of AI systems using observations or interventions.

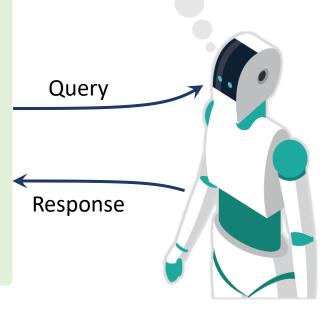
- Not interpretable, assume access to predefined options: Konidaris et al. (JAIR'18)
- Assume precise user-vocabulary: AIA Verma et al. (AAAI'21)
- Needs hand-coding of states: Zhang et al. (ICML'18)
- Require lot of data: Schema Networks Kansky et al. (ICML'17), Agarwal et al. (NIPS'16)
- Use passive observations: LOCM Cresswell et al. (ICAPS'09), ARMS Yang et al. (AIJ 2007), LOUGA Kučera and Barták (KMAIS 2018), FAMA Aineto et al. (AIJ 2019)

Our Approach: Symbolic Description Learner





Symbolic Description Learner



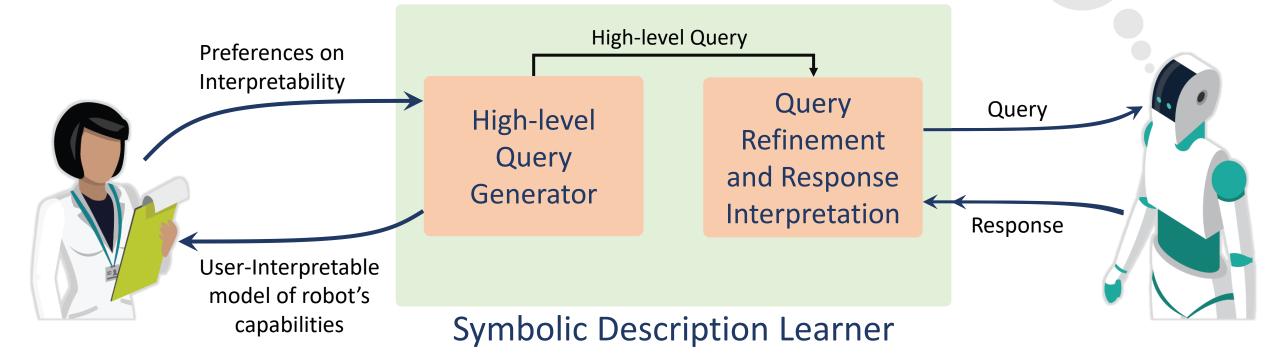
Our Approach: Symbolic Description Learner

In terms that the user understands. **Query**: Initial State and Plan **Response**: Length of plan that can be executed successfully and the final state

What do you think will happen if you don't have key and you move to a location near the door and then open the door?

I can execute only the first step. After this I will still be in a location near the door.



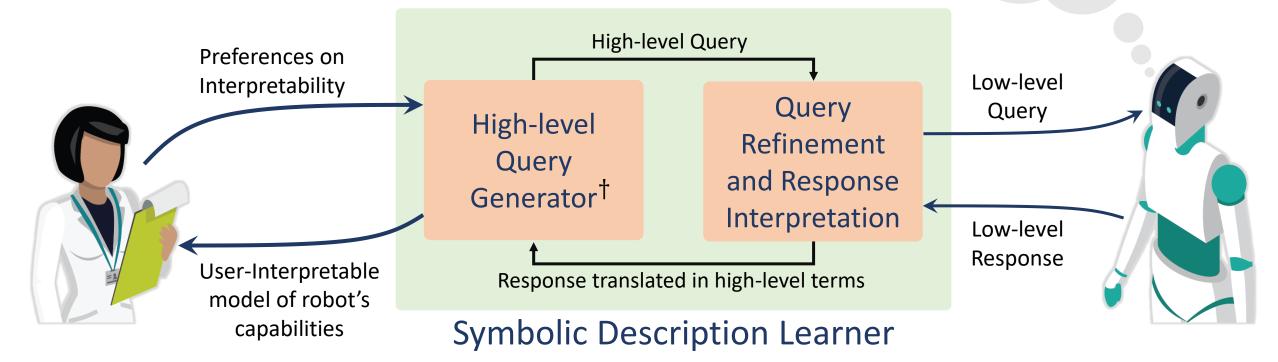


Our Approach: Symbolic Description Learner

In terms that the agent uses. **Query**: Initial State and Goal State **Response**: Yes/No. Representing if it can reach from the initial state to the goal state.

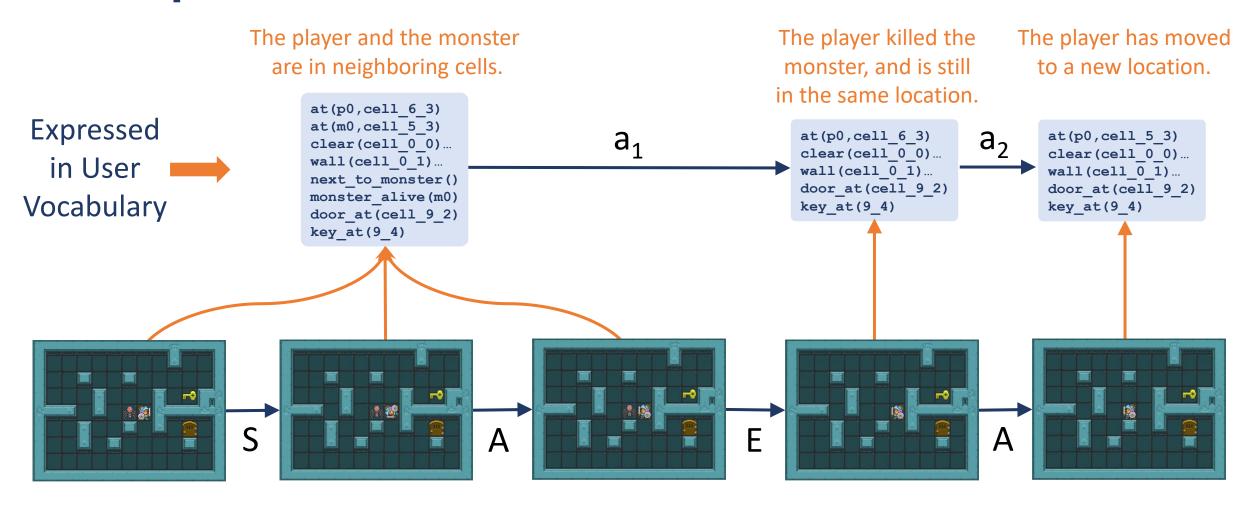
Can you reach from state <initial_state> to <goal_state>?

Yes/No



†Verma, P.; Marpally S. R.; and Srivastava, S. Asking the Right Questions: Learning Interpretable Action Models through Query Answering. In AAAI 2021.

Temporal Abstraction



Experimental Setup

- Randomly generate an environment from one of four GVGAI Games.
- Initialize two kinds of agents
 - Search Agent: Use search algorithms.
 - Policy Agent: Use black-box policies.
- Vary grid size to see variations in number of queries and time taken per query.

Results

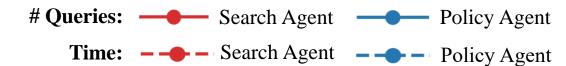
Some of the actions learned for Zelda

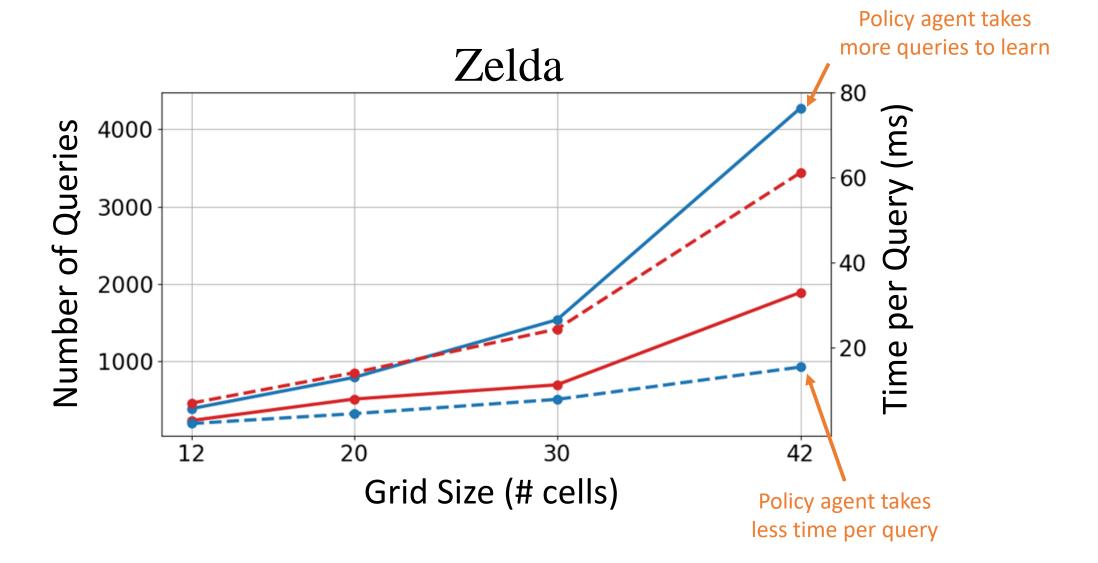


```
(:action kill-monster
  :parameters ()
  :precondition (and
        (at p1 6-3)
        (at m1 5-3)
        (monster_alive m1)
        (next_to_monster))
  :effect (and
        (clear 5-3)
        (not (at m1 5-3))
        (not (monster_alive m1))
        (not (next_to_monster))))
```

```
(:action escape-door
 :parameters ()
 :precondition (and
    (at p1 1-1)
    (at door 2-1)
    (clear 2-1)
    (has_key)
    (not (monster_alive m1)))
:effect (and
    (not (at p1 1-1))
    (not (clear 2-1))
    (clear 1-1)
    (escaped)
    (at p1 2-1)))
```

Results





Conclusions

The proposed approach:

- Efficiently learns internal model of an agent in a STRIPS-like form.
- Needs no prior knowledge of the agent model.
- Only requires an agent to have rudimentary query answering capabilities.
- Learns the model in terms of concepts that the user understands.



