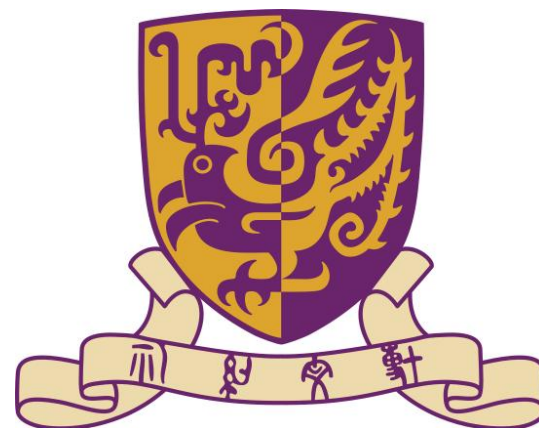


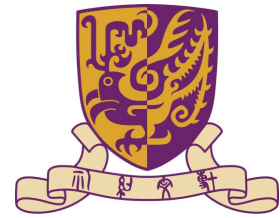


**ICLR**

# Enhancing Human-AI Collaboration Through Logic-Guided Reasoning

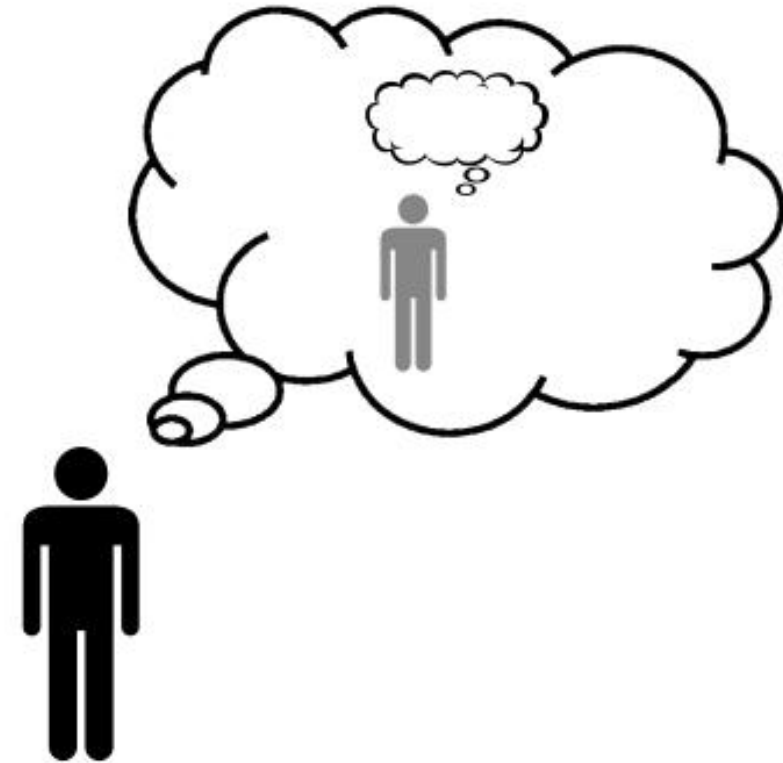
Chengzhi Cao, Yinghao Fu, Sheng Xu, Ruimao Zhang, Shuang Li



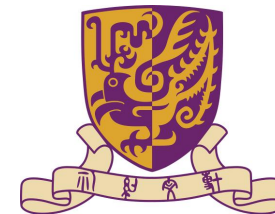
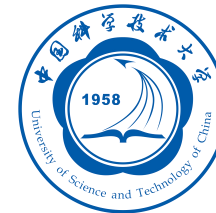


## Theory of Mind

*The ability to understand others' mental states and act upon them*



# Introduction



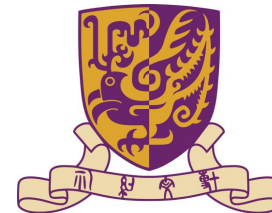
**Main agent:** setup a table



**Helper agent:** help on the **inferred** goal



# Introduction



**Spatial-Temporal Predicate** In our paper, we extend the above static predicates to spatial-temporal predicates, which include spatial-temporal *property* predicates and spatial-temporal *relation* predicates.

Specifically, the spatial-temporal *property* predicates are defined as

$$X(v) : \mathcal{C} \times \cdots \times \mathcal{C} \times \mathcal{T} \times \mathcal{S} \mapsto \{0, 1\}.$$

We will consider spatial-temporal logic rules where the body part contain spatial-temporal predicates as relation constraints. For example, a sensible rule will look like

$$f : Y_{\text{TurnAround}}(c, t, s) \leftarrow X_{\text{PickUpKey}}(c, t, s) \bigwedge R_{\text{InFront}}((c', t, s'), (c, t, s)) \bigwedge R_{\text{Behind}}((c'', t, s''), (c, t, s))$$

where  $c \in \mathcal{C}_{\text{person}}$ ,  $c' \in \mathcal{C}_{\text{block}}$ , and  $c'' \in \mathcal{C}_{\text{key}}$ . In general, the *spatial-temporal logic rule* in our paper is defined as a logical connectives of predicates, including property predicates and spatial-temporal relation predicates,

$$f : Y(v) \leftarrow \bigwedge_{X_{\text{property}} \in \mathcal{X}_f} X_{\text{property}}(v') \bigwedge_{R_{\text{spatial-temporal}} \in \mathcal{R}_f} R_{\text{spatial-temporal}}(v'', v) \quad (1)$$

where  $Y(v)$  is the *head predicate* evaluated at the entity-time-location triplet  $v$ ,  $\mathcal{X}_f$  is the set of property predicates defined in rule  $f$ , and  $\mathcal{R}_f$  denotes the set of spatial-temporal relation predicates defined in rule  $f$ .

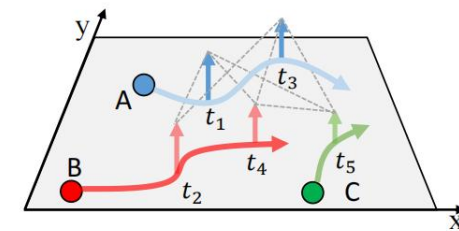
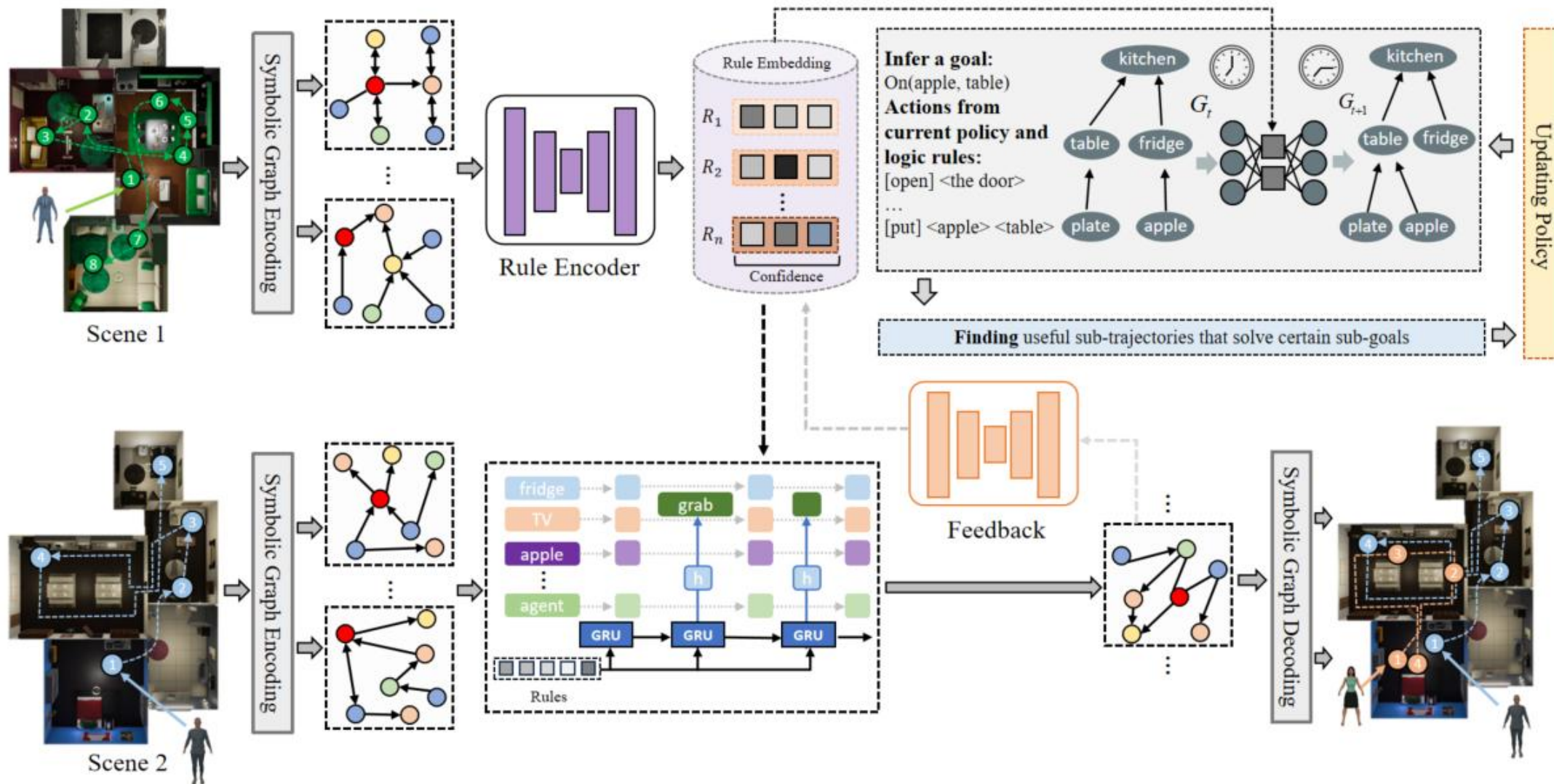
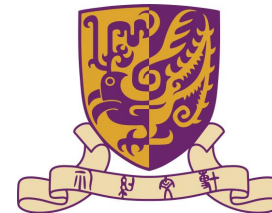
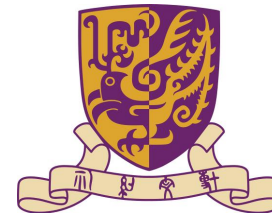


Figure 1: Illustration of feature construction using a simple logic formula with temporal relation predicate ( $t_1 < t_2$ ),  $f : Y \leftarrow A \wedge B \wedge C \wedge (A \text{ Before } B)$ . The rule defines the template to gather combinations of the body predicate history events. Here predicate A has 2 events and predicate B has 1 event, the temporal relation constraint would lead to valid combinations (also called "paths"). This type of feature construction can be extended to spatial-temporal cases, where we count the valid paths as the feature.

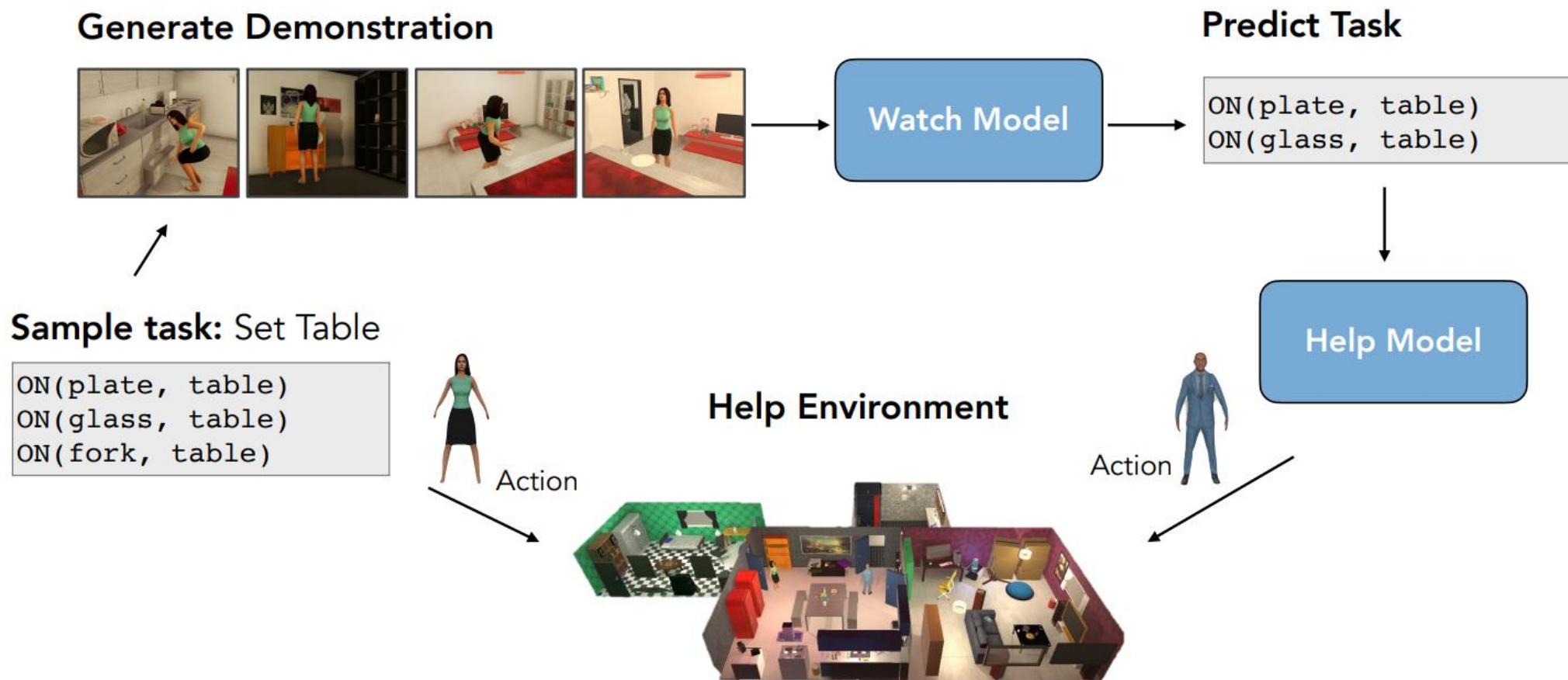
# Framework



# Experiment



## Watch-and-Help Dataset



# Result

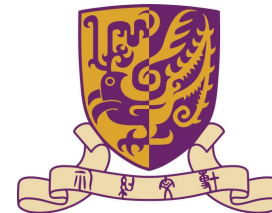
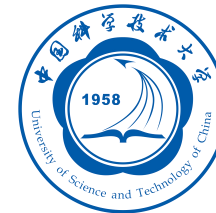


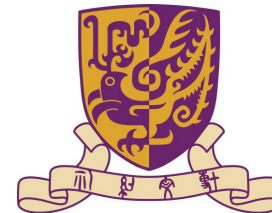
Table 1: Comparative performance of various rule encoder backbones on the Watch-and-Help Dataset: An evaluation across four difficulty tiers using three metrics—Average Number of Moves (AN, lower is better) for successful episodes, Success Rate (SR, higher is better); and Speedup (SU, higher is better).

Methods		Partially Observable			
		Level 1	Level 2	Level 3	Level 4
WAH	AN ↓	15.28±0.06	22.24±0.15	28.88±0.15	51.20±0.12
	SR ↑	78.61±0.02	71.02±0.08	60.13±0.11	51.48±0.04
	SU ↑	0.17±0.012	0.14±0.019	0.13±0.016	0.08±0.016
GEM	AN ↓	14.08±0.04	20.40±0.10	28.71±0.05	42.12±0.04
	SR ↑	79.27±0.02	71.32±0.04	60.99±0.07	53.97±0.06
	SU ↑	0.24±0.017	0.22±0.016	0.19±0.012	0.15±0.015
DRRN	AN ↓	13.47±0.13	20.10±0.14	28.68±0.15	38.25±0.01
	SR ↑	80.86±0.09	71.74±0.06	66.27±0.13	57.53±0.18
	SU ↑	0.25±0.015	0.19±0.017	0.17±0.004	0.16±0.009
Seq2Seq	AN ↓	12.76±0.06	19.92±0.06	28.32±0.02	34.16±0.04
	SR ↑	81.12±0.01	72.45±0.18	66.69±0.18	57.86±0.09
	SU ↑	0.38±0.003	0.28±0.008	0.26±0.007	0.21±0.014
ReQueST	AN ↓	12.68±0.03	18.72±0.10	28.12±0.12	32.92±0.18
	SR ↑	81.12±0.05	77.44±0.17	68.52±0.04	55.34±0.07
	SU ↑	0.38±0.019	0.30±0.015	0.25±0.009	0.22±0.011
Ours	AN ↓	11.40±0.04	16.21±0.15	26.16±0.02	31.10±0.05
	SR ↑	82.43±0.17	78.14±0.03	69.84±0.17	58.91±0.16
	SU ↑	0.48±0.015	0.44±0.009	0.36±0.013	0.33±0.017

Table 2: Experiment results on HandMeThat dataset. Each model is evaluated on 4 hardness levels with 3 metrics: the Average Score (AS, higher is better), the Success Rate (SR, higher is better), and the Average Number of Moves (AN, lower is better) in successful episodes.

Methods		Partially Observable			
		Level 1	Level 2	Level 3	Level 4
Random	AS ↑	-40.0	-40.0	-40.0	-40.0
	SR ↑	0.0	0.0	0.0	0.0
	AN ↓	N/A	N/A	N/A	N/A
Seq2Seq	AS ↑	-5.1±0.07	-25.3±0.09	-34.5±0.15	-32.0±0.20
	SR ↑	25.50±0.20	10.40±0.14	3.95±0.23	5.30±0.26
	AN ↓	4.21±0.01	4.17±0.04	4.19±0.02	4.12±0.06
GEM	AS ↑	-6.3±0.11	-22.5±0.12	-30.9±0.14	-29.6±0.18
	SR ↑	24.64±0.40	12.47±0.21	5.74±0.15	7.21±0.13
	AN ↓	4.34±0.04	4.36±0.02	4.24±0.00	4.22±0.00
DRRN	AS ↑	-40.0	-40.0	-40.0	-40.0
	SR ↑	0.0	0.0	0.0	0.0
	AN ↓	N/A	N/A	N/A	N/A
Ours	AS ↑	-1.4±0.03	-5.8±0.06	-9.1±0.17	-11.9±0.23
	SR ↑	27.73±0.29	24.81±0.36	20.97±0.12	21.66±0.18
	AN ↓	4.05±0.01	4.09±0.01	4.14±0.03	4.21±0.02

# Visualization



Rule:  $\text{Walk\_to}(\text{Bob}, \text{plate}) \wedge \text{Grab}(\text{Bob}, \text{plate}) \wedge \text{Walk\_to}(\text{Bob}, \text{microwave}) \wedge \text{Open}(\text{Bob}, \text{microwave}) \rightarrow \text{Inside}(\text{plate}, \text{microwave})$

Explanation:

- **Bob** walks towards the **plate** and grabs it
- **Bob** walks towards the **microwave**
- **Alice** infers his goal as “put the **plate** into the **microwave**”
- **Alice** opens the **microwave**



Rule:  $\text{Walk\_to}(\text{Bob}, \text{bedroom}) \wedge \text{Switch\_on}(\text{Alice}, \text{TV}) \wedge \text{Walk\_to}(\text{Bob}, \text{sofa}) \rightarrow \text{Watch}(\text{Bob}, \text{TV})$

Explanation:

- **Bob** walks towards the bedroom
- **Bob** walks towards the **sofa**
- **Alice** infers his goal as “watch TV”
- **Alice** switches on TV



Rule:  $\text{Walk\_to}(\text{Bob}, \text{sofa}) \wedge \text{Block}(\text{Alice}, \text{Bob}) \wedge \text{Make\_way}(\text{Alice}, \text{Bob}) \rightarrow \text{Pass\_by}(\text{Bob}, \text{Alice})$

Explanation:

- **Bob** walks towards the **sofa**
- **Alice** blocks **Bob's** way by chance
- **Alice** makes way for **Bob** so he can walk quickly
- **Bob** walks by Alice's side

Figure 4: Visualization and explanation of logic rules in Watch-and-help dataset.



Thanks for Listening!