



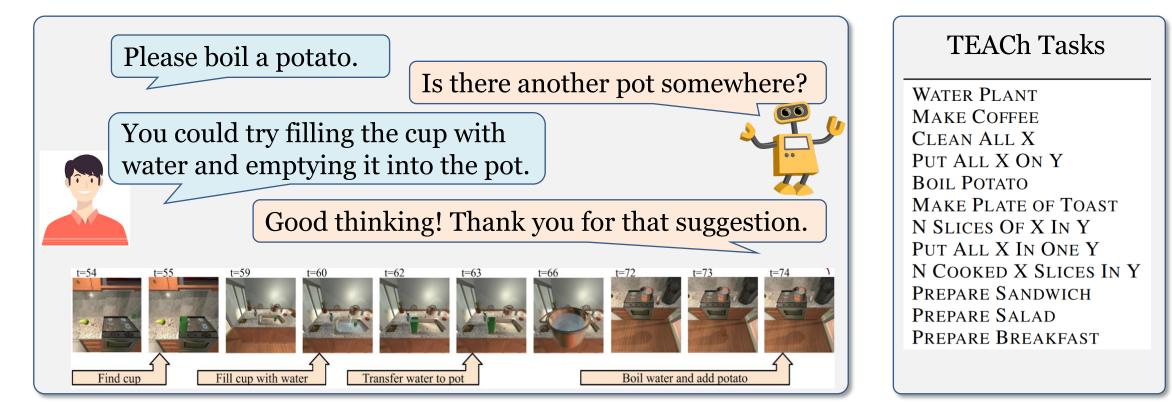
DANLI: Deliberative Agent for Following Natural Language Instructions

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Problem Definition

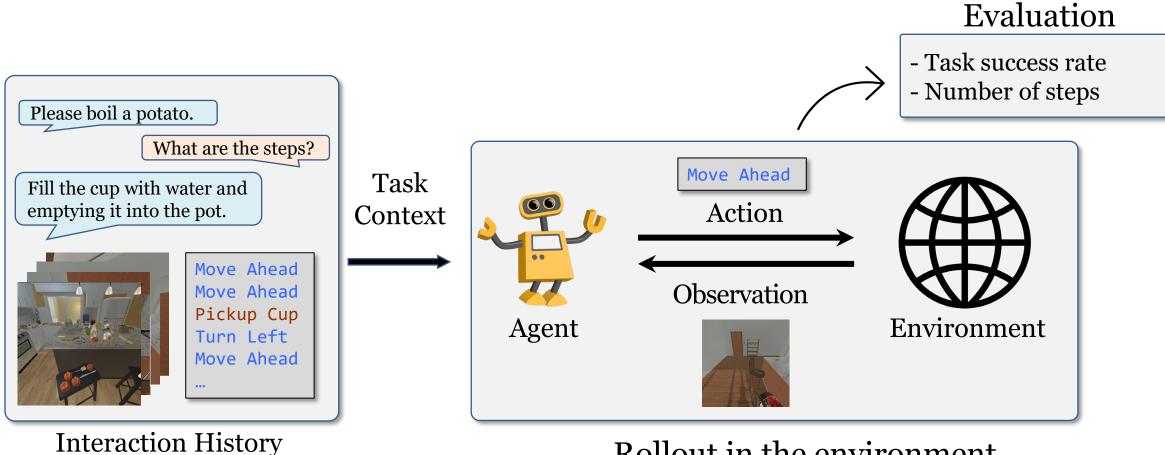
• TEACh^[1]: The agent is given a dialog history as input, and is expected to execute a sequence of actions to achieve the goal set out by the human commander.



[1] Padmakumar A, Thomason J, Shrivastava A, Lange P, Narayan-Chen A, Gella S, Piramuthu R, Tur G, Hakkani-Tur D. Teach: Task-driven embodied agents that chat. AAAI 2022.

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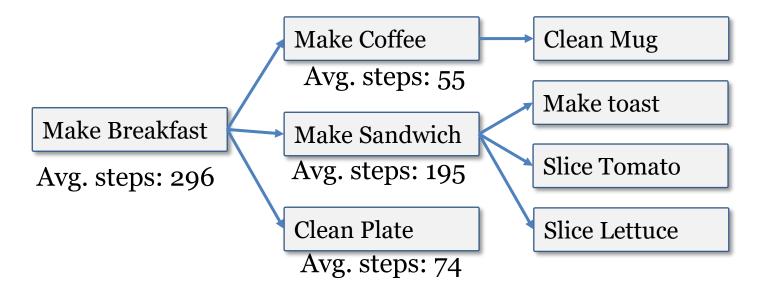
Execution From Dialog History (EDH)



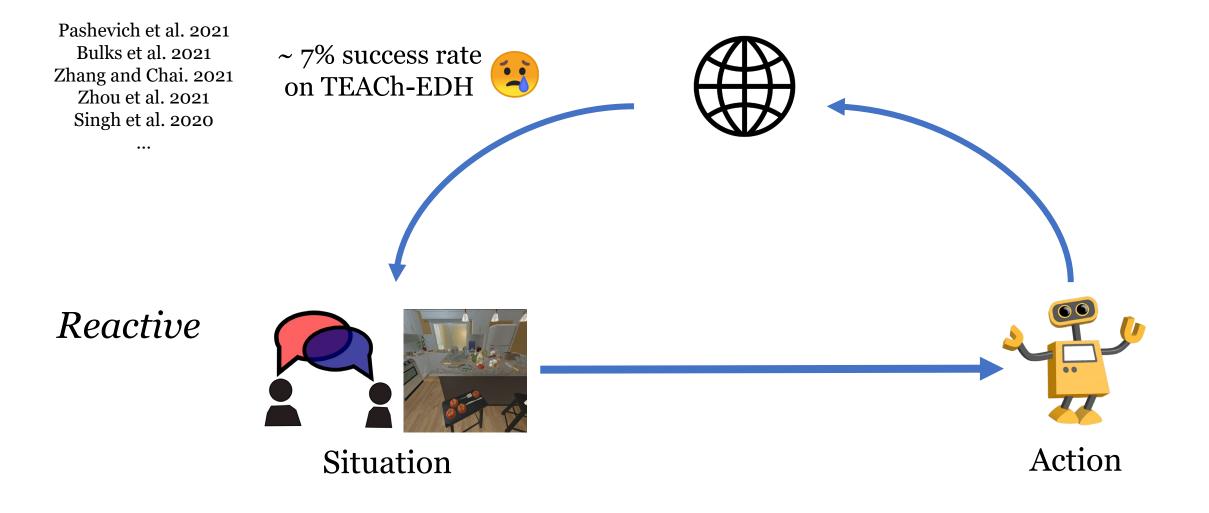
Rollout in the environment



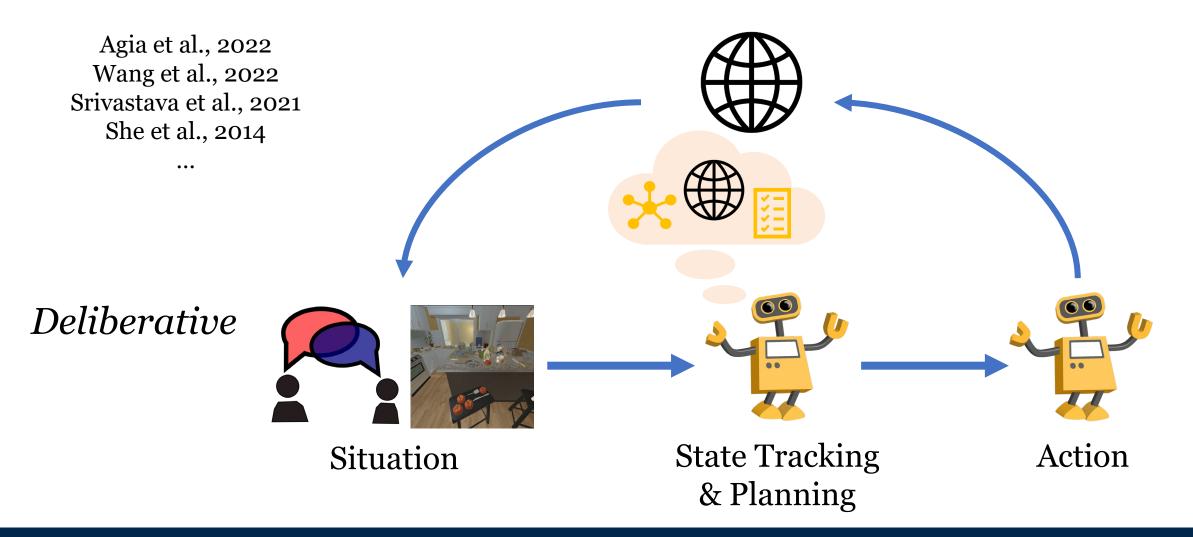
- Multi-modal situation and task understanding
- Partially observable and high-dimensional state
- Long-horizon and compositional tasks



Motivation: From Reactive to Deliberative



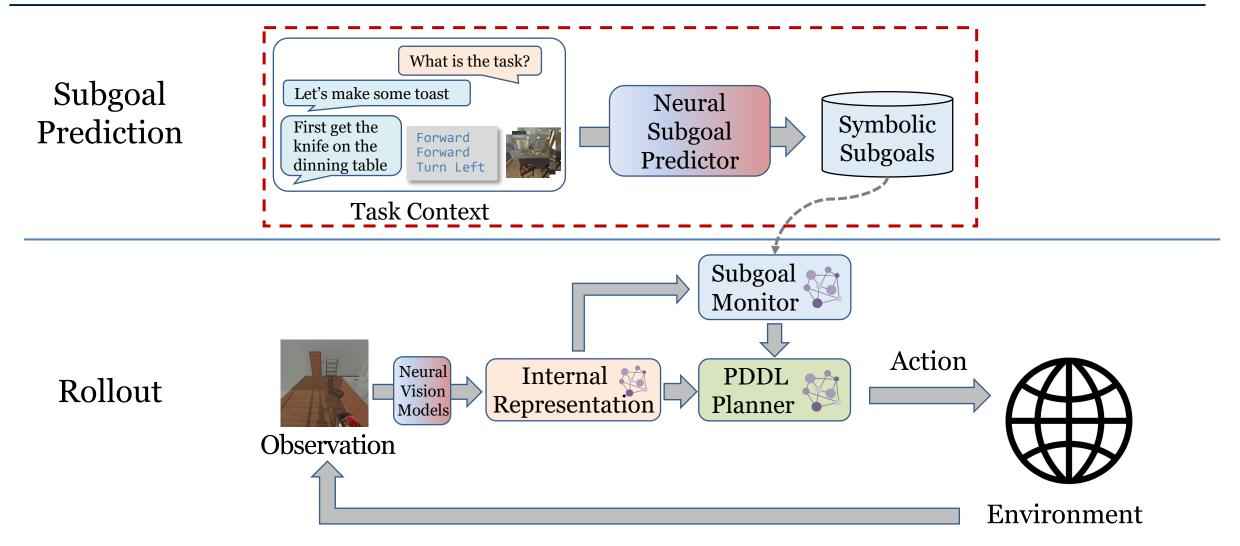
Motivation: From Reactive to Deliberative



Deliberative Agent for following Natural Language Instructions (**DANLI**):

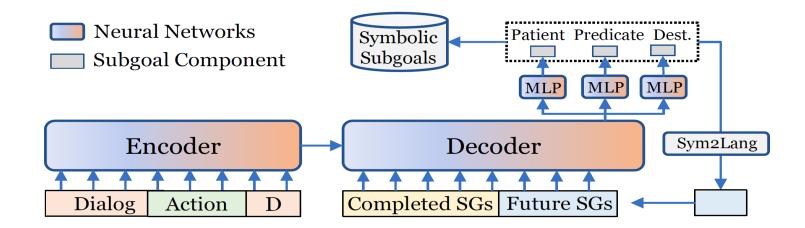
- **1. Neural subgoal predictor** uses a language model to predict symbolic subgoals from situated dialog history
- **2. Spatial-Symbolic world representation** tracks and grounds objects in a 3D voxel map to facilitate both path and task planning
- **3. Online symbolic planner** generates efficient, interpretable plans while allowing online exception handling

Neuro-Symbolic System Architecture



Subgoal Learning From Dialog

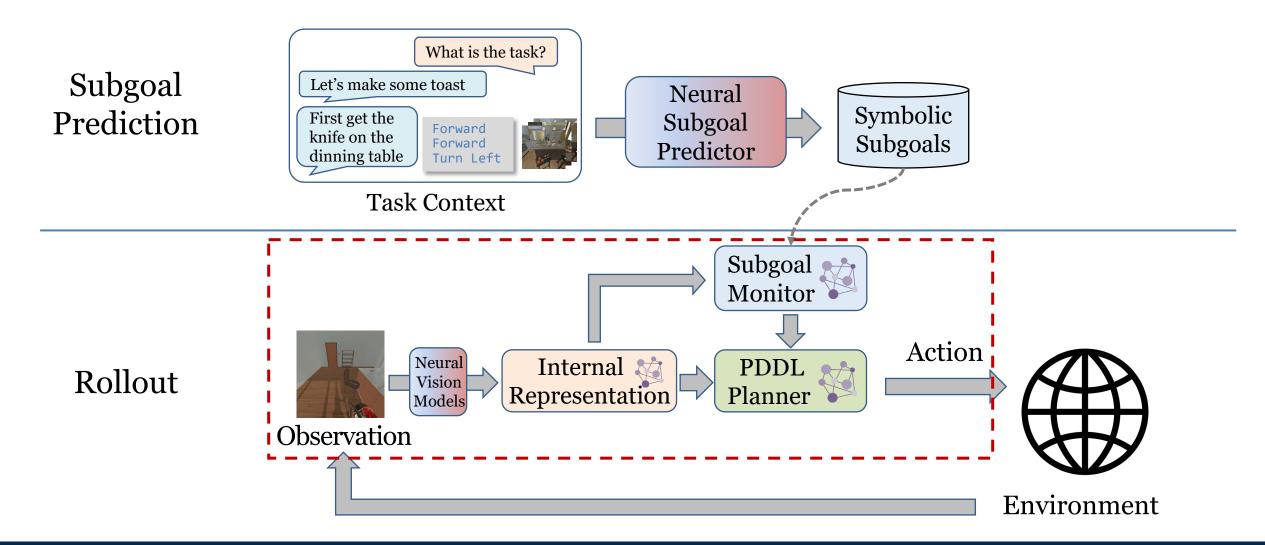
- Neural encoder-decoder network for subgoal prediction from interaction history
- Joint progress estimation and future subgoal prediction



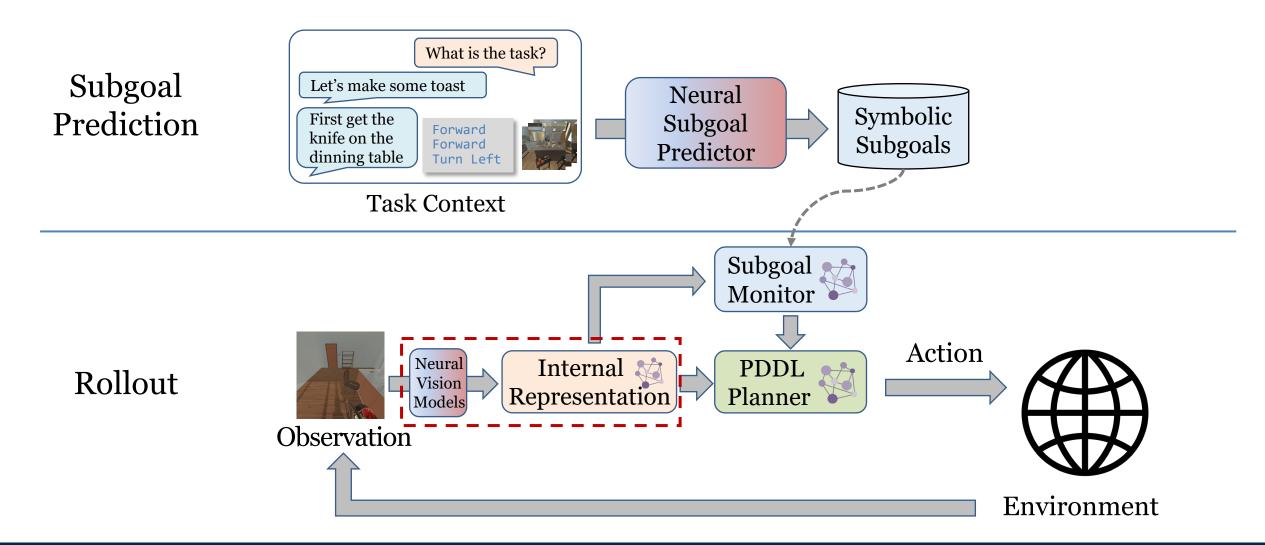
Example Dialog & Action History Input

Follower: Hi. What can I do for you? Commander: Find a cup. Follower go for cup, pick up cup. Commander : Put it on the table. Example Completed & Future Subgoal Output Completed SG: (Cup,isPickedUp) Future SG: (Cup,isPlacedTo,Table)

Neuro-Symbolic System Architecture



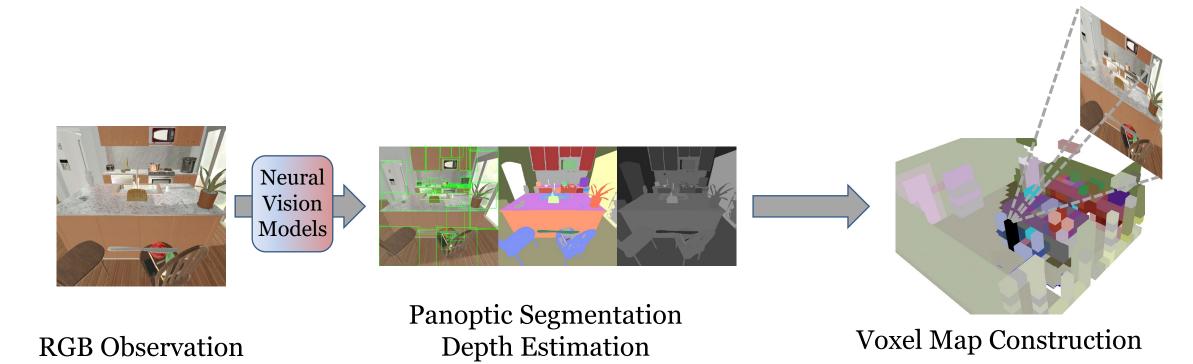
Neuro-Symbolic System Architecture





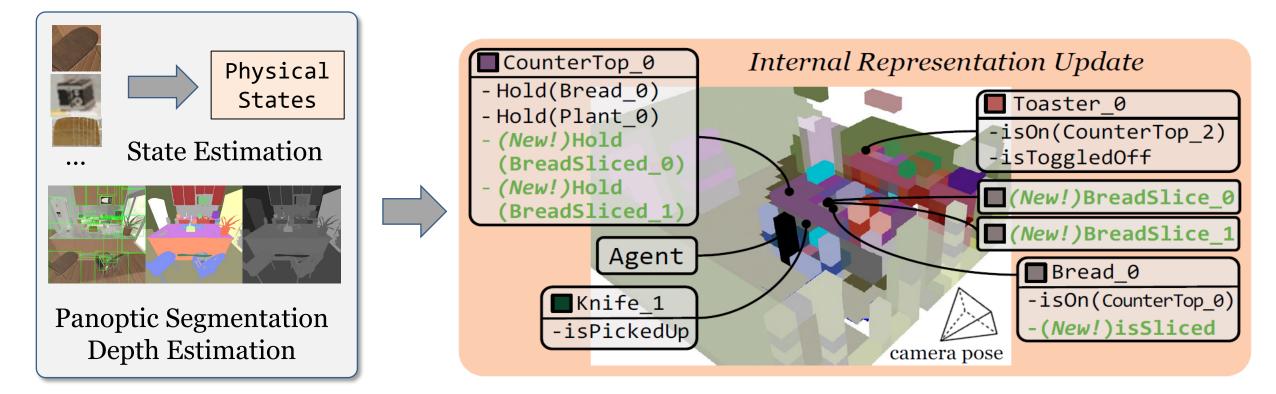
Spatial-Symbolic World Representation

• Neural-powered 3D voxel map construction from ego-centric observations

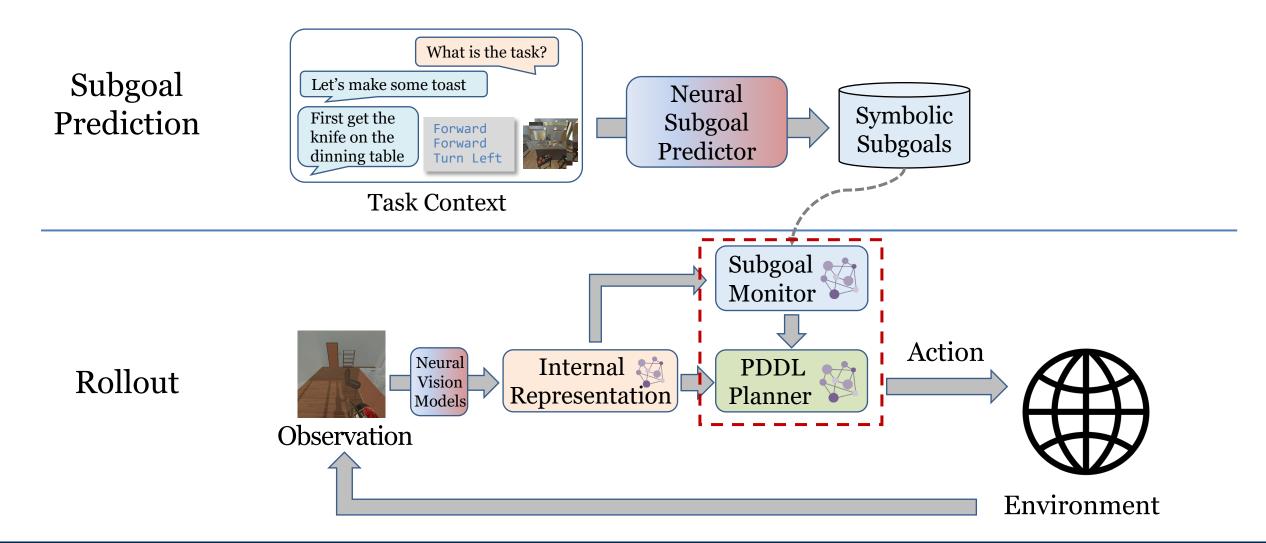


Spatial-Symbolic World Representation

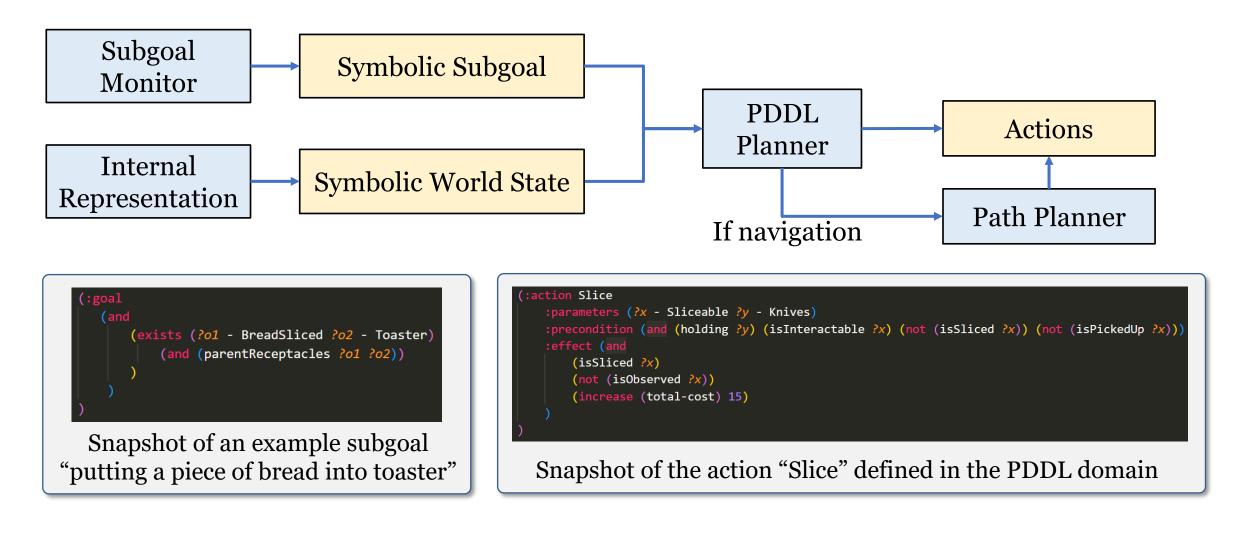
- Recognize physical states of each object instance
- Update the spatial-symbolic world representation at every step



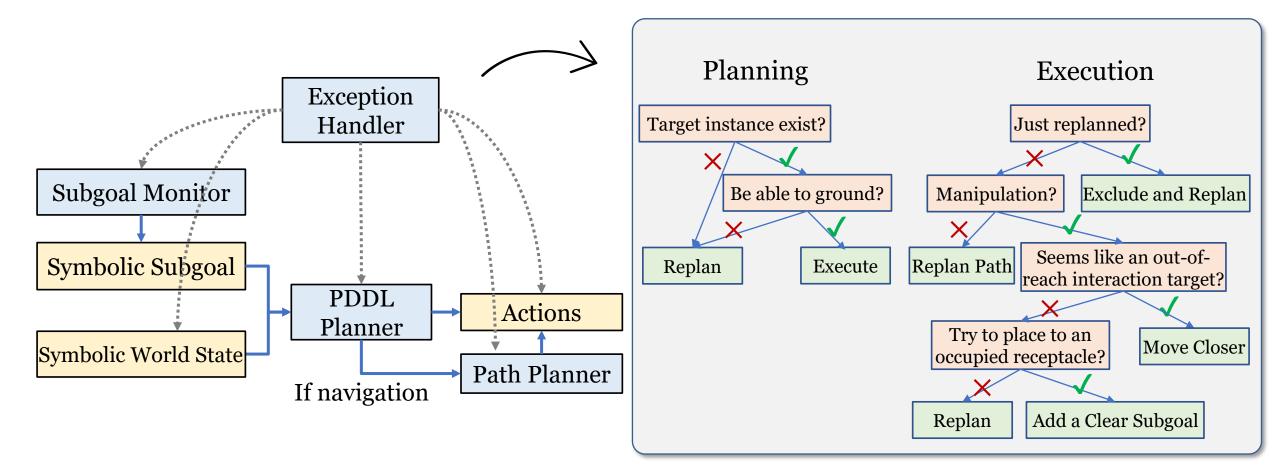
Neuro-Symbolic System Architecture



Symbolic Planning Pipeline

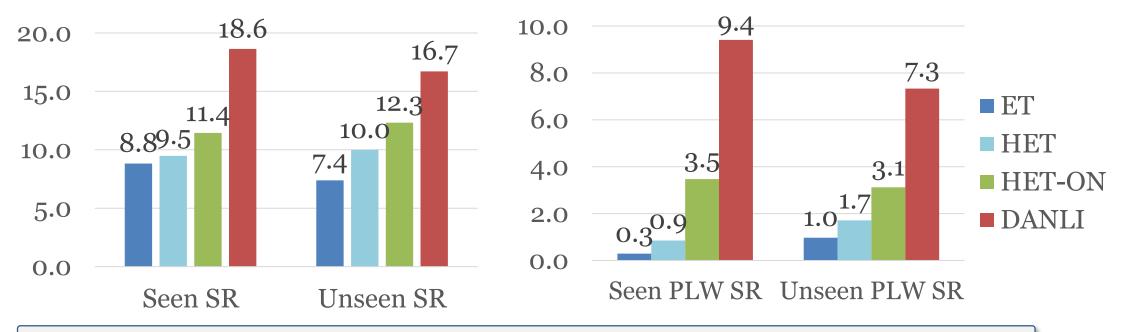


Exception Handling In Online Planning



Higher Success Rates

• DANLI outperforms reactive agents (ET, HET, HET-ON) by a large margin, especially when the task completion efficiency is considered



- Success Rate (SR): the proportion of successfully completed tasks
- Seen/Unseen: whether the evaluated scene is seen or unseen during training
- PLW SR: success rate weighted by the relative trajectory path length compared to humans

Higher Efficiency

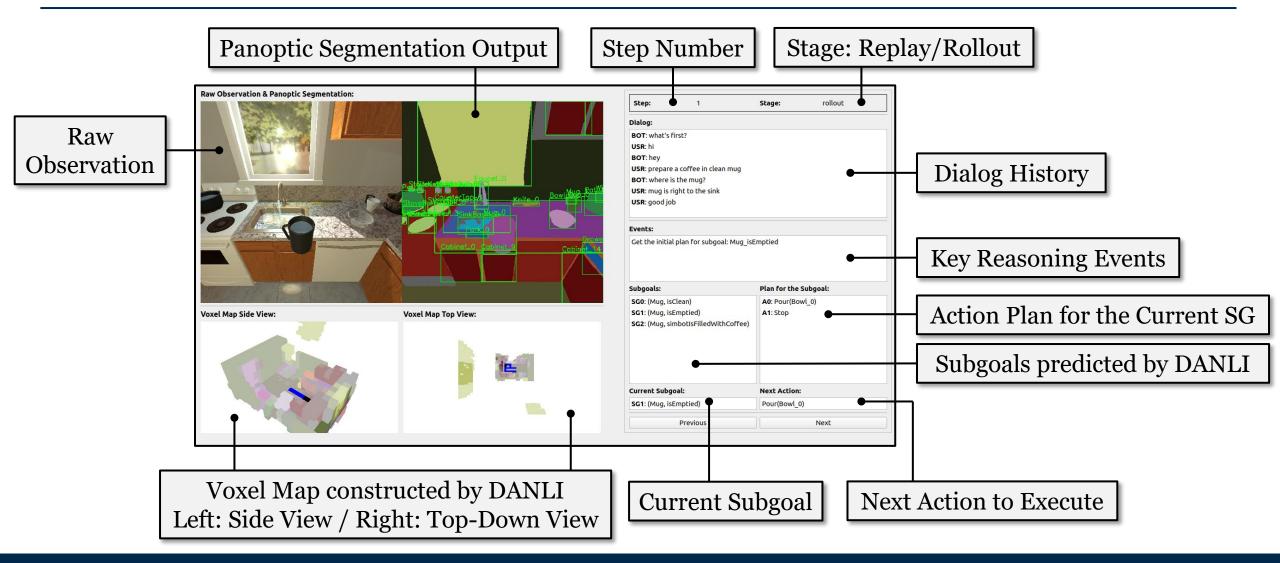
• DANLI completes tasks in fewer steps than reactive agents

Shorter path than human trajectories HET 0.4 Ours Density :0 0.1 0.0 -2 0 2 6 4 Log(ModelPathLength/HumanPathLength)

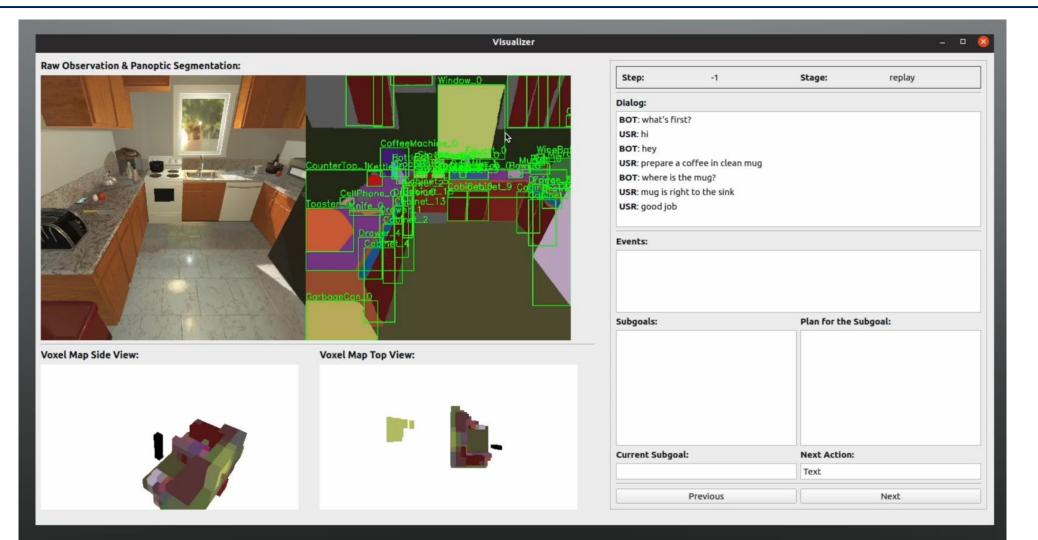
Interpretable Exception Handling

• Deliberative agent produces interpretable plans which allow fine-grain exception handling. **Exception:** grounding failed 20.7% (13.1%) Handling: replan **Exception**: manipulation failed 12.5% (1.5%) Handling: replan **Exception**: non-existent navi. target 9.0% (5.2%) Handling: replan **Exception**: failed again with replan 6.1% (0.8%) Handling: exclude and replan **Exception**: navigation blocked 20.3% (9.0%) Handling: update and path replan **Exception:** no room for placing 9.4% (4.9%) Handling: clear the receptacle **Exception**: mani. target unreached 22.0% (11.4%) Handling: move closer

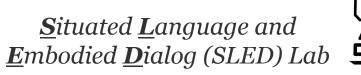
Demo



Demo



Summary - DANLI





Neural Subgoal Predictor

Spatial-Symbolic World Representation

Symbolic Planner with Online Exception Handling

