

Vision and Language Navigation in the Real World via Online Visual Language Mapping

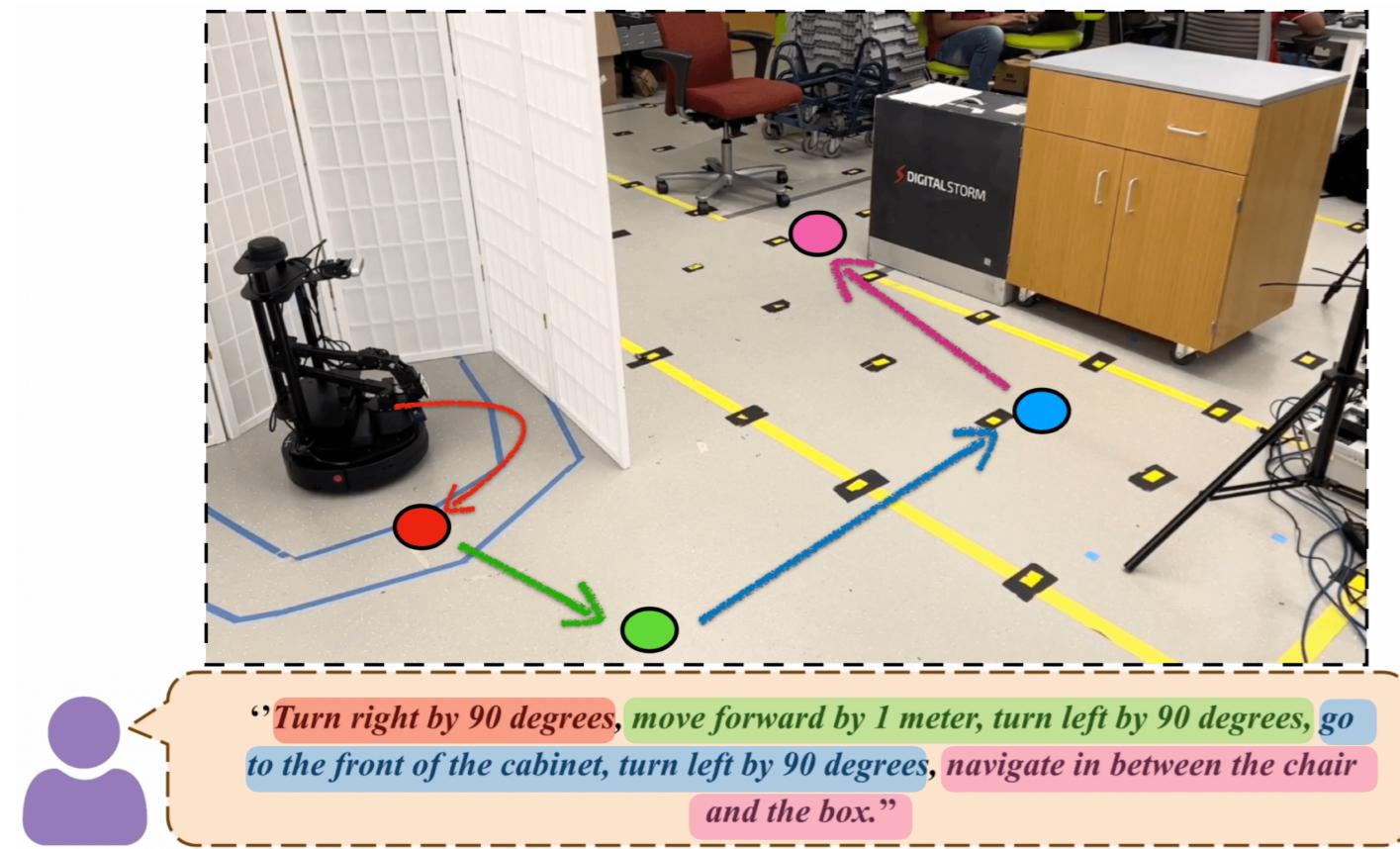


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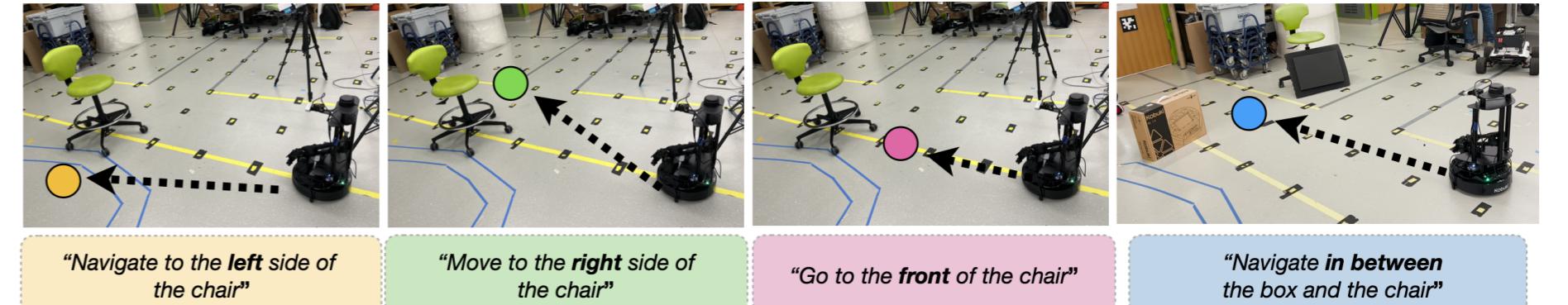


Vision and Language Navigation in Continuous Environments



The robot is asked to navigate in unseen environments by following instructions in natural languages. The robot takes **RGB-D** images and **camera poses** as observations and outputs one of four discrete actions (i.e. **move_forward**, **turn_right**, **turn_left**, and **stop**).

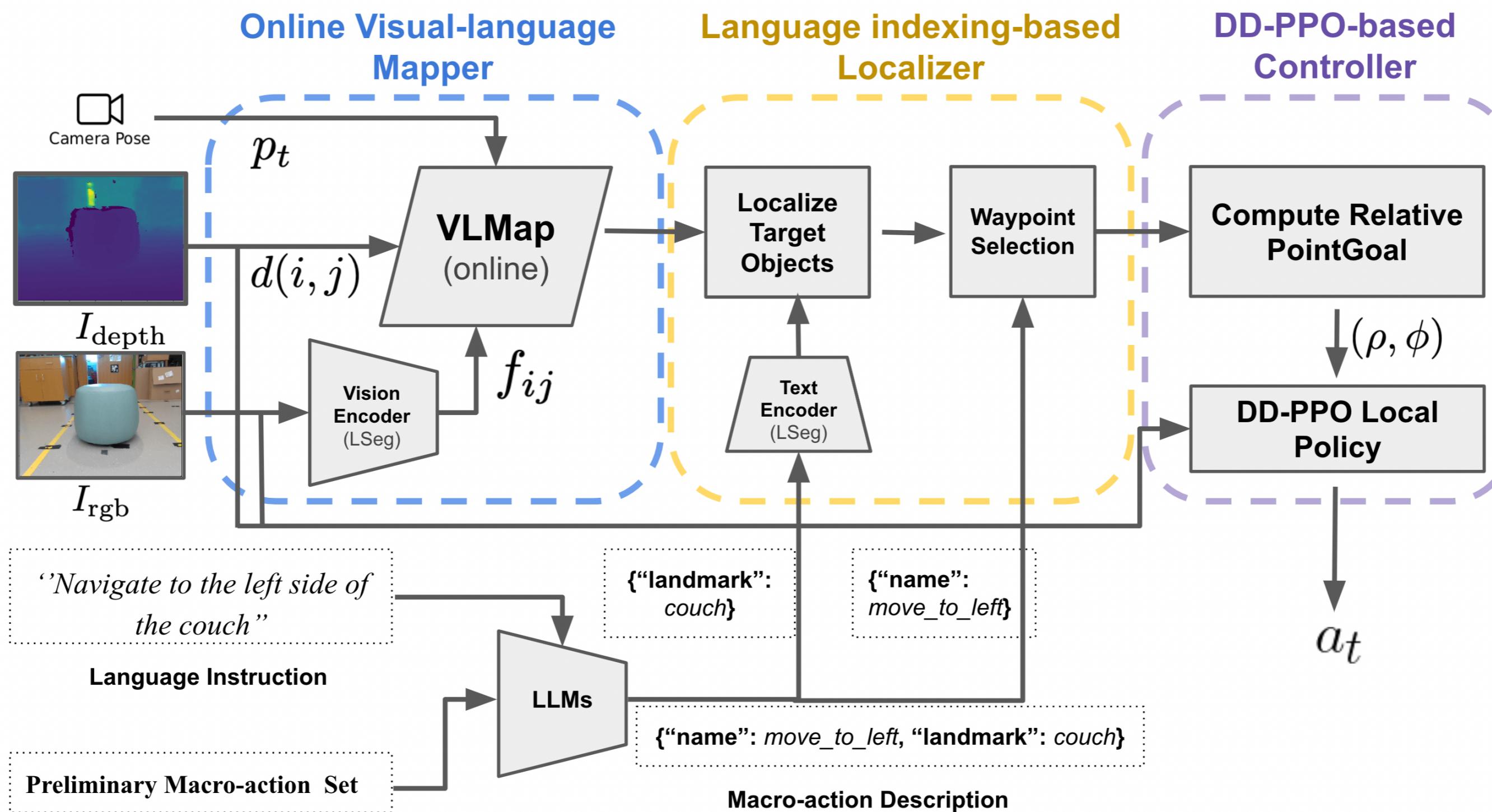
Single Instruction Following



Complex Instruction Following



Navigation Framework Overview



We first use large language models (**LLMs**) to parse the instruction into a sequence of preliminary macro-action descriptions containing both the macro-action name and the related landmark. Then, the online visual-language mapper maintains a visual-language map from the front-view RGB-D observations using visual-language models (**VLMs**). Based on the latest map and the macro-action description, the language-indexing-based localizer outputs the waypoint location, represented as a point goal, on the map. The controller takes in both the RGB-D observation and a relative point goal, computed from the waypoint location and the agent location on the map, and predicts the next action.

Quantitative Results

Table 1: Results of Pure Motion Task

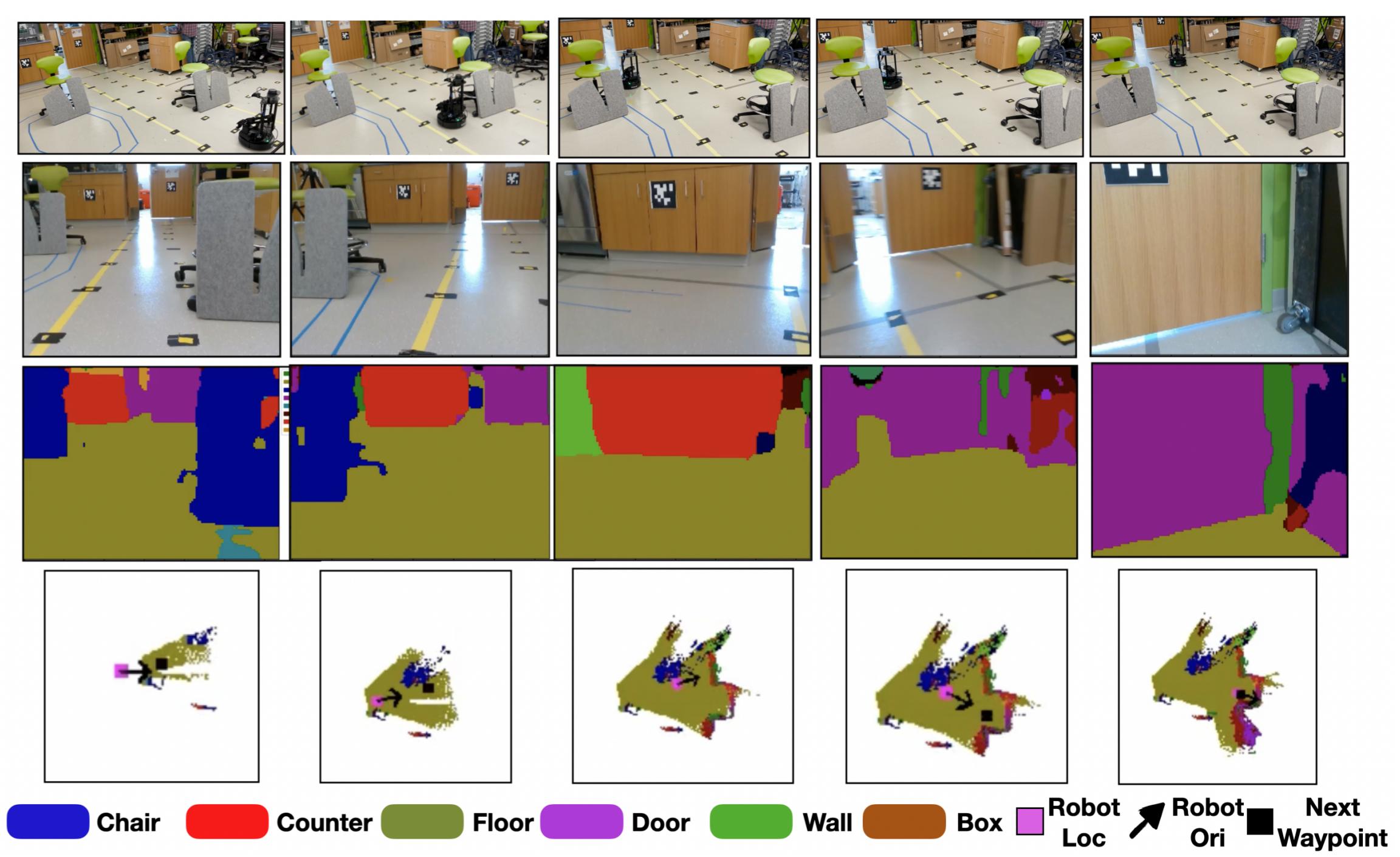
Target Dist (m)	Actual Dist (m)	Est Dist (m)	Err Dist (m)
0.5	0.426	-	-
1.0	0.748	0.238	0.014
2.0	1.678	0.308	0.014

Table 2: Results of Landmark-associated Motion Task

Method	CM2 [7]	
	Instruction	SR (%)
"Navigate to the <i>left</i> side of the chair"	60	0.88
"Navigate to the <i>right</i> side of the chair"	40	0.97
"Navigate to the <i>front</i> of the chair"	20	1.37
"Move in between the box and the chair"	0	2.06
Average	30	1.32
Method	Ours	
"Navigate to the <i>left</i> side of the chair"	100	0.79
"Navigate to the <i>right</i> side of the chair"	100	0.83
"Navigate to the <i>front</i> of the chair"	100	0.81
"Move in between the box and the chair"	80	0.20
Average	95	0.66

Method	SR (%)	Dist to Goal (m)	Time steps
CM2 [7]	0	4.9	203.4
Ours	100	0.256	88.6

Qualitative Results



With no fine-tuning, the proposed framework generalizes well to the unseen lab environment.