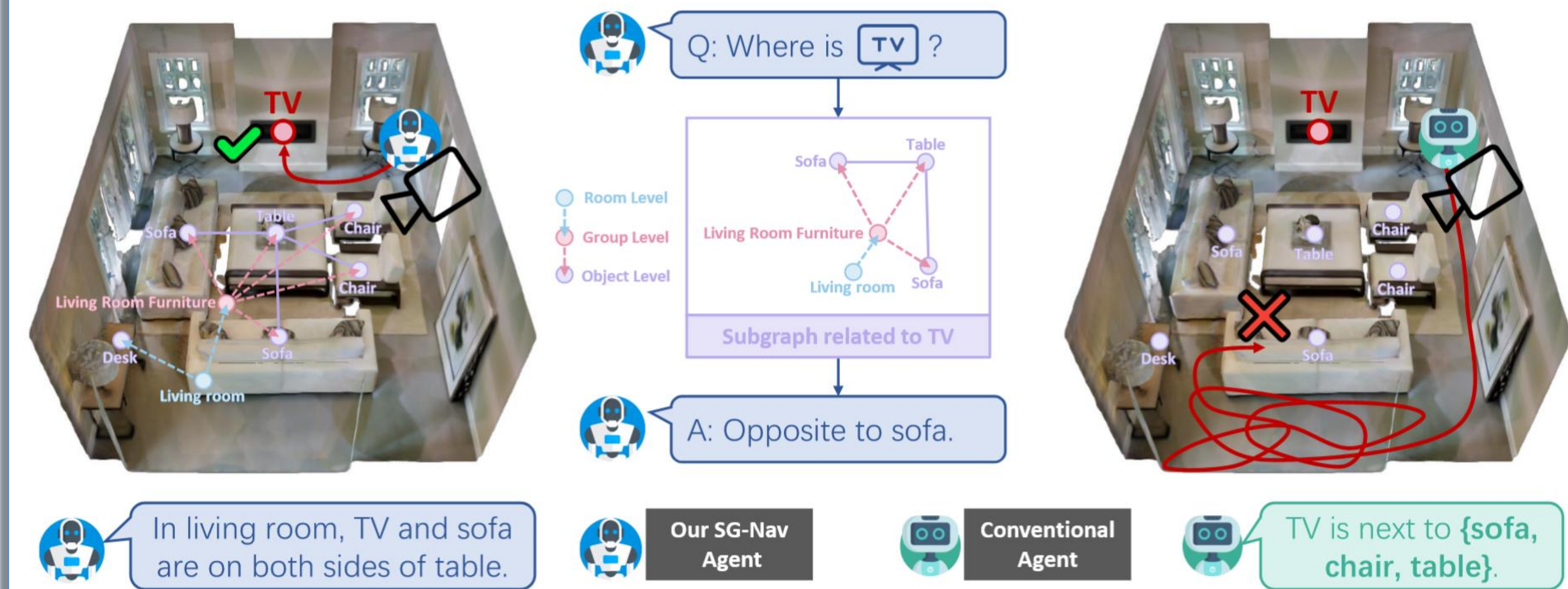


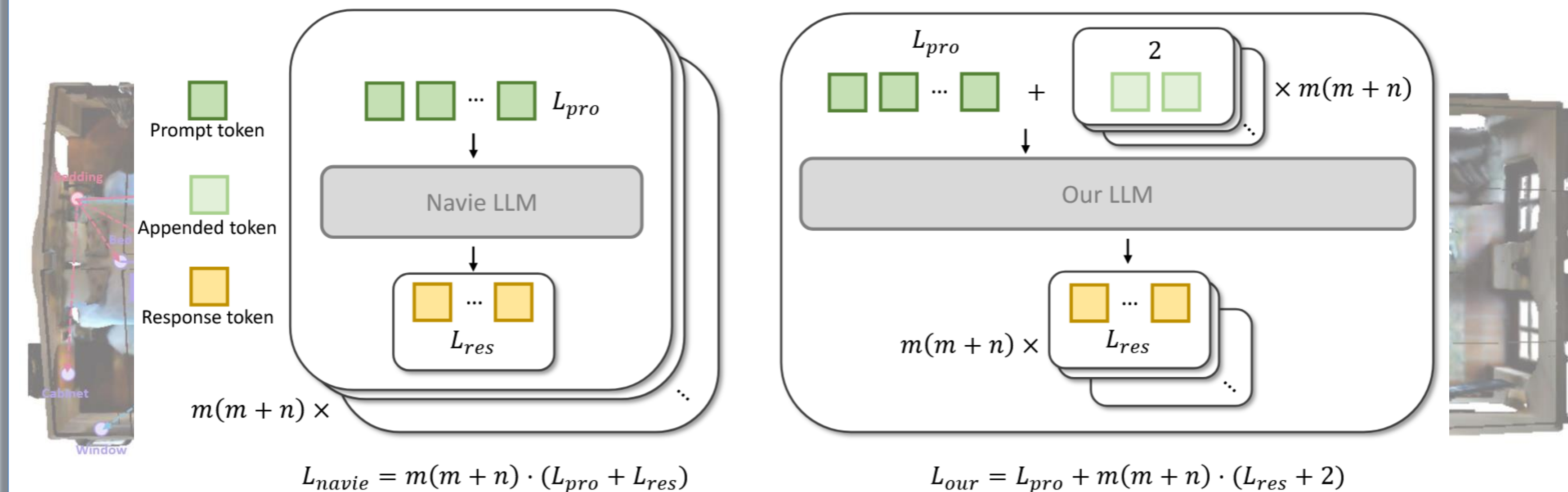
Highlight

- Conventional object-goal navigation methods rely on time-consuming training. Our SG-Nav online construct **hierarchical scene graph** and **prompt LLM** to reason the position of object goal in a **zero-shot** manner, with higher generalization ability. We even surpass some learning-based methods.



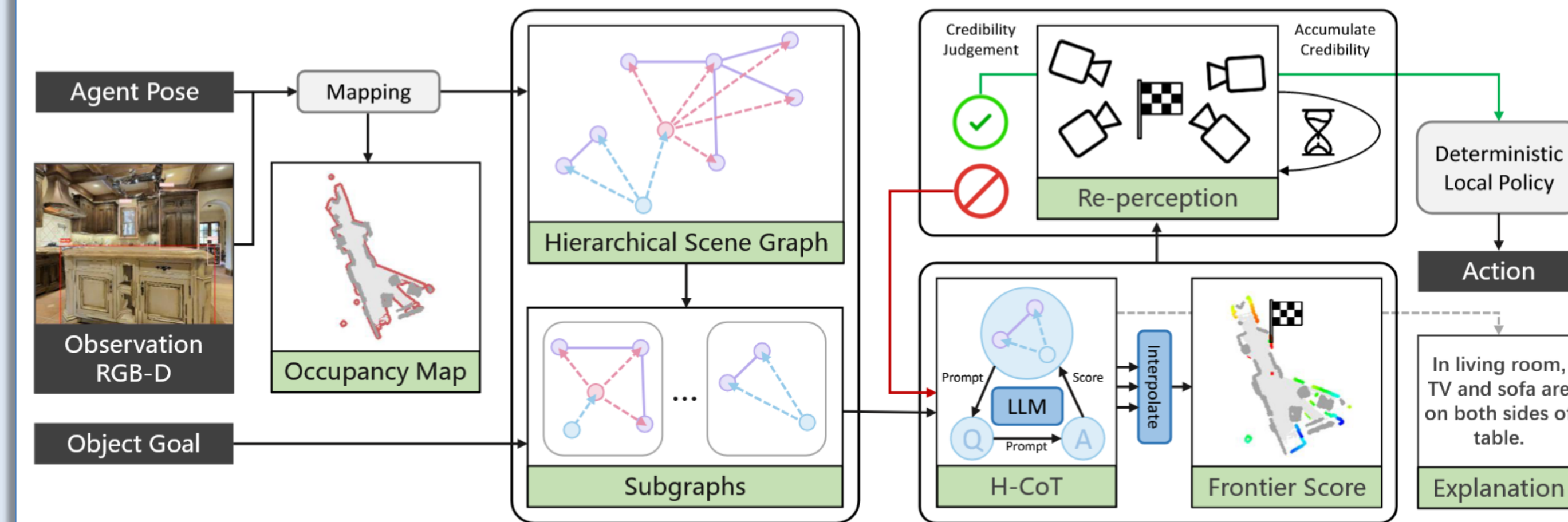
Analysis

- To **speed up** the dense connecting process for real-time processing, we propose a **new form of prompt**. This enables LLM to generate relationships between all pairs of nodes in **one-shot** with much **less computational cost**, comparing to navie generation approach.

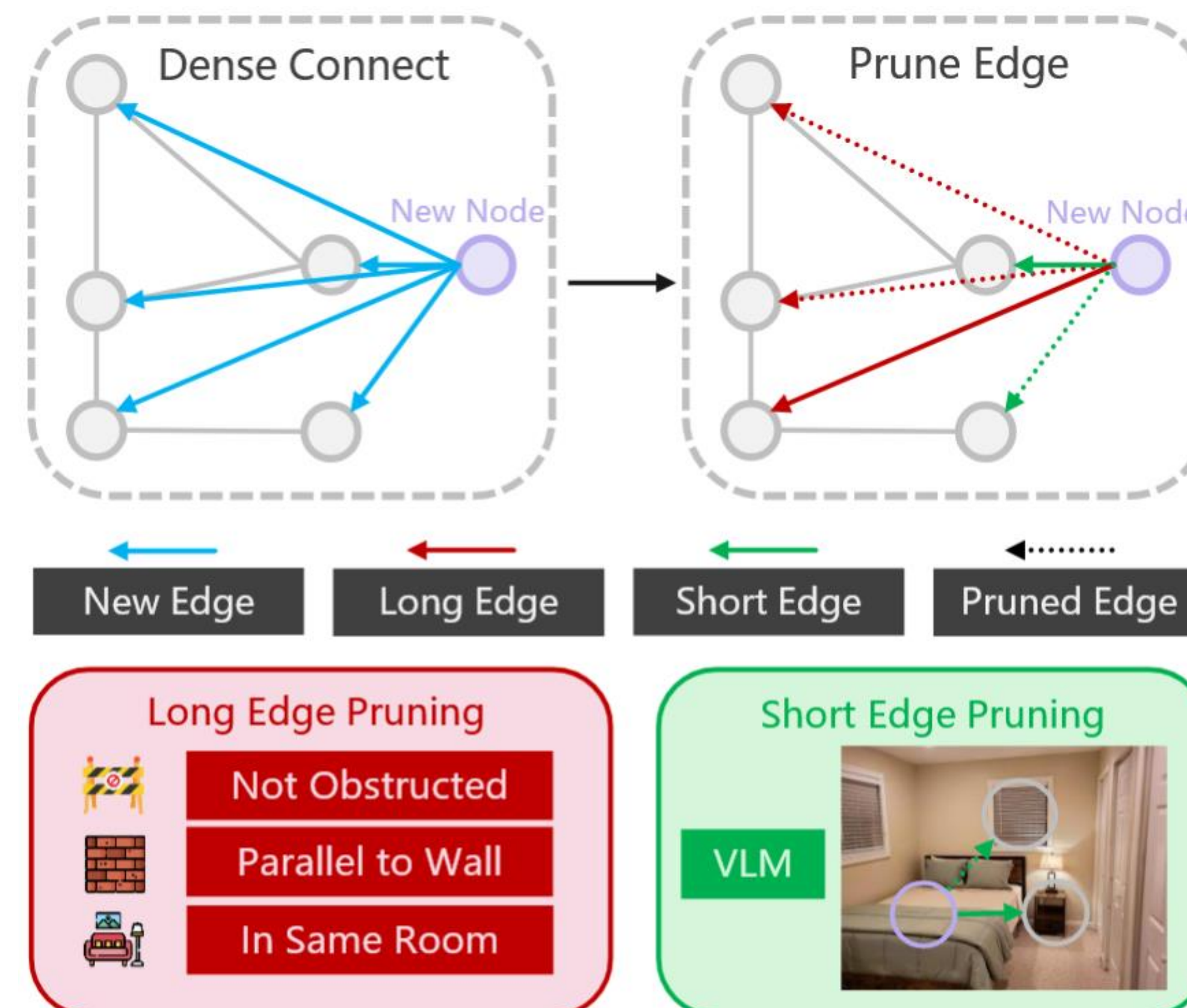


Approach

- We propose to online construct a **hierarchical scene graph** and decompose the scene graph and use the subgraphs to **prompt LLM** in a **chain-of-thought** manner. Then we **score the frontier points** and select a long-term goal. Besides, we propose **re-perception** to solve the problem of perception errors.



- We **densely connect** the newly detected nodes to previous nodes in an **incremental** manner. We further **prune** the dense connected edges to make the 3D scene graph more precise.



Experiments

- Our SG-Nav achieves leading performance comparing to most zero-shot methods.

Method	Unsupervised	Zero-shot	MP3D		HM3D		RoboTHOR	
			SR	SPL	SR	SPL	SR	SPL
CoW [10]	Yes	Yes	7.4	3.7	-	-	26.7	16.9
ESC [51]	Yes	Yes	28.7	14.2	39.2	22.3	38.1	22.2
L3MVN [45]	Yes	Yes	34.9	14.5	48.7	23.0	41.2	22.5
OpenFMNav [18]	Yes	Yes	37.2	15.7	52.5	24.1	44.1	23.3
VLFM [44]	Yes	Yes	36.2	15.9	52.4	30.3	42.3	23.0
SG-Nav-LLaMA	Yes	Yes	40.1	16.0	53.9	24.8	47.3	23.7
SG-Nav-GPT	Yes	Yes	40.2	16.0	54.0	24.9	47.5	24.0

- Visualization of Navigation Process:

