



# GroundFlow: A Plug-in Module for Temporal Reasoning on 3D Point Cloud Sequential Grounding

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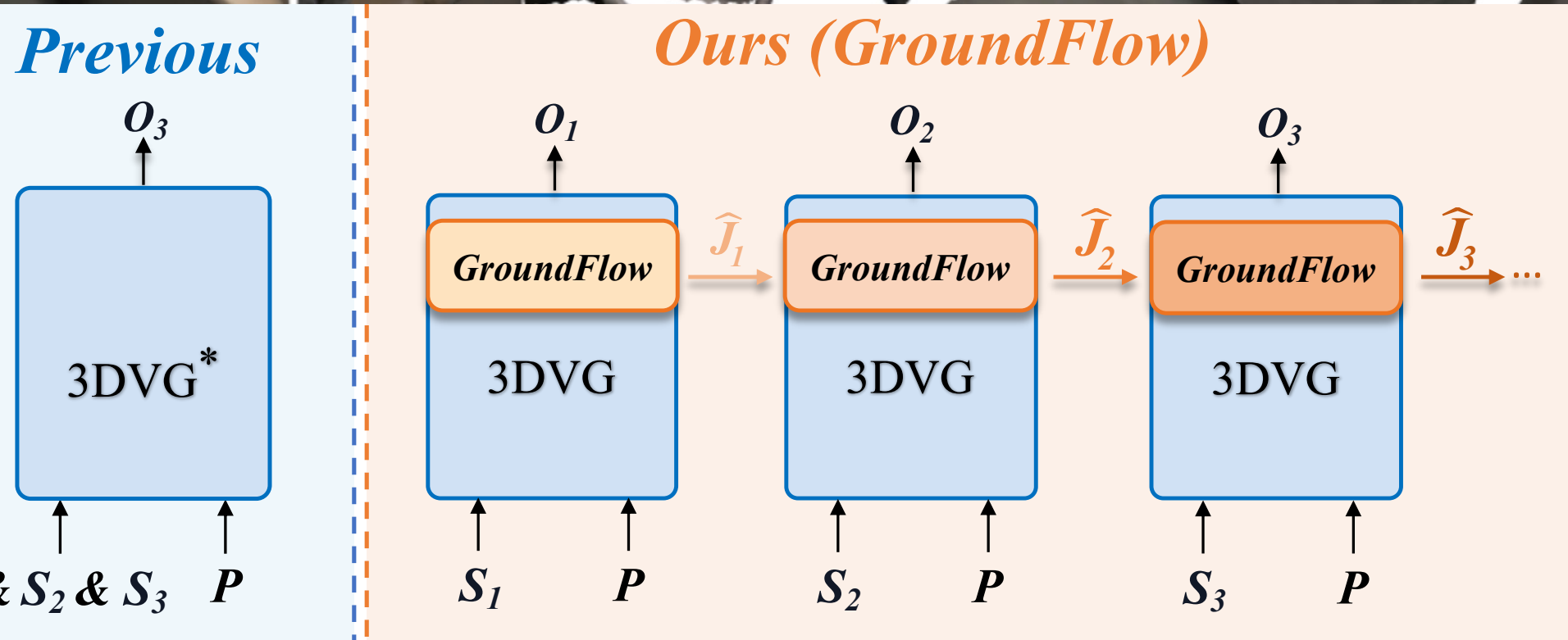
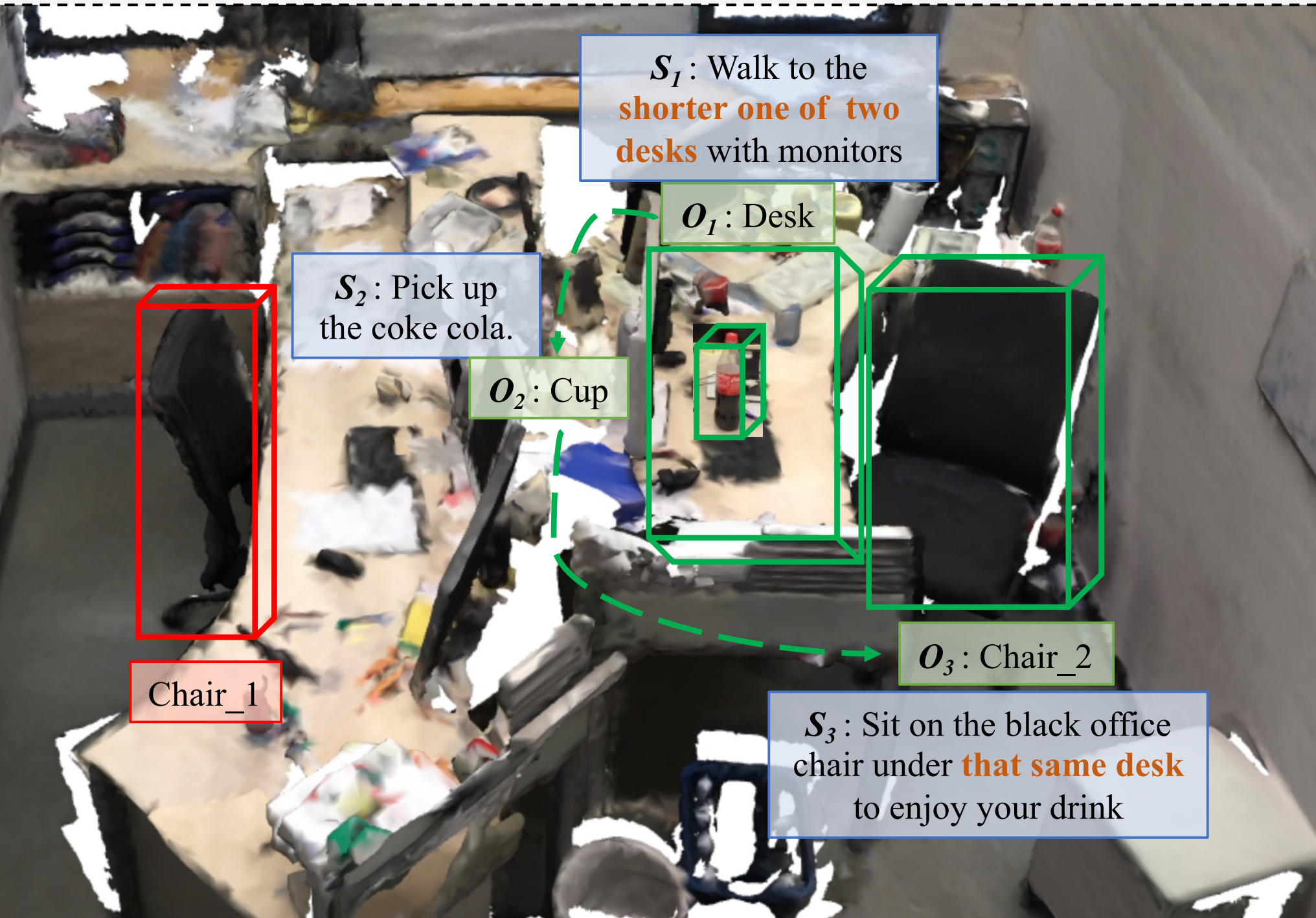


## Motivation

$T$ : Refresh yourself with a beverage

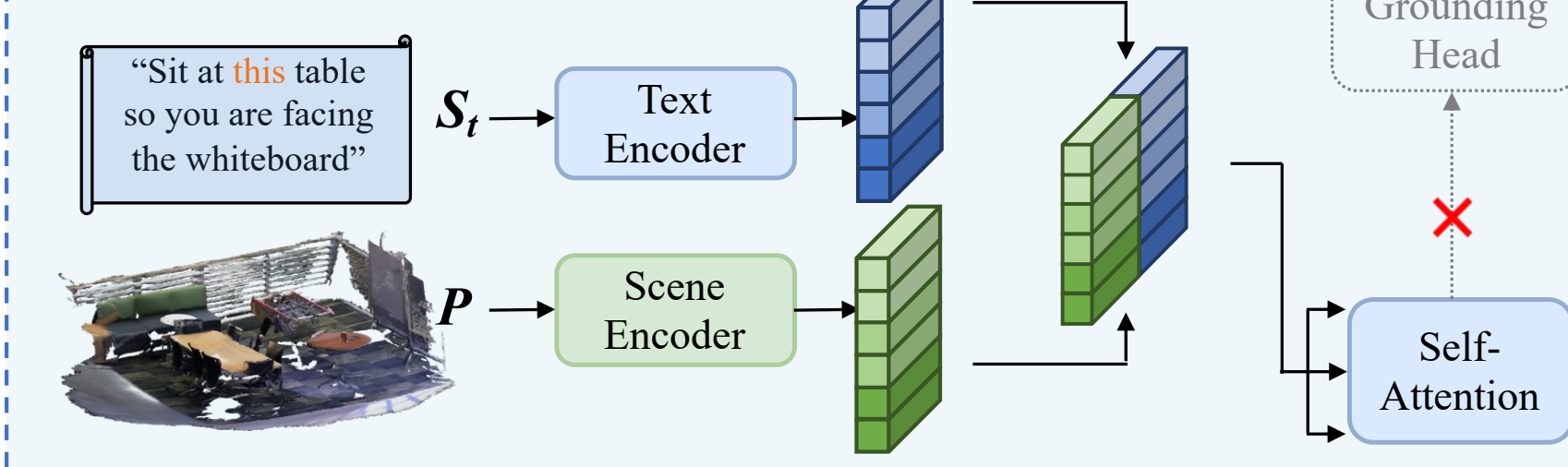
$T$ : Task Description  $P$ : 3D Point Cloud  $S_t$ : Step  $t$ 's Instruction  $O_t$ : Step  $t$ 's Target Object

— Grounding Sequences — Correct Predictions — Distractor \* 3DVG-based Method

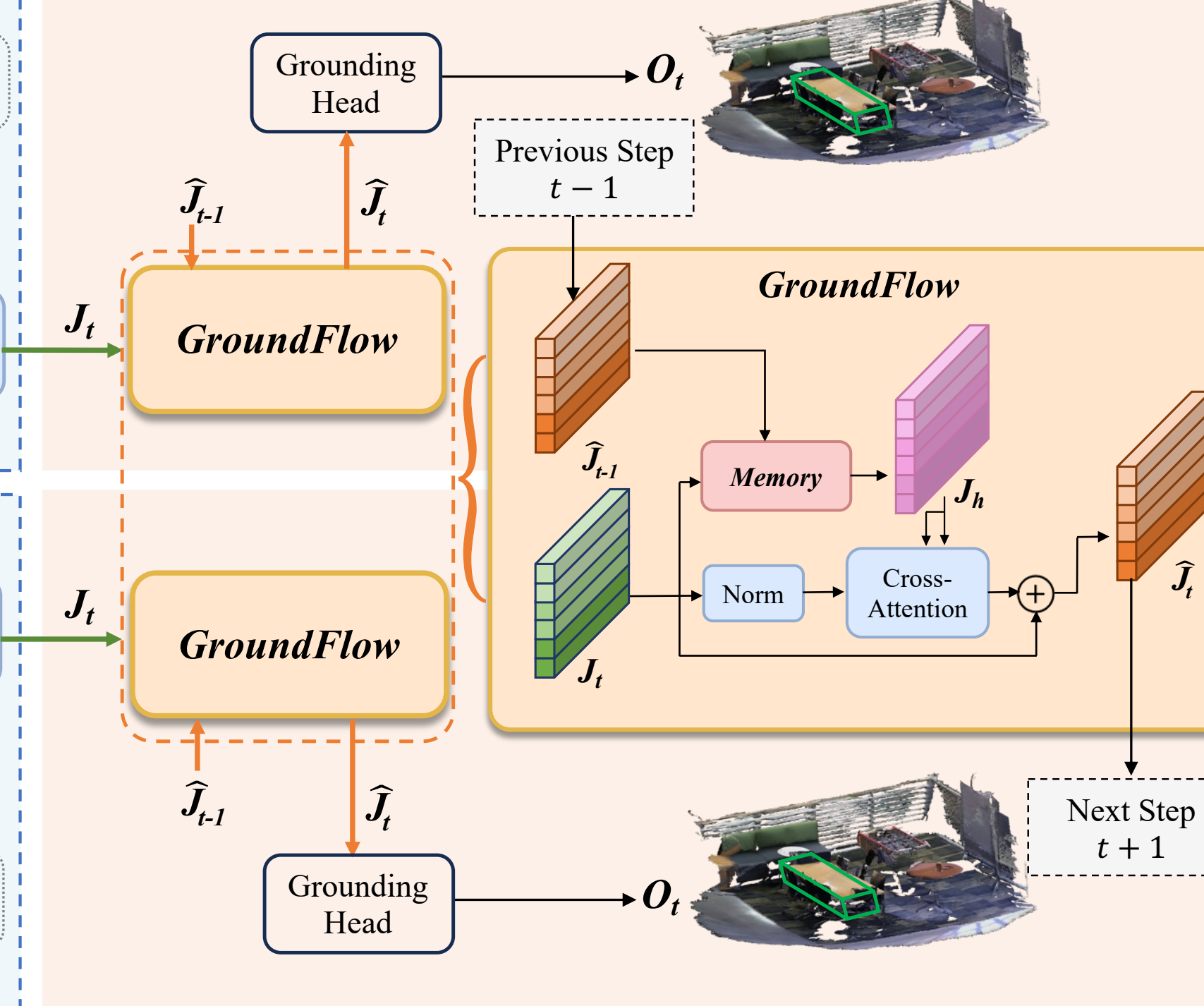
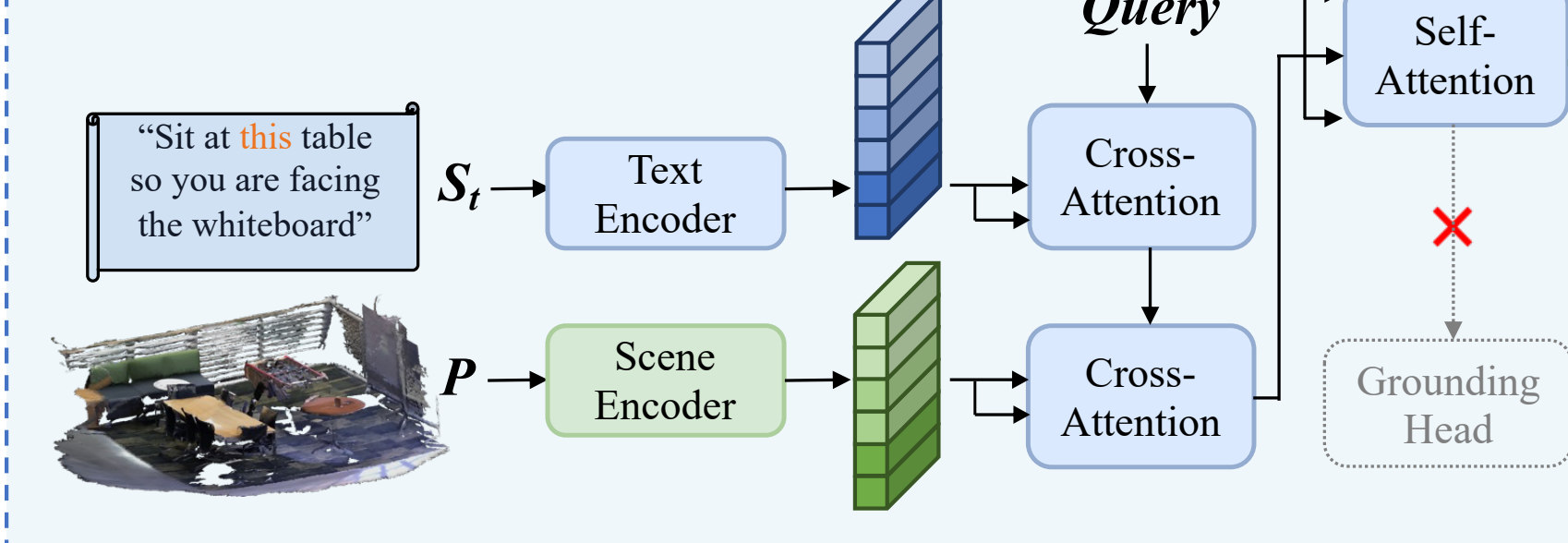


## GroundFlow

### Dual-stream Model



### Query-based Model



GroundFlow could be built on any 3DVG models as a plug-in in a recurrent framework.

## Experiments

Model Type	Method	ScanNet s-acc t-acc	3RScan s-acc t-acc	MultiScan s-acc t-acc	ARKitScenes s-acc t-acc	HM3D s-acc t-acc	Overall s-acc t-acc
LLM-based	GPT4 + PointNet++ (Zero-shot) [52]	42.6 10.9	25.5 2.4	27.0 0.0	27.6 6.0	20.8 7.7	27.3 7.6
	LEO (3DLLM) [24]	61.2 25.7	55.8 16.0	52.7 7.6	69.6 41.5	61.5 35.7	62.8 34.1
Dual-stream	3D-VisTA [56]	60.1 24.7	52.7 13.5	47.6 7.0	68.4 37.8	57.5 30.6	60.3 28.8
	3D-VisTA+ GroundFlow	63.0 26.6	56.8 21.7	57.1 14.0	71.9 46.0	62.3 36.9	64.1 35.1
	MiKASA [5]	57.8 19.4	53.0 10.9	48.7 2.3	67.1 35.7	57.3 30.1	60.8 31.9
	MiKASA + GroundFlow	62.7 28.9	58.9 17.4	54.0 11.6	70.2 42.9	61.8 36.2	63.5 34.2
Query-based	PQ3D [57]	53.7 17.9	50.2 9.9	43.5 4.7	64.9 32.0	56.9 30.6	57.3 25.9
	PQ3D + GroundFlow	62.0 28.2	60.1 21.0	51.3 7.0	73.0 48.1	63.6 38.0	64.8 36.1
	Vi3DRel [8]	59.3 19.9	55.9 15.2	50.9 4.7	69.3 38.6	58.7 31.0	61.1 28.6
	Vi3DRel + GroundFlow	63.1 27.8	58.8 22.5	57.6 20.9	72.4 45.1	62.3 36.6	64.4 35.2

GroundFlow improves the task accuracy of baseline methods by large margin (+7.5% in dual-stream and +10.2% in query-based).

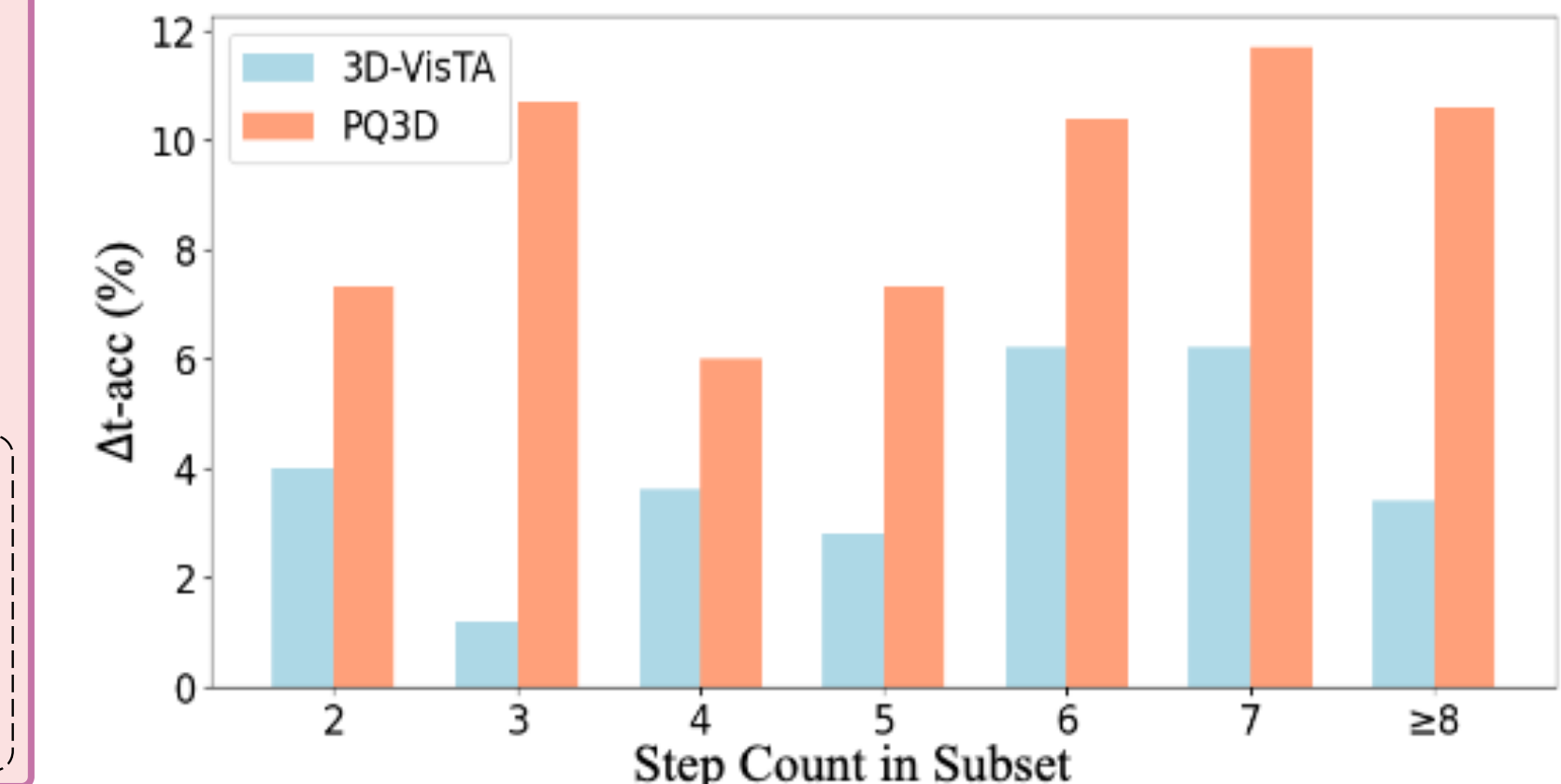
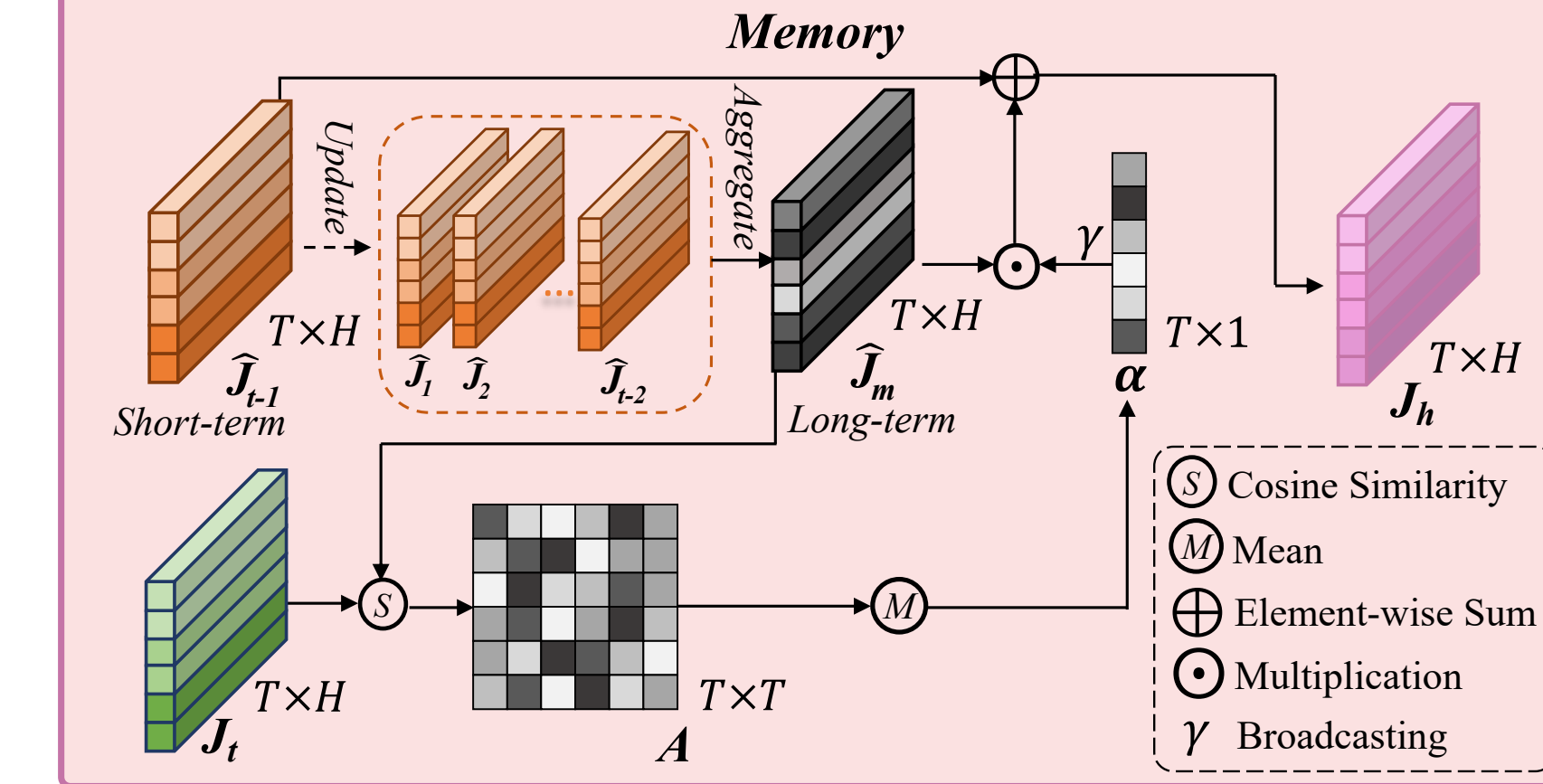
3DVG + GroundFlow outperforms 3DLLM LEO, achieving SOTA results.

Models	Temporal Fusion Methods	s-acc	t-acc	$\Delta$ s-acc	$\Delta$ t-acc
3D-VisTA	LSTM	61.4	29.5	+1.1	+0.7
	GRU	62.0	28.8	+1.7	+0.0
	Transformer	62.9	33.5	+2.6	+4.7
	GroundFlow	64.1	35.1	+3.8	+6.3
PQ3D	LSTM	63.1	30.8	+5.8	+4.9
	GRU	63.8	30.7	+6.5	+4.8
	Transformer	63.4	33.6	+6.1	+7.7
	GroundFlow	64.8	36.1	+7.5	+10.2

Models	#params	Speed	s-acc	t-acc
LEO	6.9B	11.3ms	62.8	34.1
3D-VisTA	101.1M	5.2ms	60.3	28.8
3D-VisTA+ GroundFlow	123.1M	5.6ms	64.1	35.1
PQ3D	167.4M	6.8ms	57.3	25.9
PQ3D+ GroundFlow	189.4M	6.9ms	64.8	36.1

GroundFlow performs temporal fusion more effectively than traditional methods (LSTM, GRU or Transformer) with only 22M parameters and a marginal increase in inference time.

## Memory Structure



GroundFlow selectively extract short-term and long-term information based on its relevance to the current instruction, maintaining its temporal understanding advantage as step counts increase.

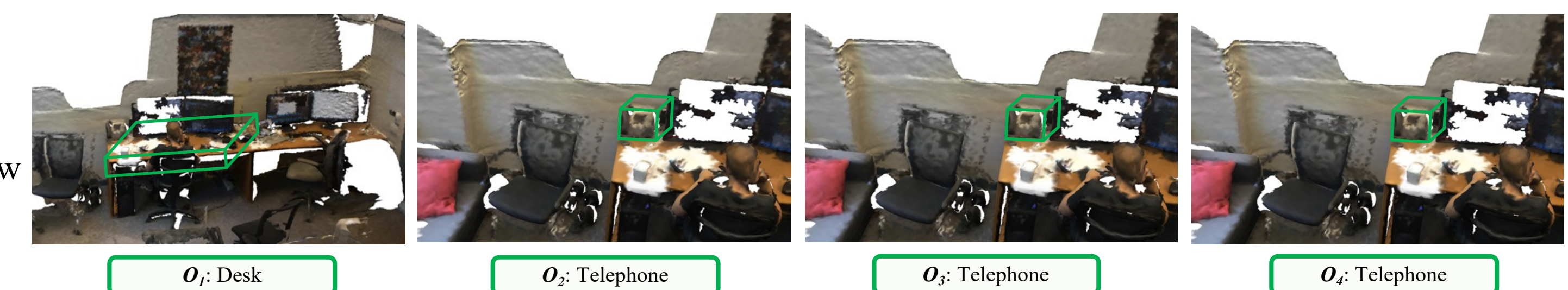
## Qualitative Visualization

$T$ : Call a client using the office phone  
 $S_1$ : Walk over the desk where a black telephone sits on top near the monitors  
 $S_2$ : Pick up the telephone receiver  
 $S_3$ : Dial the client's number  
 $S_4$ : Begin the conversation when the client answers

PQ3D



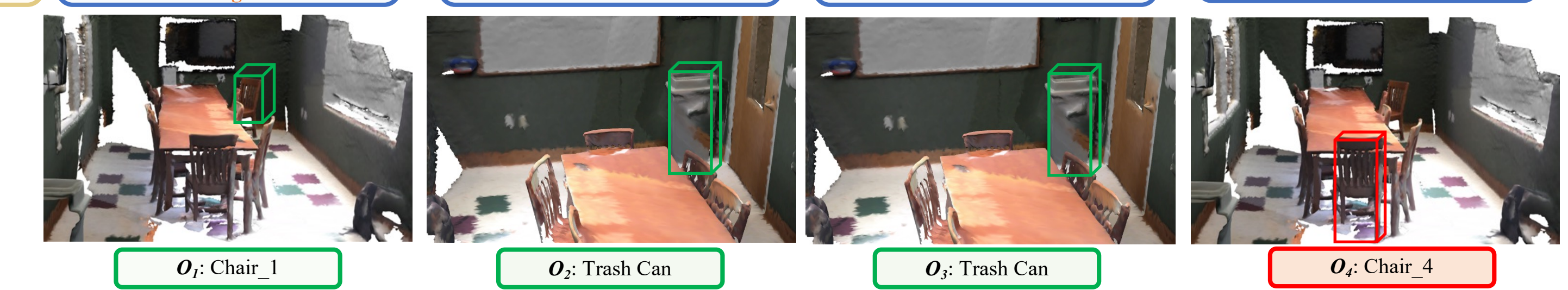
PQ3D + GroundFlow



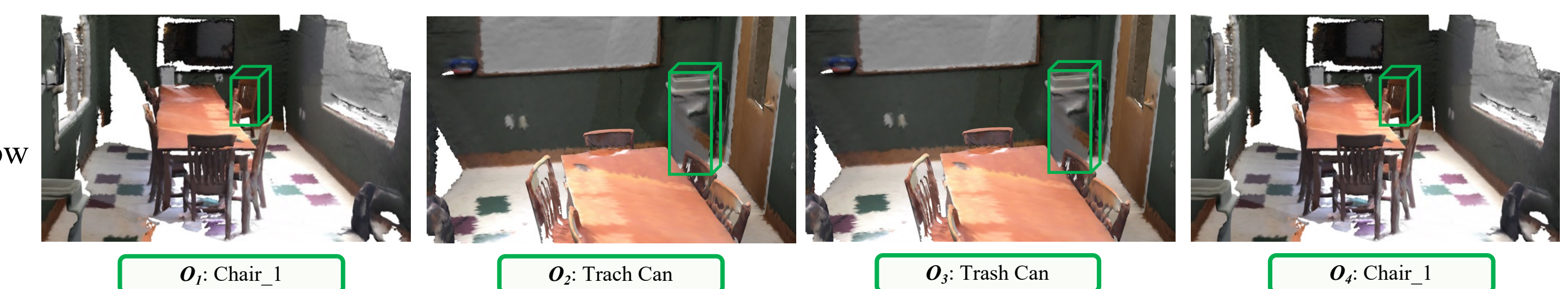
$T$ : Throw Away your lunch trash

$S_1$ : Stand up from the wooden chair, which is under the tv and on the right side  
 $S_2$ : Walk to the gray trash can near the table  
 $S_3$ : Open the lid of the trash can and dispose of the leftovers  
 $S_4$ : Walk back to your chair and continue your work

PQ3D



PQ3D + GroundFlow



- SG3D task instruction often contains **pronouns** such as “it”, “the other”, “the same”.
- It requires grounding method to understand the **context** and retrieve **relevant history information**.
- Previous 3DVG method: Simply **concatenate** multiple step instructions **without extracting temporal information**.
- We propose **GroundFlow** – a **plug-in** module for **temporal reasoning** on SG3D.