

# NaVid: Video-based VLM Plans the Next Step for Vision-and-Language Navigation

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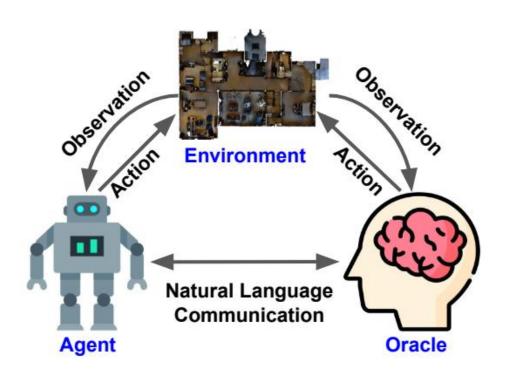


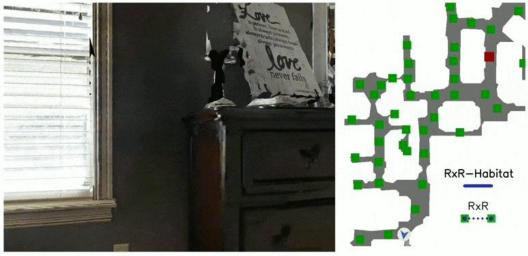
GALBOT

## Vision-and-Language Navigation (VLN)

Given free-form instruction, the robot is required to follow the instruction to navigate in the unseen environments.

"Leave the bedroom, and enter the kitchen. Walk forward and take a left at the couch. Stop in fornt of the window"

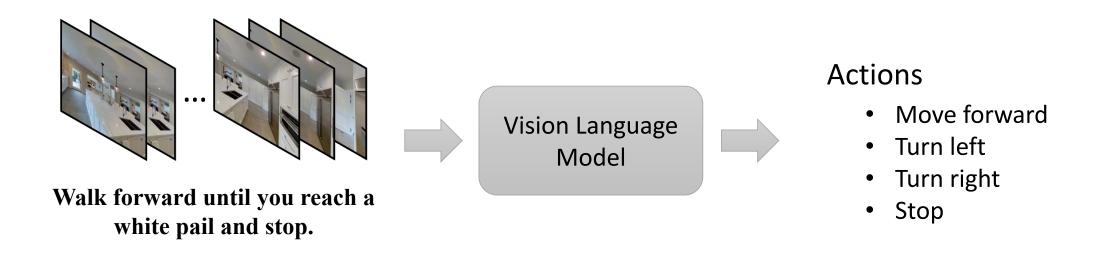




You are in a bedroom. Turn around to the left until you see a door leading out into a hallway, go through it. Hang a right and walk between the island and the couch on your left. When you are between the second and third chairs for the island stop.

#### Key Insight 1 – VLM-Driven Real-World VLN

 Leverage the power of <u>foundational VLMs</u> to extend VLN to <u>real-world</u> applications, using <u>pretrained</u> large model, and <u>co-tuning</u> with <u>web-based data</u>.



510k Navigation Data + 763k Web-based Data → Total 1.2M training data

## Key Insight 2 – Video-Based VLN Agent

 Navigate in a <u>human-like</u> manner, relying <u>solely</u> on real-time <u>video streams</u> from a monocular camera, <u>without the need for maps, odometers, or depth inputs</u>.

Walk out of the bedroom, turn right, stop before the stairs.

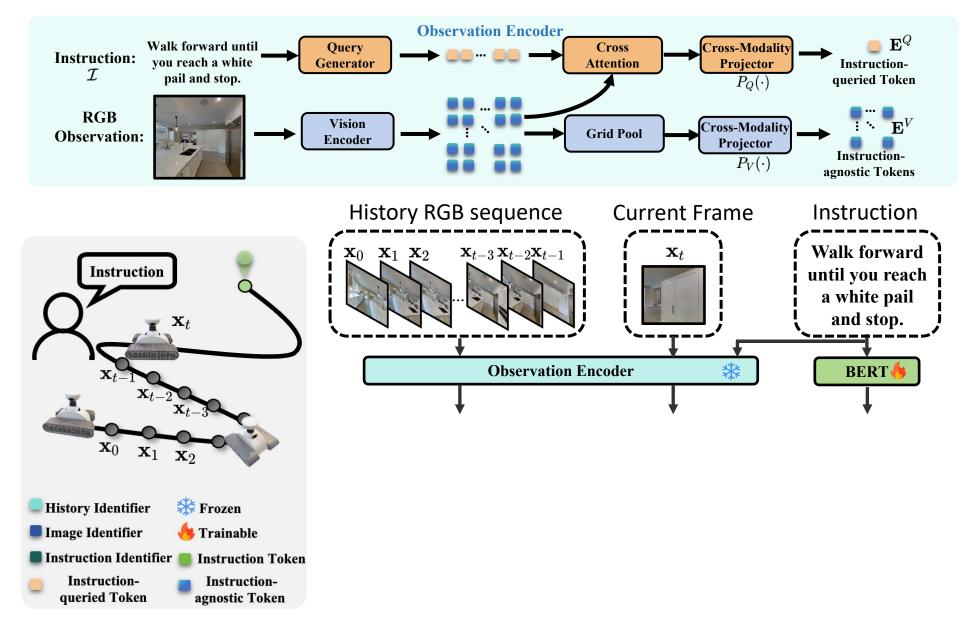
Walk out of the bedroom

Turn right

Stop before the stairs

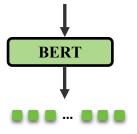
On-the-fly Video as Input

## Pipeline

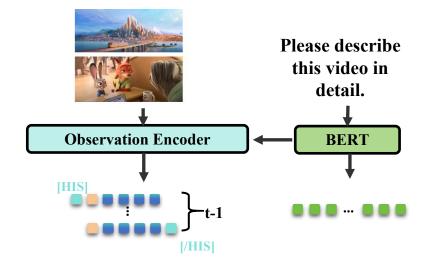


#### Text:

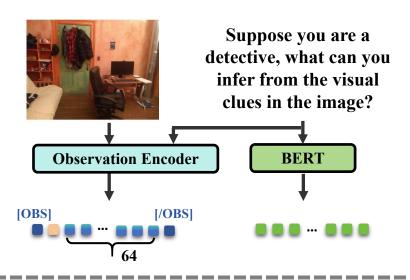
What is large language model?



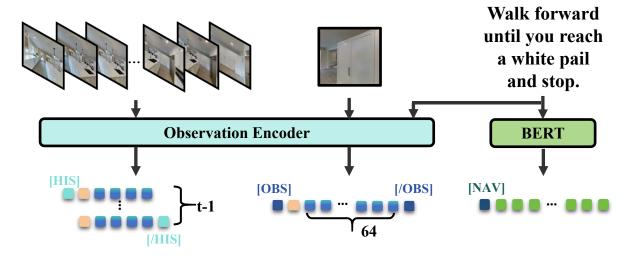
#### Video:



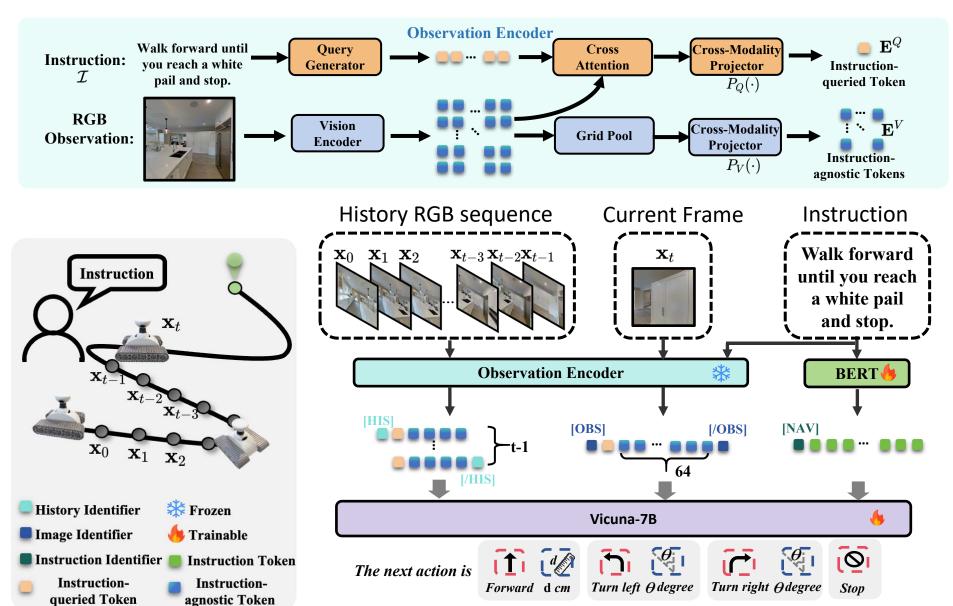
#### Image:



#### Navigation:

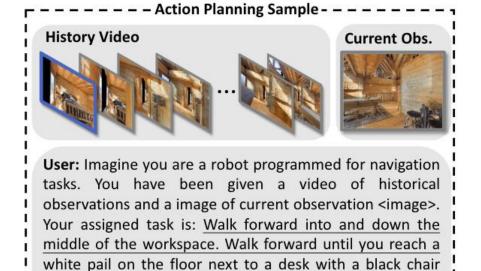


## Pipeline



#### Data collection

We collect the navigation data based on R2R dataset training-split on VLN-CE simulator: 10819 episodes, 61 scenes (MP3D).



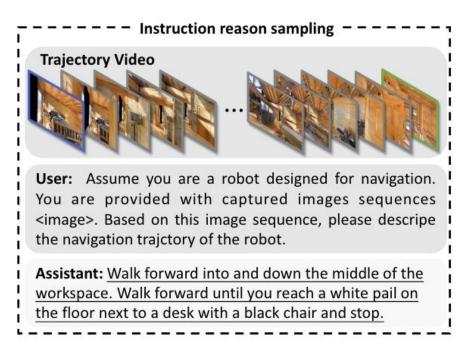
and stop. Analyze this series of images to decide your next

move, which could involve turning left or right by a

Assistant: The next action is move forward 75 cm.

specific degree or moving forward a certain distance.

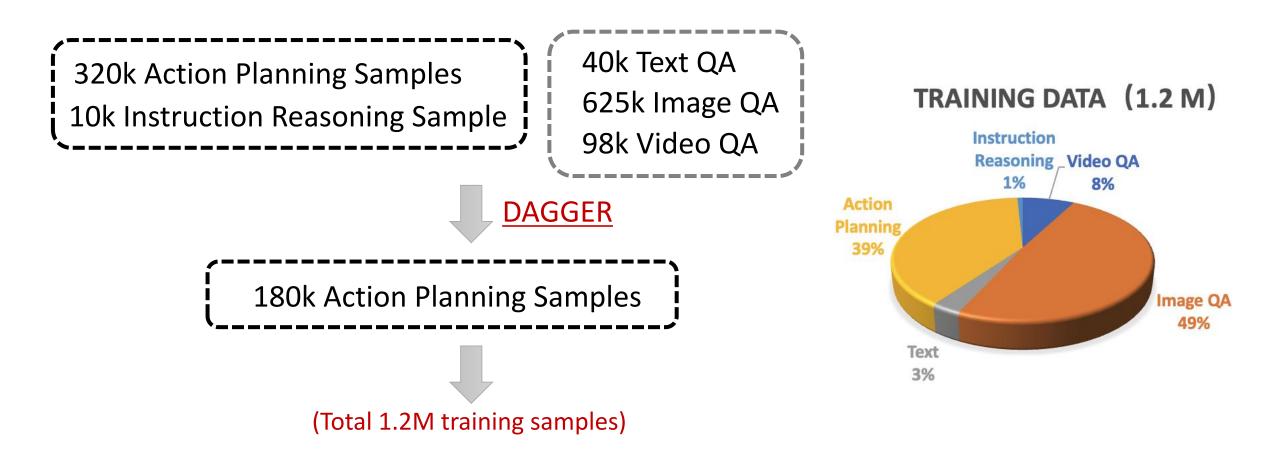
Sample video segment + action (Action Planning Sample)



Video + instruction (Instruction Reasoning Sample)

#### Data collection

We collect the navigation data based on R2R dataset training-split on VLN-CE simulator: 10819 episodes, 61 scenes (MP3D).



## R2R train -> R2R val-unseen (cross split)

		Obser	vation		VLN-CE R2R Val-Unseen						
	Pan.	S.RGB	Depth	Odo.	TL	NE↓	OS↑	SR↑	<b>SPL</b> ↑		
AG-CMTP [15]	<b>V</b>		<b>√</b>	<b>√</b>	53 <del></del> 2	7.90	39.2	23.1	19.1		
R2R-CMTP [15]	✓		$\checkmark$	$\checkmark$	·	7.90	38.0	26.4	22.7		
LAW [73]		$\checkmark$	$\checkmark$	$\checkmark$	8.89	6.83	44.0	35.0	31.0		
CM2 [29]		$\checkmark$	$\checkmark$	$\checkmark$	11.54	7.02	41.5	34.3	27.6		
WS-MGMap [16]		$\checkmark$	$\checkmark$	$\checkmark$	10.00	6.28	47.6	38.9	34.3		
Seq2Seq [43]		$\checkmark$	$\checkmark$		9.30	7.77	37.0	25.0	22.0		
CMA [43]		$\checkmark$	$\checkmark$		8.64	7.37	40.0	32.0	30.0		
RGB-Seq2Seq		$\checkmark$			4.86	10.1	8.10	0.00	0.00		
RGB-CMA		$\checkmark$			6.28	9.55	10.8	5.00	4.43		
Ours		$\checkmark$			7.63	5.47	49.1	37.4	35.9		

↑ SR (success rate)

↑ OS (oracle success rate)

↑ SPL (success weighted by path length)

**↓** NE (Navigation error)

SOTA level performance with only RGB video inputs

## R2R train -> RxR val-unseen (cross dataset)

	Ob	servatio	n	VLN-CE RxR Val-Unseen							
	S.RGB	Depth	Odo.	TL	NE↓	OS↑	SR↑	<b>SPL</b> ↑			
LAW [73]	<b>√</b>	<b>√</b>	<b>√</b>	4.01	10.87	21.0	8.0	8.0			
CM2 [29]	✓	$\checkmark$	$\checkmark$	12.29	8.98	25.3	14.4	9.2			
WS-MGMap [16]	✓	$\checkmark$	$\checkmark$	10.80	9.83	29.8	15.0	12.1			
Seq2Seq [43]	✓	$\checkmark$		1.16	11.8	5.02	3.51	3.43			
CMA [43]	✓	$\checkmark$		5.09	11.7	10.7	4.41	2.47			
RGB-Seq2Seq	✓			4.43	11.2	12.2	0.0	0.0			
RGB-CMA	✓			13.56	9.55	14.8	0.0	0.0			
$A^2$ Nav [17]	✓			_	_	_	16.8	6.3			
Ours	✓			10.59	8.41	34.5	23.8	21.2			

↑ SR (success rate)

↑ OS (oracle success rate)

↑ SPL (success weighted by path length)

**↓** NE (Navigation error)

Our method consistently demonstrates SOTA performance, significantly surpassing baseline metrics.

## R2R train -> Real world (Sim-to-real)

	Meeting Room			Office				Lab				Lounge				
	Simple I.F.   Complex I.F.		Simple I.F. C		Complex I.F.		Simple I.F.		Complex I.F.		Simple I.F.		Complex I.F.			
	SR↑	NE↓	SR↑	NE	SR↑	NE↓	SR↑	NE↓	SR↑	NE↓	SR↓	NE↓	SR↑	NE↓	SR↑	NE↓
Seq2Seq [42]	4%	4.45	0%	7.21	0%	4.28	0%	6.92	0%	4.58	0%	6.61	0%	5.95	0%	6.82
CMA [42]	0%	4.27	0%	7.30	8%	4.62	0%	5.71	4%	4.35	0%	5.67	0%	4.63	0%	5.46
WS-MGMap [45]	52%	1.18	24%	2.20	60%	0.96	20%	2.94	44%	1.85	12%	3.18	48%	1.66	32%	2.88
Ours	92%	0.55	56%	0.98	84%	0.63	48%	0.71	76%	0.83	40%	1.89	88%	0.72	44%	1.37

#### **Example:**



## Real-world experiment – Simple Instruction

Simple Instruction following

Speed up x10

Walk towards the door then stop.

Walk towards the white box then stop.







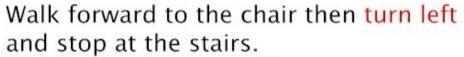


#### Real-world experiment - Outdoor Scenes

#### Simple Instruction following

Speed up x10

Walk forward to the chair then turn right, and stop at the stairs.











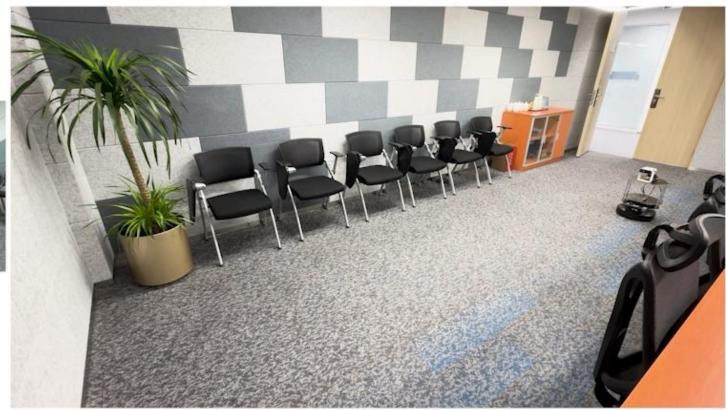
## Real-world experiment – Complex Instruction

#### Complex Instruction following

Speed up x10

Go straight and move close to the plant, then turn right facing the door, then walk to the door and stop.





# Thanks for your attention

Project page: https://pku-epic.github.io/NaVid/











GALBOT

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