



**Carnegie Mellon University**

Language Technologies Institute

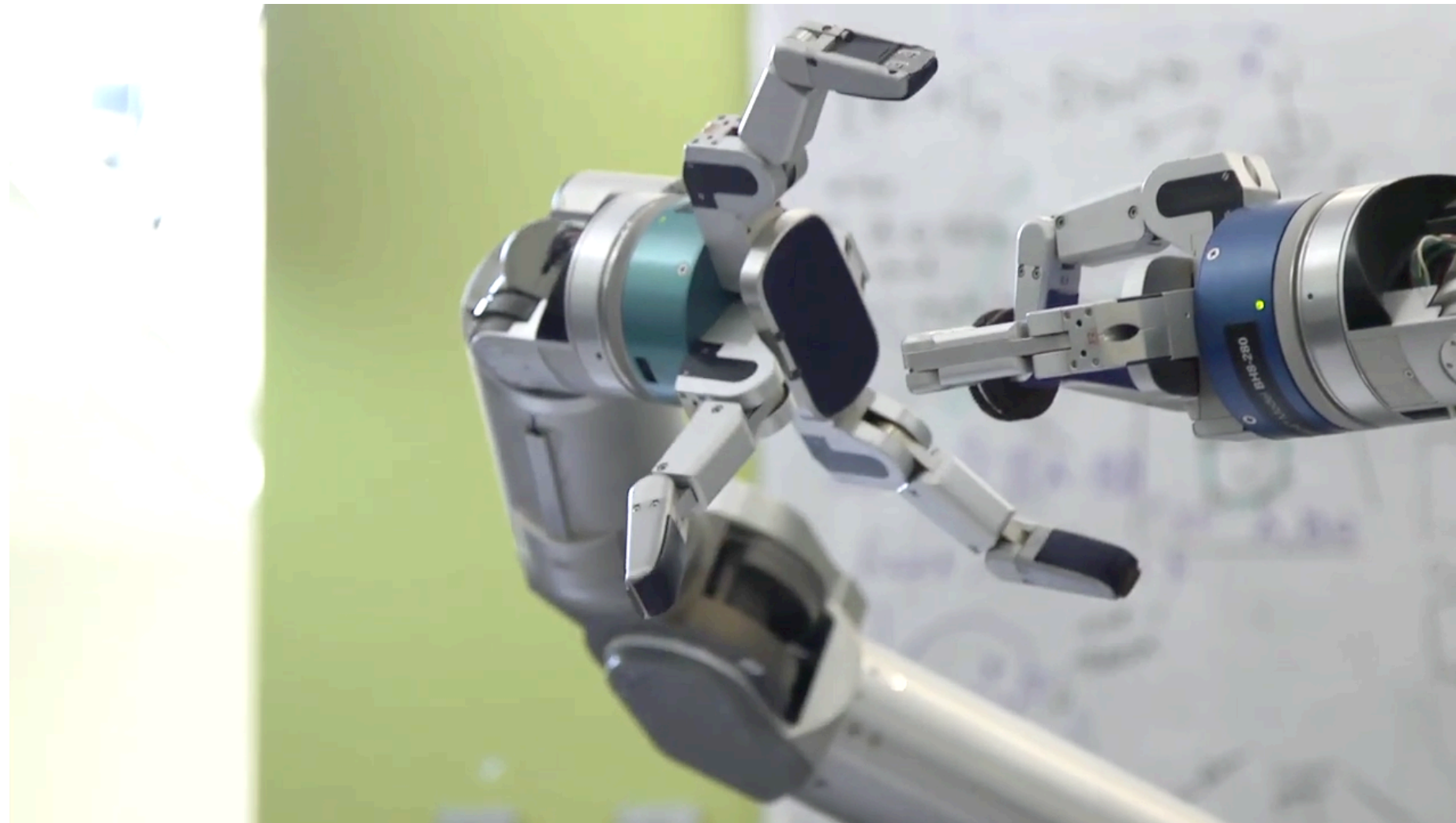
# Connecting Language to Actions

An MMLL choose your own adventure game

# Why?

Language that affects the world

*Remove the cream from the middle of the Oreo...*



HERB (Siddhartha Srinivasa)

Access to Broader Semantics

What's it like to drive a bus?



How many hours of watching to achieve same level of performance as 30m of practice?

# What does interaction mean?

## Grid World?



Reinforcement Learning: Crash Course AI#9  
<https://www.youtube.com/watch?v=nlgiv4lfJ6s>

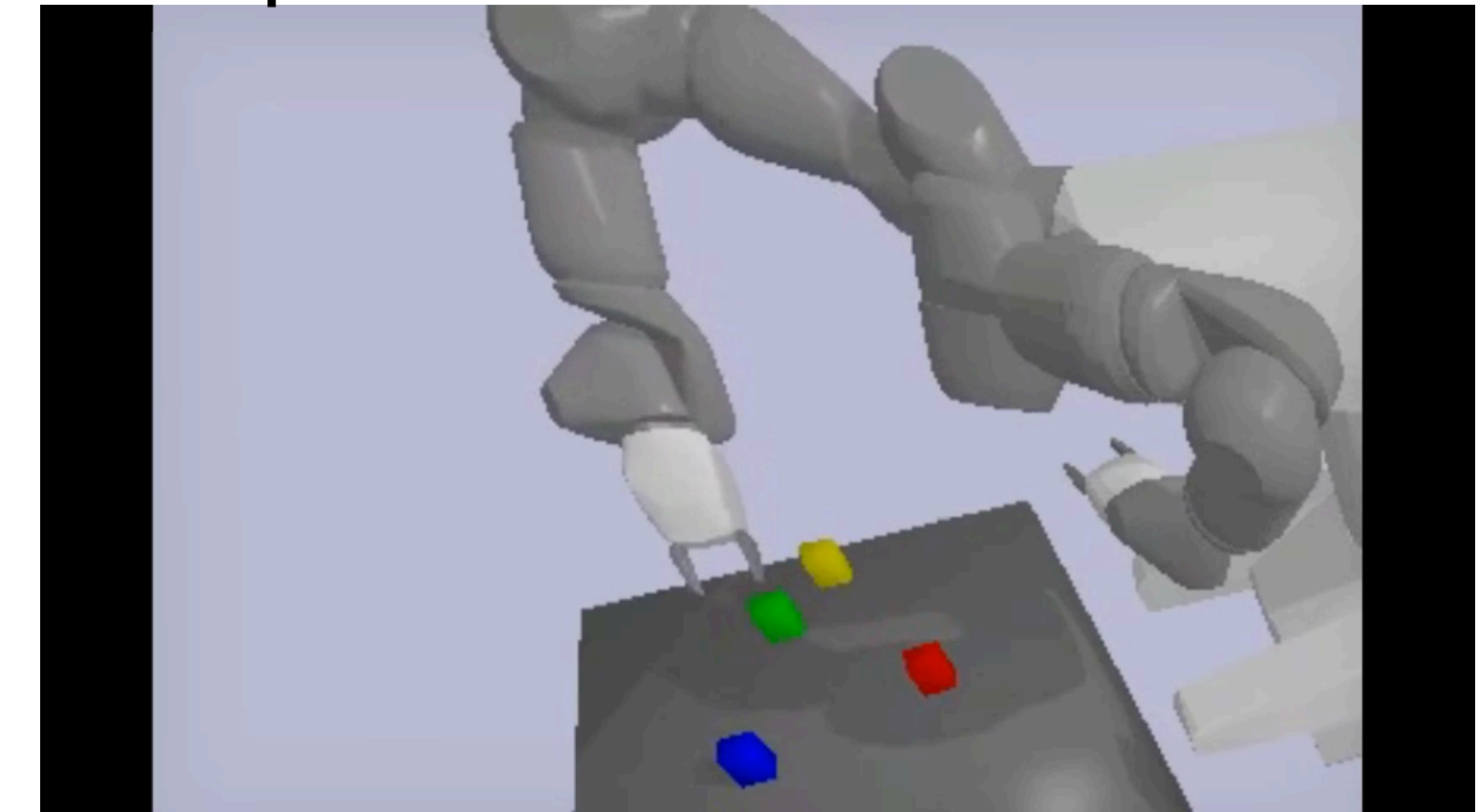
## Graph Navigation?



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

Anderson 2018

## Manipulation?

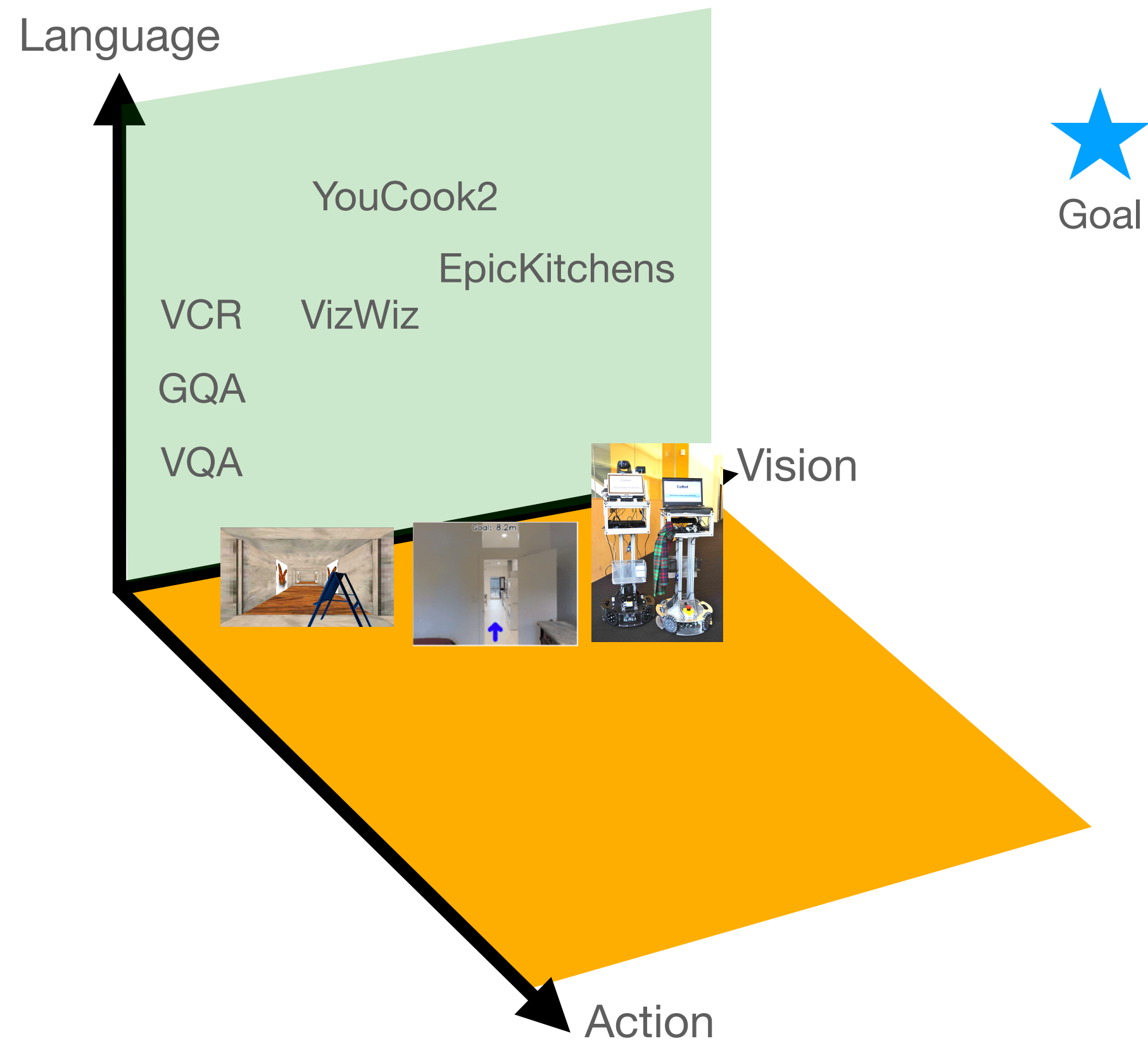


Paxton 2019

1. How does the agent move?
2. How many arms or legs does it have?
3. How many fingers (if any) do the grippers have?
4. How many joints do the limbs have?
5. What about physics? Real motor noise?

...

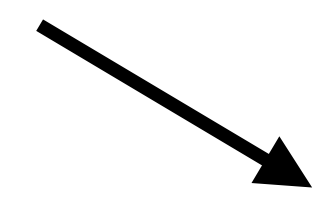
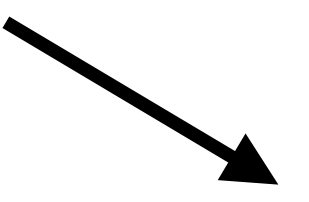
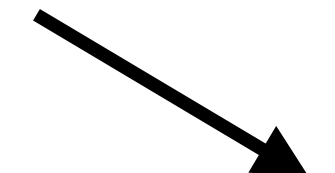
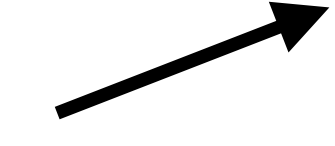
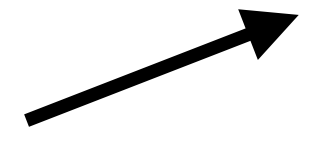
# Every Dimension Interacts



1. How rich or abstract is the language?
2. How complex is the visual field?
3. Is the vision 2D, 3D, Lidar, ... ?
4. What kind of supervision do you have?

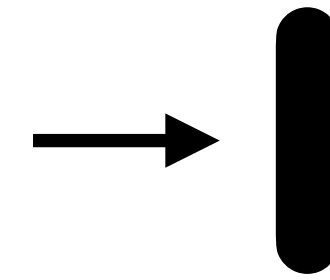
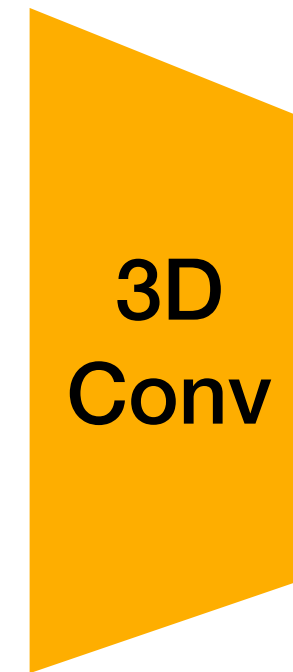
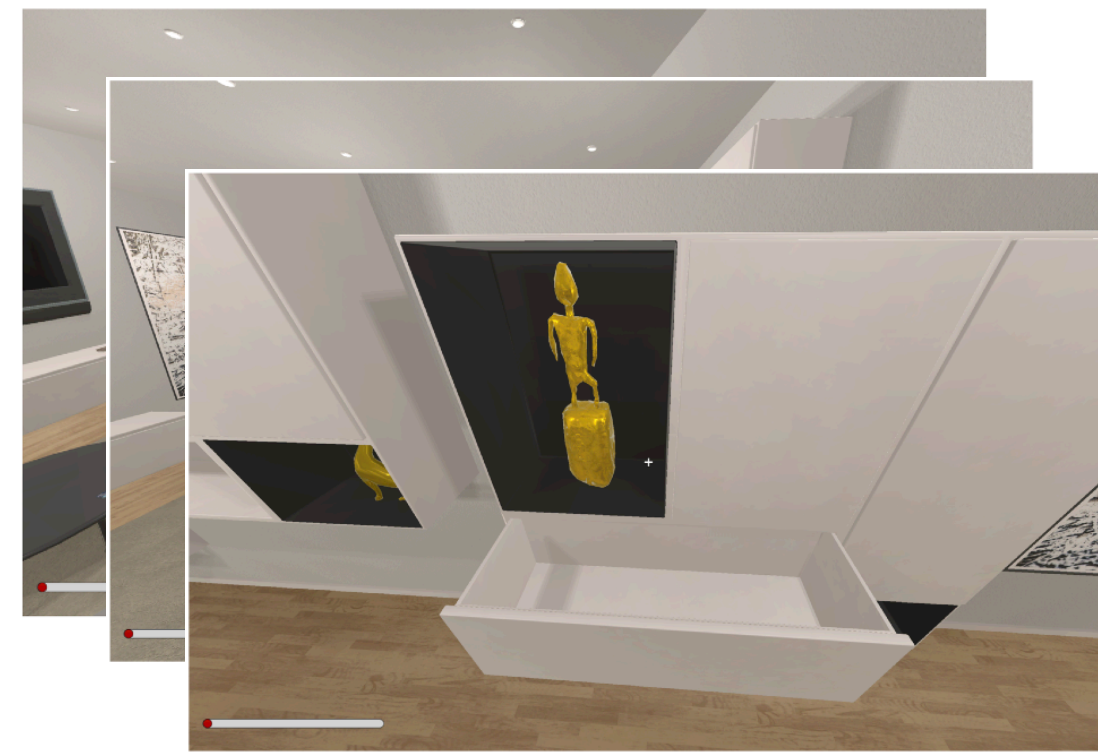
...

# Choose your own adventure



# Sequential and Online Modeling

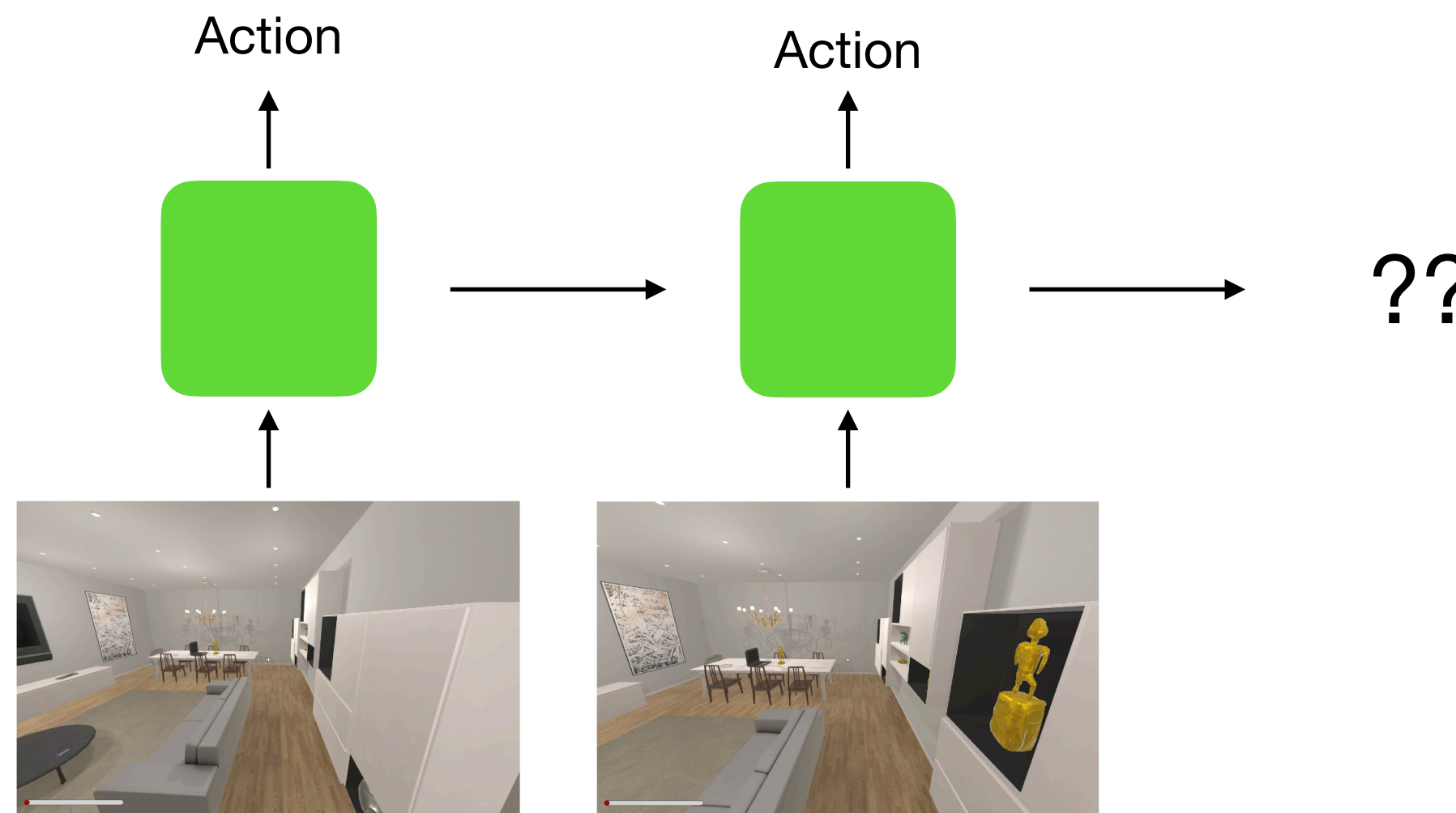
Action Recognition



“Action Summary”

$$p(\text{Action} | v_0, \dots, v_t)$$

Embodied



$$p(v_t | v_0, \dots, \text{Action})$$

Requirement: Have a goal

# What is a “goal”?

“Put the green dog on the table”

$$p(v_t | v_0, \dots, \text{Action})$$

$$p(v_t | v_0, \dots, v_{t-1}, a_0, \dots, a_t)$$

$v_t =$



# Planning

## Pre- and Post-Conditions

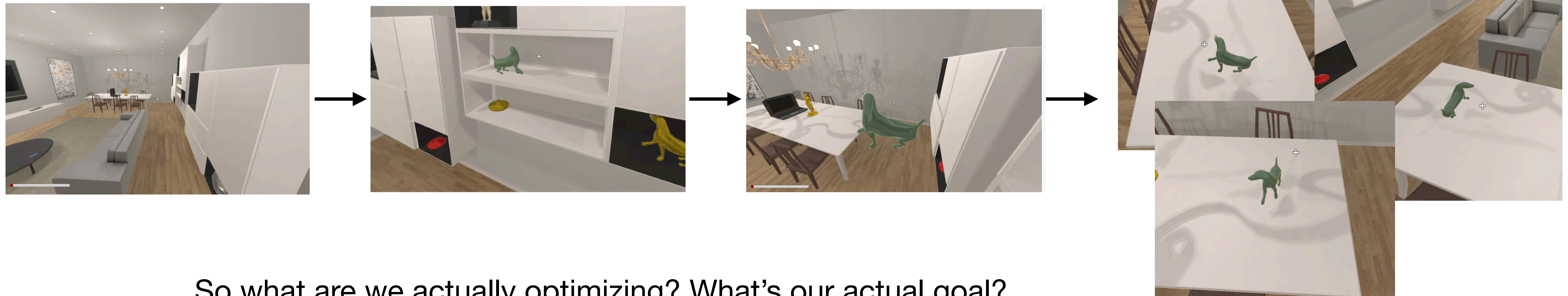
Task 4: Must locate object,  
to move to object

Task 3: Must move to object,  
to hold object

Task 2: Must hold object,  
to place object

Task 1: Recognize Success

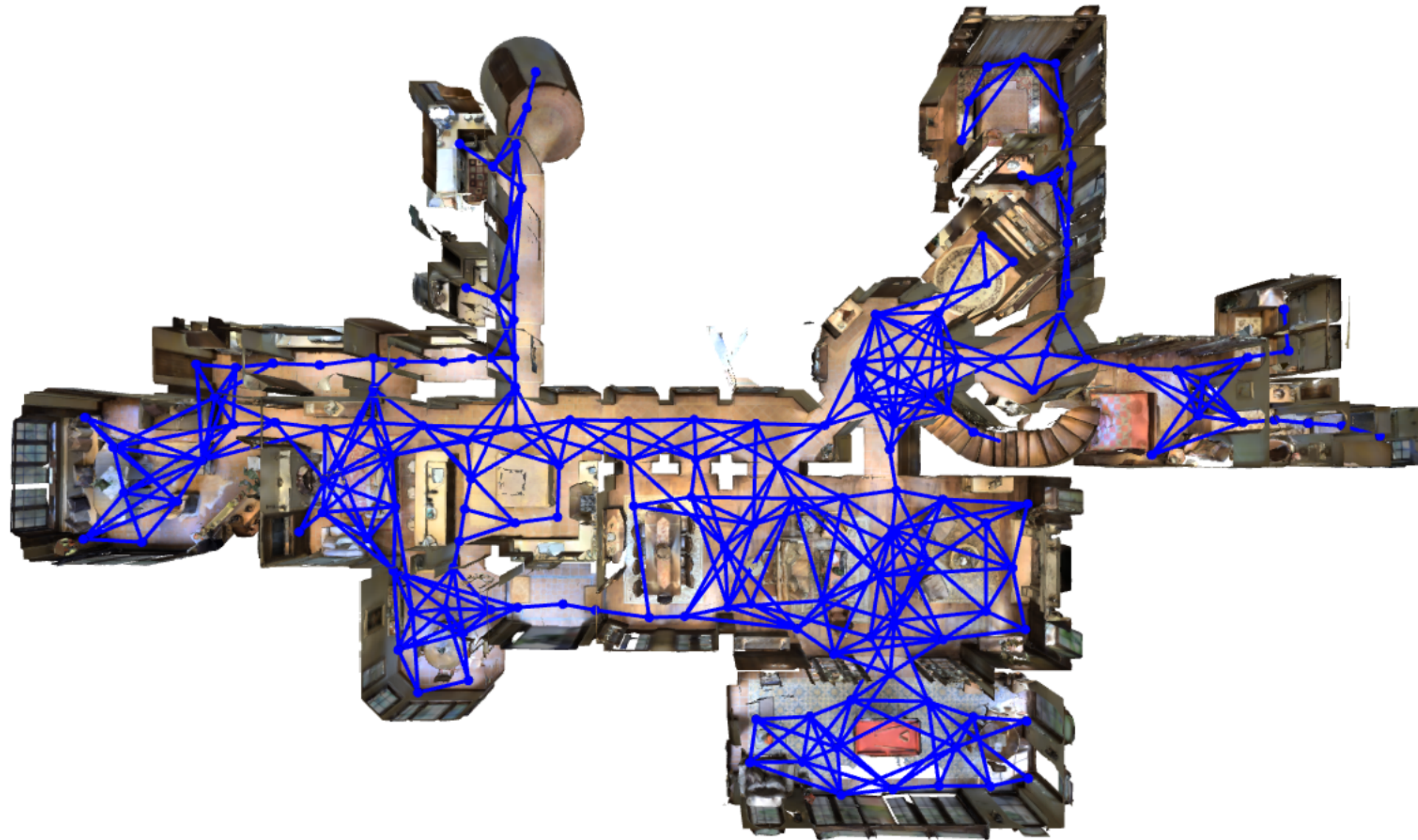
Instances of “green dog sculpture on table”



So what are we actually optimizing? What's our actual goal?



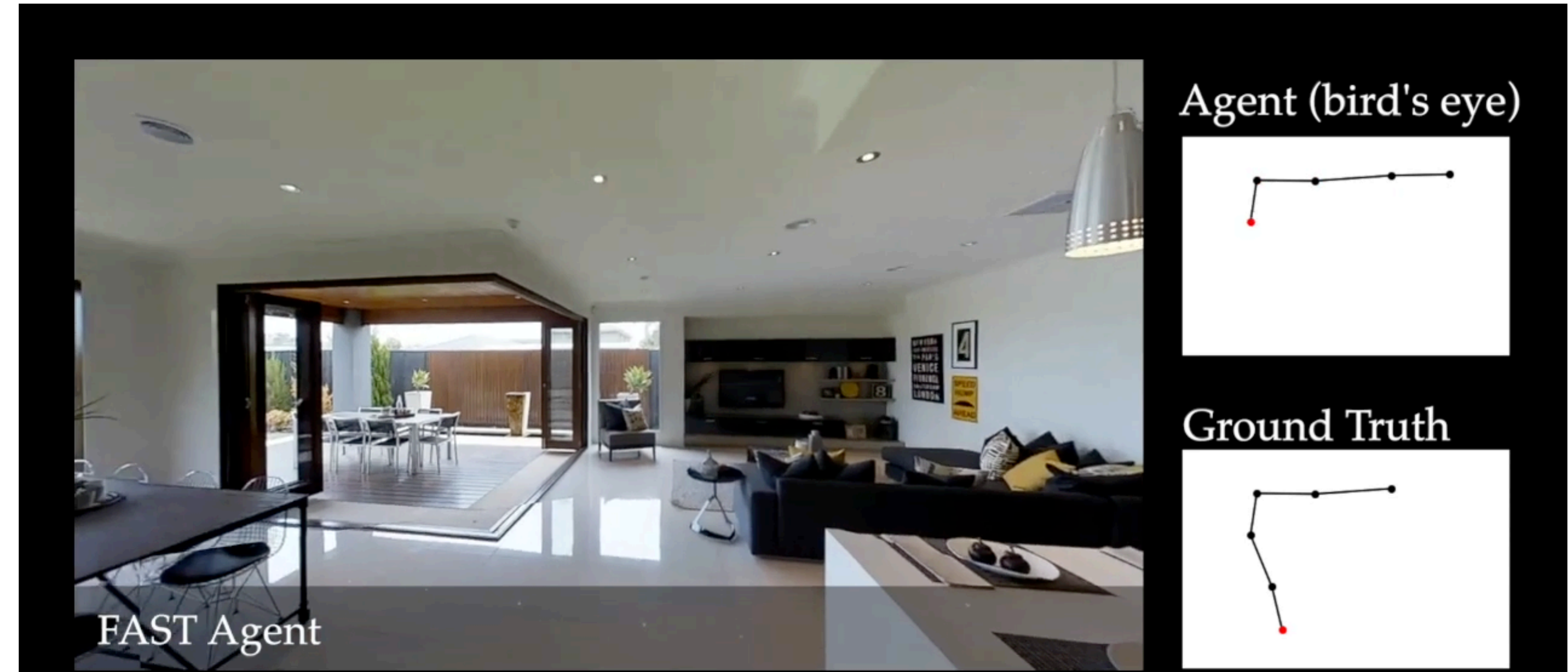
# Let's Start Simple



# Instruction Following

## Explicit Action Supervision

- Walk out of the bedroom through the open door into the hallway
- Turn the corner and walk into the dining area.
- Pass the dining table and walk into the living room area towards the television.
- Stop near the chair and open sliding doors to outside



$V+L \rightarrow A$



Does this actually need vision?

Does this understand plans?

No, this is ~Semantic Parsing

$V+L \rightarrow A$

Walk out of the bedroom through the open door into the hallway



Does this actually need vision?

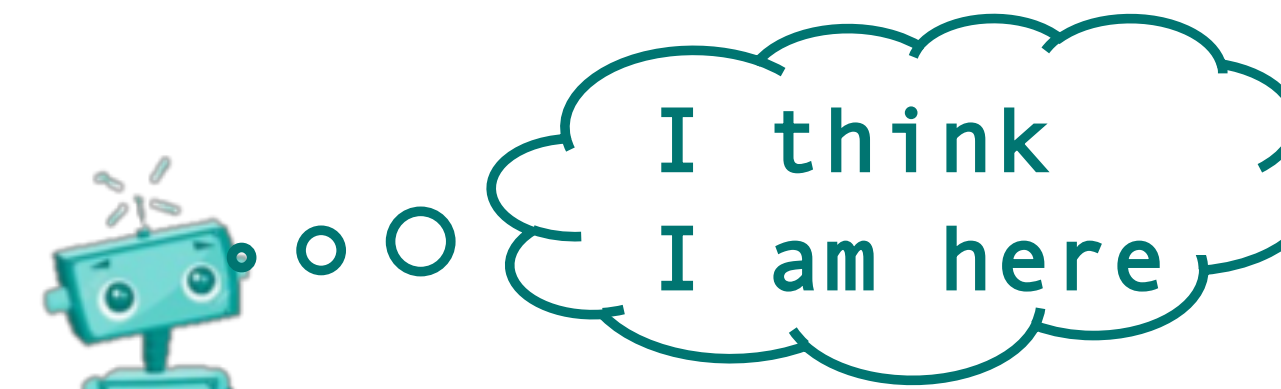
Yes

Does this understand plans?

Maybe, probably not

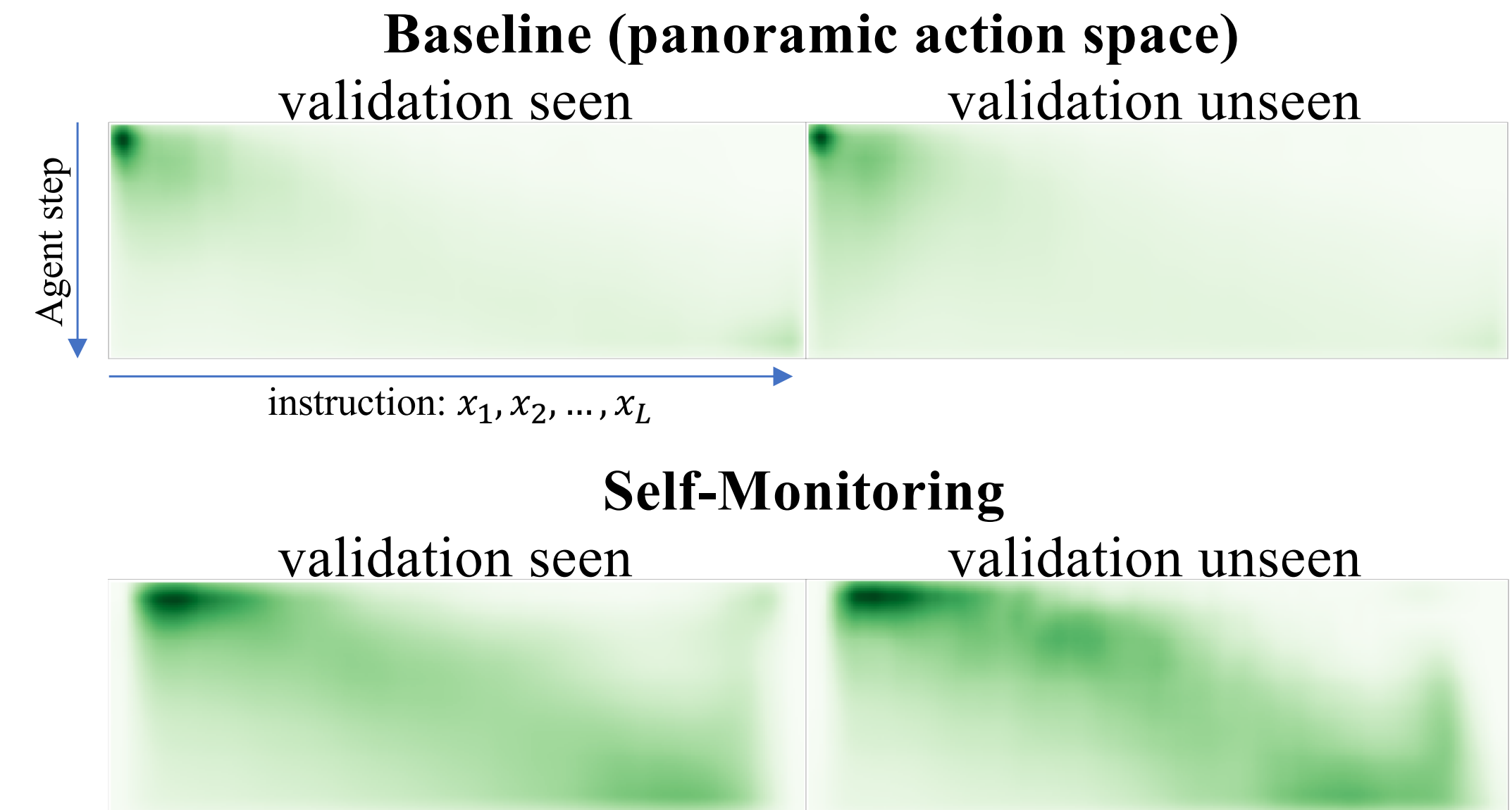
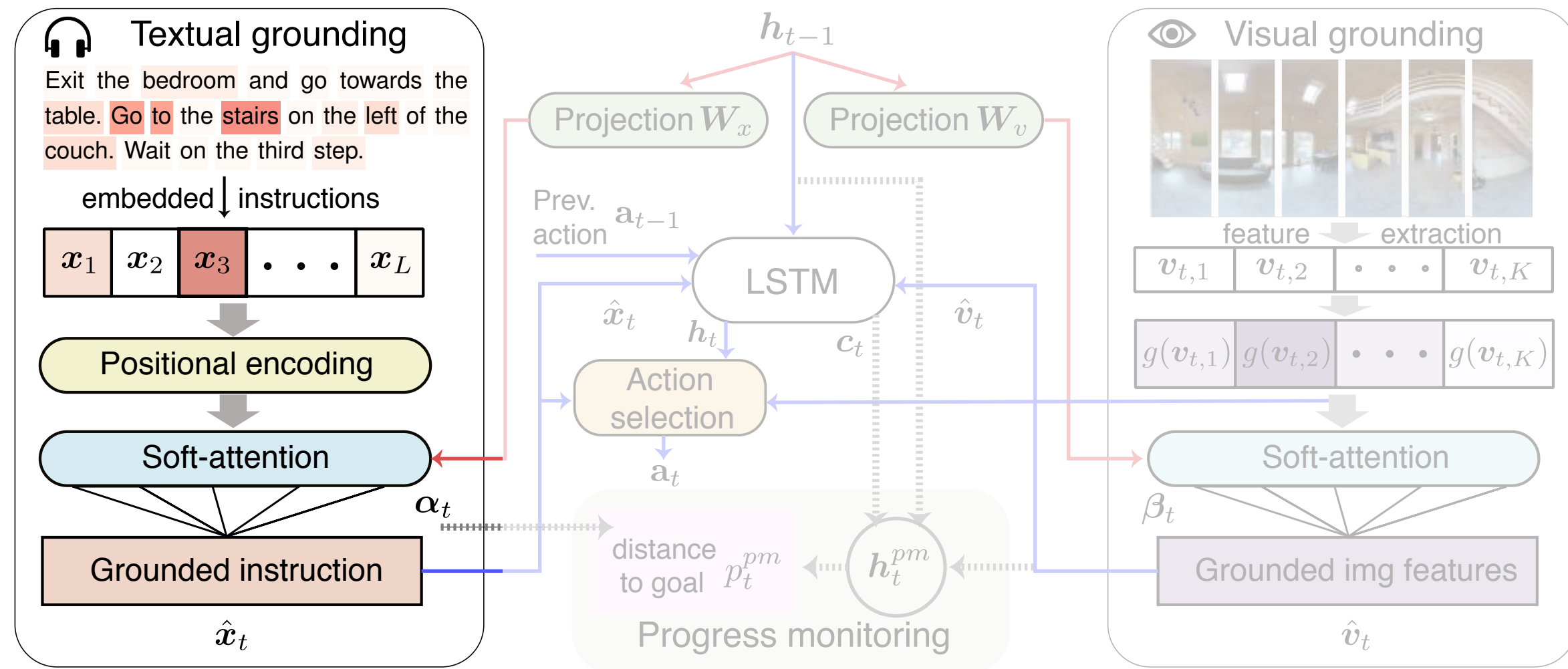
# First Major Question: Alignment

Exit the bedroom and go towards the table. Go to the stairs on the left of the couch. Wait on the third step.

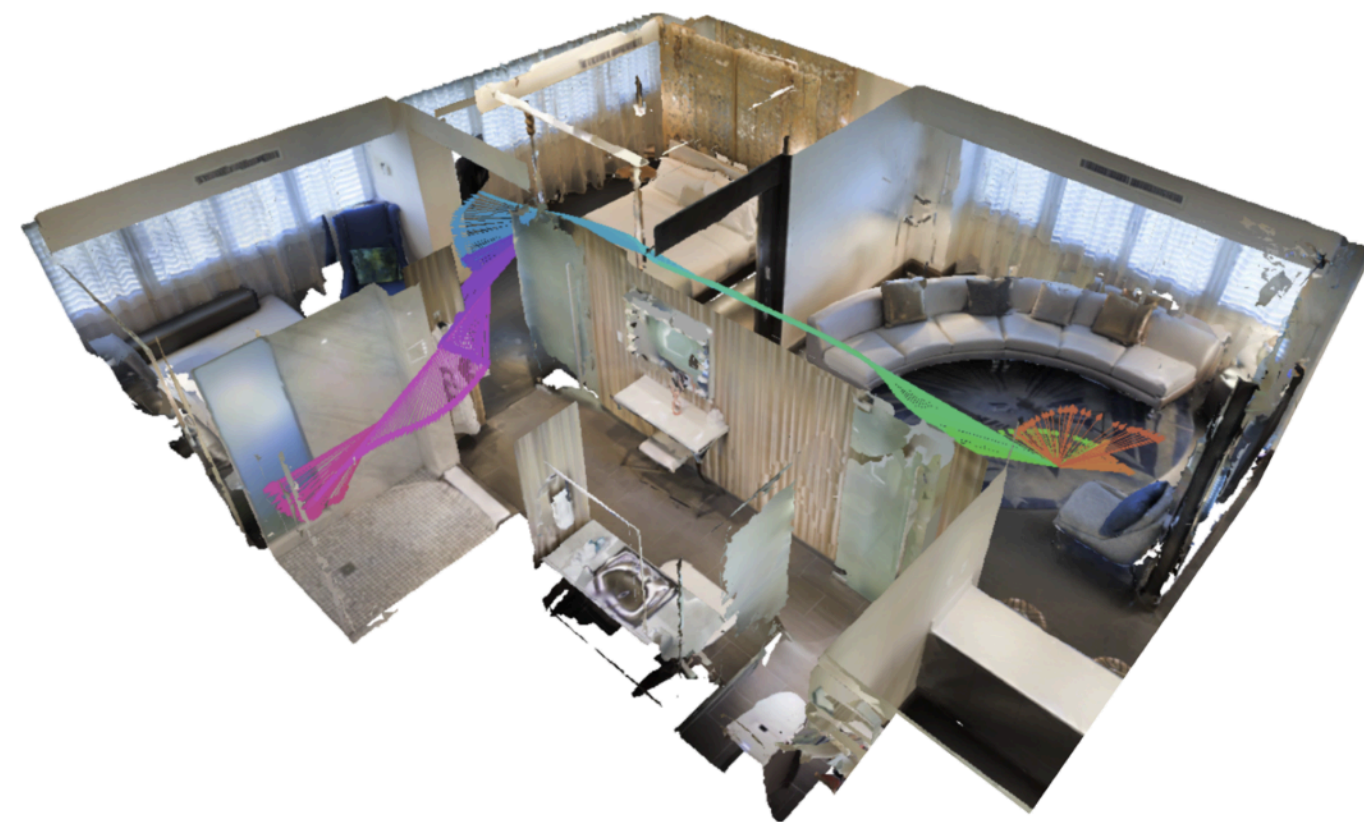


# Alignment

~~Exit the bedroom and go towards the table.~~ Go to the stairs on the left of the couch. Wait on the third step.



# Lots of Data



Our starting point is in a living room, we're facing towards a long beige sofa, and in front of the sofa there are three glass coffee tables, turn around and exit through the doorway that's in front of you, walk pass the bed that's on your right and then turn left, we're now facing towards another living room, and on the left there's an open door, walk towards that open door enter the bathroom that's in front of you, turn towards the right into the shower area. and that's your destination.

		Number of:		Includes:		
	Lang	Instruct	Words	Paths	Text	Ground Demos
CVDN	1	2K <sup>†</sup>	167K	7K	✓	
R2R	1	22K	625K	7K	✓	
Touchdown	1	9K	1.0M	9K	✓	✓ <sup>‡</sup>
REVERIE	1	22K	388K	7K	✓	✓ <sup>‡</sup>
RxR	3	126K	9.8M	16.5K	✓	✓ ✓

<sup>†</sup>The number of dialogues. <sup>‡</sup>Grounding limited to one object per instruction.

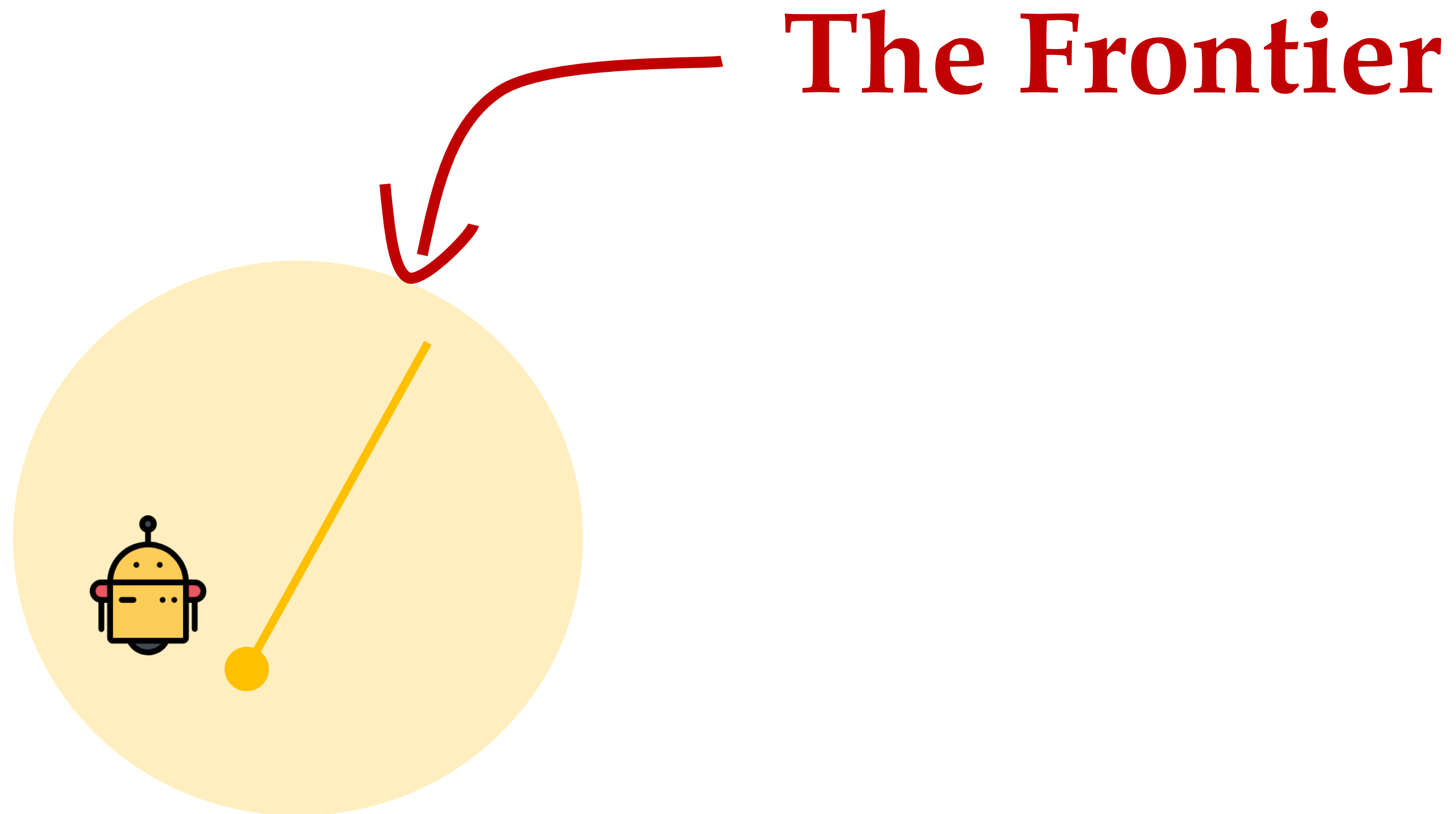
Ku et al. Room-Across-Room: Multilingual Vision-and-Language Navigation with Dense Spatiotemporal Grounding — EMNLP 2020

# Lots and lots of aligned data?

Wait, remember the bus driver question?



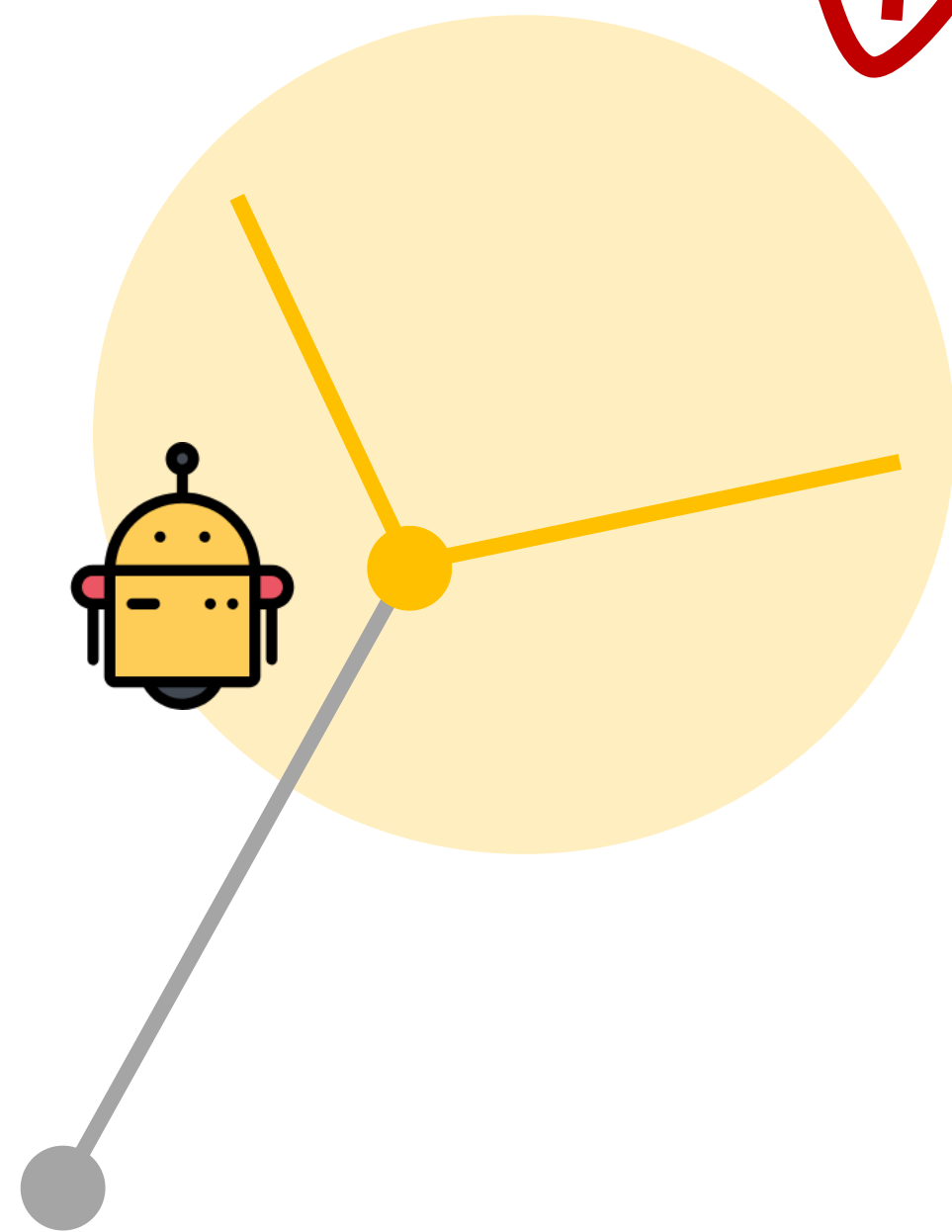
# What if you make a mistake?





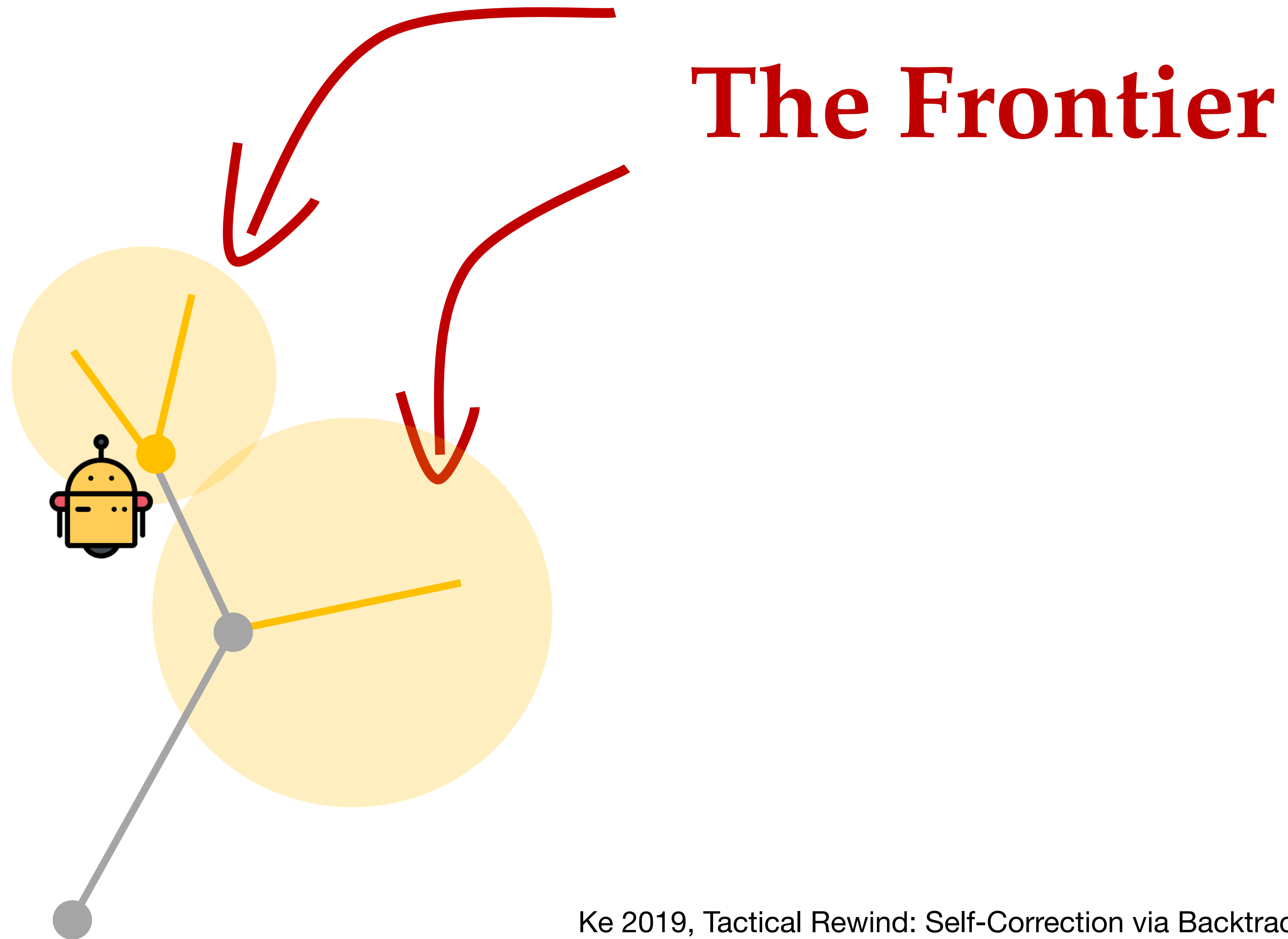
# What if you make a mistake?

## The New Frontier



Ke 2019, Tactical Rewind: Self-Correction via Backtracking in Vision-and-Language Navigation - CVPR 2019

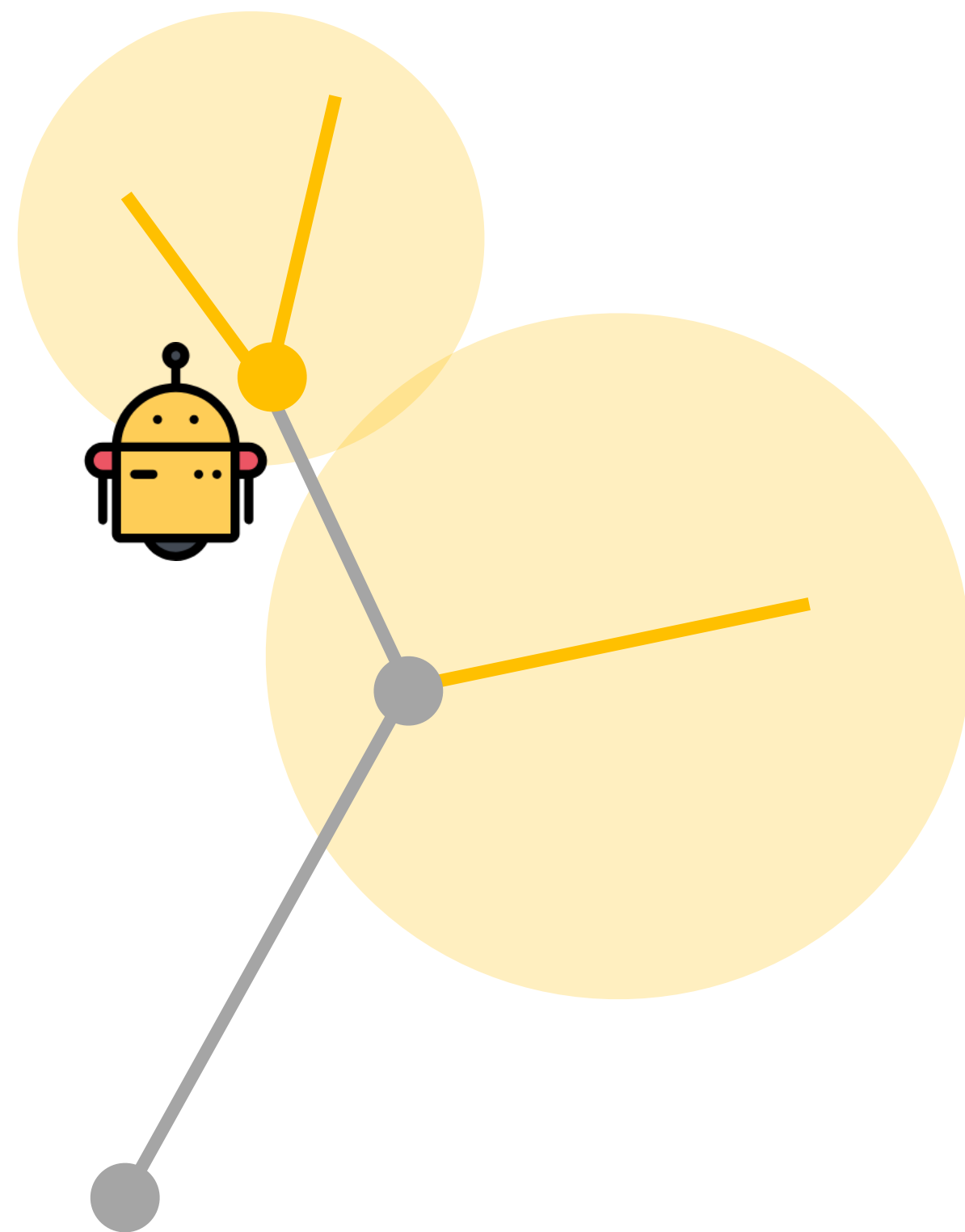
# What if you make a mistake?



Ke 2019, Tactical Rewind: Self-Correction via Backtracking in Vision-and-Language Navigation - CVPR 2019

# What if you make a mistake?

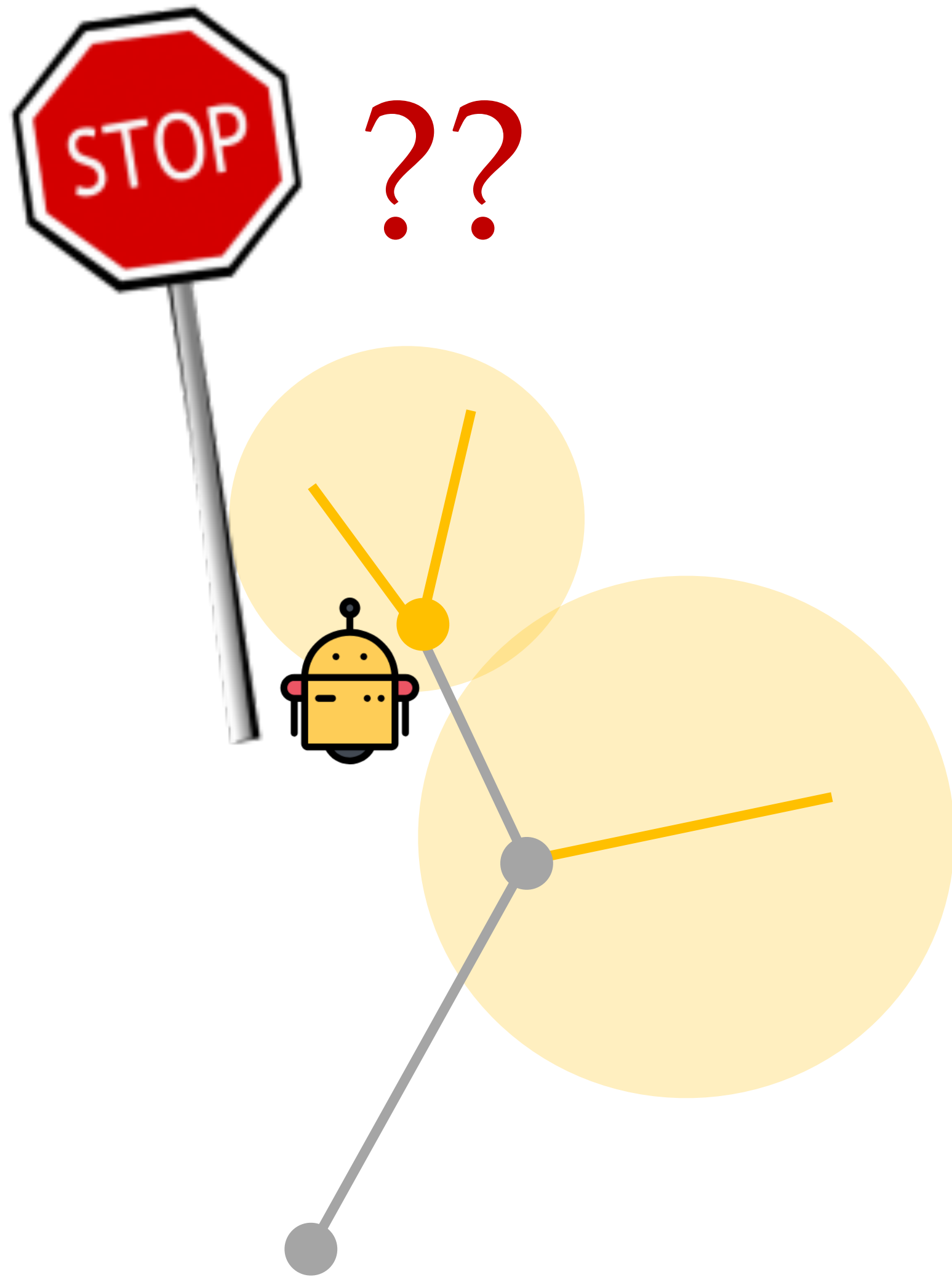
## Eventually ...



Ke 2019, Tactical Rewind: Self-Correction via Backtracking in Vision-and-Language Navigation - CVPR 2019

# What if you make a mistake?

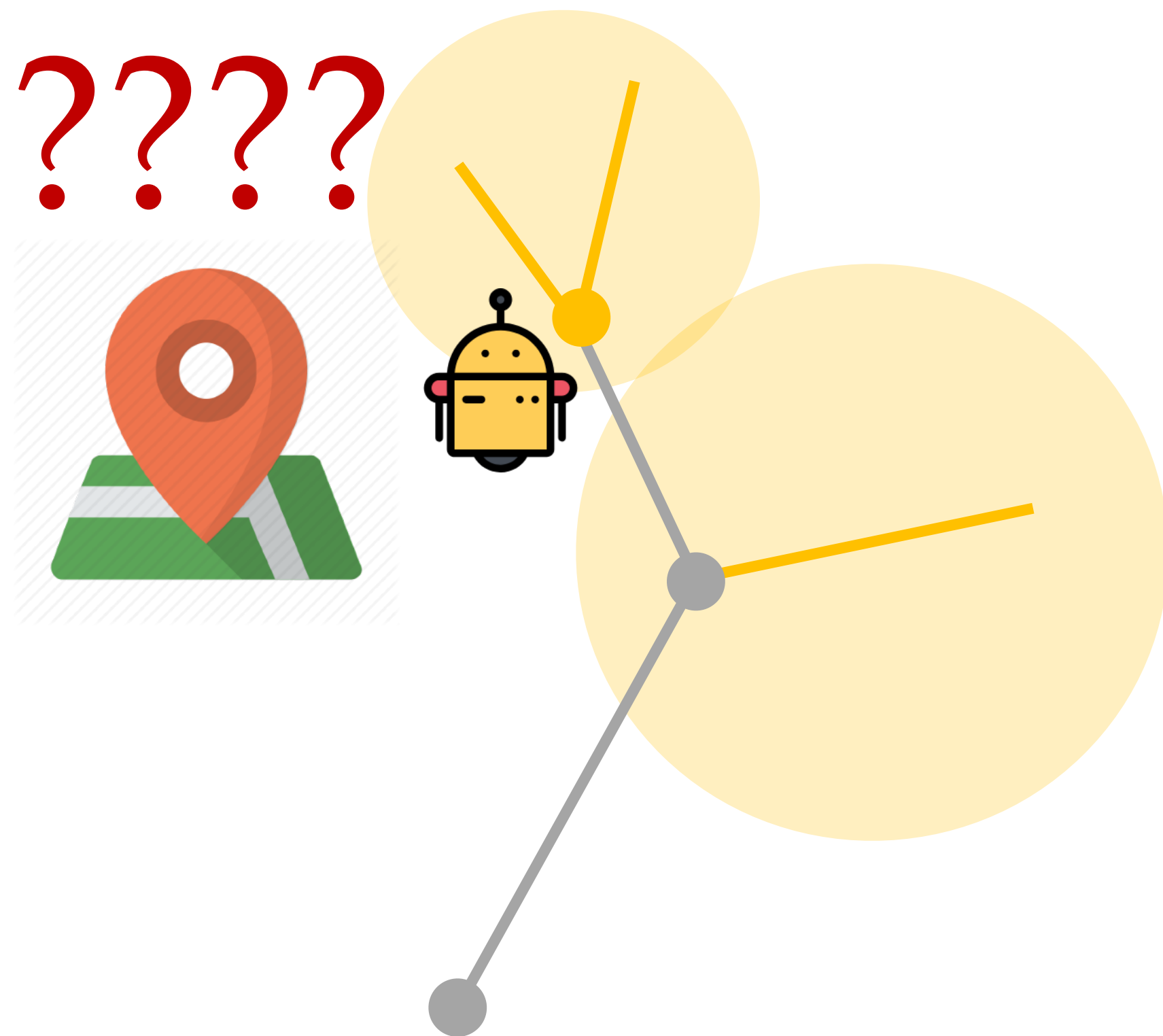
## 1. Did I reach the target?



Ke 2019, Tactical Rewind: Self-Correction via Backtracking in Vision-and-Language Navigation - CVPR 2019

# What if you make a mistake?

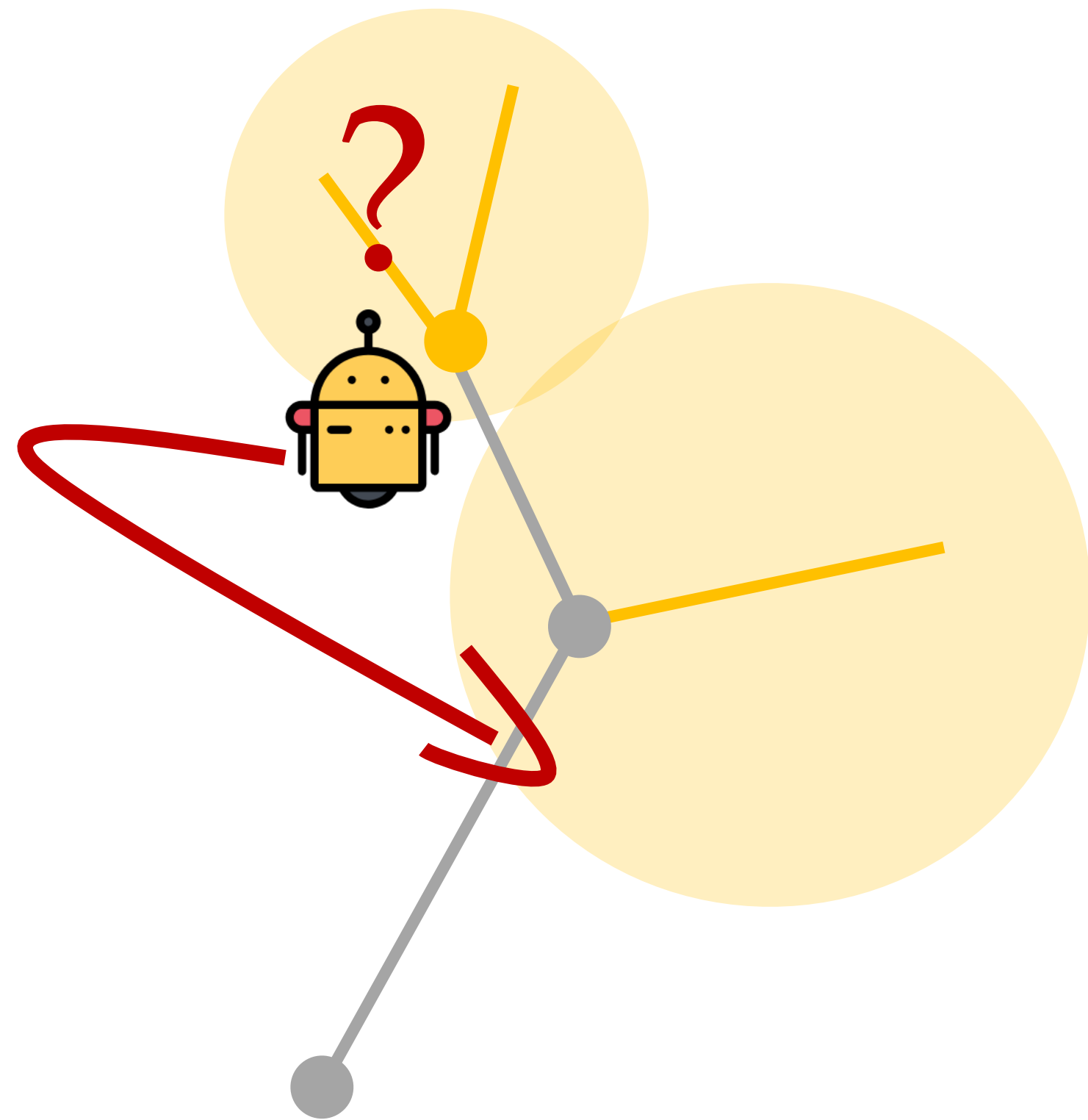
1. Did I reach the target?
2. Am I lost?



Ke 2019, Tactical Rewind: Self-Correction via Backtracking in Vision-and-Language Navigation - CVPR 2019

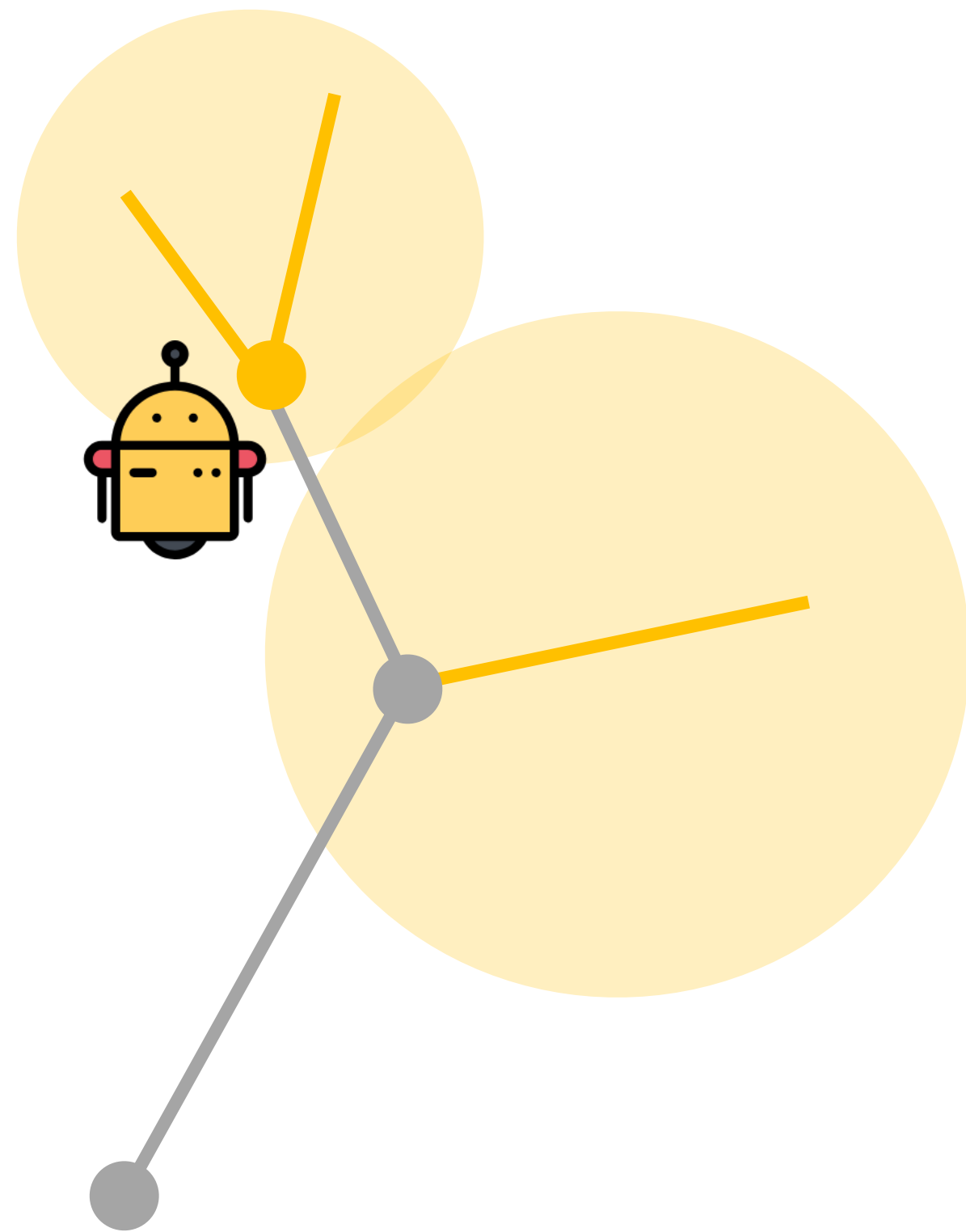
# What if you make a mistake?

1. Did I reach the target?
2. Am I lost?
3. Should I backtrack?

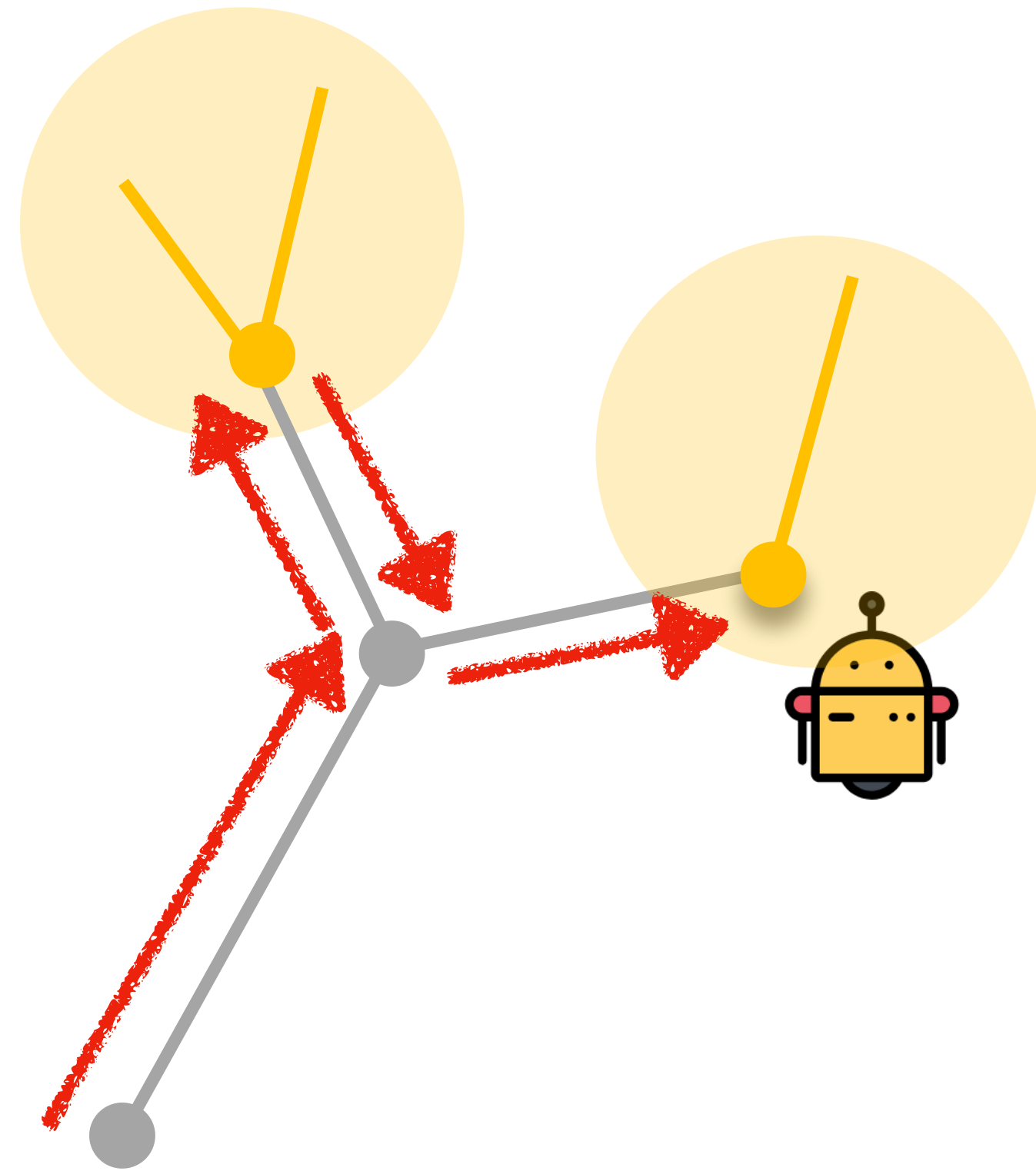


# What if you make a mistake?

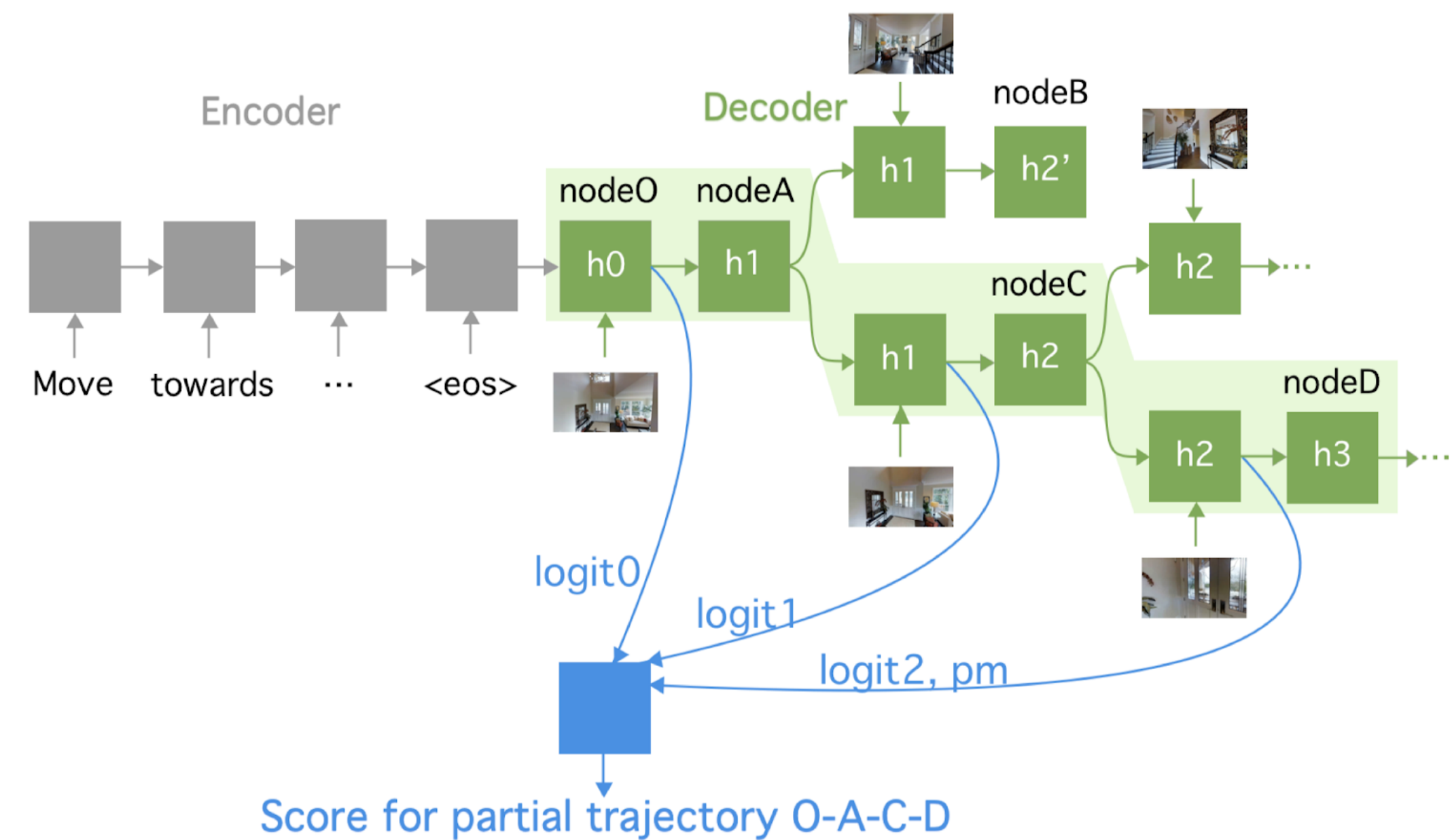
1. Did I reach the target?
2. Am I lost?
3. Should I backtrack?
4. **Where to backtrack to?**



# What if you make a mistake?

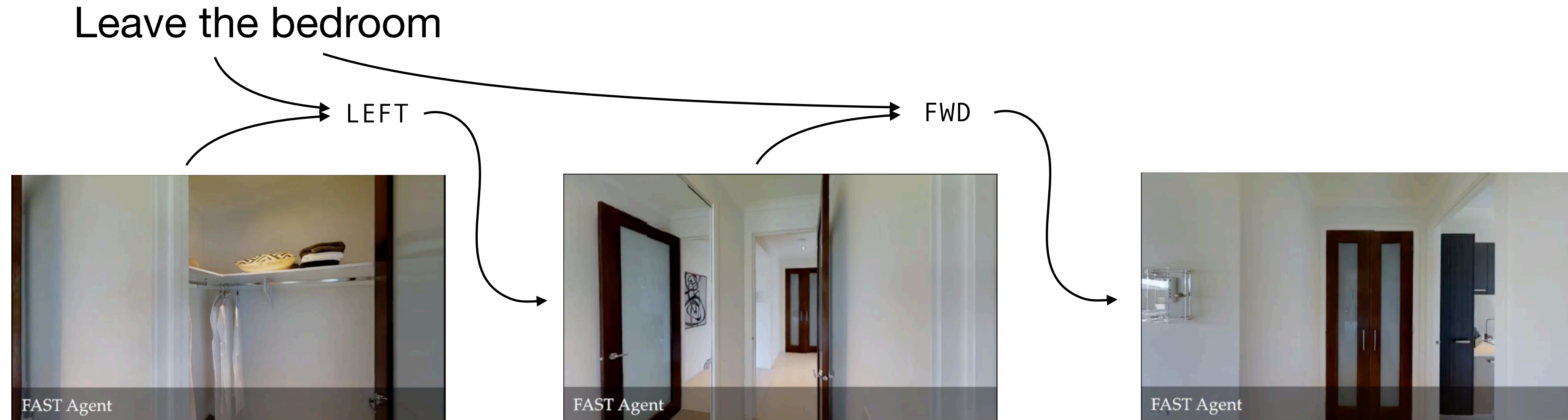


A lot of the visual observations and actions have no correspondence to the language





# Underspecification



Does this actually need vision?

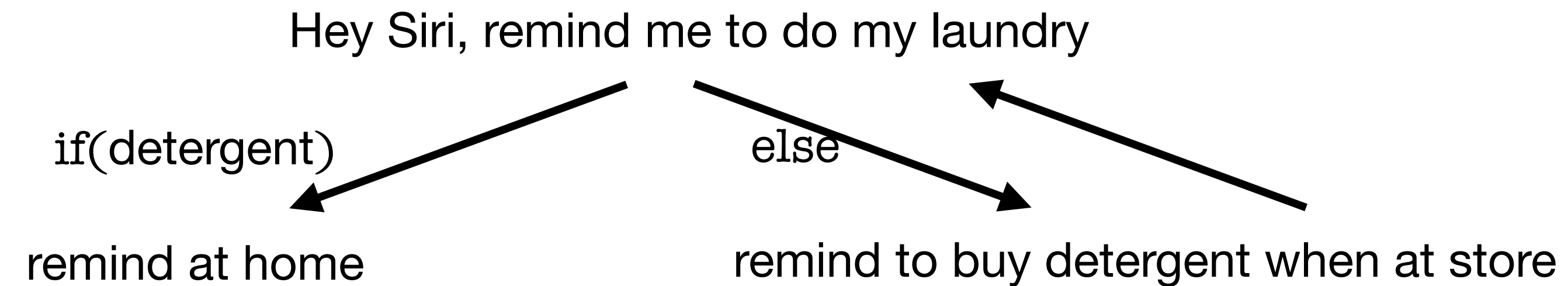
Yes

Does this understand plans?

Maybe?

# Why does this question matter?

Because in general, we can't supervise everything



Hey Siri-bot, do my laundry



Go to hamper...

# ALFRED

Action Learning From Realistic Environments and Directives



# Seven High-level Tasks

Paths are generated by planner



Pick & Place



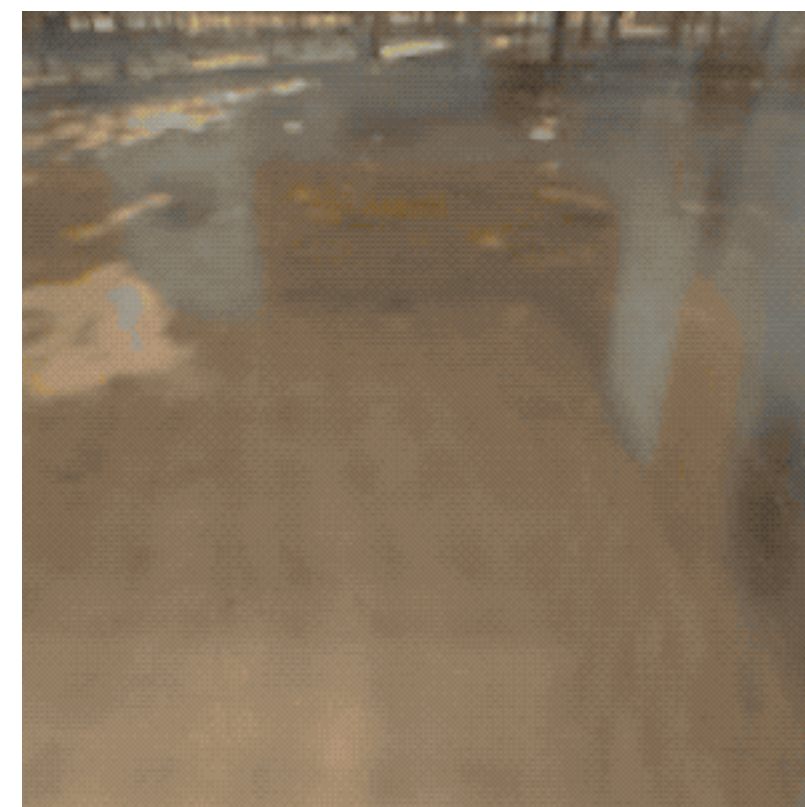
Double Place



Stack



Examine



Heat



Cool



Rinse

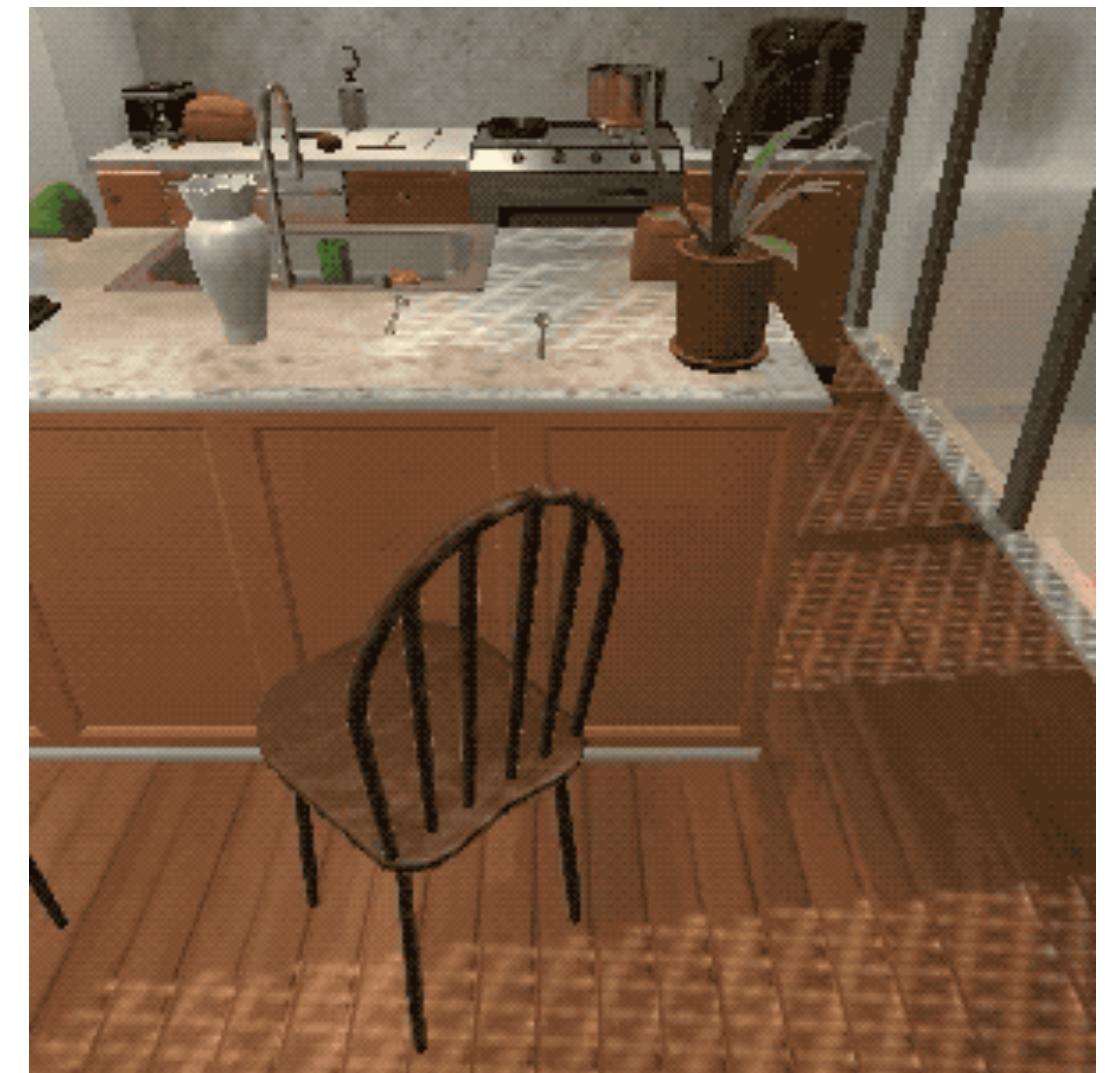
# Data collection

Tuple (Stack, Fork, Cup, CounterTop, Kitchen3)

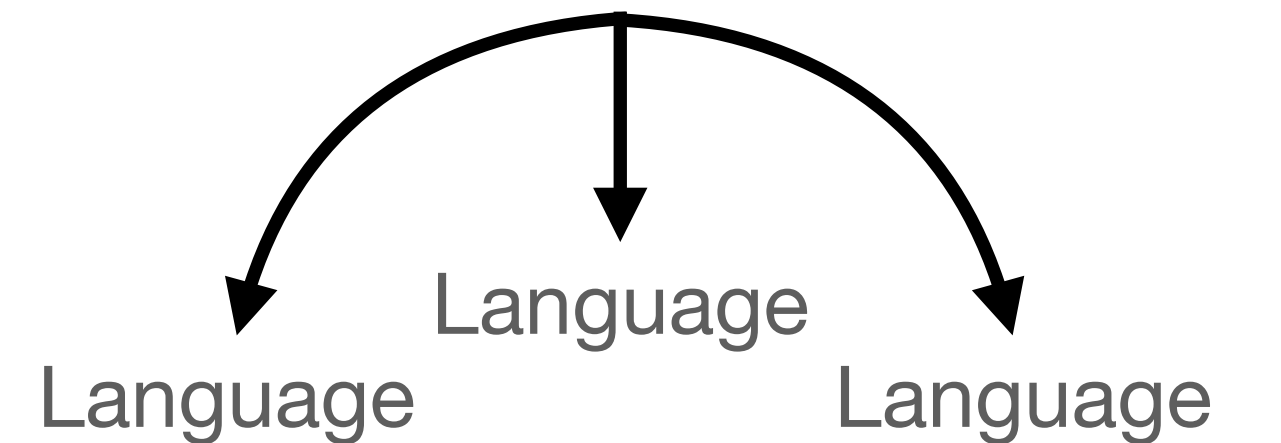
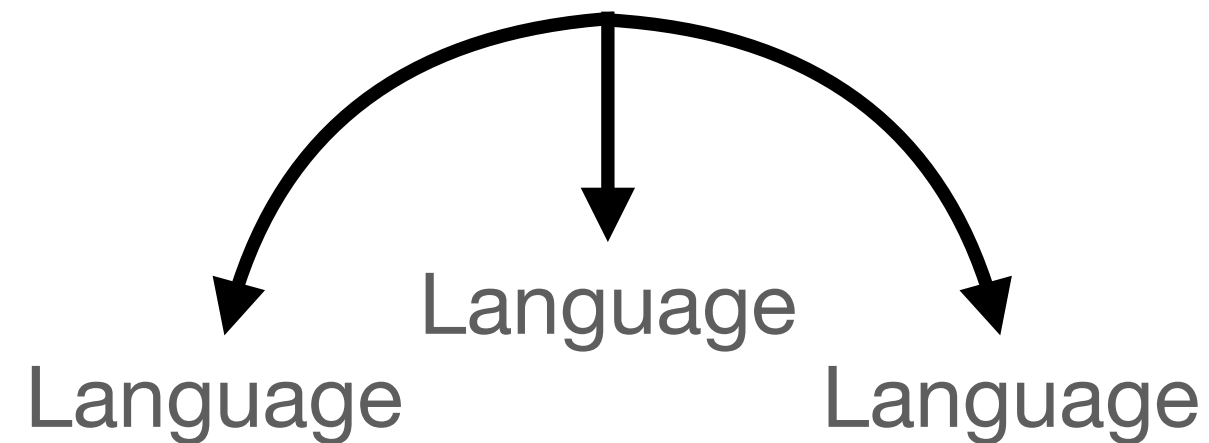
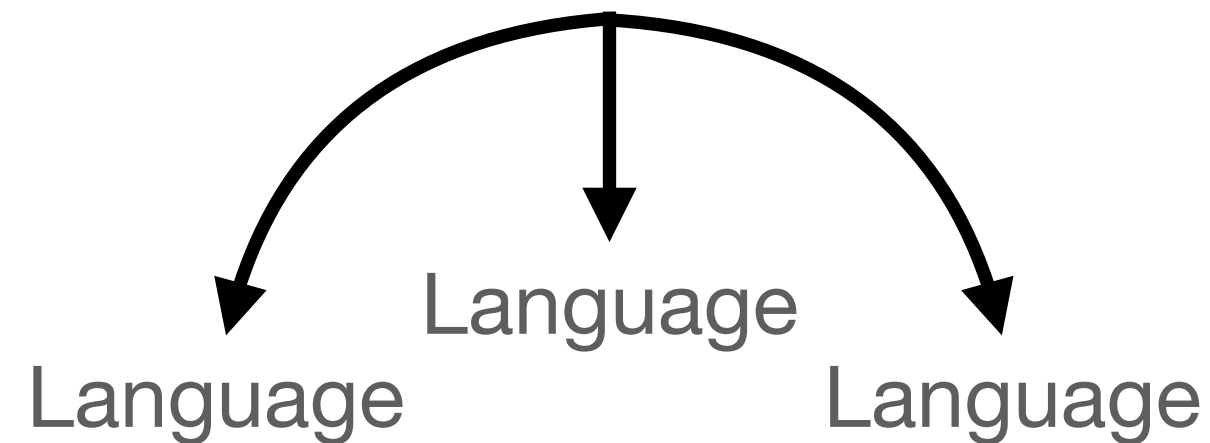
Planner  $(x,y,z) \mid \text{is\_fork}(x) \wedge \text{is\_cup}(y) \wedge \text{on}(x, y) \wedge \text{is\_counter}(z) \wedge \text{on}(y, z)$

Sample

Execute



Annotate



# Example Language

Goal: "Put a clean bowl of water on the kitchen island"



*Instructions:*

"**Turn right and begin walking across the room, then hang a left and walk over to the far side of the kitchen island.** Pick up the dirty bowl that is closest to the bottle of wine on the kitchen island. Turn left and take a step forward, then turn left and walk up to the sink. Put the dirty bowl in the sink and turn on the water, after a couple seconds turn the water off and remove the now clean bowl filled with water. Turn around and take a step forward so you are facing the kitchen island. Put the clean bowl of water on the island on the left corner."

# Action Space

*Wash the cup*

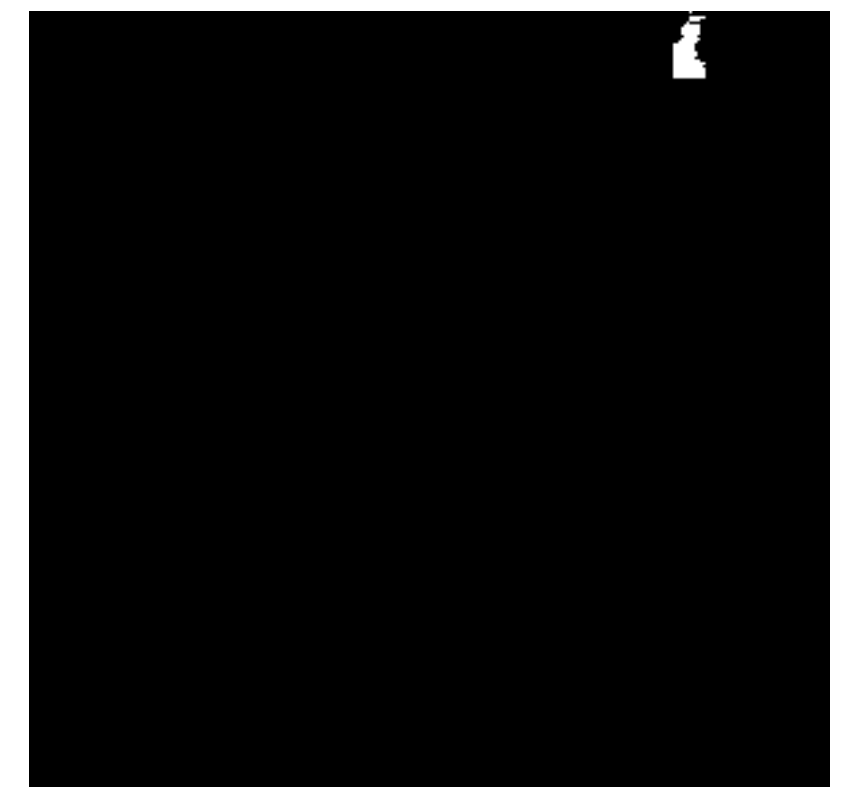
- Masks for object interaction
- Discrete actions (no torques)



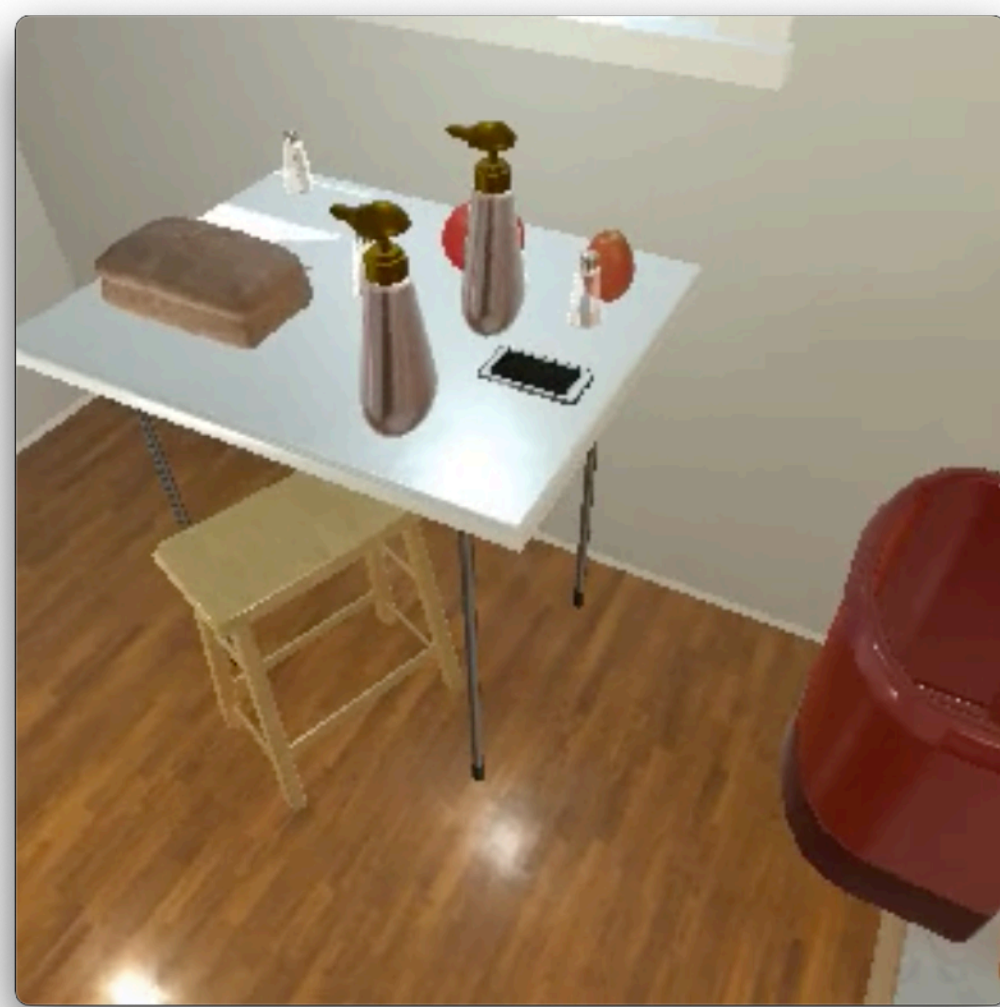
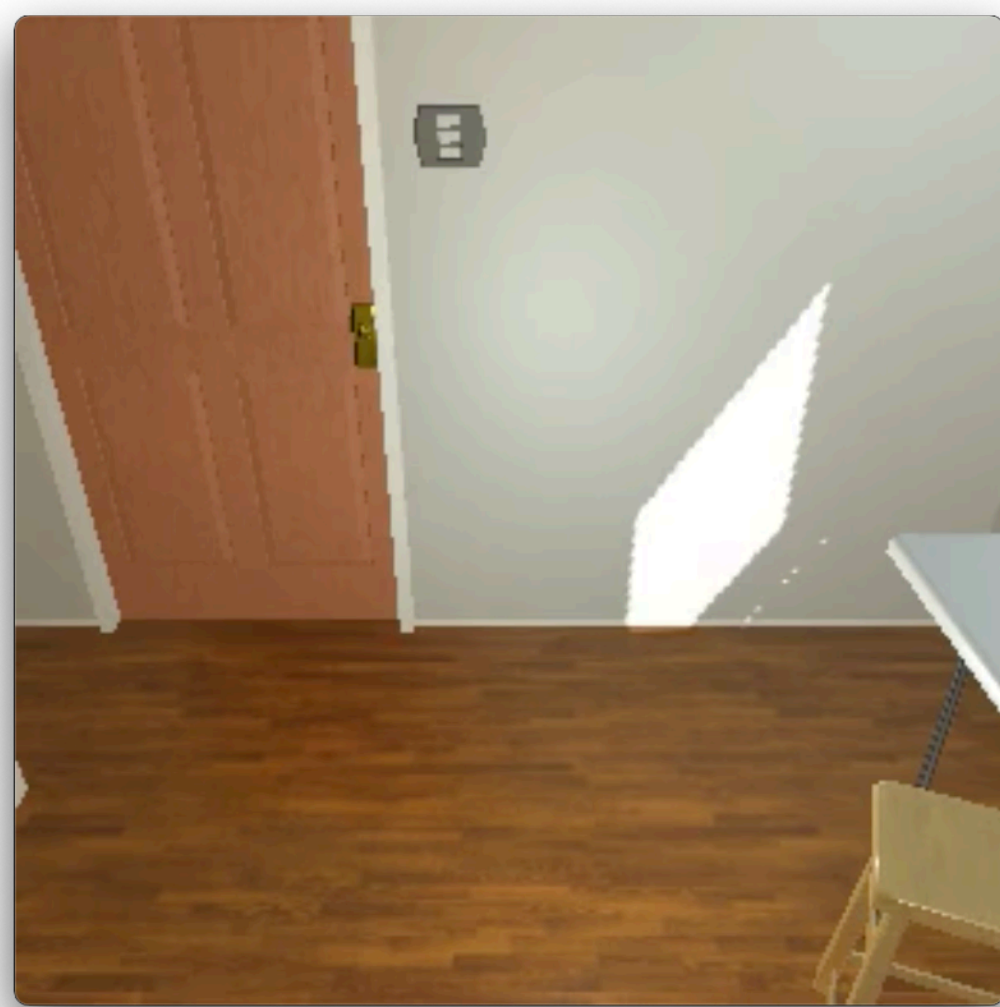
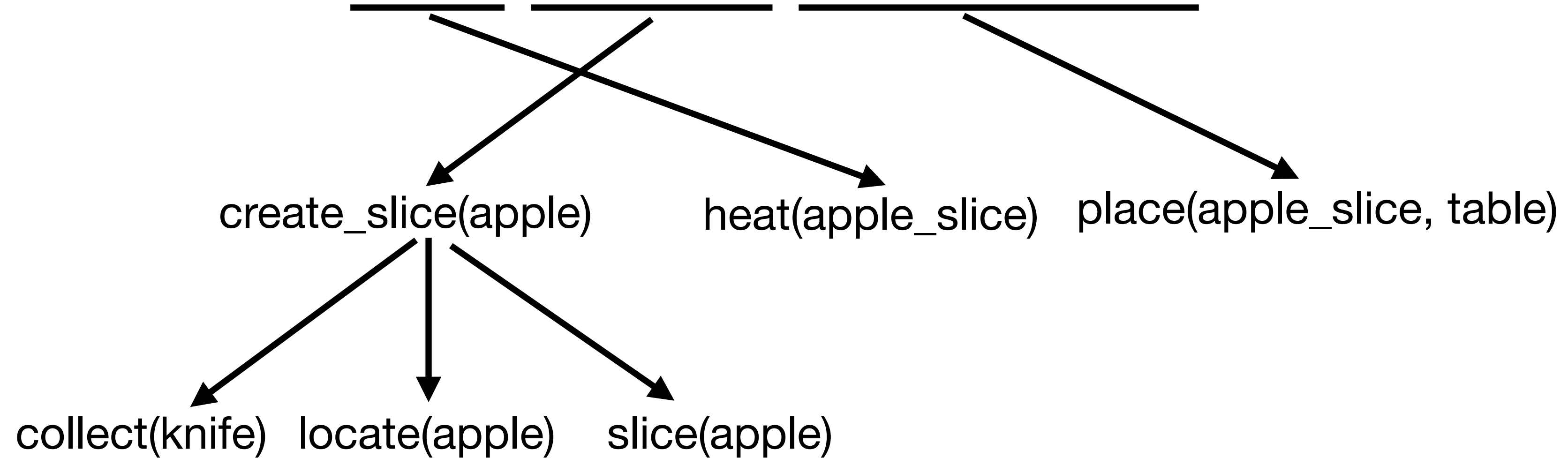
**Put In**



**Toggle**



"Place a heated apple slice on the large table"

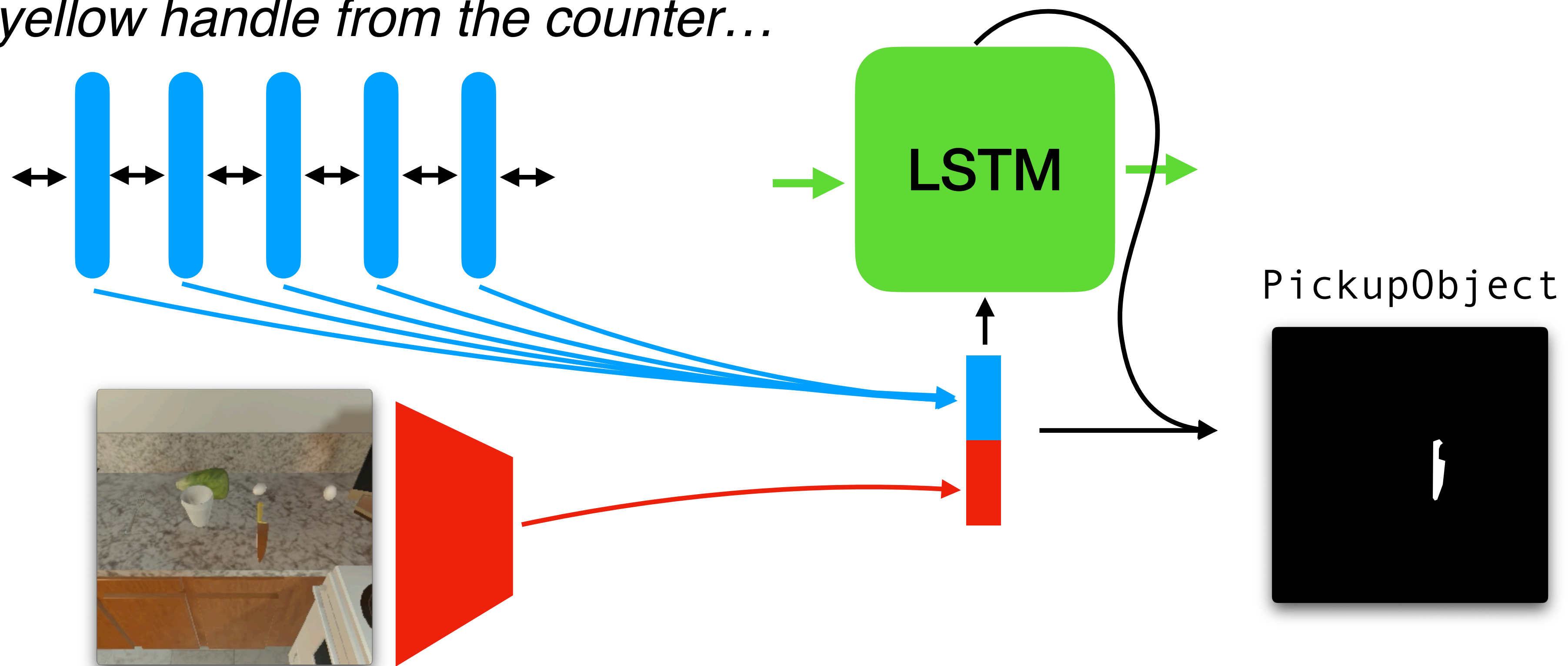


...



# End-to-End Models

*Turn around and move to the stove, then turn left to face the counter to the left of the stove. Pick up the sharp knife with the yellow handle from the counter...*



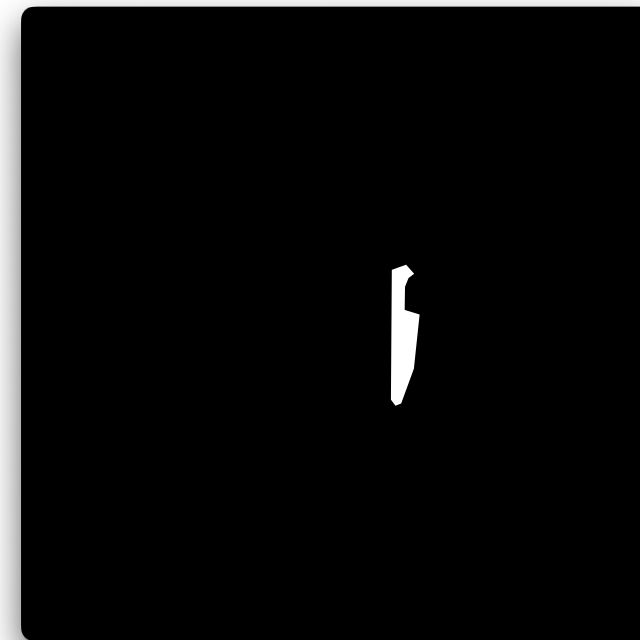
# Action Spaces

Choose a view

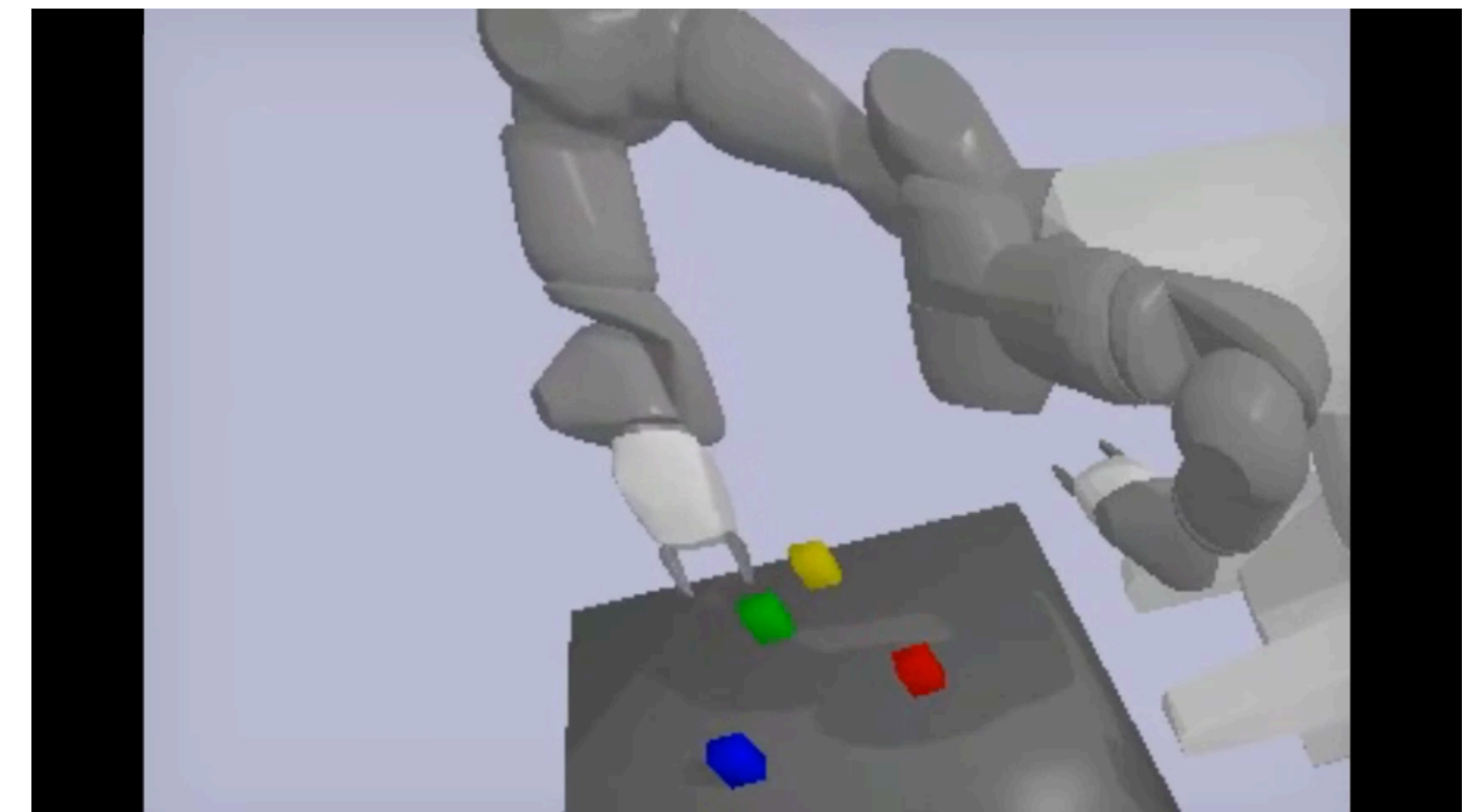


Outline an Object

PickupObject



Grasp an Object

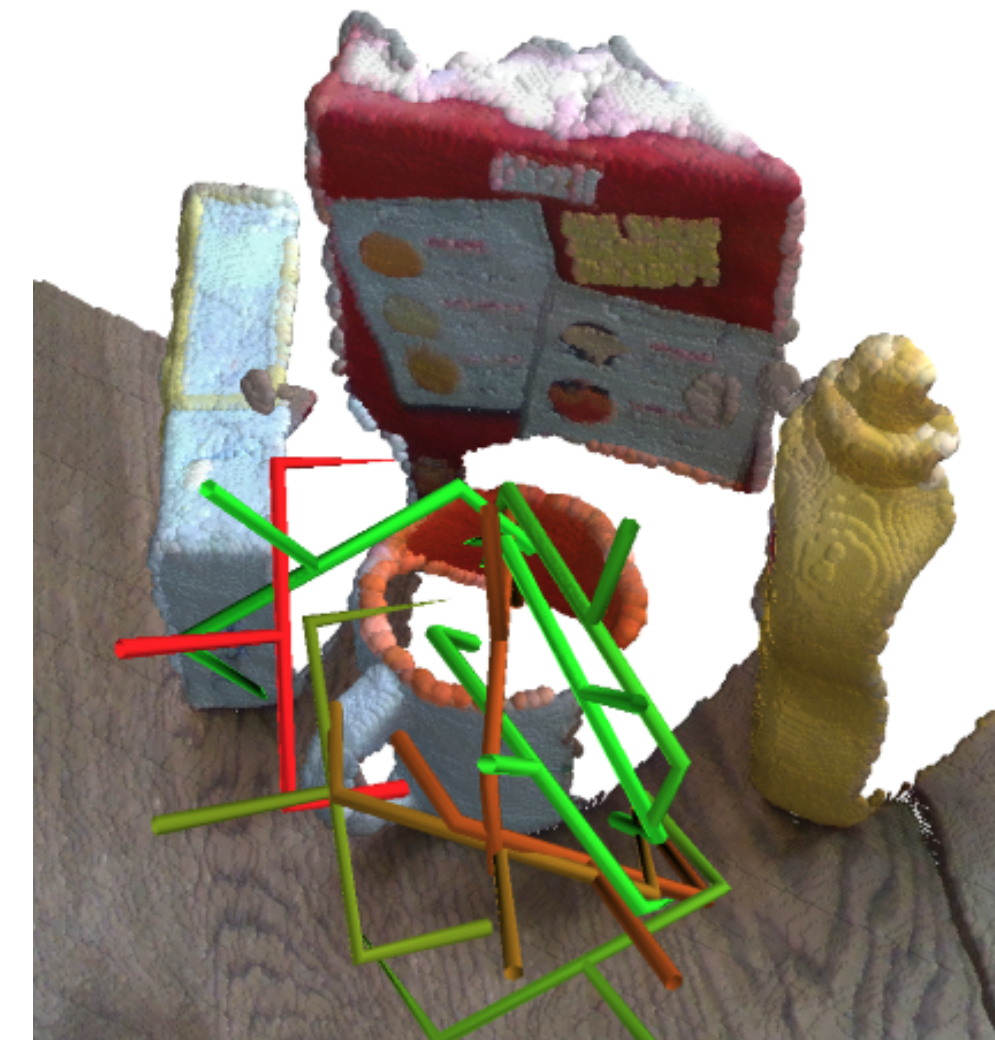


# Pick-up

What's hidden in that?

If I gave you one of these and labeled it, could you abstract to the others?

Does “pick up” mean the same thing for all of these?

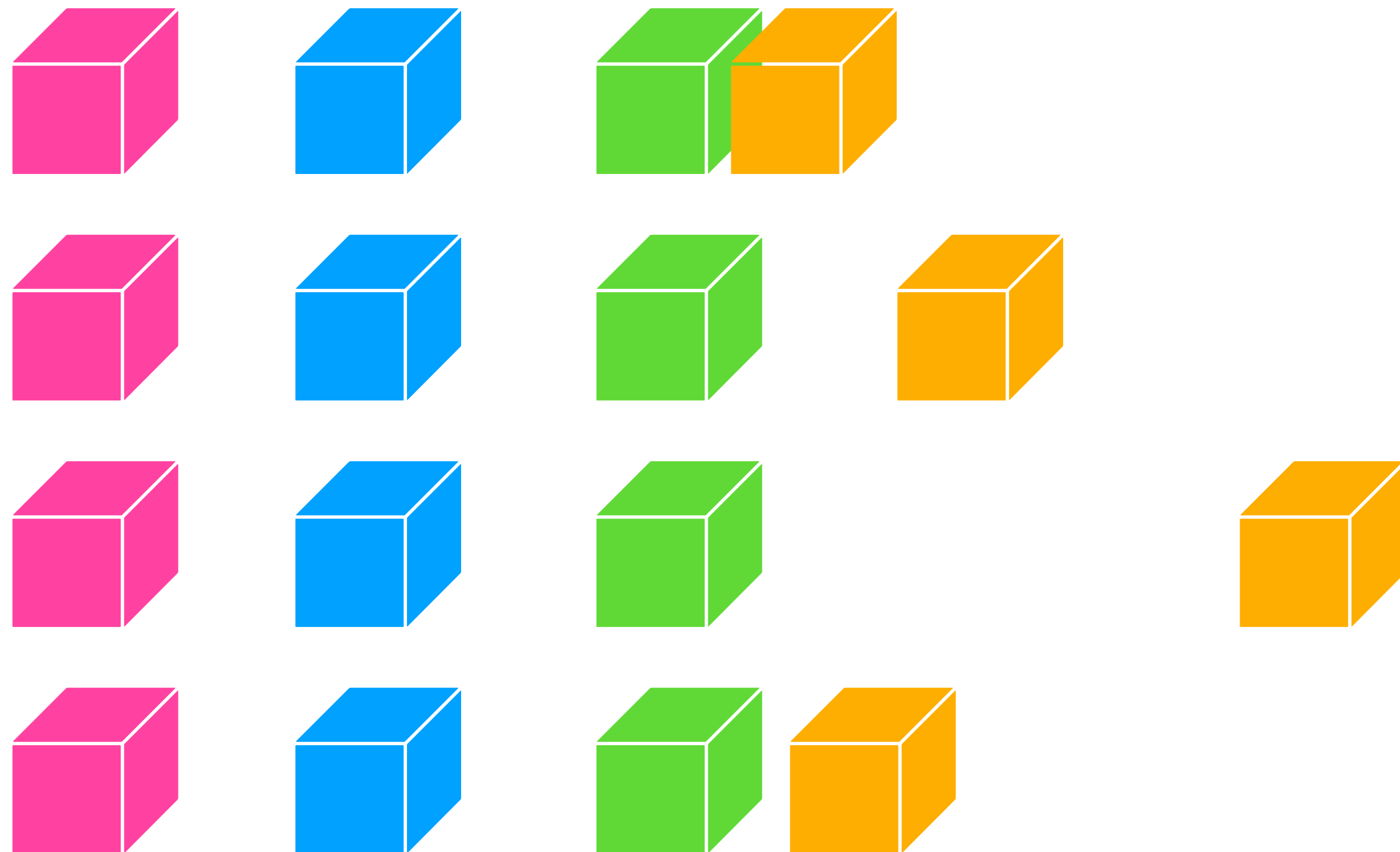


Does “pick up” correspond to a specific action sequence?

Mousavian et al. 6-DOF GraspNet: Variational Grasp Generation for Object Manipulation — ICCV 2019

# Simplify with Blocks and Coordinates

*Put the orange block to the right of the green block*

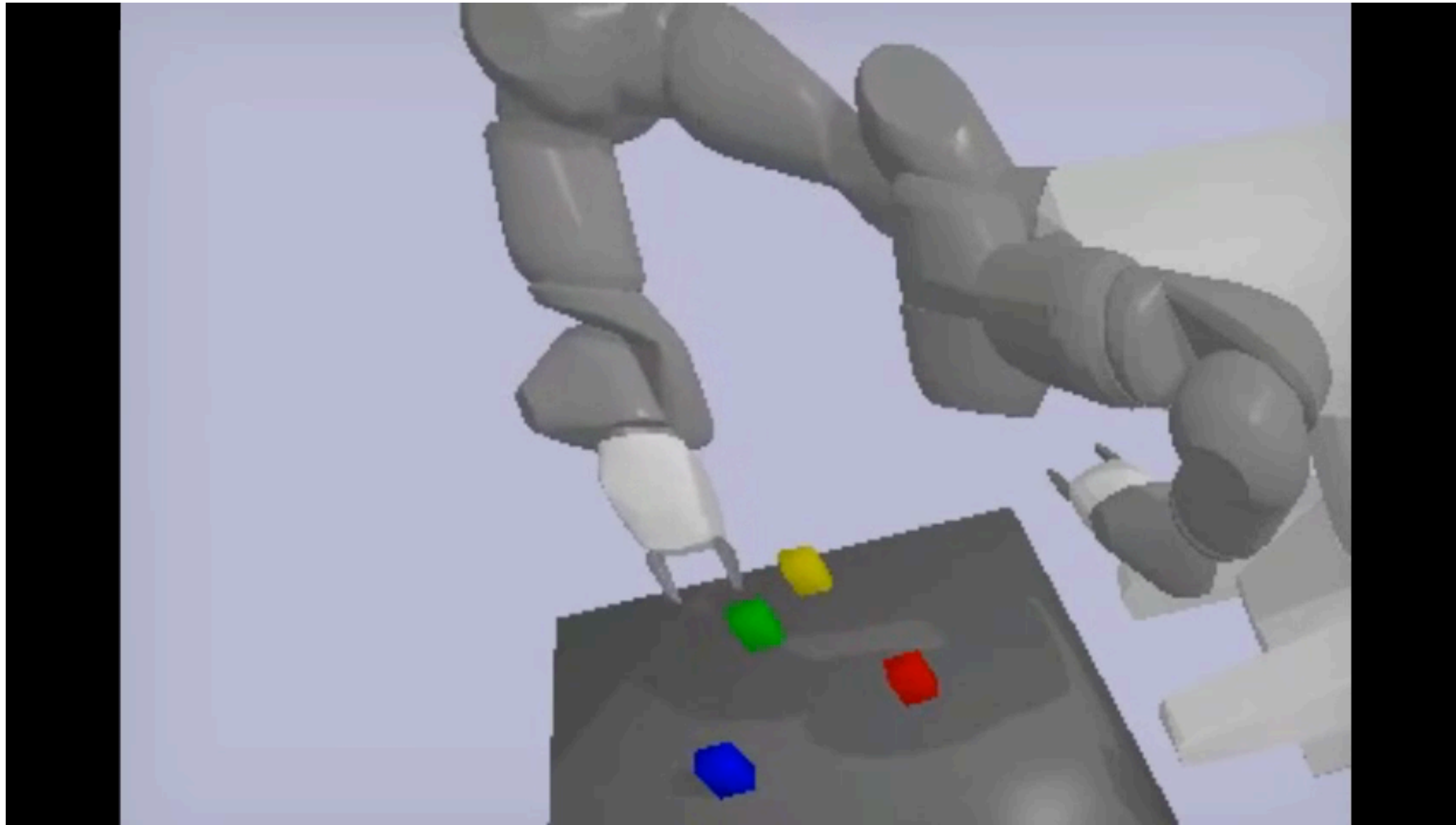


## Why?

Is this a useful training datum?  
("Put the orange block to right of the green block",  
0.35)

We no longer have a discrete grounding

# Simple Blocks



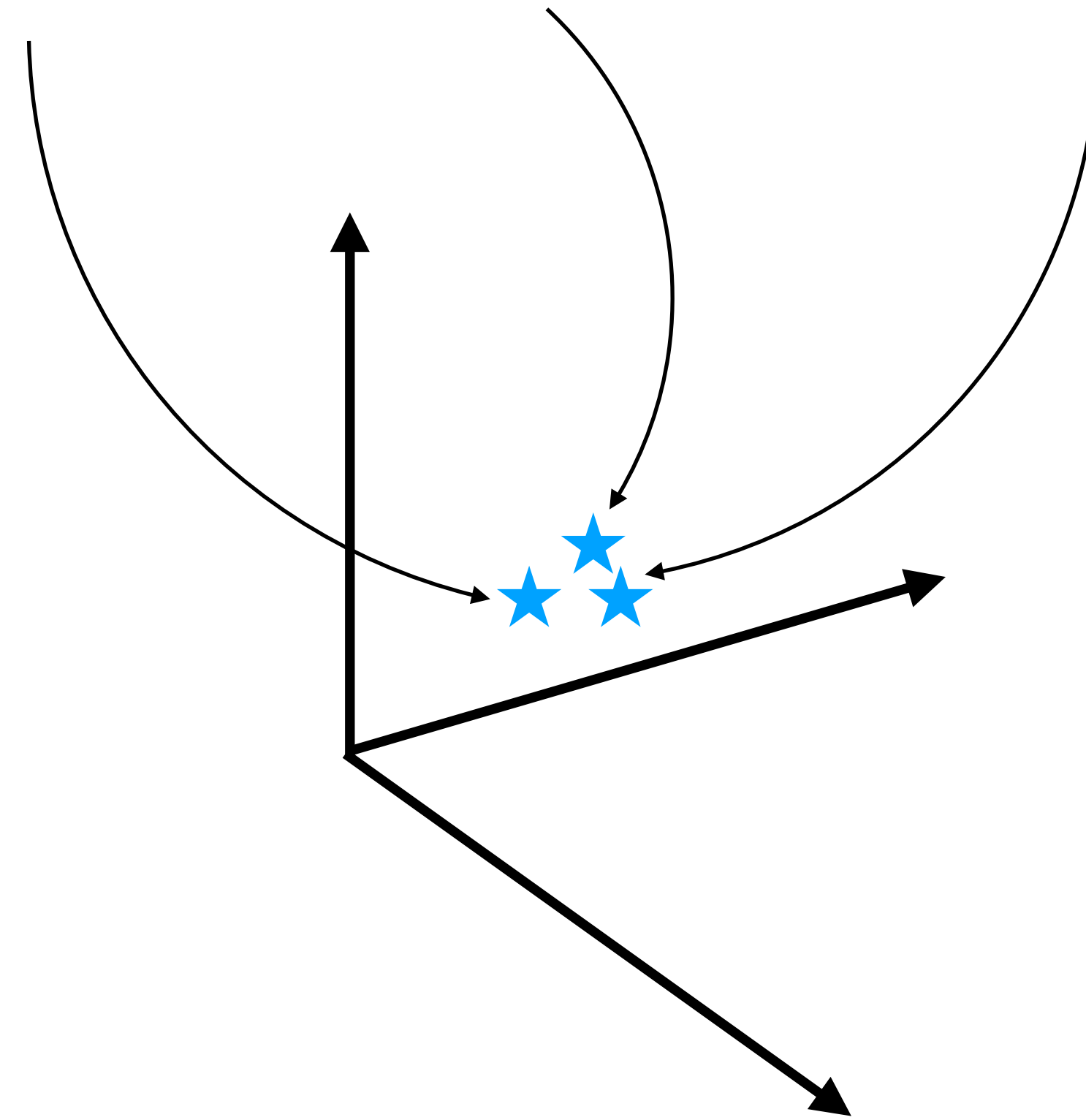
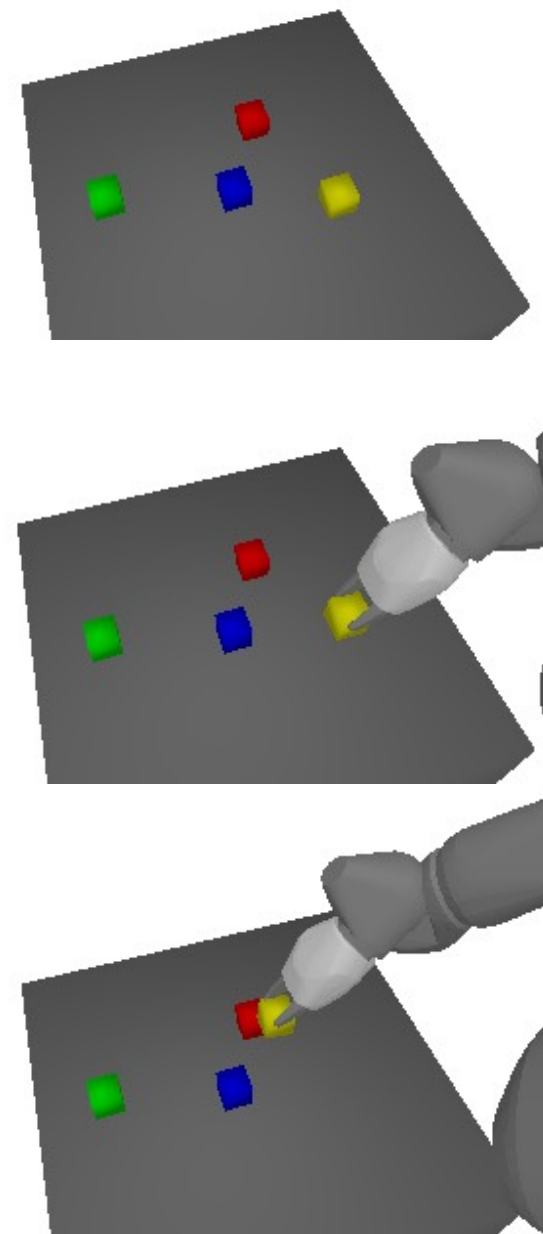
# A Shared Semantic Space

## Language

*“take the yellow object from the table and place it on top of the red object”*

`move_to(yellow) grasp(yellow) ... release(yellow)`

## Observations



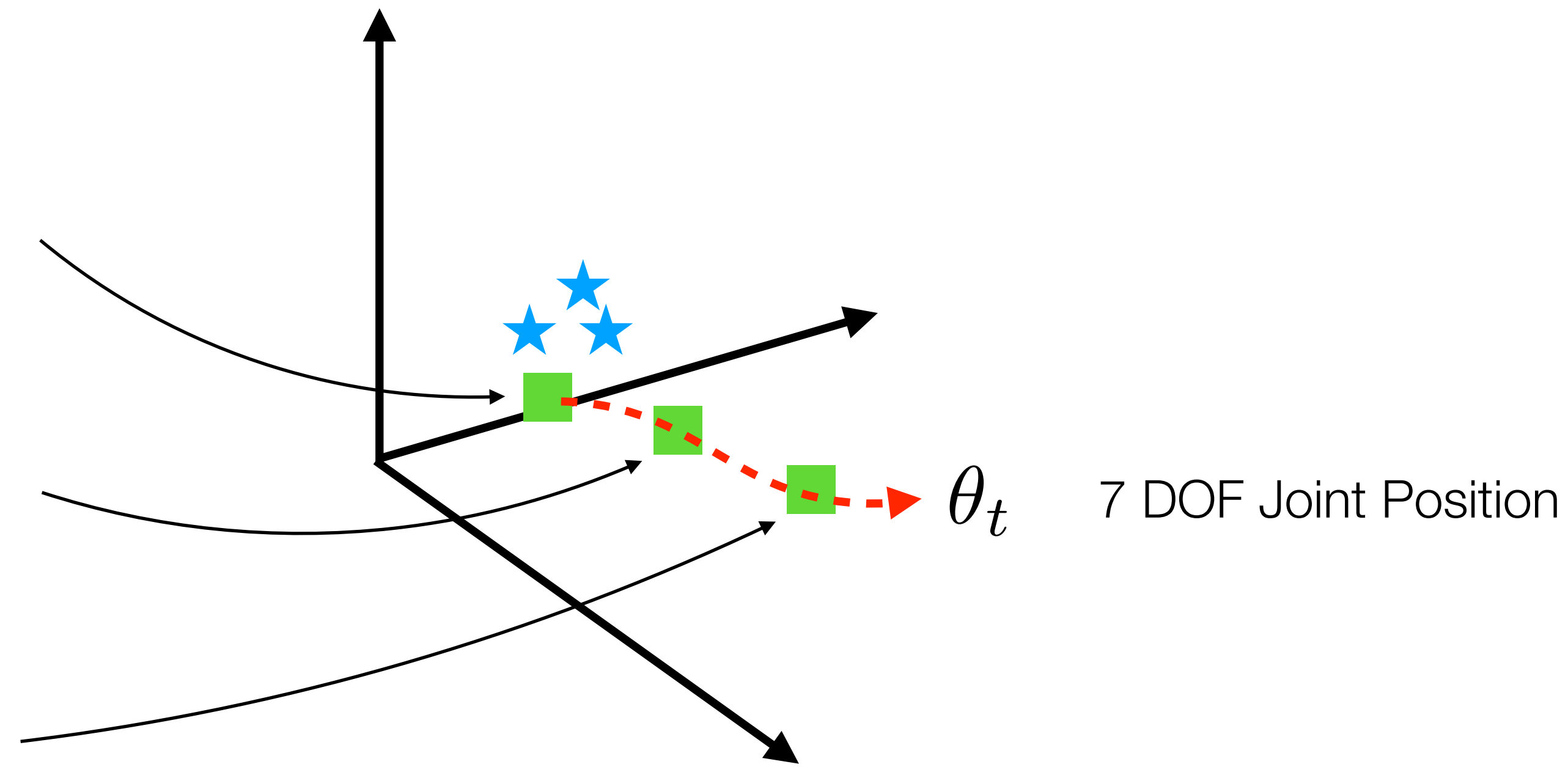
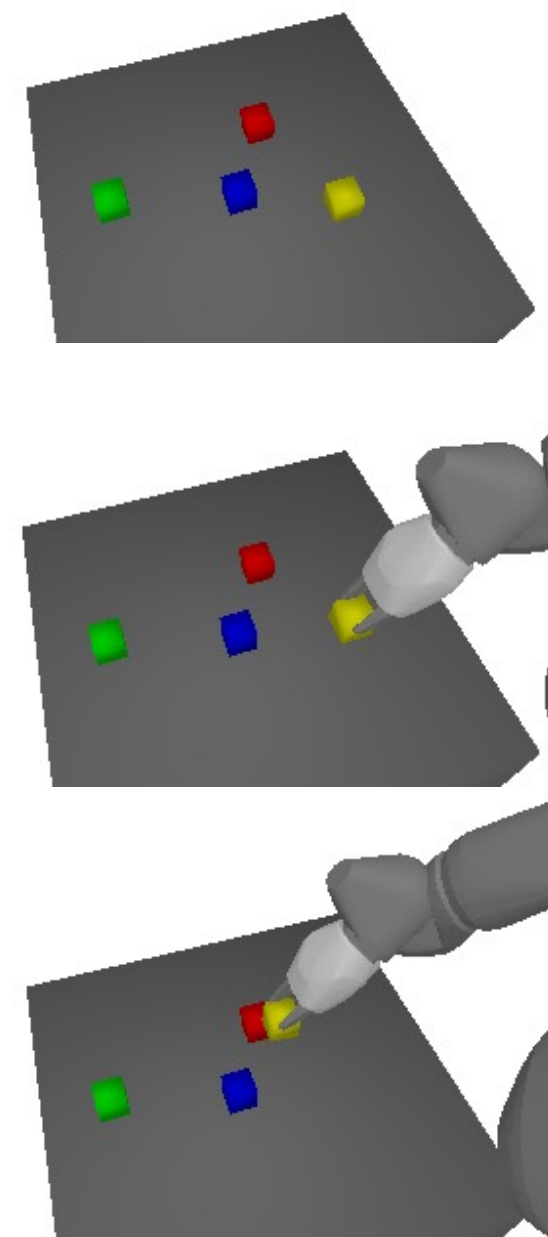
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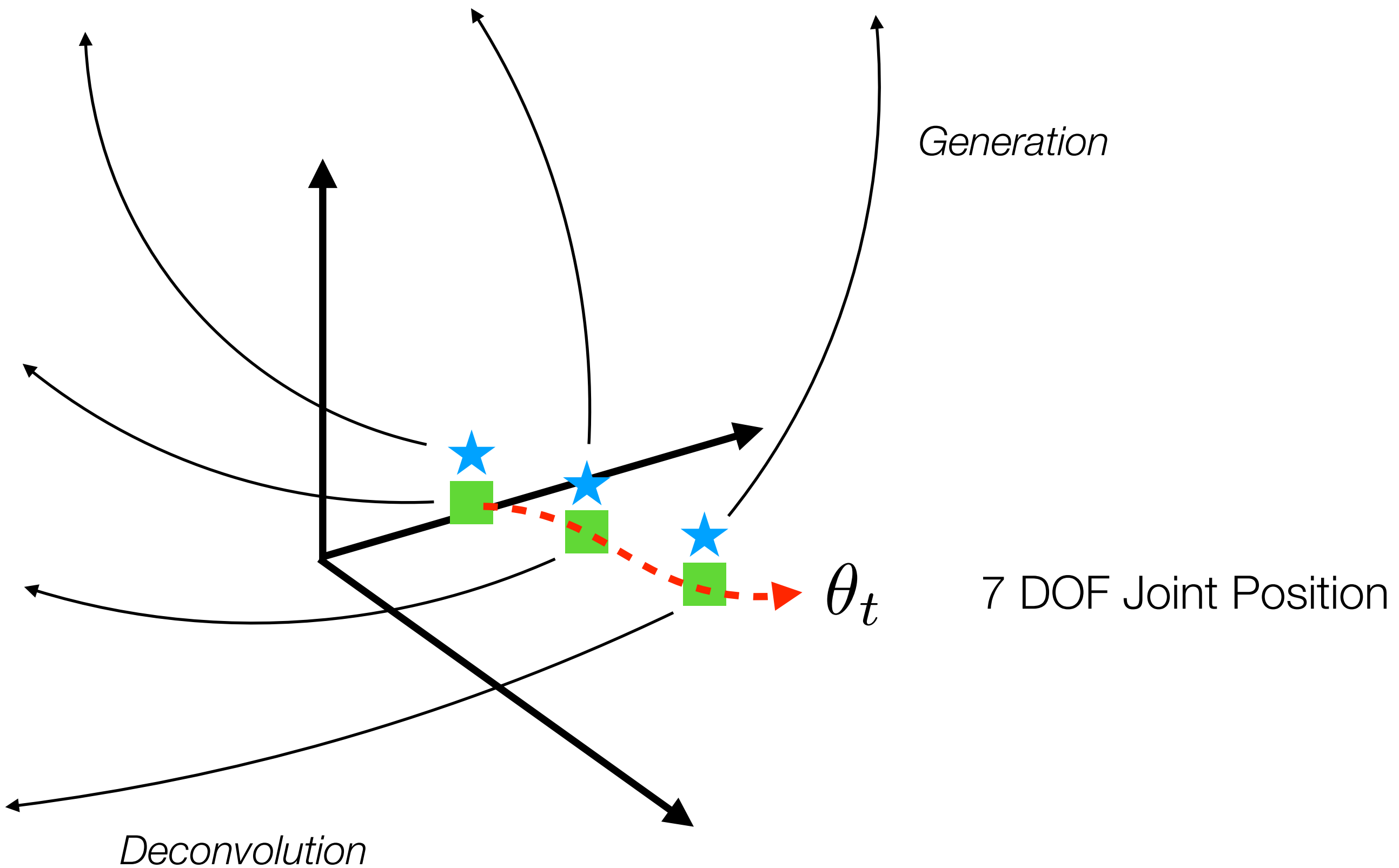
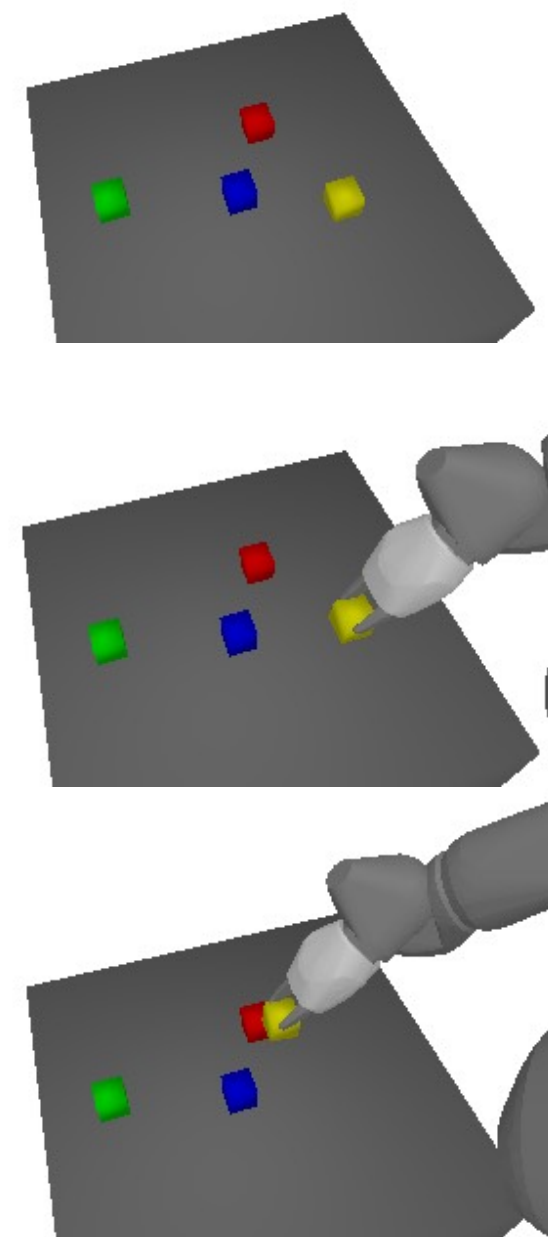
# A Shared Semantic Space

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`move_to(yellow) grasp(yellow) ... release(yellow)`

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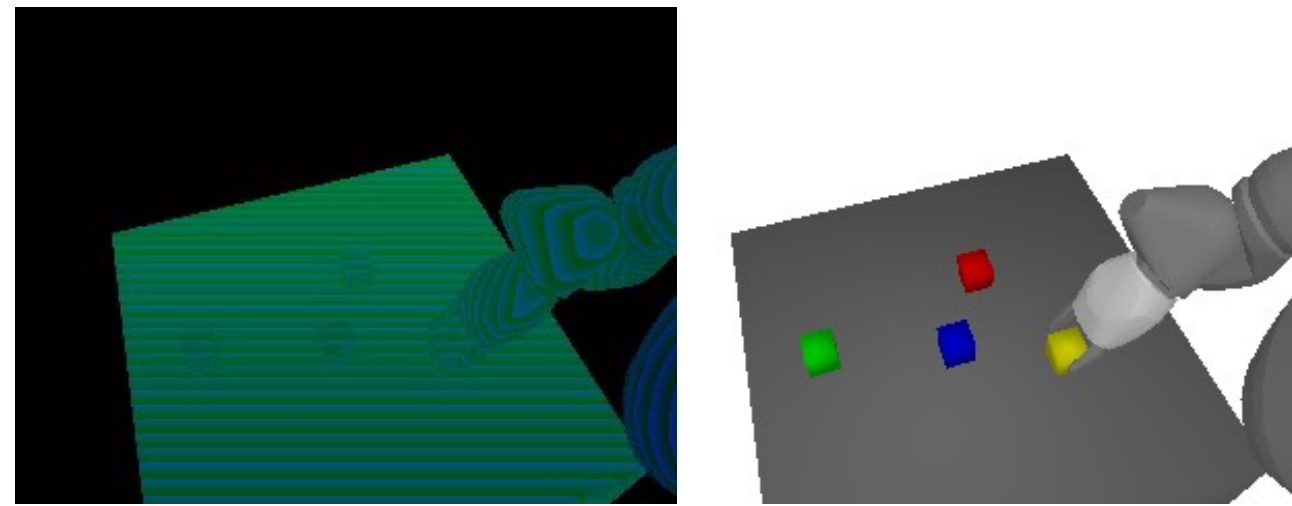


# Predicting the Future

## Goal:

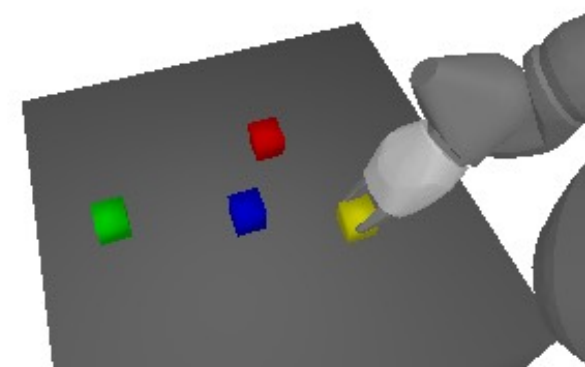
*take the yellow object from the table and place it on top of the red object*

## Current World



$h_t$

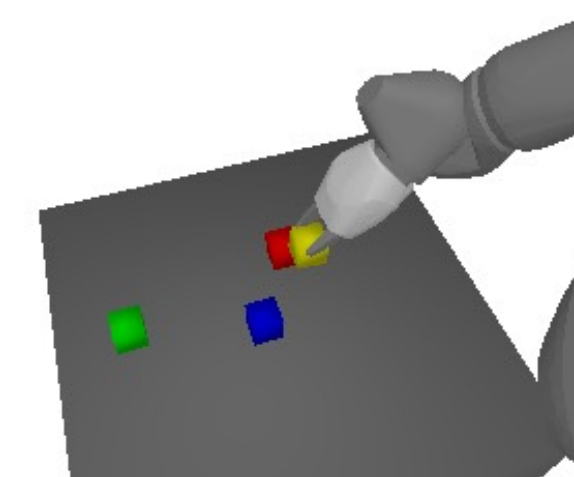
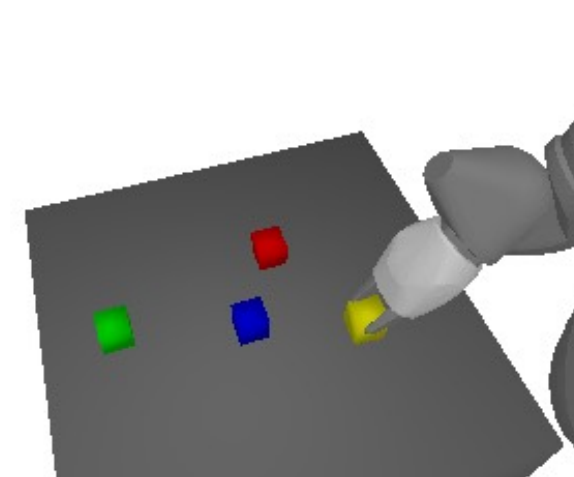
grasp (yellow)



## Interpretable Possible Futures

lift (yellow)

move (yellow, red)



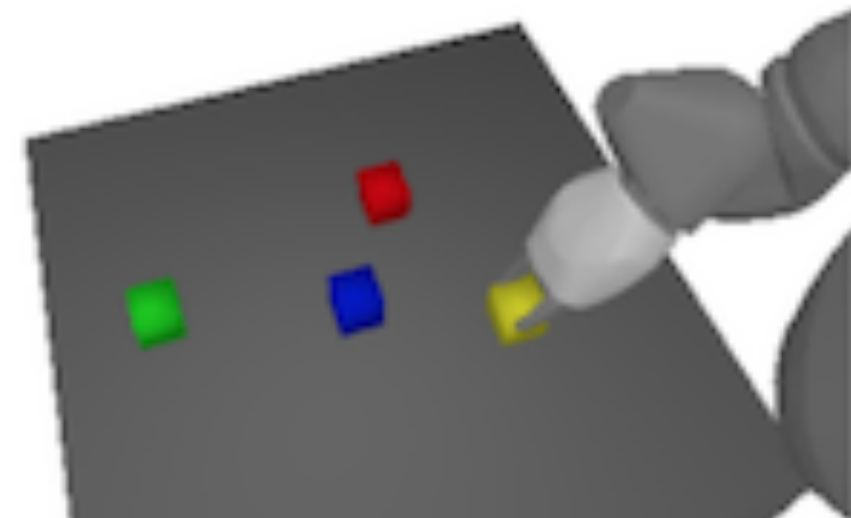
.....

# Objectives

Latent Space  $Z_t$

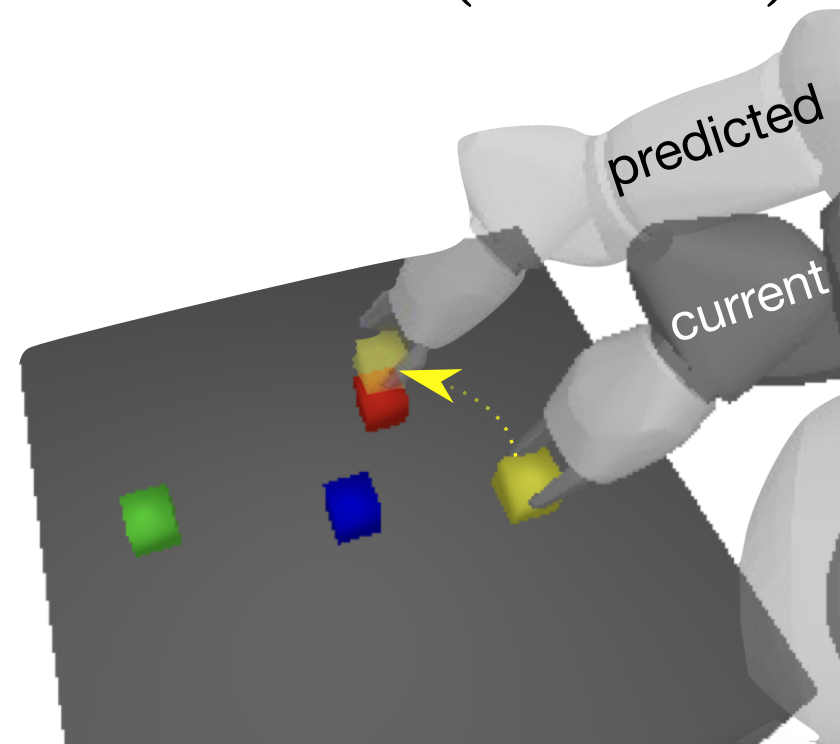
Reconstruction

$$\|\hat{W}_t - W_t\|_2^2$$



Pose

$$C_{actor}(\hat{\theta}_t, \theta_t)$$



SubGoal

$$C_G(\hat{G}_t, G_t)$$

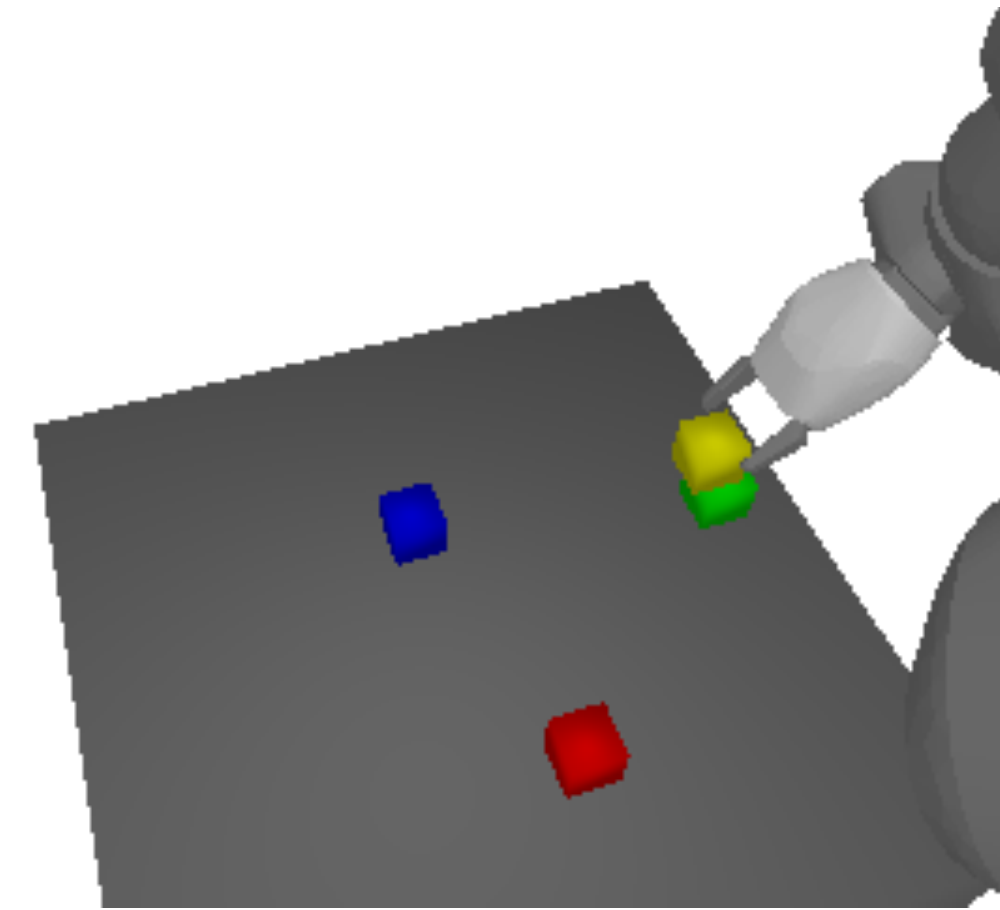
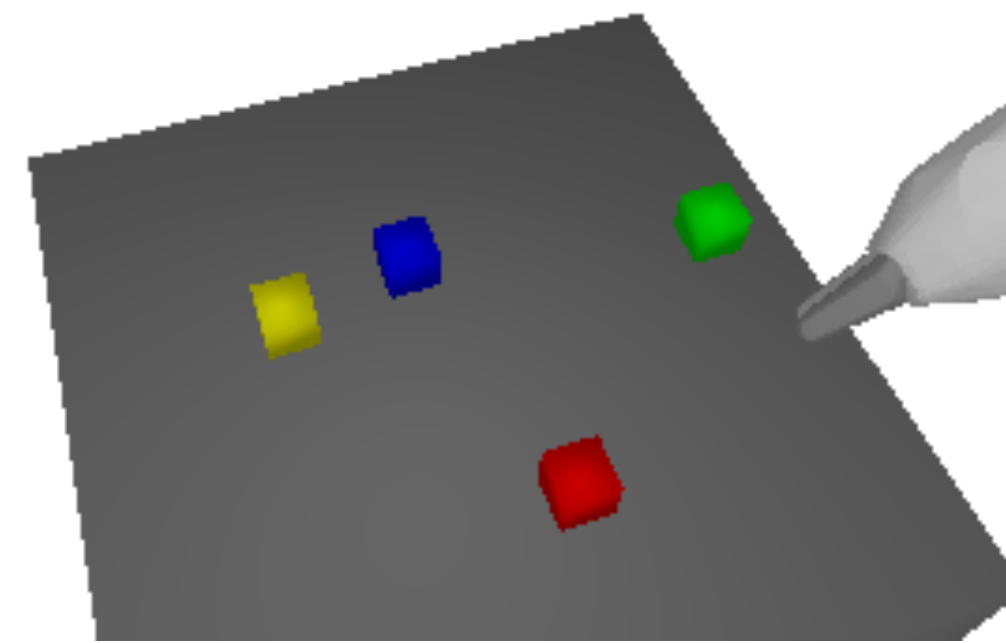
*move (yellow,  
red)*

Block pos

$$C_{obj}(z_t)$$

x #steps in horizon

# Long Tails



## Templates:

put the yellow one on the green block

## Humans:

move the yellow cube to the right until it is on top of the green cube with the front half of the yellow cube touching the far half of the top of the green cube

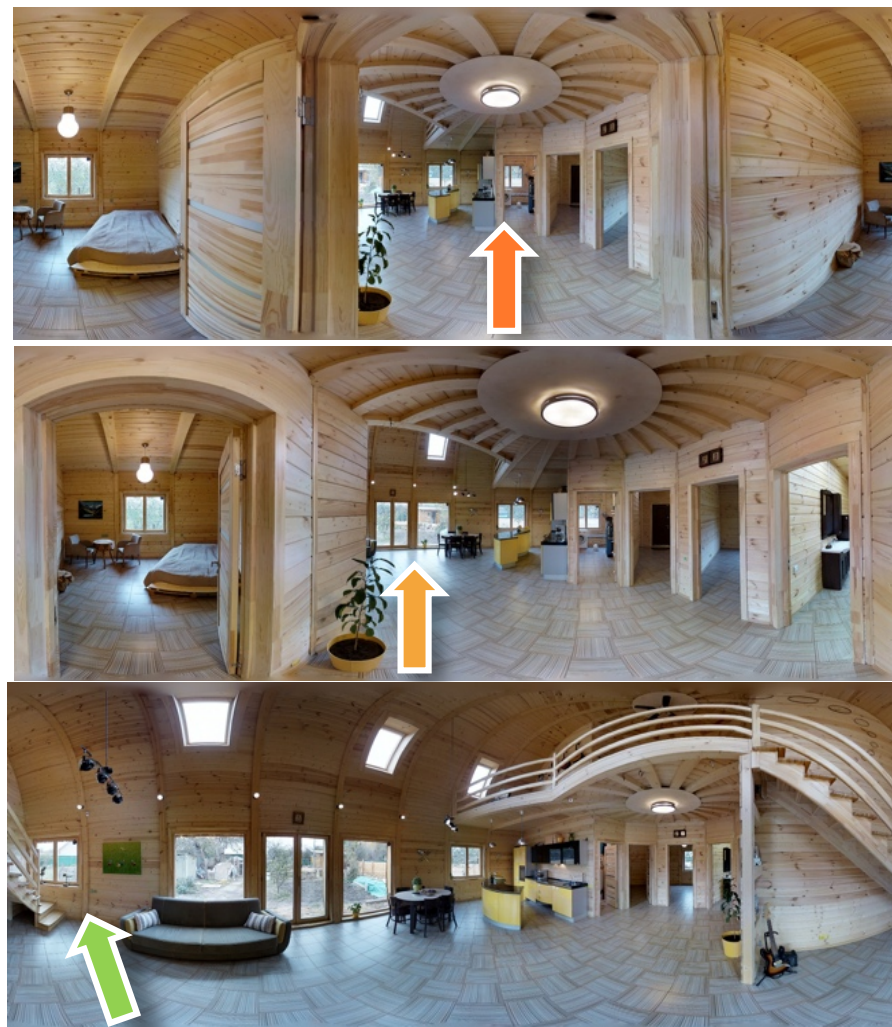
# Where does semantics come from?

Someone labeled it?

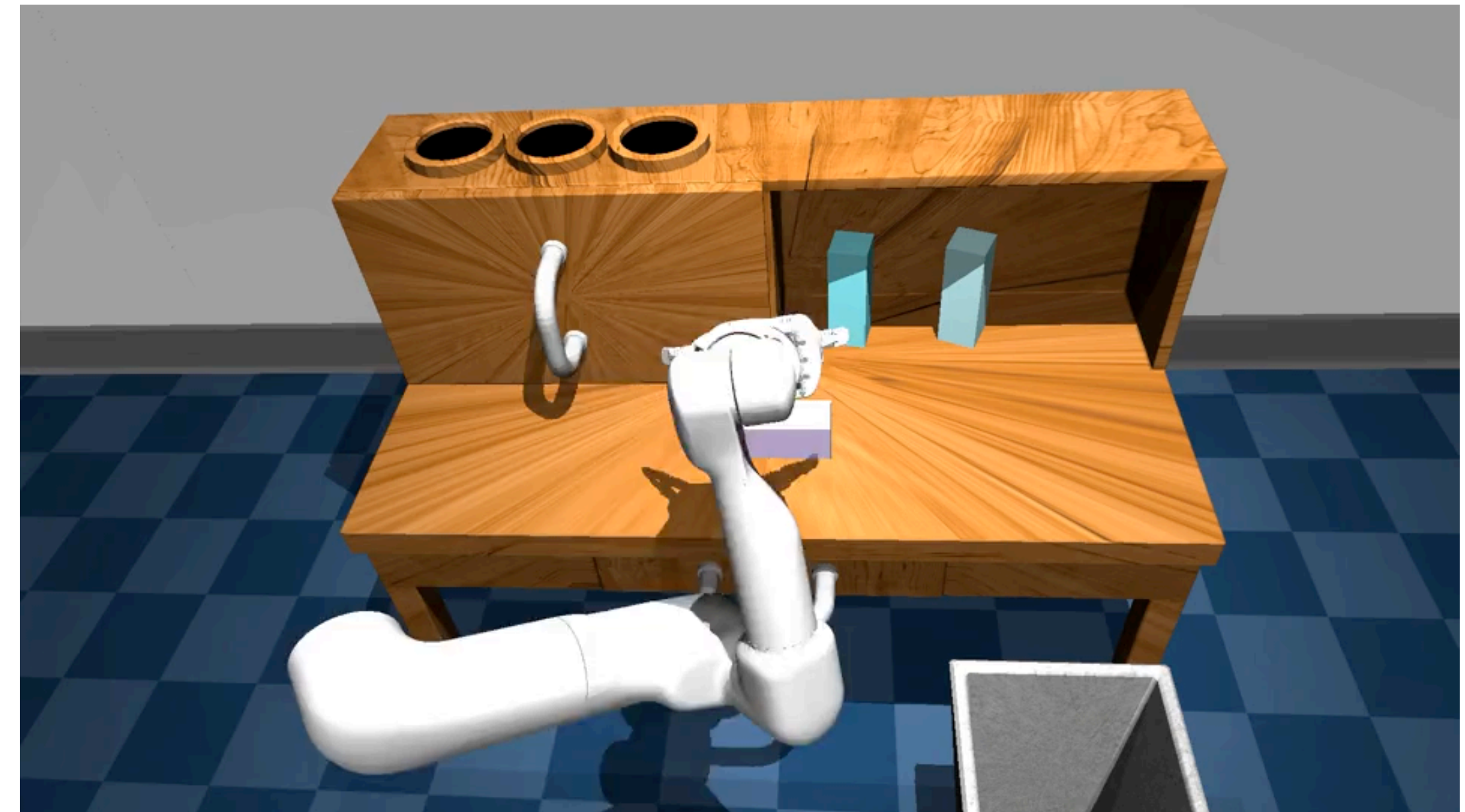
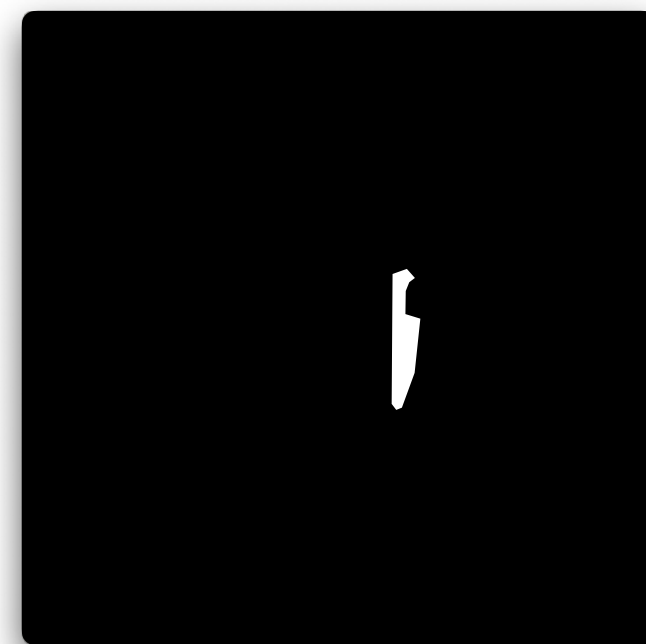
$$p(a|v_0, \dots, v_t)$$

Self-Play and Physical Affordances?

Simulator Definitions?



PickupObject



Lynch et al. — Learning Latent Plans from Play — CoRL 2019

# Embodiment

- Choose your own adventure — Lots of noise
- What does it mean to succeed?
- Where do concepts come from?
- What's the role of exploration?
- Language is woefully underspecified

All of these are the “same” verb

