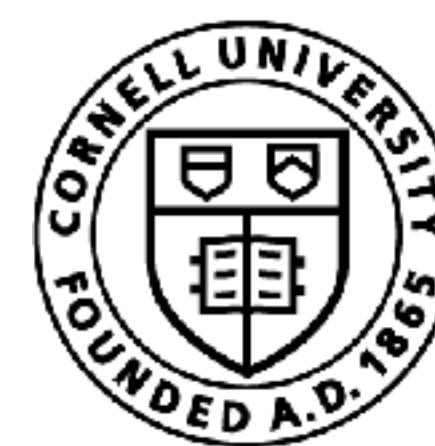


Controlling Robots via Large Language Models

Sanjiban Choudhury



Cornell Bowers CIS
Computer Science

Today is the last day.

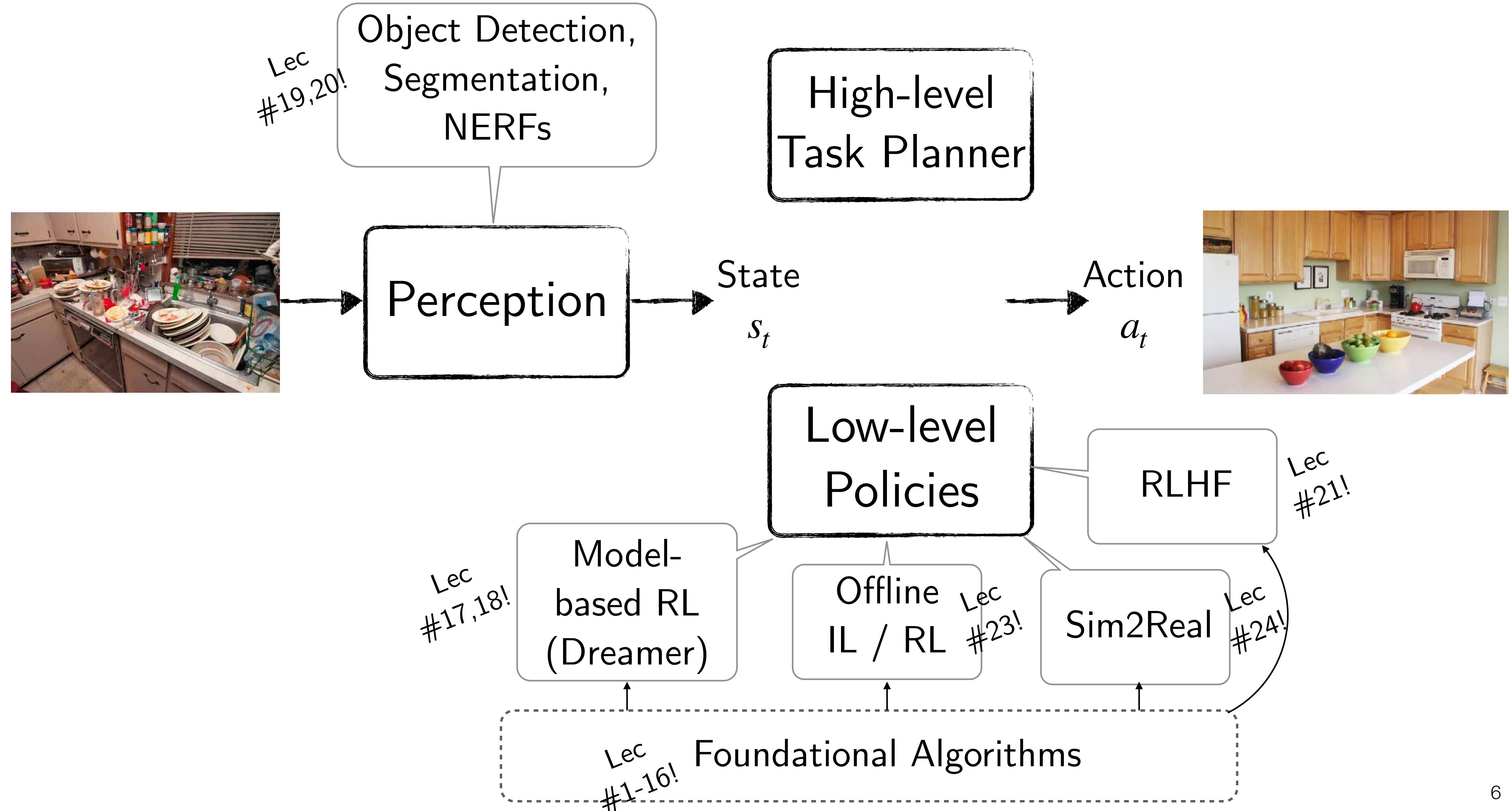
Have we learnt anything
useful at all?

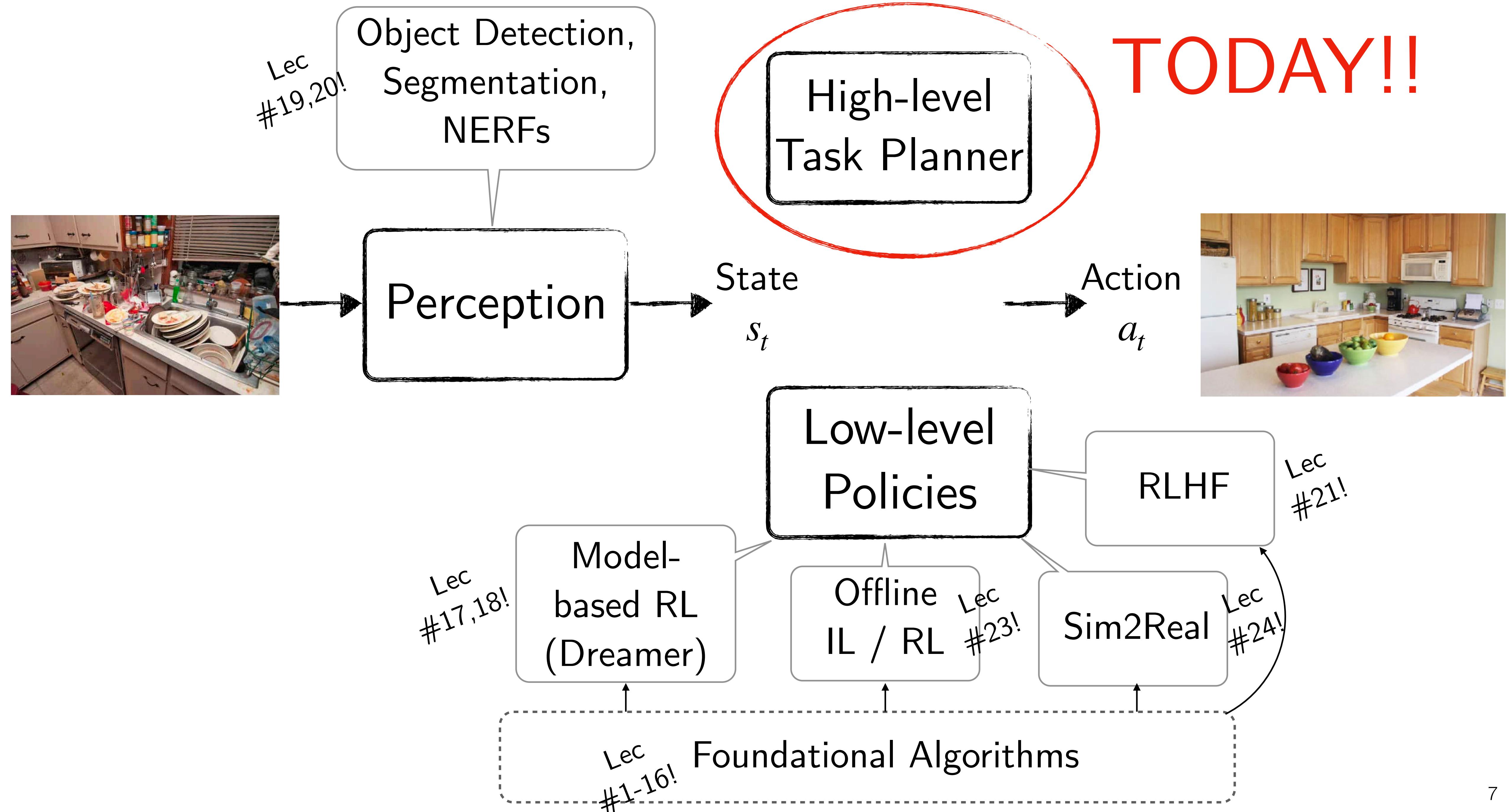
Let's go back to
Episode #1

The Problem: Real world is complex!









Poll!

When poll is active respond at PollEv.com/sc2582

Send **sc2582** to **22333**



The Problem:

Many tasks are *personal*

Cooking is personal



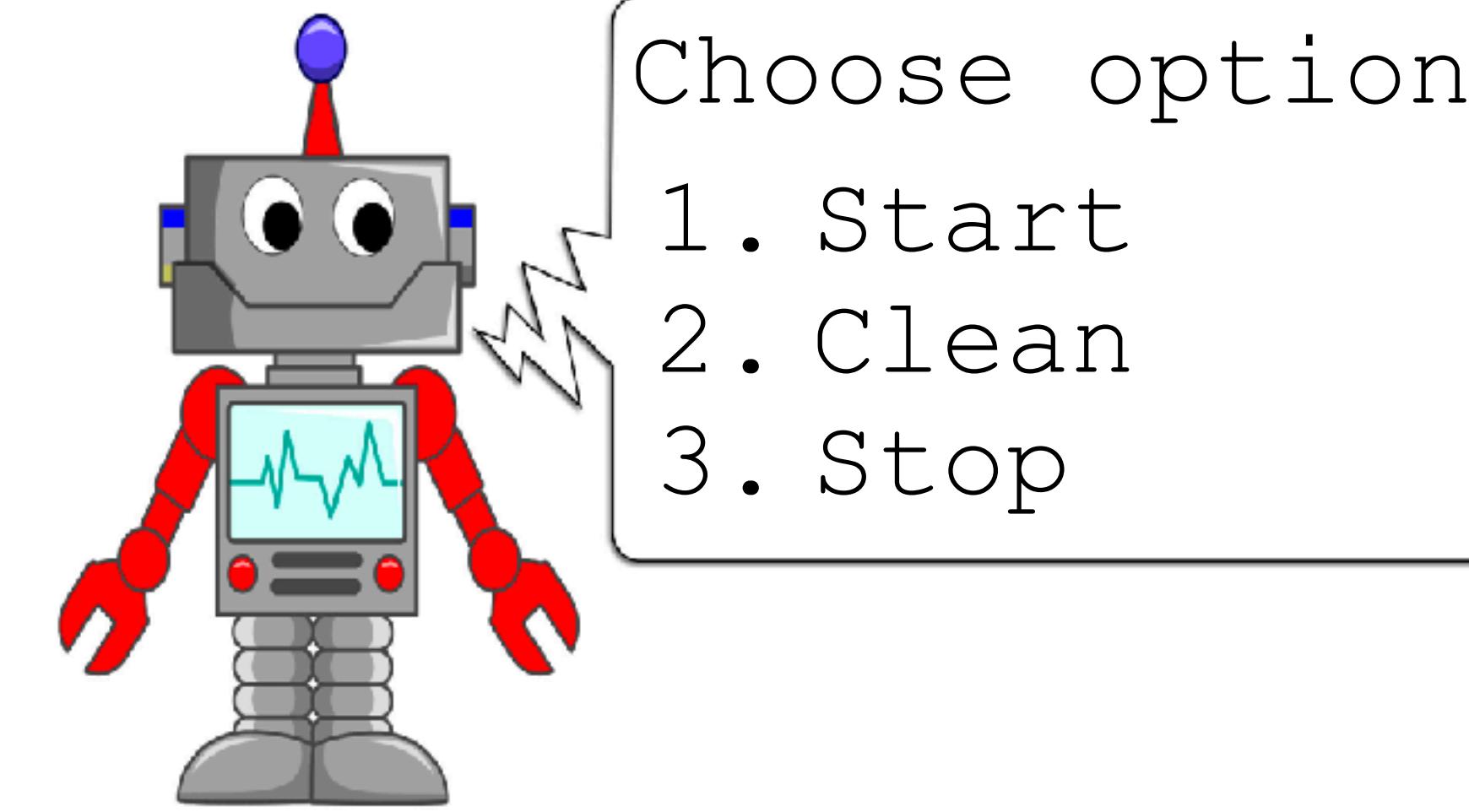
Home organization is personal



Robots today are NOT personal



Engineers program behaviors



Ship robot



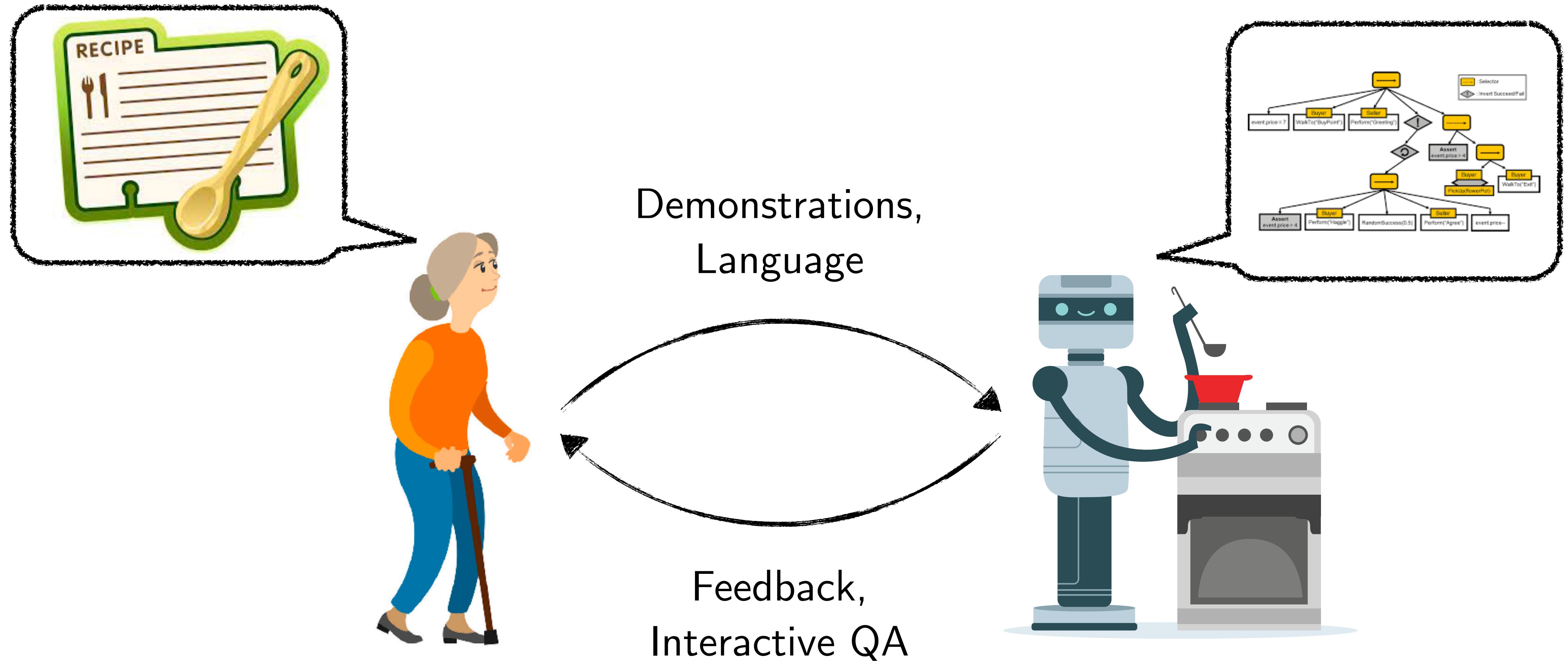
Frustrate users!

Cannot be **flexibly re-programmed** by **everyday** users

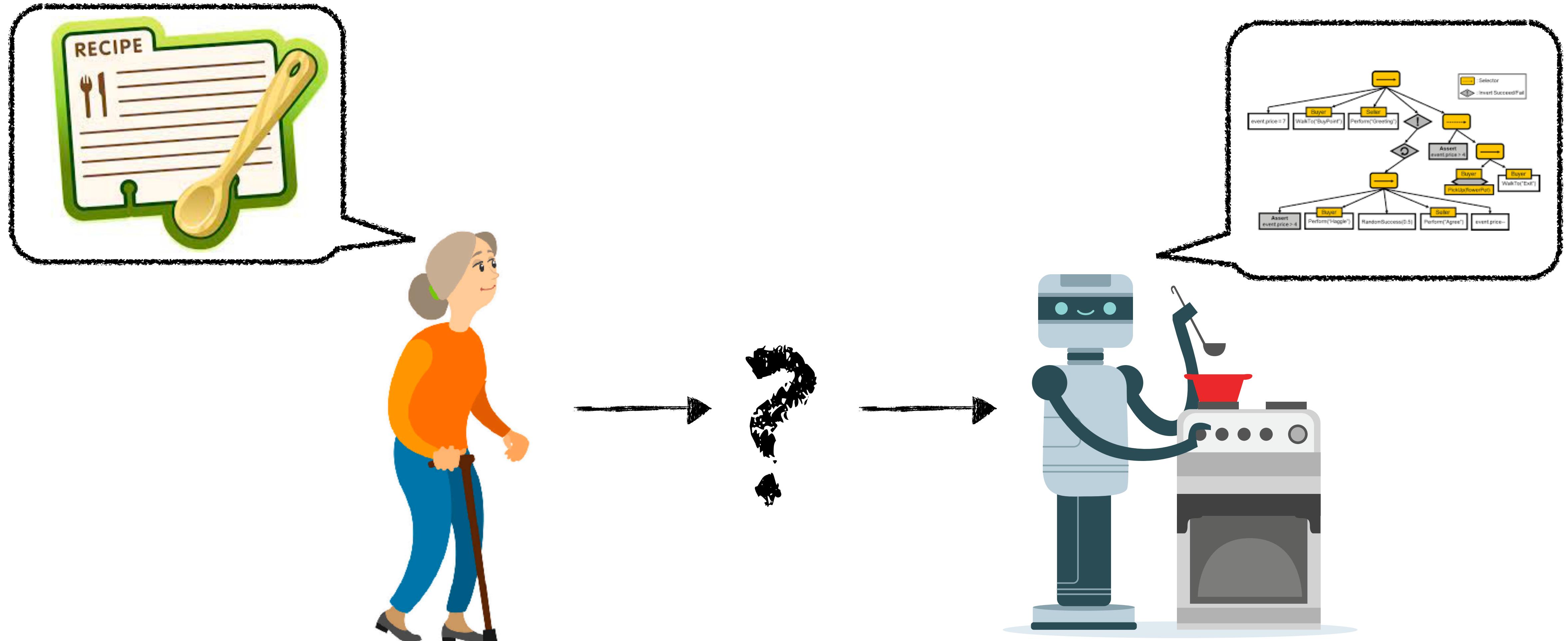


Can we **implicitly** program
robots via natural interactions?

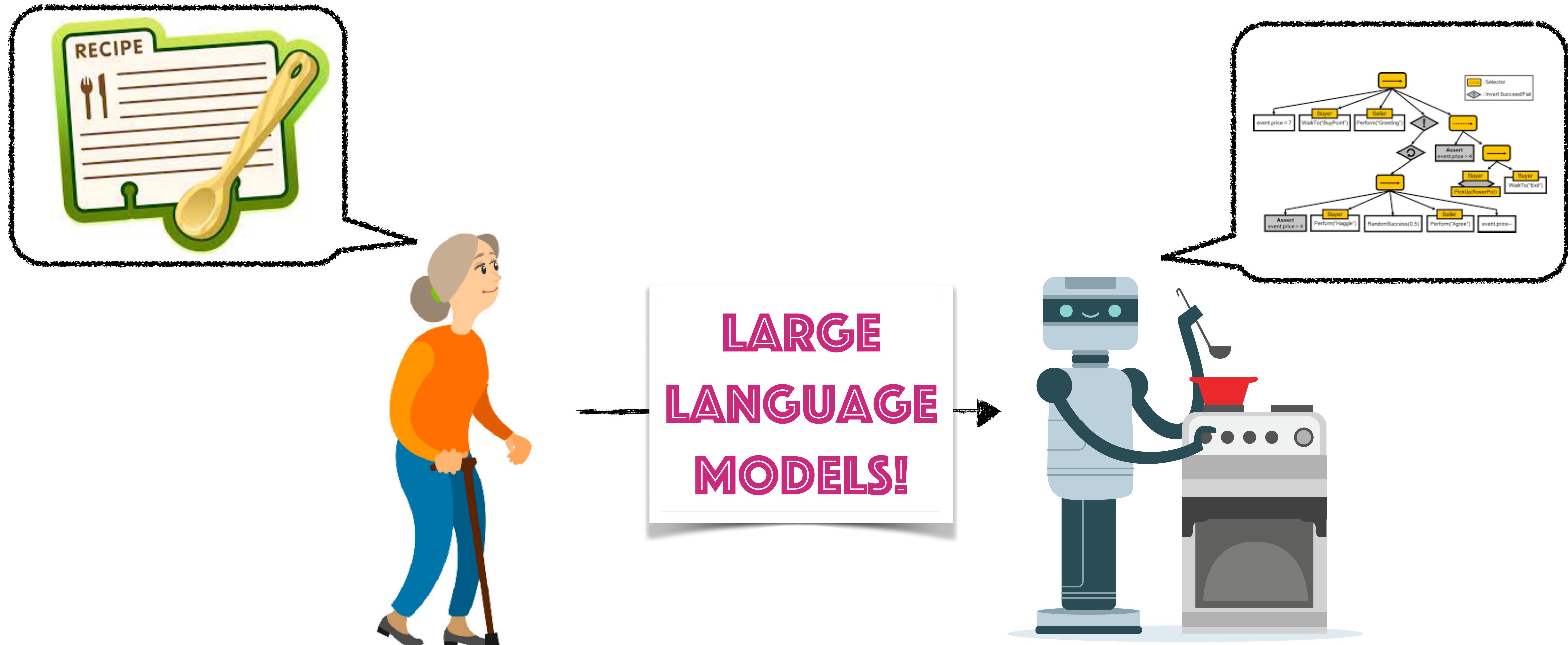
Programming via natural interactions



Question: How do we translate between humans and robots?



Large Language Models to the rescue!



An Example

MOSAIC

A Modular System for Assistive and Interactive Cooking



MOSAIC

A Modular System
for Assistive and Interactive Cooking

Hey MOSAIC! It's so cold today! Can you suggest some soup recipe?



Chicken tomato soup would be a great choice for a cold day. Let's start cooking!



No, I don't want chicken in it. How about mixed vegetable soup? I got a bunch of vegetables.



If you are looking to use a bunch of vegetables and want a delicious soup, the mixed vegetable soup from our recipe list would be perfect! Let's get started on that!



Shall Alice pour salt into the pot and then stir it for you?
Shall Bob go get the soup base for you?

Activity!



Think-Pair-Share!

Think (30 sec): What are some of the challenges to convert natural language conversation to code that a robot can execute?

Pair: Find a partner

Share (45 sec):
Partners exchange
ideas





How things
worked
pre-LLM

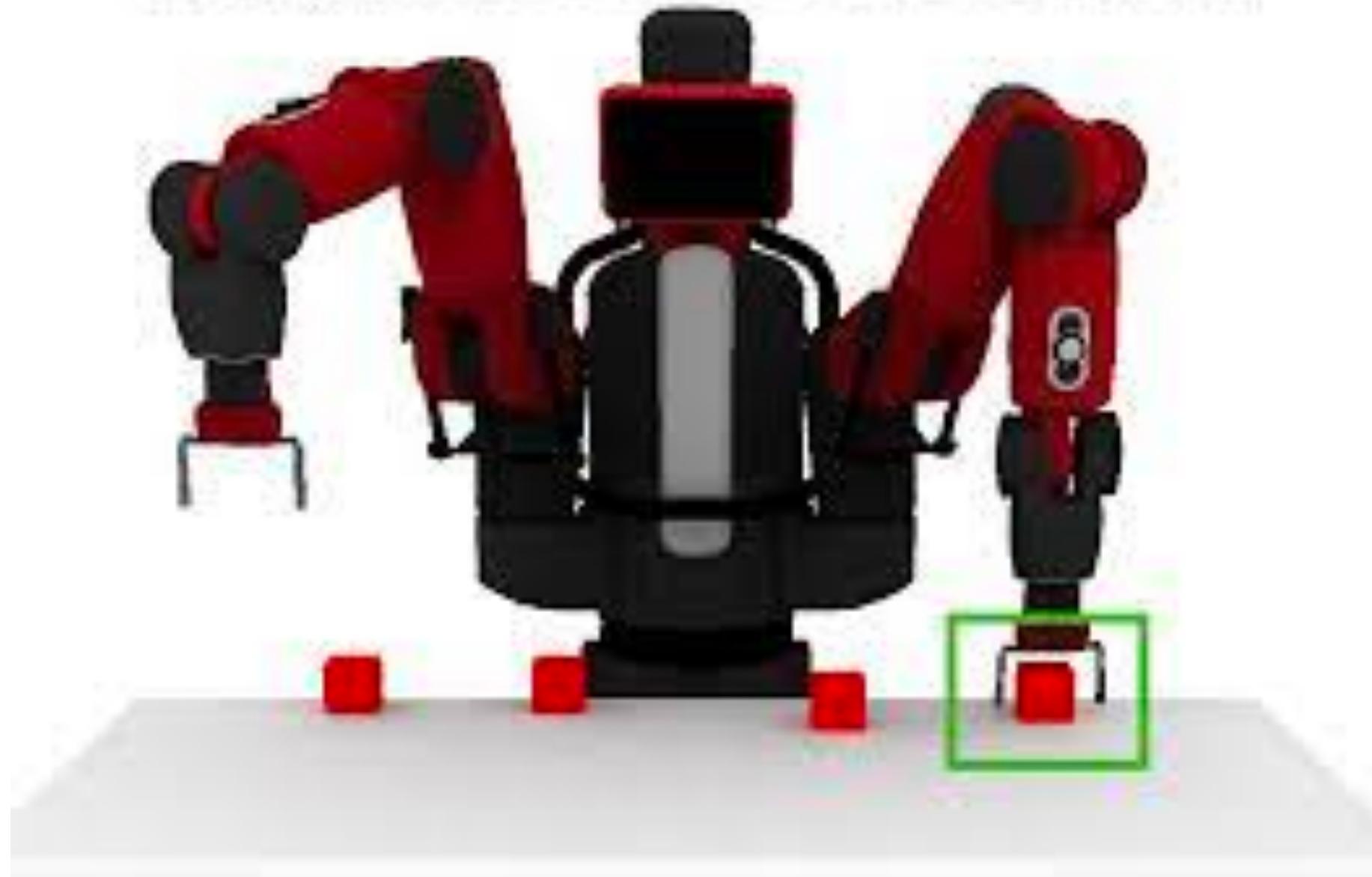
Two Fundamental Challenges

Two Fundamental Challenges

Challenge 1:

Ground natural language
in robot state

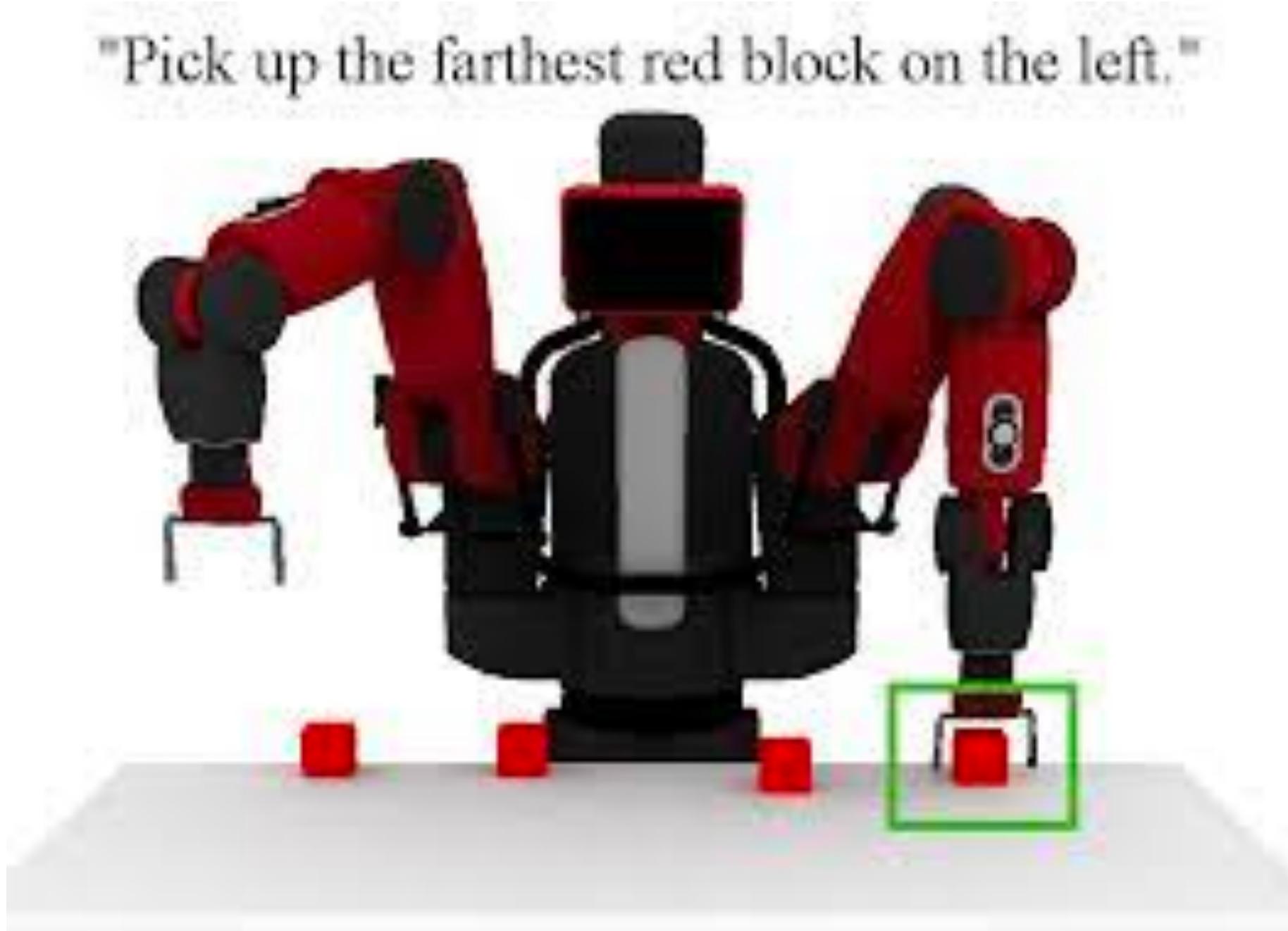
"Pick up the farthest red block on the left."



Two Fundamental Challenges

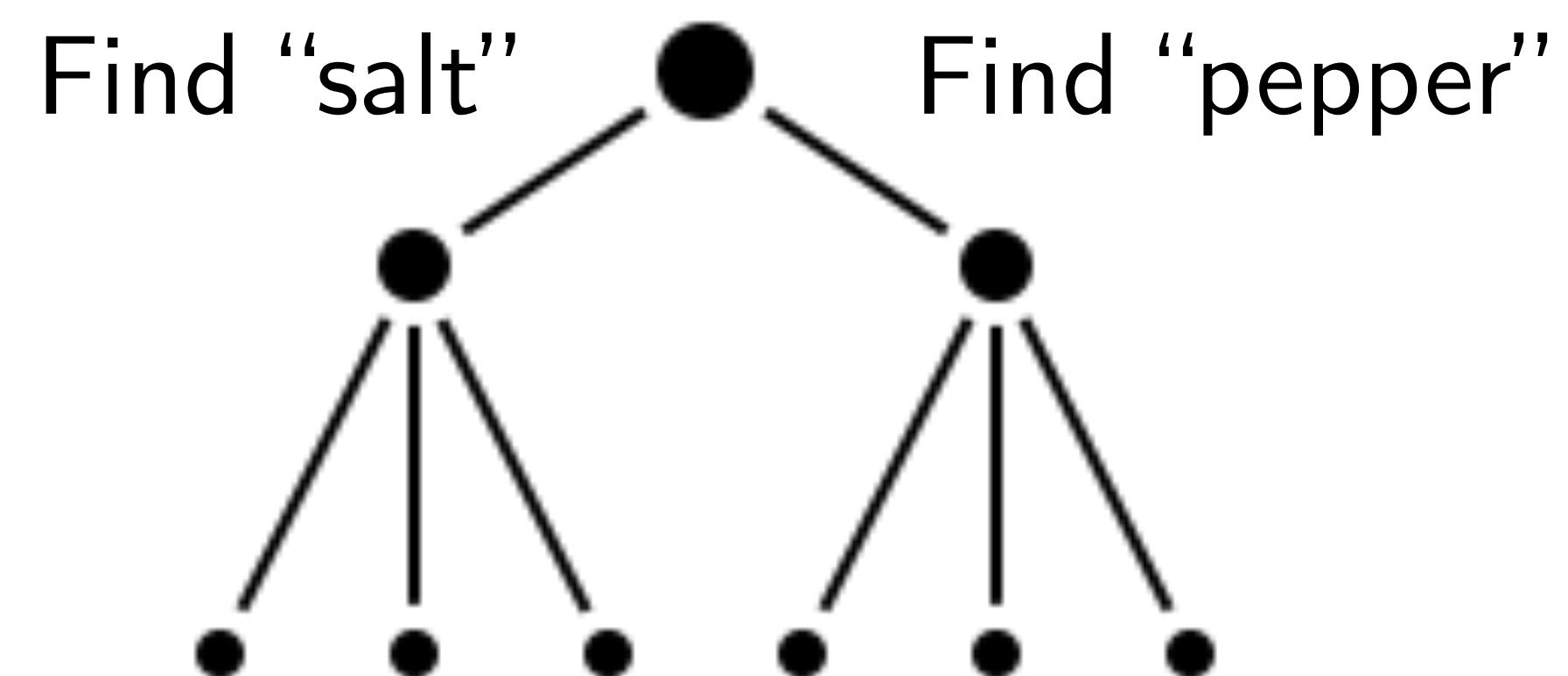
Challenge 1:

Ground natural language
in robot state



Challenge 2:

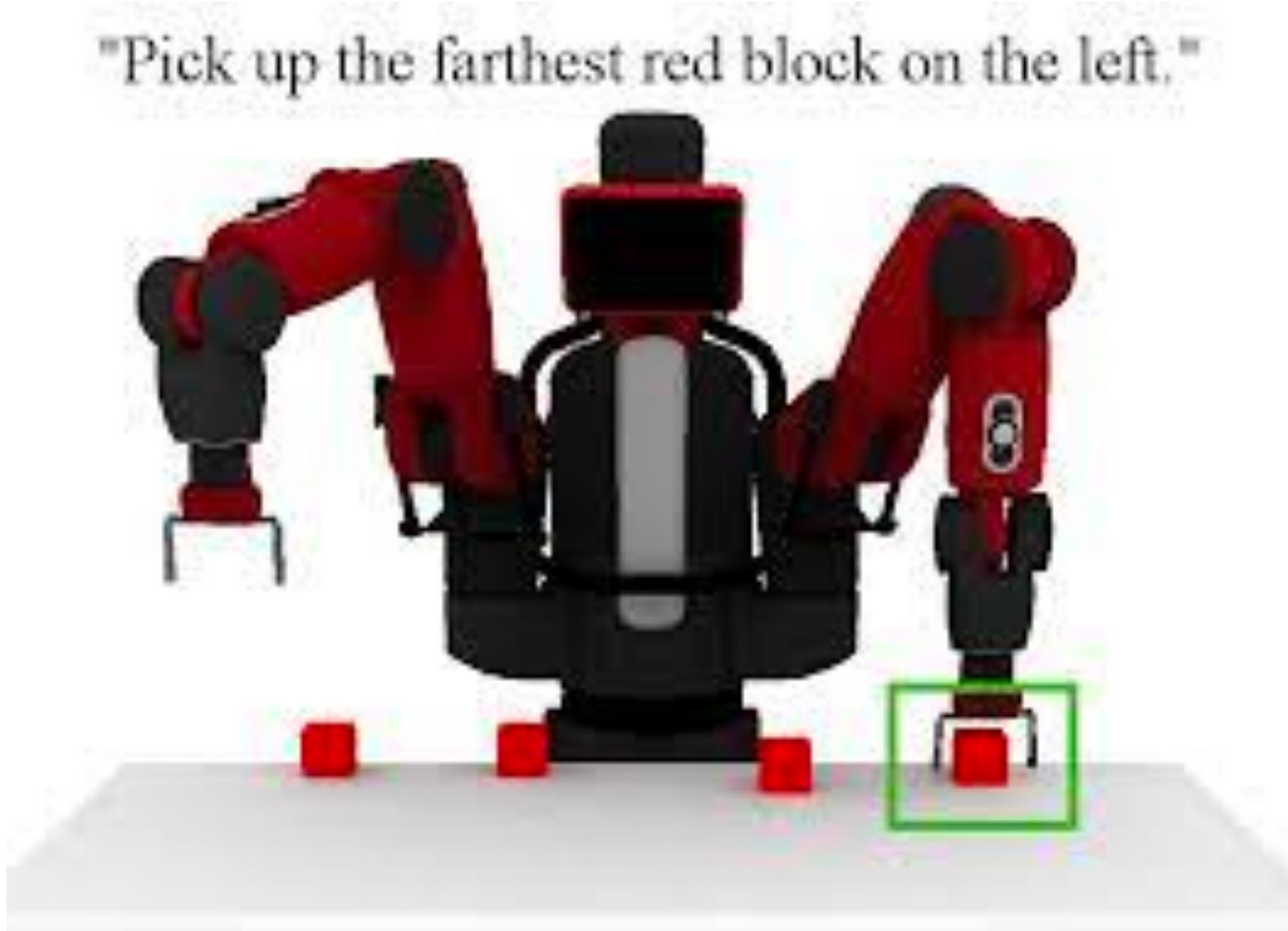
Planning actions to
solve a task



Two Fundamental Challenges

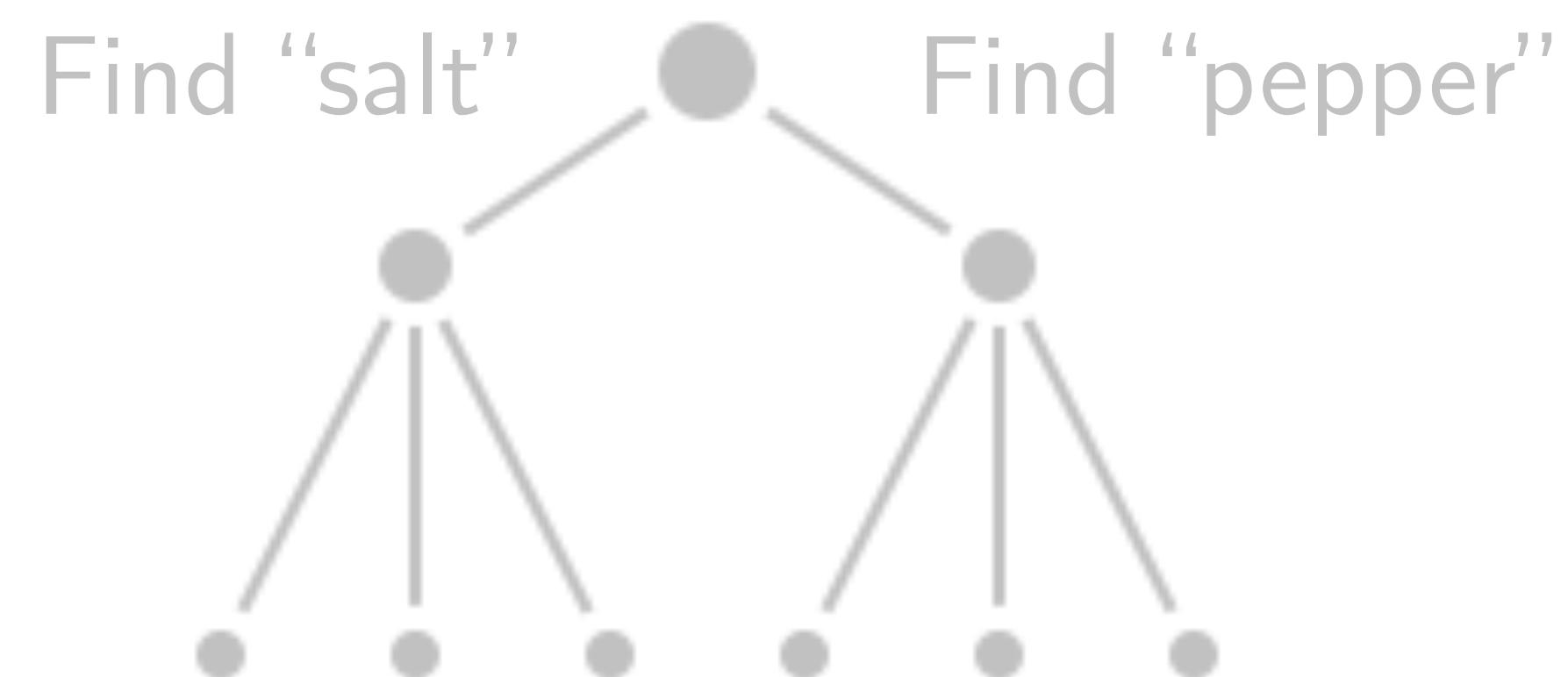
Challenge 1:

Ground natural language
in robot state



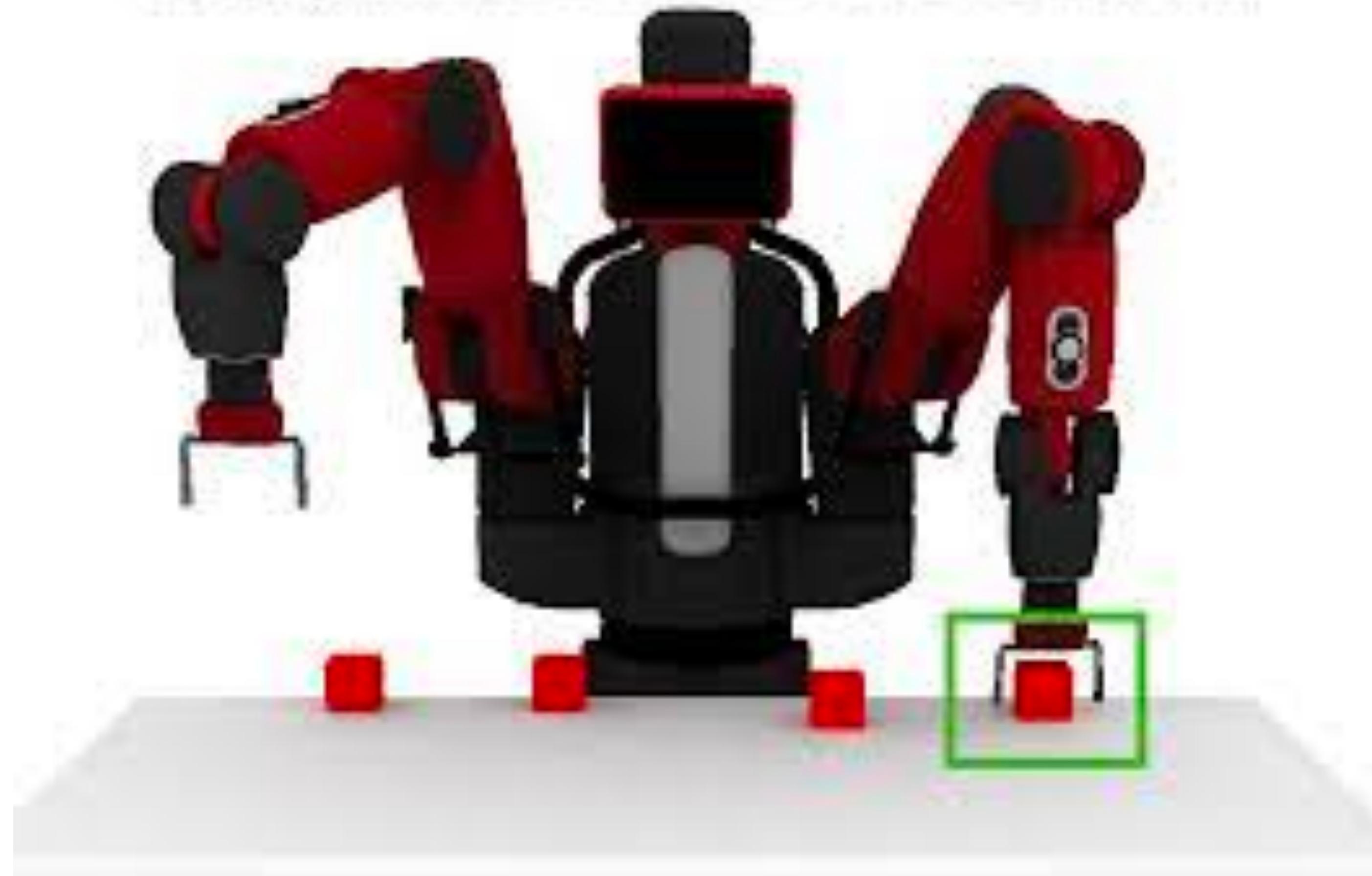
Challenge 2:

Planning actions to
solve a task



What is grounding? Why is it hard?

"Pick up the farthest red block on the left."



Grounding: Mapping language to robot's internal state

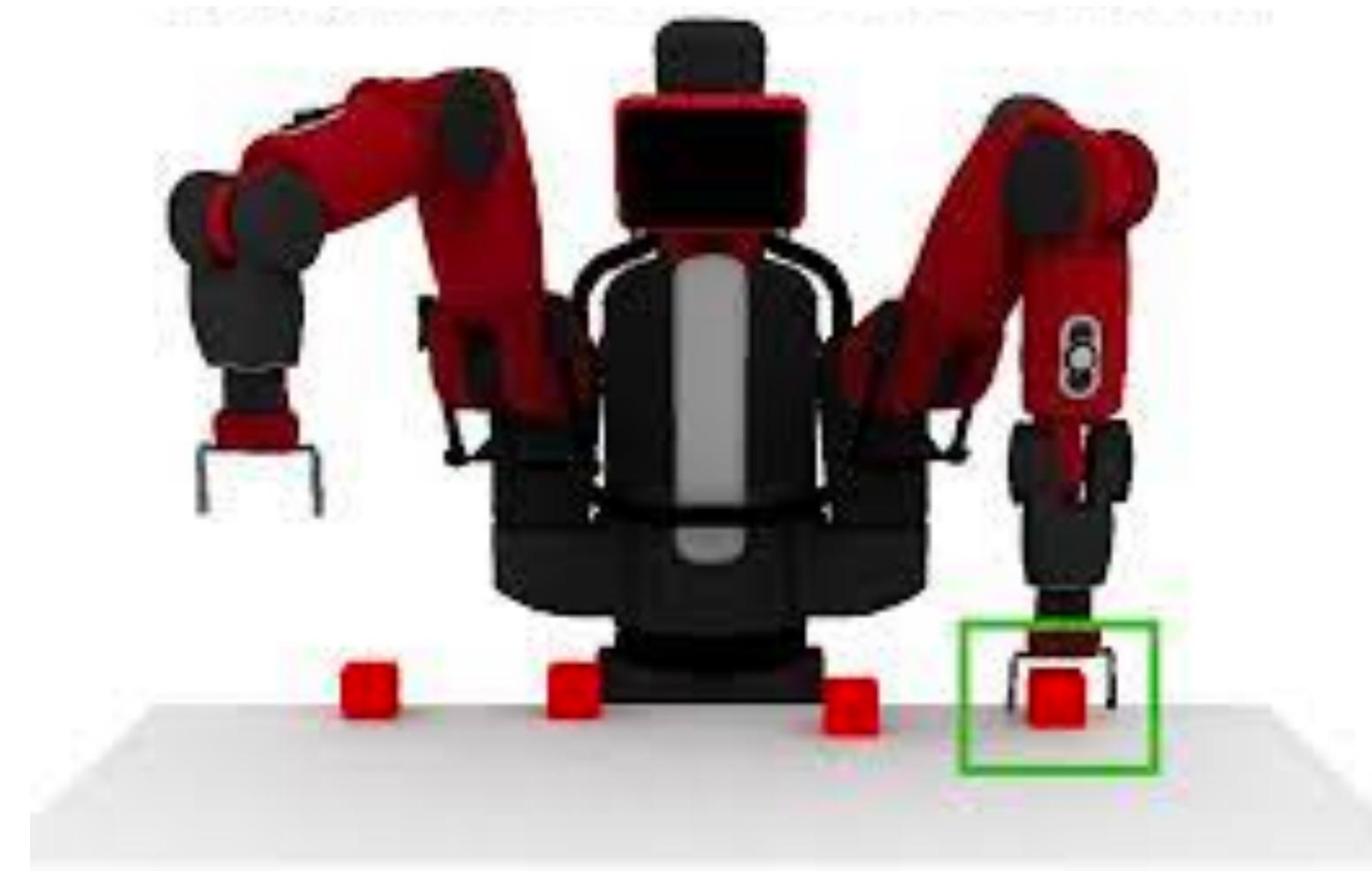
Natural Language



MDP

“Pick up the farthest
red block”

$$\langle S, A, R, \mathcal{T} \rangle$$



Grounding: Mapping language to robot's internal state

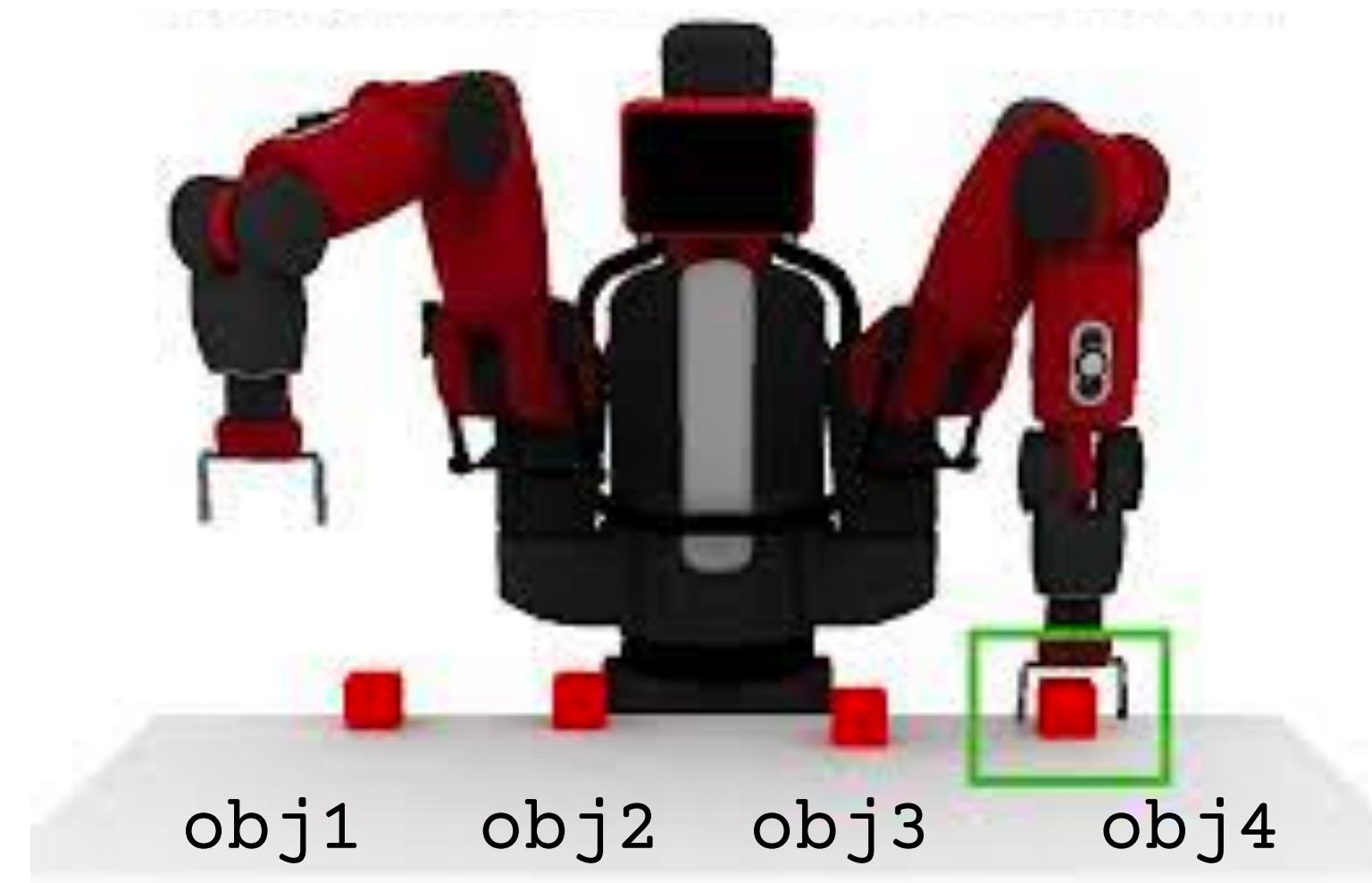
Natural Language



MDP

“Pick up the farthest red block”

$\langle S, A, R, \mathcal{T} \rangle$



```
on('obj1','table')
on('obj2','table')
on('obj3','table')
on('obj4','table')
left('obj2','obj1')
left('obj3','obj2')
left('obj4','obj3')
...
```

Grounding: Mapping language to robot's internal state

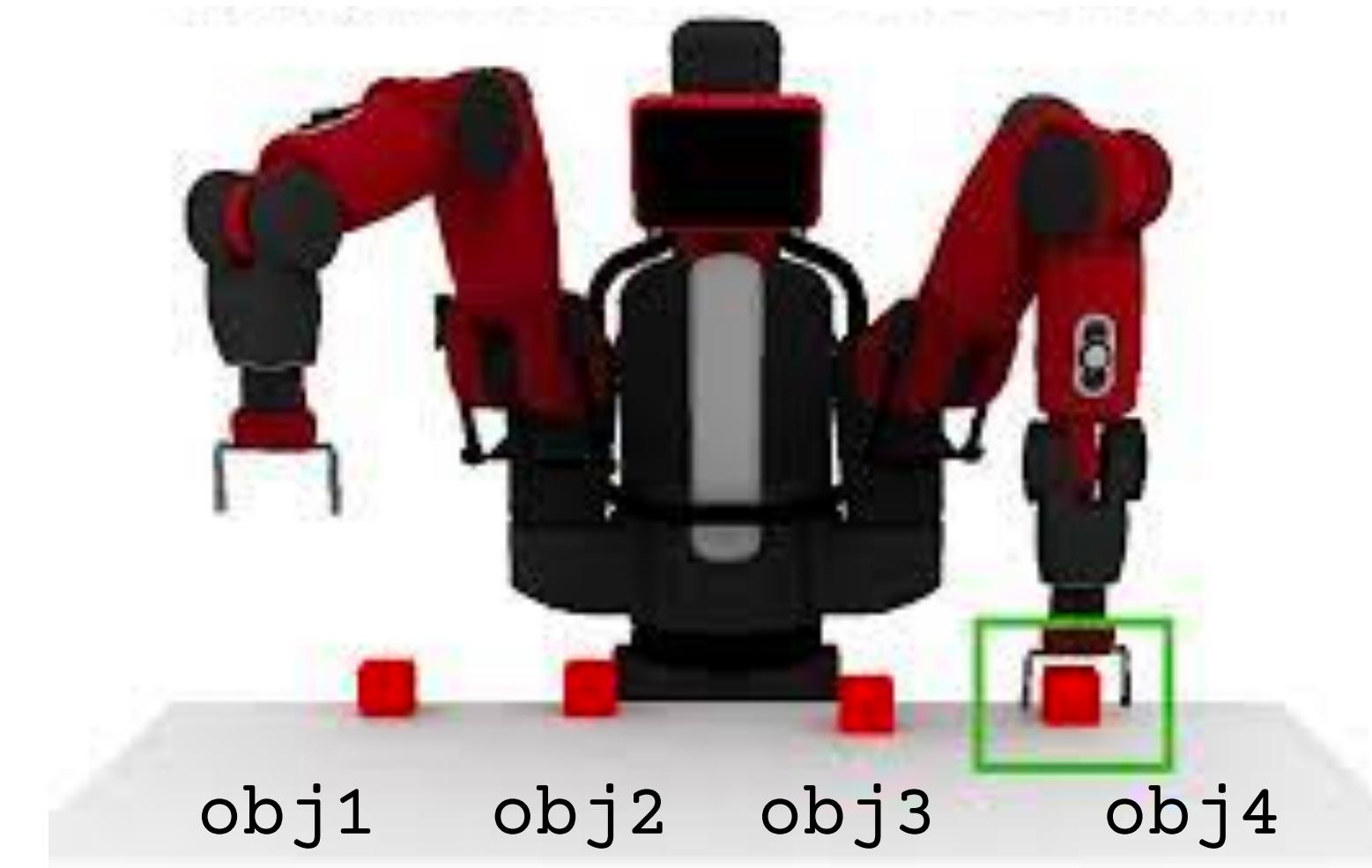
Natural Language



MDP

“Pick up the farthest red block”

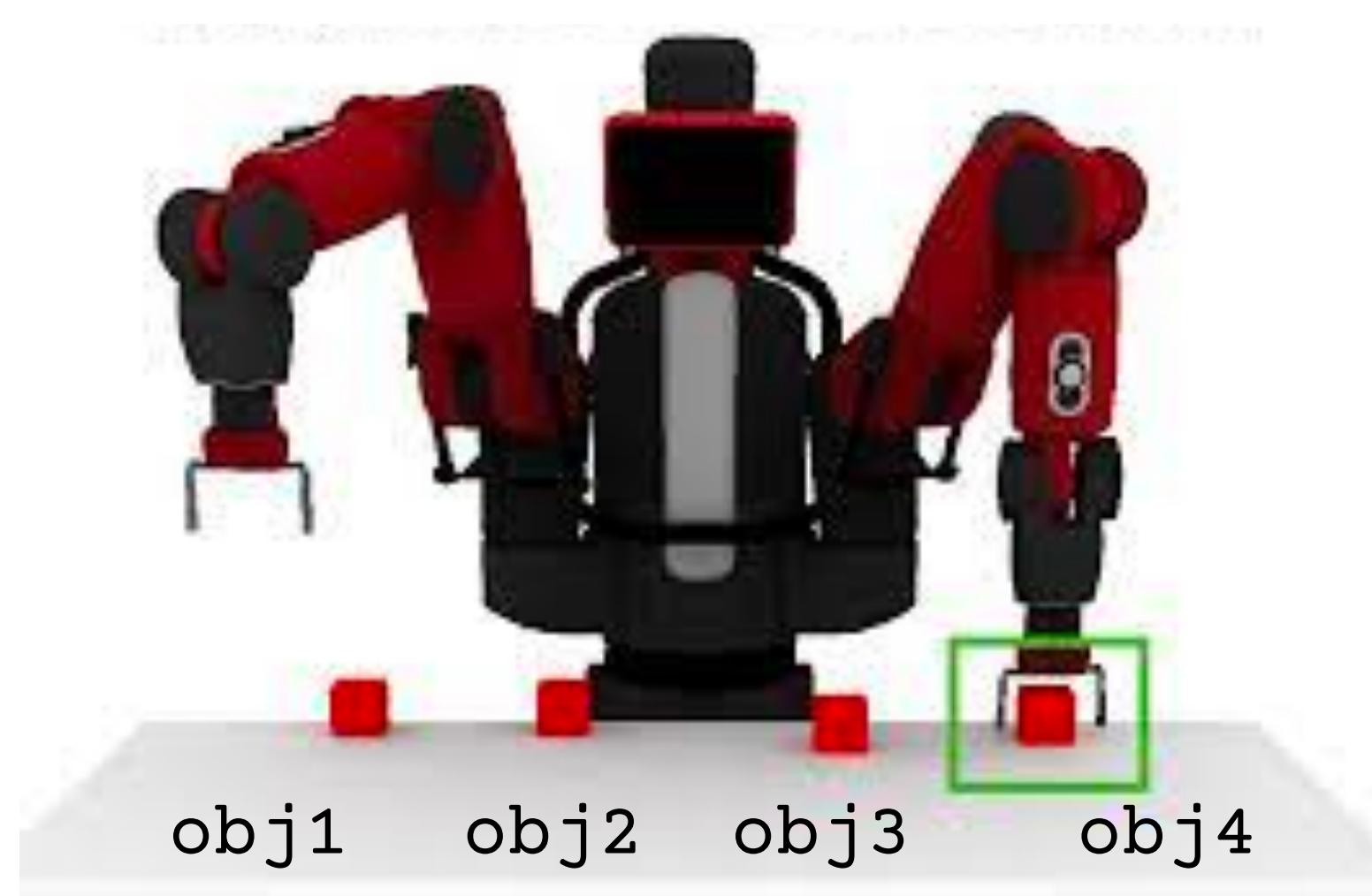
$\langle S, A, R, \mathcal{T} \rangle$



$$R(\text{in}(\text{obj4}, \text{hand})) = +1$$

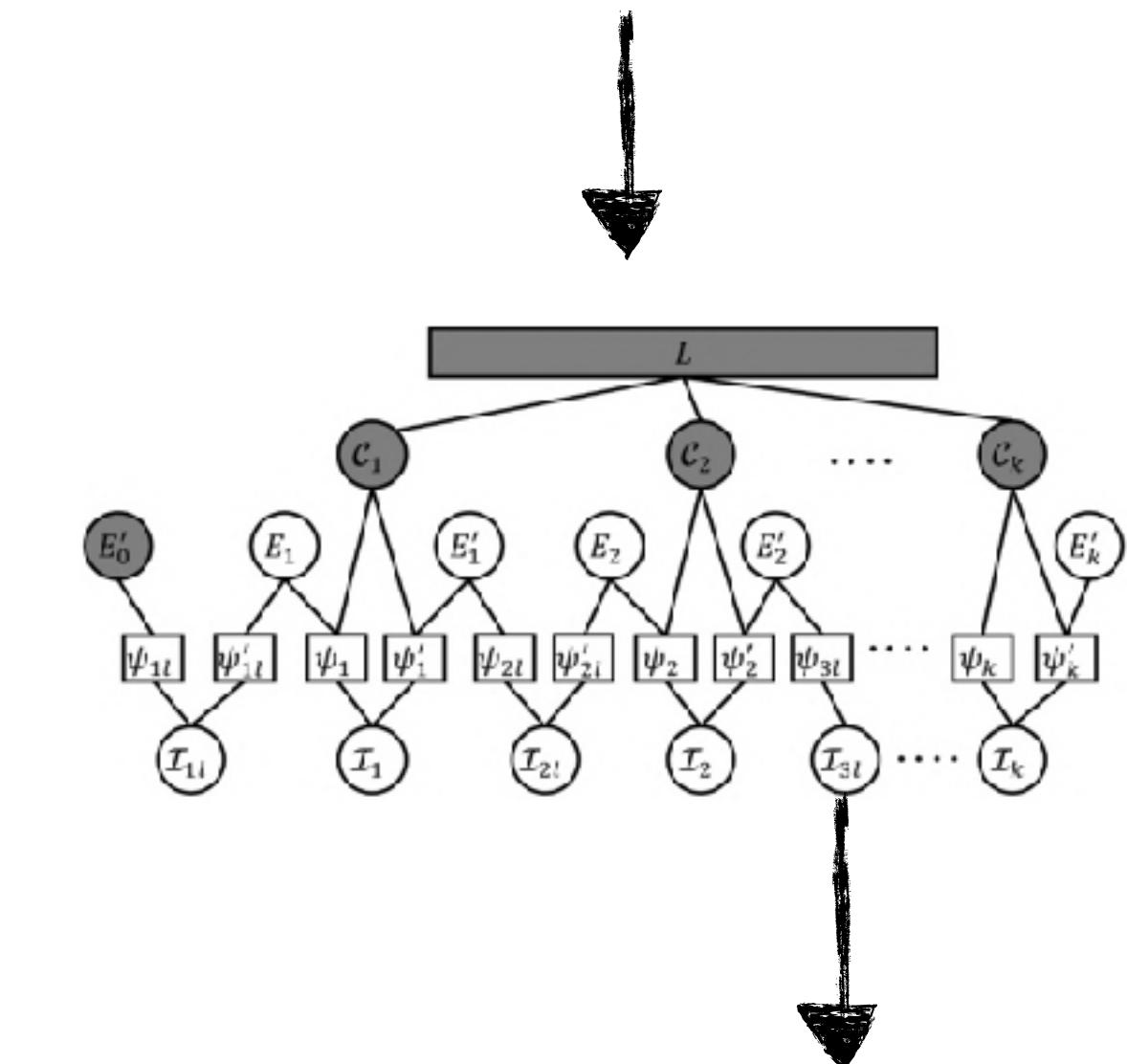
How did we *solve* grounding?

Train this on small, custom robot datasets!



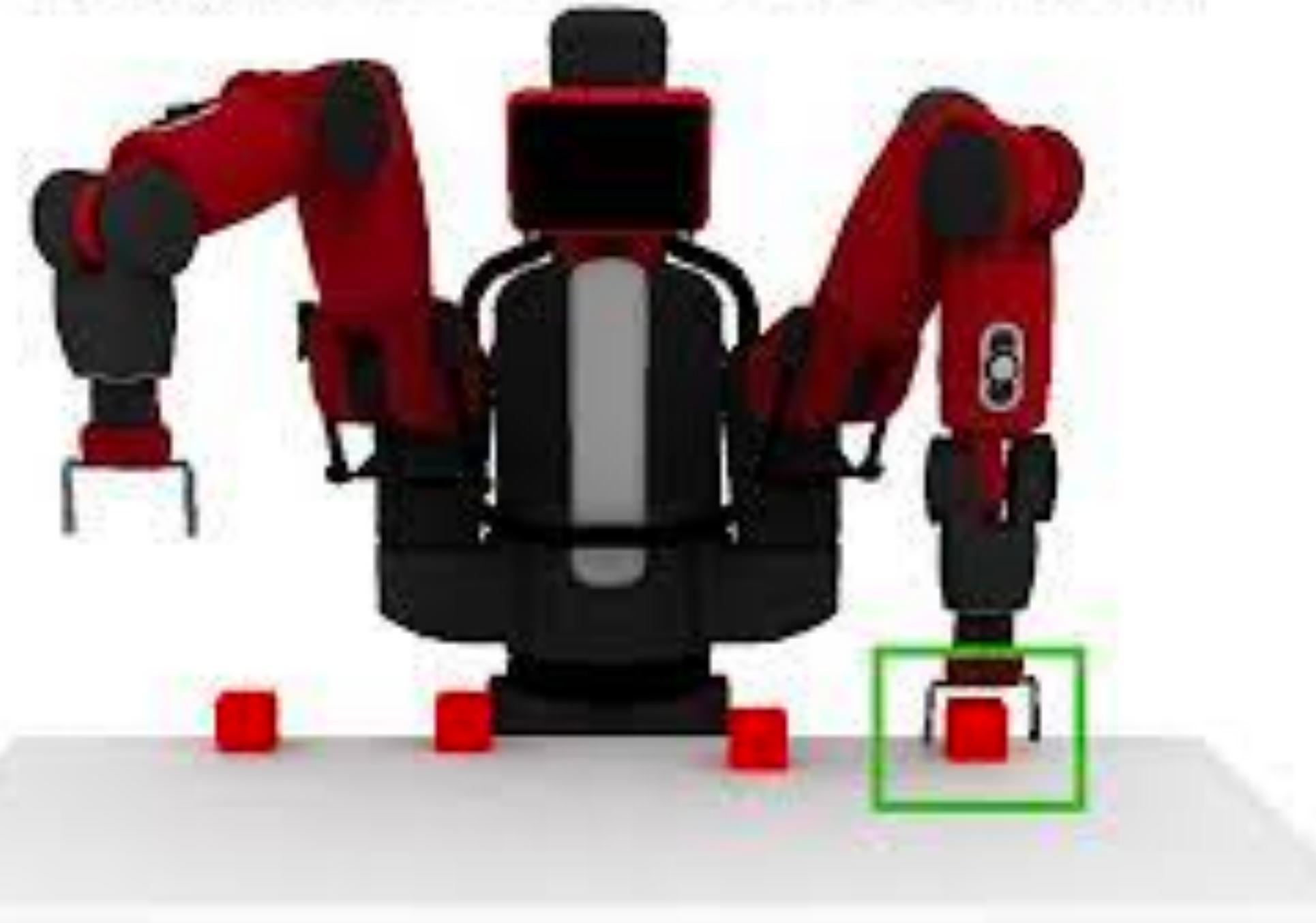
Complex graphical models!

“Pick up the farthest red block”



Why did this not scale?

"Pick up the farthest red block on the left."

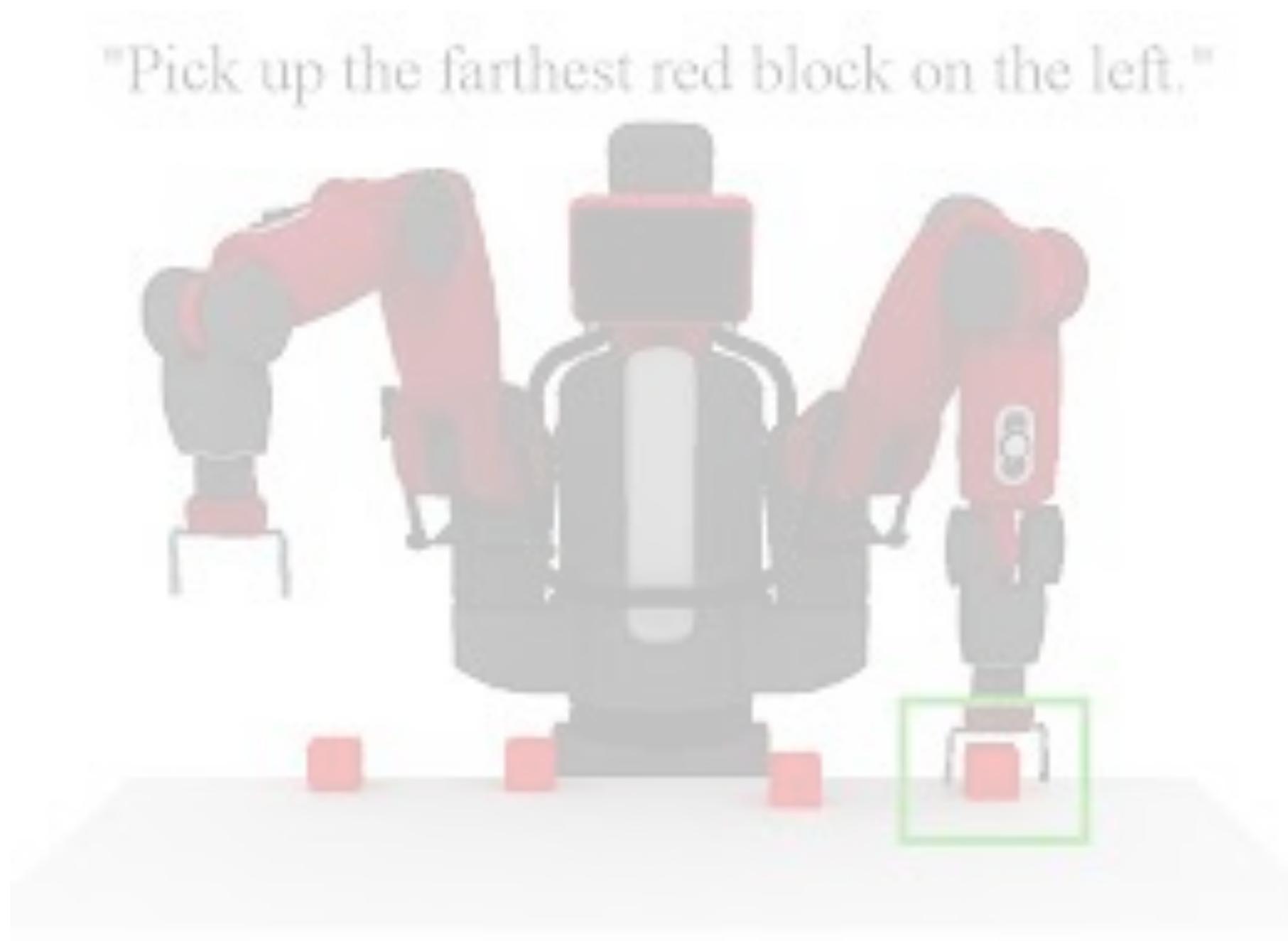


1. Failure to generalize to different human utterances
2. Failure to capture common sense
3. Failure to capture complex instructions (while loops)

Two Fundamental Challenges

Challenge 1:

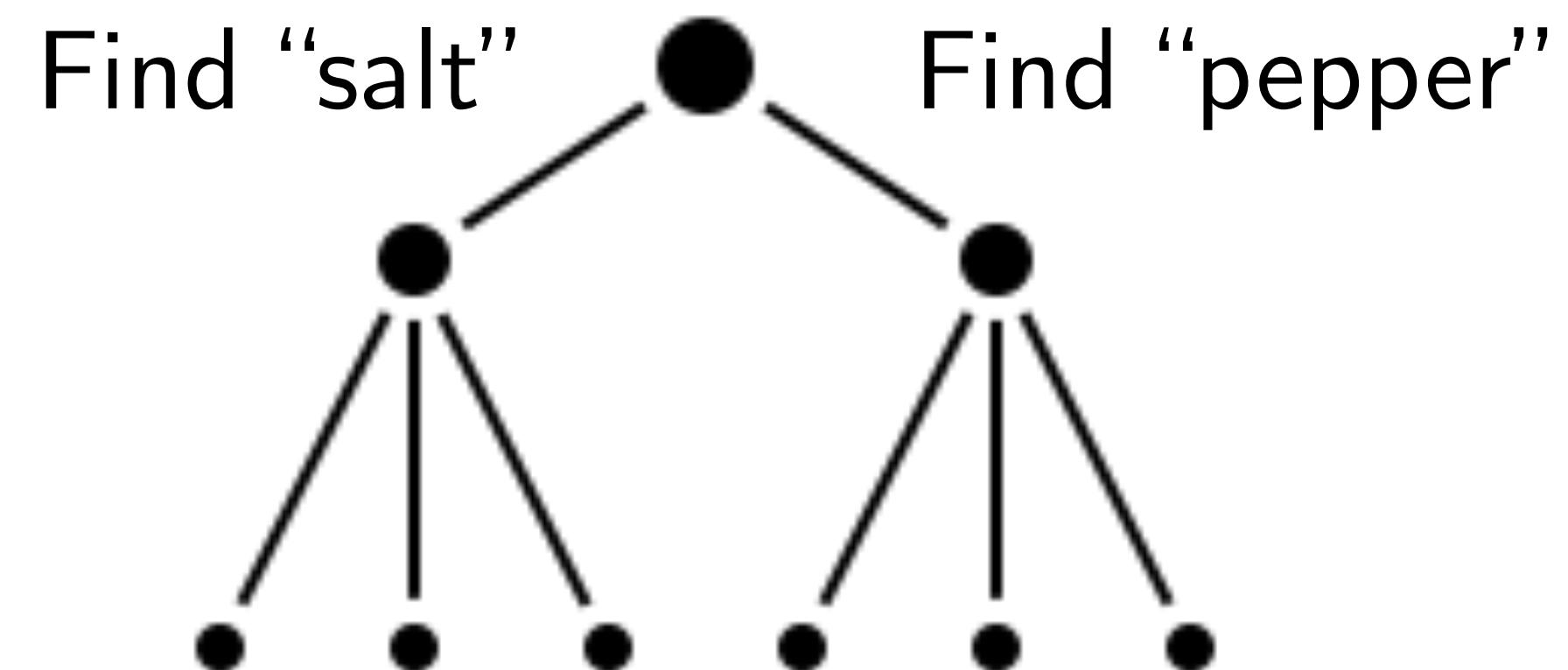
Ground natural language
in robot state



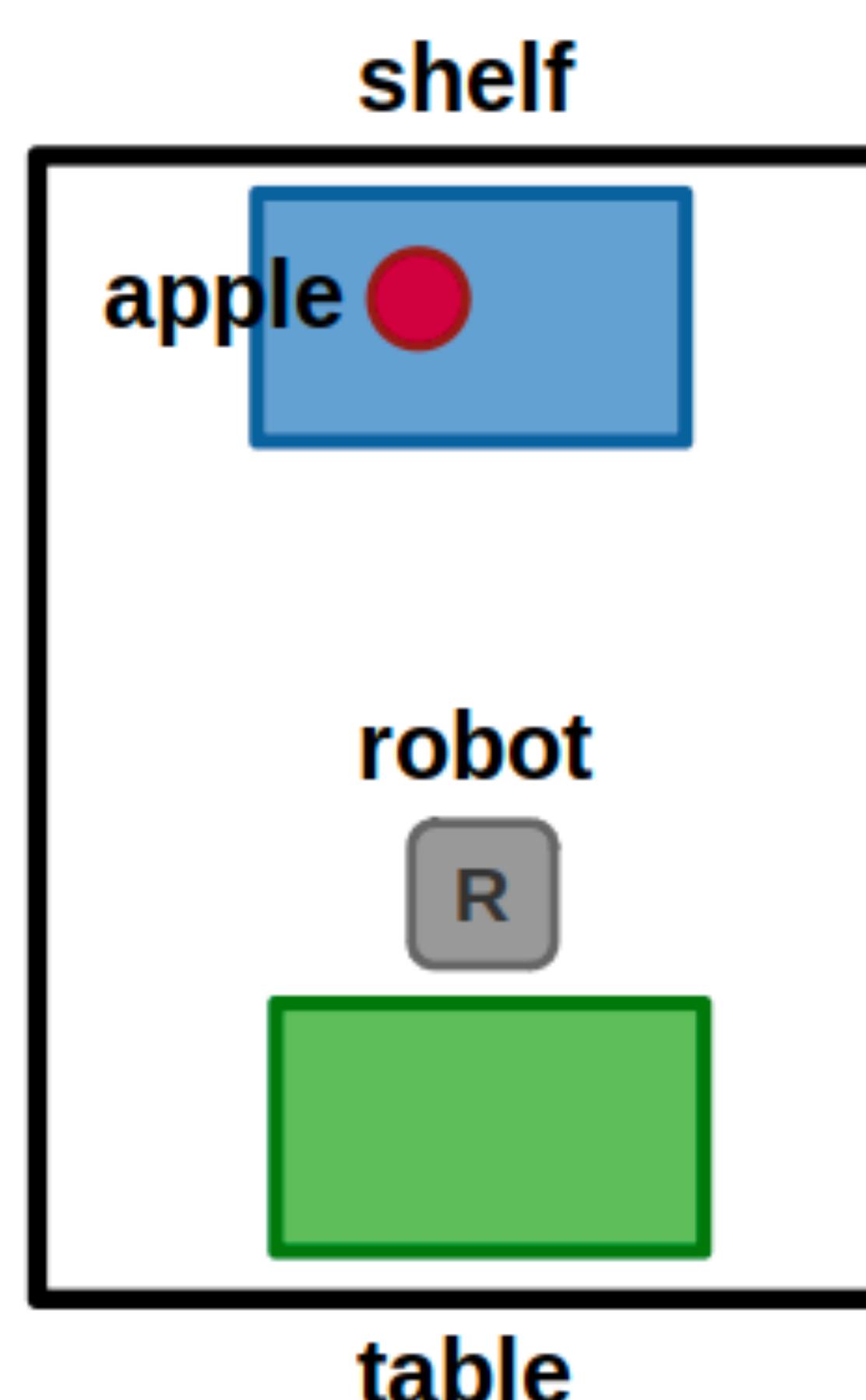
"Pick up the farthest red block on the left."

Challenge 2:

Planning actions to
solve a task

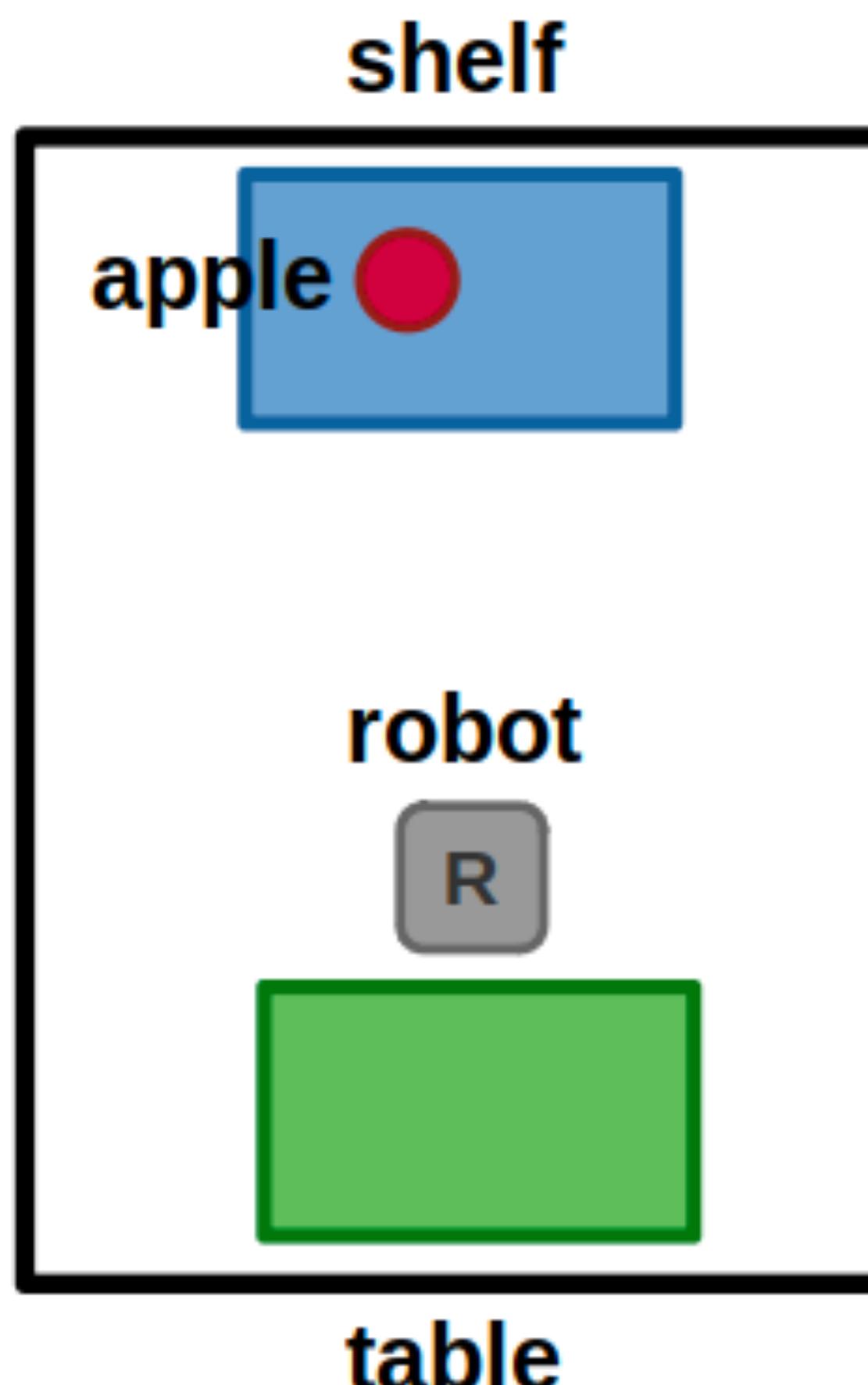


What is task planning? Why is it hard?



Take the apple from the shelf and put it on the table

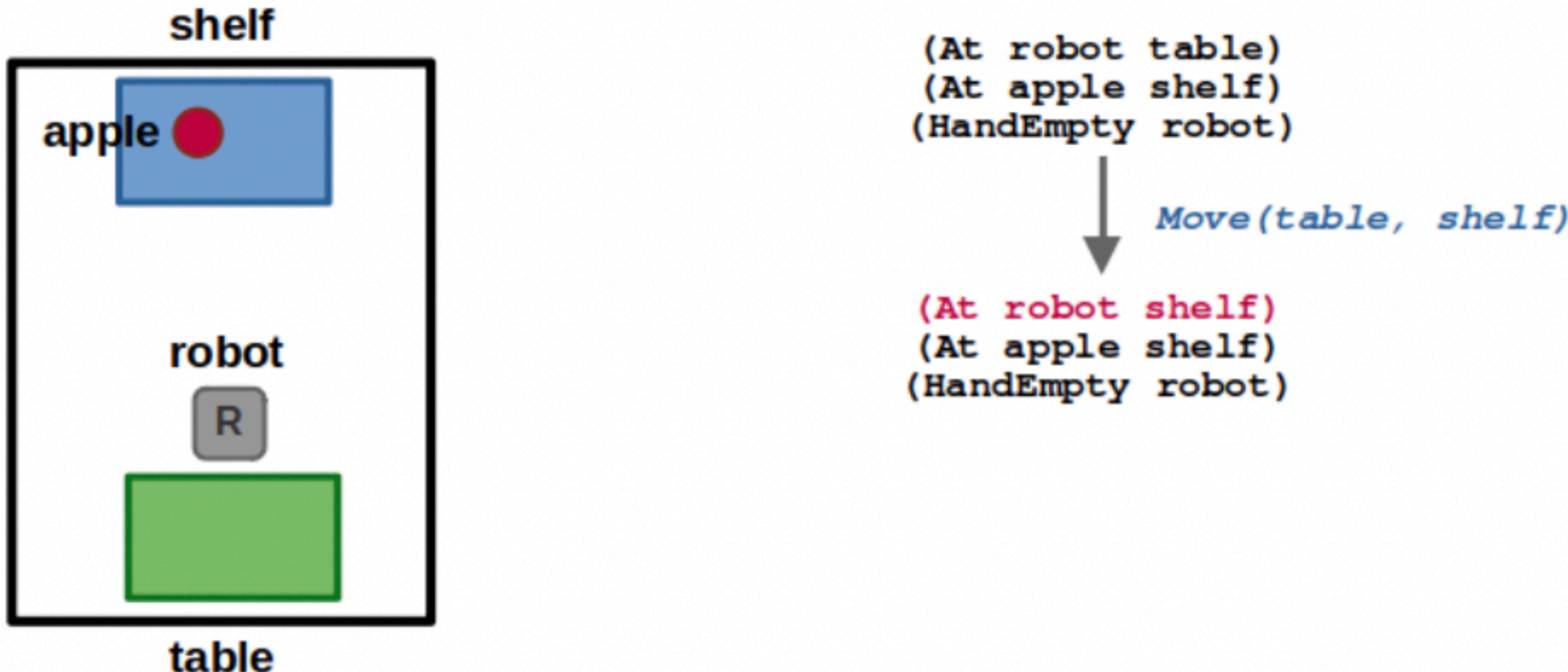
What is task planning? Why is it hard?



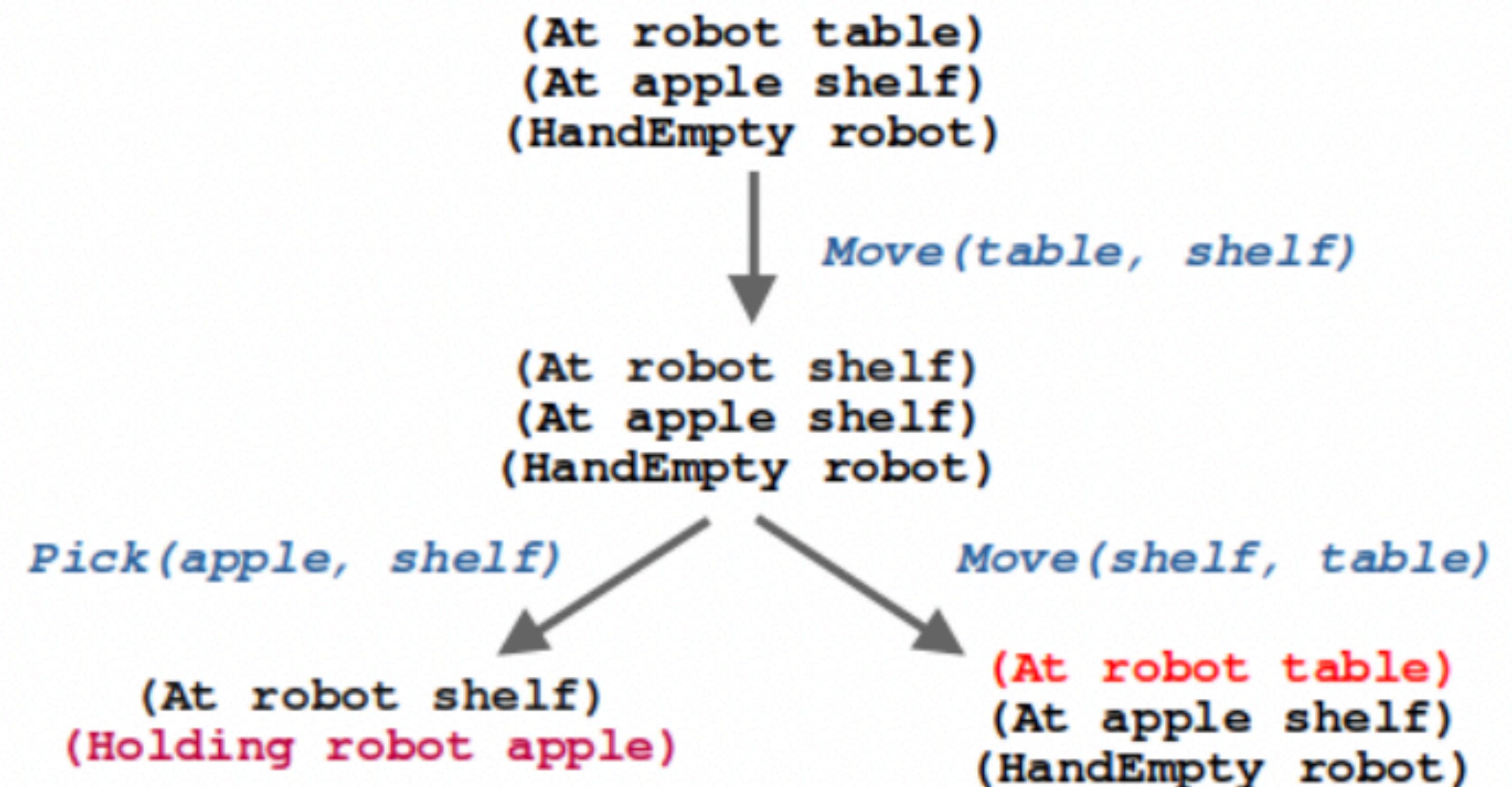
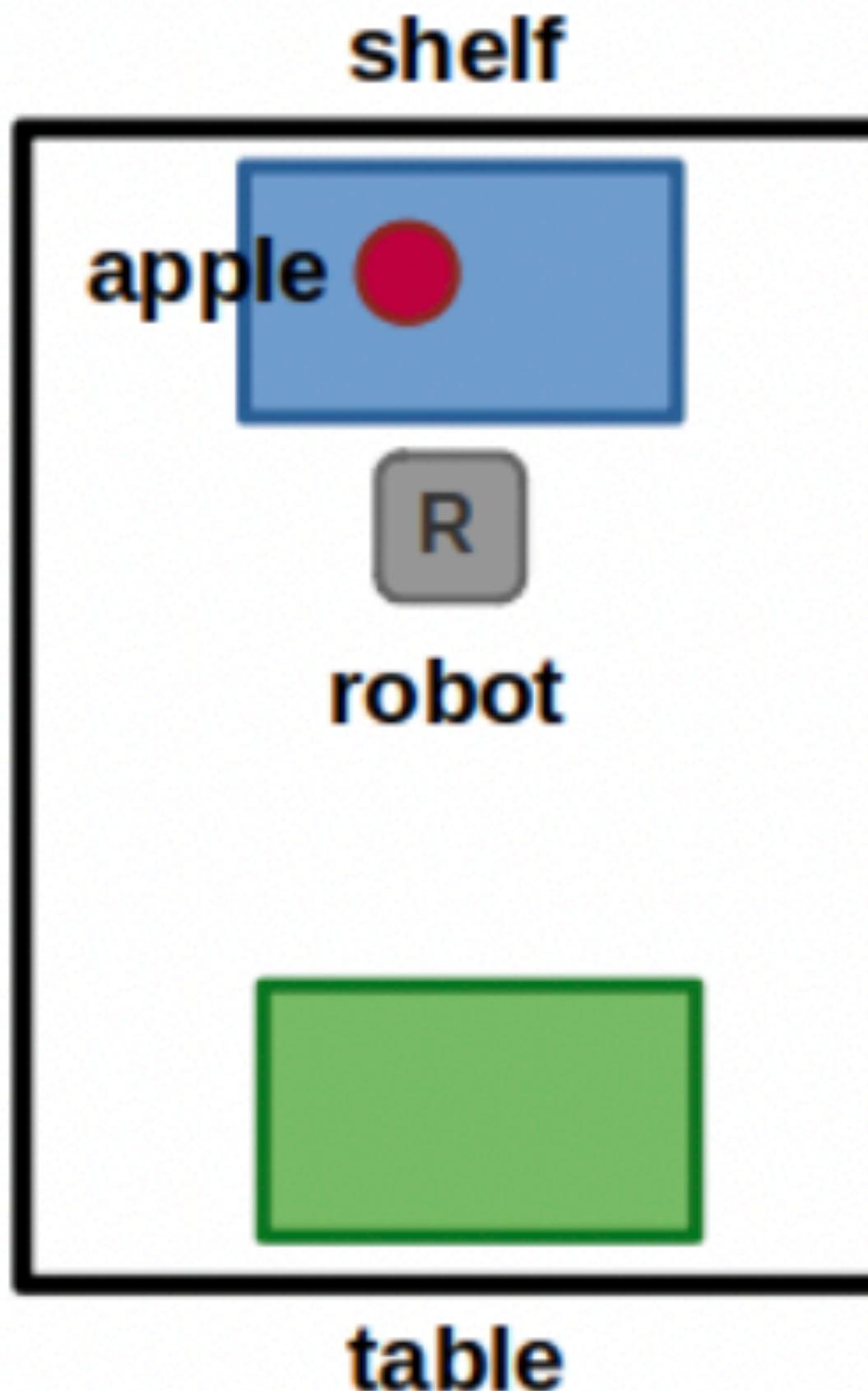
Take the apple from the shelf and put it on the table

1. Move to the shelf
2. Pick up the apple
3. Move back to the table
4. Place the apple

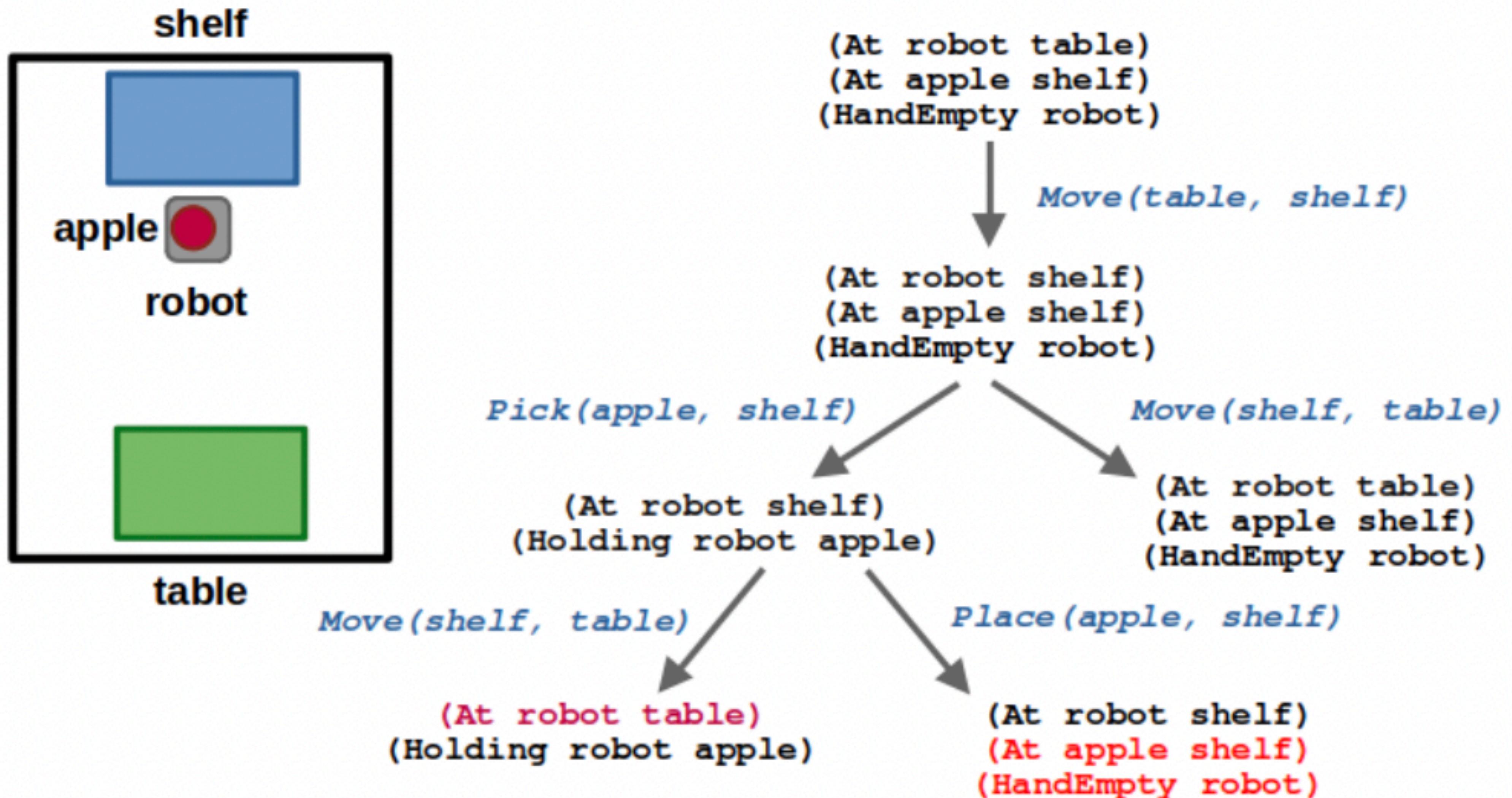
What is task planning? Why is it hard?



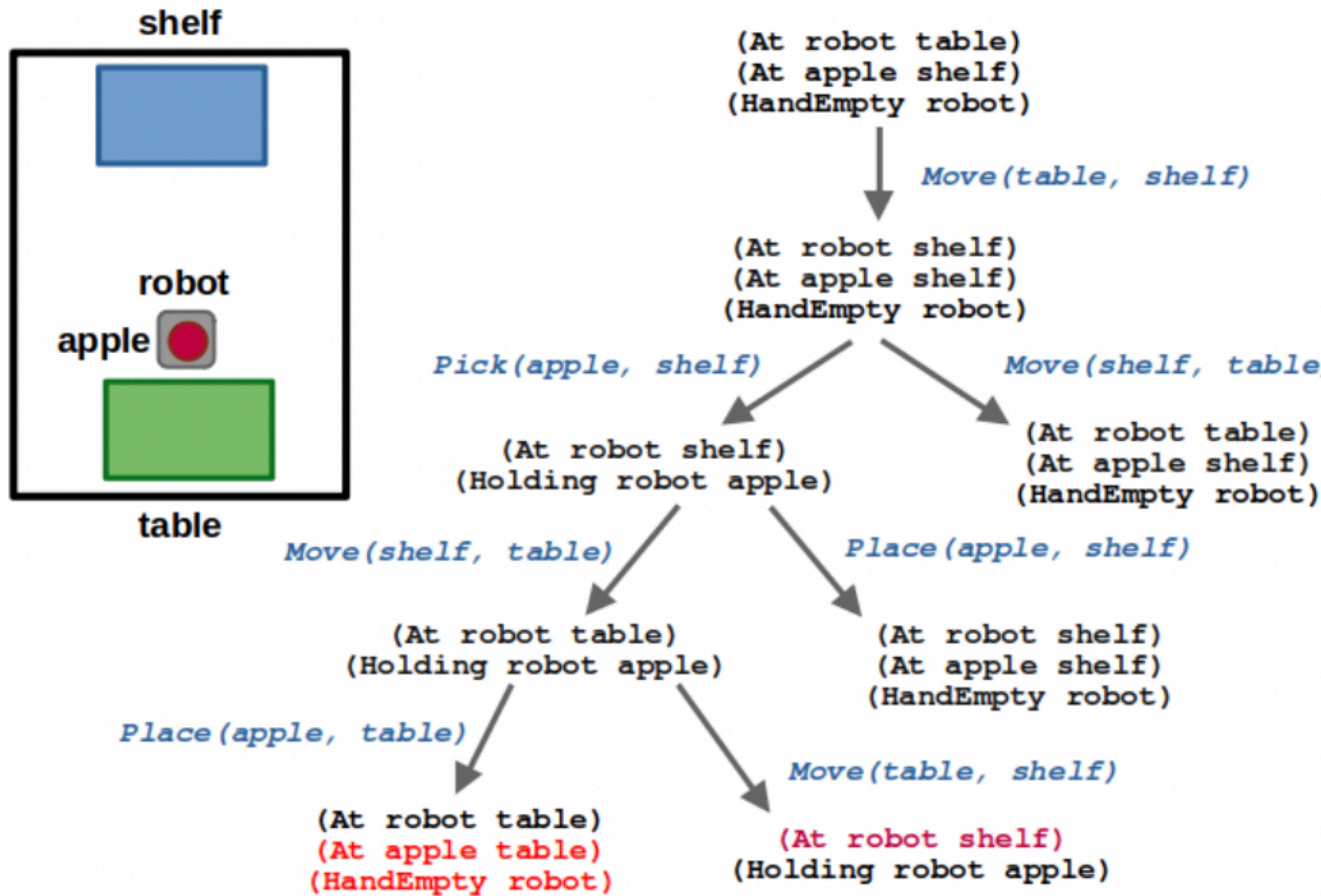
What is task planning? Why is it hard?



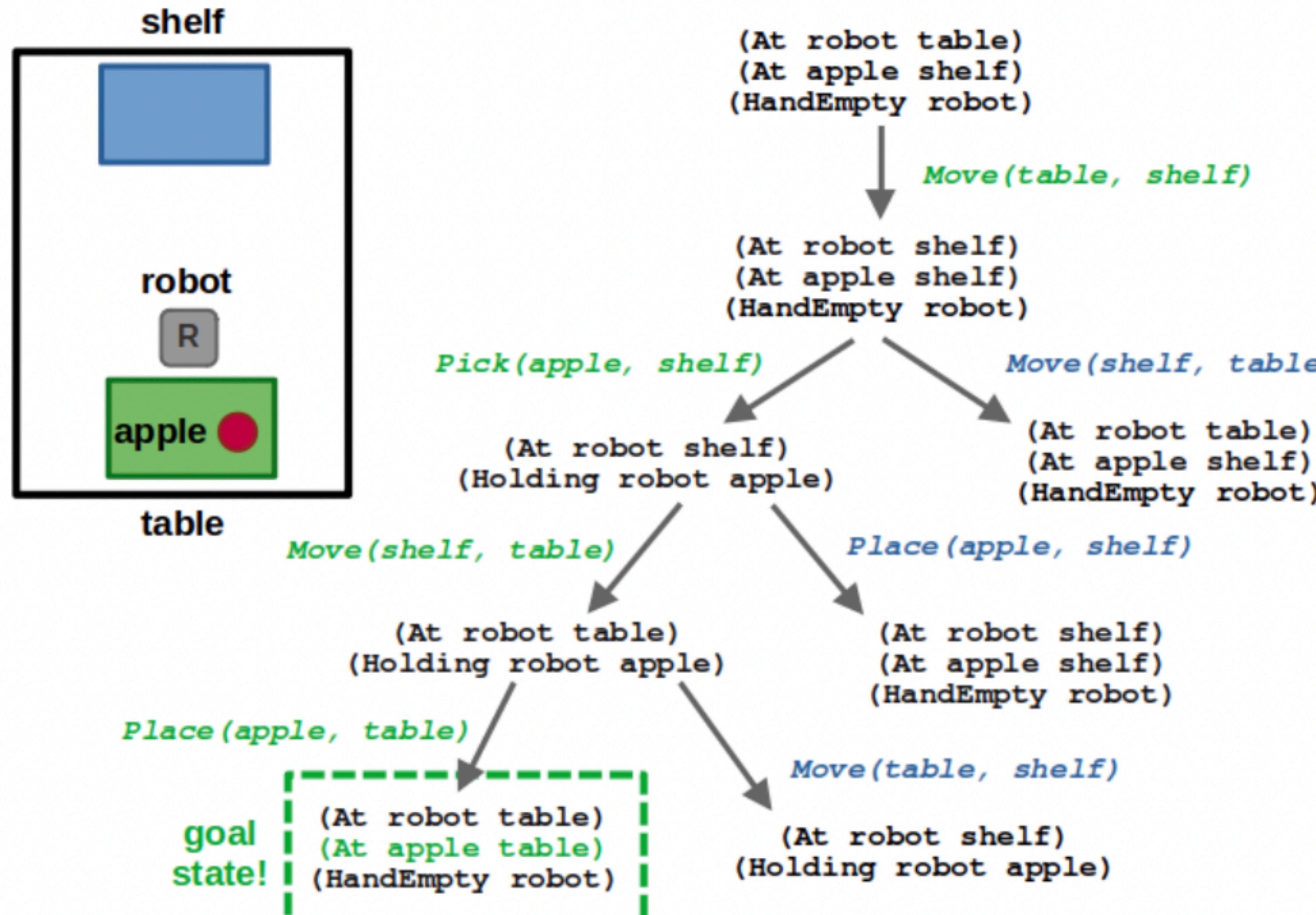
What is task planning? Why is it hard?



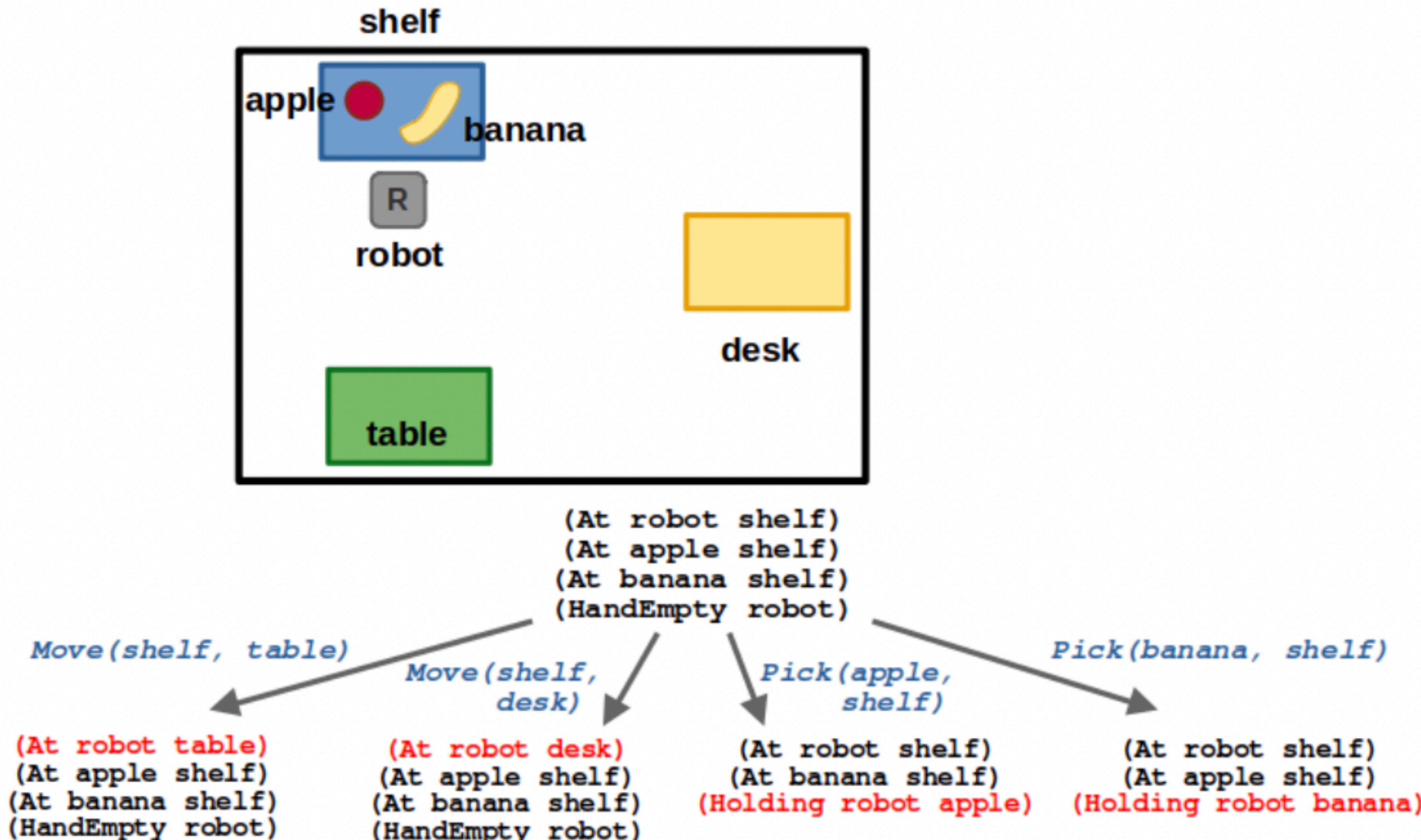
What is task planning? Why is it hard?



What is task planning? Why is it hard?



What is task planning? Why is it hard?



How did we **solve** it?

Good old fashioned search!

Lots of heuristics to make it real time

Why did it not **scale**?

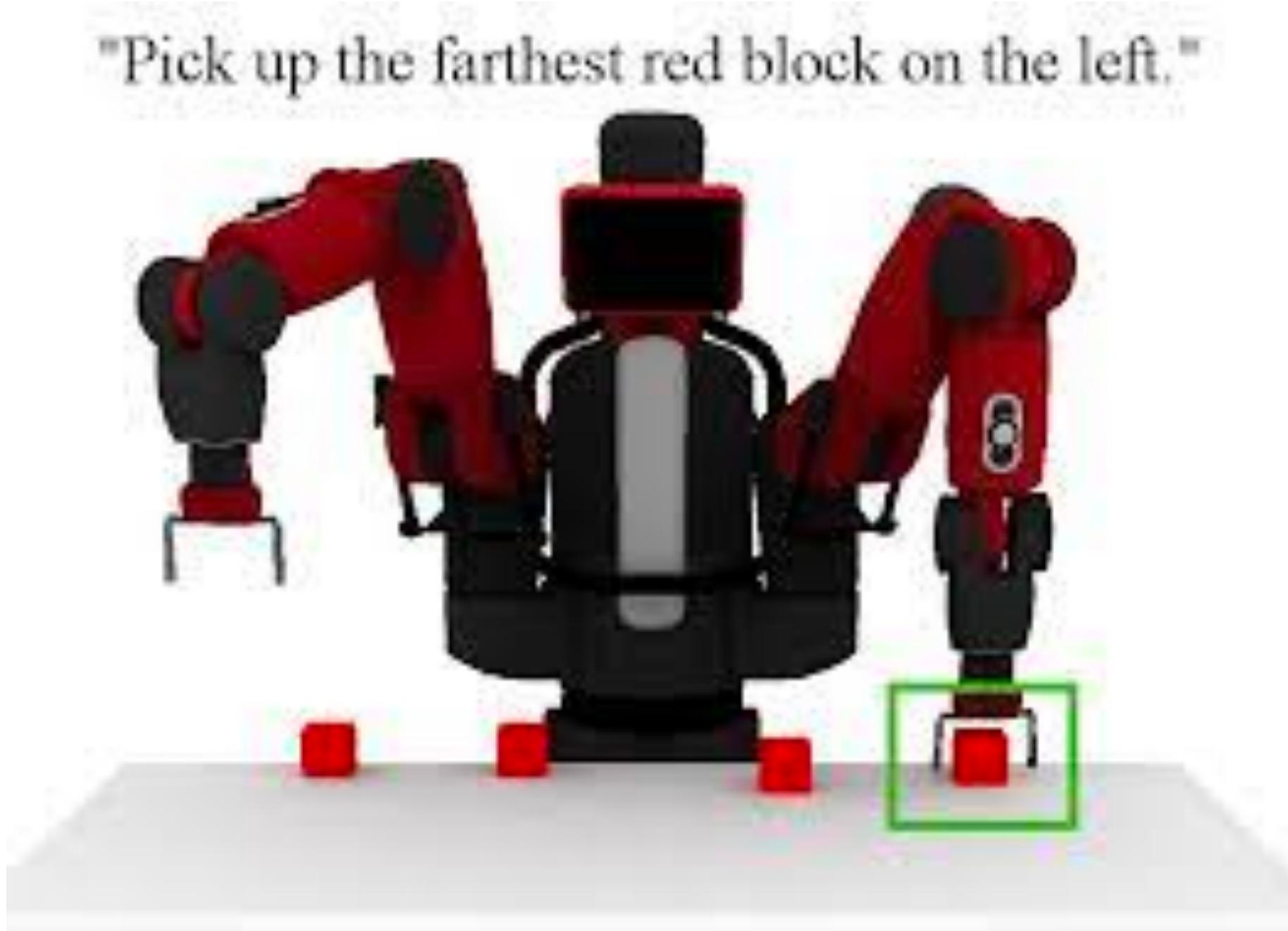
Combinatorially large search tree

Had no notion of common sense

Two Fundamental Challenges

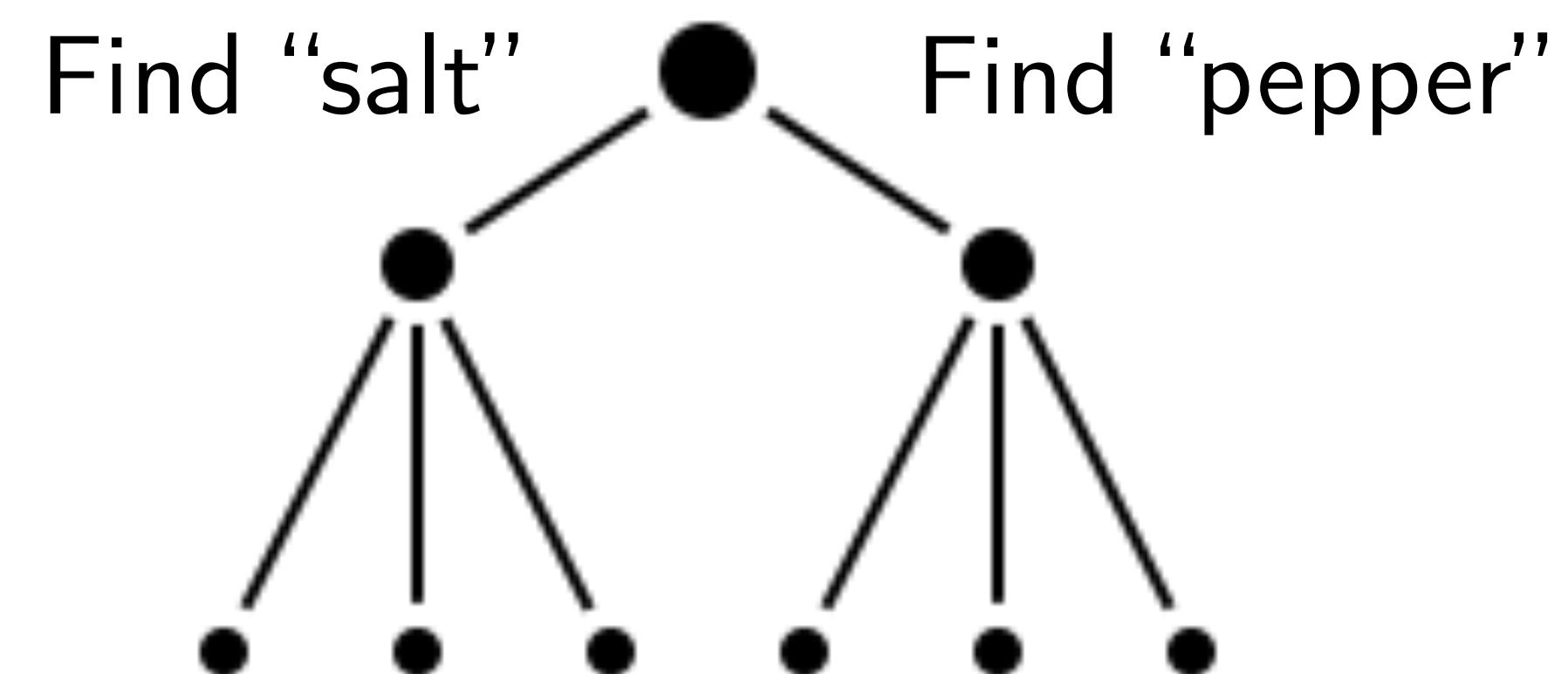
Challenge 1:

Ground natural language
in robot state



Challenge 2:

Planning actions to
solve a task



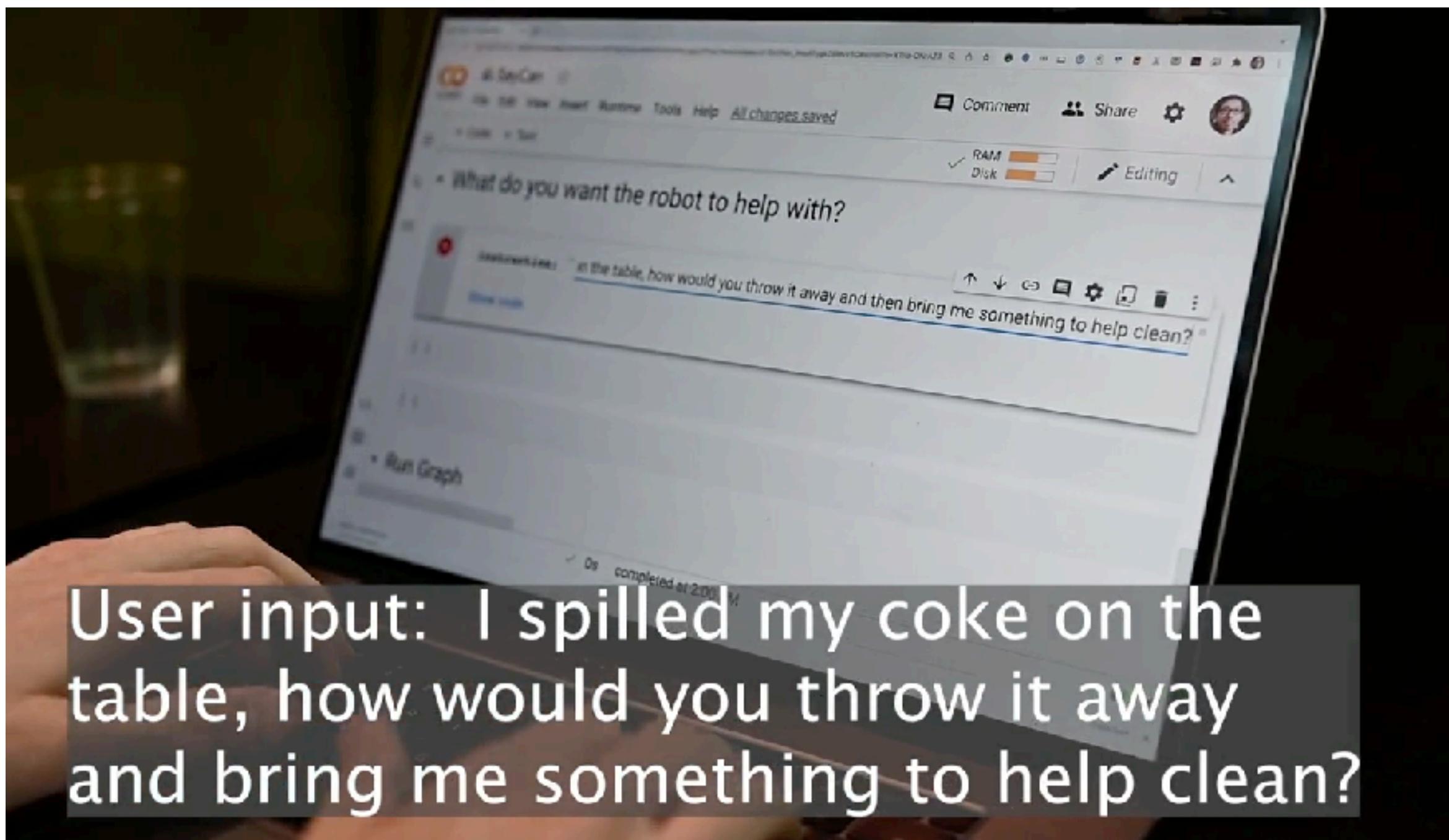
LARGE LANGUAGE MODELS

Episode IV

A NEW HOPE

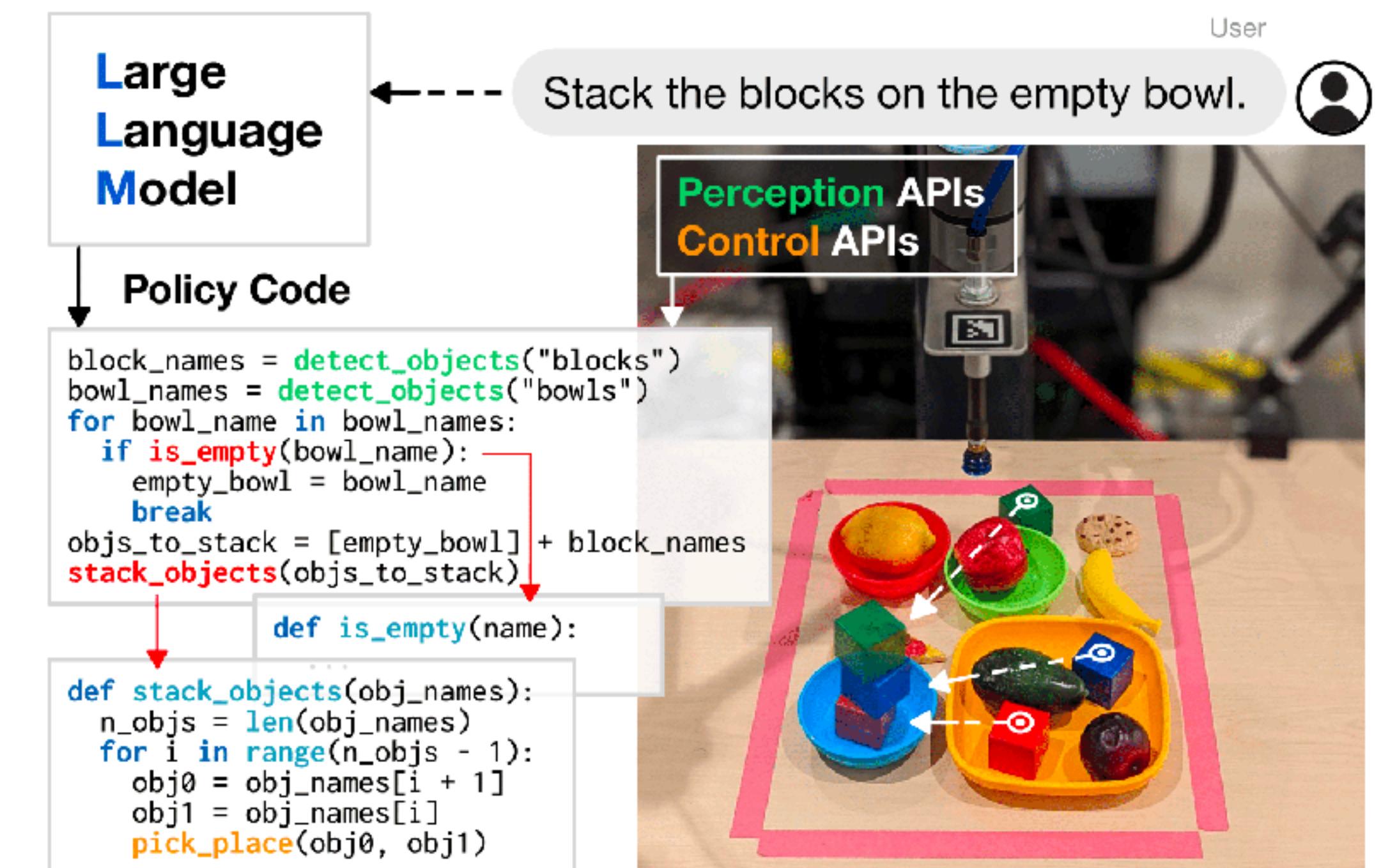
Many recent papers on LLM+Task Planning

SayCan [Ichter et al.'22]



User input: I spilled my coke on the table, how would you throw it away and bring me something to help clean?

Code-As-Policies [Liang et al.'22]



Also ProgPrompt [Singh et al. '22], InnerMonologue [Huang et al.'22], Socrative [Zeng et al.'22], TidyBot [Wu et al'23], CLARIFY [Skreta et al.'23], Text2Motion [Lin et al. '23], ...

Can LLMs directly
predict robot action?

Do As I Can, Not As I Say:

Grounding Language in Robotic Affordances

Michael Ahn* Anthony Brohan* Noah Brown* Yevgen Chebotar* Omar Cortes* Byron David* Chelsea Finn*
Chuyuan Fu* Keerthana Gopalakrishnan* Karol Hausman* Alex Herzog* Daniel Ho* Jasmine Hsu* Julian Ibarz*
Brian Ichter* Alex Irpan* Eric Jang* Rosario Jauregui Ruano* Kyle Jeffrey* Sally Jesmonth* Nikhil Joshi*
Ryan Julian* Dmitry Kalashnikov* Yuheng Kuang* Kuang-Huei Lee* Sergey Levine* Yao Lu* Linda Luu* Carolina Parada*
Peter Pastor* Jornell Quiambao* Kanishka Rao* Jarek Rettinghouse* Diego Reyes* Pierre Sermanet* Nicolas Sievers*
Clayton Tan* Alexander Toshev* Vincent Vanhoucke* Fei Xia* Ted Xiao* Peng Xu* Sichun Xu* Mengyuan Yan* Andy Zeng*



Robotics at Google

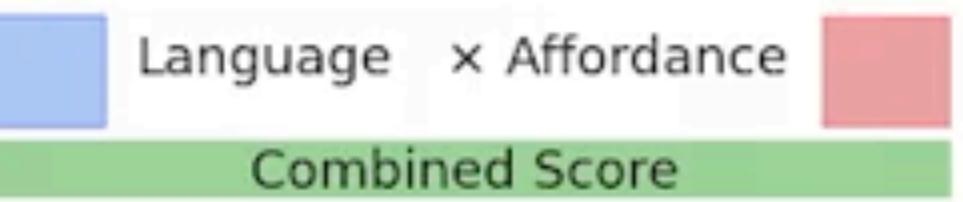


Everyday Robots

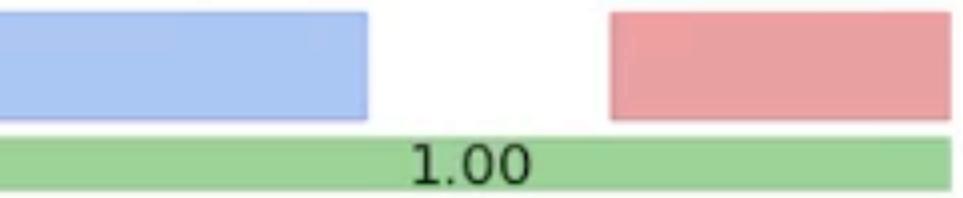
10x speed



I spilled my coke on the table, how would you throw it away and bring me something to help clean?
Robot: I would: 1. find a coke can, 2. __



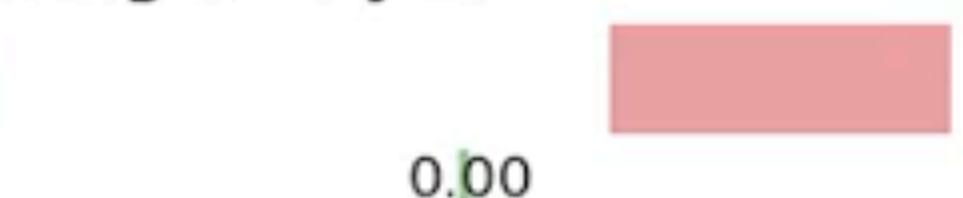
pick up the coke can



put down the coke can



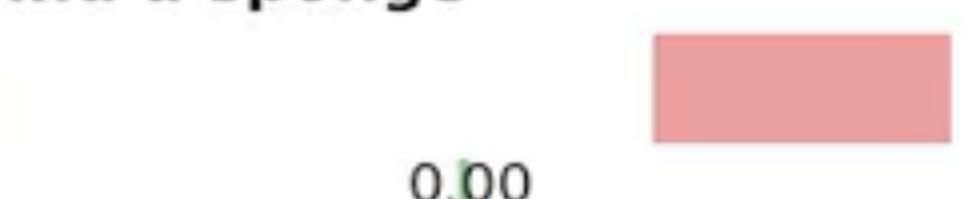
bring it to you



go to the trash can



find a sponge



So ... we just ask an
LLM to tell us what to
do?



No! LLMs can say *anything* ..

I spilled my drink, can you help?

GPT3

You could try using a
vacuum cleaner.

LaMDA

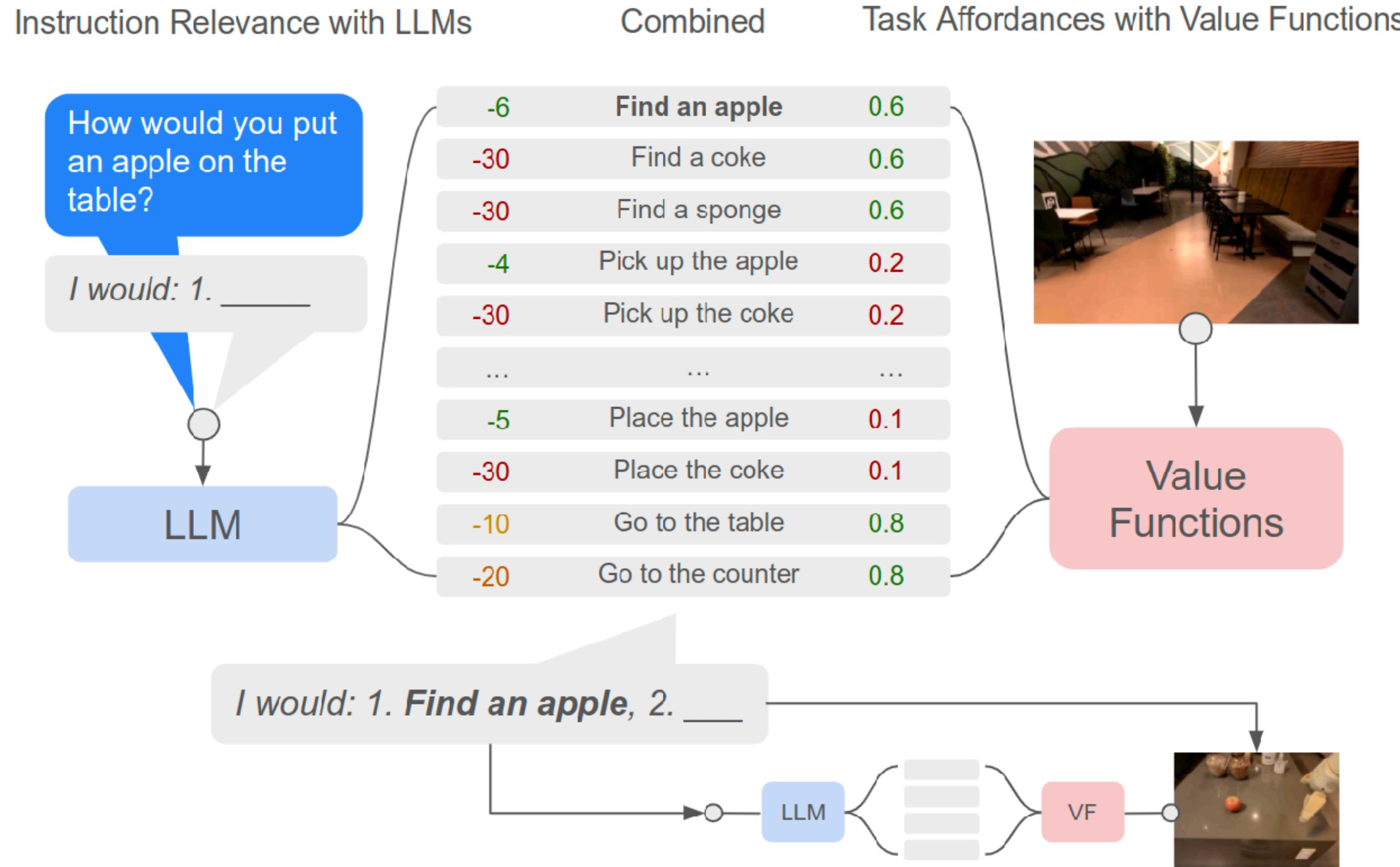
Do you want me to find
a cleaner?

FLAN

I'm sorry, I didn't mean
to spill it.

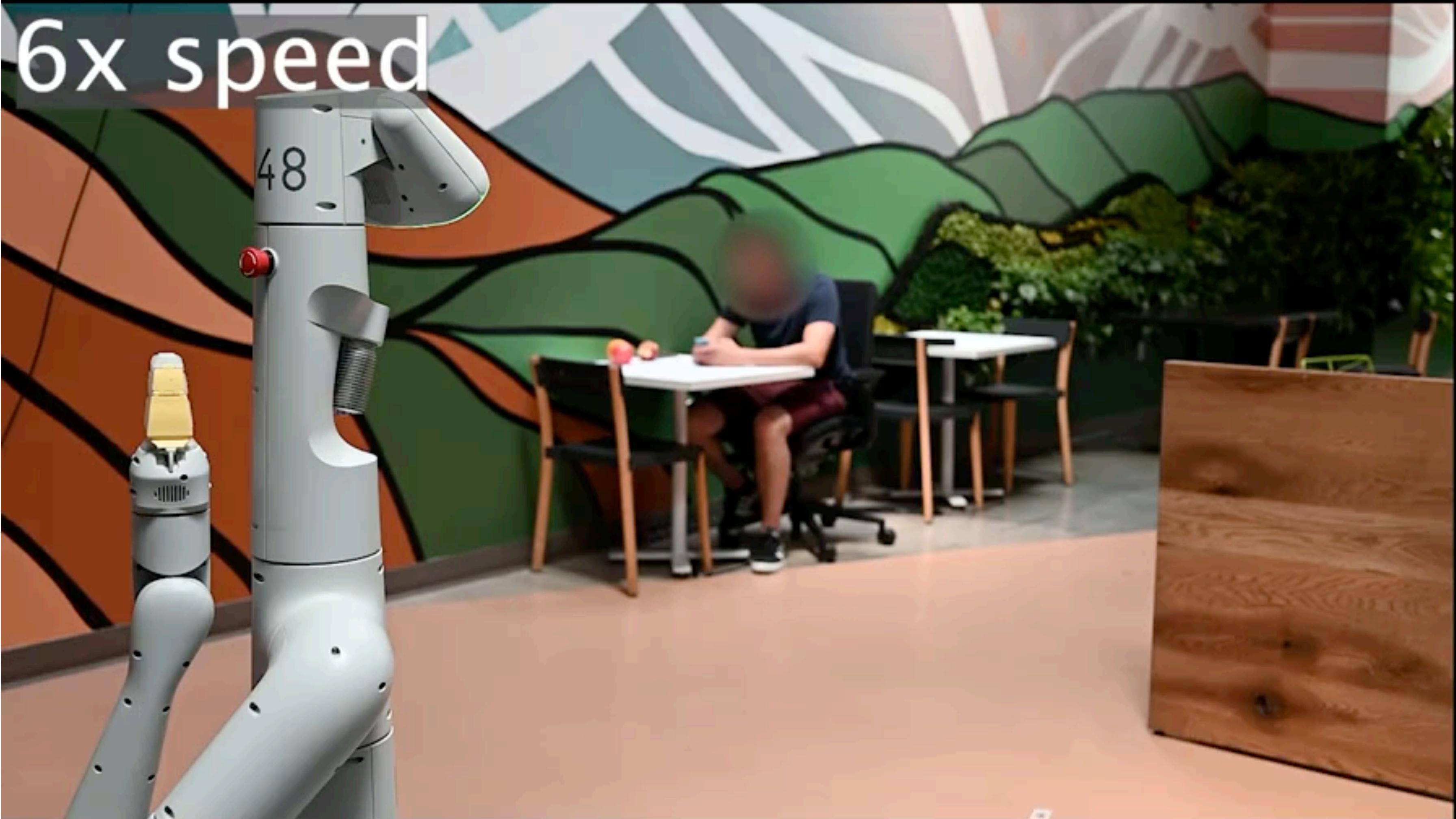
Idea: Constrain LLM by what the robot can do
(affordance)

The “SayCan” Approach



10x speed





User input: Bring me a fruit flavoured drink without caffeine.

Robot: 1.



Can LLMs predict
robot code?

Code as Policies:

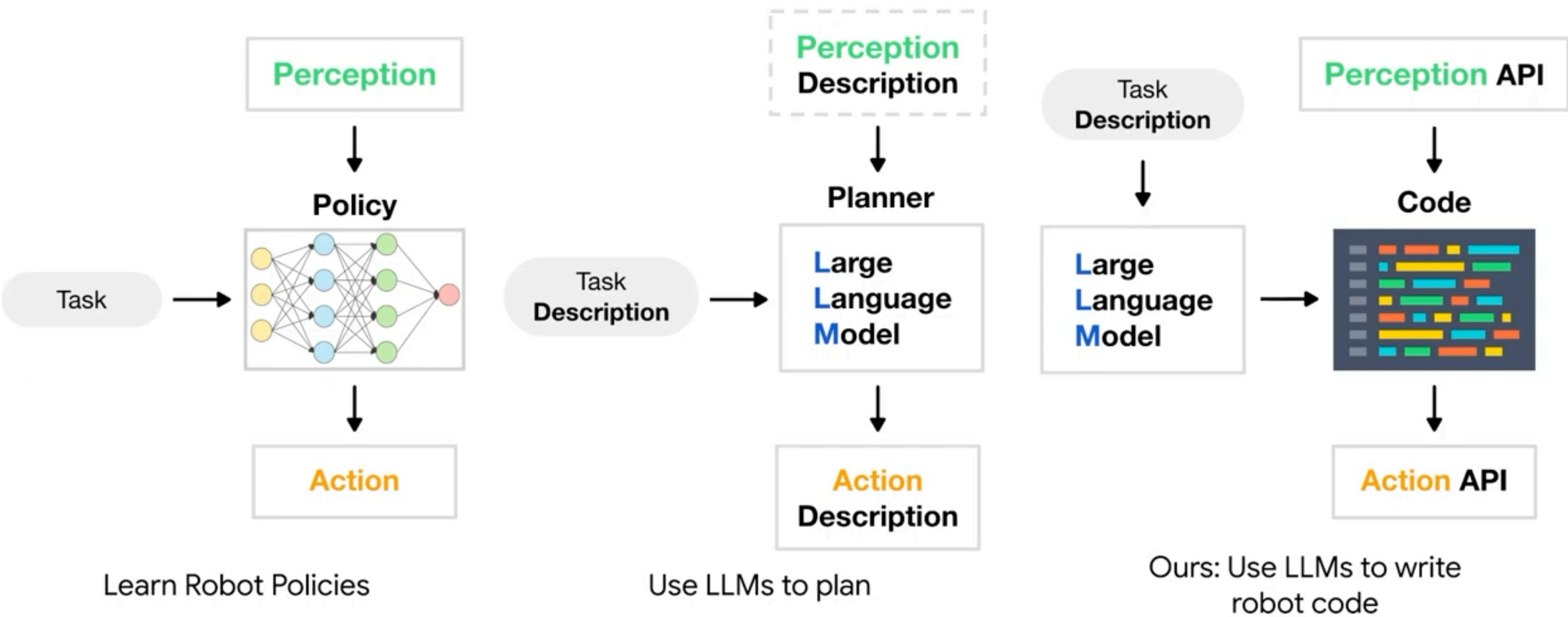
Language Model Programs for Embodied Control

Jacky Liang Wenlong Huang Fei Xia Peng Xu Karol Hausman Brian Ichter Pete Florence Andy Zeng



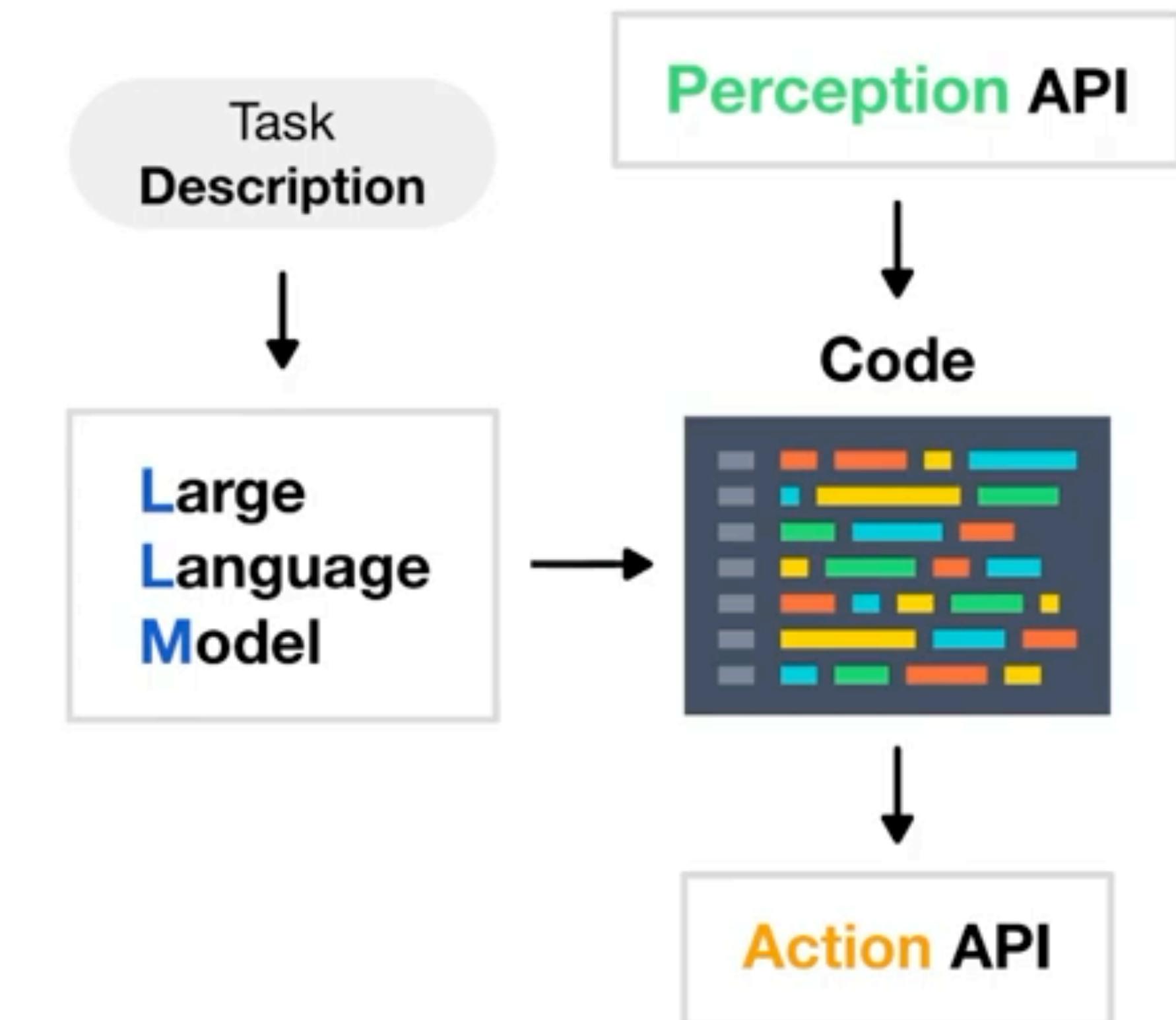
Robotics at Google

Different policy representations



Why choose code as a representation?

Interpretable
Verifiable
Composable



Ours: Use LLMs to write
robot code

User



Large Language Model



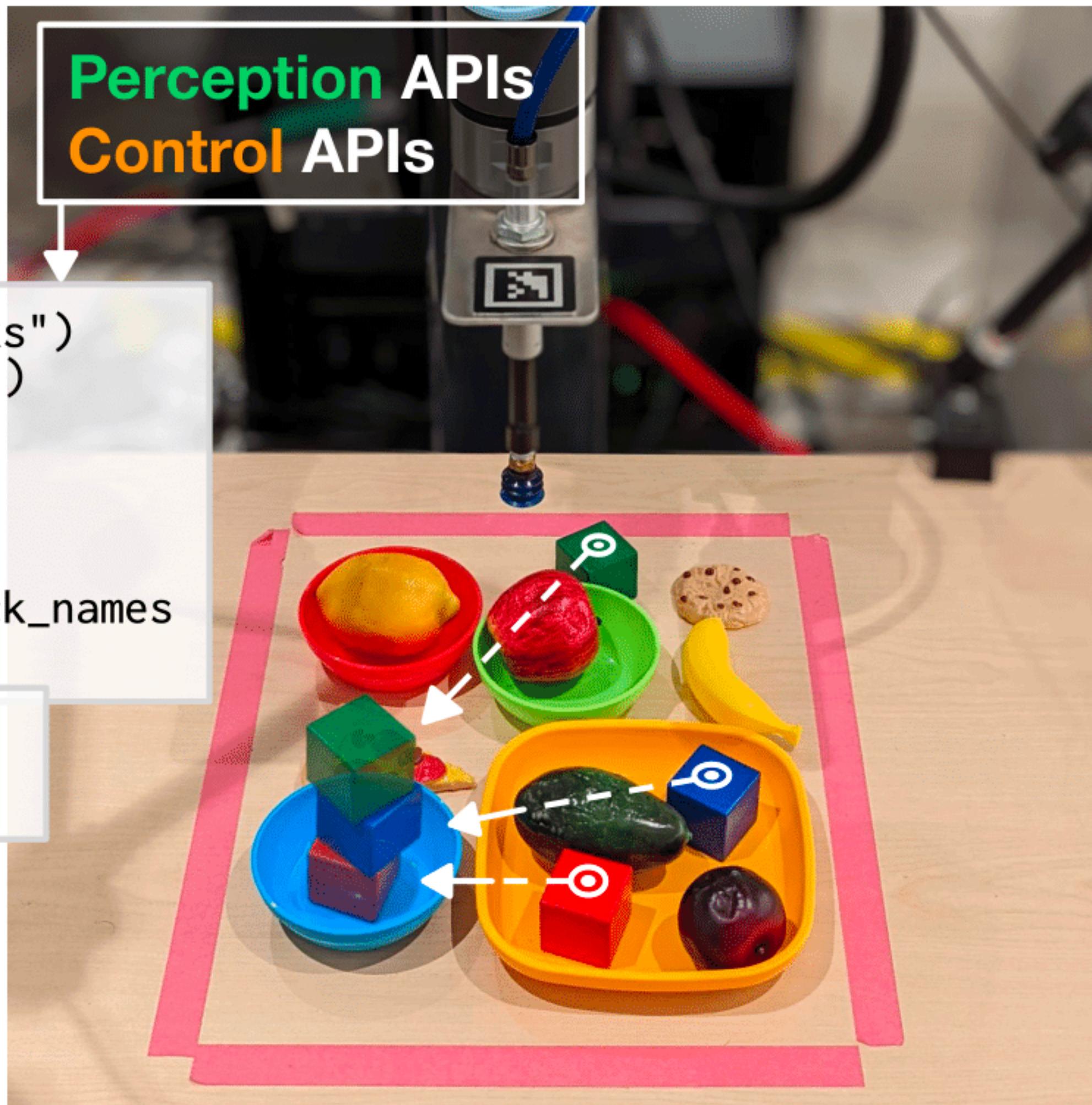
Stack the blocks on the empty bowl.

Policy Code

```
block_names = detect_objects("blocks")
bowl_names = detect_objects("bowls")
for bowl_name in bowl_names:
    if is_empty(bowl_name):
        empty_bowl = bowl_name
        break
objs_to_stack = [empty_bowl] + block_names
stack_objects(objs_to_stack)

def is_empty(name):

def stack_objects(obj_names):
    n_objs = len(obj_names)
    for i in range(n_objs - 1):
        obj0 = obj_names[i + 1]
        obj1 = obj_names[i]
        pick_place(obj0, obj1)
```



Simple code generation examples

```
# if you see an orange, move backwards.  
if detect_object("orange"):  
    robot.set_velocity(x=-0.1, y=0, z=0)  
# move rightwards until you see the apple.  
while not detect_object("apple"):  
    robot.set_velocity(x=0, y=0.1, z=0)
```

```
# do it again but faster, to the left, and with a banana.  
while not detect_object("banana"):  
    robot.set_velocity(x=0, y=-0.2, z=0)
```

How do we prompt LLMs to generate robot code?

1. Instructions

You are an AI assistant writing robot code given natural language instructions. Please refer to the following API guidelines ...

2. Import Hints

```
import numpy as np  
from utils import get_obj_names, put_first_on_second
```

3. Few-shot Examples

Example: Using imported functions

```
from utils import get_pos, put_first_on_second  
...  
# move the purple bowl toward the left.  
target_pos = get_pos('purple bowl') + [-0.3, 0]  
put_first_on_second('purple bowl', target_pos)  
objs = ['blue bowl', 'red block', 'red bowl', 'blue block']  
# move the red block a bit to the right.  
target_pos = get_pos('red block') + [0.1, 0]  
put_first_on_second('red block', target_pos)  
# put the blue block on the bowl with the same color.  
put_first_on_second('blue block', 'blue bowl')
```

Example: Using control flows

```
# while the red block is to the left of the blue bowl, move it to the  
right 5cm at a time.
```

```
while get_pos('red block')[0] < get_pos('blue bowl')[0]:  
    target_pos = get_pos('red block') + [0.05, 0]  
    put_first_on_second('red block', target_pos)
```

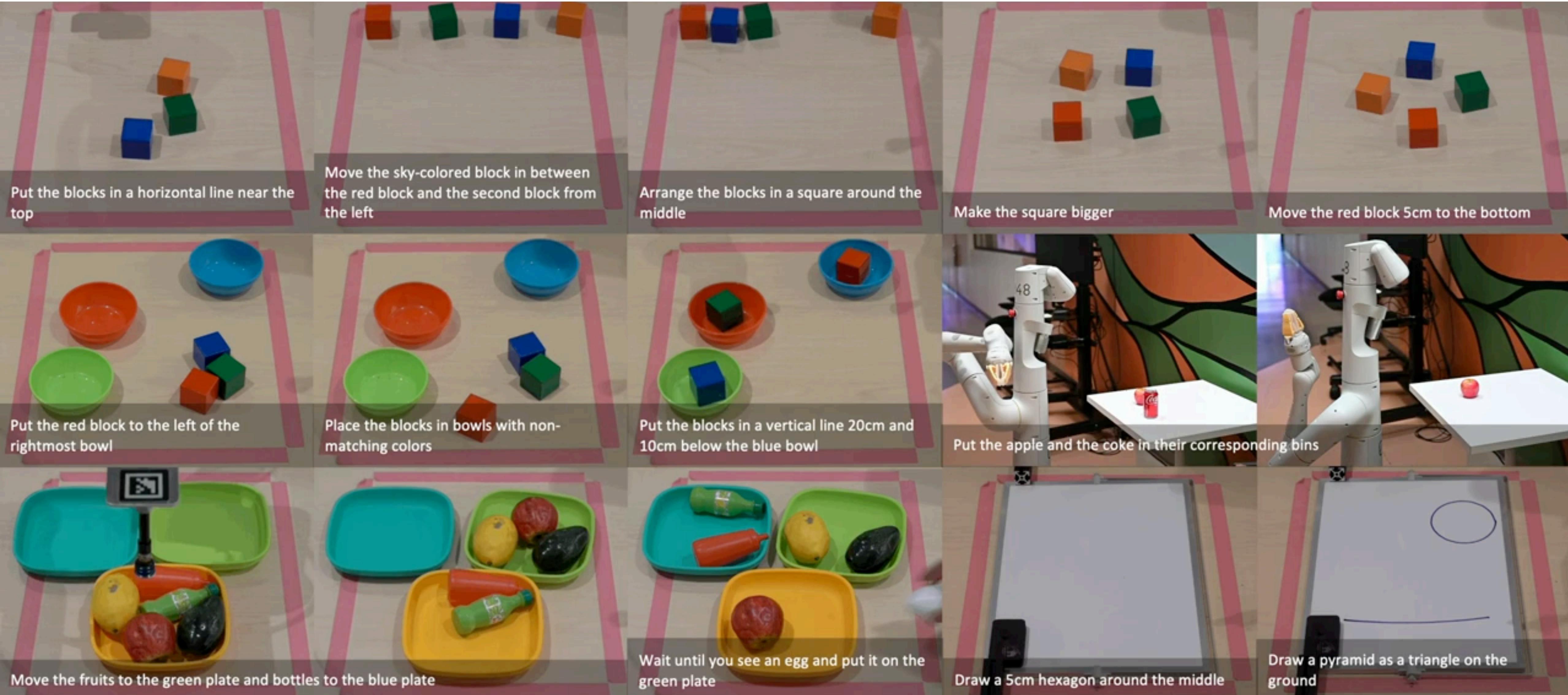
Example: Hierarchical Code Generation

```
import numpy as np
from utils import get_obj_bbox_xyxy
# define function: total = get_total(xs).
def get_total(xs):
    return np.sum(xs)
# define function: get_objs_bigger_than_area_th(obj_names, bbox_area_th).
def get_objs_bigger_than_area_th(obj_names, bbox_area_th):
    return [name for name in obj_names
            if get_obj_bbox_area(name) > bbox_area_th]
```

Have the LLM recursively define functions!

```
# define function: get_obj_bbox_area(obj_name).
def get_obj_bbox_area(obj_name):
    x1, y1, x2, y2 = get_obj_bbox_xyxy(obj_name)
    return (x2 - x1) * (y2 - y1)
```

Verifiably solve a number of tasks!



Can LLMs convert
demonstrations (non-language)
to code?

Demo2Code: From Summarizing Demonstrations to Synthesizing Code via Extended Chain-of-Thought

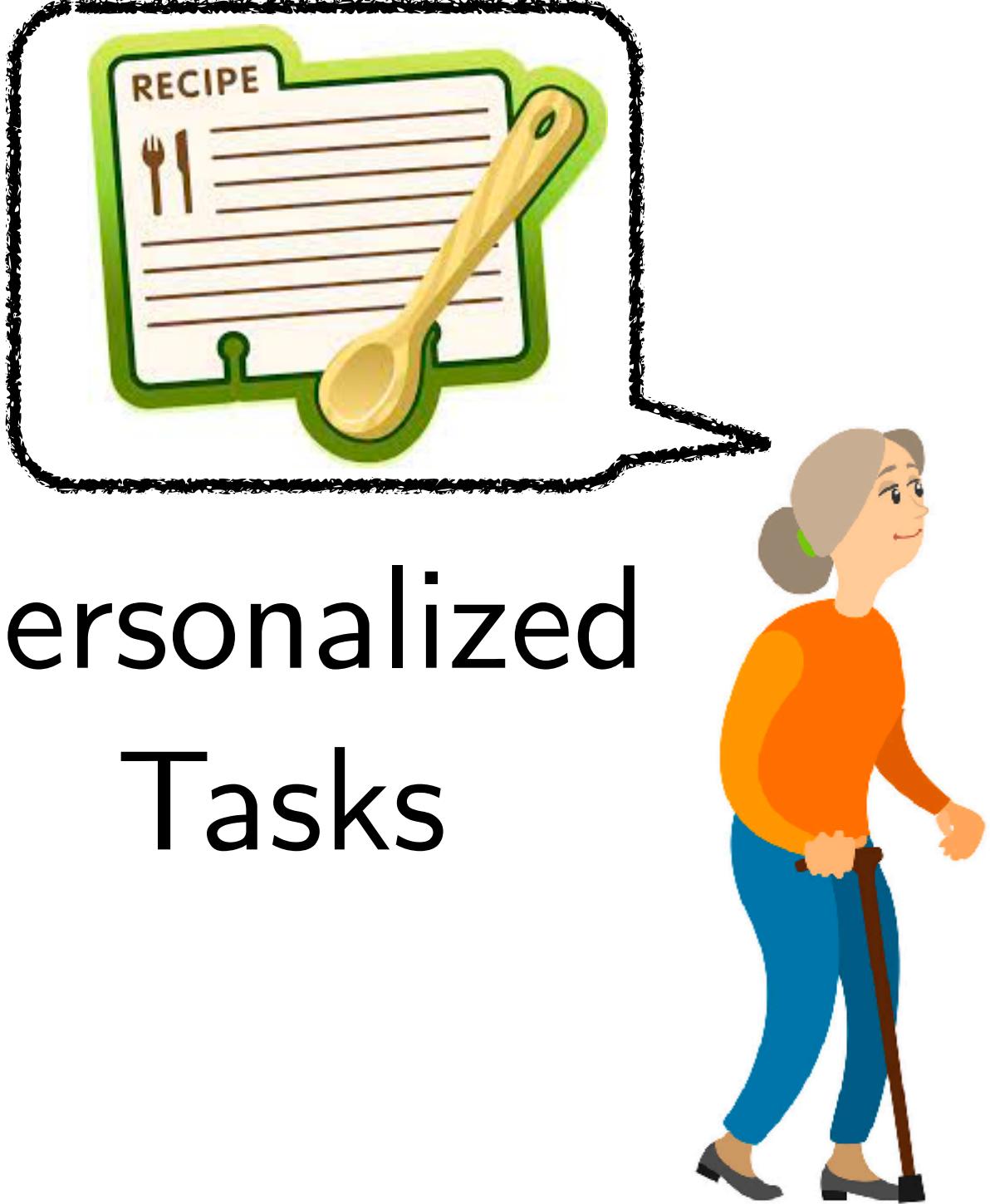
NeurIPS 2023

Huaxiaoyue Wang, Gonzalo Gonzalez-Pumariega, Yash Sharma, Sanjiban Choudhury

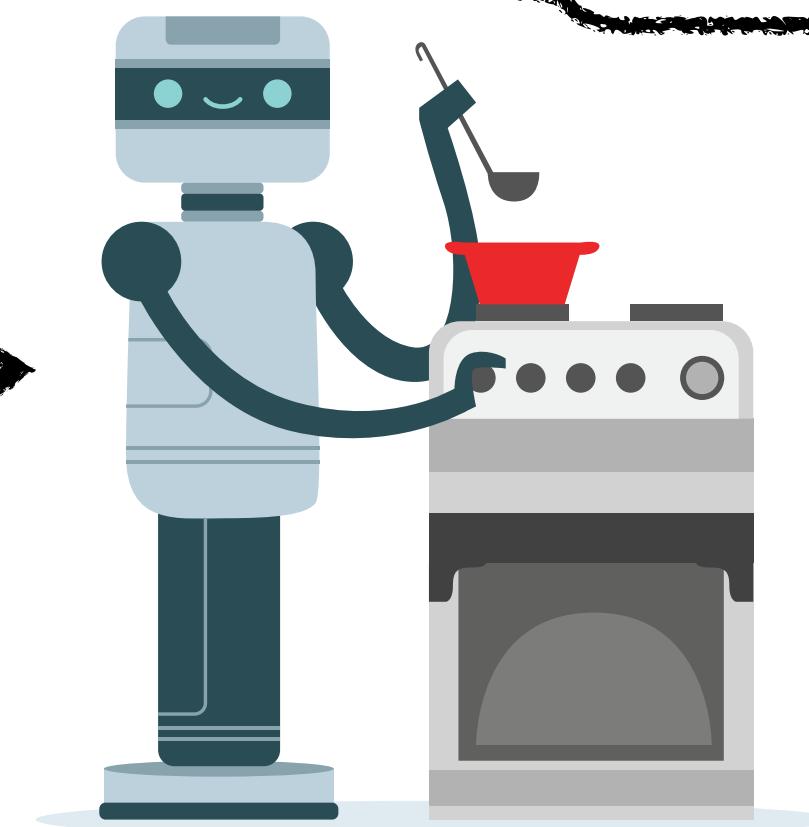
Cornell University

How can we teach robots *personalized* tasks?

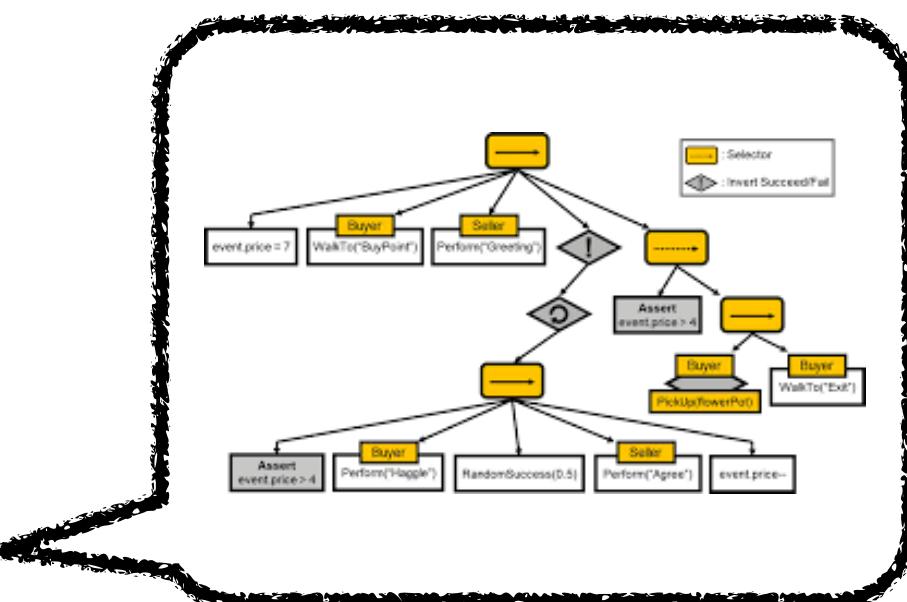
Personalized
Tasks



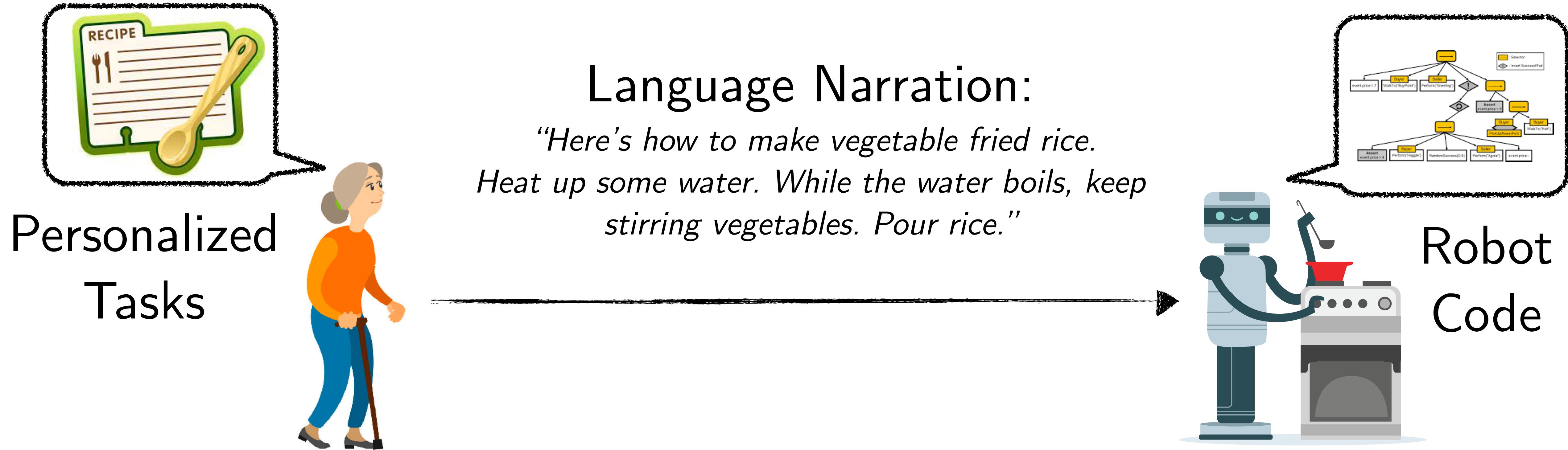
Language Narration:
*"Here's how to make vegetable fried rice.
Heat up some water. While the water boils, keep
stirring vegetables. Pour rice."*



Robot
Code



How can we teach robots *personalized* tasks?



Language alone is insufficient to communicate the task

✗ Lacks specificity (e.g. Heat up water how? Pour rice where?)

✗ Leaves out implicit preferences (e.g. Personal style of stirring?)

How can we teach robots *personalized* tasks?



Personalized
Tasks

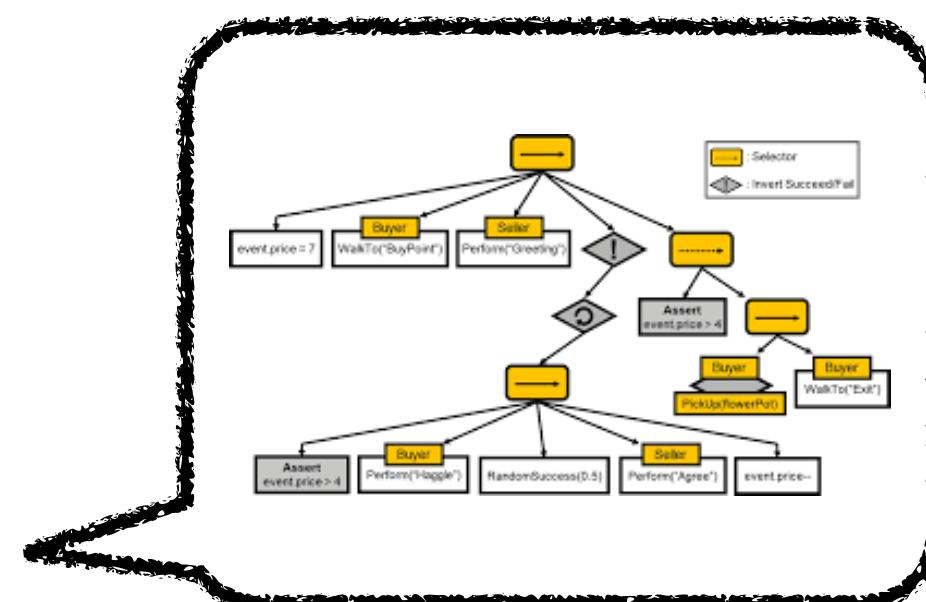


Demonstrations
convey dense
information on how
states change

Language Narration:

"Here's how to make vegetable fried rice."

*Heat up some water. While the water boils, keep
stirring vegetables. Pour rice."*



Robot
Code

Demonstrations:



over('kettle',
'left_pan')

in('spatula',
'hand')

over('rice',
'left_pan')

Language:

"Here's how to make vegetable fried rice.

*Heat up some water. While the water
boils, keep stirring vegetables. Pour rice."*

+

Demonstrations

(Sequence of states
represented as text)



s_1

s_2

s_3

```
over('kettle',  in('spatula',  over('rice',  
'left_pan'))      'hand')        'left_pan')
```



Robot
Code

Challenges

Challenge 1: Long-horizons



state₁

state₂

⋮

state_T



state₁

state₂

⋮

state_T

.....

Each demonstration
 \geq hundreds of
states.

Multiple such
demonstrations.

Challenge 2: Complex Task Code

```
def main():
    # Get a list of all the lettuces in the kitchen.
    lettuces = get_all_obj_names_that_match_type('lettuce')
    # Get a list of all the bottom buns in the kitchen.
    bottom_buns = get_all_obj_names_that_match_type('bottom_bun')
    # Get a list of all the patties in the kitchen.
    patties = get_all_obj_names_that_match_type('patty')
    # Get a list of all the top buns in the kitchen.
    top_buns = get_all_obj_names_that_match_type('top_bun')

    # Decide a lettuce to use.
    lettuce_to_use = lettuces[0]
    # Get a list of all the available cutting boards in the kitchen.
    cutting_boards = get_all_location_names_that_match_type('cutting_board')
    # Decide a cutting board to go to.
    cutting_board_to_cut_at = cutting_boards[0]
    # Cut that lettuce at that cutting board.
    cut_object_at_location(obj=lettuce_to_use, location=cutting_board_to_cut_at)

    # Decide a bottom bun to use.
    bottom_bun_to_use = bottom_buns[0]
    # Stack the lettuce on top of the bottom bun.
    # obj1 should be the lettuce, obj2 should be the bottom bun.
    stack_obj1_on_obj2(obj1=lettuce_to_use, obj2=bottom_bun_to_use)

    # Decide a patty to use.
    patty_to_use = patties[0]
    # Get a list of all the available stoves in the kitchen.
    stoves = get_all_location_names_that_match_type('stove')
    # Decide a stove to go to.
    stove_to_cook_at = stoves[0]
    # Cook that patty at that stove.
    cook_object_at_location(obj=patty_to_use, location=stove_to_cook_at)

    # Stack the patty on top of the lettuce.
    # obj1 should be the patty, obj2 should be the lettuce.
    stack_obj1_on_obj2(obj1=patty_to_use, obj2=lettuce_to_use)

    # Decide a top bun to use.
    top_bun_to_use = top_buns[0]
    # Stack the top bun on top of the patty.
    # obj1 should be the top bun, obj2 should be the patty.
    stack_obj1_on_obj2(obj1=top_bun_to_use, obj2=patty_to_use)
```

```
def cook_object_at_location(obj, location):
    # To cook an object, the robot first needs to be holding obj
    if not is_holding(obj):
        # If the robot is not holding obj, there are 2 scenarios:
        # (1) if obj is in a stack ,unstack obj
        # (2) else, pick up obj.
        if is_in_a_stack(obj):
            # Because obj is in a stack, robot need to move then unstack the obj
            obj_at_bottom = get_obj_that_is_underneath(obj_at_top=obj)
            # move then unstack: first you move to the obj_at_bottom's location
            move_then_unstack(obj_to_unstack=obj, obj_at_bottom=obj_at_bottom, unstack_location=location)
        else:
            # move_then_pick: first you move to obj's location, then you pick obj
            move_then_pick(obj=obj, pick_location=get_obj_location(obj))
    # move then place: first you move to the location to cook at, then you place obj
    move_then_place(obj=obj, place_location=location)
    # cook the object until it is cooked
    cook_until_is_cooked(obj=obj)

def stack_obj1_on_obj2(obj1, obj2):
    # To stack obj1 on obj2, the robot needs to be holding obj1
    if not is_holding(obj1):
        # If the robot is not holding obj1, there are 2 scenarios:
        # (1) if obj1 is in a stack ,unstack obj1
        # (2) else, pick up obj1.
        if is_in_a_stack(obj1):
            # Because obj1 is in a stack, robot need to move then unstack the obj1
            obj_at_bottom = get_obj_that_is_underneath(obj_at_top=obj1)
            # move then unstack: first you move to the obj_at_bottom's location, then you unstack obj from obj_at_bottom
            move_then_unstack(obj_to_unstack=obj1, obj_at_bottom=obj_at_bottom, unstack_location=get_obj_location(obj_at_bottom))
        else:
            # move_then_pick: first you move to obj's location, then you pick obj up
            move_then_pick(obj=obj1, pick_location=get_obj_location(obj1))
    # determine the location of obj2 to stack on
    obj2_location = get_obj_location(obj2)
    # move then stack: first you move to obj2's location, then you stack obj1 on obj2
    move_then_stack(obj_to_stack=obj1, obj_at_bottom=obj2, stack_location=obj2_location)

def cut_object_at_location(obj, location):
    # To cut an object, the robot first needs to be holding obj
    if not is_holding(obj):
        # If the robot is not holding obj, there are 2 scenarios:
        # (1) if obj is in a stack ,unstack obj
```

```
def move_then_unstack(obj_to_unstack, obj_at_bottom, unstack_location):
    if get_curr_location() != unstack_location:
        move(get_curr_location(), unstack_location)
    unstack(obj_to_unstack, obj_at_bottom)

def move_then_pick(obj, pick_location):
    if get_curr_location() != pick_location:
        move(get_curr_location(), pick_location)
    pick_up(obj, pick_location)

def move_then_place(obj, place_location):
    if get_curr_location() != place_location:
        move(get_curr_location(), place_location)
    place(obj, place_location)

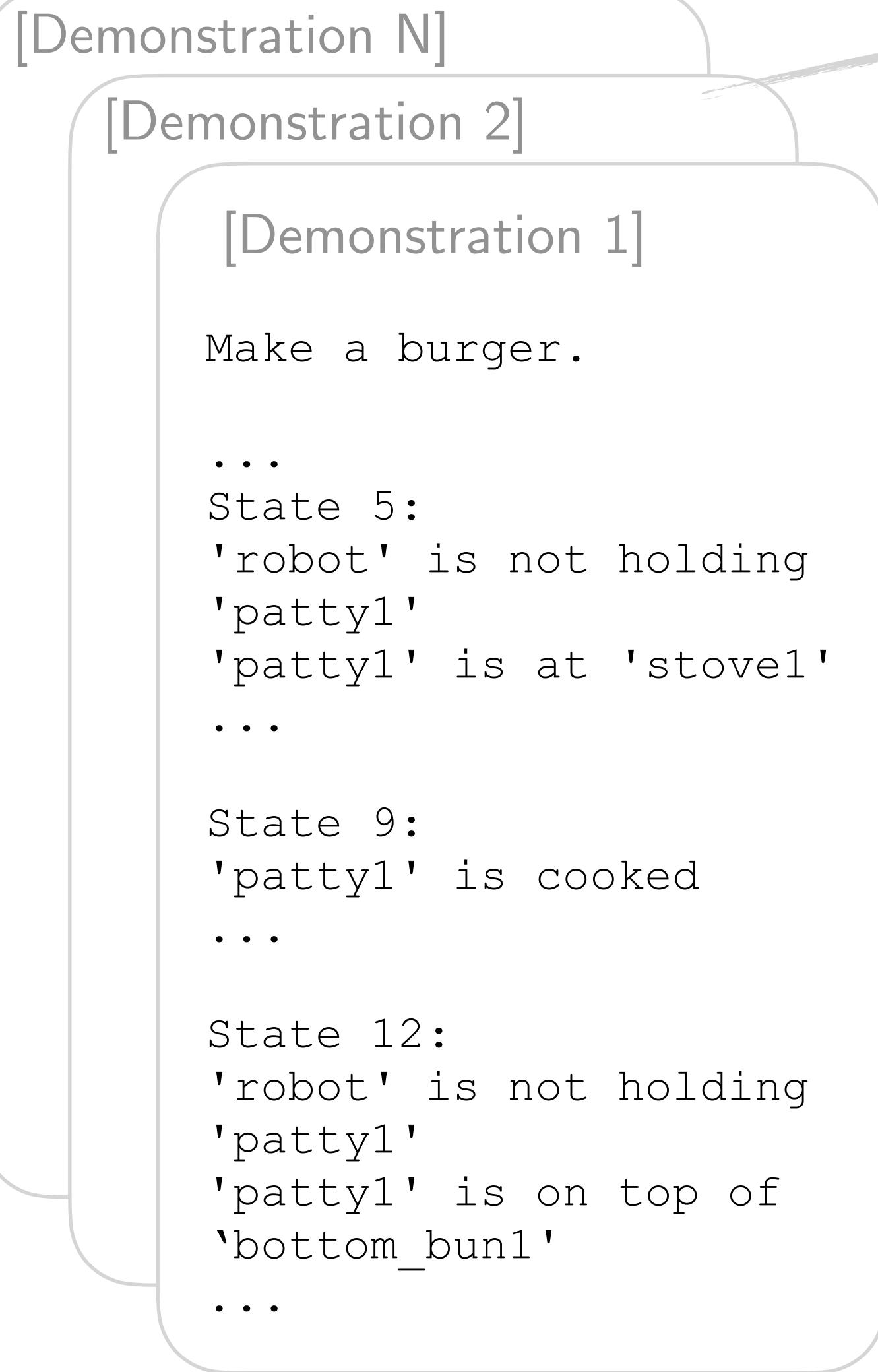
def cook_until_is_cooked(obj):
    start_cooking(obj)
    while not is_cooked(obj):
        noop()

def move_then_stack(obj_to_stack, obj_at_bottom, stack_location):
    if get_curr_location() != stack_location:
        move(get_curr_location(), stack_location)
    stack(obj_to_stack, obj_at_bottom)

def cut_until_is_cut(obj):
    while not is_cut(obj):
        cut(obj)
```

Loops, checks, and calls to custom robot libraries ..

Directly going from demo to code is hard ...



```
# Cook object at location
def cook_object_at_loc(obj,
                      loc):
    if not is_holding(obj):
        ...
    move_then_place(obj, loc)
    cook_until_is_cooked(obj)

# Move to a location and place
# object
def move_then_place(obj, loc):
    curr_loc = get_curr_loc()
    if curr_loc != loc:
        move(curr_loc, loc)
    place(obj, place_location)
    ...

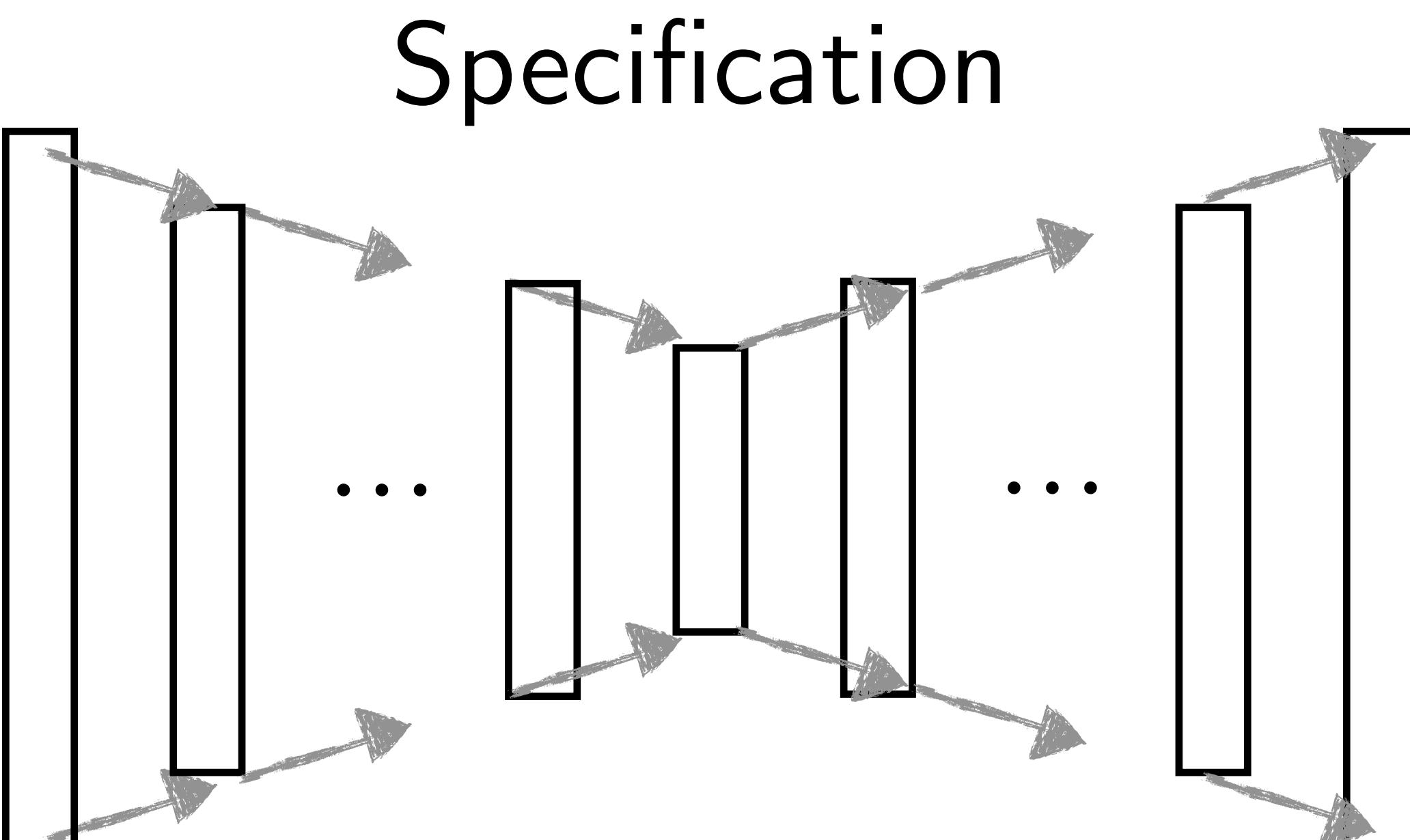
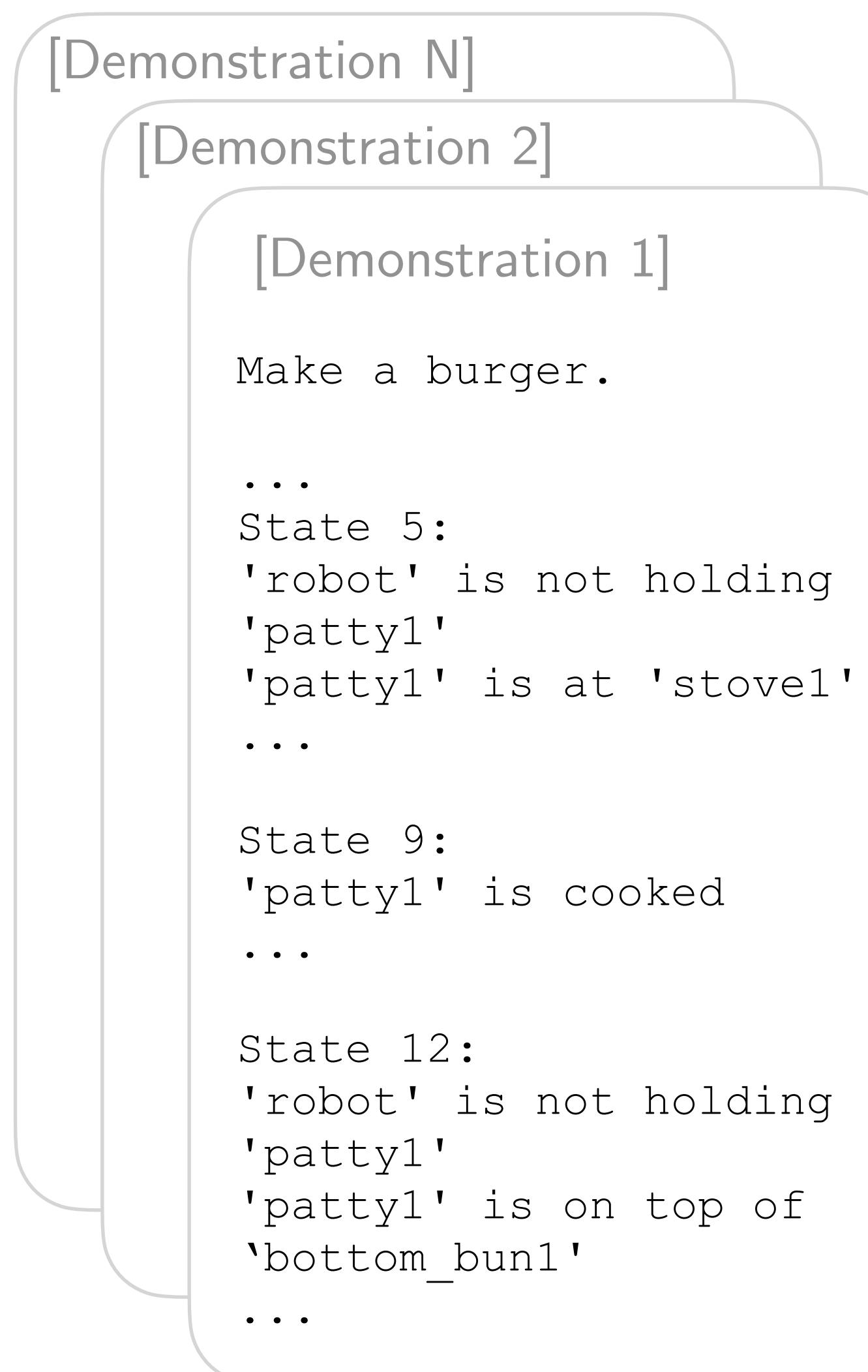
def main():
    ...
    cook_object_at_loc(patty,
                       stove)
    ...
    stack_objects(top_bun,
                  lettuce)
```



Demonstrations can be
rationalized by
a latent, compact
specification

(Like Reward Functions in IRL)

Key Insight: Extended chain-of-thought



Specification

Every step along the chain
is small and easy for LLM

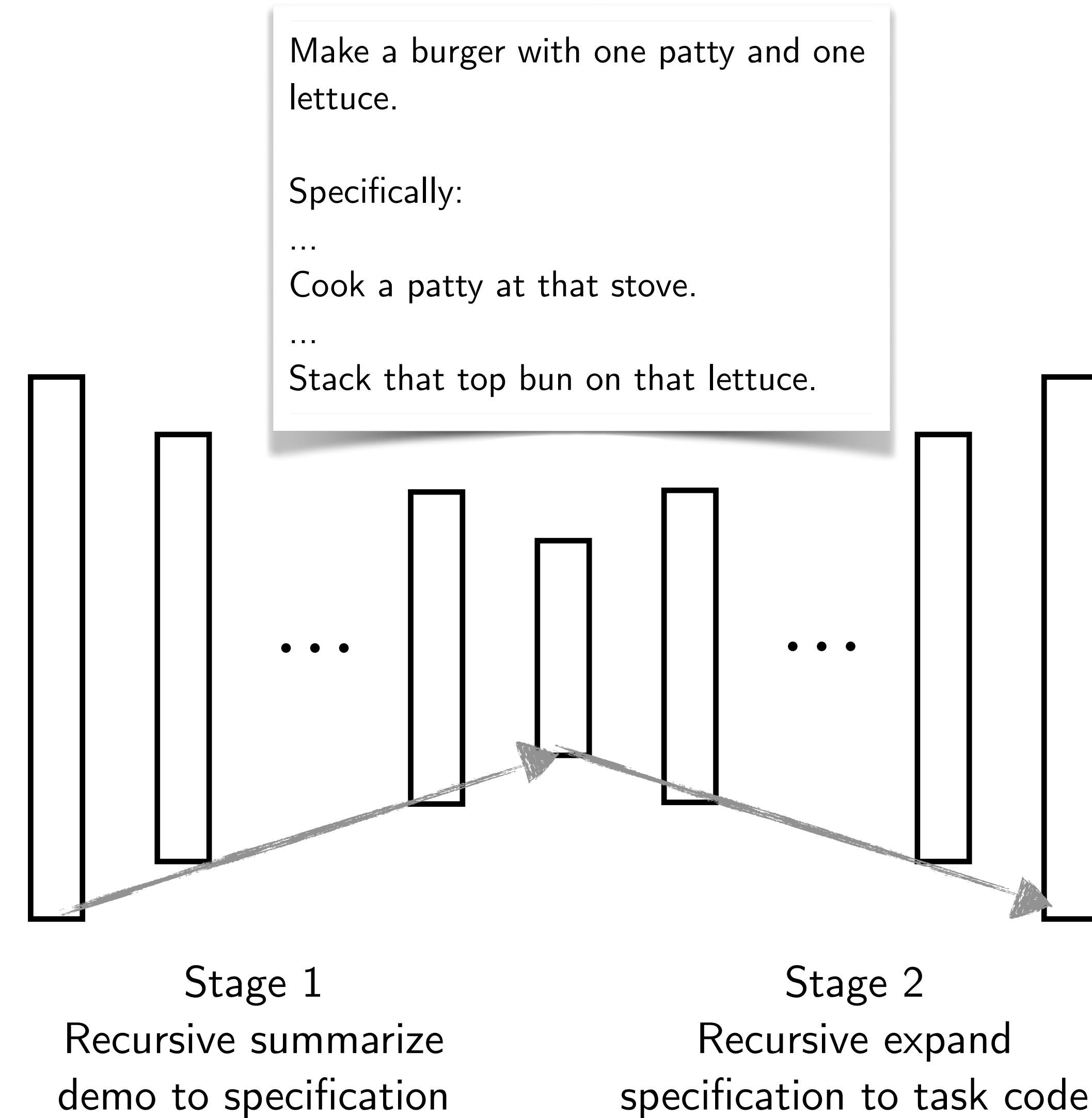
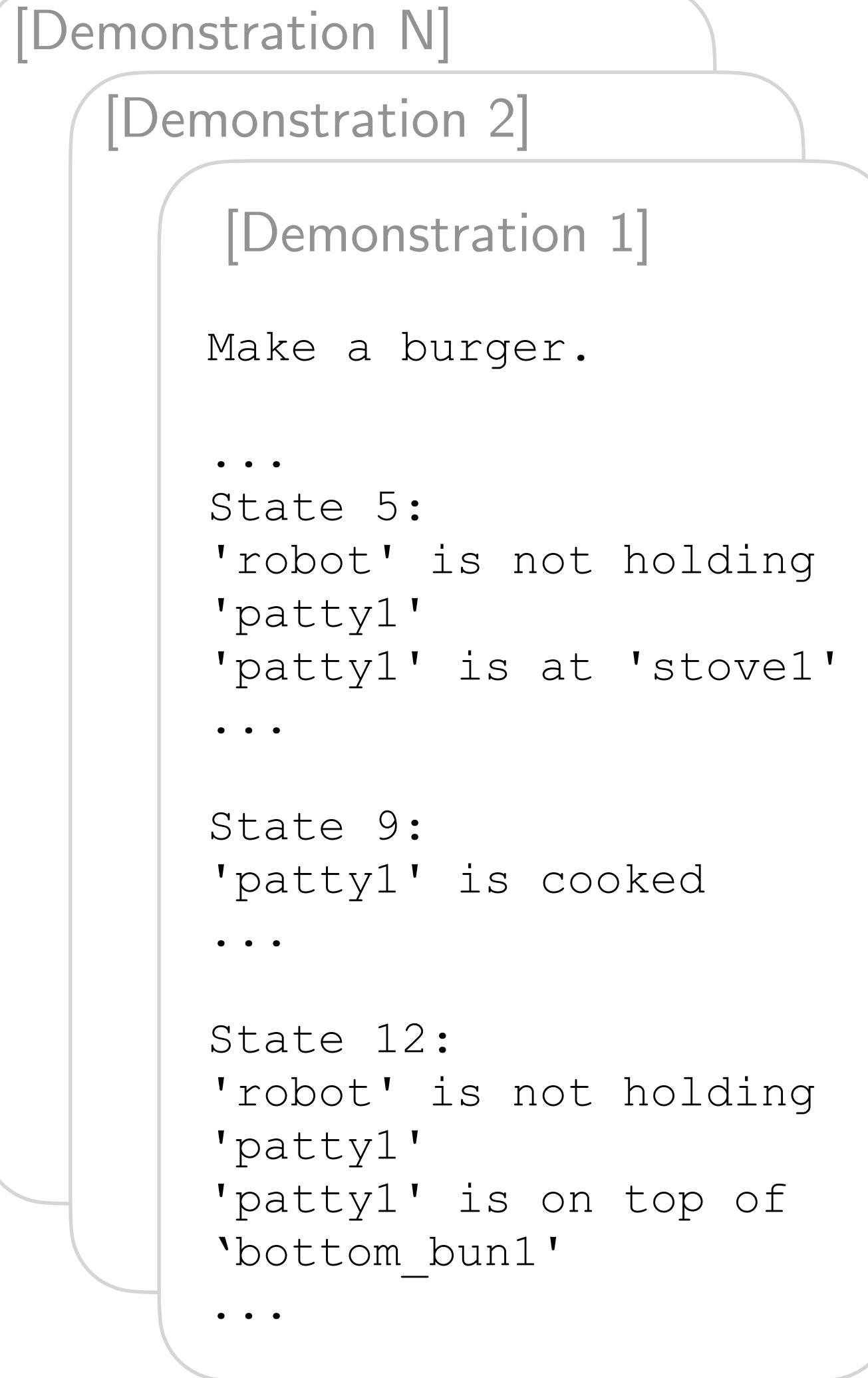
```
# Cook object at location
def cook_object_at_loc(obj,
                      loc):
    if not is_holding(obj):
        ...
    move_then_place(obj, loc)
    cook_until_is_cooked(obj)

# Move to a location and place
# object
def move_then_place(obj, loc):
    curr_loc = get_curr_loc()
    if curr_loc != loc:
        move(curr_loc, loc)
    place(obj, place_location)
    ...

def main():
    ...
    cook_object_at_loc(patty,
                        stove)
    ...
    stack_objects(top_bun,
                  lettuce)
```

Demo2Code

Demo2Code: Recursive Summarization and Expansion



```
# Cook object at location
def cook_object_at_loc(obj, loc):
    if not is_holding(obj):
        ...
    move_then_place(obj, loc)
    cook_until_is_cooked(obj)

# Move to a location and place object
def move_then_place(obj, loc):
    curr_loc = get_curr_loc()
    if curr_loc != loc:
        move(curr_loc, loc)
    place(obj, place_location)
    ...

def main():
    ...
    cook_object_at_loc(patty, stove)
    ...
    stack_objects(top_bun, lettuce)
```

Experiments

We made a new game: Robotouille!



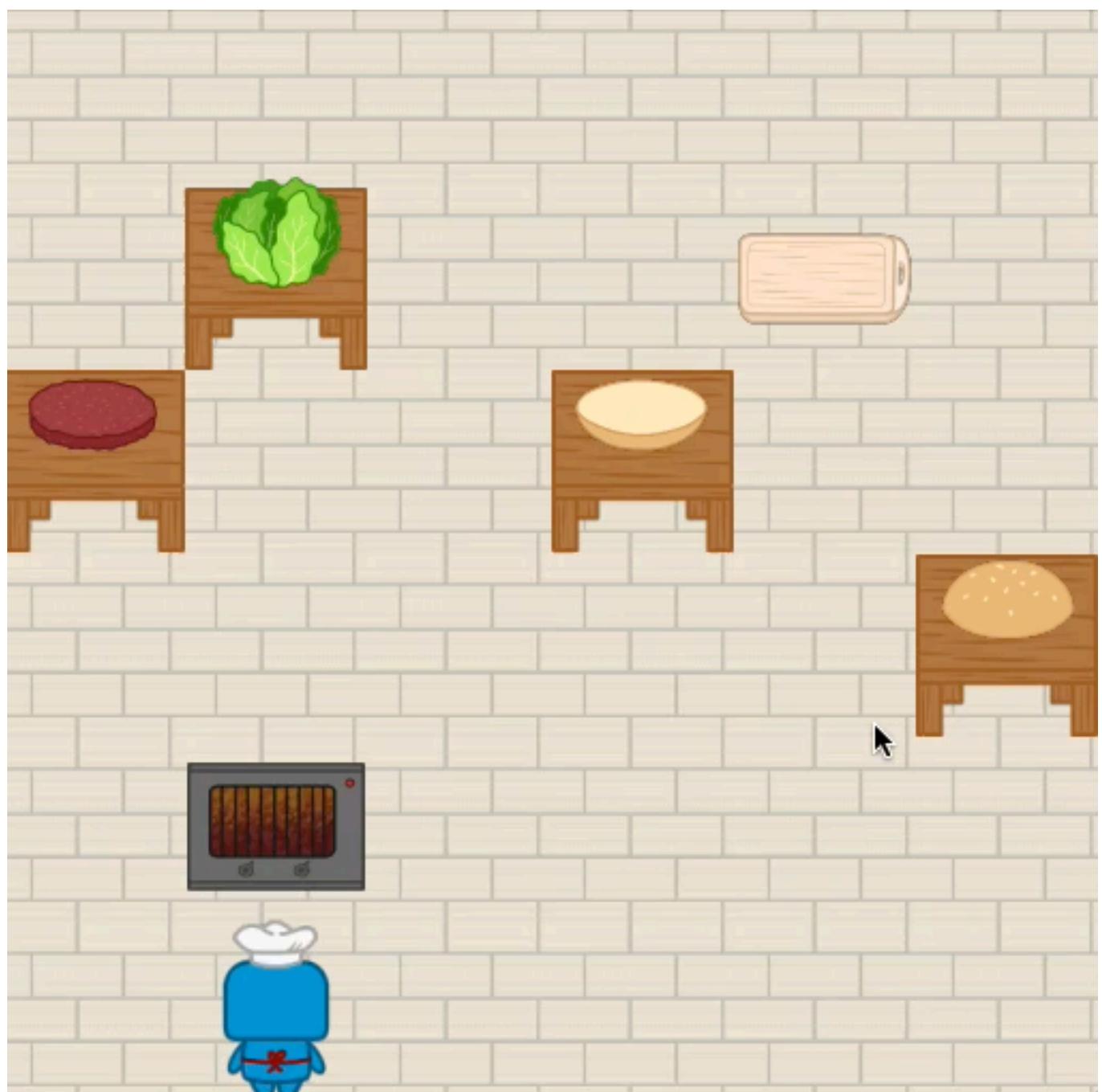
Open-source game coming soon
on iOS / Android

Human runs a restaurant
with robot sous-chef

Fun way to learn how humans
plan and communicate tasks!

Demo2Code can generalize to new, complex environments

Make a burger.



User provides a demonstration in a simple environment

Make a burger.

Cut that lettuce at that cutting board.

Stack the lettuce on top of the bottom bun.

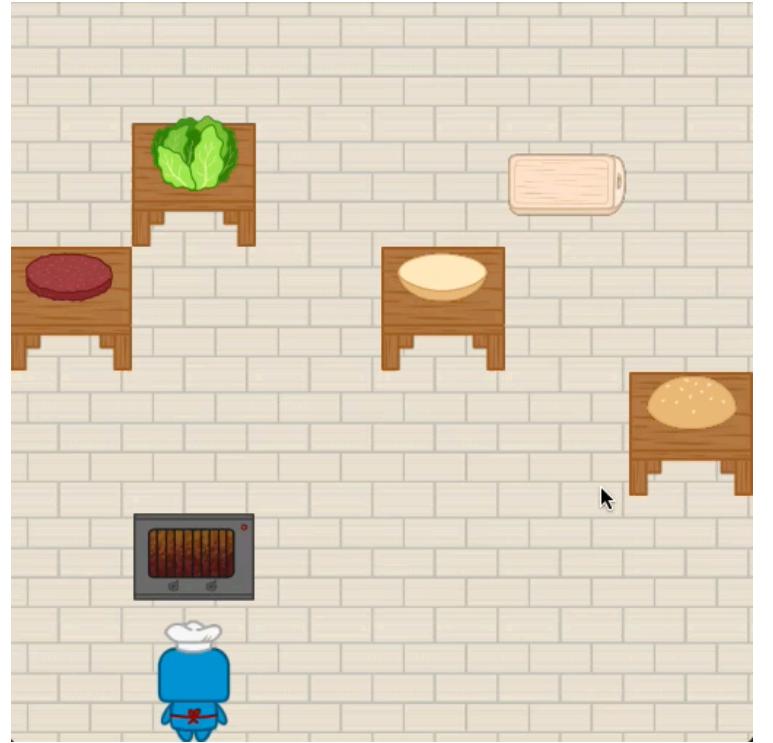
Cook that patty at that stove.

Stack the patty on top of the lettuce.

Stack the top bun on top of the patty.

```
def cook_object_at_loc(obj, loc):  
    if not is_holding(obj):  
        ...  
    move_then_place(obj, loc)  
    cook_until_is_cooked(obj)  
  
def move_then_place(obj, loc):  
    curr_loc = get_curr_loc()  
    if curr_loc != loc:  
        move(curr_loc, loc)  
    place(obj, place_location)  
  
def main():  
    ...  
    cut_object_at_loc(lettuce, stove)  
    ...  
    stack_objects(lettuce, bottom_bun)  
    ...  
    cook_object_at_loc(patty, stove)  
    ...  
    stack_objects(top_bun, patty)
```

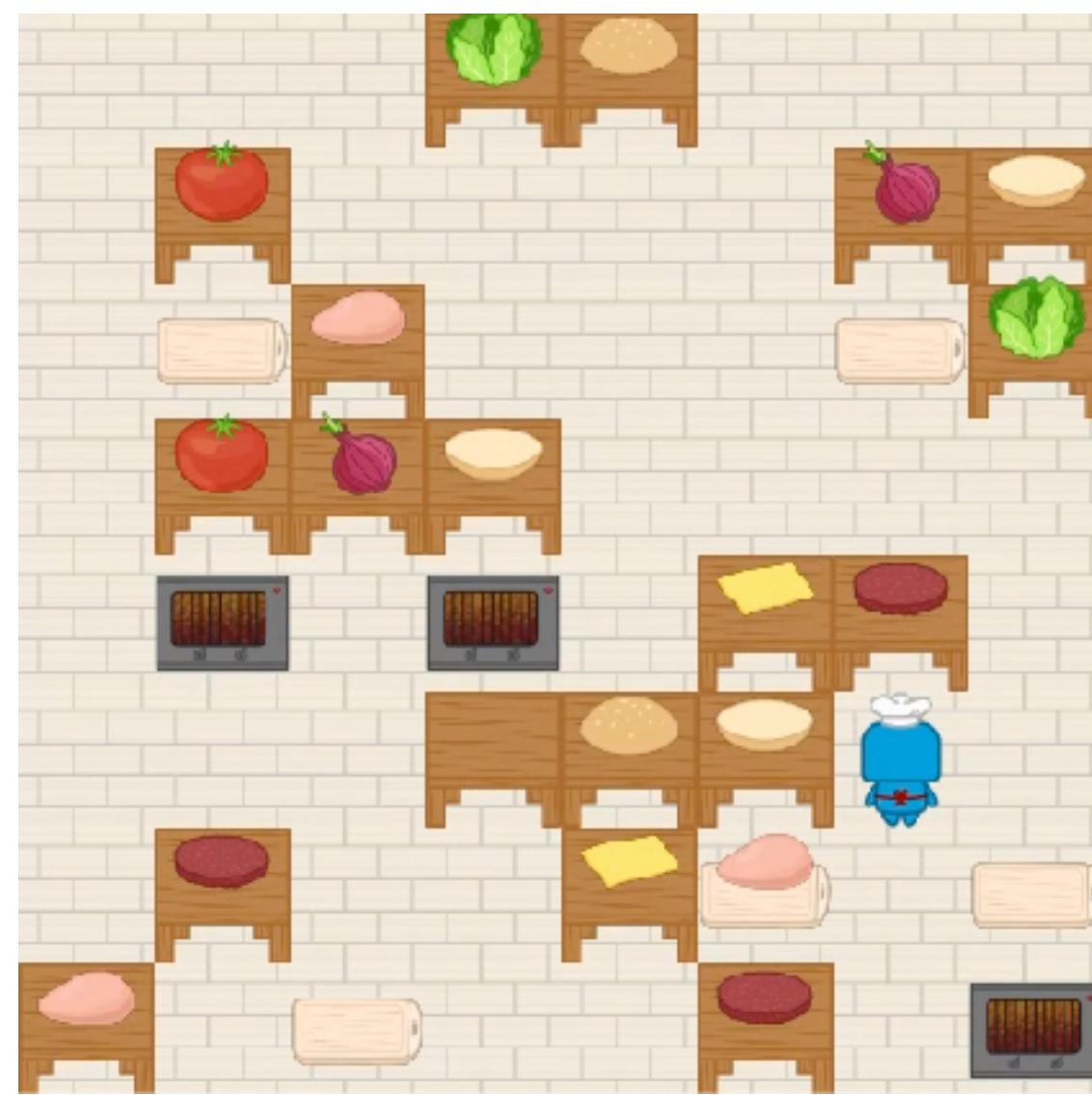
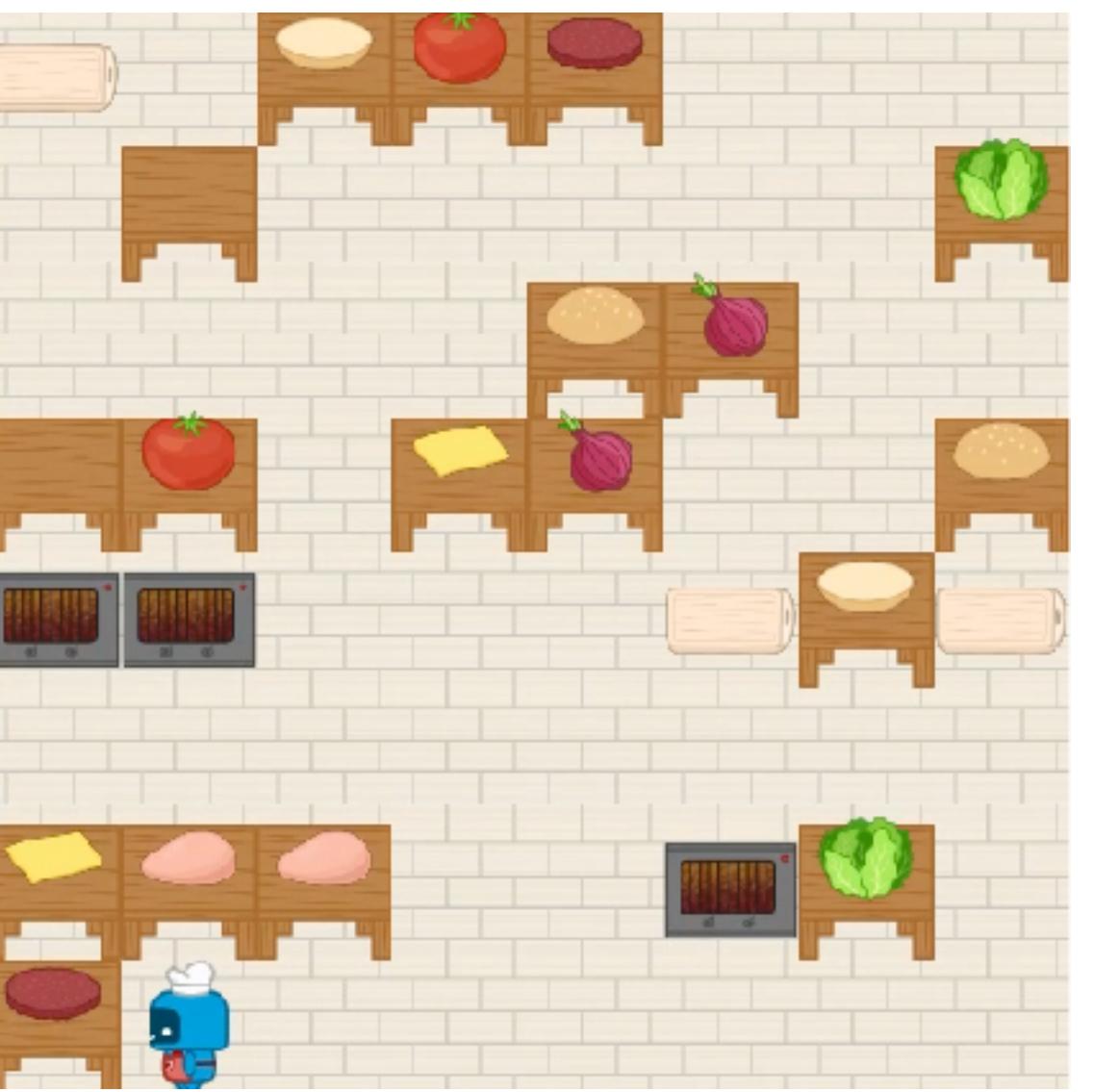
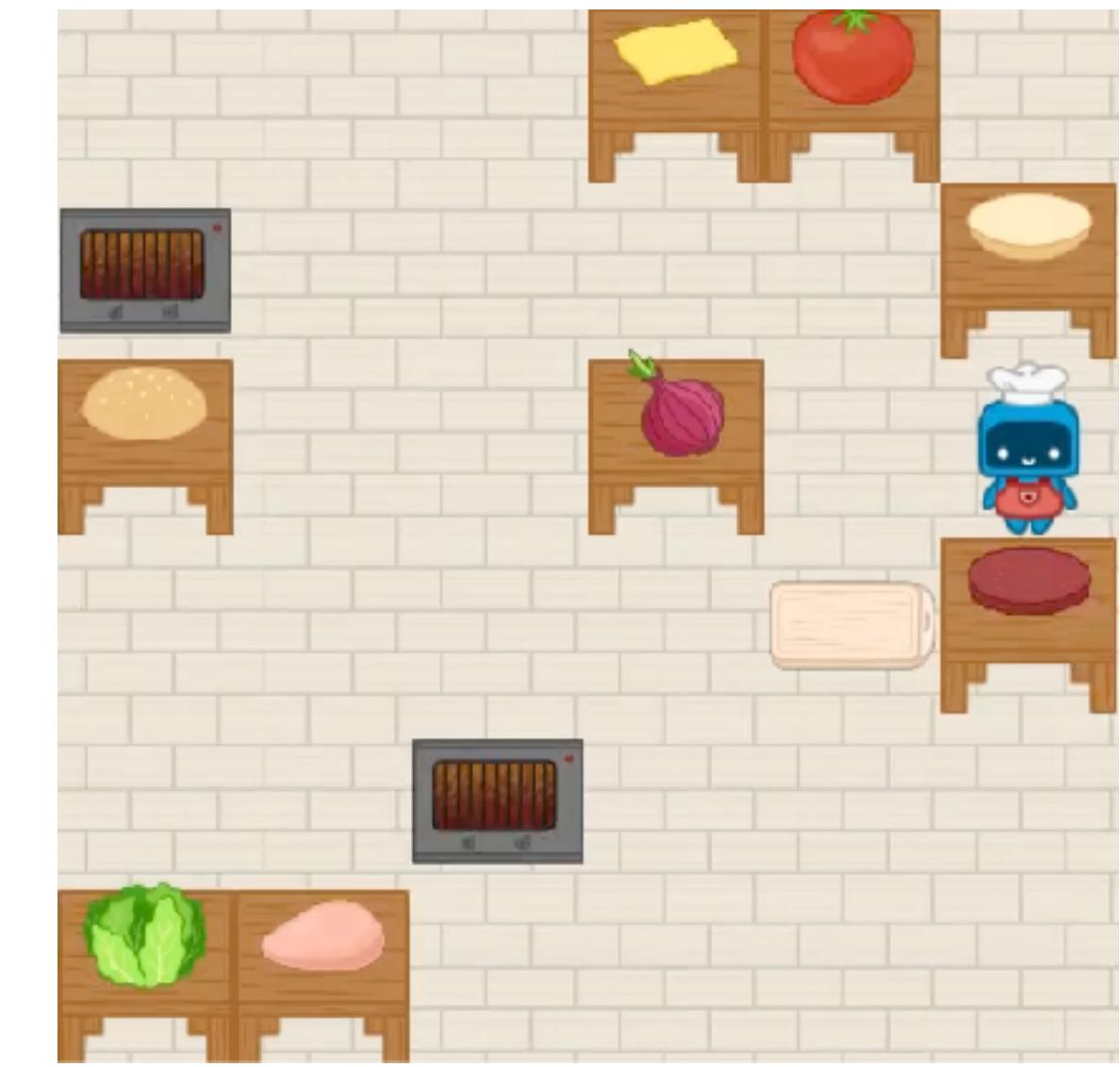
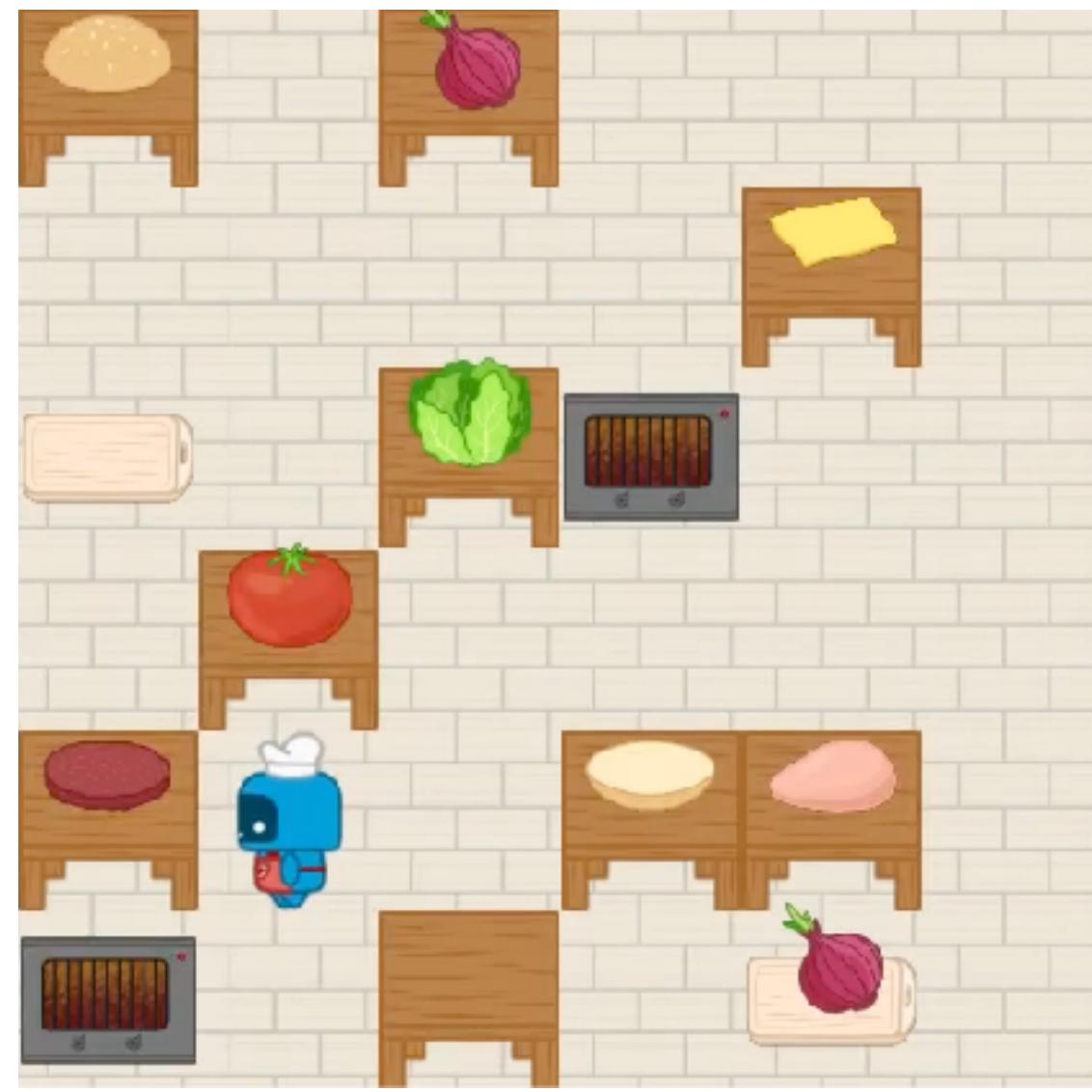
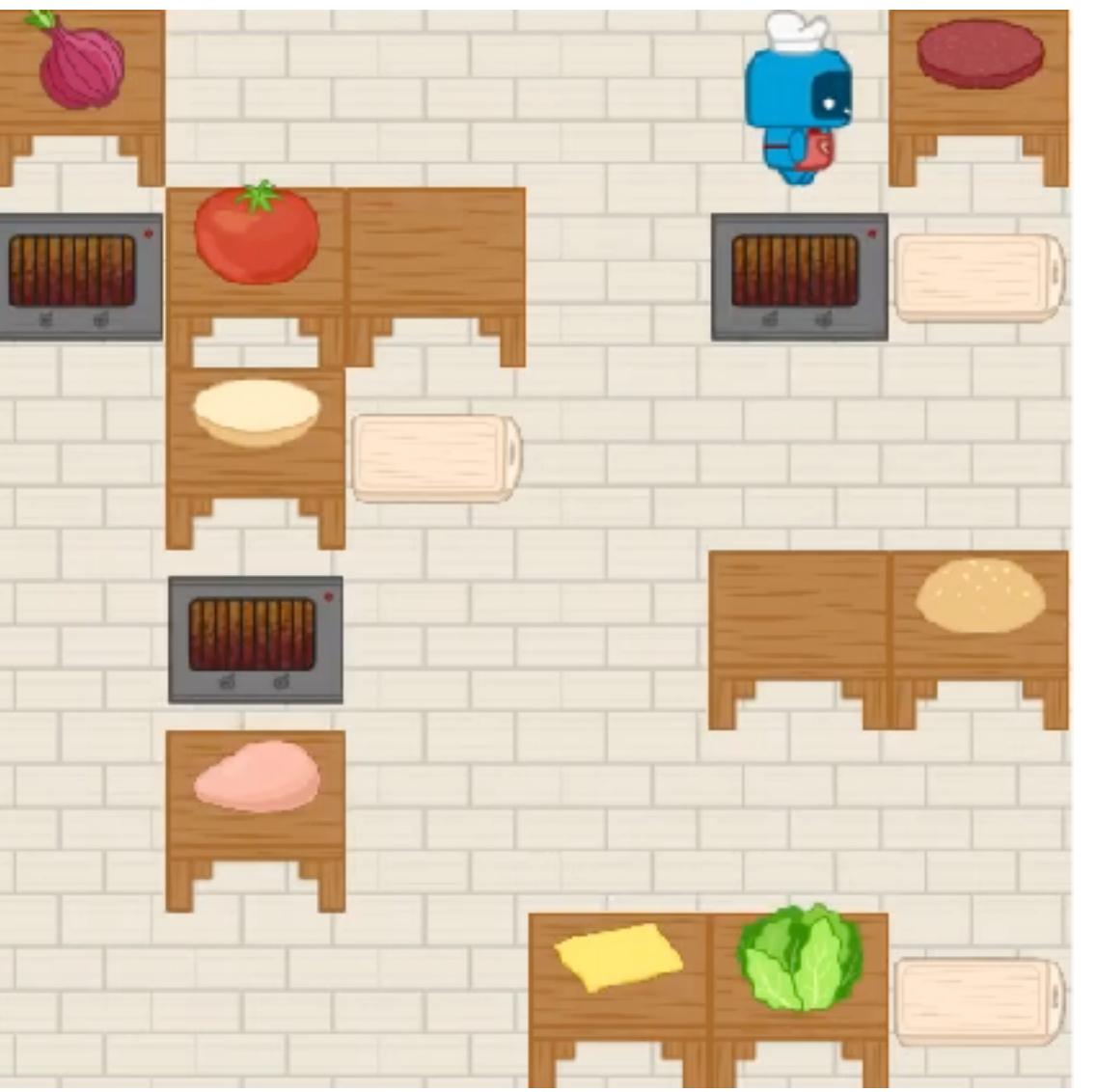
Demo2Code can generalize to new, complex environments



```
def cook_object_at_loc(obj, loc):
    if not is_holding(obj):
        ...
    move_then_place(obj, loc)
    cook_until_is_cooked(obj)

def move_then_place(obj, loc):
    curr_loc = get_curr_loc()
    if curr_loc != loc:
        move(curr_loc, loc)
    place(obj, place_location)

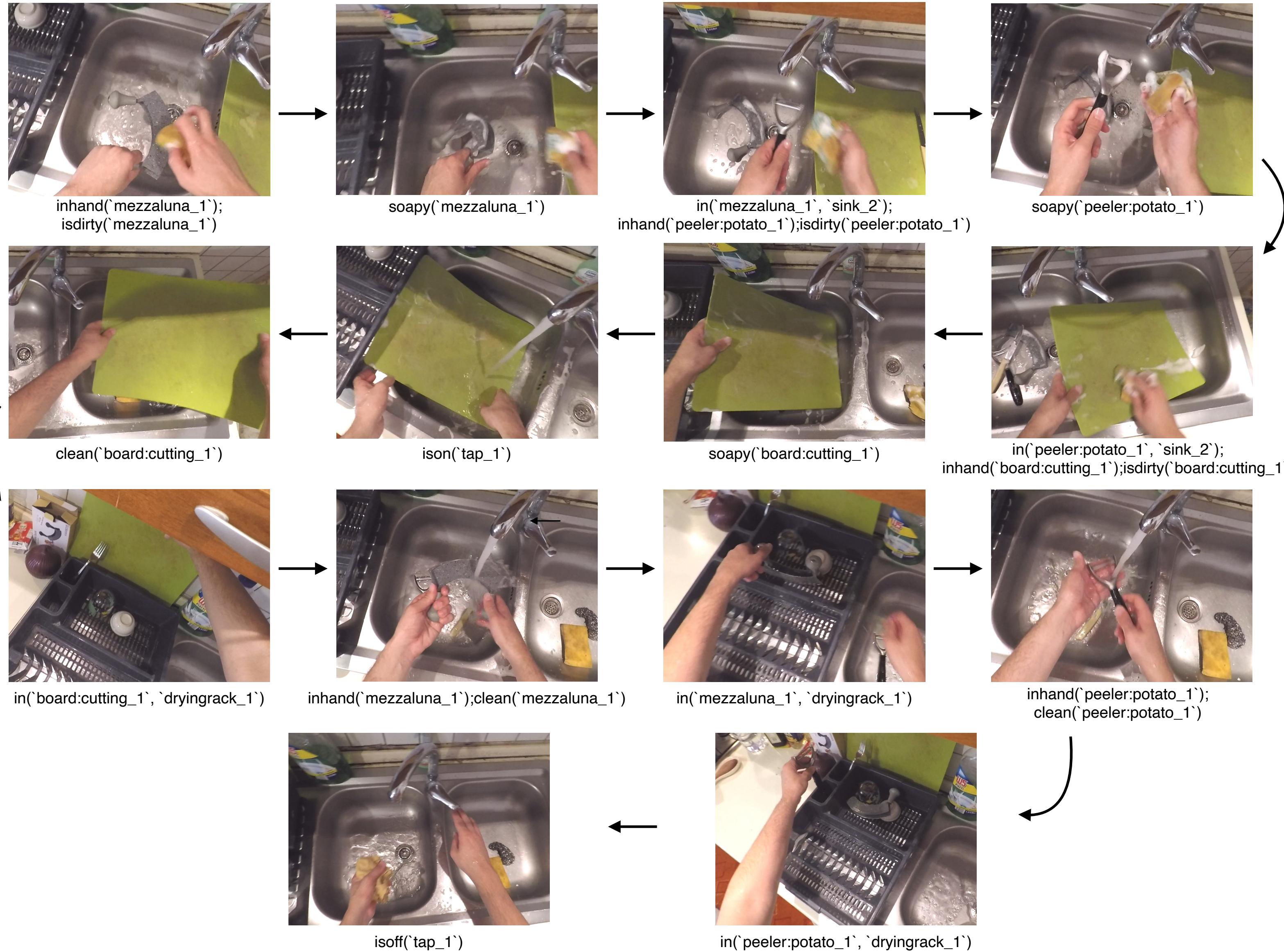
def main():
    ...
    cut_object_at_loc(lettuce, stove)
    ...
    stack_objects(lettuce, bottom_bun)
    ...
    cook_object_at_loc(patty, stove)
    ...
    stack_objects(top_bun, patty)
```



Demo2Code generates correct code that passes unit tests

Task	Lang2Code[33]			DemoNoLang2Code			Demo2Code(<i>ours</i>)			Horizon Length
	Exec.	Pass.	BLEU.	Exec.	Pass.	BLEU.	Exec.	Pass.	BLEU.	
Cook a patty	1.00	1.00	0.90	1.00	1.00	0.90	1.00	1.00	0.90	8.0
Cook two patties	0.80	0.80	0.92	0.80	0.80	0.92	0.80	0.80	0.92	16.0
Stack a top bun on top of a cut lettuce on top of a bottom bun	1.00	1.00	0.70	0.00	0.00	0.75	1.00	1.00	0.60	14.0
Cut a lettuce	1.00	1.00	0.87	0.00	0.00	0.76	1.00	1.00	0.87	7.0
Cut two lettuces	0.80	0.80	0.92	0.00	0.00	0.72	0.80	0.80	0.92	14.0
Cook first then cut	1.00	1.00	0.88	1.00	1.00	0.88	1.00	1.00	0.88	14.0
Cut first then cook	1.00	1.00	0.88	0.00	0.00	0.82	1.00	1.00	0.88	15.0
Assemble two burgers one by one	0.00	0.00	0.34	1.00	1.00	0.77	1.00	1.00	0.76	15.0
Assemble two burgers in parallel	0.00	0.00	0.25	1.00	1.00	0.51	0.00	0.00	0.71	15.0
Make a cheese burger	1.00	0.00	0.24	1.00	1.00	0.69	1.00	1.00	0.69	18.0
Make a chicken burger	0.00	0.00	0.57	0.00	0.00	0.64	0.90	0.90	0.69	25.0
Make a burger stacking lettuce atop patty immediately	1.00	0.00	0.74	0.20	0.00	0.71	0.00	0.00	0.71	24.5
Make a burger stacking patty atop lettuce immediately	0.00	0.00	0.74	0.20	0.00	0.71	1.00	1.00	0.74	25.0
Make a burger stacking lettuce atop patty after preparation	1.00	0.00	0.67	0.10	0.00	0.65	0.00	0.00	0.66	26.5
Make a burger stacking patty atop lettuce after preparation	1.00	0.00	0.67	0.00	0.00	0.53	1.00	0.00	0.69	27.0
Make a lettuce tomato burger	0.00	0.00	0.13	1.00	1.00	0.85	1.00	0.00	0.66	34.0
Make two cheese burgers	0.00	0.00	0.63	1.00	1.00	0.68	1.00	1.00	0.68	38.0
Make two chicken burgers	0.00	0.00	0.52	0.00	0.00	0.68	1.00	0.00	0.56	50.0
Make two burgers stacking lettuce atop patty immediately	0.80	0.00	0.66	0.80	1.00	0.69	0.00	0.00	0.66	50.0
Make two burgers stacking patty atop lettuce immediately	0.80	0.00	0.67	1.00	0.00	0.48	1.00	1.00	0.73	50.0
Make two burgers stacking lettuce atop patty after preparation	0.80	0.00	0.66	0.60	0.00	0.66	0.80	0.00	0.67	54.0
Make two burgers stacking patty atop lettuce after preparation	0.80	0.00	0.67	0.50	0.00	0.71	0.80	0.00	0.68	54.0
Make two lettuce tomato burgers	1.00	0.00	0.55	0.00	0.00	0.70	1.00	1.00	0.84	70.0
Overall	0.64	0.29	0.64	0.49	0.38	0.71	0.79	0.59	0.74	28.8

EPIC-Kitchens Tasks

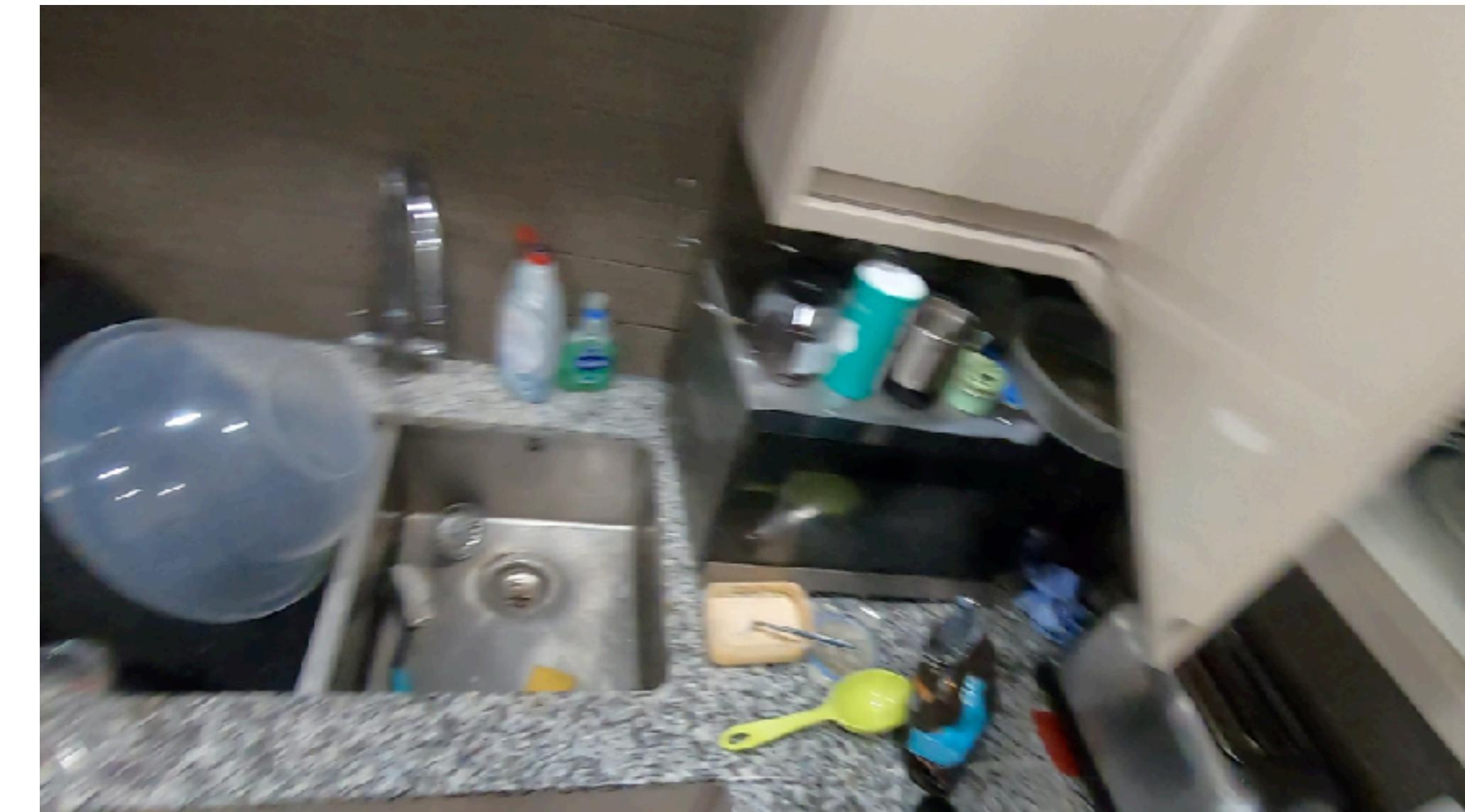


```

objects = get_all_objects()
for object in objects:
    pick_up(object)
    if check_if_dirty(object):
        while check_if_dirty(object):
            scrub(object)
            place(object, "sink_2")
turn_on("tap_1")
for object in objects:
    pick_up(object)
    rinse(object)
    place(object, "dryingrack_1")
turn_off("tap_1")

```

EPIC-Kitchens Dishwashing Tasks



	P4-101 (7)		P7-04 (17)		P7-10 (6)		P22-05 (28)		P22-07 (30)		P30-07 (11)		P30-08 (16)	
	Pass.	BLEU.	Pass.	BLEU.	Pass.	BLEU.	Pass.	BLEU.	Pass.	BLEU.	Pass.	BLEU.	Pass.	BLEU.
Lang2Code [33]	1.00	0.58	0.00	0.12	0.00	0.84	0.00	0.48	0.00	0.37	1.00	0.84	0.00	0.66
DemoNoLang2Code	0.00	0.00	0.00	0.00	1.00	0.00	0.00	0.37	0.00	0.51	1.00	0.57	0.00	0.00
Demo2Code	1.00	0.33	0.00	0.19	1.00	0.63	1.00	0.43	1.00	0.66	1.00	0.58	0.00	0.24

A photograph of a two-lane asphalt road curving through a landscape at sunset. The road is marked with white dashed lines and a solid yellow center line. The surrounding terrain is dry and hilly, with sparse vegetation. The sky is a warm orange and yellow, with the sun low on the horizon.

Exciting coming years for robot learning!

SO LONG AND...



Thanks for all the fish!