

# Object-Oriented Option Framework for Robotics Manipulation in Clutter



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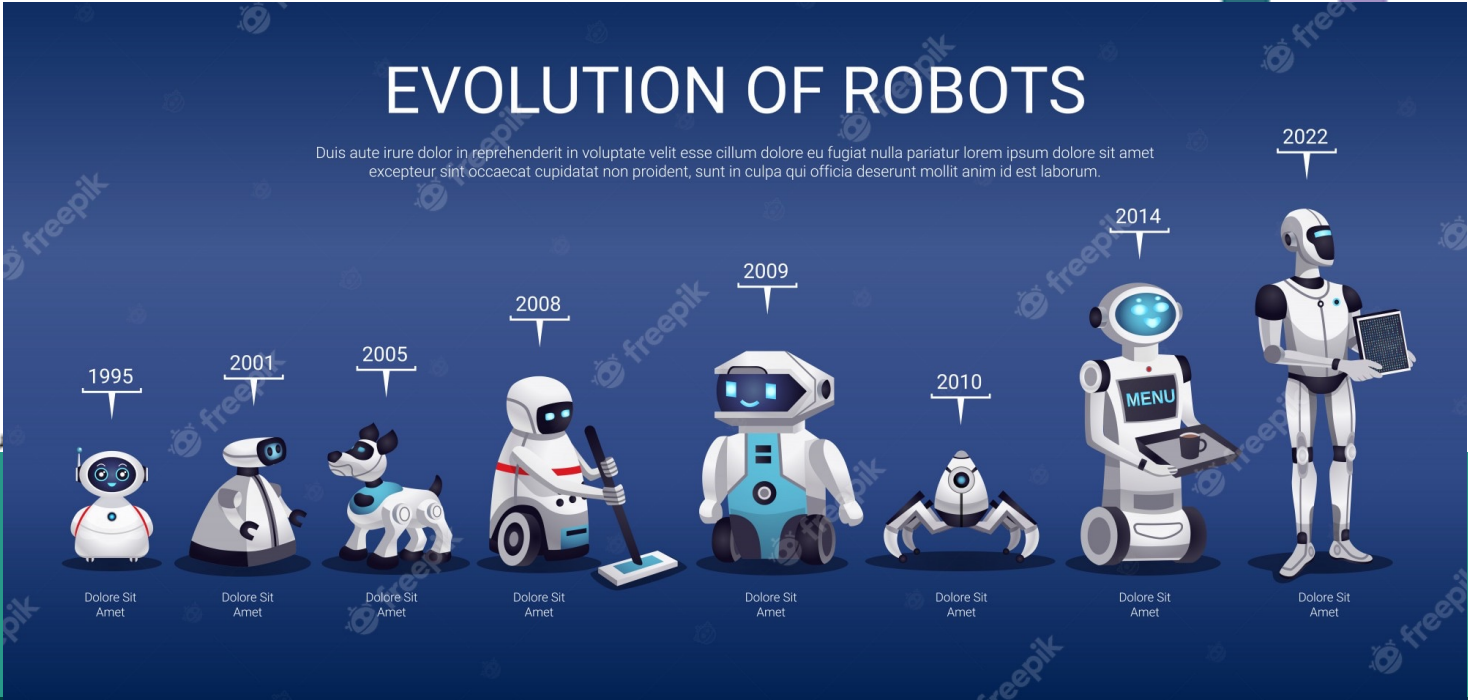
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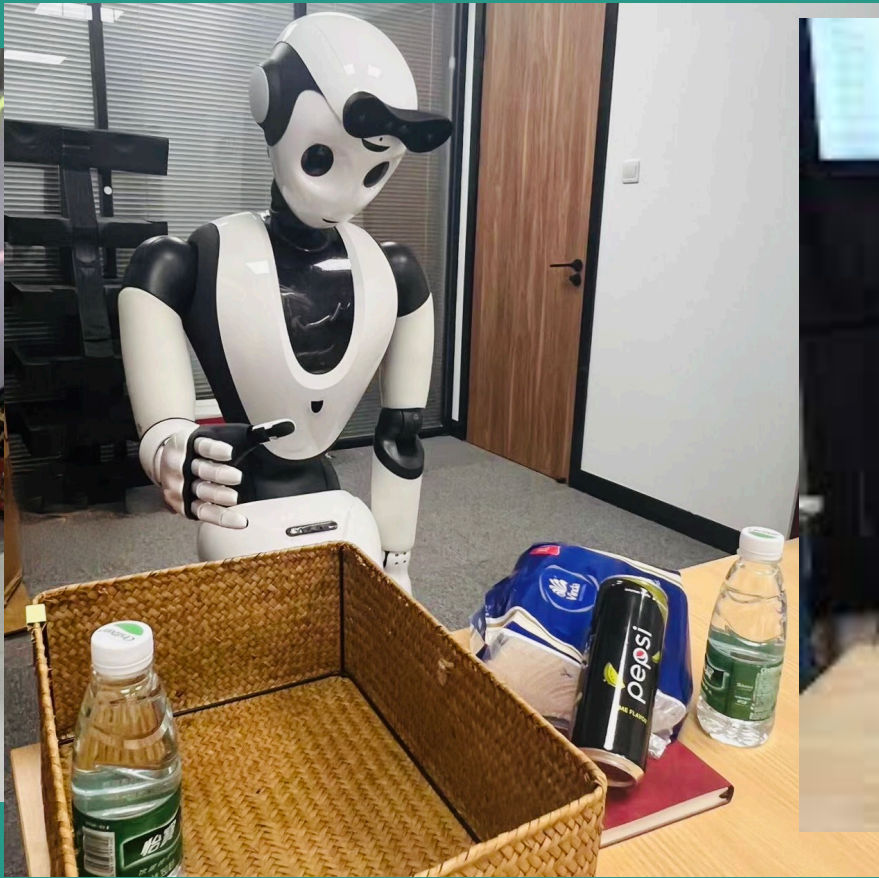
# Background

- Domestic popular.

U.S. professional service robots market size, by application, 2016 - 2027 (USD Million)



# Background



- Examples of Manipulating Objects in Clutter (MoC) problems



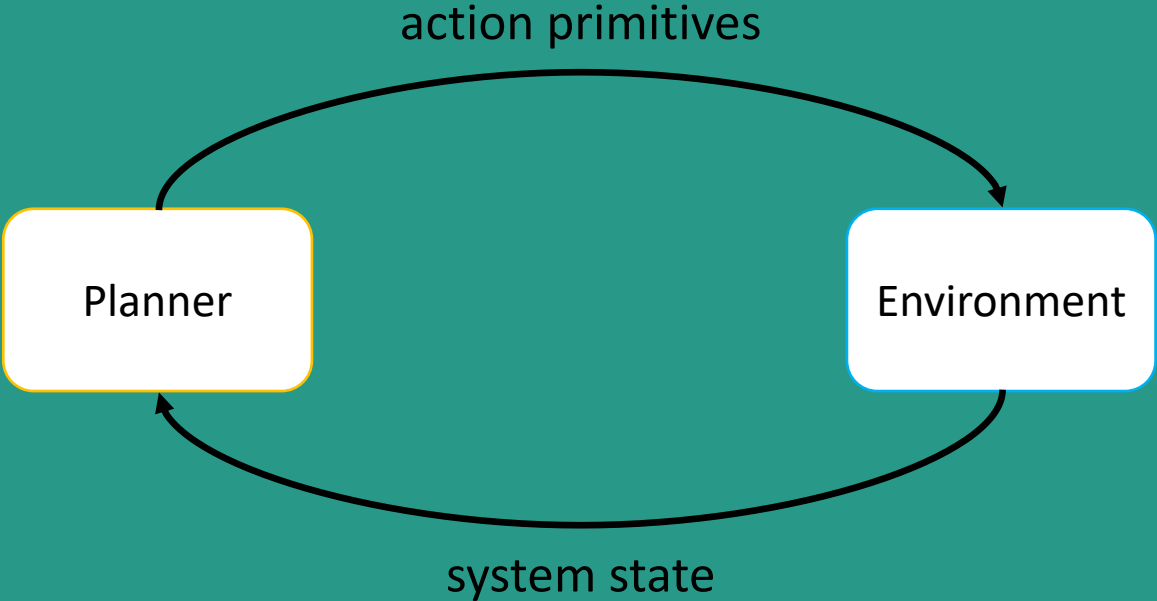


# Background

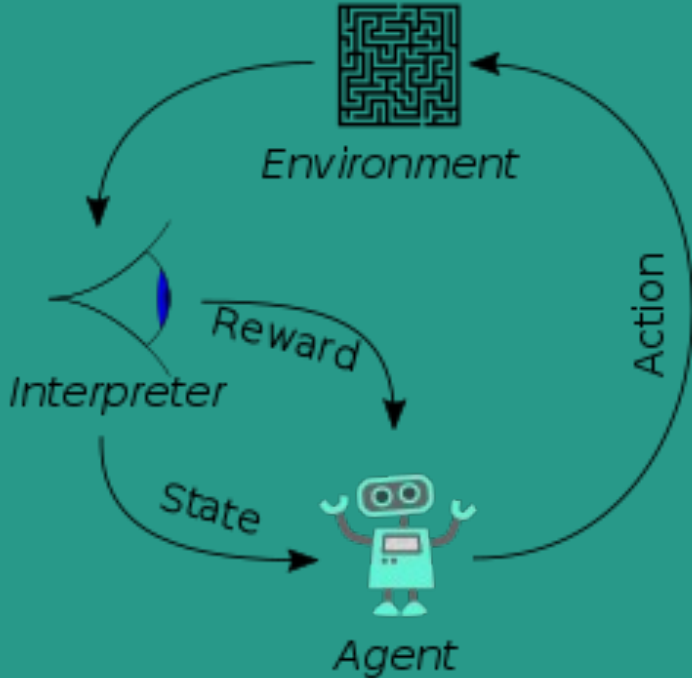
- Keys to solve MoC problems:
  - Identify objects
  - Planning
  - Robotics control/manipulation



# Related works



Action primitives + planning

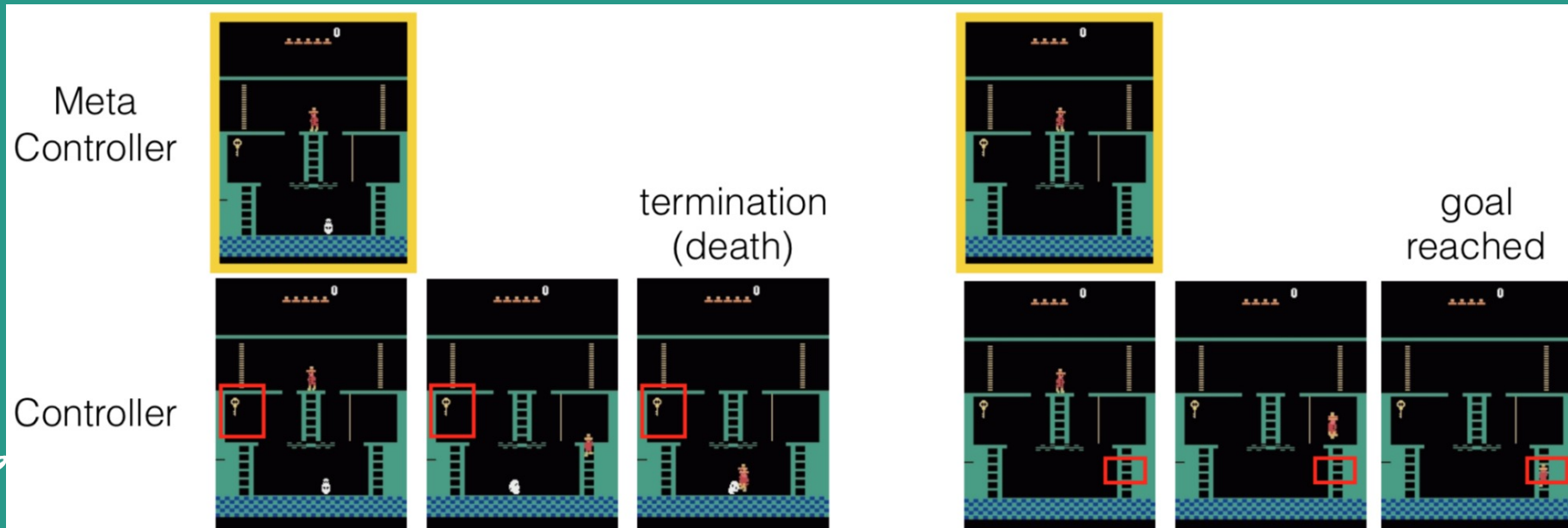


reinforcement learning



# Method

- Option Framework (OF)
  - Option: temporal abstract actions, denoted by  $\{I, \pi(a|s), \beta(s, o)\}$



# Method

- Option Framework
  - Discover options from scratch based on RL
  - May cause collapsed options and hinder policy learning.



# Method

- Object-oriented Option Space

- Each option in the space represents once object movement.

1. object to move
2. target location





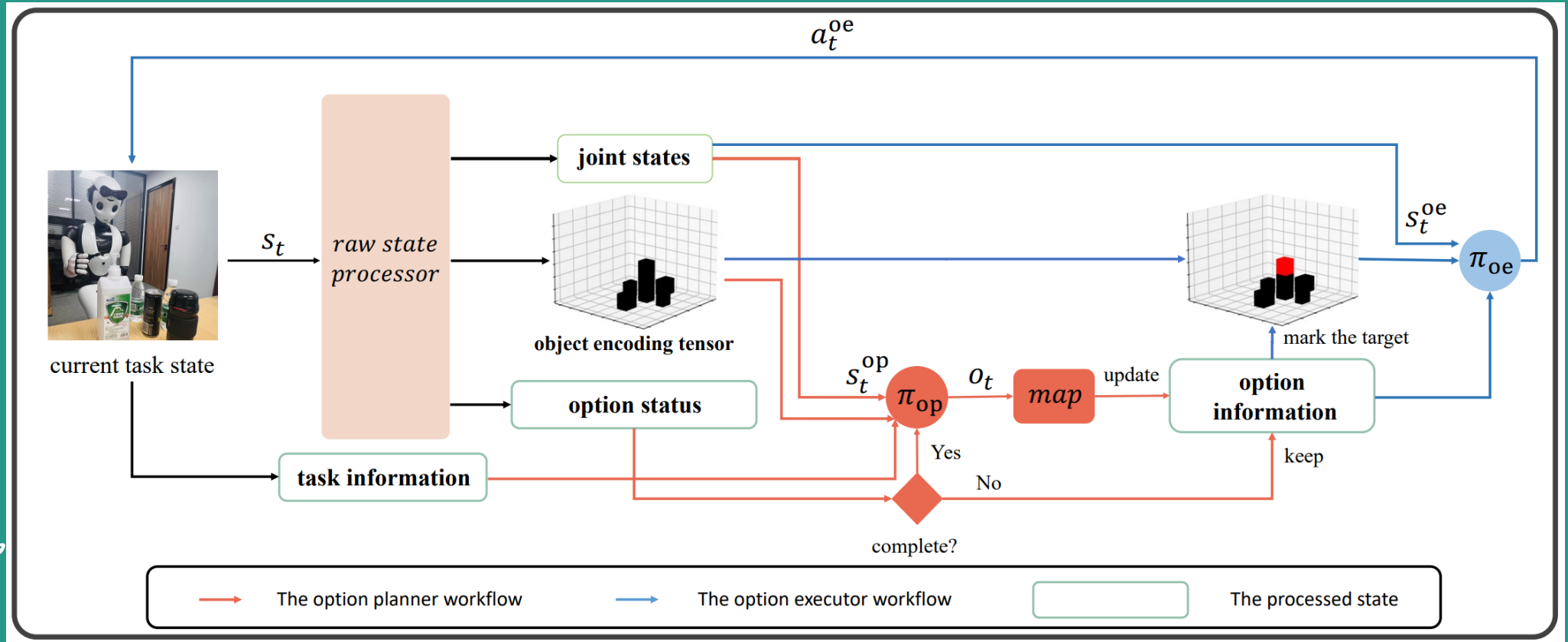
# Method

- Object-oriented Option Framework (O3F)
  1. Option planner  $\pi_{op}(o|s)$ : a policy that makes decision over option space
  2. Option executor  $\pi_{op}(o|s)$ : a universal intra-policy for all options.

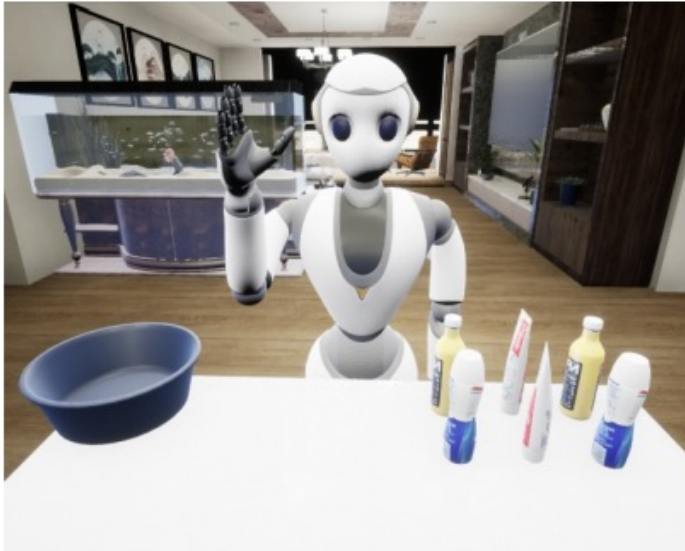


# Method

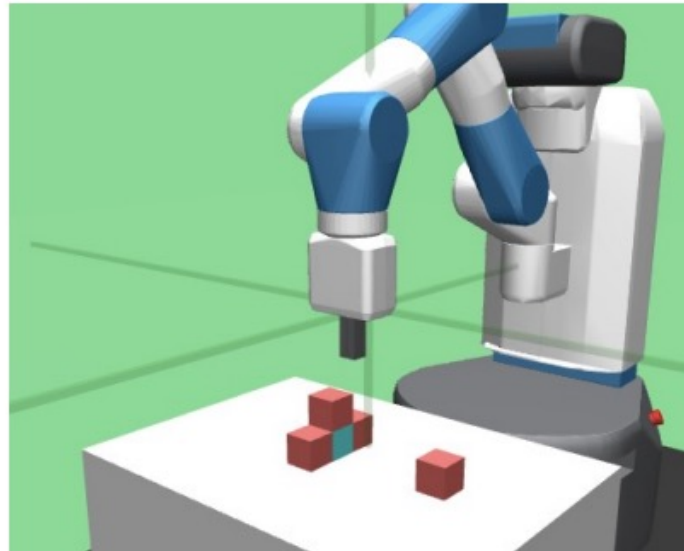
- Object-oriented Option Framework (O3F)



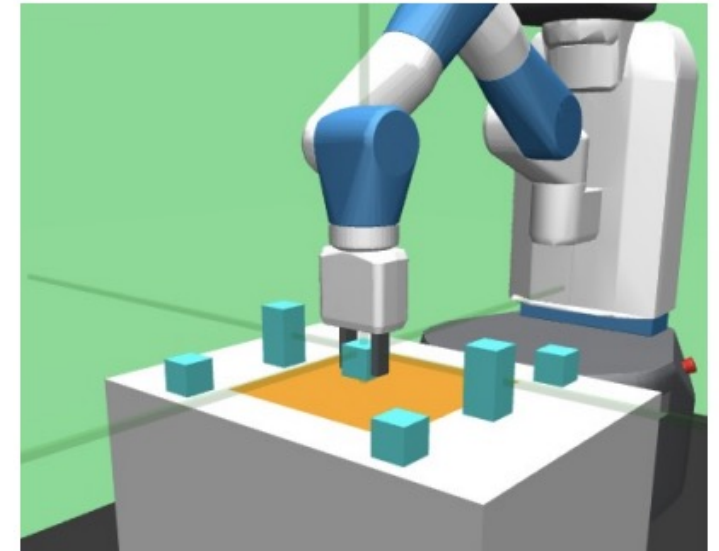
# Experiments



(a) Grasp (Ginger XR1)



(b) Grasp



(c) Collect

# Results

**TABLE I:** Success rate (%) of grasping task on the Ginger XR1 Robot.

Method	Scene 1	Scene 2	Scene 3	Average
O3F (Ours)	<b>88.0</b>	<b>95.0</b>	<b>91.0</b>	<b>91.3</b>
PPO [36]	2.0	4.0	3.0	3.0

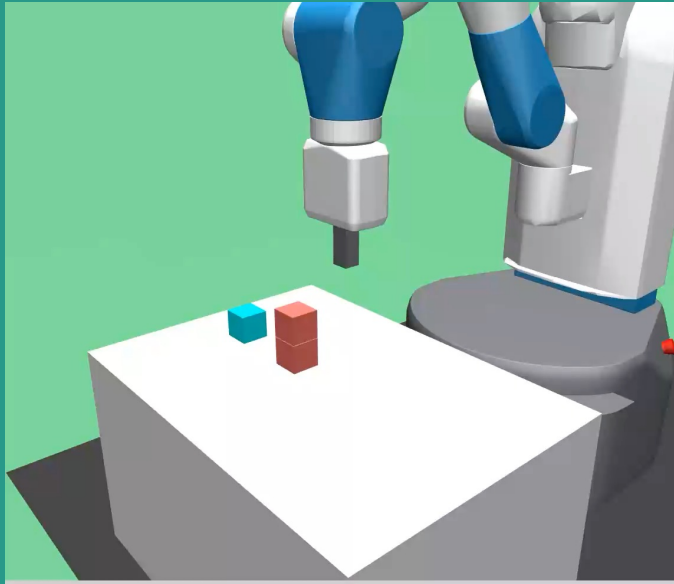
**TABLE III:** Success rate (%) of O3F and PPO in grasping and collecting tasks with the robot arm.

Method	grasping	collecting
O3F (Ours)	<b>72.4</b>	<b>90.0</b>
PPO [24]	20.6	3.33

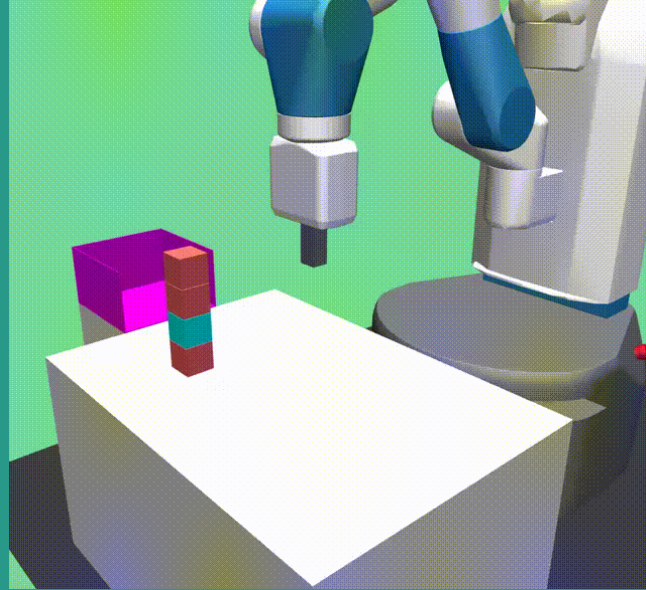




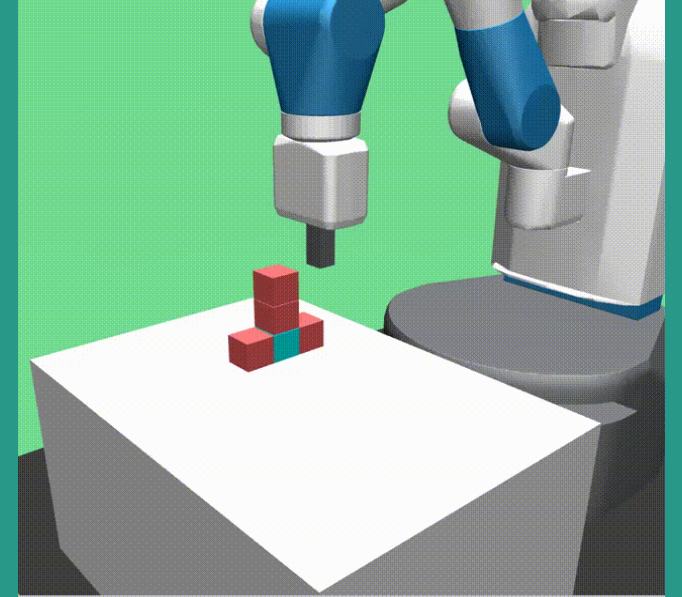
# Results



O3F



VPG<sup>[1]</sup>

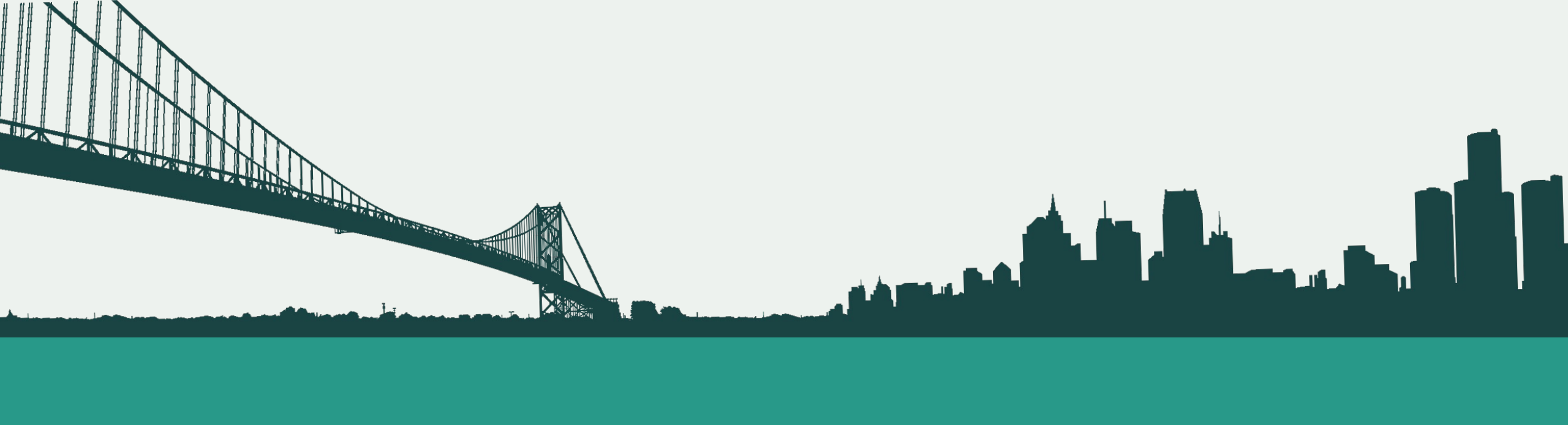


PPO<sup>[2]</sup>



[1] ZENG, A., SONG, S., WELKER, S., LEE, J., RODRIGUEZ, A., AND FUNKHOUSER, T. A. Learning synergies between pushing and grasping with self-supervised deep reinforcement learning. In International Conference on Intelligent Robots and Systems, IROS (2018).

[2] SCHULMAN, J., WOLSKI, F., DHARIWAL, P., RADFORD, A., AND KLIMOV, O. Proximal policy optimization algorithms, 2017.



**Thanks for your listening!**