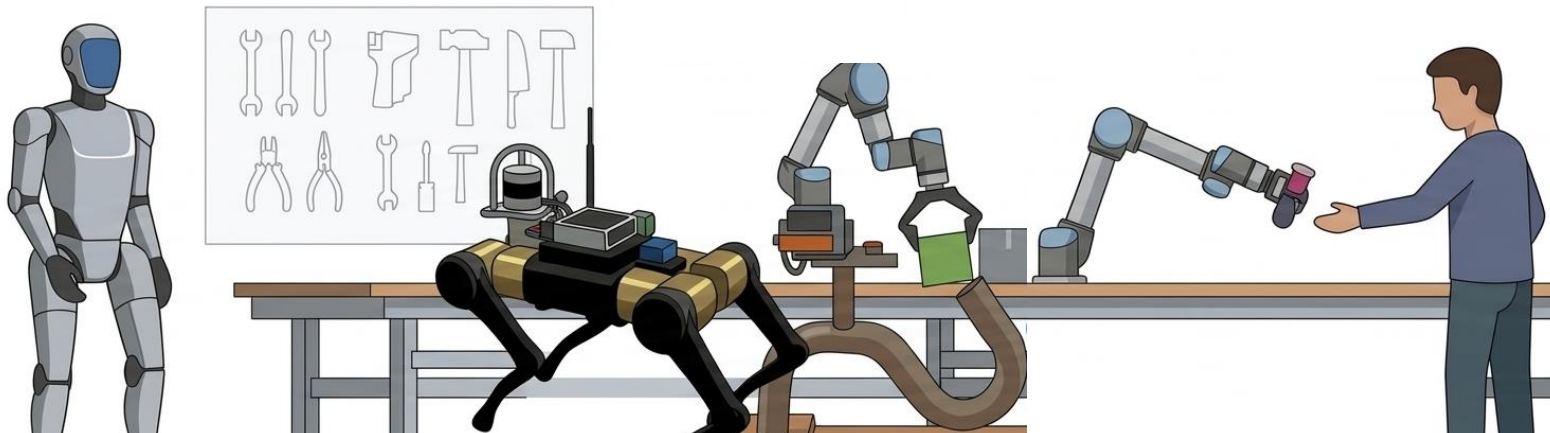


# Robust Intelligence 강인지능 Robotics Lab 로봇틱스 랩

Research Highlight



**Dr. Daehyung Park**  
Associate Professor  
School of Computing, KAIST



- Name** Daehyung Park
- Affiliation** School of Computing, KAIST
- Degree** Robotics Ph.D., Georgia Tech.
- Research Interest** Robotics, LfD, TAMP, NLG/NLU

## Major Research Experience & Awards

- ✔ Postdoctoral Associate, MIT CSAIL (2018-2020)
- ✔ Research Engineer, Samsung (2008-2012)
- ✔ ICRA Outstanding Planning Paper Award (2023)
- ✔ ICRA Outstanding Navigation Paper Award Finalist (2022)
- ✔ Google Research Scholar Award (2022)



## Representative Work



## Professional Activities

- ✔ Director, Korean Robotics Society (2025~Present)
- ✔ Demo Chair, CoRL (2025)
- ✔ Program Chair, RiTA (2024)
- ✔ Organizer, Workshop at Humanoids (2024)
- ✔ Organizer, Workshop at RSS (2023)

## Funding Agencies



**Opened**  
in 1971

**Located**  
in Daejeon, S. Korea



**Students**  
12,348

**Comprise of 6 Colleges**  
7 Schools  
33 Departments  
55 Programs

**Faculty and Staff**

Faculty	731
Staff	907

**International Population**

Faculty	137
Students	929



# Robust Intelligence & Robotics (RIRO) Lab

## Opened

in 2020 Fall

## Affiliation

School of Computing

## Members

4 Ph.D. (+1 soon)

11 M.S.

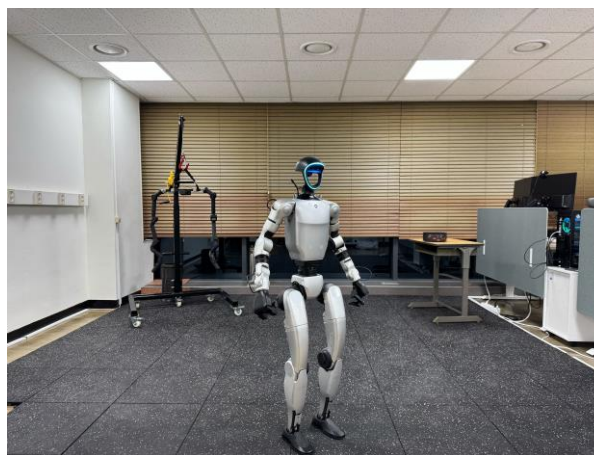
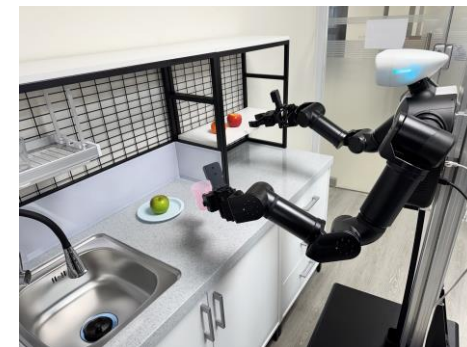
1 visiting scholar

## Research Area

Manipulation,  
Skill Learning,  
HRI



KAIST



Robust Intelligence & Robotics Lab



# Robust Intelligence & Robotics (RIRO) Lab

## Projects



Gov. Proj 7



Industry Proj 3

## Facilities



Office room 3



Robot Lab 3



Server Workstation 6ea. 8ea.



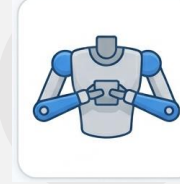
3D Printer 2ea.



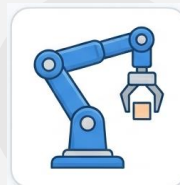
## Equipments



Humanoid Robot 1ea.



Dual Arm 1ea.



Robot Arm 2ea.



Mobile Arm 2ea.



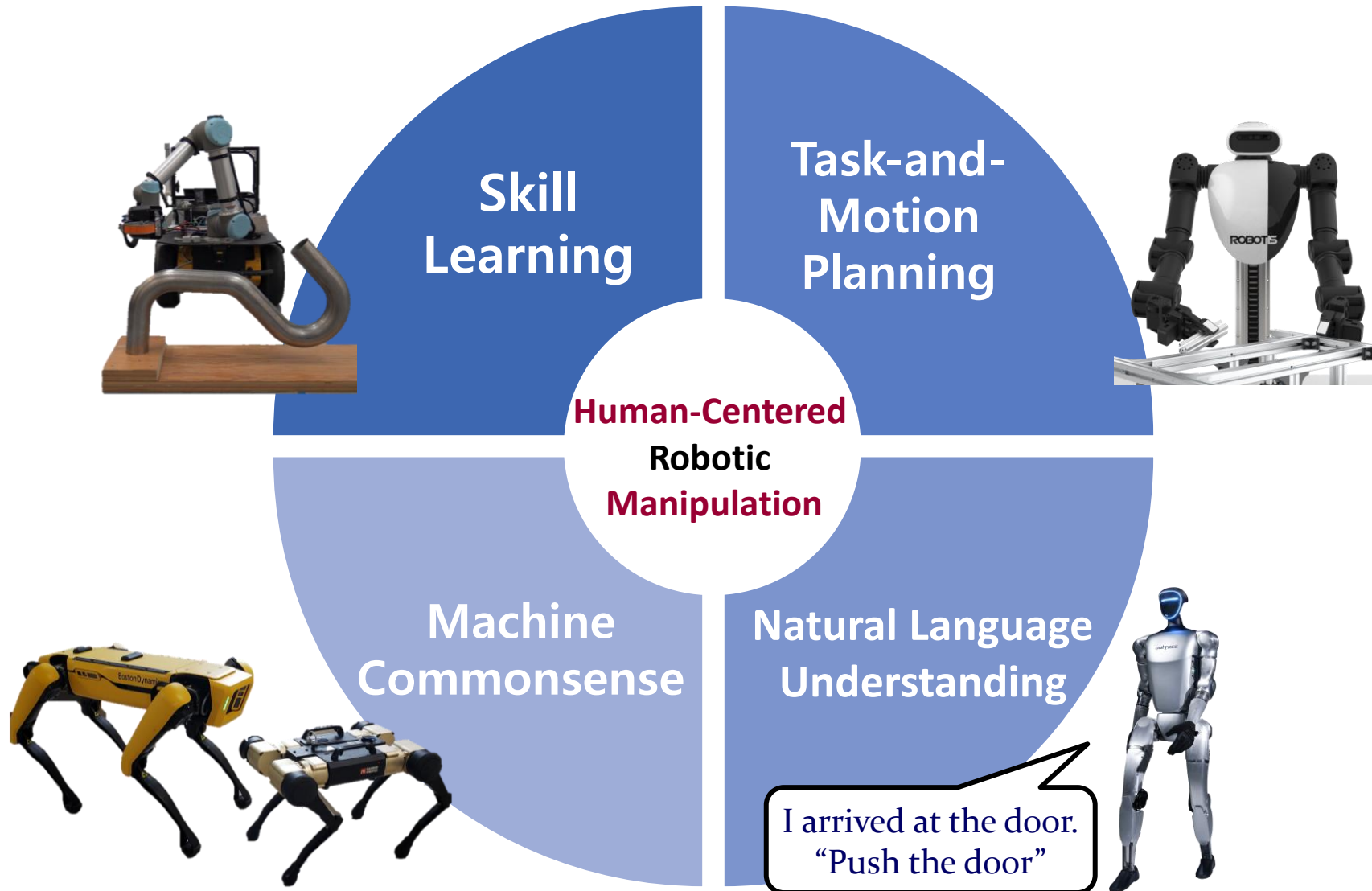
Robot Hand 1ea.



Quadruped Robot 1ea.

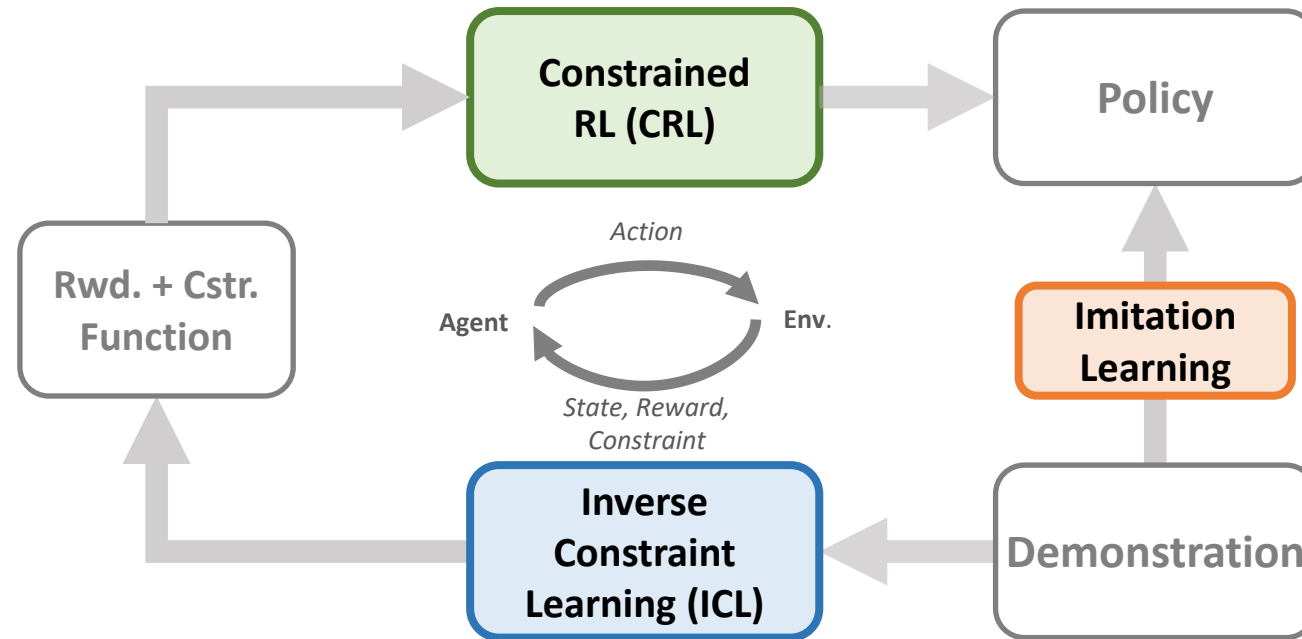
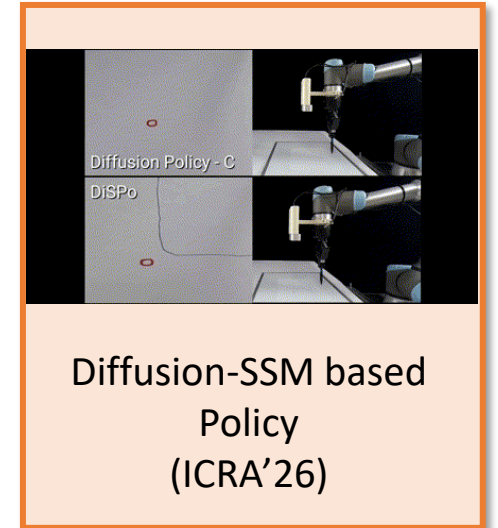
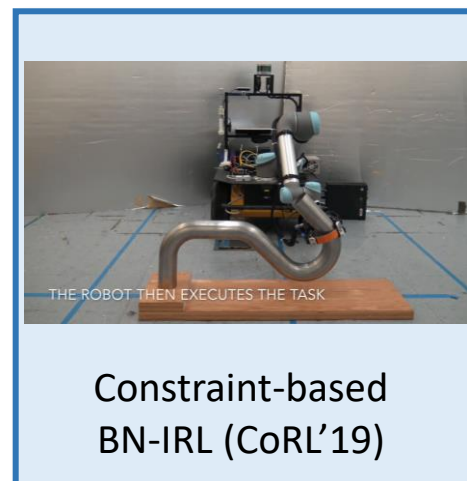
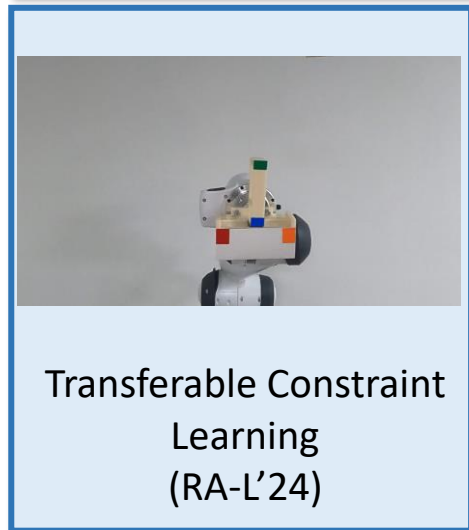
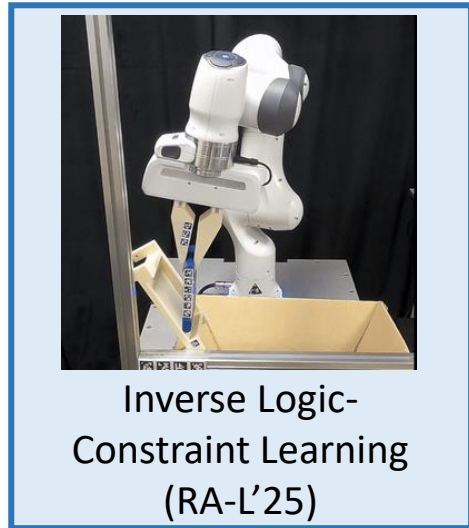
# Human-Centered Robotics:

## How to bridge the gap between humans and robots?





# Inverse Skill Learning



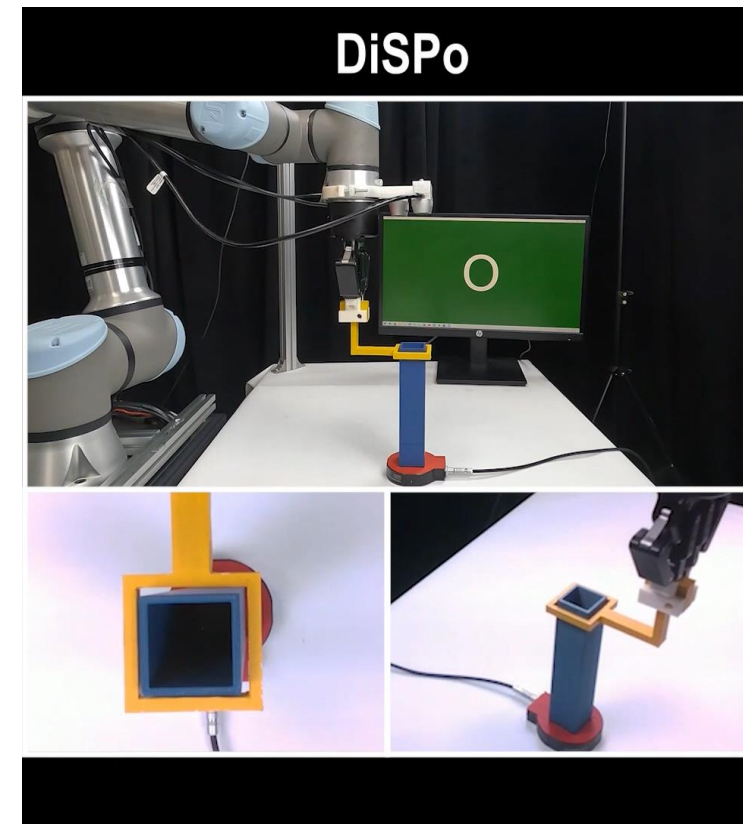
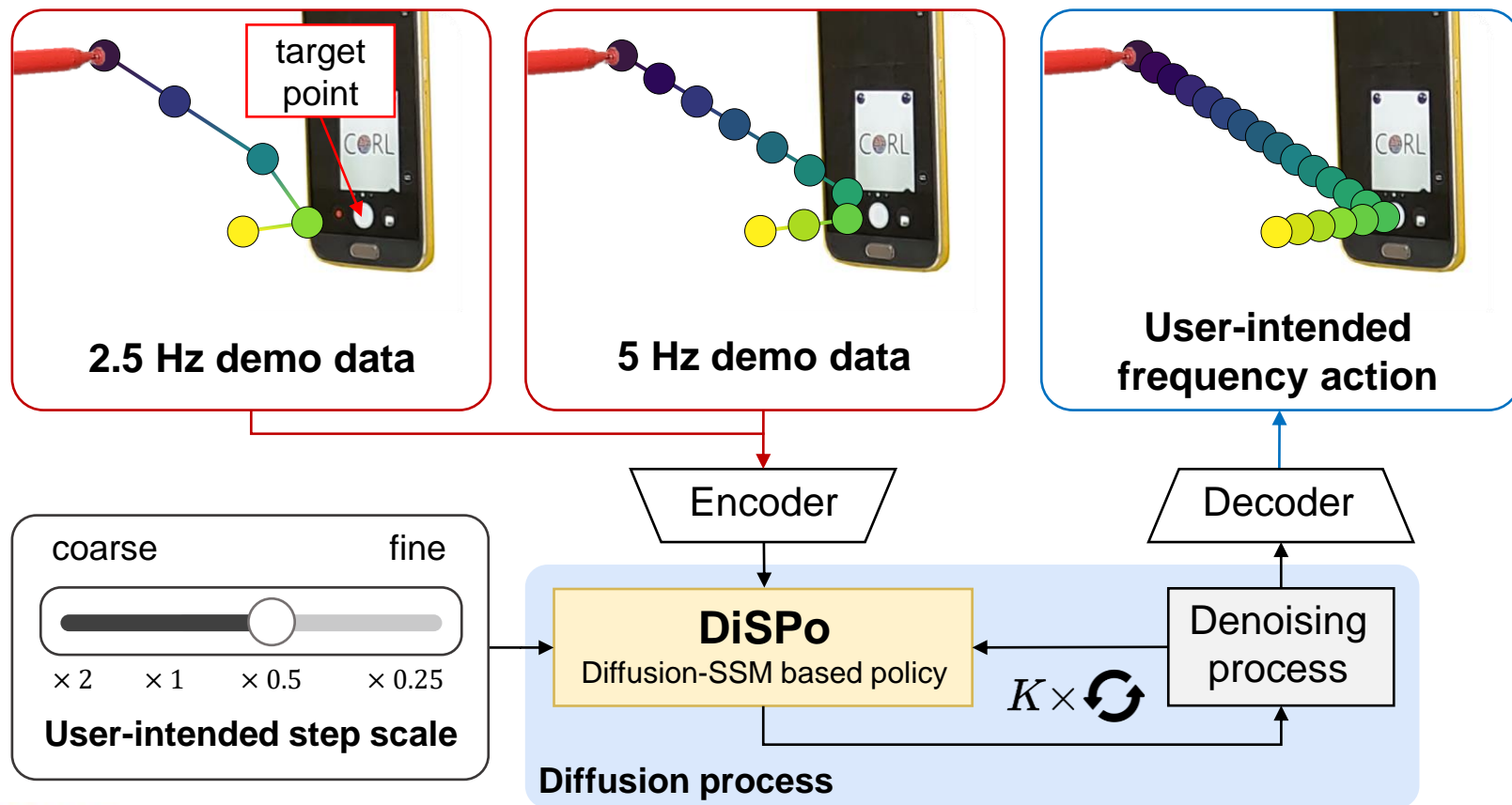
# Coarse-to-fine manipulation

## DiSPo: Diffusion-SSM based Policy



Jaehyeong

DiSPo that leverages a **state-space model**, Mamba, to learn from diverse coarse demonstrations and generate **multi-scale actions**.



# Coarse-to-fine manipulation

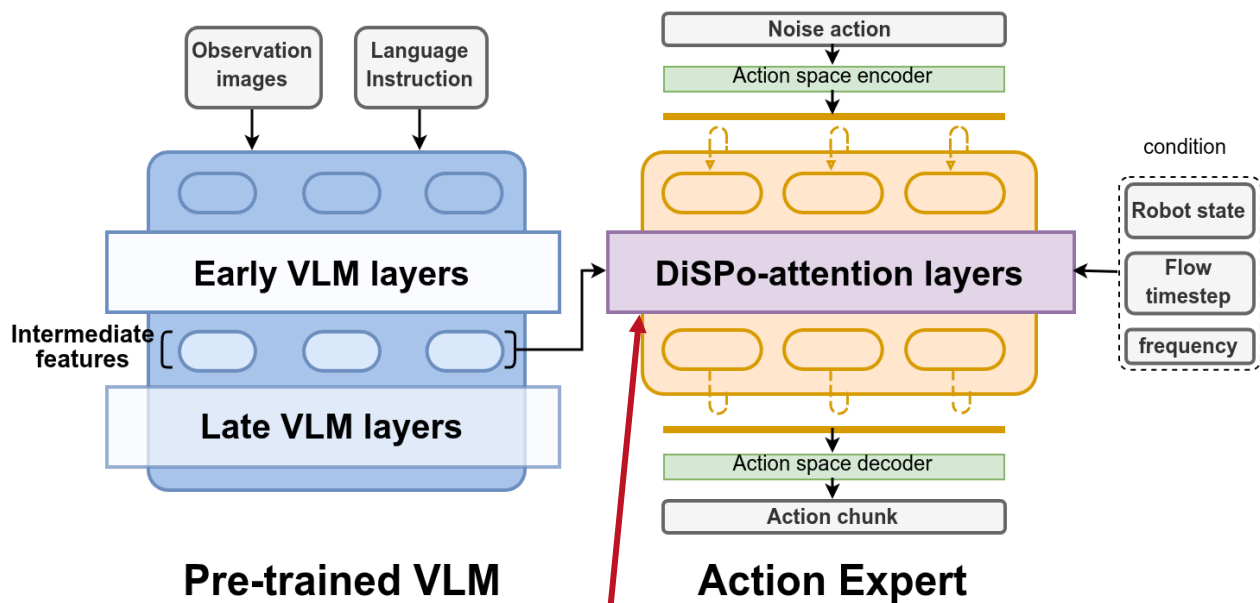
## DiSPo-VLA (1B)



Sangyun

(to be submitted)

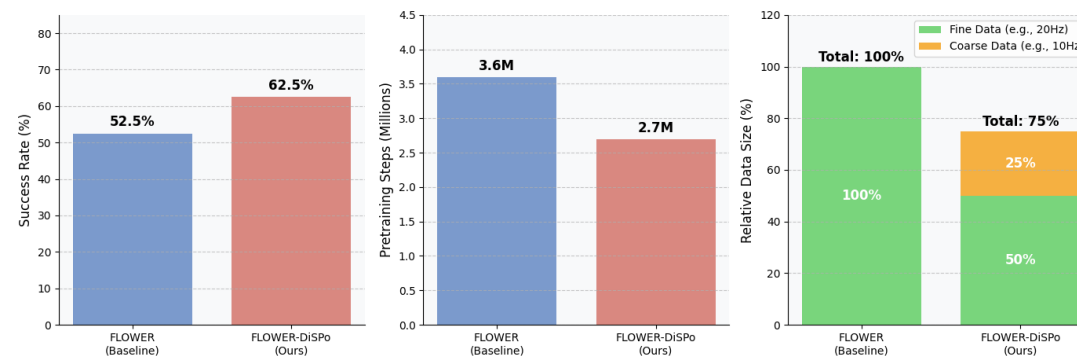
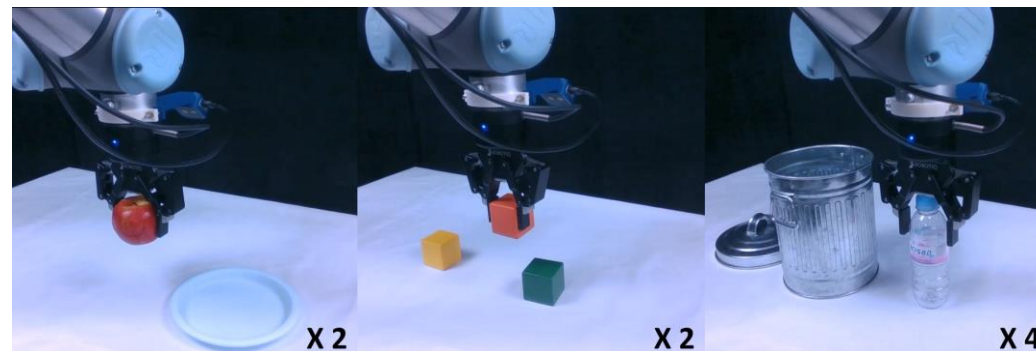
DiSPo's multi-granularity learning reduces **time & storage complexities** in large scale training.



**Self-attention replacement**

$$O(N^2) \rightarrow O(N)$$

(DiSPo-VLA~1B)



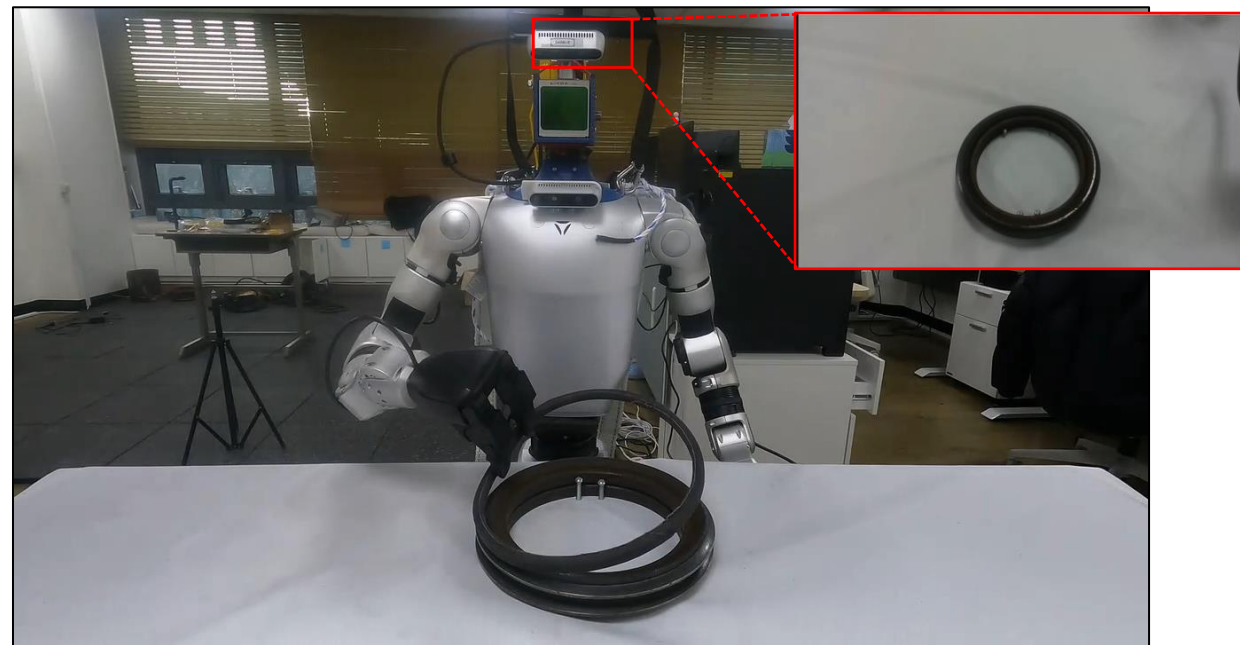
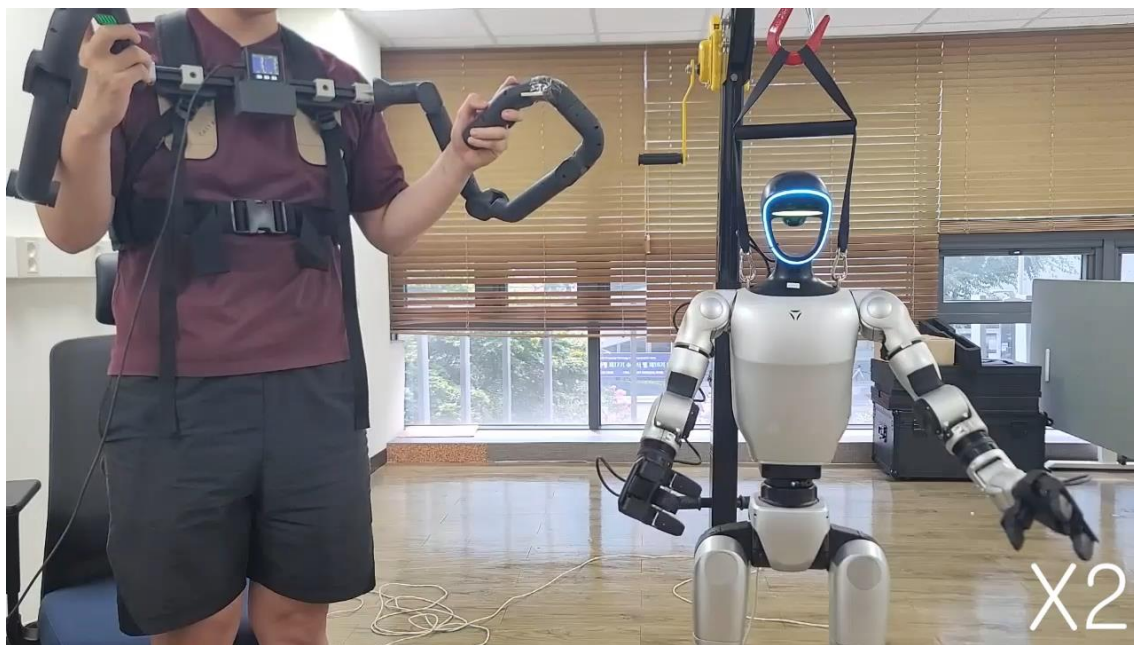
Success Rates

Pretraining steps

Relative dataset size



We learn manipulation skills for highly elastic deformable objects using a humanoid's dual arms via reinforcement learning.



# Constraints in Everyday Tasks

---



Credit: Ljupco

## A wide **variety** of constraints

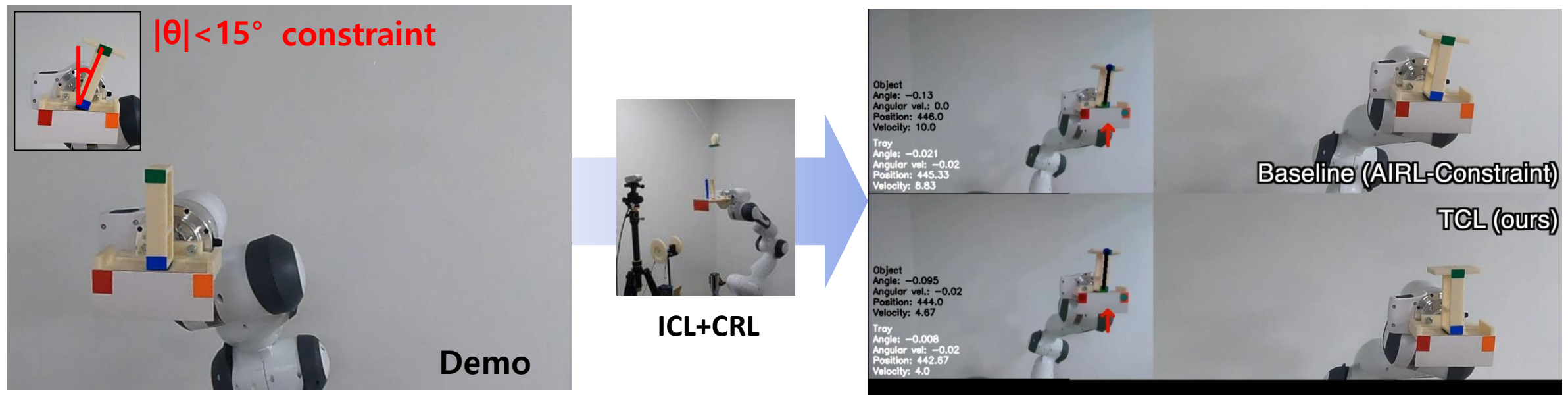
- Safety constraints
- Operational constraints
- Task-specific constraints
- Preference, legal, privacy, etc.

You may fail to complete the carrying without satisfying constraints.

# Inverse Constraint Learning

## Transferable Constraint Learning (TCL)

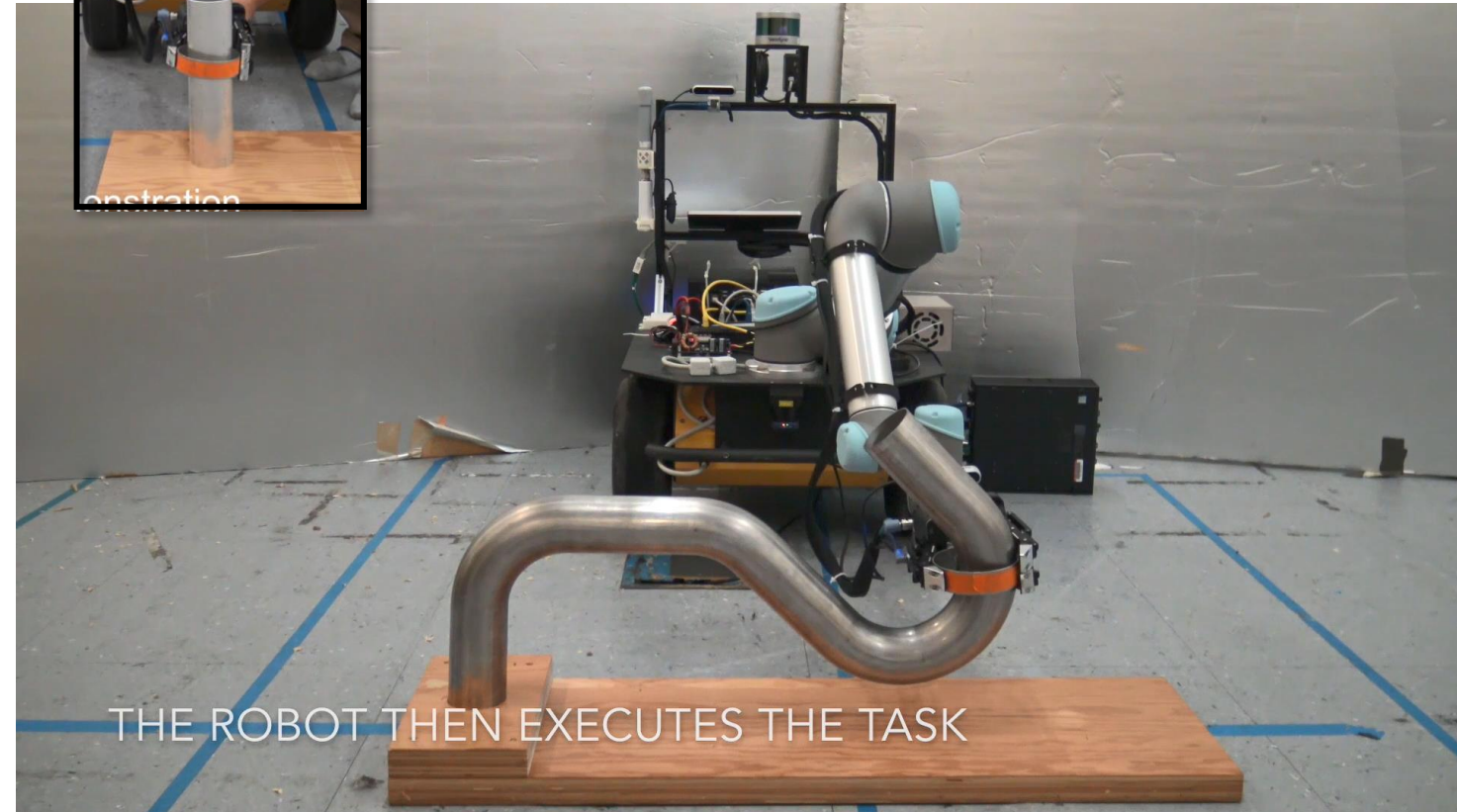
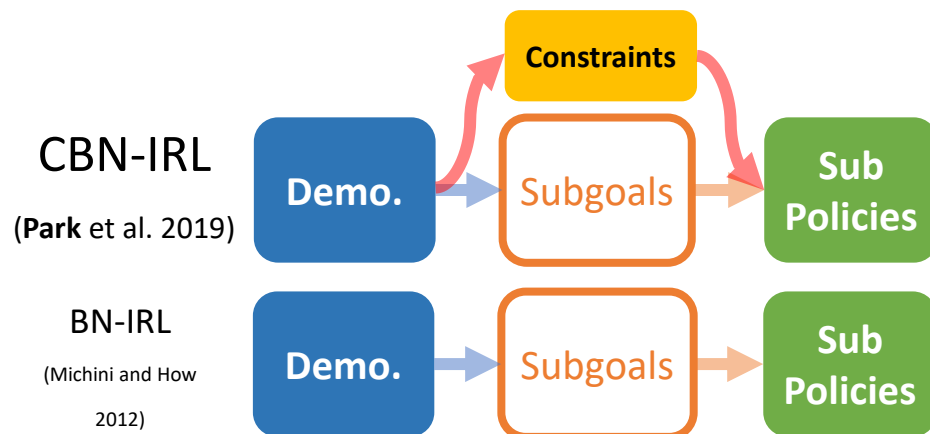
A transferable constraint learning (TCL) algorithm that jointly infers a **task-relevant reward** and **task-agnostic constraint**.



“

**Locally active constraints** are common in manipulation.

”



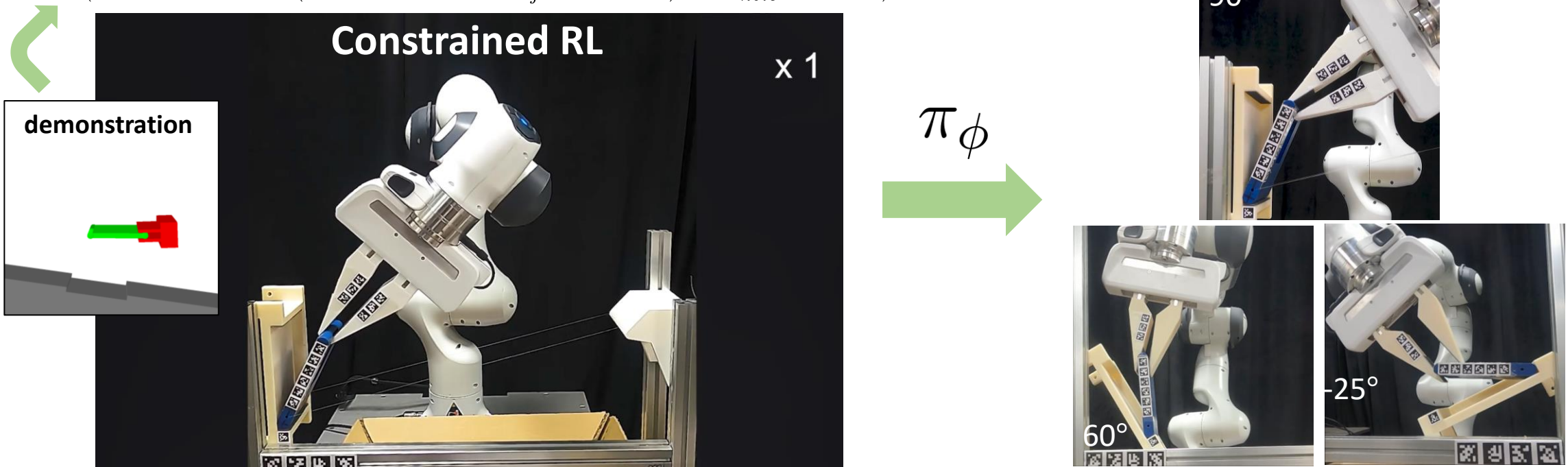
Park, Daehyung, et al. "Inferring task goals and constraints using bayesian nonparametric inverse reinforcement learning." CoRL, 2020.

# Inverse Constraint Learning

## ILCL : Inverse Logic-Constraint Learning

ILCL learns **temporal logic constraints** from temporally constrained demonstration leveraging two-player zero-sum game framework.

$$\diamond (\Sigma_h \theta^{peg} > 34.88^\circ \wedge (\Sigma_h \theta^{peg} < 4.217^\circ \mathcal{R} d_{jaw}^{peg} < 0.0053) \wedge \square d_{hole}^{peg} < 0.0072)$$

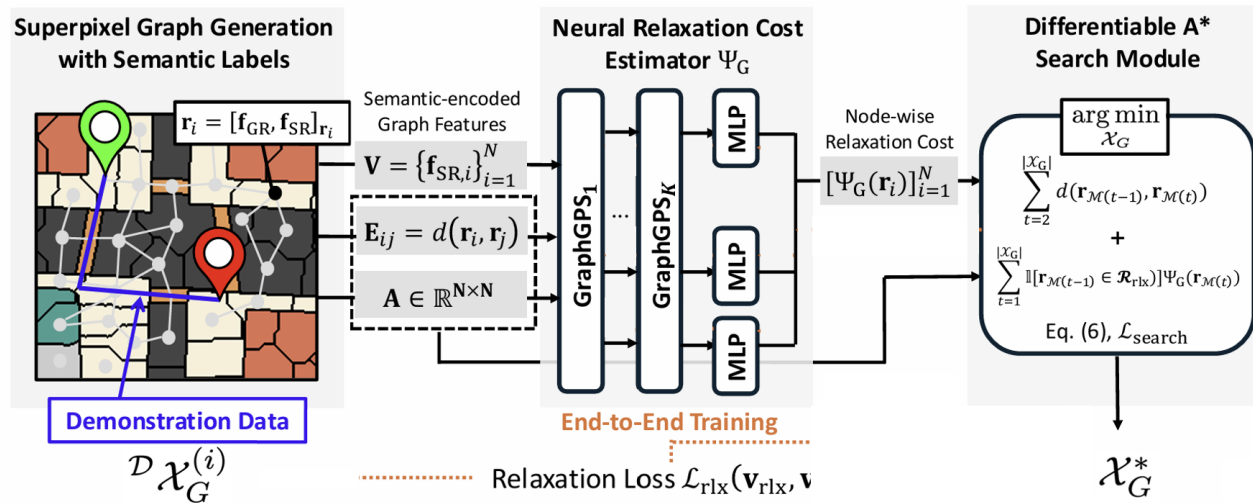


# SuReNav: Superpixel Graph-based Constraint Relaxation for Navigation in Over-constrained Environments



Can we find a best-effort solution that avoids all hard constraint regions while minimally traversing the least risky areas?

SuReNav learns to **automatically relaxes regional constraint** while planning a graph path.



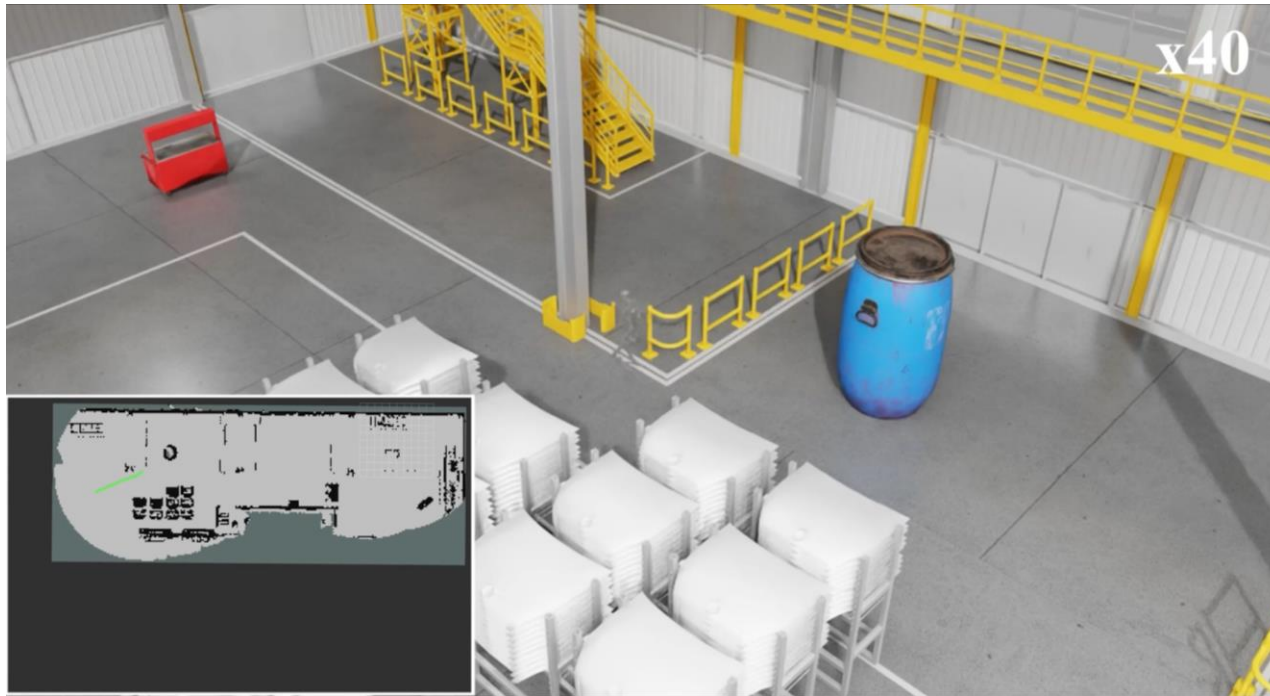
# SuReNav+: Human-like Humanoid Nav.



Woosong Hyungjin



We extend SuReNav by integrating **human preference-aware constraint relaxation** and navigation on a superpixel graph in and outdoor environments.



# Implicit Neural-Representation Learning for Elastic Deformable-Object Manipulations

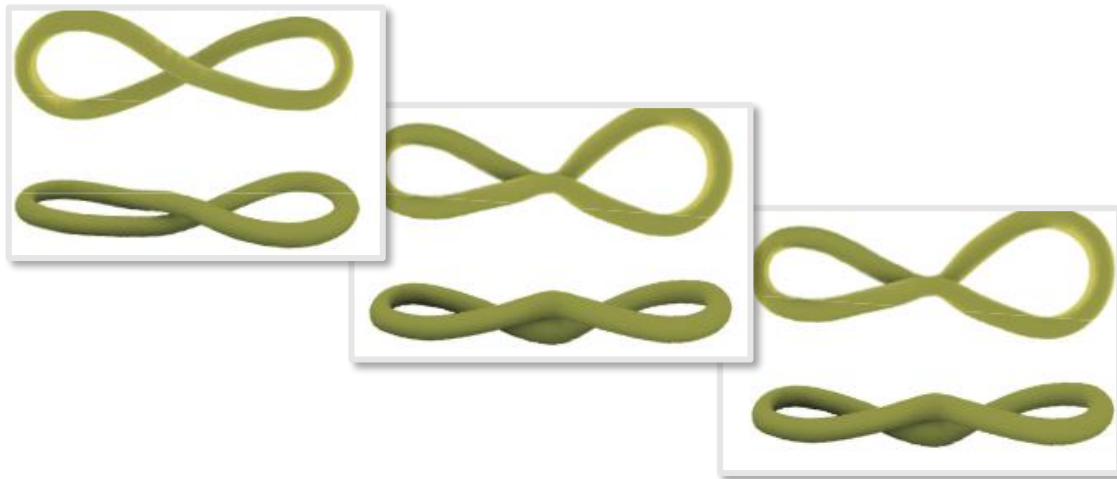


Minseok

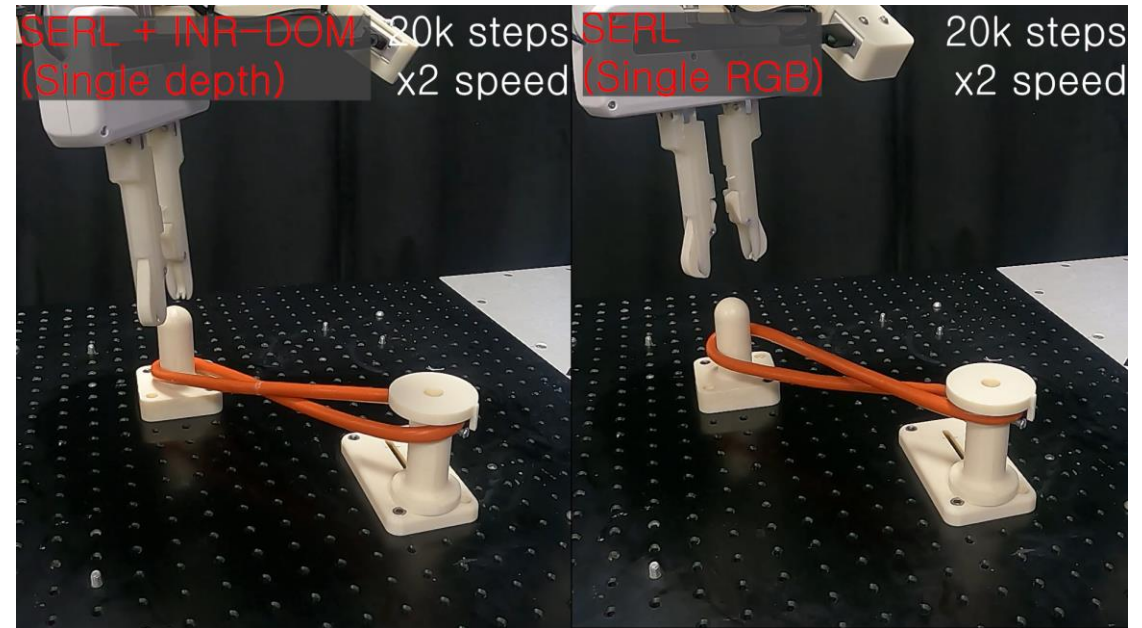
**Problem:** Deformable elastic-object manipulation in real world

**Challenge:** Partially-observable unlimited DoF → sampling complexity in RL

**Solution:** Implicit neural representation learning with contrastive RL



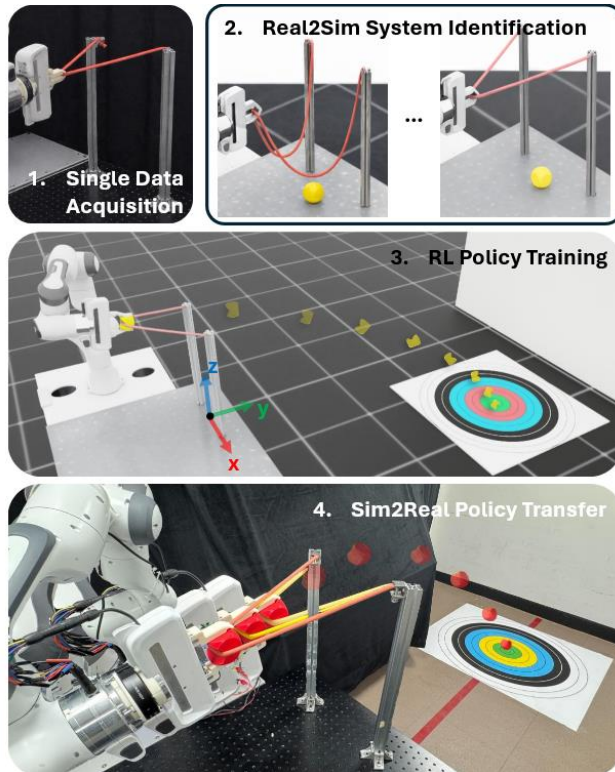
Visually similar, but geometrically largely different DLOs  
→ **precise & consistent representations**



# Sling2Sim2Real: One-Shot Elastic System Identification for Non-Destructive Slingshot Policy Learning

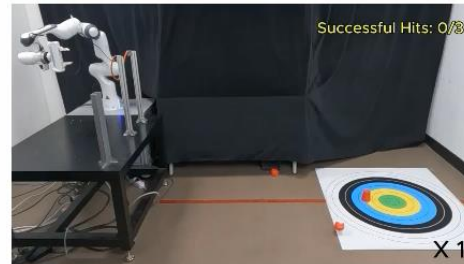
(under review)

Our **one-shot Real2Sim2Real framework** that identifies elastic parameters from a single non-destructive interaction and enables policy learning in simulation.

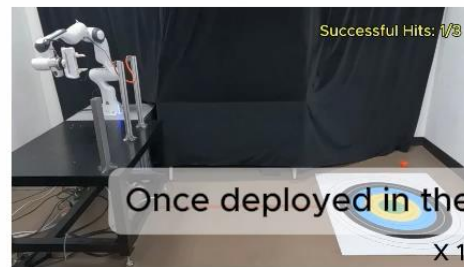


## Policy Transfer (Sim2Real)

Real2Sim2Real (Baseline)

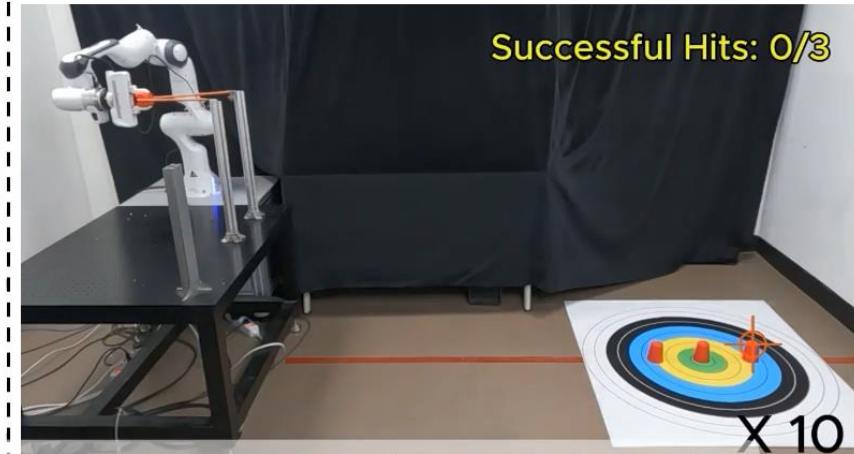


Package SI (Baseline)



Once deployed in the real world, the policy trained with the physical

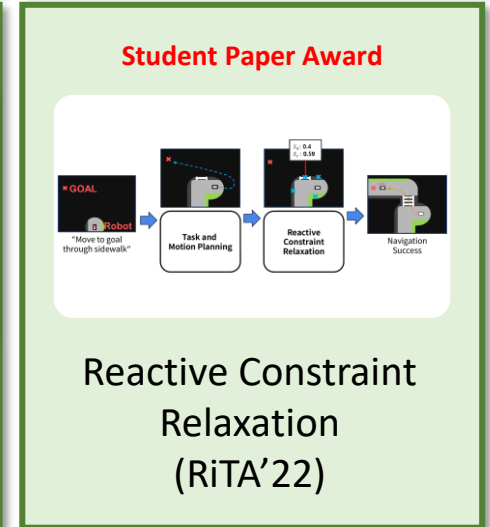
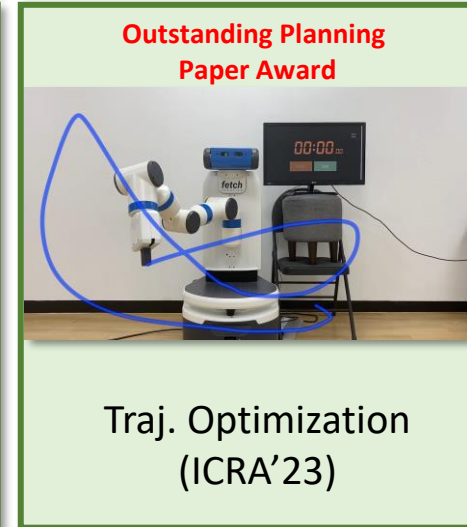
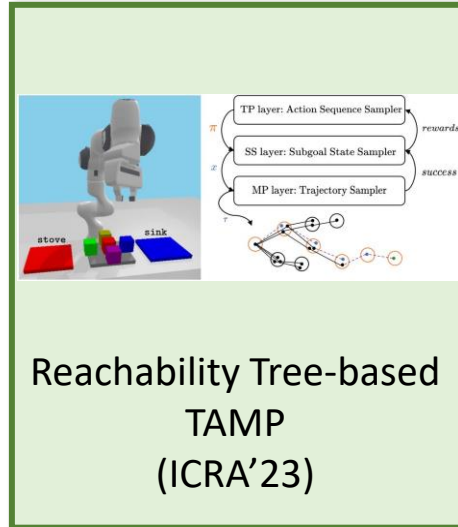
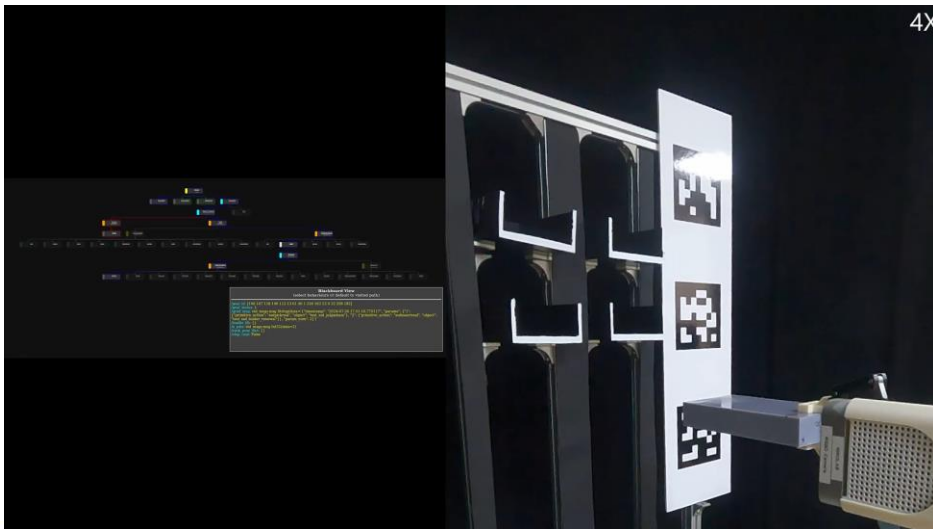
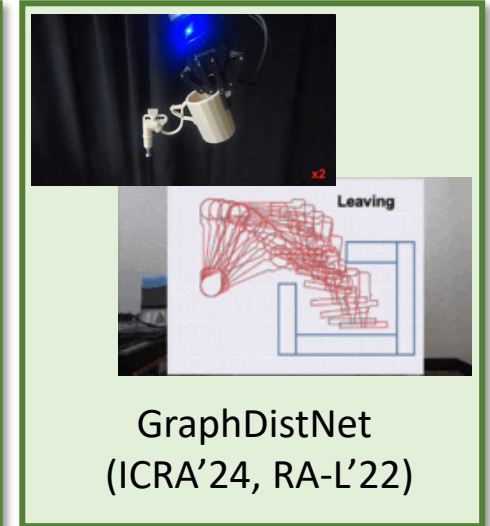
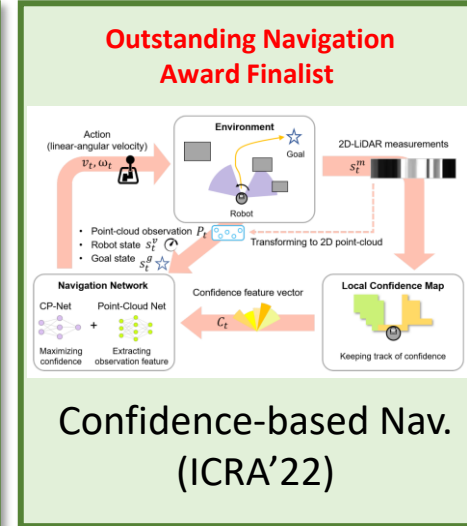
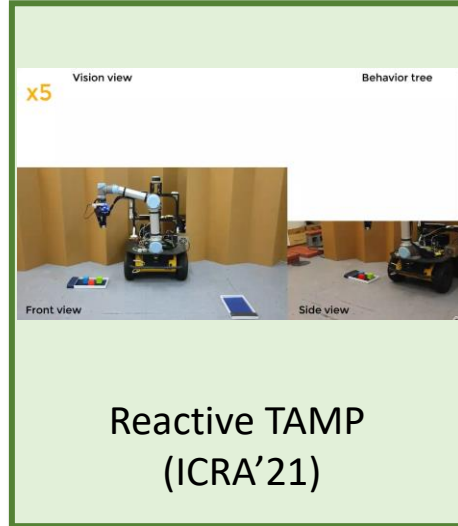
## Sling2Sim2Real (Ours)





# Task and/or Motion Planning

- Motion planning under environmental changes or occlusions
- Task (re)-planning under changes
- Online-recovery with behavior tree
- Learning-based collision check



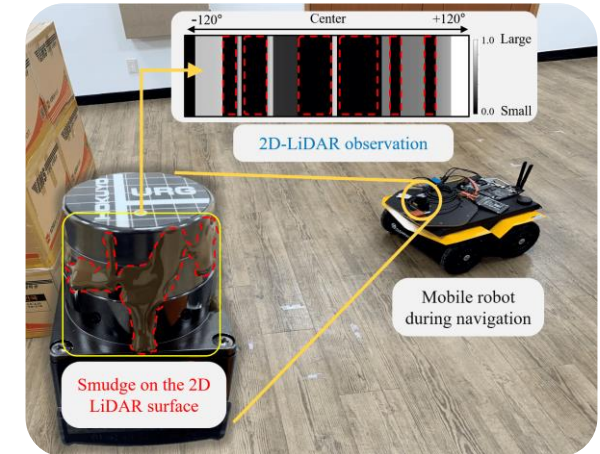
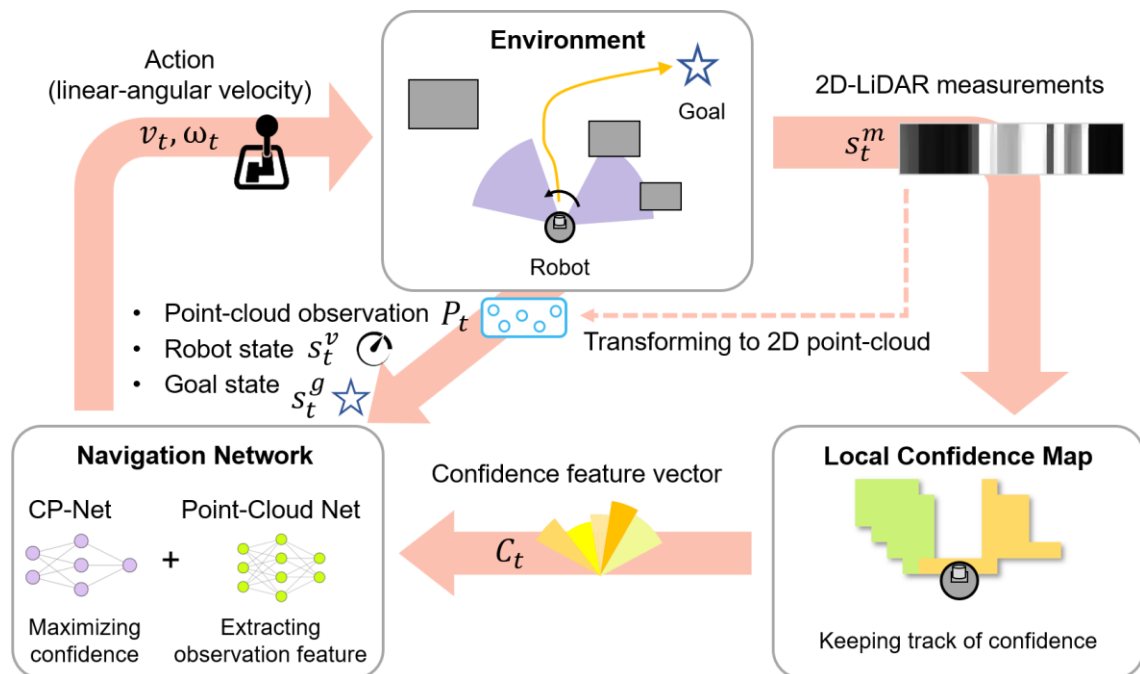
Recovery with behavior tree

# Confidence-Based Robot Navigation Under Sensor Occlusion

ICRA22 – **Outstanding Navigation Paper Finalist**

Consider unexpected occlusions on navigation sensors.

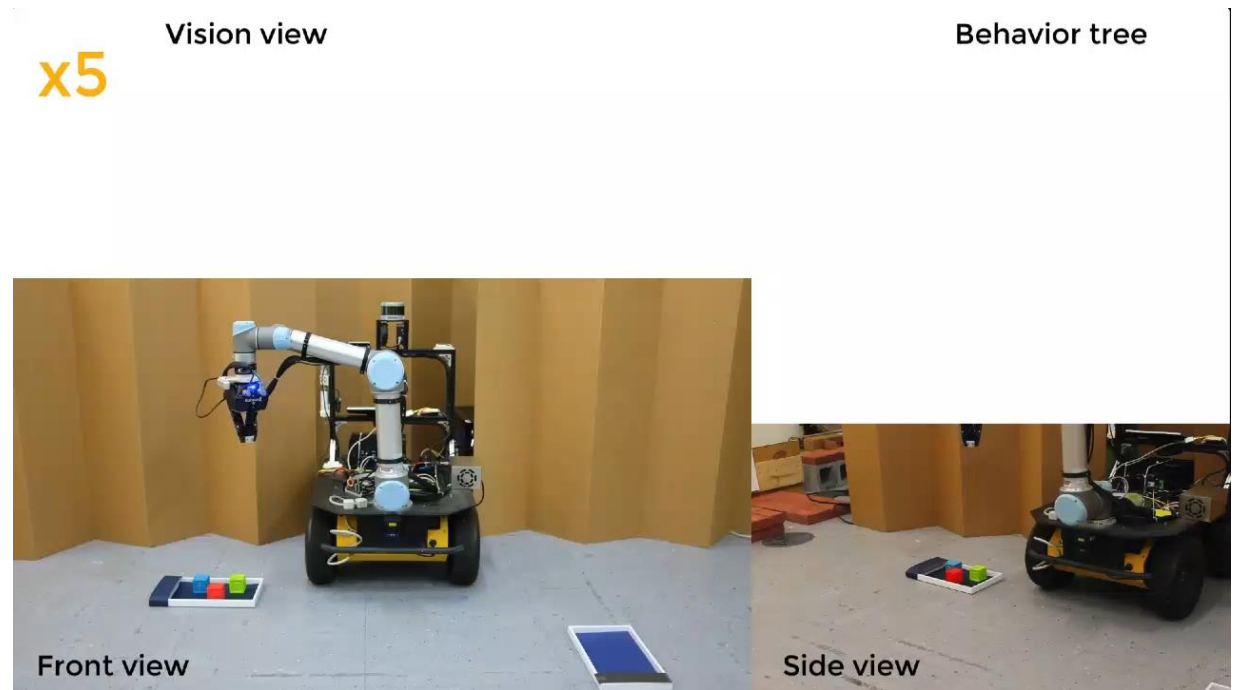
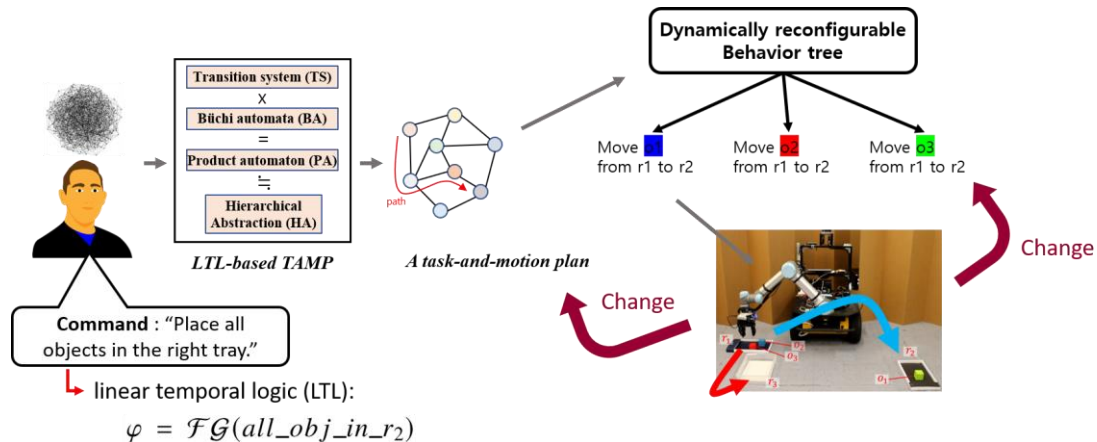
We propose a **confidence-based navigation method** trained through DRL. (Collaboration with Prof. Sung-eui Yoon)



# Task and Motion Planning under Changes

We propose **TAMP with Linear Temporal Logic (LTL)**. (Collaboration with Dr. Shen Li and Dr. Yoonchang Sung)

The hierarchical architecture for efficient replanning and execution to handle environmental changes



# ForeSight: Autoregressive Plan Monitoring for Efficient Long-Horizon Replanning and Execution



Yeseung



Gyungjin

(under review)

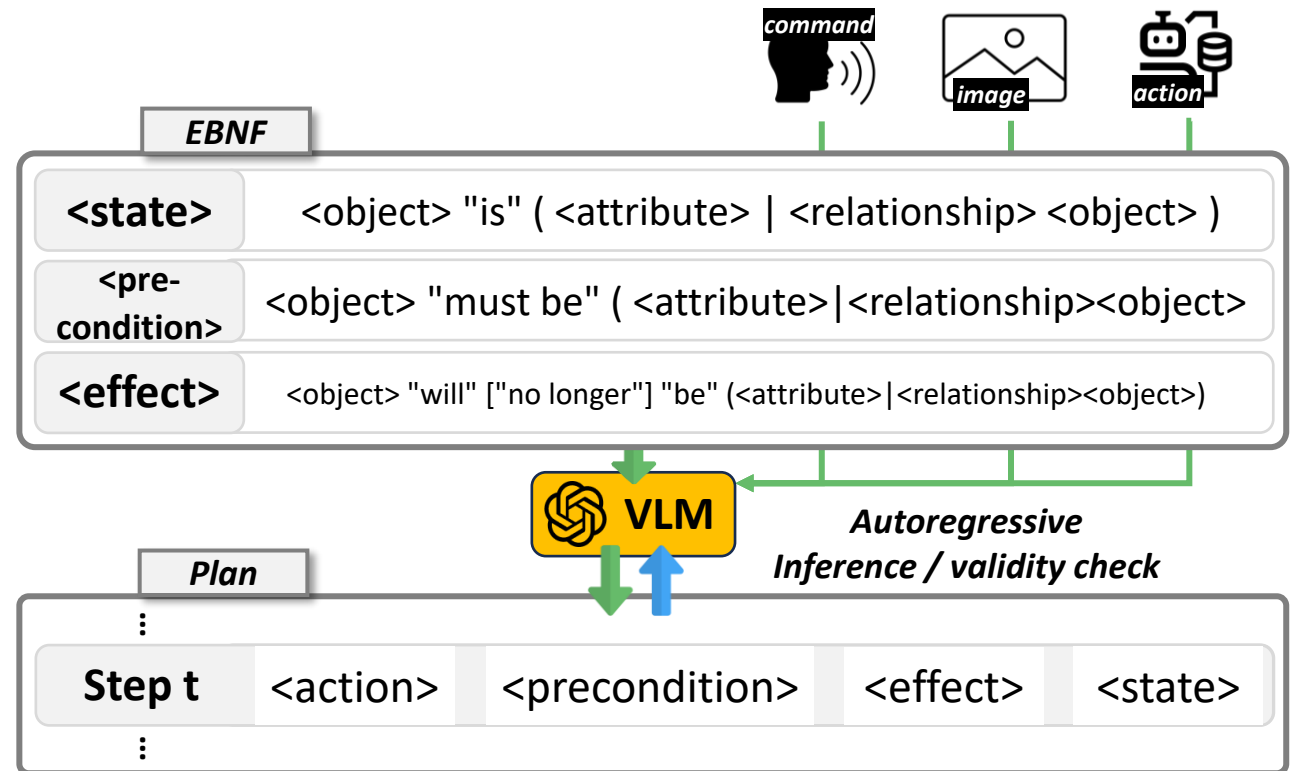
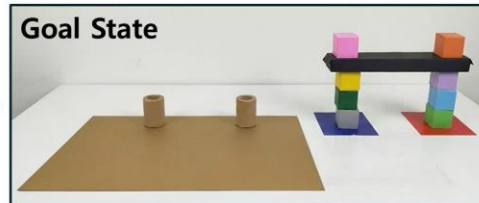
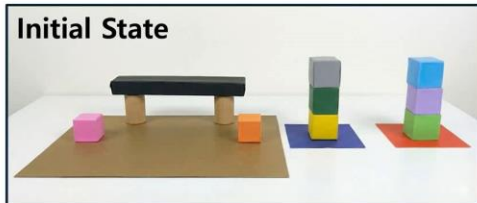
- **Extended Backus–Naur Form (EBNF)-based Chain-of-State prompt**

→ Consistent representation for long-horizon task planning

- **Autoregressive plan validity checking**

→ Early detection + Reactive planning

Task: Build a tower using cubes and one bar.

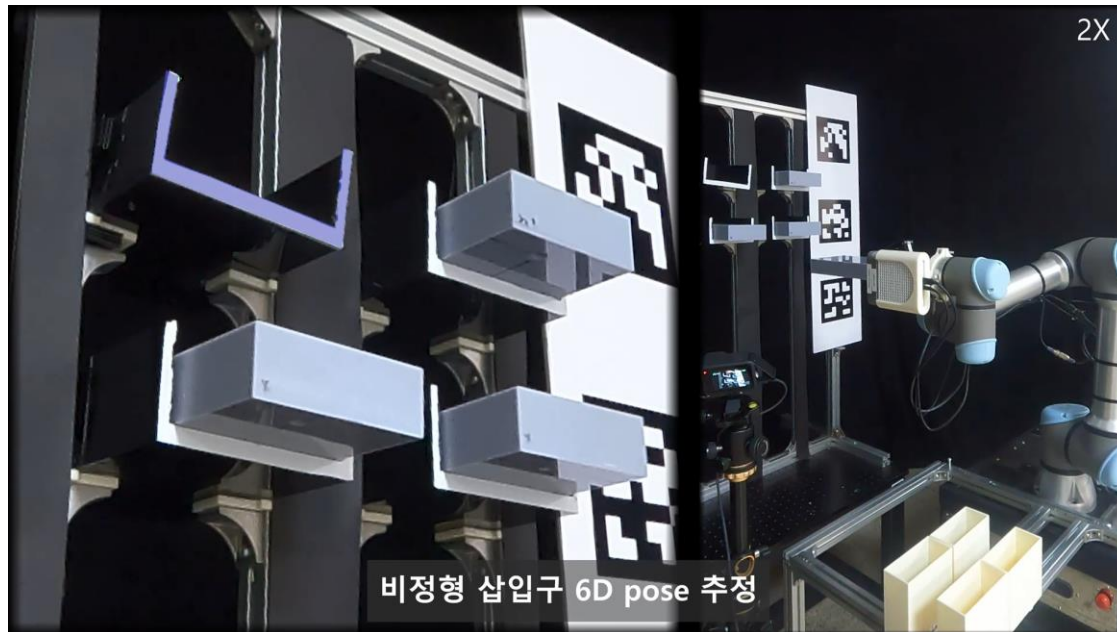


# Multimodal Task-and-Motion Planning for multi-pack insertation task

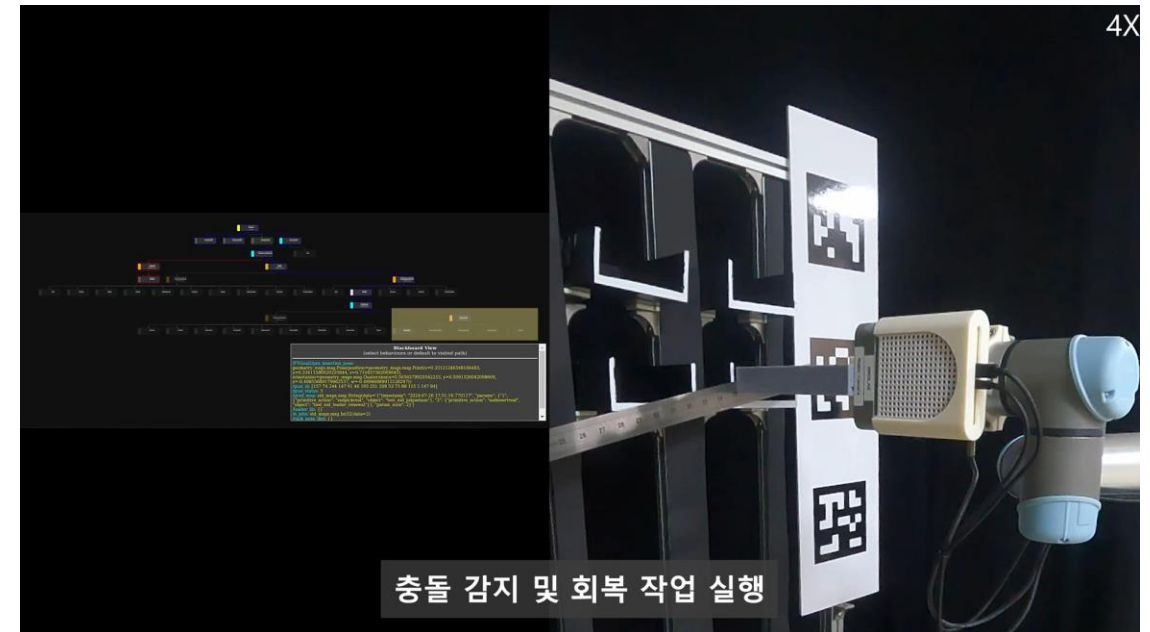
SEMES

## Task planning and multi-modal recovery using behavior tree

- 6D pose estimation for multi holes in real-world
- Safe motion planning and simulation



Automatic HDD/SDD quality check system



Recovery with behavior tree



# Humanoid Precision Task Execution

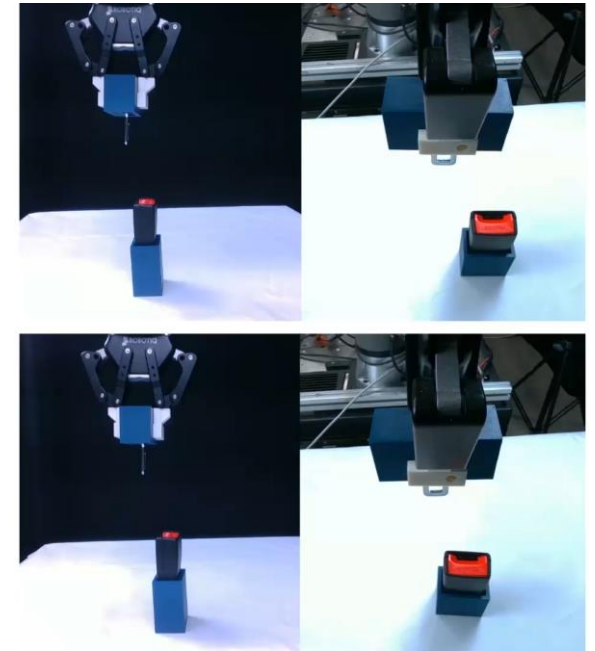
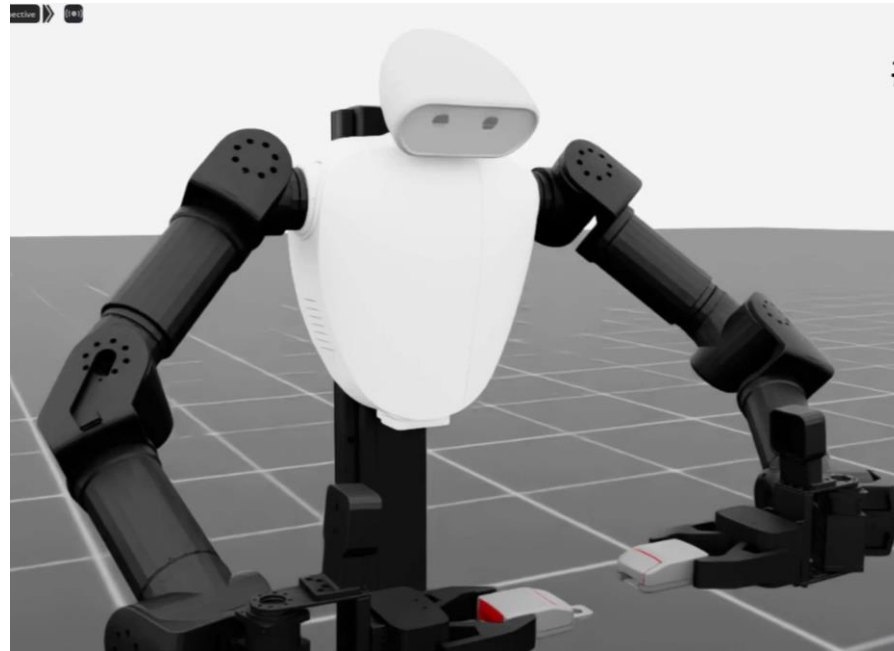


Jun



Jaehyeong

We develop a behavior tree–based system for executing precision tasks on humanoid robots.



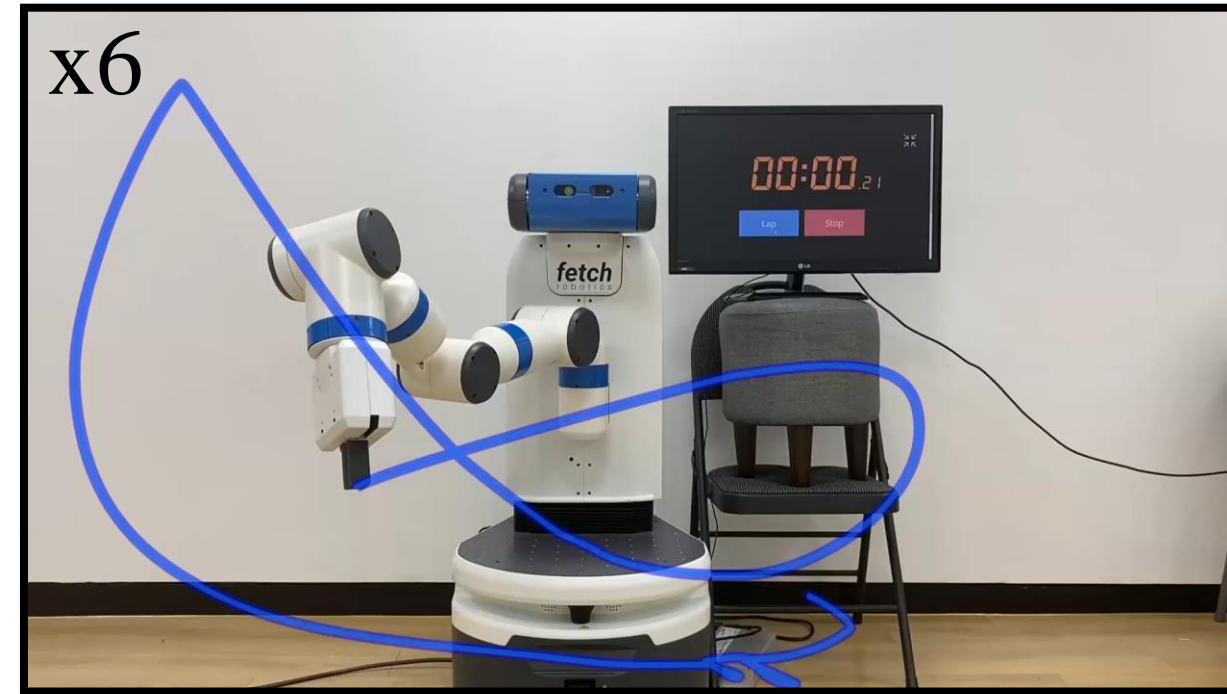
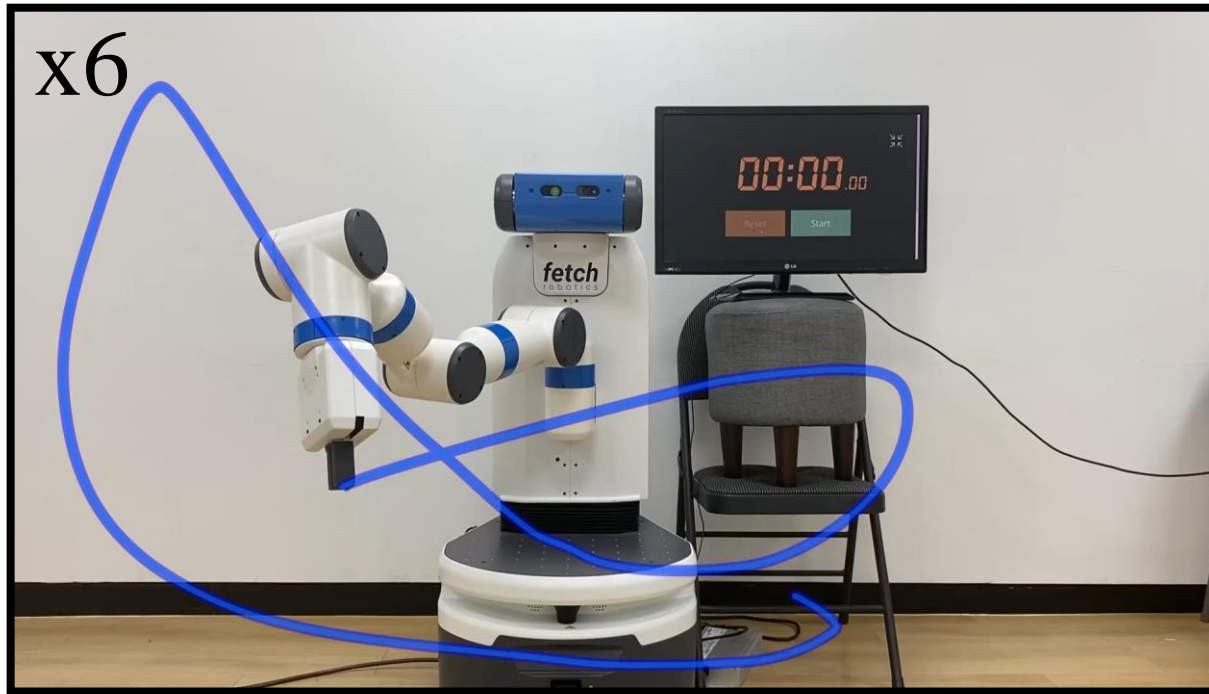
# Learning-based Initialization of Trajectory Optimization

ICRA23 – **Outstanding Planning Paper Award**

- Benchmark name: 'Random #64' | Trajectory optimizer: TORM [Kang et al. 2020]

with **RL-ITG** (ours)

with **Greedy** (baseline)



- Total execution time: **00:46.76**

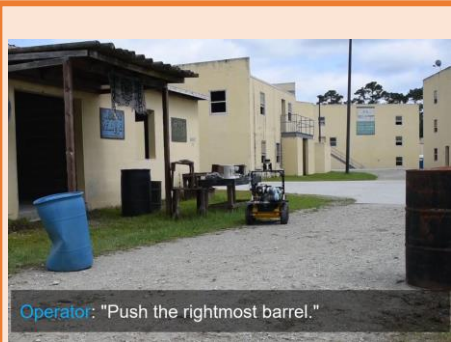
about **220%** faster than *Greedy*

**01:43.16**

min : sec



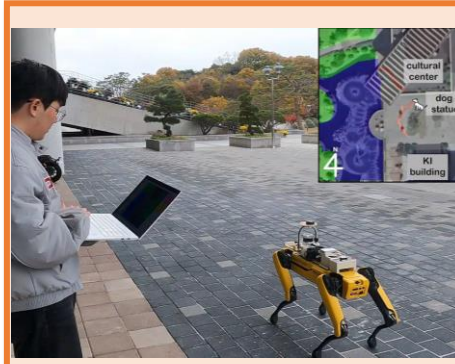
# Natural-Language Grounding



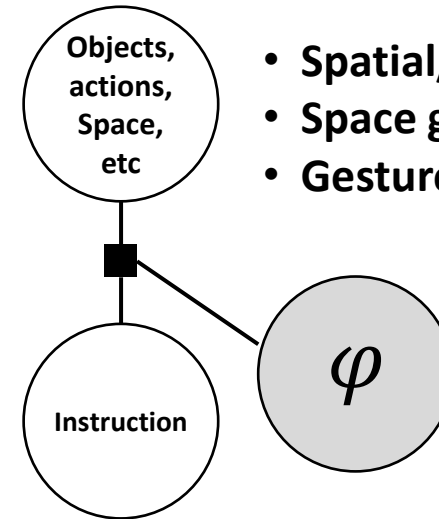
Multi-modal Estimation  
& Communication  
(IJRR'20)



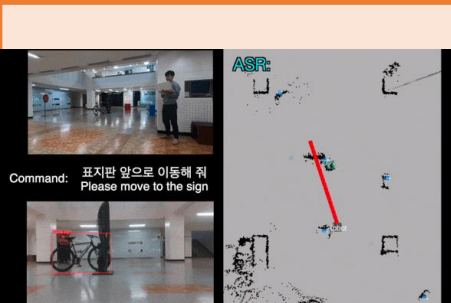
LINGO-Space for  
Language Grounding  
(AAAI'24)



Language-guided  
Navigation  
(Proj. 23-24)



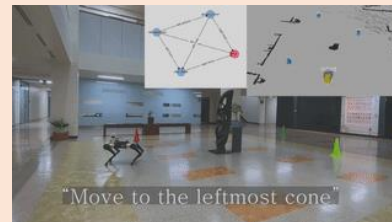
- Spatial/space grounding
- Space grounding-based VLA
- Gesture grounding



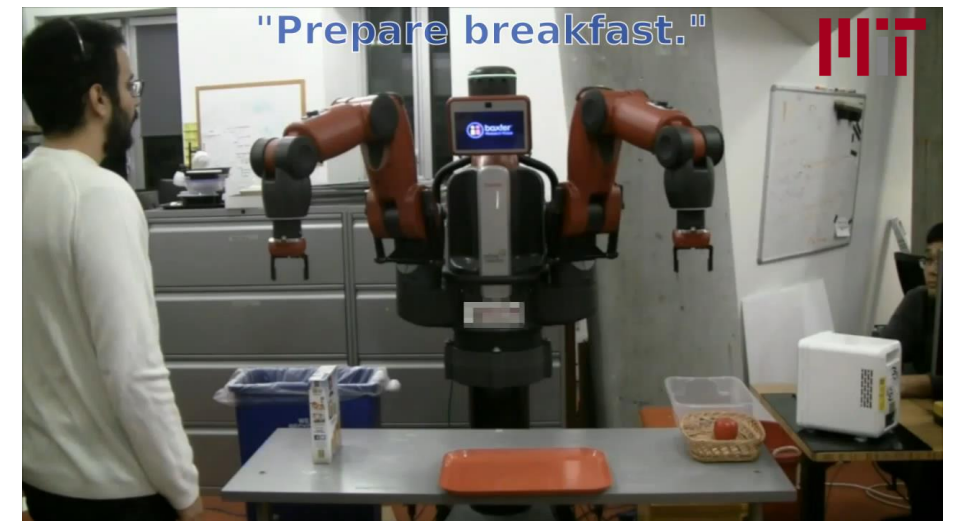
SGGNet2: Speech-guided Navigation  
(RO-MAN'23)



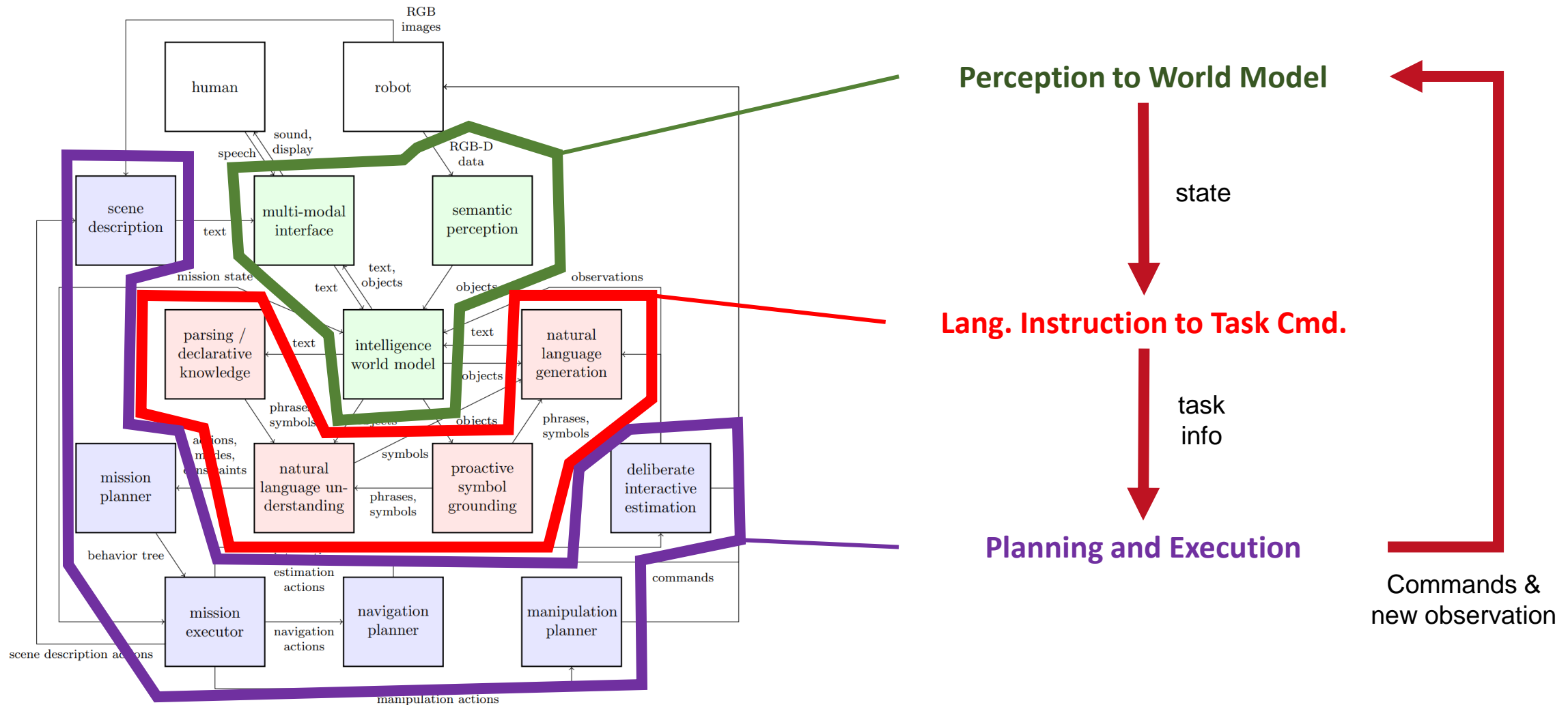
Grounding Robot Plans  
(CoRL'18)



Language-guided  
Navigation  
(Ro-MAN'23)



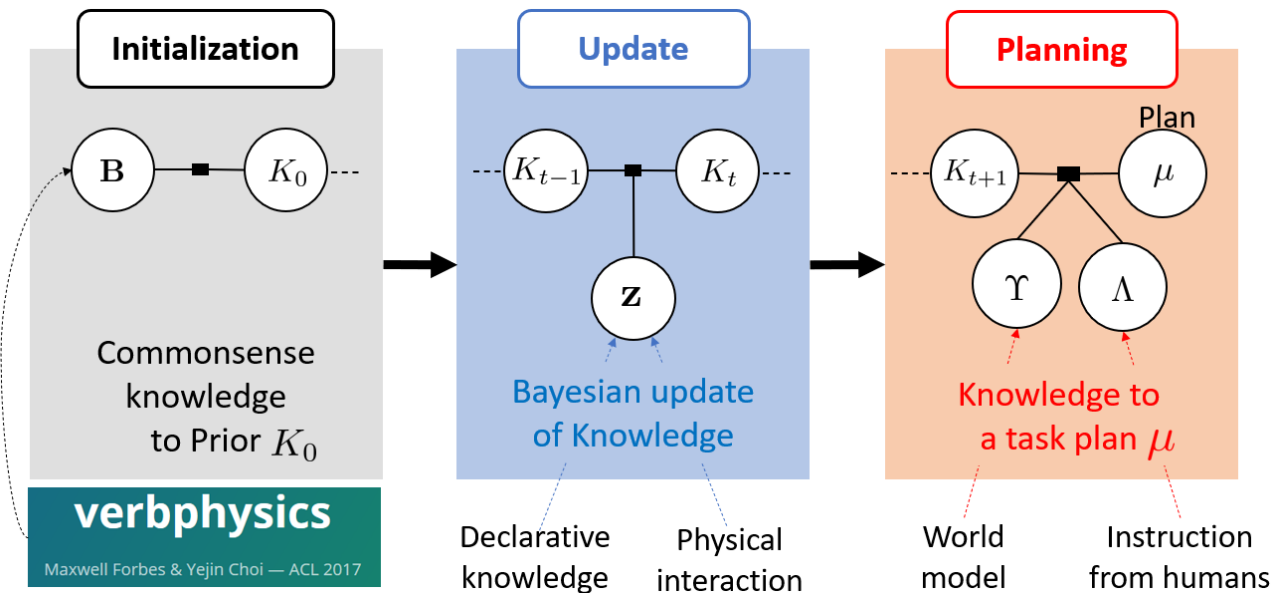
# Intelligence Architecture for Grounded Language



# Multimodal Estimation and Communication

## Natural language-driven mobile manipulation

- Infer semantic knowledge of the world using a Bayesian framework
- Correct invalid knowledge and generate a safe plan

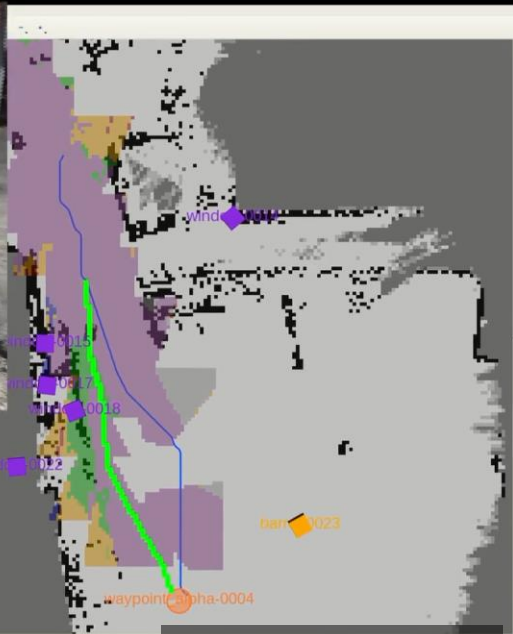




KAIST



x4



Time: ROS Time: 1570745340.47 ROS Elapsed: 128.74 Wall Time: 1570745340.50 Wall Elapsed: 128.64  
 Reset Left-Click: Move XY, Middle-Click: Rotate, Right-Click: Zoom, Shift: More options.

[FR22, IJRR20]



“교직원 회관 남동쪽으로 가”

1X

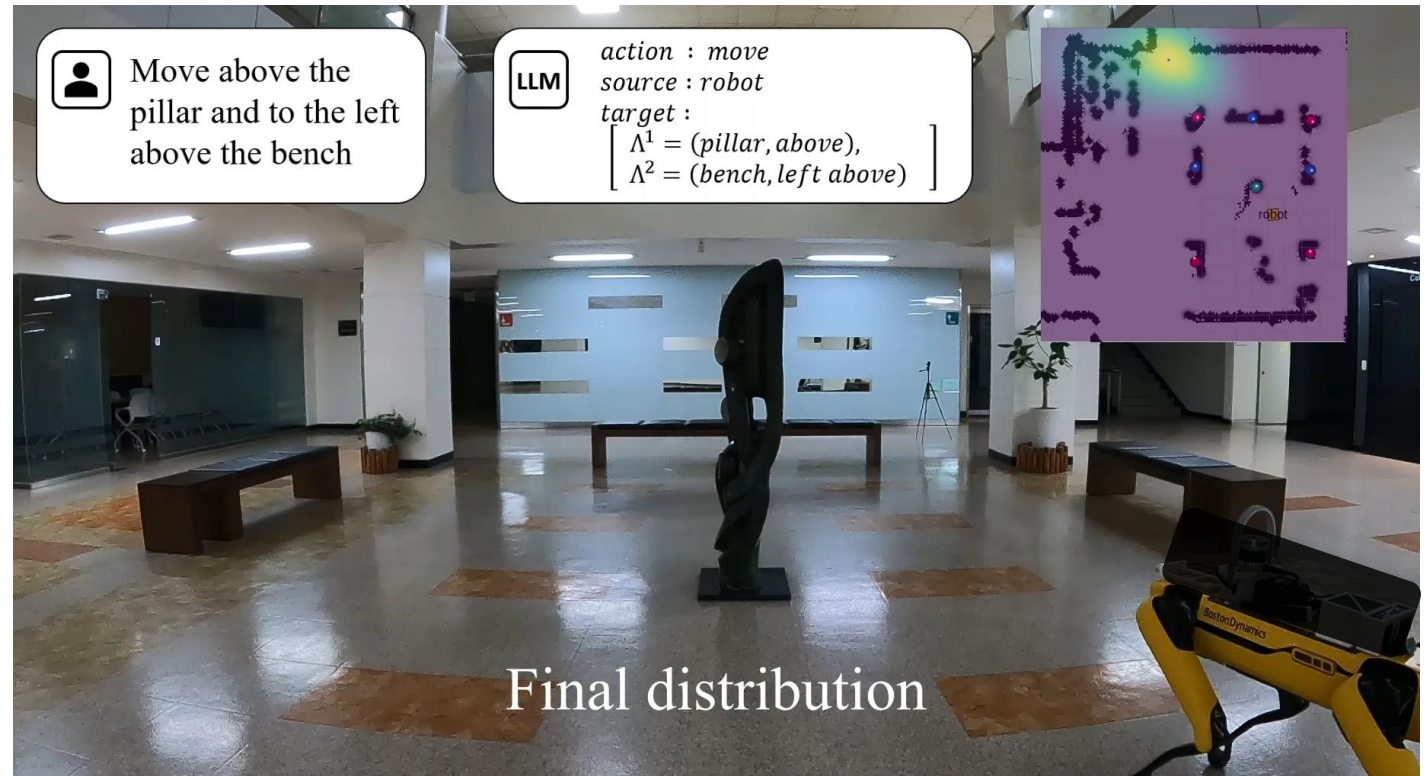
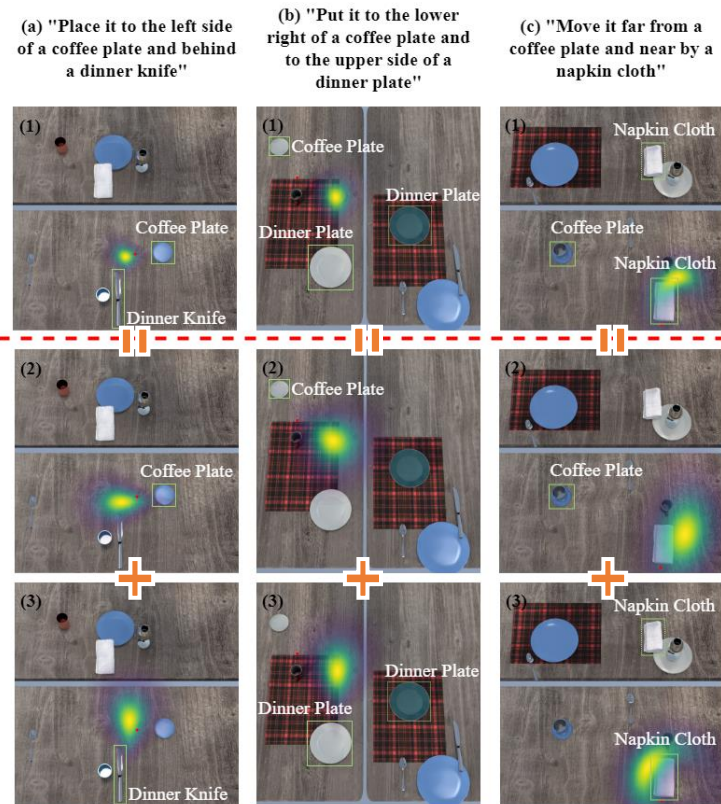


Dohyun

# Model-based Space Grounding

## LINGO-Space: Language-Conditioned Incremental Grounding

- Our mixture of polar distributions represents positions from combined instructions.
- **The probabilistic grounding model** infers multiple placement candidates !!





Dohyun Junhyeong

(under review)

# Model-free Space Grounding

## C2F-SPACE: Coarse-to-Fine Space Grounding for Spatial Instructions using Vision-Language Models

(Collaboration with Dr. Rohan Paul, IIT Delhi)



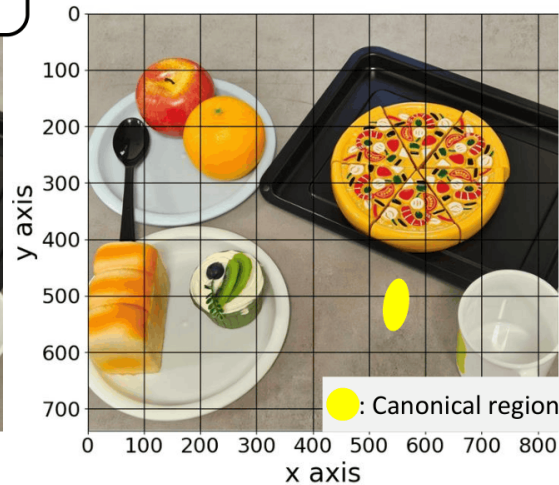
KAIST

- Estimates an approximated yet spatially consistent region using a VLM.
- Refines the region to align with the local environment through superpixelization

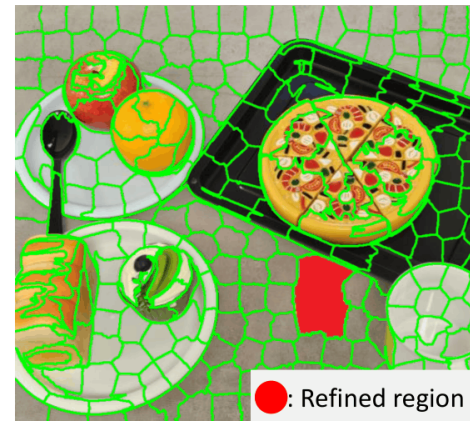
Place the spoon to the **right** of the cupcake at **twice** the distance between the cup and the pizza.



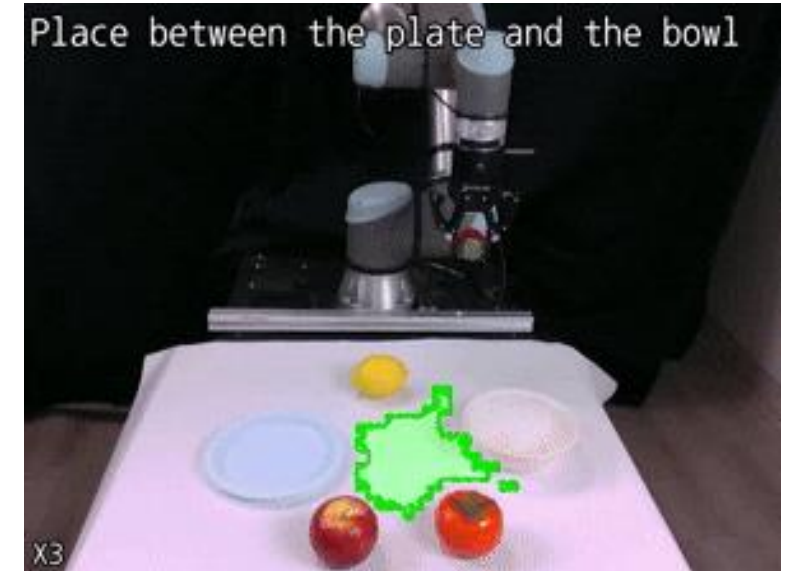
**Stage 1:** Space reasoning with physical & semantic validators



**Stage 2:** Superpixel based space refinement



Two stage grounding



Pixel-grounding with robot manipulation

# Robots may not work as we want

## Take home message

**Next-token prediction** is now a core part of robot intelligence, but it does not truly model human intelligence yet! (from Prof. Yann LeCun)



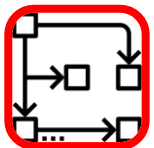
Lack of understanding physical world



Limited persistent memory

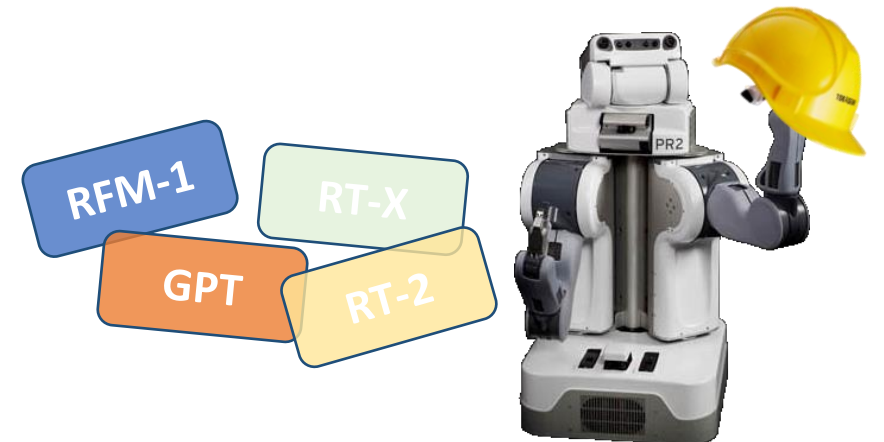


Poor reasoning



Absence of planning ability

Put specialized intelligence!



# Thank you. Any questions?



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