

**3<sup>rd</sup> Class / Jan 15 (Wed)**

# **Modern Robot Learning: Hands-on Tutorial**

Haoshu Fang, Younghyo Park, Jagdeep Bhatia, Lars Ankile, Pulkit Agrawal

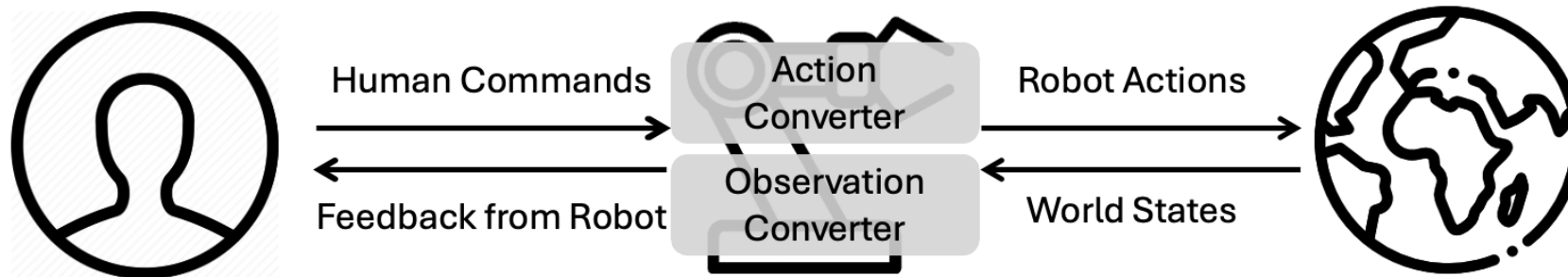


# Last Week...

## Robot Teleoperation

**4 Key Elements** of  
Teleoperation System

1. **Designing** command space for humans
2. **Converting** commands to robot actions
3. **Designing** feedback space for humans
4. **Converting** robot perceptions to human feedback



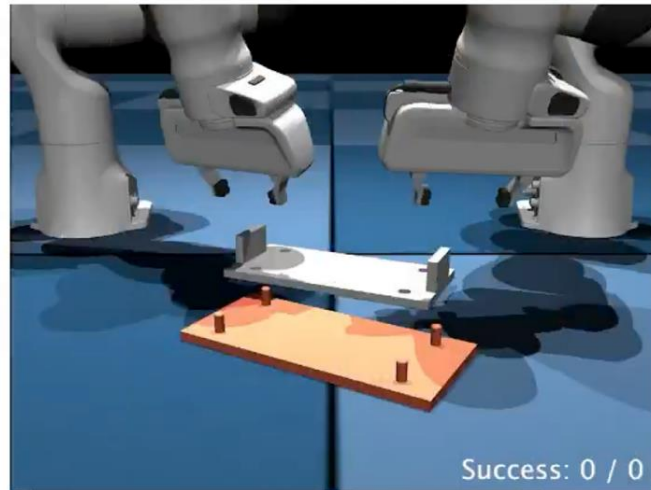
*Most of the robot datasets are created by “teleoperation”*

# Last Week...

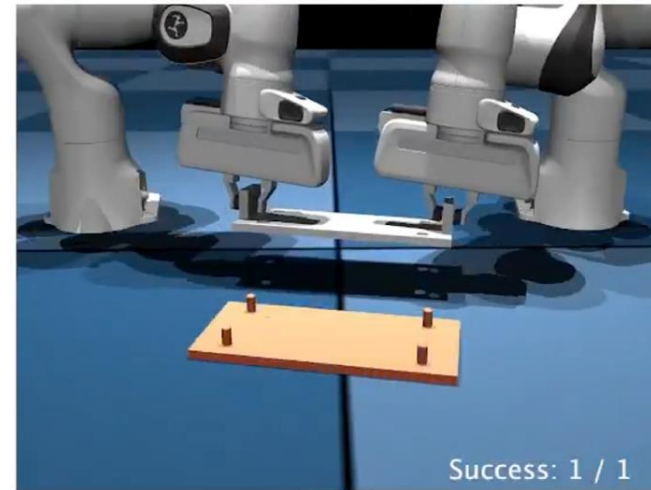
## Role of Simulation

Demos collected in simulation supports last-mile performance improvement through **RL finetuning**.

Imitation only



With a sprinkle of reactivity



*How **simulation** can help data collection*

Today



Collecting Robot  
Data in Virtual World



Learning from  
Videos



Interaction  
Interfaces

# Today

- Learning from videos
  - Policy learning from human video
    - Observation mismatch
    - Get action
  - World modeling
    - Data and learning
    - How to apply to robots
- Drawbacks of three different methods
- Interaction interface Case Studies: In-Depth Analysis
  - Policy learning with interaction interface

# Today

- **Learning from videos**
  - Policy learning from human video
    - Observation mismatch
    - Get action
  - World modeling
    - Data and learning
    - How to apply to robots
- Drawbacks of three different methods
- Interaction interface Case Studies: In-Depth Analysis
  - Policy learning with interaction interface

# Learning from videos

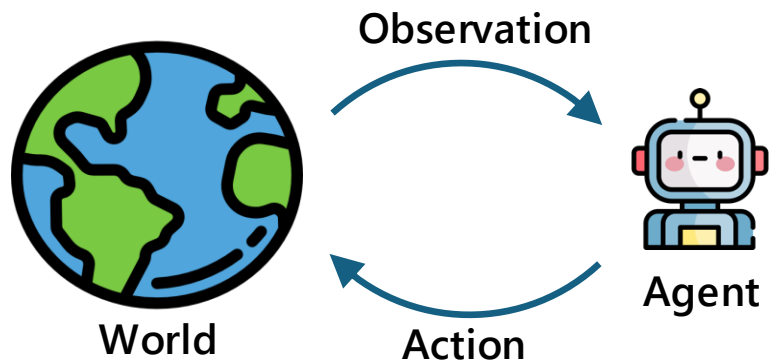
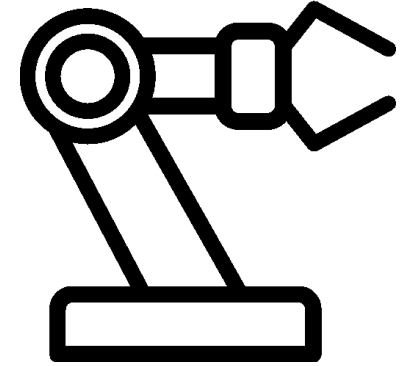
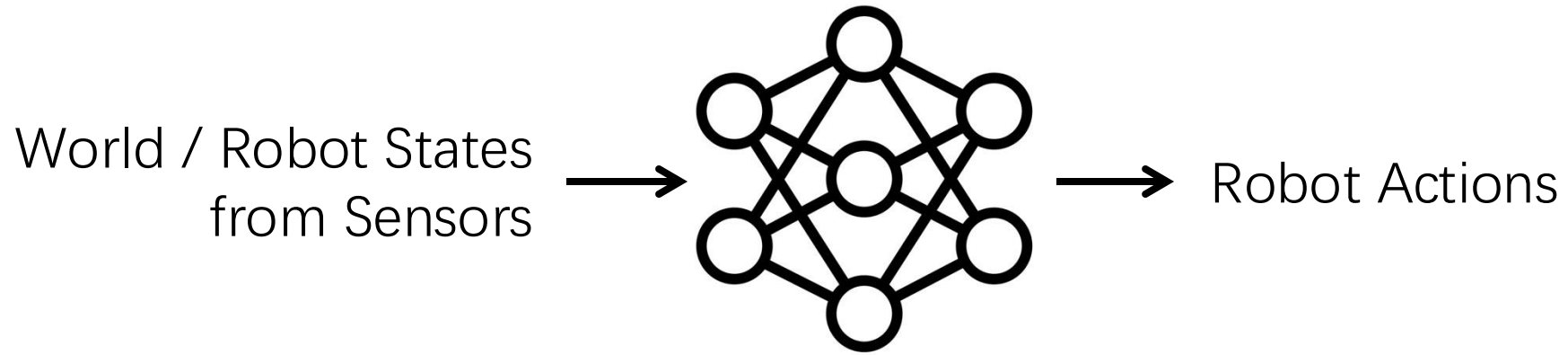
[A] Policy learning from human videos



[B] World modeling from all videos

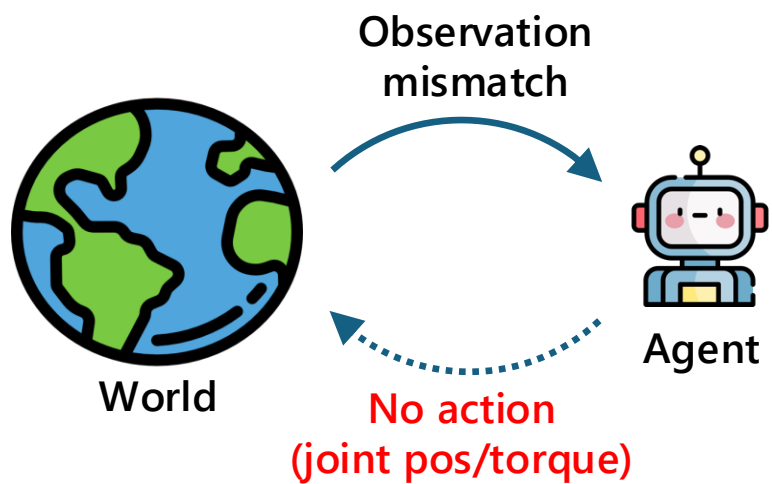
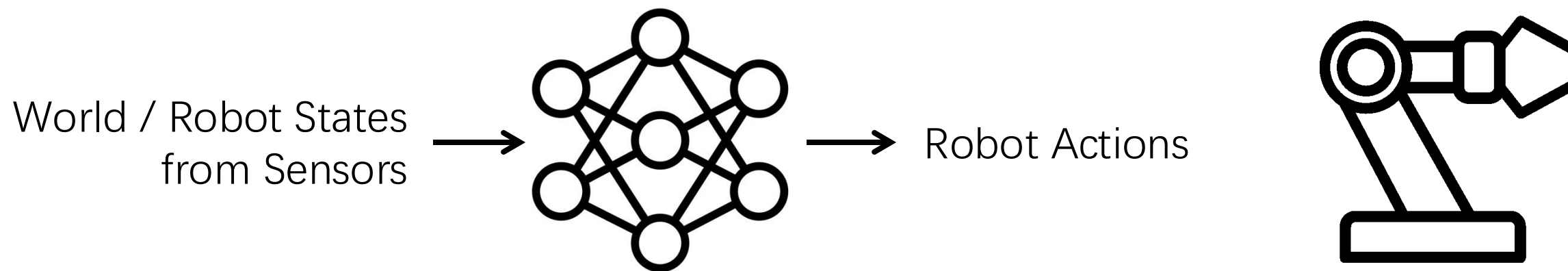


# Policy learning



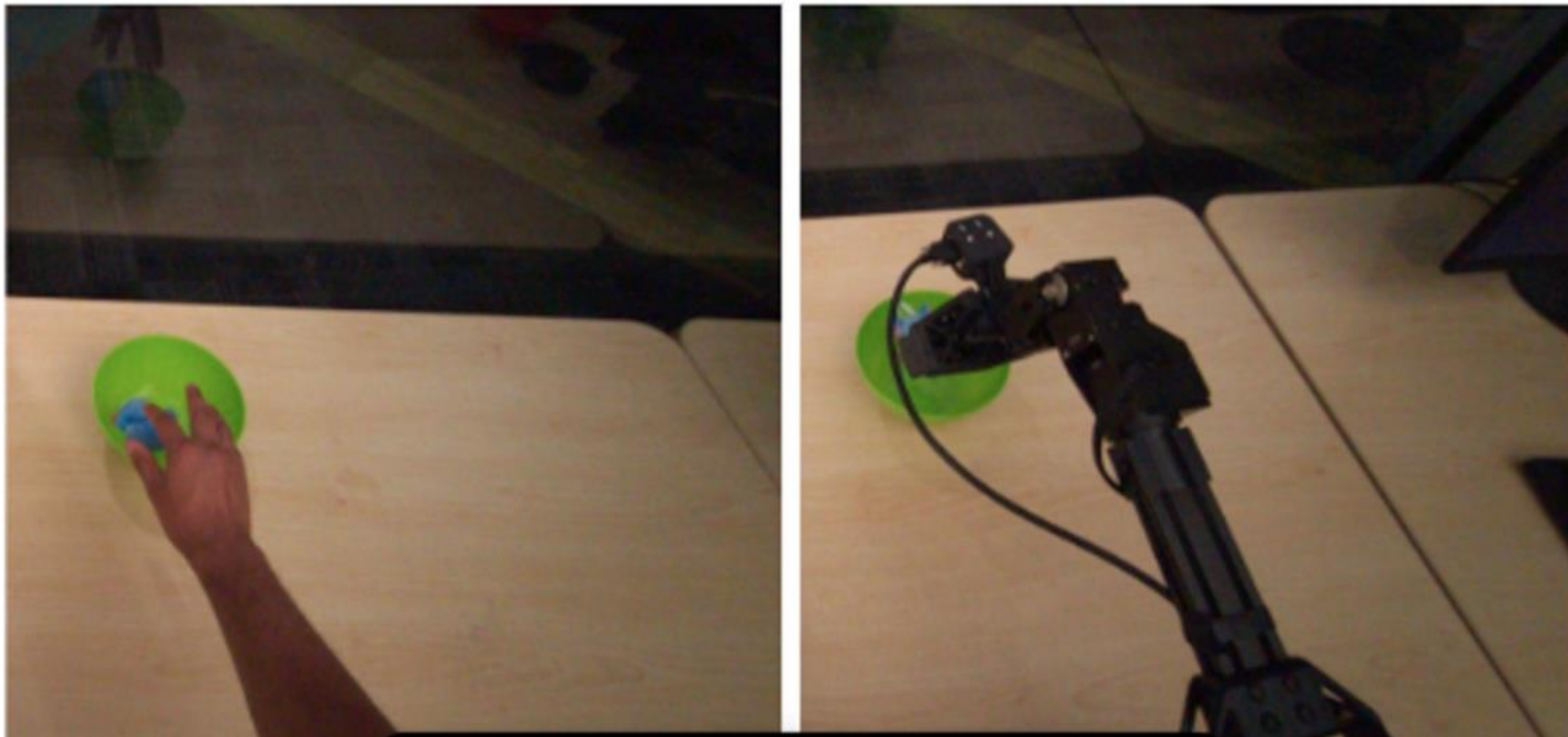


# Policy learning

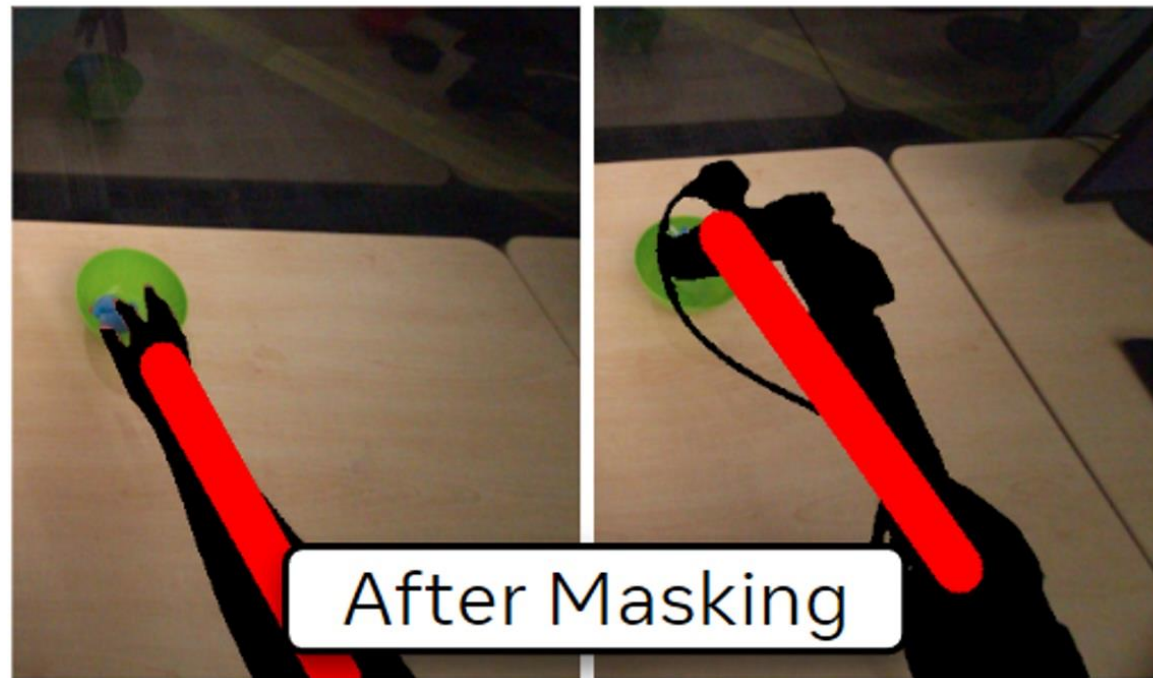


**EPIC-KITCHENS**

# Observation mismatch



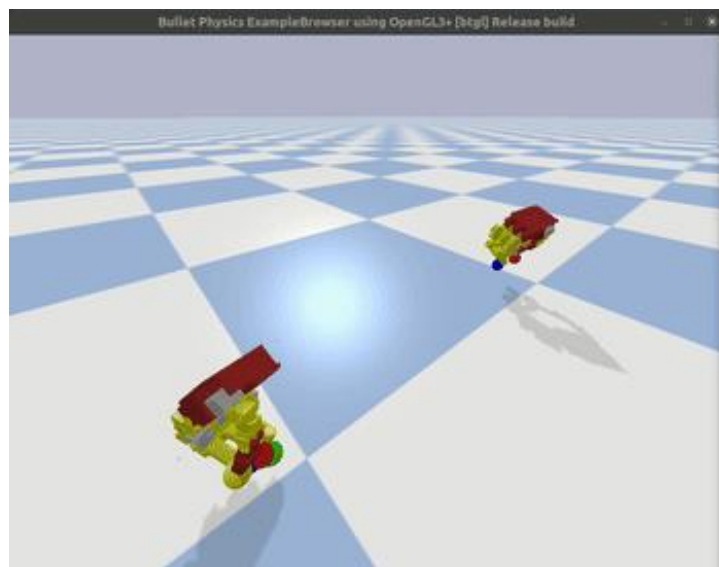
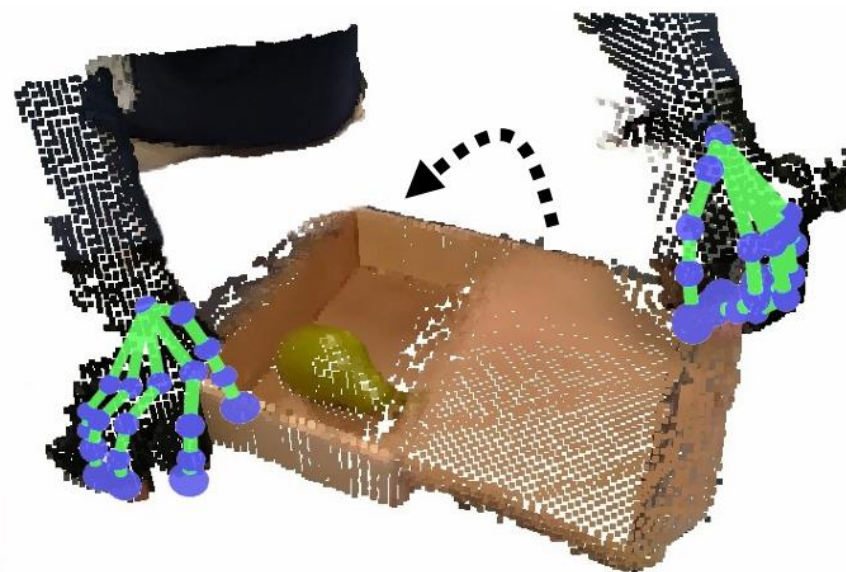
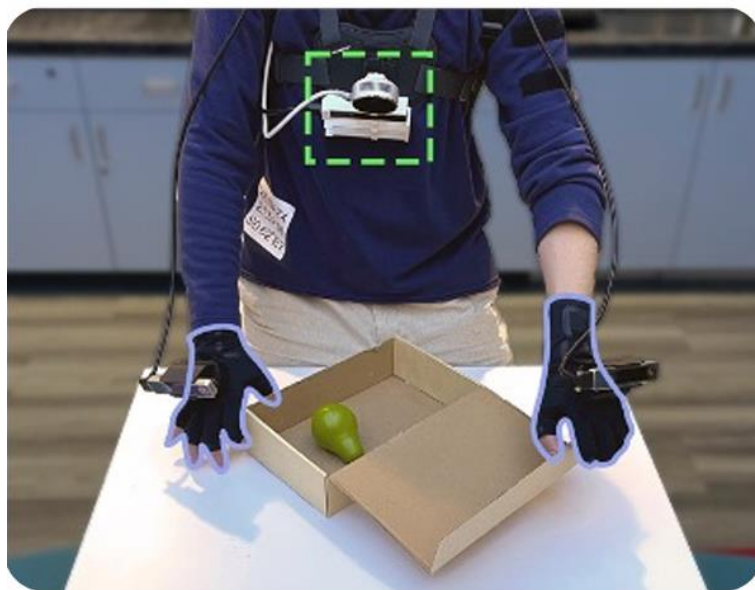
Observation retargeting



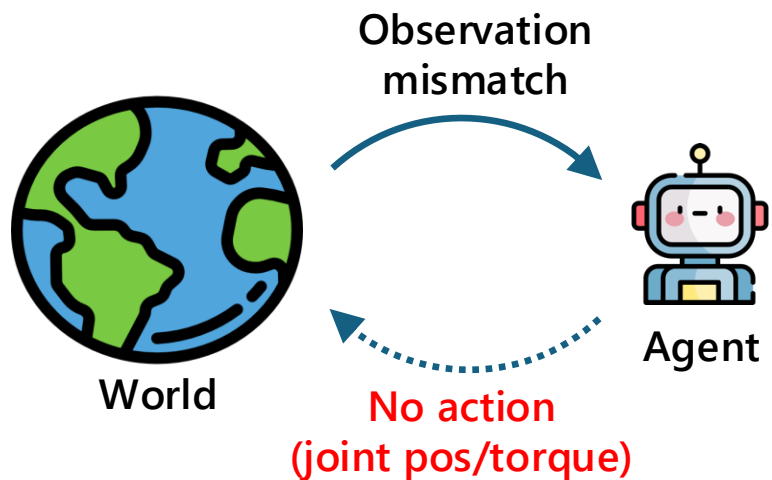
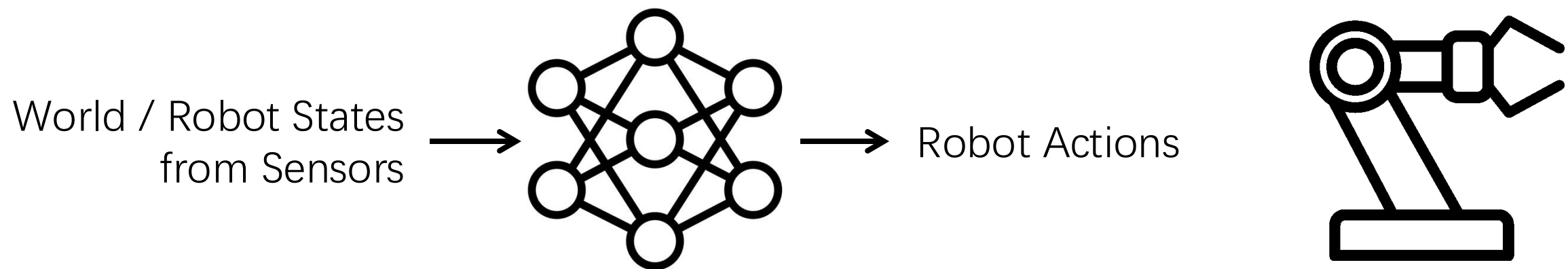
Mask out both human and robot arms,  
replace with a red line



Overlay based  
on hand pose

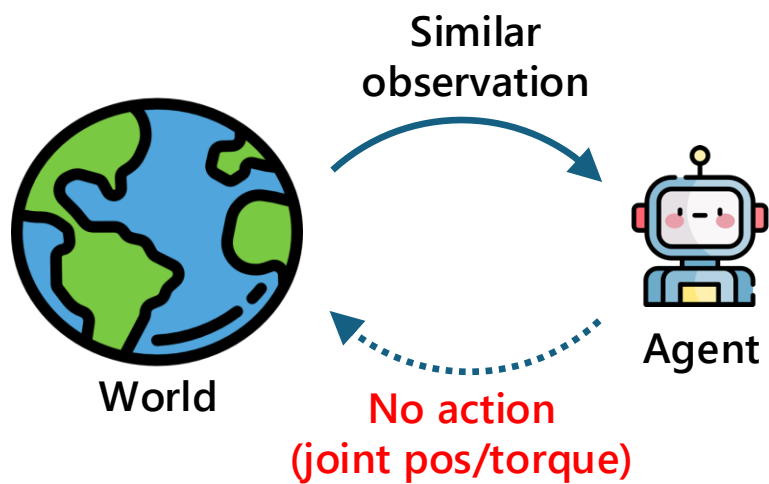
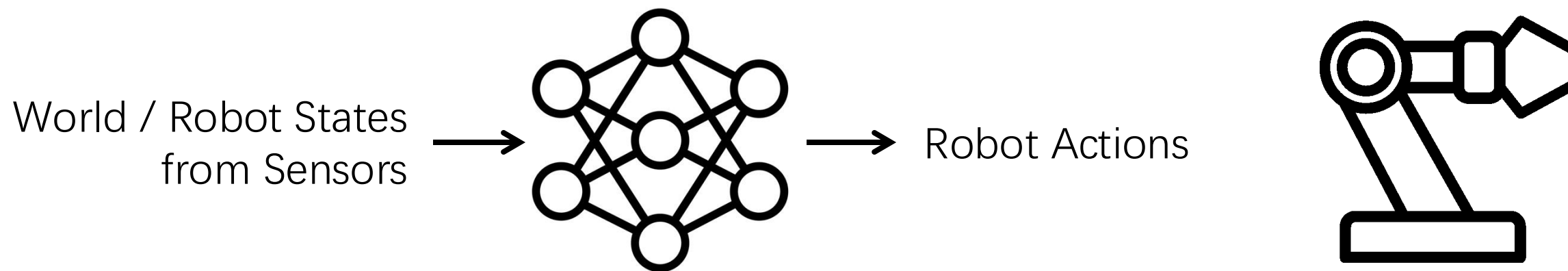


# Policy learning



By retargeting, we can get an observation close to the robot would have.

# Policy learning

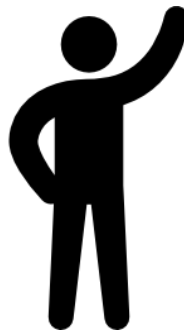


**EPIC-KITCHENS**

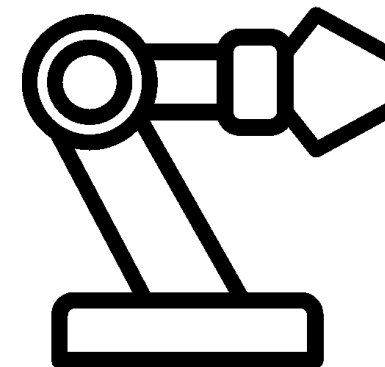
# Action generation from videos



Human video



Human action



Robot action

$$D_1 = \{(s_0, s_1, \dots, s_n)\} \quad D_2 = \{(s_0, a_0, s_1, a_1, \dots, s_n)\} \quad D_3 = \{(s_0, \hat{a}_0, s_1, \hat{a}_1, \dots, s_n)\}$$

$$a_t = \mathbf{g}(s_t, s_{t+1})$$

$$\hat{a}_t = \mathbf{h}(a_t)$$

# From videos to human action



Human video



Human action

$$D_1 = \{(s_0, s_1, \dots, s_n)\} \quad D_2 = \{(s_0, a_0, s_1, a_1, \dots, s_n)\}$$

$$a_t = \mathbf{g}(s_t, s_{t+1})$$



# From videos to human action



AlphaPose



OpenPose

## Human pose estimation



ITALIAN HAND GESTURE

I AM FULL



ITALIAN HAND GESTURE:

Right hand I AM FULL

Top view



Left hand

Top view



Reconstructing Hands in 3D with Transformers

# From videos to human action

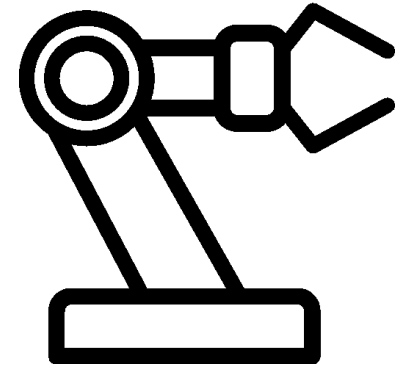


## 3D Human pose estimation

# Action generation from videos



Human action

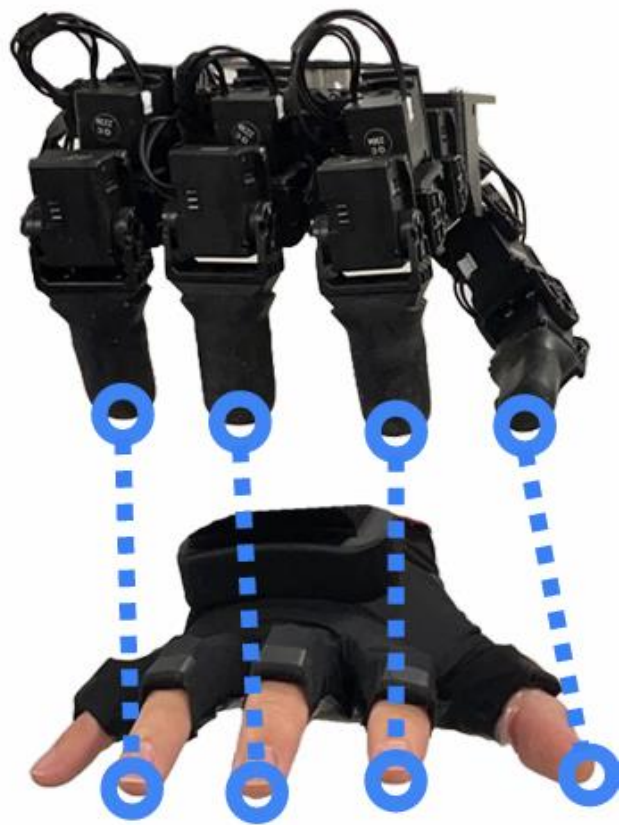


Robot action

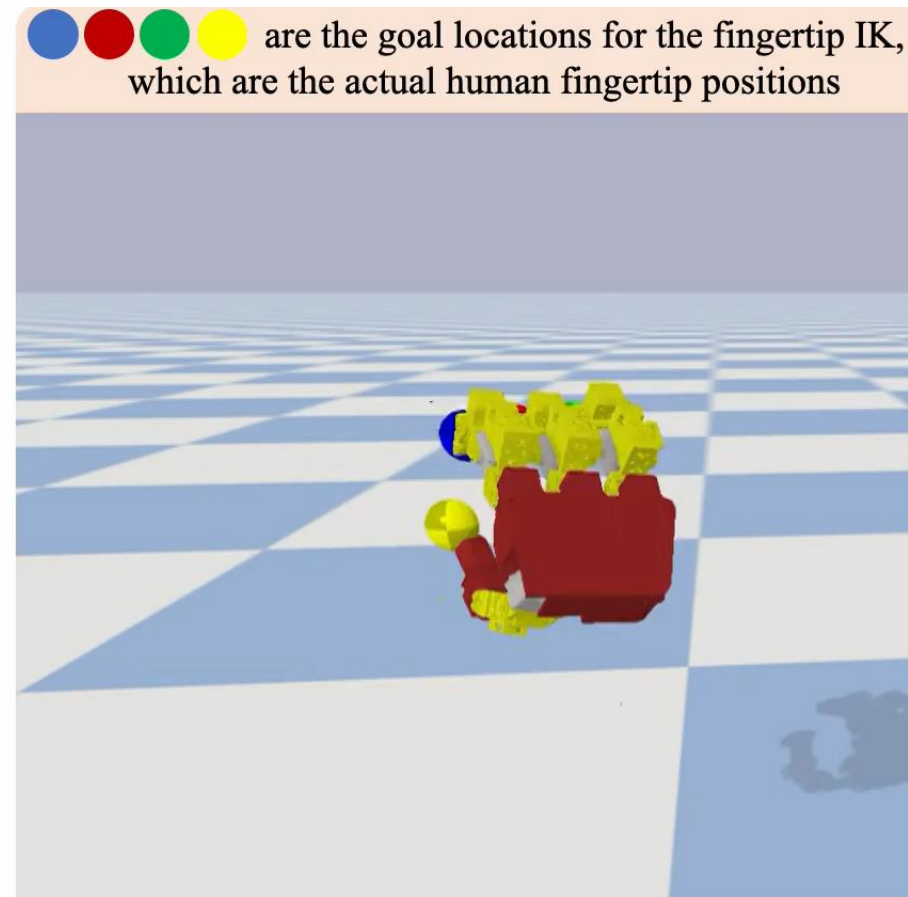
$$D_2 = \{(s_0, a_0, s_1, a_1, \dots, s_n)\} \quad D_3 = \{(s_0, \hat{a}_0, s_1, \hat{a}_1, \dots, s_n)\}$$

$$\hat{a}_t = \mathbf{h}(a_t)$$

# Action retargeting

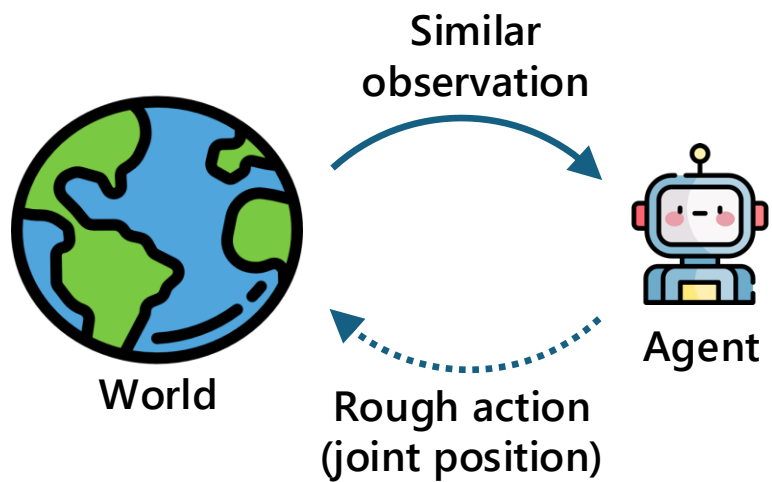
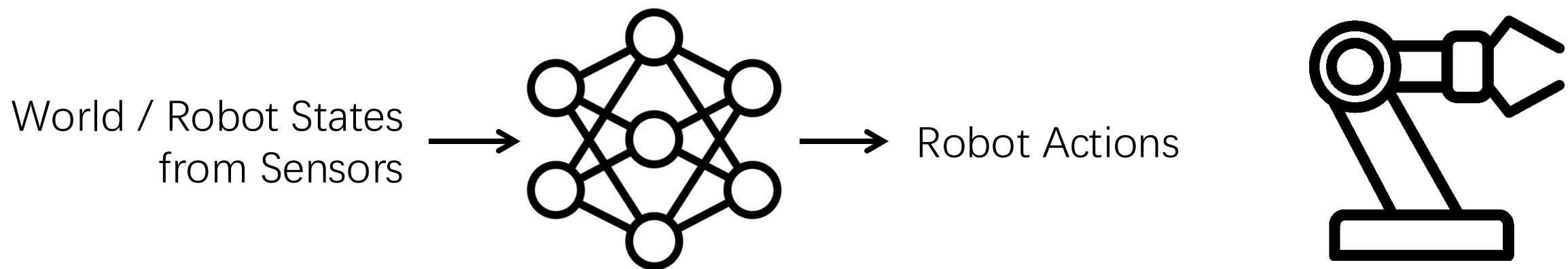


Human hand



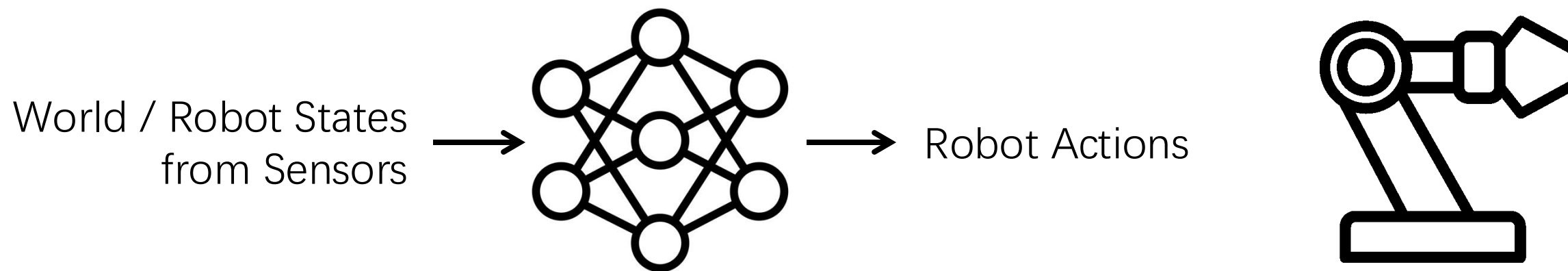
LEAP robot hand (1.5x larger in size)

# Policy learning

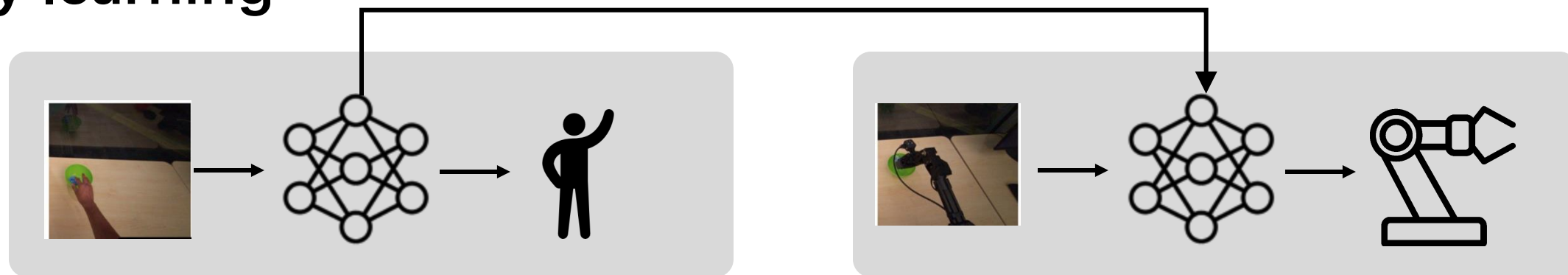


By action retargeting, we can get rough robot action (joint position)

# Policy learning

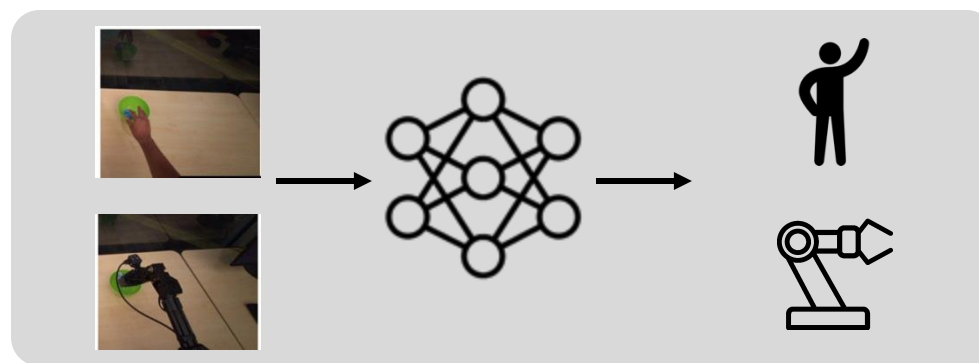


# Policy learning

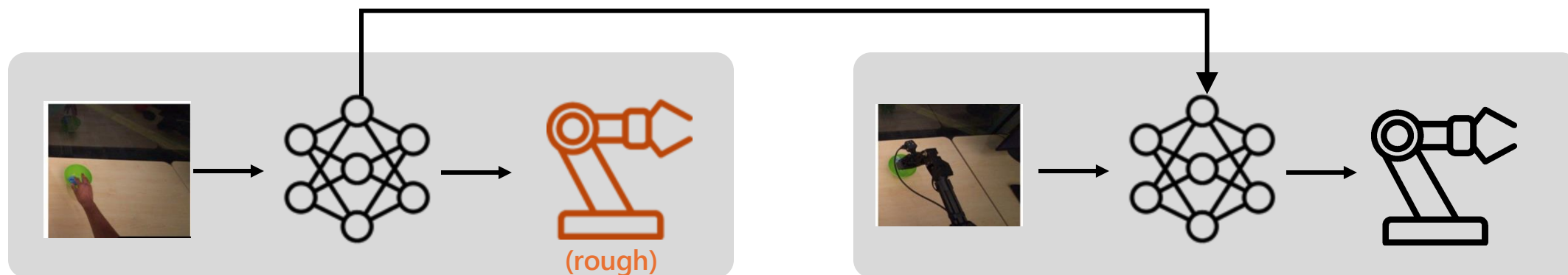


Pretrain

Finetune



Mix training



Human-in-the-loop correction



# Learning from videos

[A] Policy learning from human videos



[B] World modeling from all videos





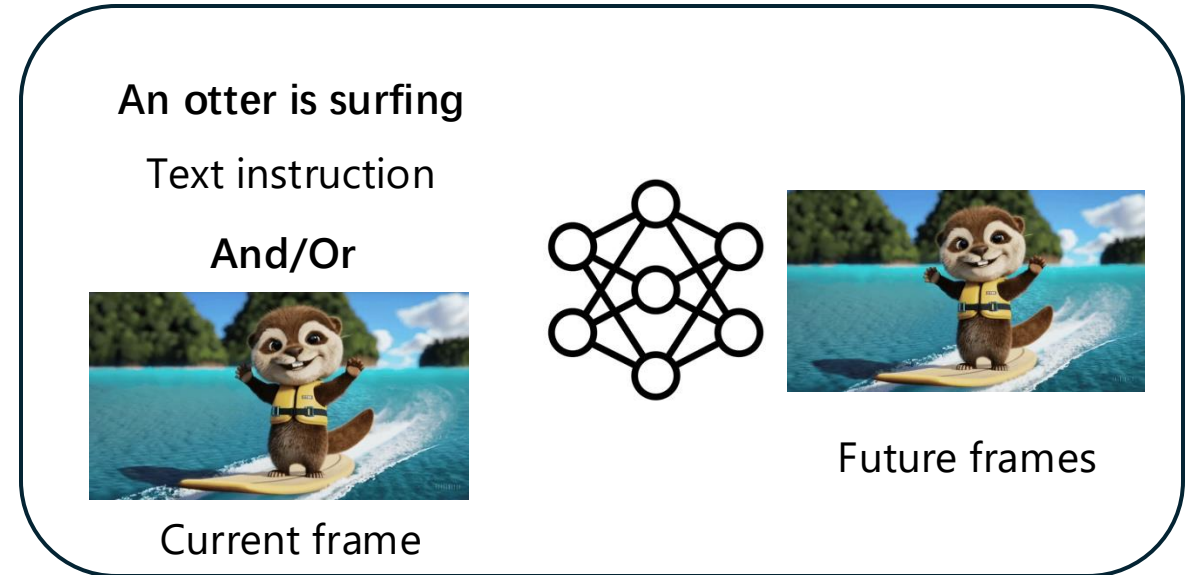
# GenAI



# Data and learning



Data: all internet videos/games



Learning: future video generation

# Cosmos



Fold a green fabric item on a table.



Organize books by placing them vertically on a shelf.



Pick up an electronic device from a table and place it in a bin.

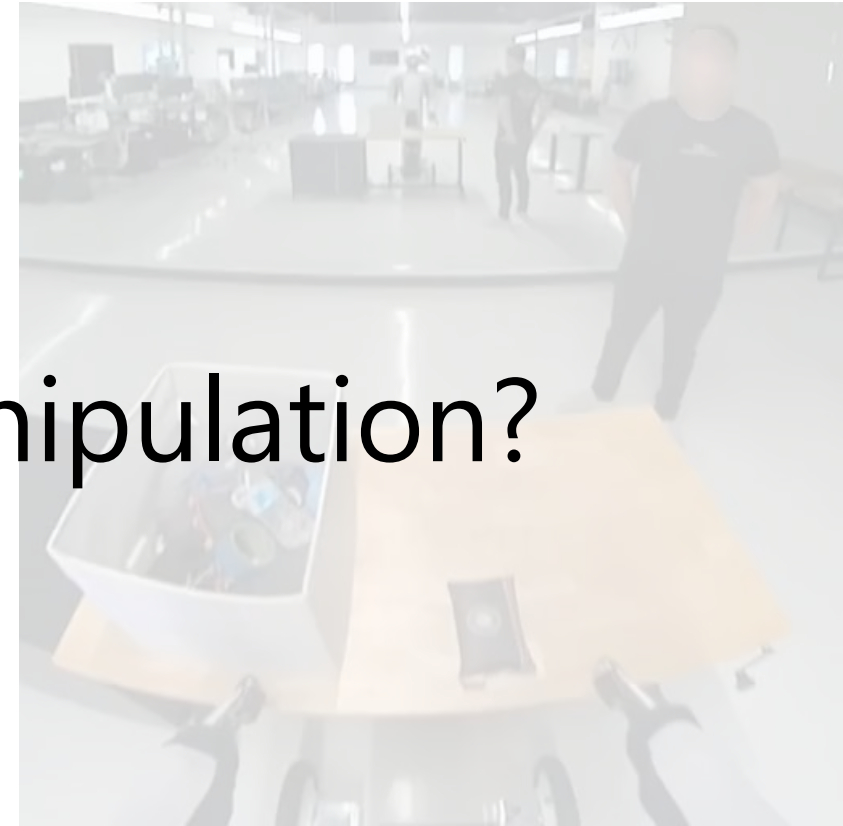
# Cosmos



Fold a green fabric item on a table.



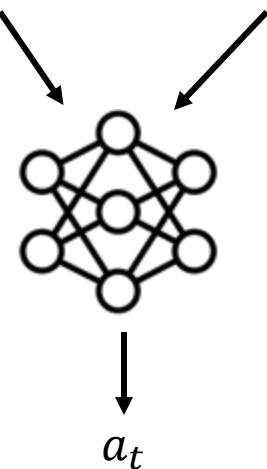
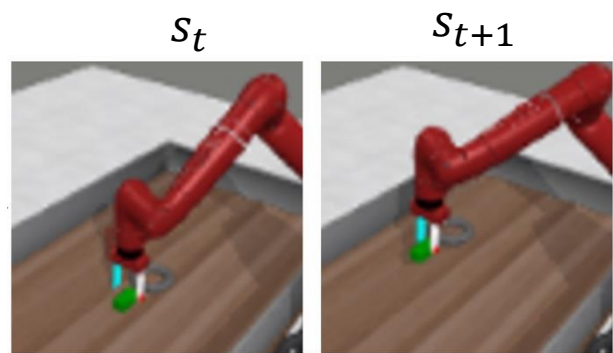
Organize books by placing them vertically on a shelf.



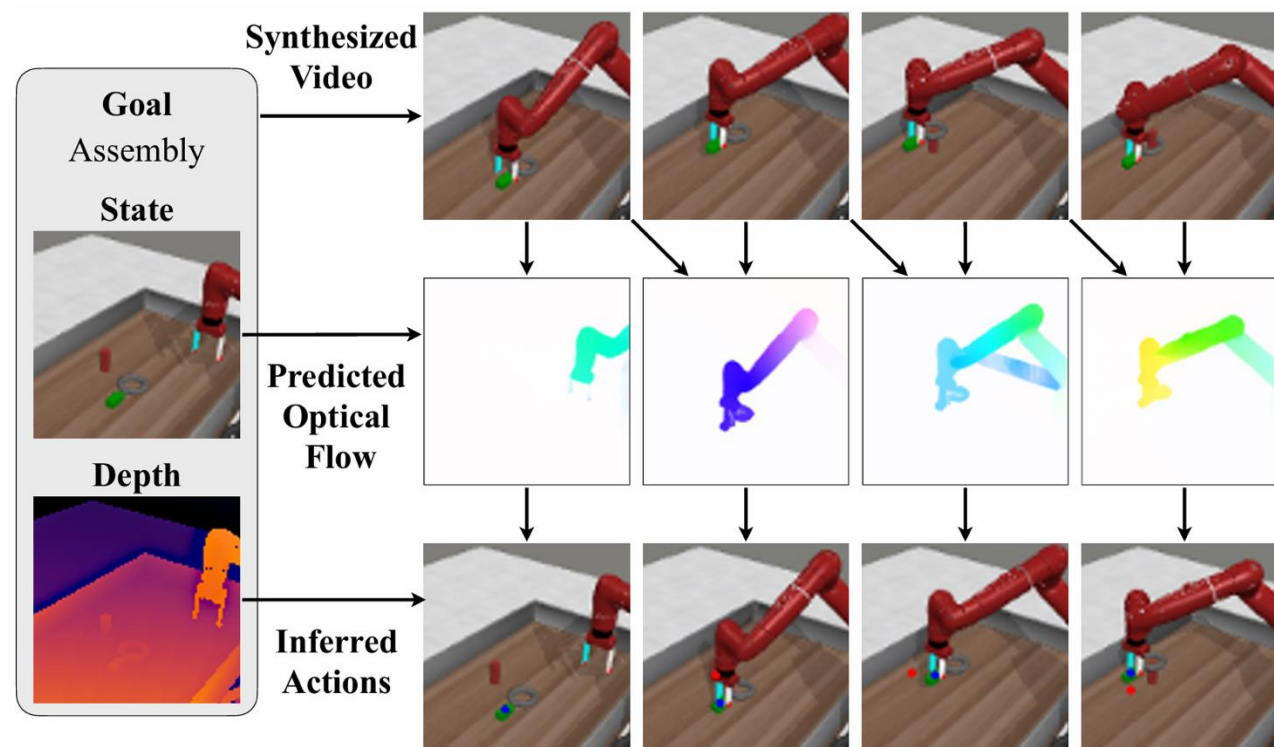
Pick up an electronic device from a table and place it in a bin.

**Problem of applying to manipulation?  
No action!**

# Applying to manipulation



Inverse dynamic model



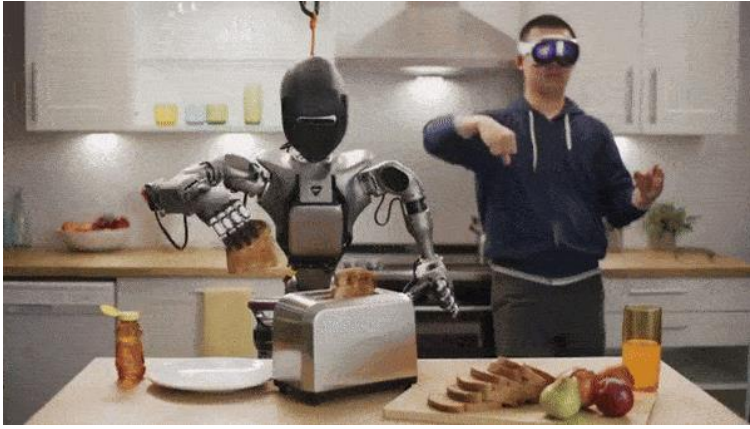
Correspondence

# Today

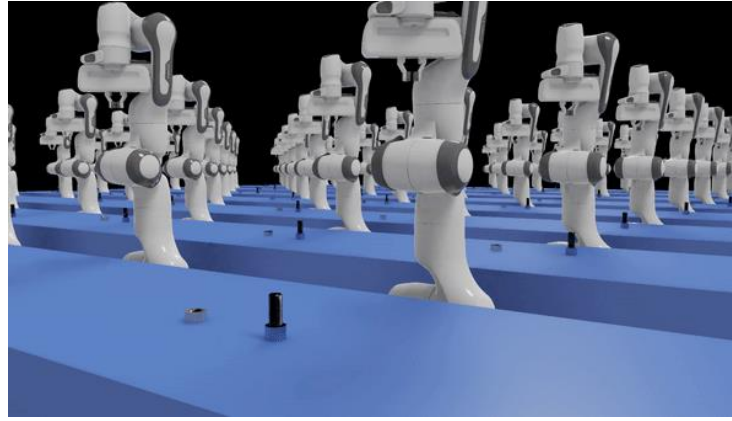
- Learning from videos
  - Policy learning from human video
    - Observation mismatch
    - Get action
  - World modeling
    - Data and learning
    - How to apply to robots
- **Drawbacks of three different methods**
- Interaction interface Case Studies: In-Depth Analysis
  - Policy learning with interaction interface



# How to collect robotic data?



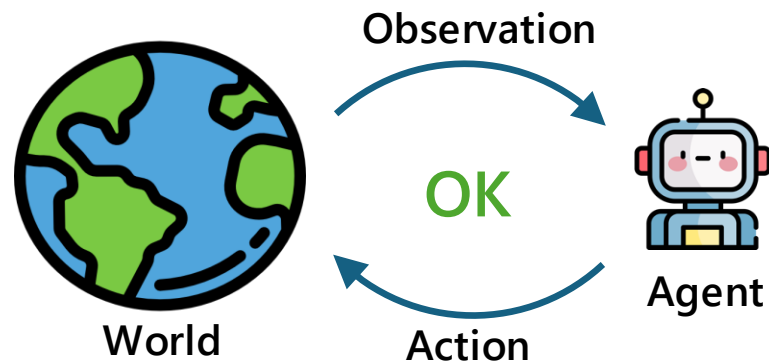
Teleoperation



Simulation



Video



# Is teleoperation good?

\$\$\$\$\$

Expensive



Inefficient



Limited capability

The screenshot shows a competition interface for XPRIZE AVATAR. It features three main images: a top-left photo of two people operating a robot, a bottom-left photo of a robot arm, and a large central photo of two yellow DeWalt power drills on a table. At the bottom, a scoreboard displays the following data:

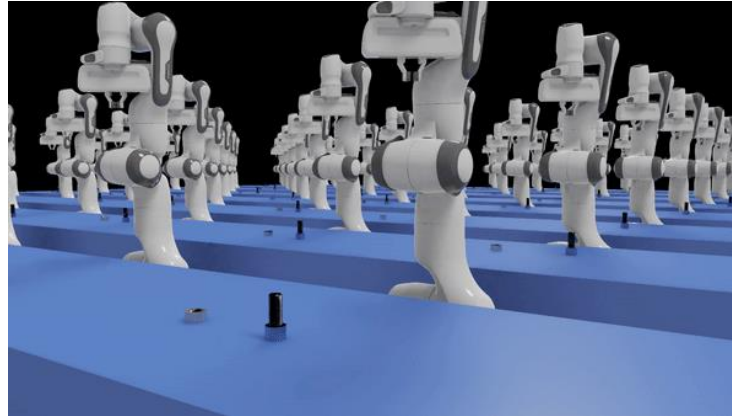
RANK	TEAM	TIME ON COURSE	TASK	POINTS JUDGED	TOTAL	LEADER STATS TO BEAT
6	TEAM NORTHEASTERN	07:33	8	2	10	08:15 15

Top teleoperation platform motivated by \$10M prize:  
Took 40 seconds to adjust the drill and failed

# How to collect robotic data?



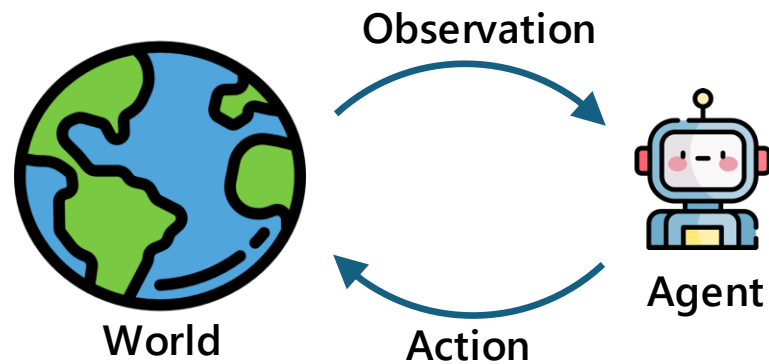
Teleoperation



Simulation



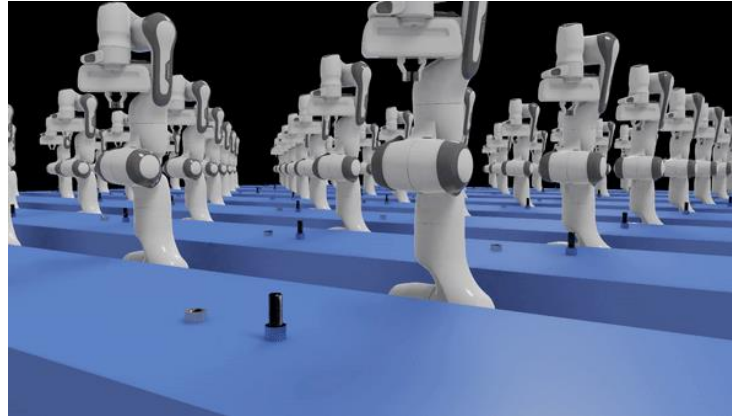
Video



# How to collect robotic data?



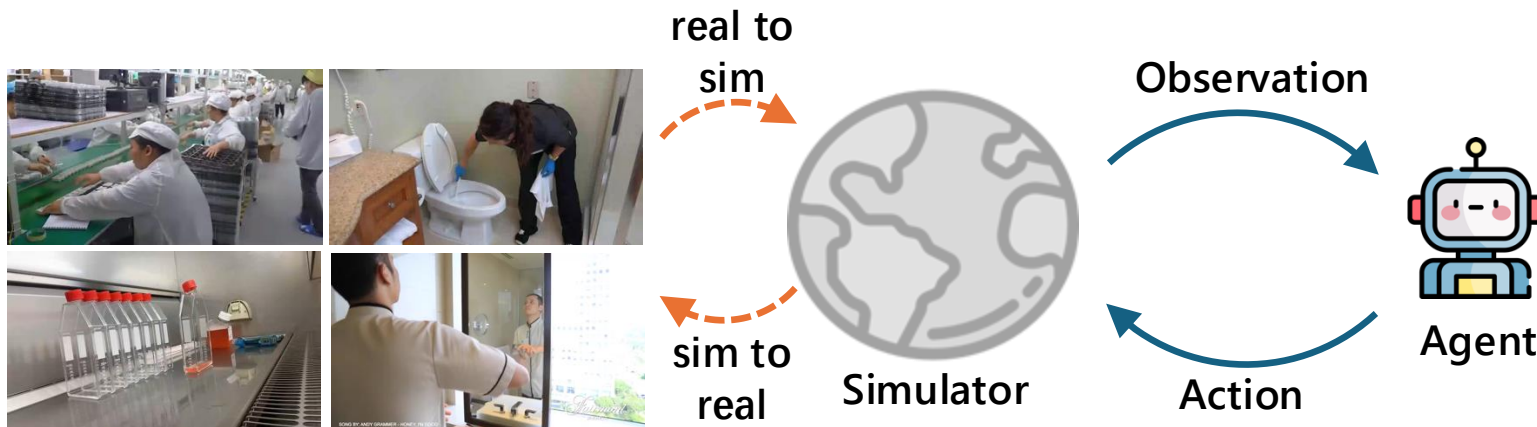
Teleoperation



Simulation



Video



# How to collect robotic data?



Teleoperation



Simulation



Video

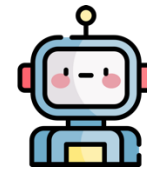


Heavily occluded  
Tactile is needed



World

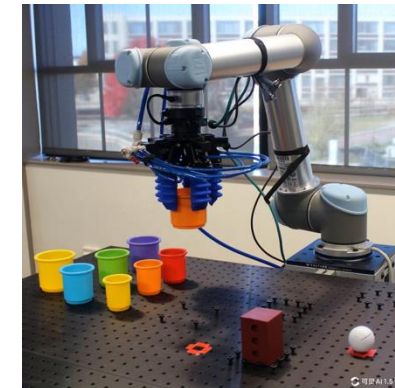
Only visual observation



Agent



No accurate action data

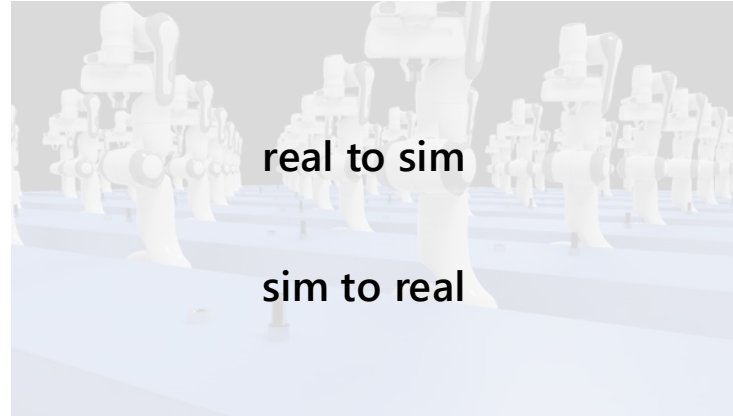


The robotic arm pushes the red block - KlingAI

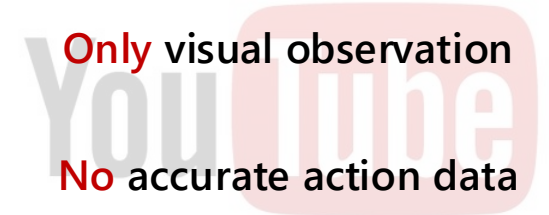
# How to collect robotic data?



Teleoperation



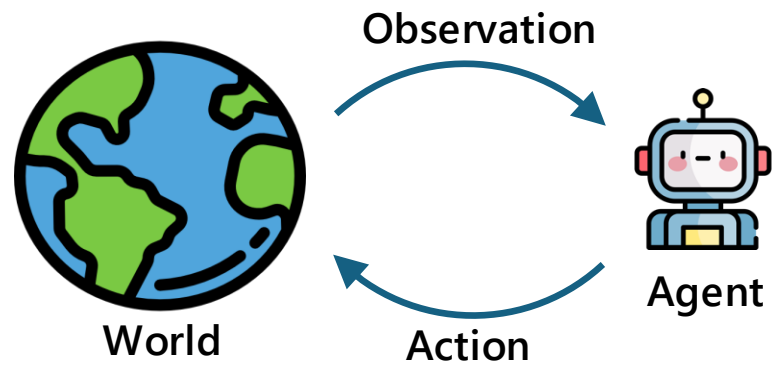
Simulation



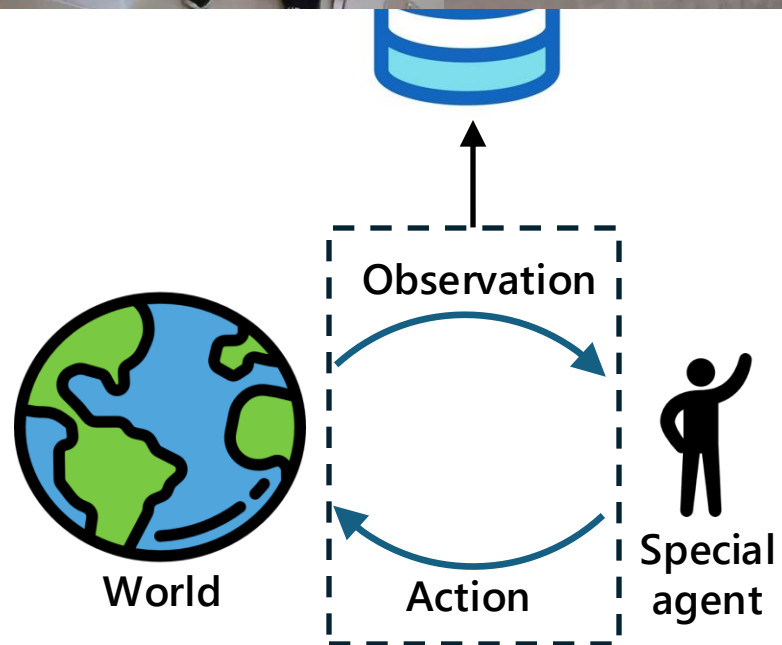
Video

**A new paradigm is needed for data collection!**

# How to collect robotic data?

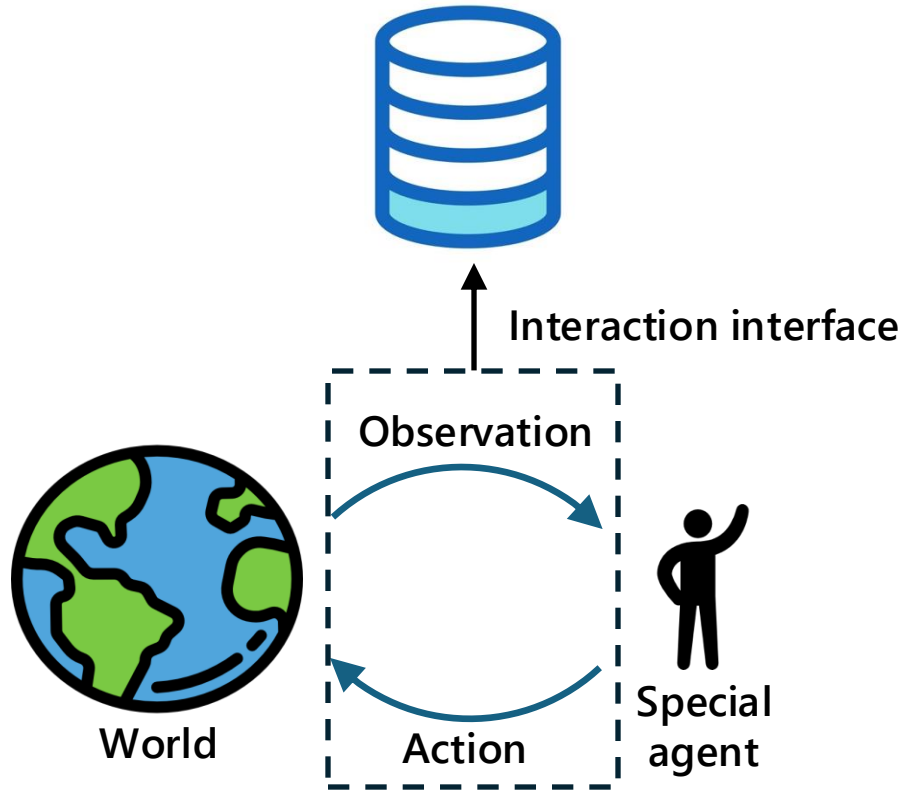


# How to collect robotic data?





# Interaction interface



Devices that can record **observation** and **action** data when humans are interacting with the world

# Today

- Learning from videos
  - Policy learning from human video
    - Observation mismatch
    - Get action
  - World modeling
    - Data and learning
    - How to apply to robots
- Drawbacks of three different methods
- **Interaction interface Case Studies: In-Depth Analysis**
  - Policy learning with interaction interface

# Interaction interface



**AirExo**



**UMI**



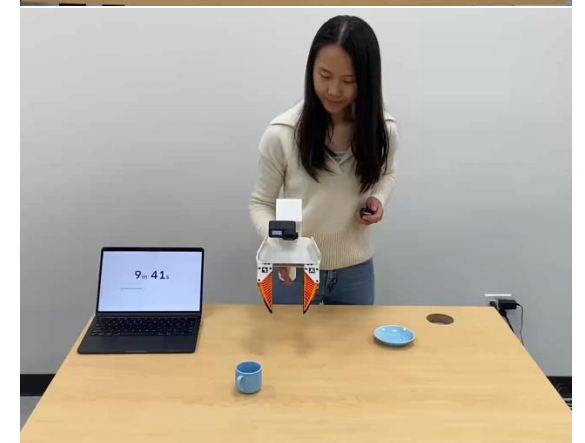
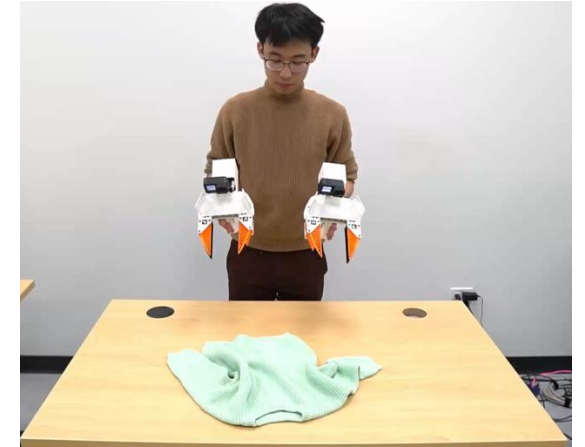
**DEXO**

AirExo: Low-Cost Exoskeletons for Learning Whole-Arm Manipulation in the Wild

Universal Manipulation Interface: In-The-Wild Robot Teaching Without In-The-Wild Robots

DEXO: Hand Exoskeleton System for Teaching Robot Dexterous Manipulation In-The-Wild

# Interaction interface



**AirExo**

**UMI**

AirExo: Low-Cost Exoskeletons for Learning Whole-Arm Manipulation in the Wild

Universal Manipulation Interface: In-The-Wild Robot Teaching Without In-The-Wild Robots

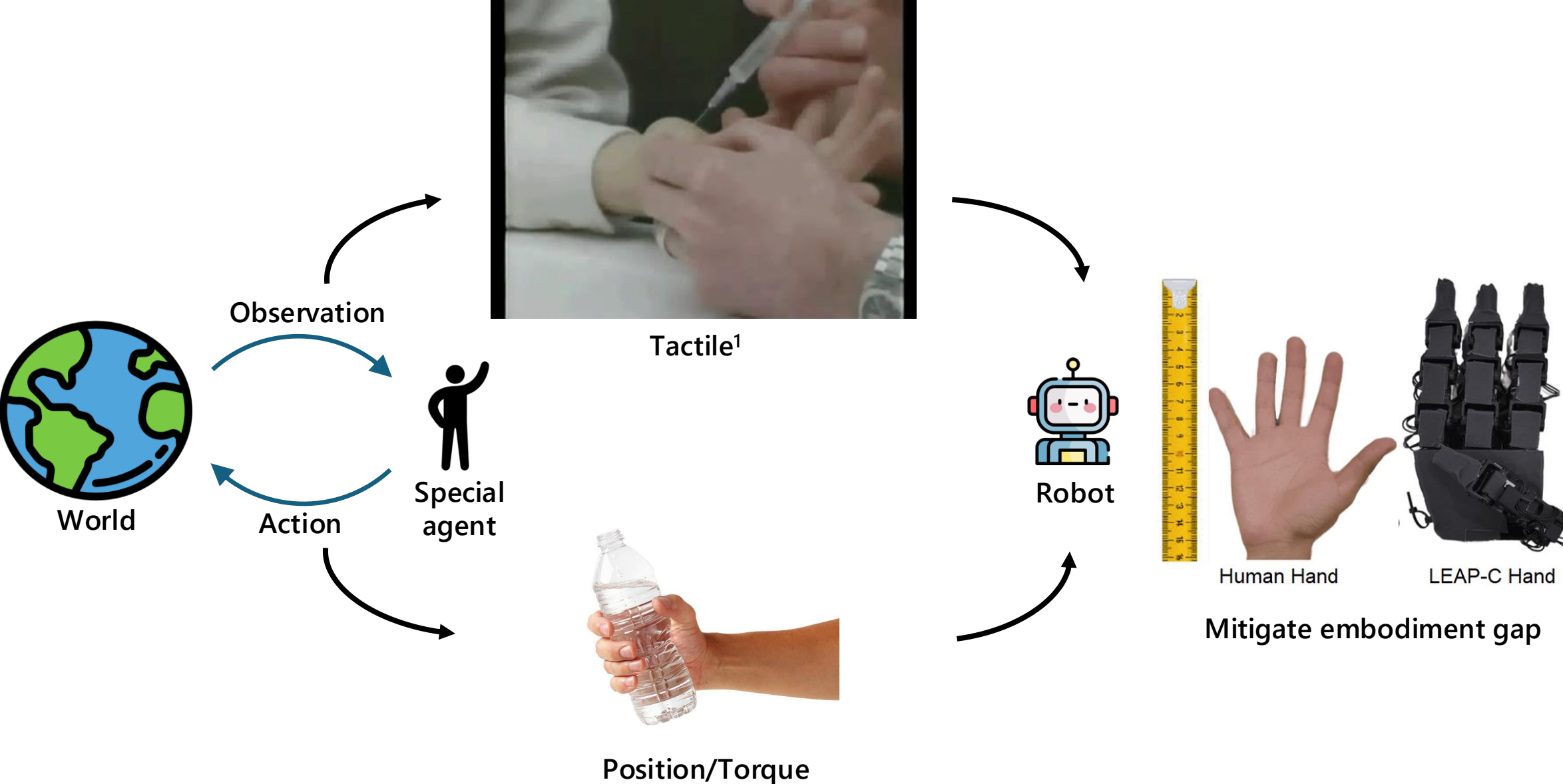
# Dexterous interaction interface

**What** is needed?  
**How** to achieve?  
**Why** is it better?



DEXO

# What is needed?



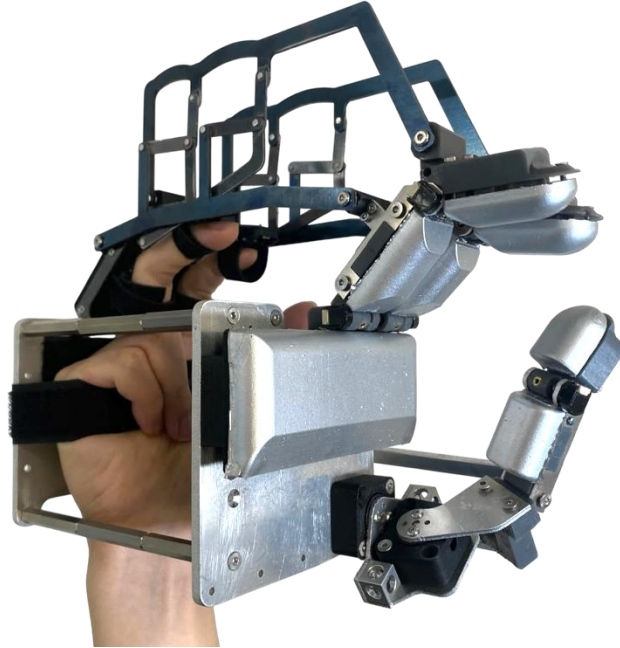
1: In courtesy of Johansson et al.

# How to achieve?



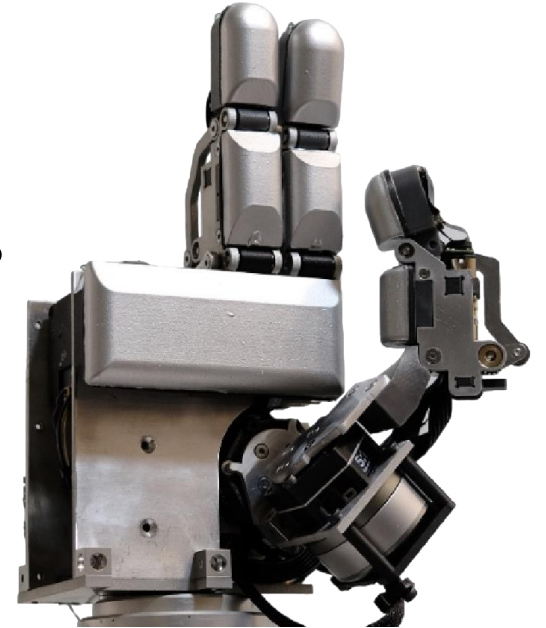
Human hand

Constraints



Interaction  
interface

Constraints



Robotic  
hardware



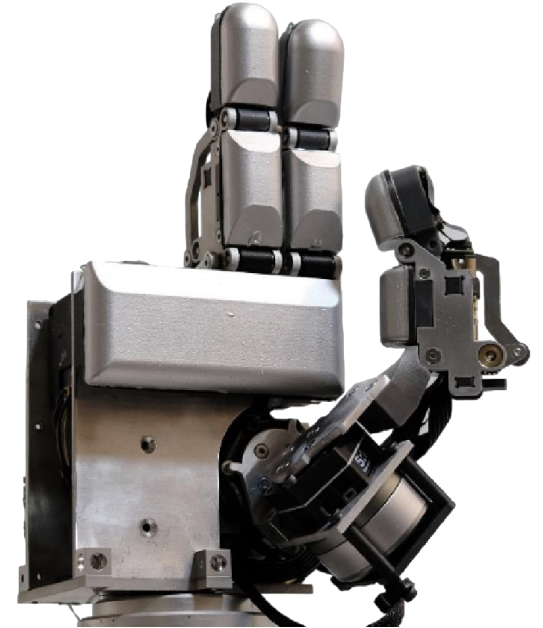
Co-design

# How to achieve?



Human hand

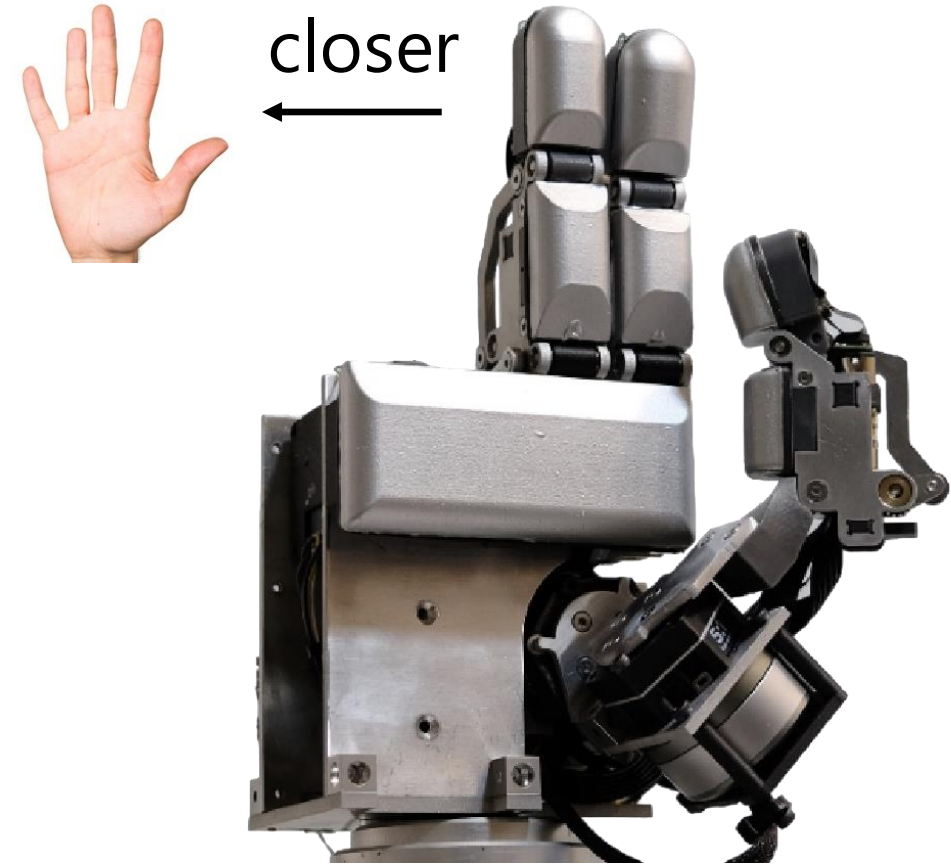
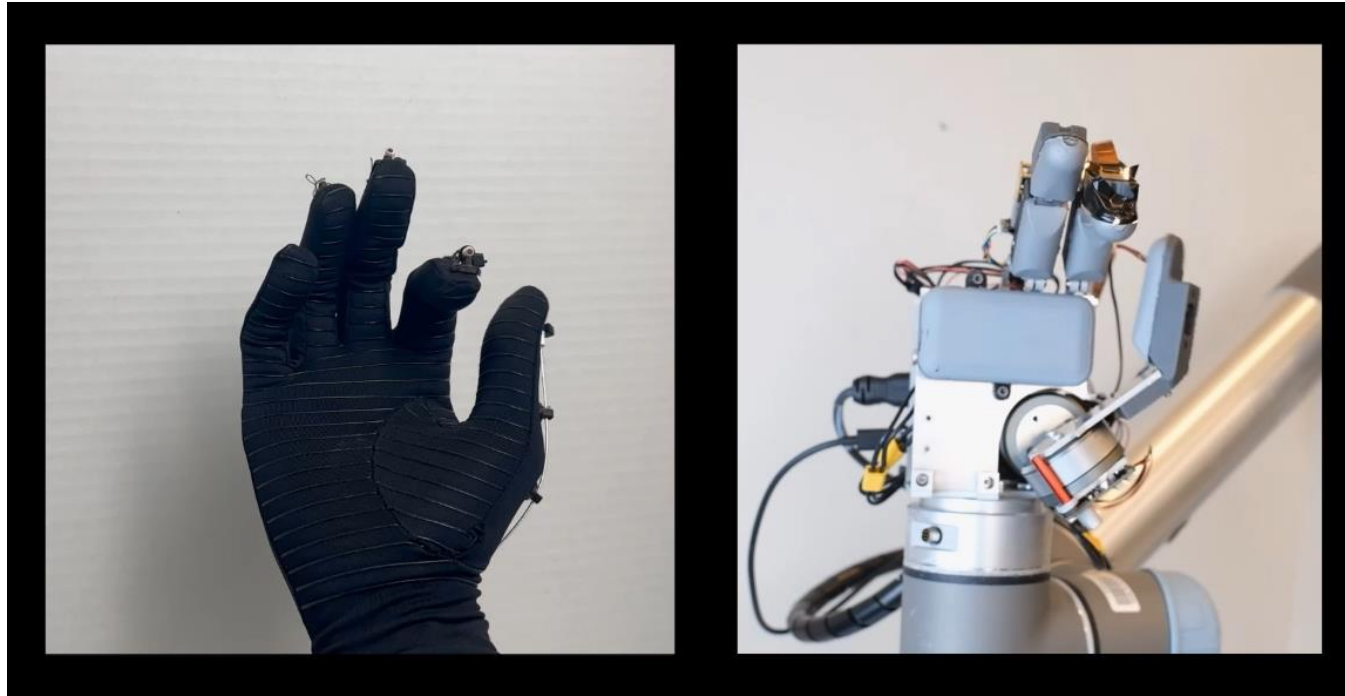
Closer to human hand



Robotic hardware

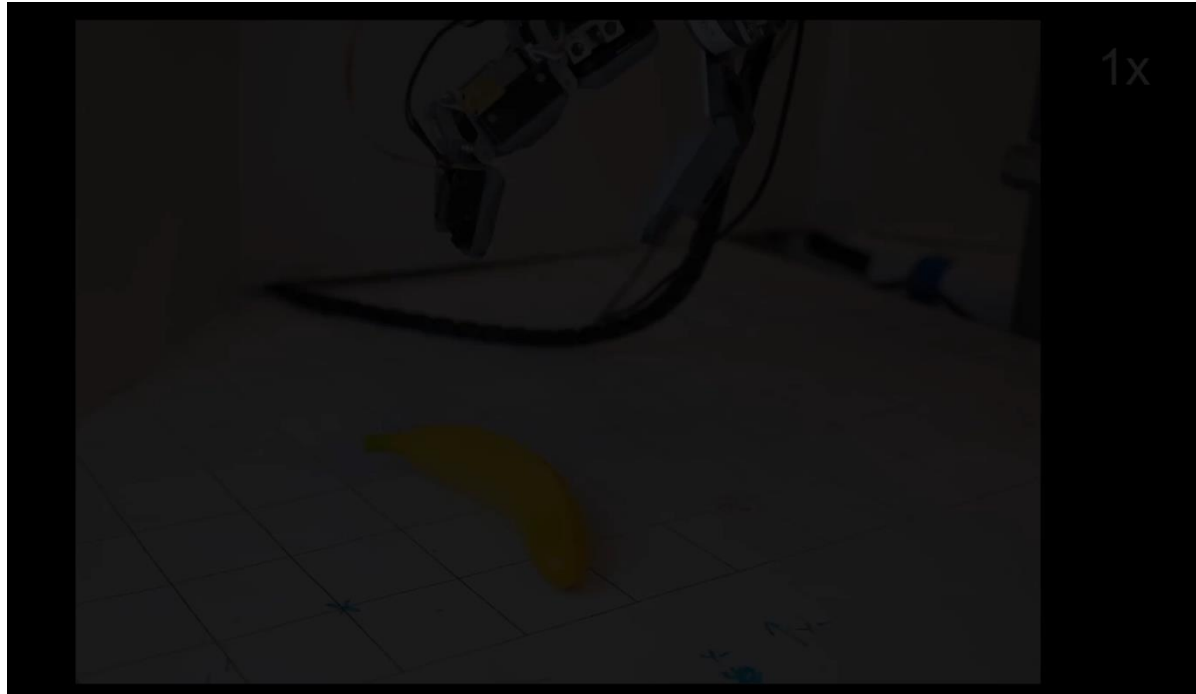


# How to achieve?

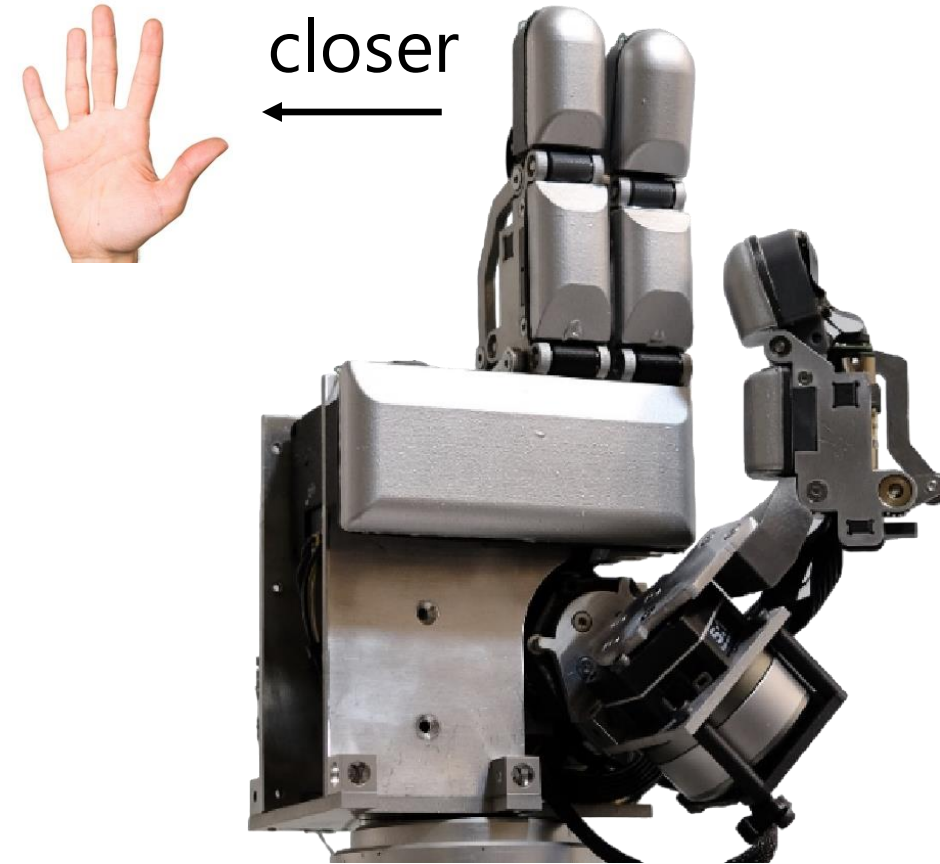


Make **tactile sensing** closer to human

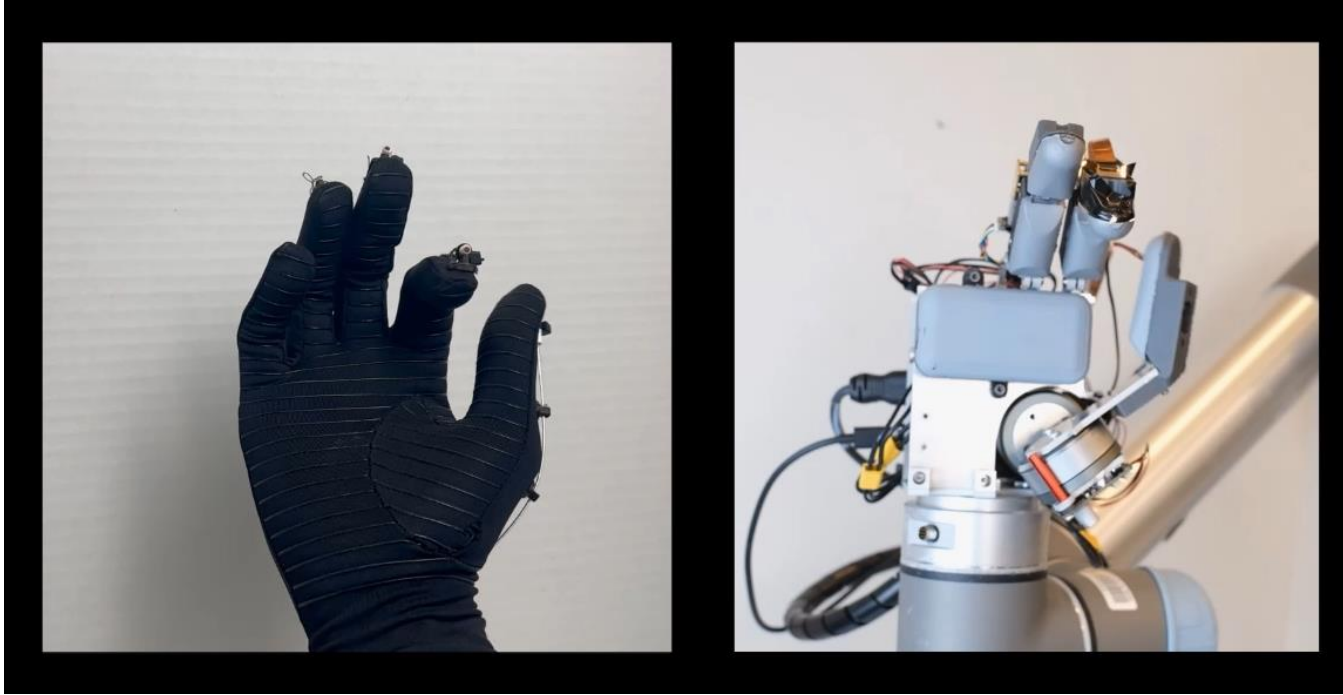
# How to achieve?



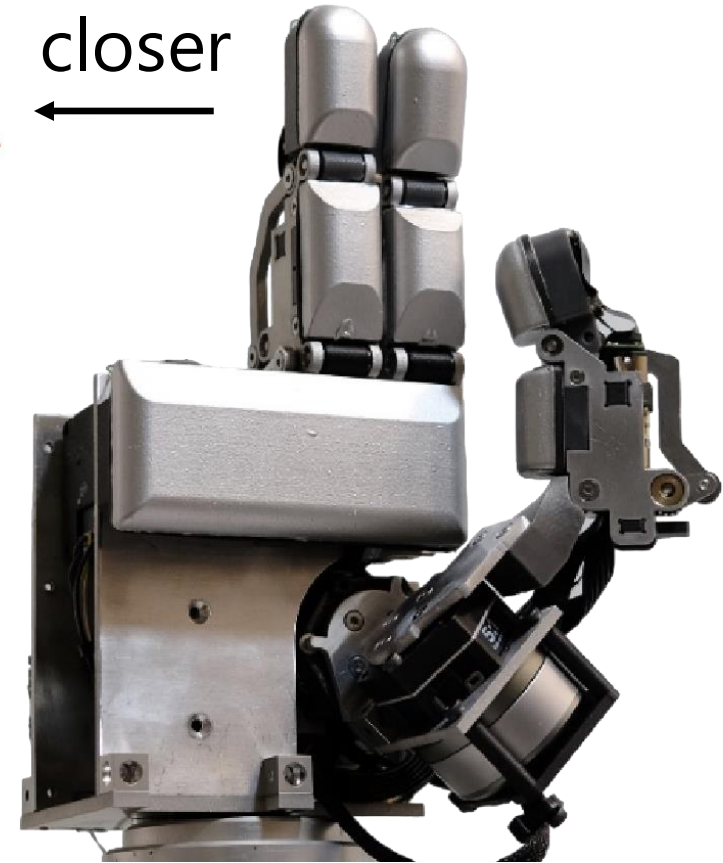
Make **dynamic** closer to human



# How to achieve?



← closer



Make **form factor** closer to human

# How to achieve?



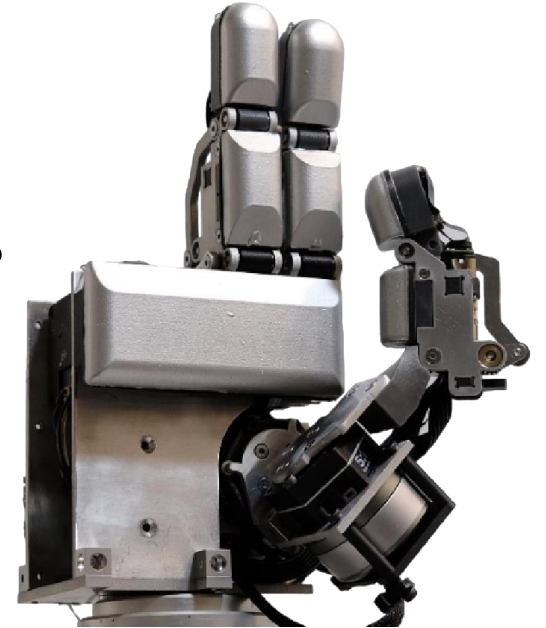
Human hand

Constraints



Interaction  
interface

Constraints



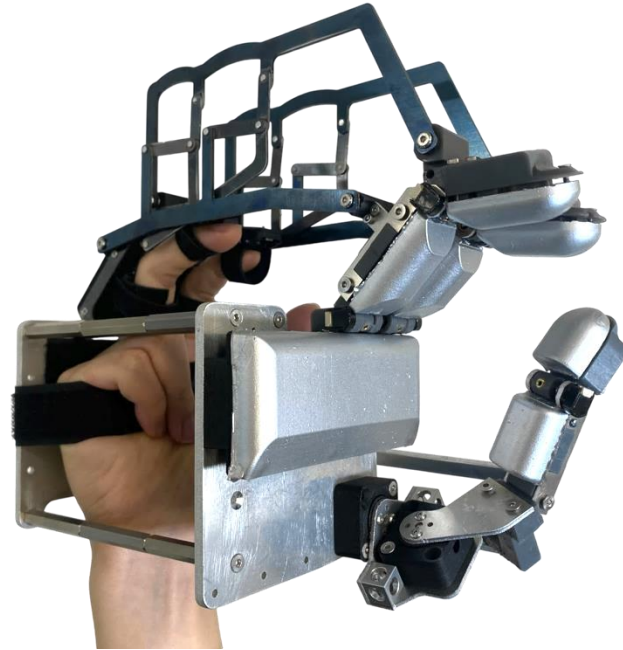
Robotic  
hardware

# How to achieve?



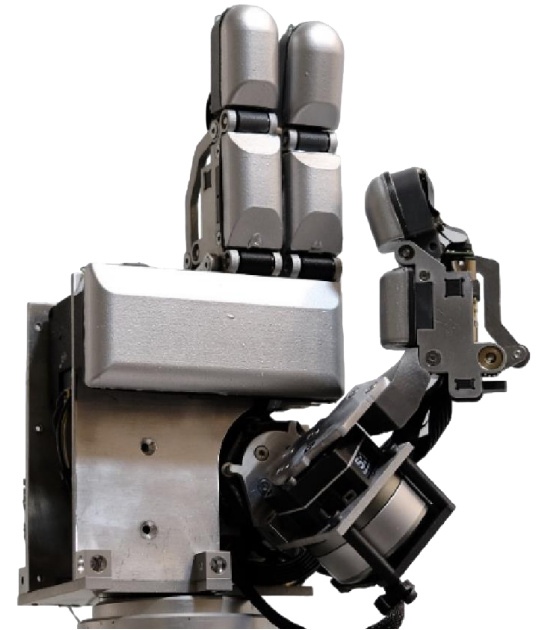
Human hand

Bridge



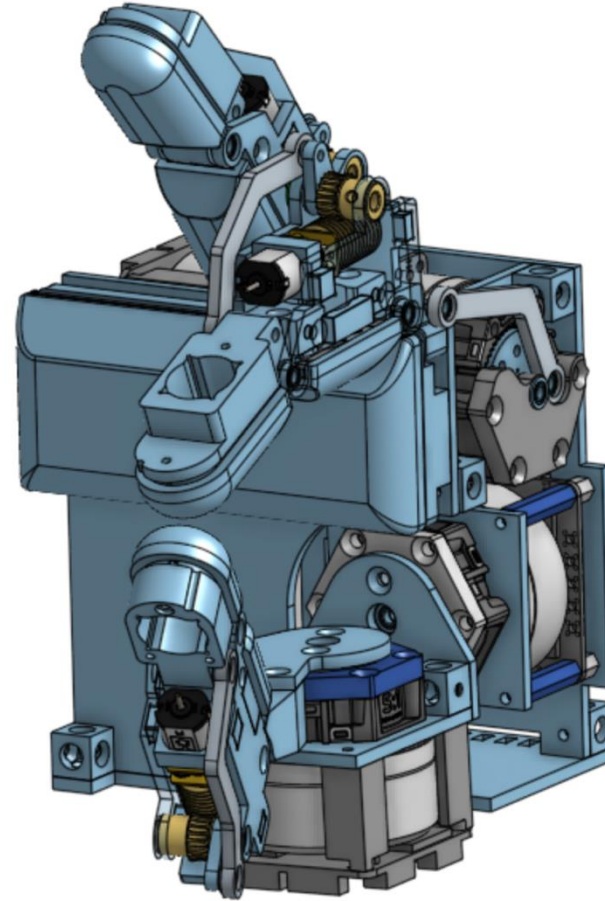
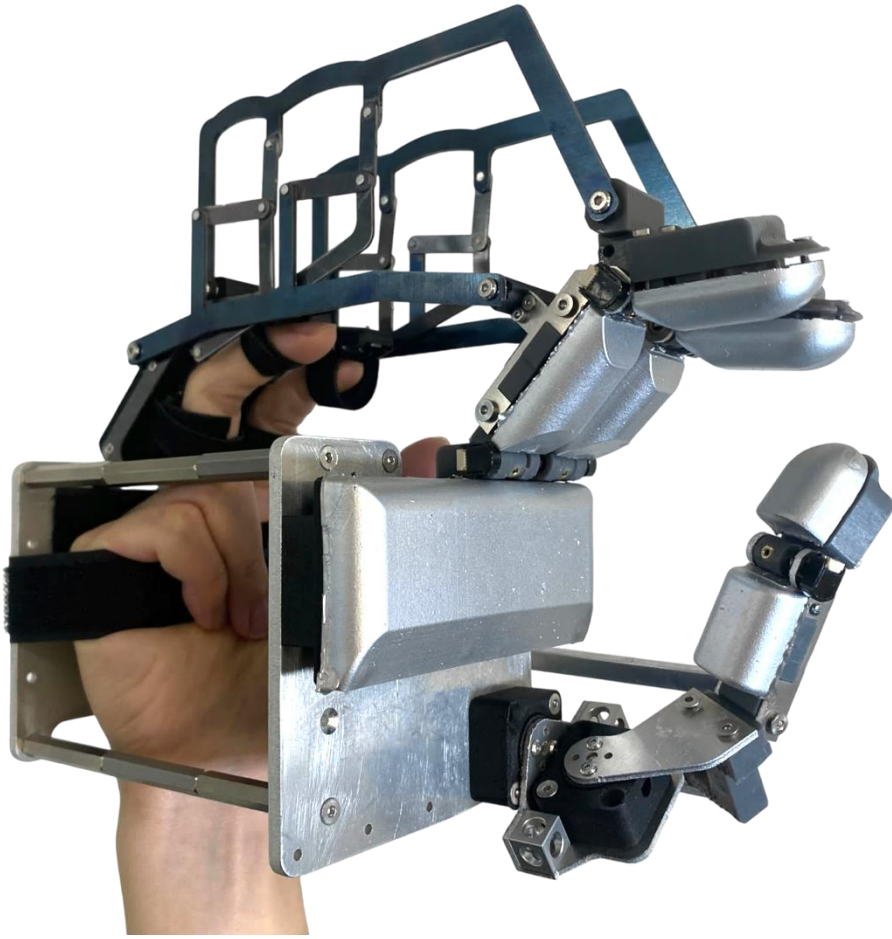
Interaction  
interface

Bridge

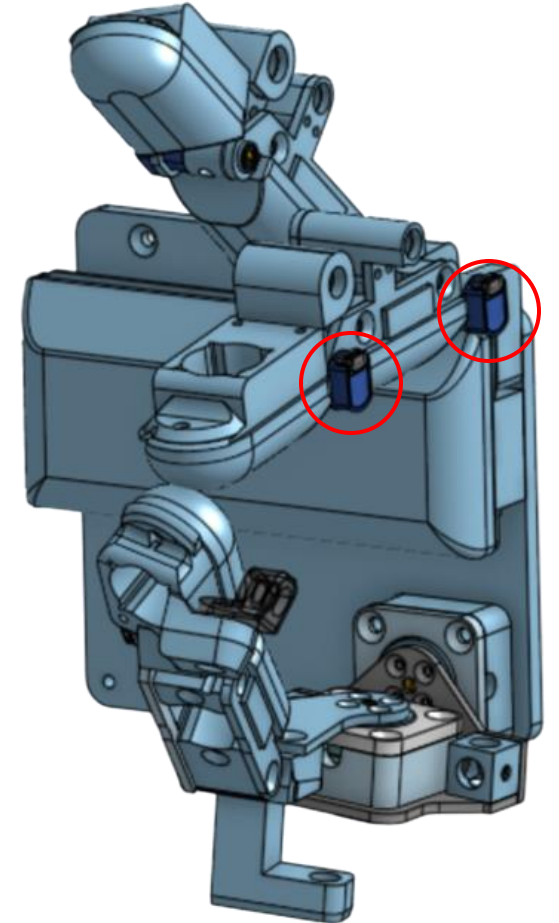


Robotic  
hardware

# How to achieve?

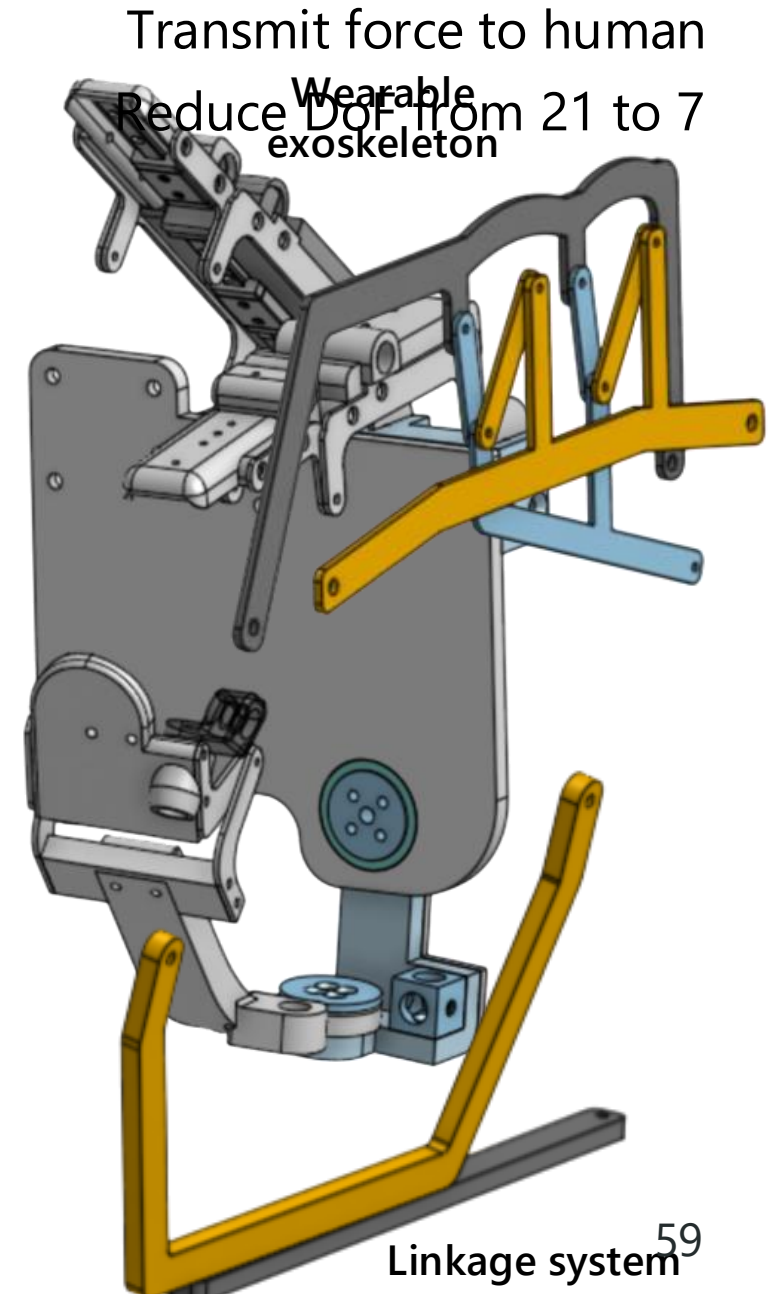
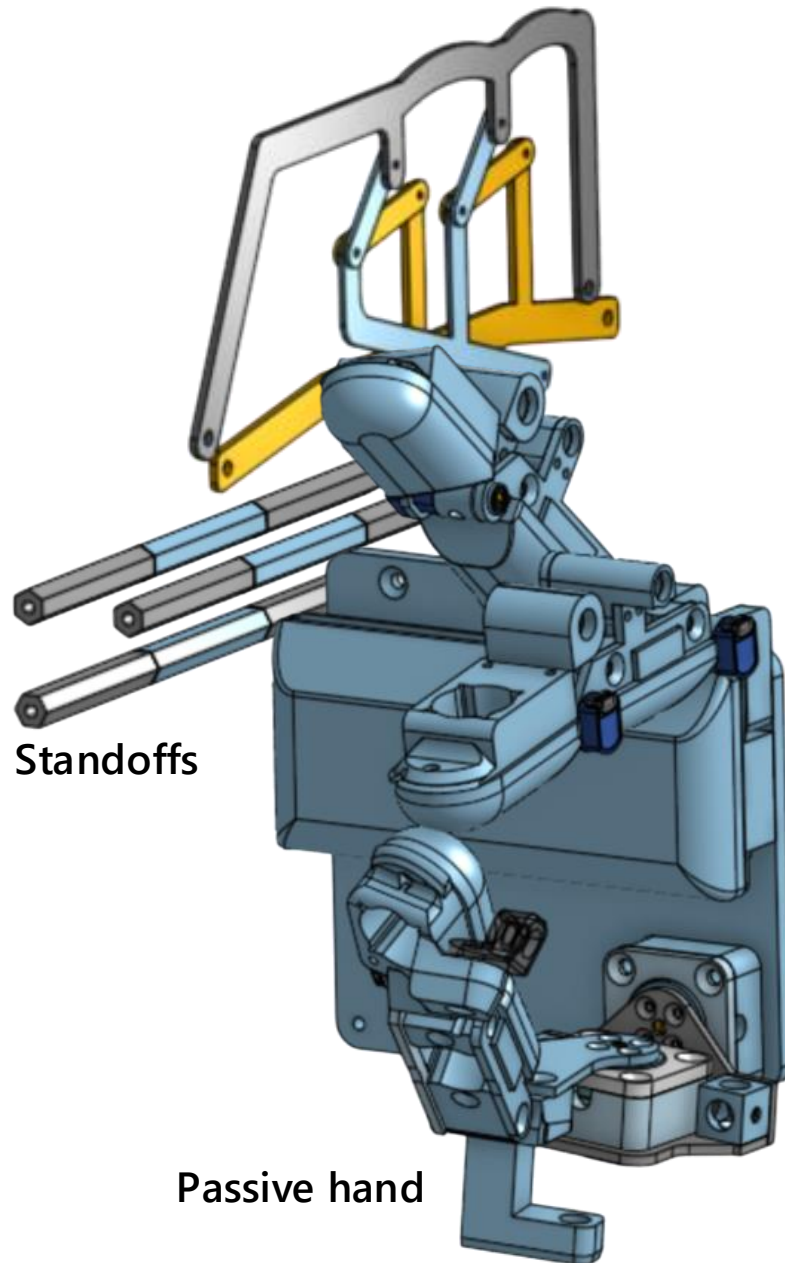
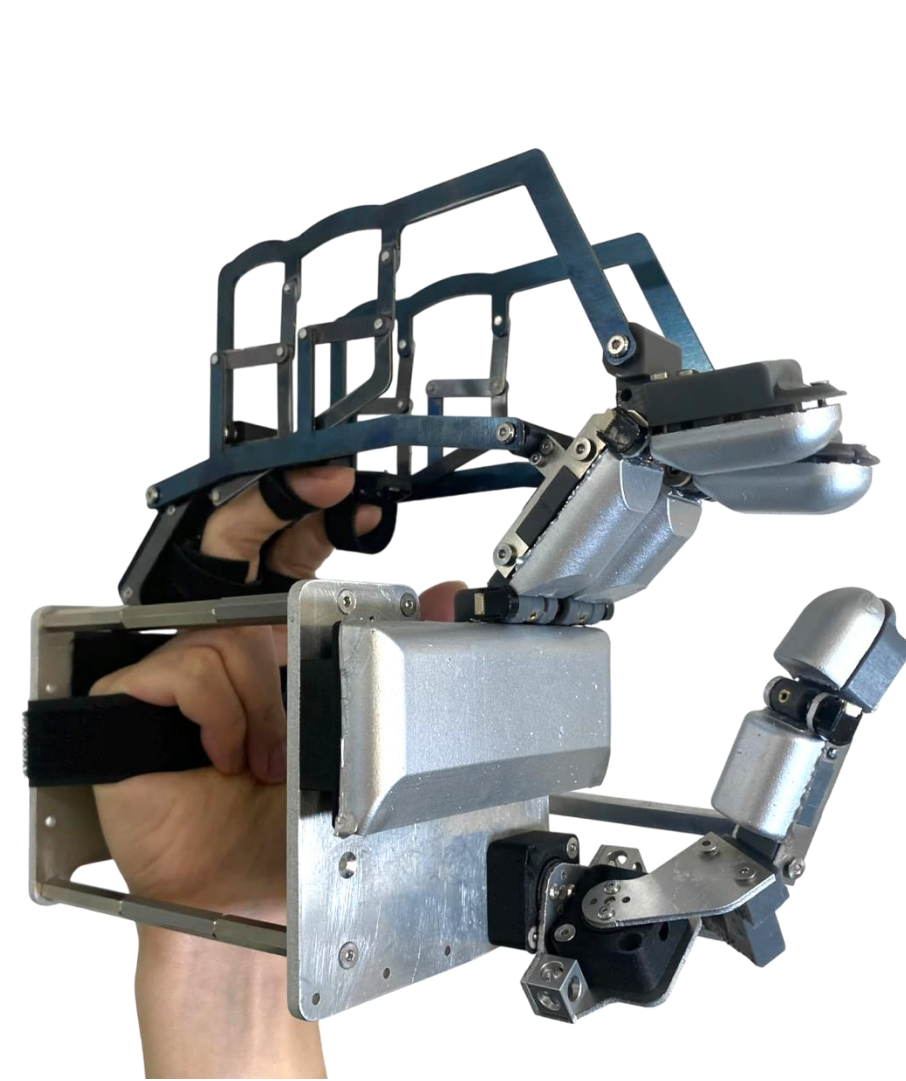


Actual hand



Passive hand

# How to achieve?



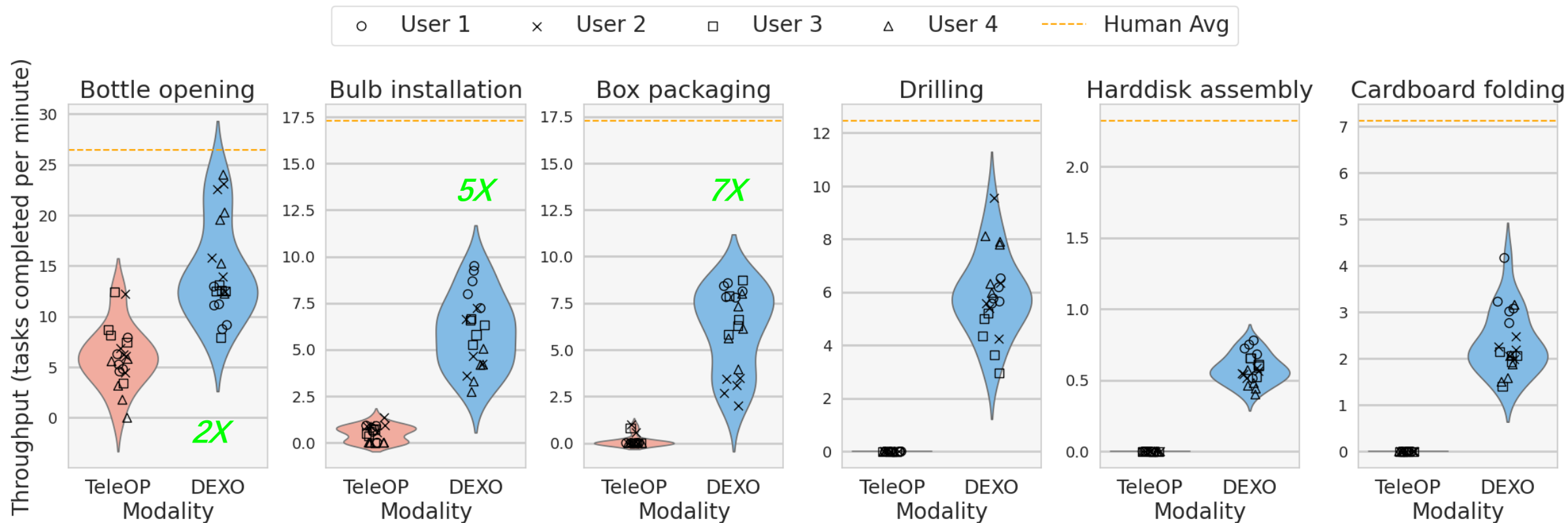
# Why is it better?



- Observation
  - Tactile: same
  - Wrist cam: same
  - Head cam: transferable
- Action
  - Angle: same
  - Torque: computable

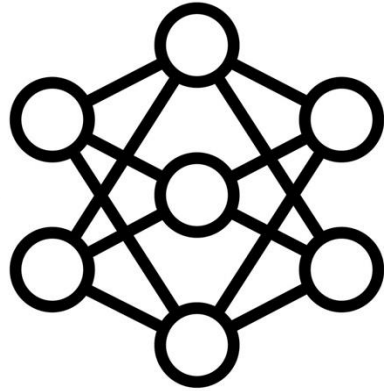


# Why is it better?

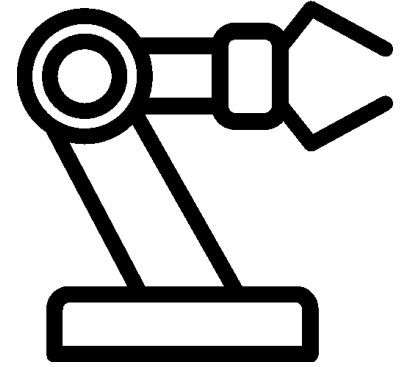


# Policy learning

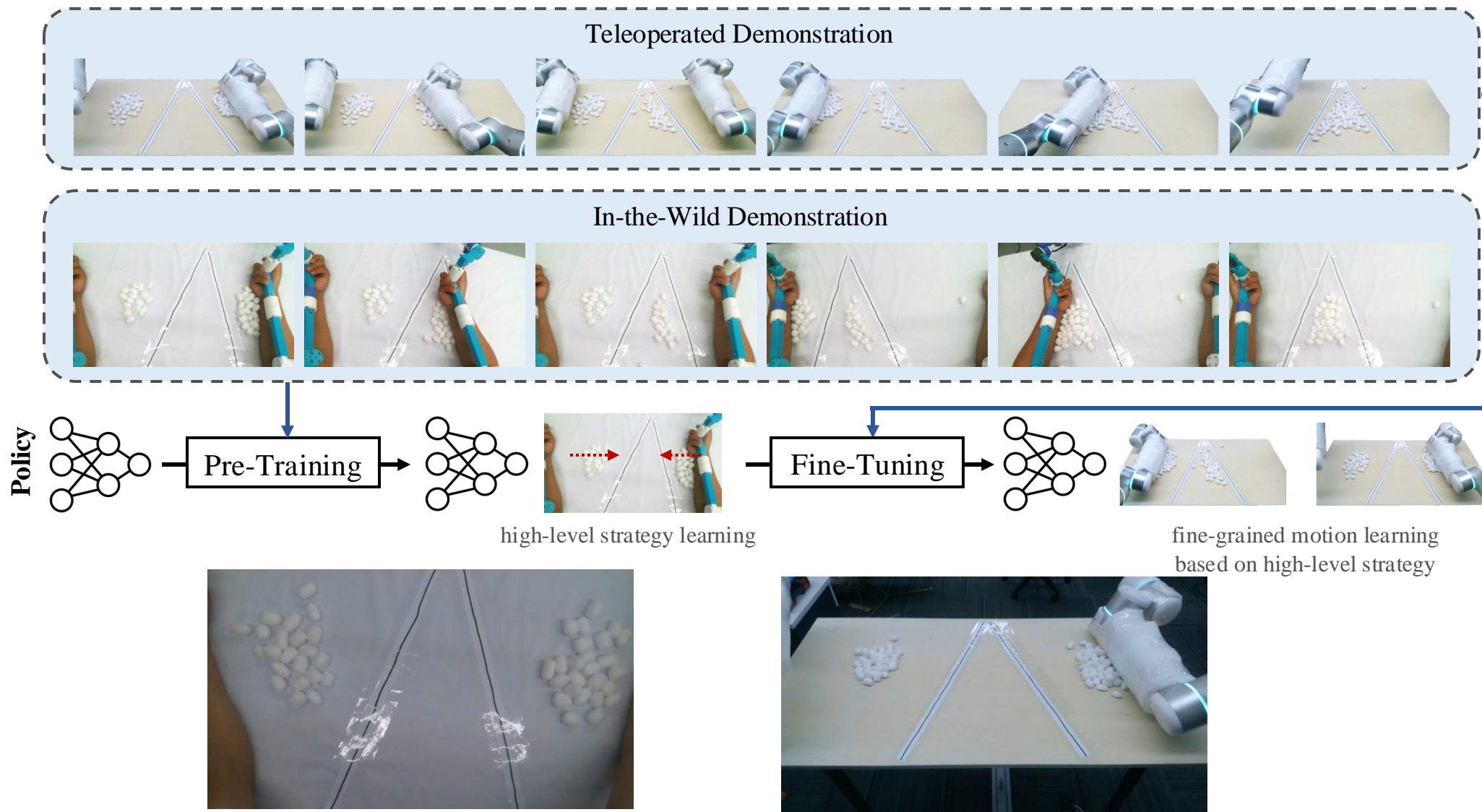
World / Robot States  
from Sensors



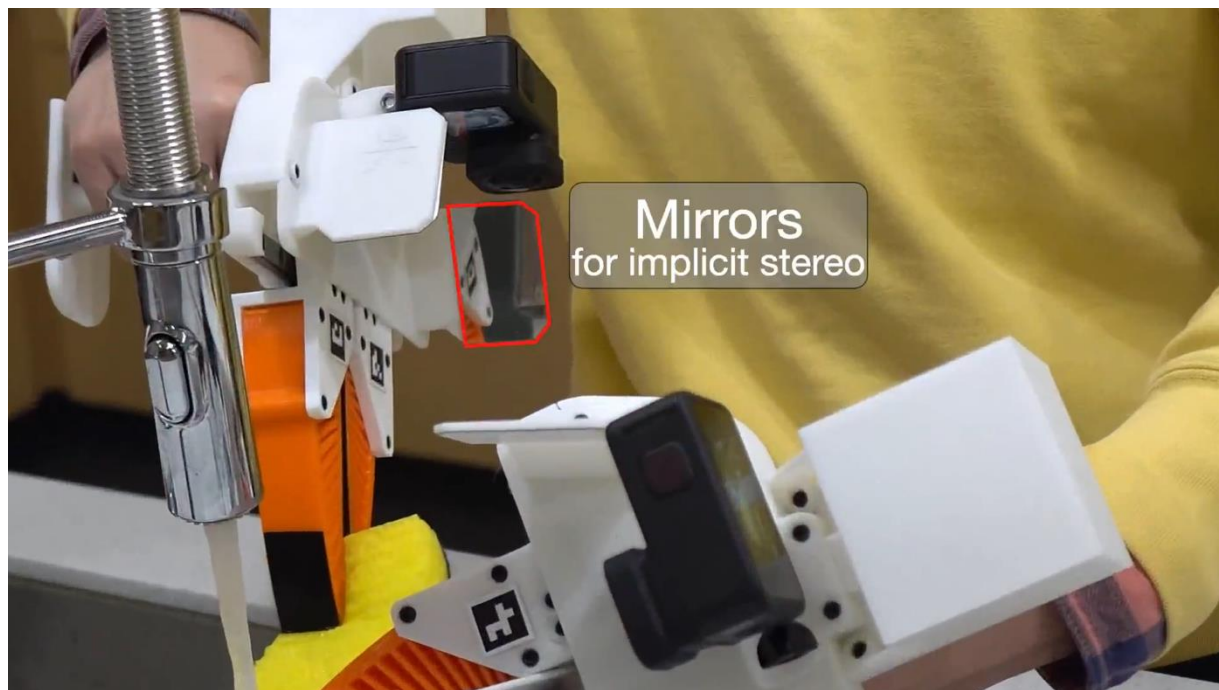
Robot Actions



# Policy learning



# Policy learning

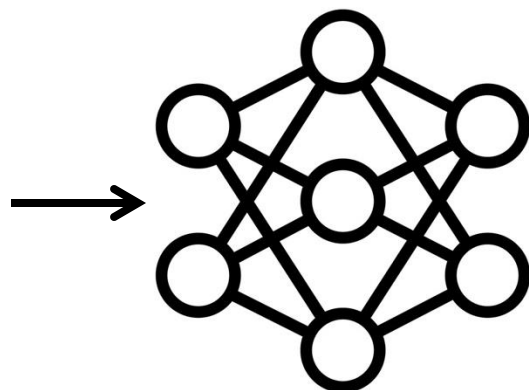


Same observation space

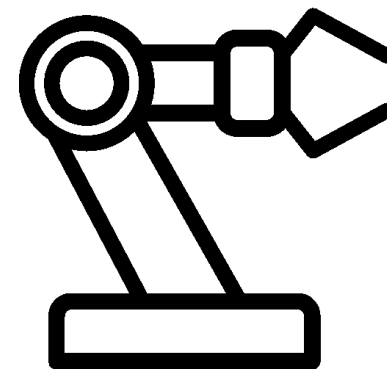


Recover action through SLAM

World / Robot States  
from Sensors



Robot Actions



# Today

- Learning from videos
  - Policy learning from human video
    - Observation mismatch
    - Get action
  - World modeling
    - Data and learning
    - How to apply to robots
- Drawbacks of three different methods
- Interaction interface Case Studies: In-Depth Analysis
  - Policy learning with interaction interface

# Next Class: Hands-on Tutorial

## Sneak Peak

Iterate to get best performance

Teleoperate Robots  
in Virtual Reality

Dataset  
Creation / Curation

Policy Training

Policy  
Evaluation

