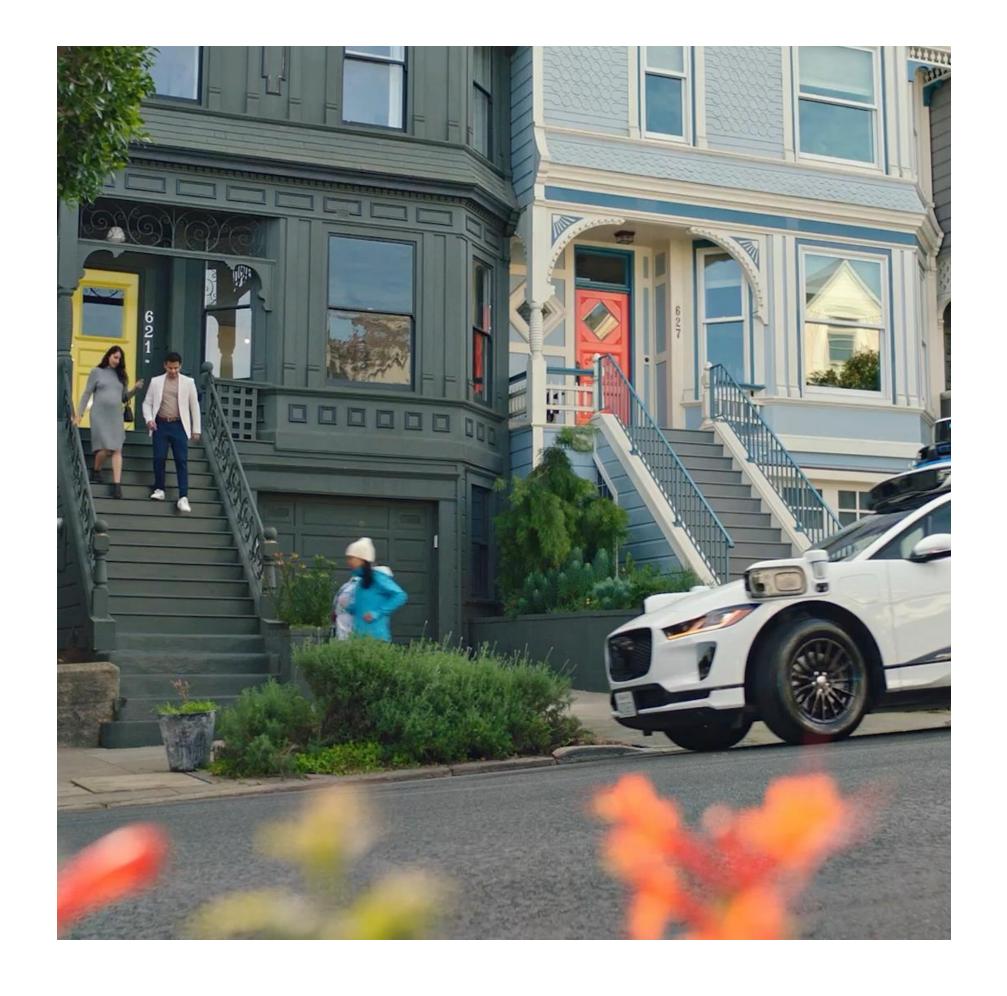
# Scaling Up Robotic Data with Minimal Supervision

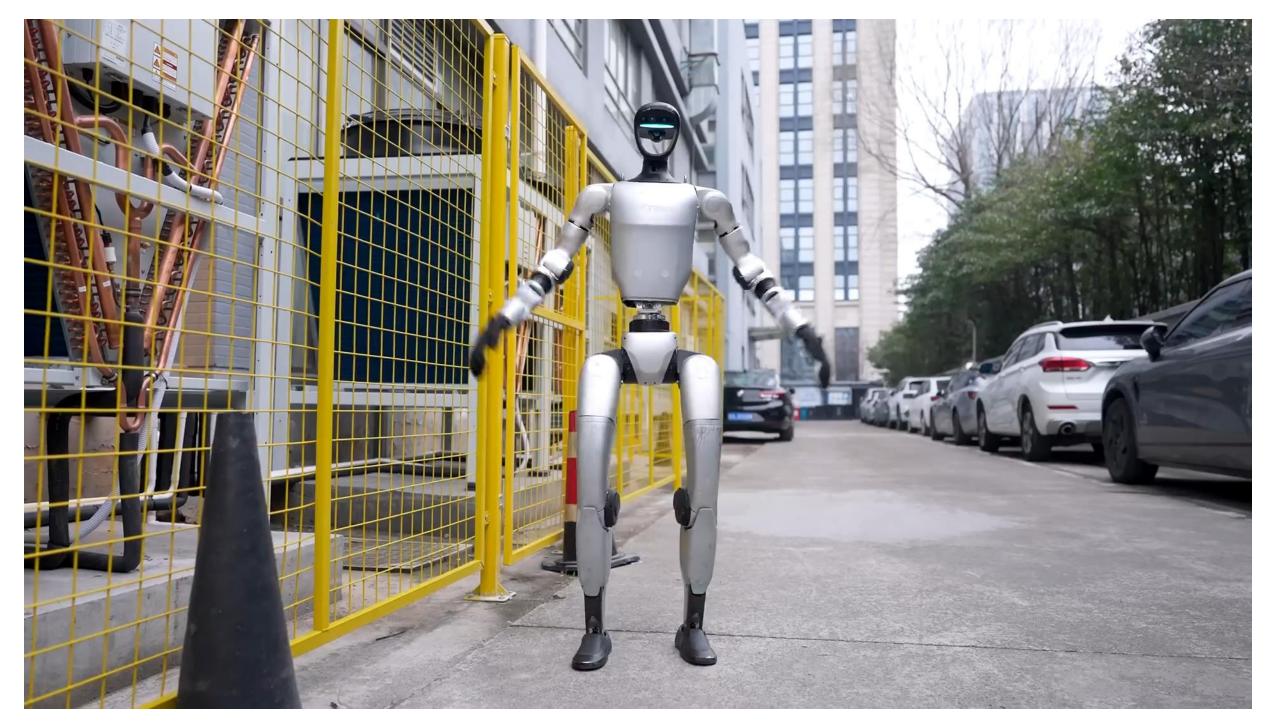
Yue Wang Space Robotics Workshop | July 29th, 2025





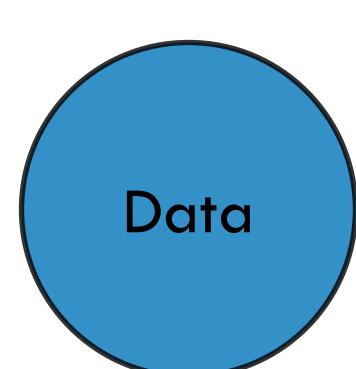


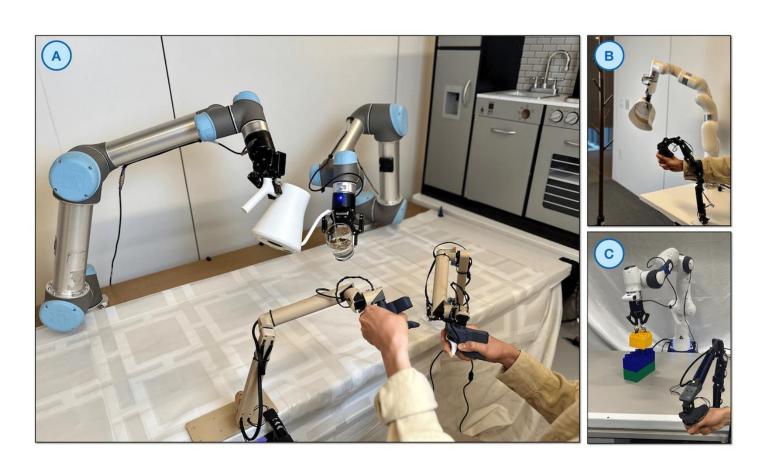
Cambrian Explosion of Robotics





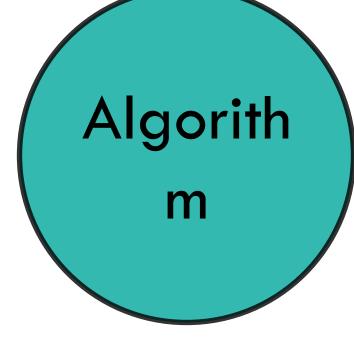


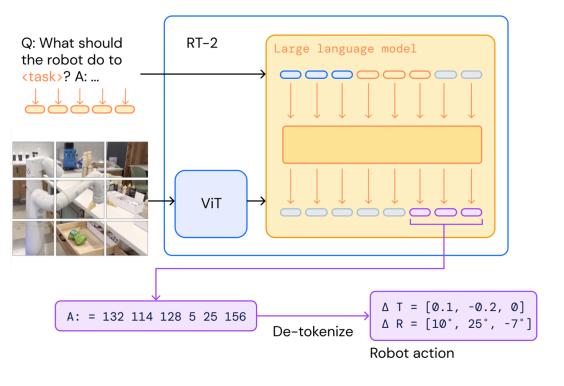






Hardwa re



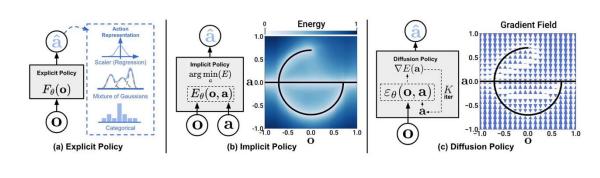


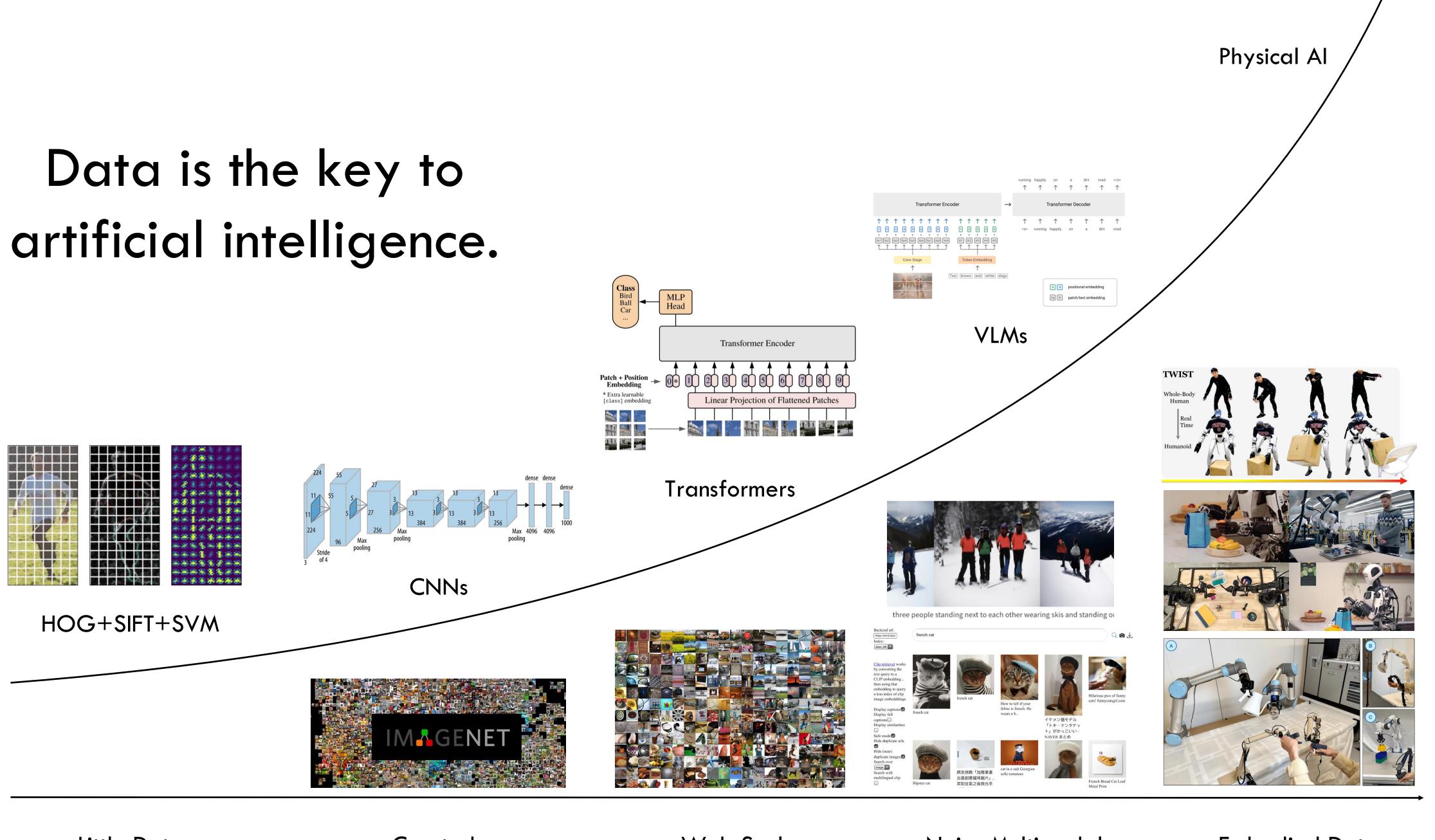




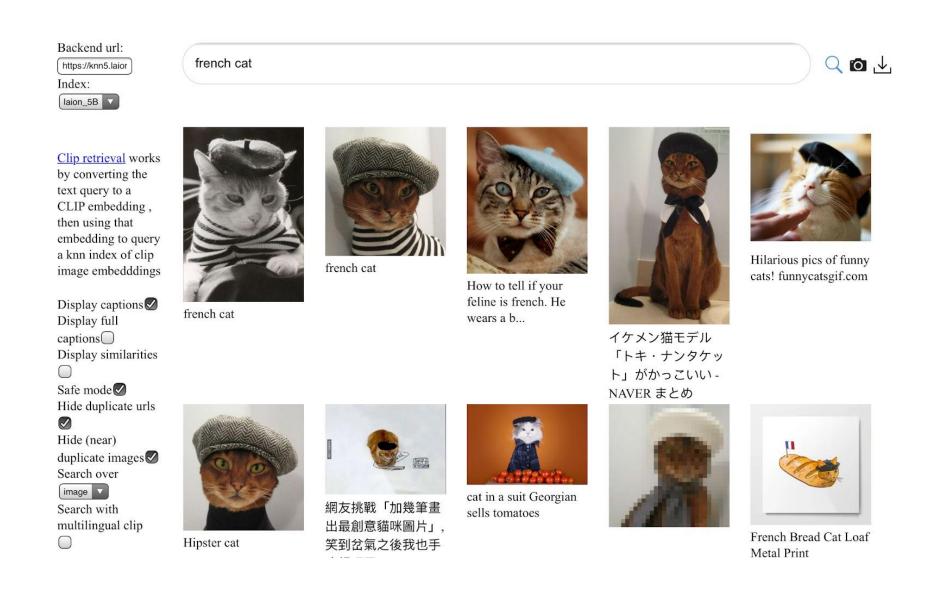
#### **Diffusion Policy**

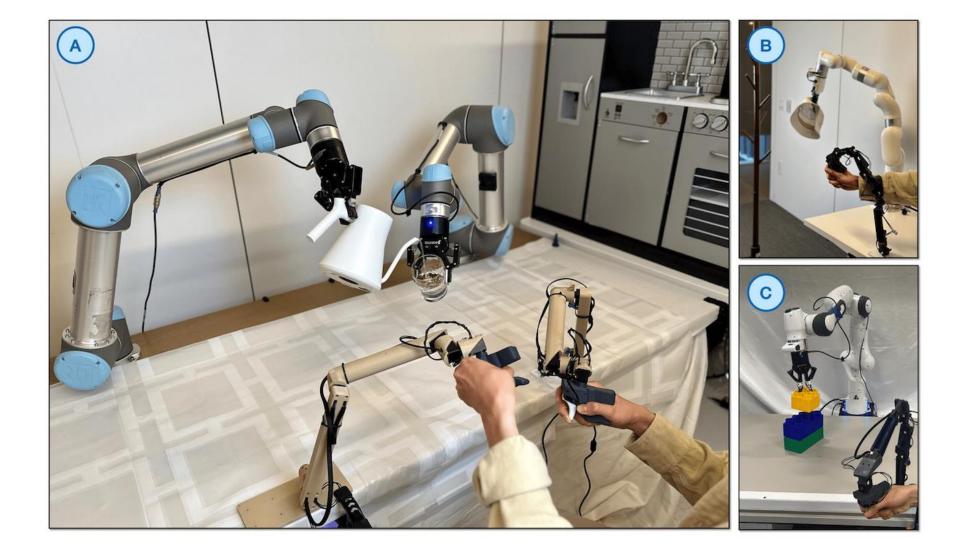
Visuomotor Policy Learning via Action Diffusion





Little Data Curated Web Scale Noisy Multimodal Embodied Data





< 1s

Ubiquitous

\$0.01 per data point

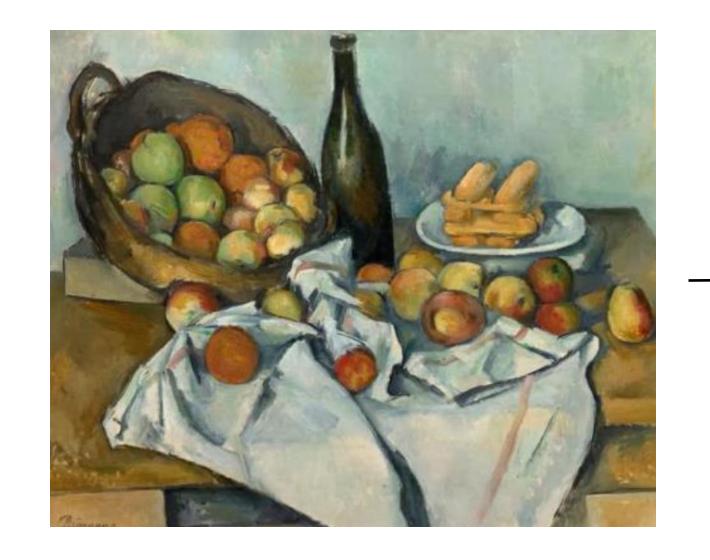
> 60s

Confined to lab environments

\$5 per data point

For space robots, data collection is even more challenging

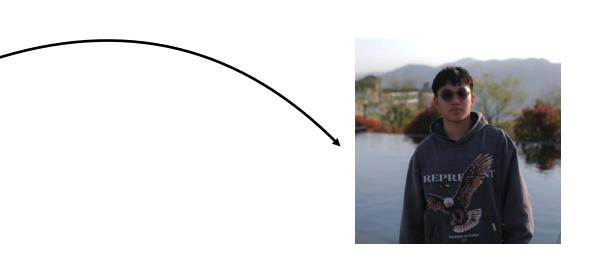
#### Can we convert a single image into a robotic data?





#### Omni Urban Scene Understanding

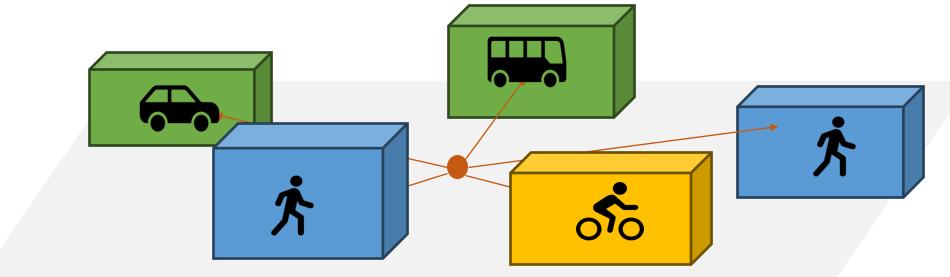




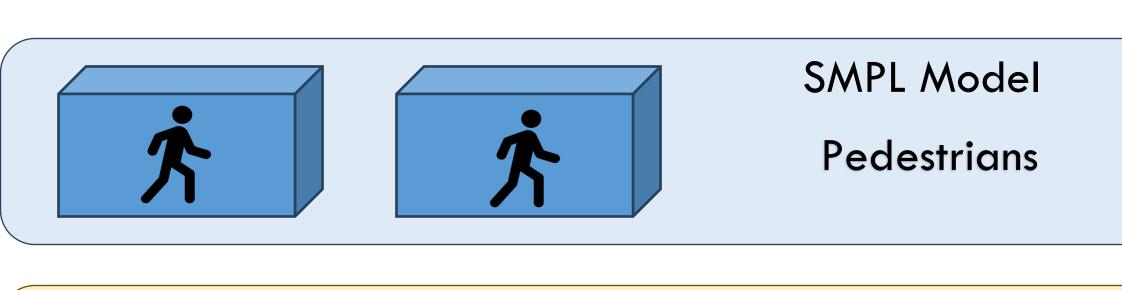
[ICLR 2025] "Omni Urban Scene Reconstruction." Chen et al.

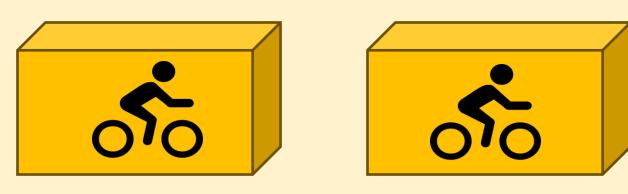
#### Scene Modeling





Sky Street Rigid Cars, Buses





Deformable Cyclists, Others

Gaussian Scene Graph

#### Applications Let People Dance!



#### Applications Driving Simulation



#### Applications Bullet time



#### Robot learning from non-robotic data



Manipulate unseen objects in unseen environments with unseen embodiments.



Learn manipulation from costly in-domain demonstrations.



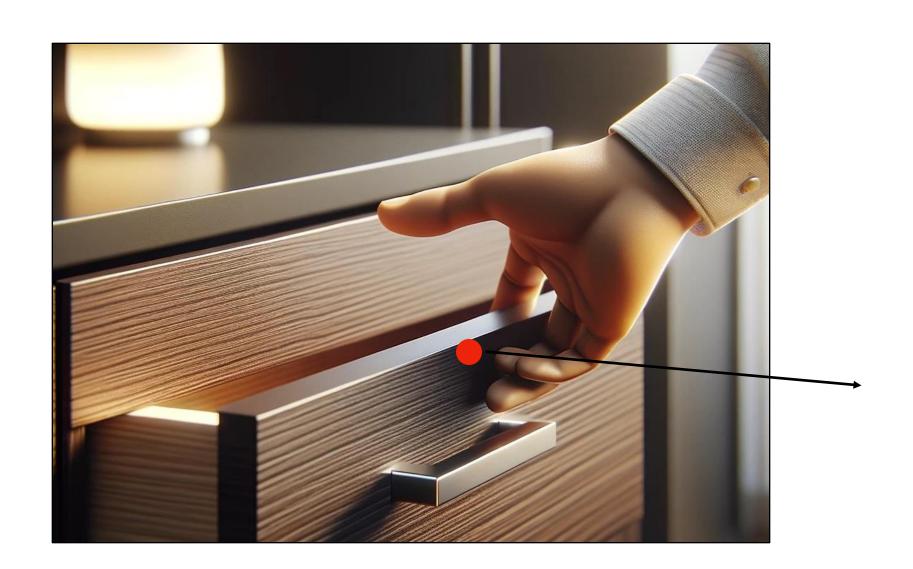
Acquire versatile manipulation capabilities from abundant out-of-domain data.

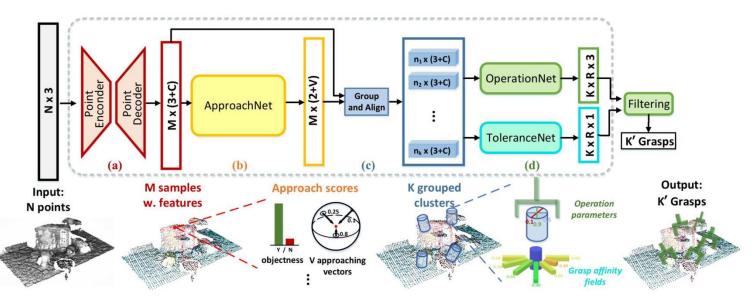


## RAM: Retrieval-Based Affordance Transfer for Generalizable Zero-Shot Robotic Manipulation

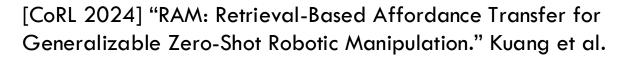


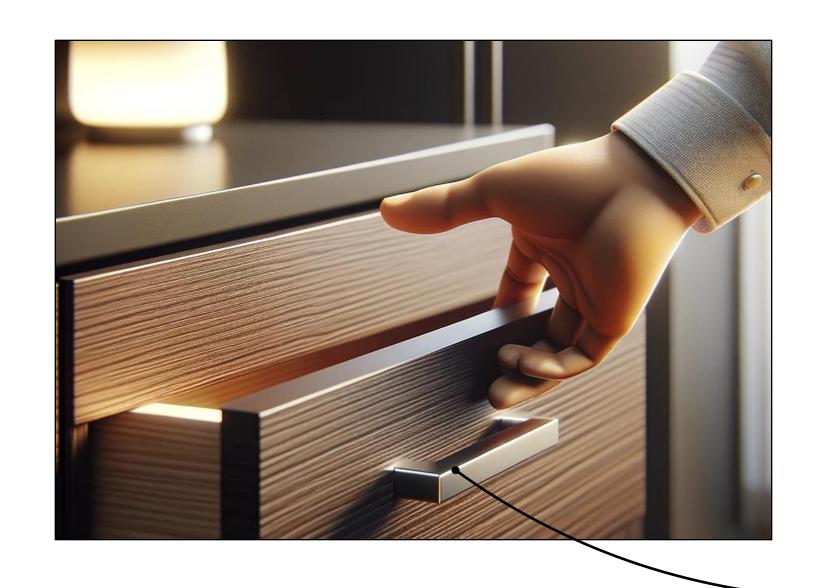
- Represent the actionable knowledge as transferrable affordance, i.e. 'where' and 'how' to act
  - 'where' to act: 3D contact point
  - 'how' to act: 3D post-contact direction
- Off-the-shelf grasp generators and motion planners for execution.











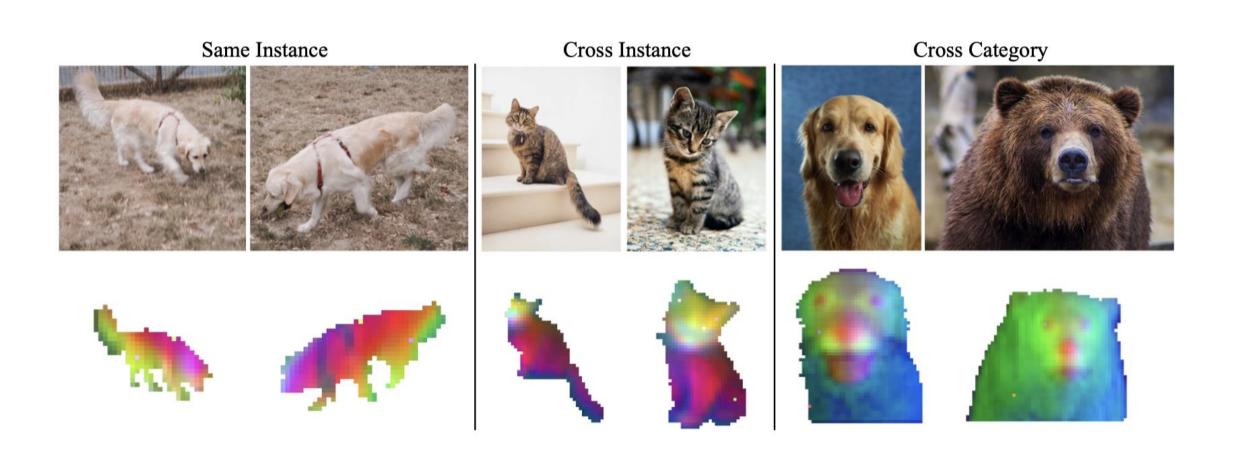


How to match points with the same semantic concepts?

Source patch Top-5 nearest neighbor cross-category target patches predicted by DIFT<sub>sd</sub>

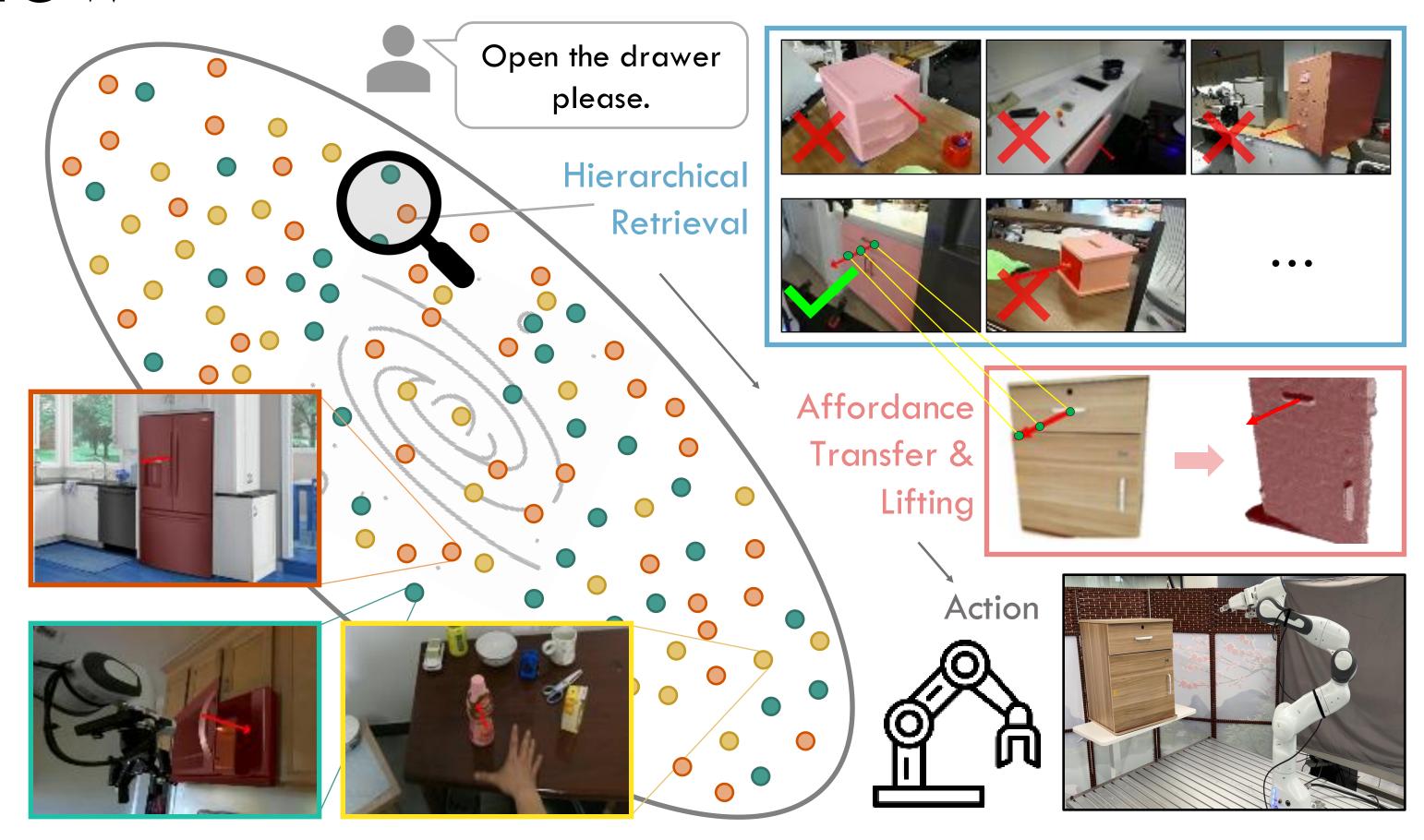


- Emergent dense correspondence of feature maps.
- Cross domain/instance/category generalization.



[NeurlPS 2023] "Emergent correspondence from image diffusion." Tang et al.

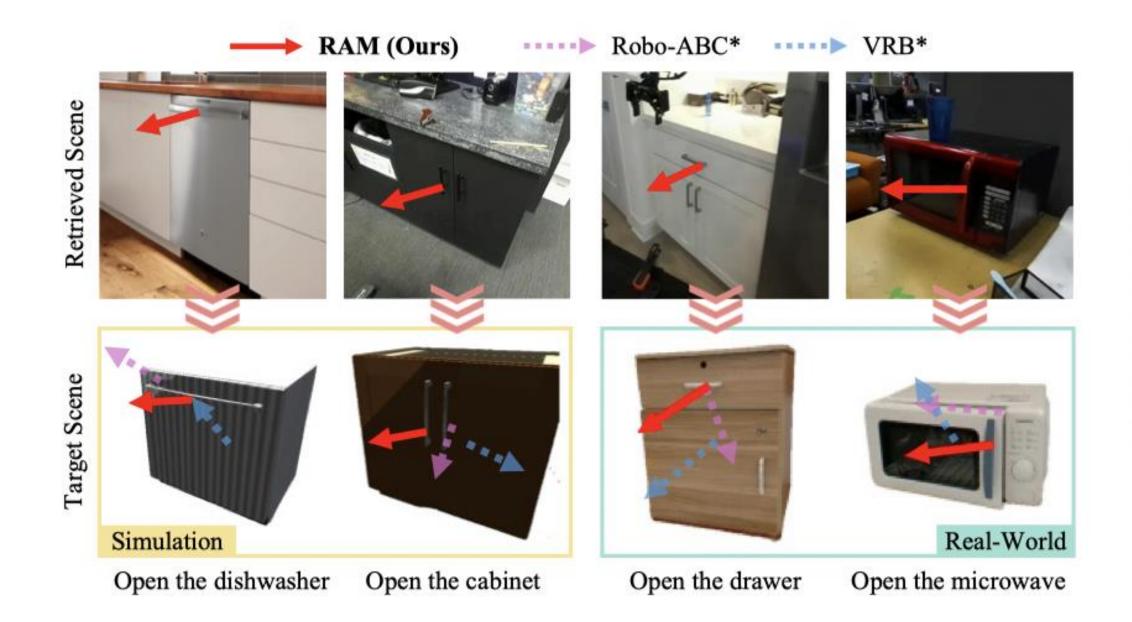
#### Overview



The Pipeline of RAM.

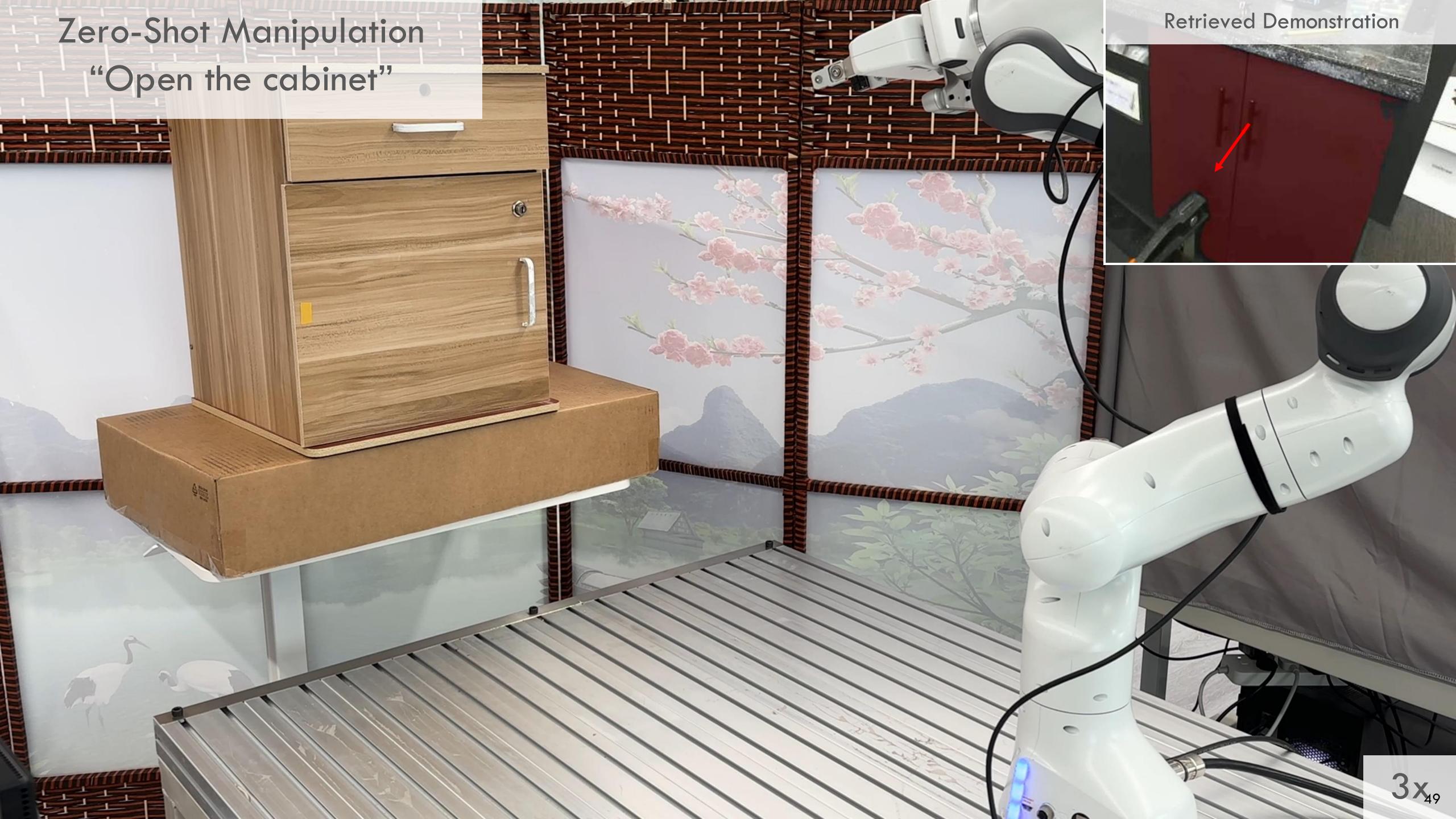
### Experiment Results

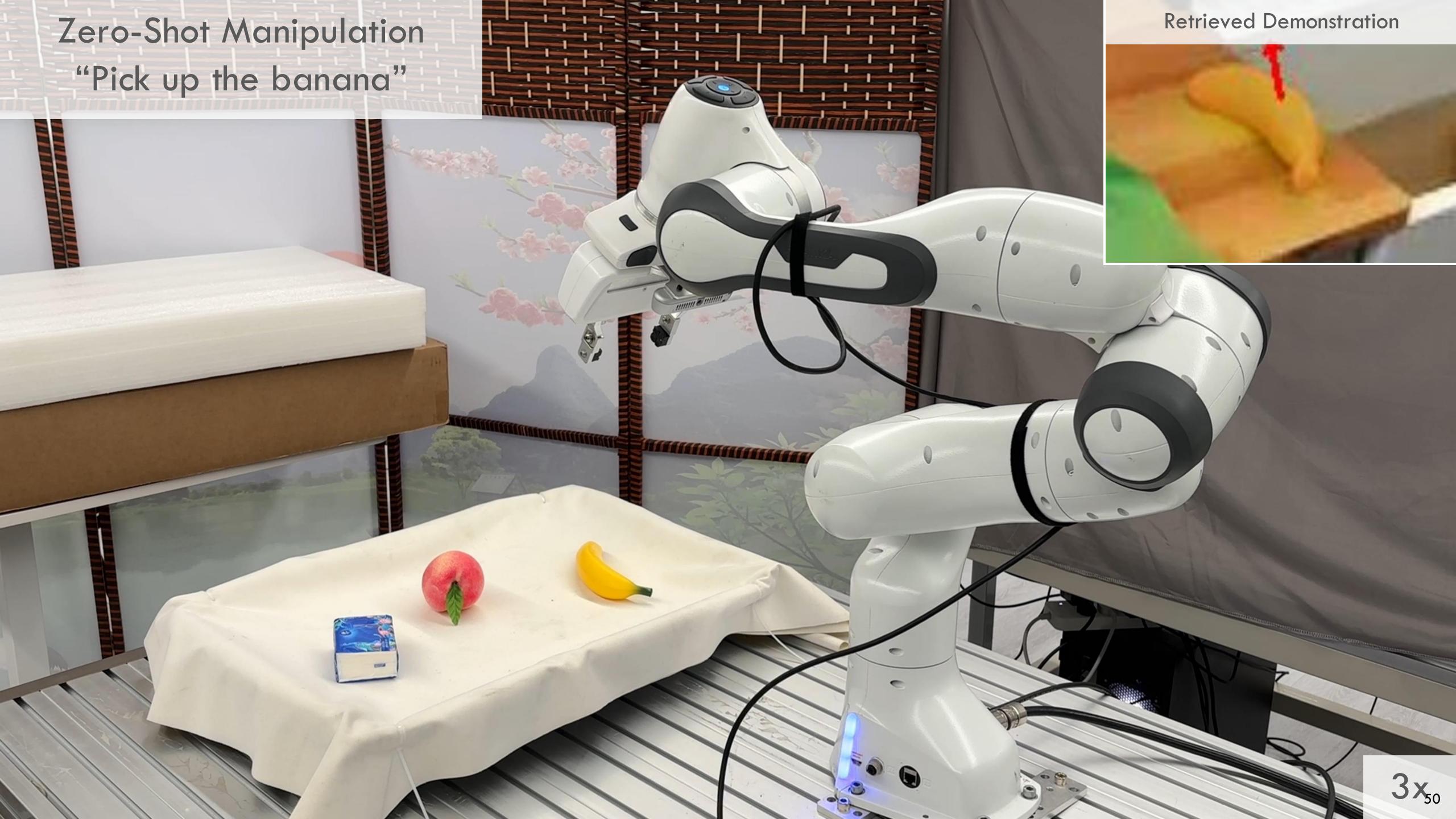
Object			اً ا					0					©iØ1	AVG
Task	0	С	0	С	0	С	0	0	0	P	P	P	P	/
Where2Act [14]	2	34	2	54	2	<b>68</b>	2 4	0	/	/	/	/	/	20.50
VRB* [12]	8	62	6	56	16	66		12	10	18	28	44	60	30.77
Robo-ABC* [44]	20	58	22	60	30	46	30	28	26	40	54	66	60	41.54
RAM (Ours)	<b>38</b>	<b>68</b>	<b>32</b>	<b>76</b>	<b>32</b>	50	<b>66</b>	<b>54</b>	<b>38</b>	<b>46</b>	<b>56</b>	<b>72</b>	<b>64</b>	<b>52.62</b>



Object						eig1	AVG
Task	0	0	0	P	P	P	/
Robo-ABC* [44] RAM (Ours)	2/5 <b>3/5</b>	1/5 <b>2/5</b>			4/5 4/5	4/5 <b>5/5</b>	50.0 <b>66.7</b>



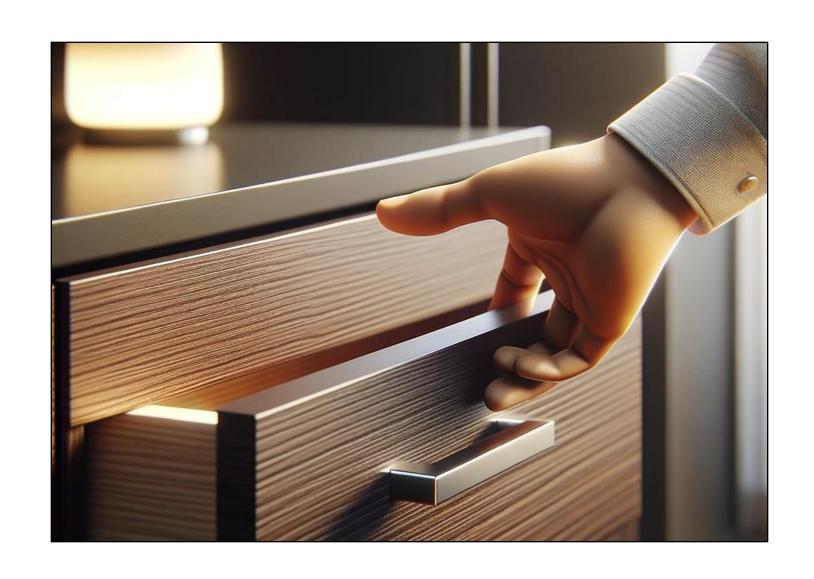




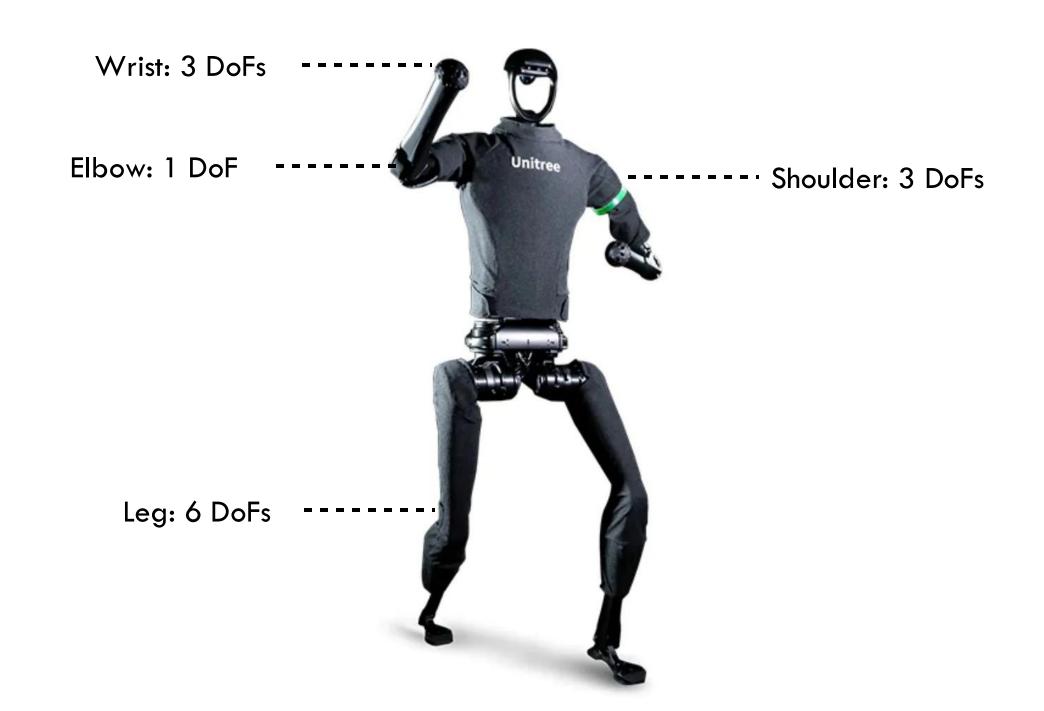








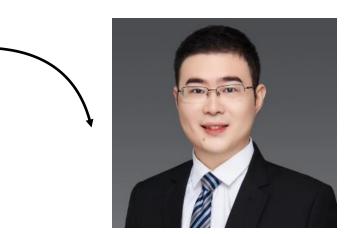


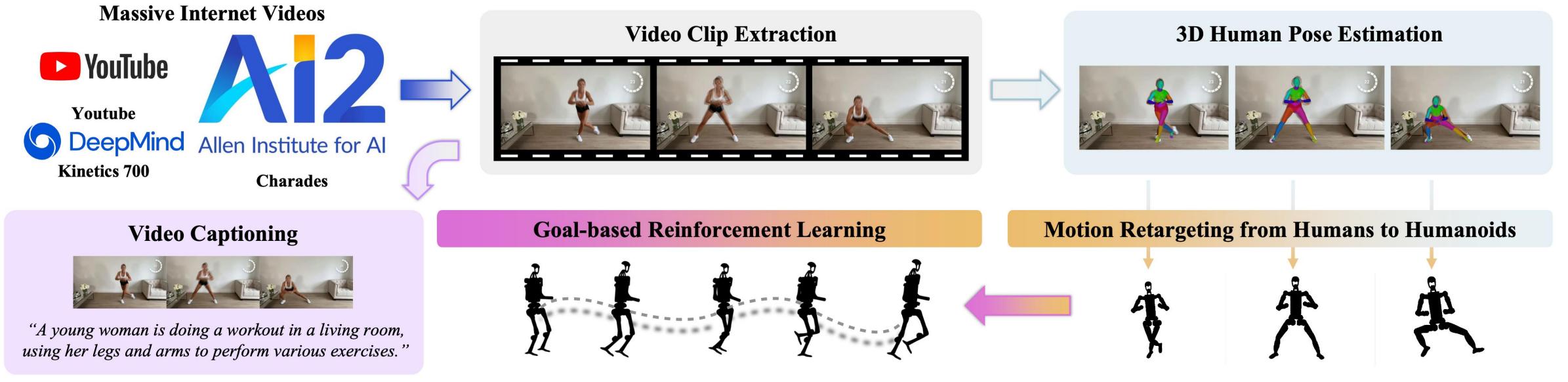


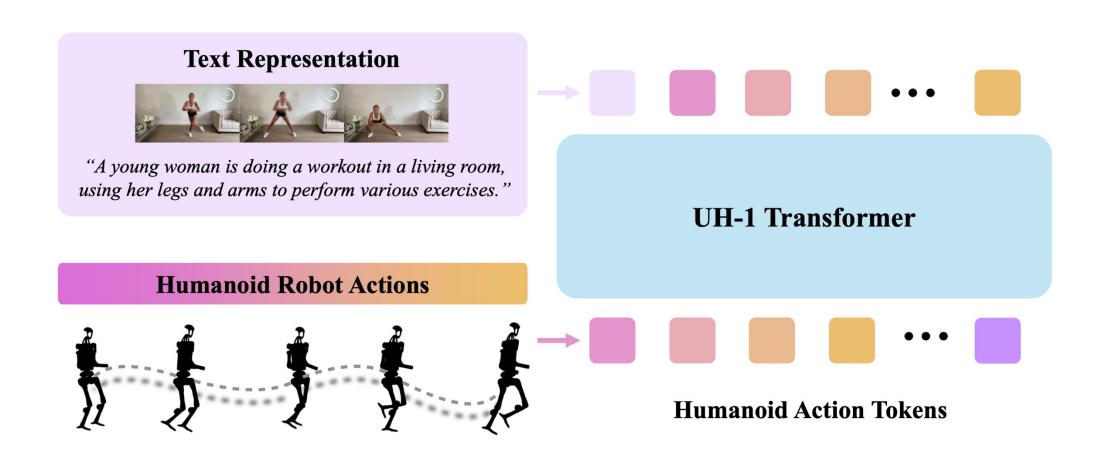
More DoFs
Not easily handled by affordance
Action retargeting is hard

How can we learn humanoid dexterity from Internet data?

## UH-1: Learning from Massive Human Videos for Universal Humanoid Pose Control







An automatic humanoid data engine

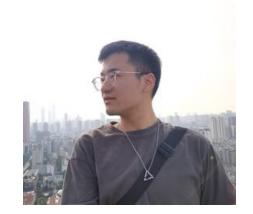
A unified whole-body control model

[In submission 2025] "Learning from Massive Human Videos for Universal Humanoid Pose Control." Mao et al.

#### Real-World Deployment of UH-1



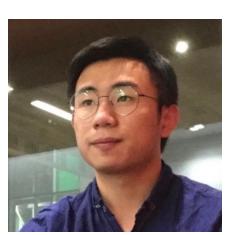
#### Acknowledgement



Junjie Ye



Jiageng Mao



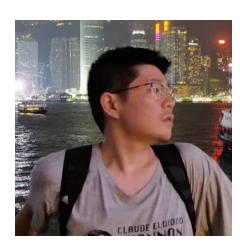
Jiawei Yang



Siheng Zhao



Cameron Smith



Yuxuan Kuang



Wei Zhou



**Boris Ivanovic** 



Sanja Fidler



Congyue Deng



Zan Gojcic



Marco Pavone



Vitor Guizilini



Leo Guibas













