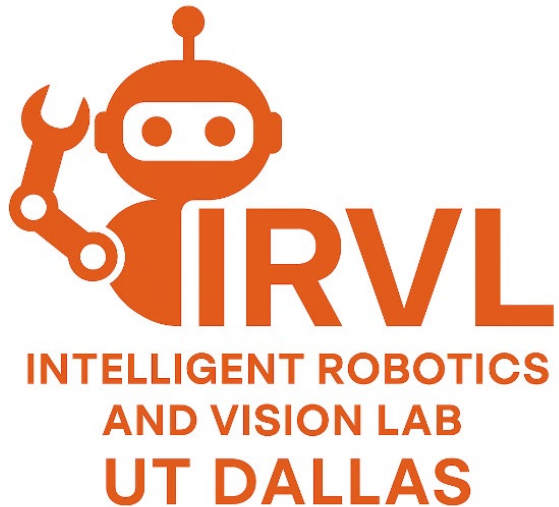


Toward Reproducible Real-World Evaluation of Robot Manipulation



Yu Xiang

Assistant Professor

Intelligent Robotics and Vision Lab

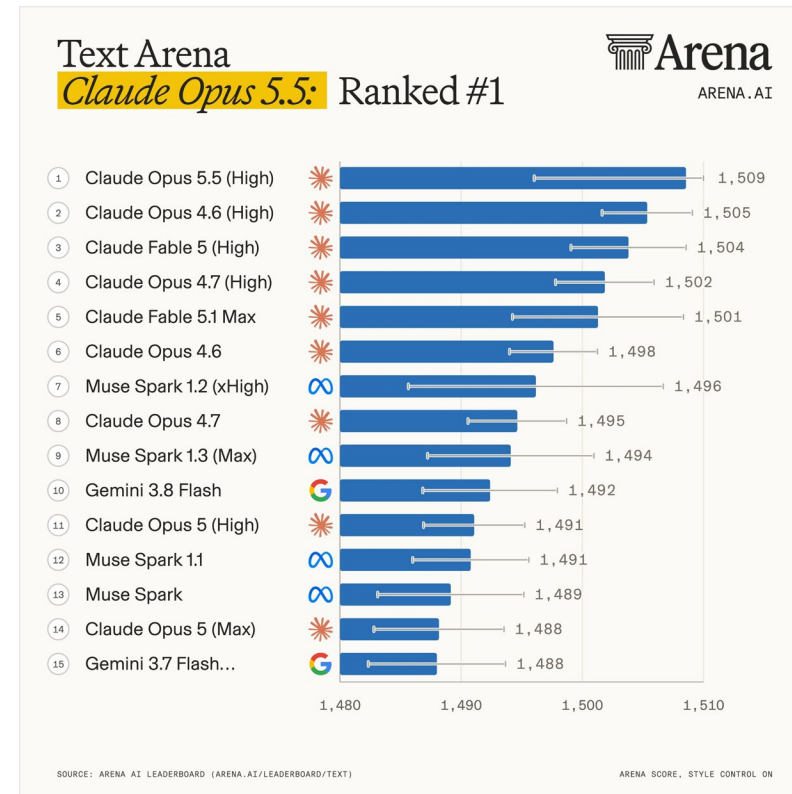
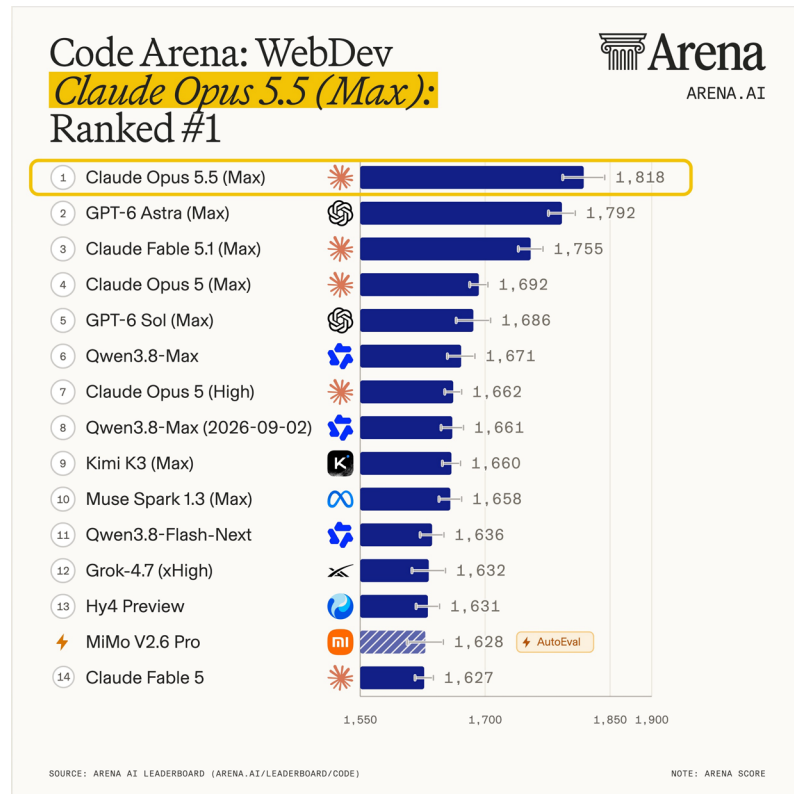
University of Texas at Dallas

9/27/2026

Reproducing Robot Manipulation Workshop @ IROS 2026

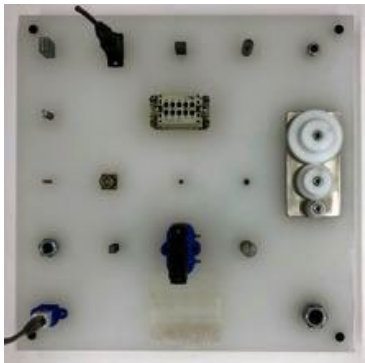
Why do we need benchmarking?

- We need to know whether we are actually making progress
 - Does it work?
 - Does it work better than existing methods?
 - Does it work beyond the conditions where it was developed?



Manipulation Benchmark

Benchmark	Type	Task	Objects	AR Tag-Free	Scene Reproducibility
Meta-World [11]	Simulation	50 tasks	Synthetic	✓	✓
RLBench [12]	Simulation	100 Tasks	Synthetic	✓	✓
robosuite [13]	Simulation	9 Tasks	Synthetic	✓	✓
Grasp Planning Protocol [10]	Real	Grasp Planning	YCB (single)	✓	✗
NIST Assembly [7]	Real	Assembly	Task Boards	✓	✓
FurnitureBench [14]	Real	Assembly	3D Printing	✗	✓
GRASPA [8]	Real	Grasping	YCB (clutter)	✗	✓
OCRTOC [15]	Real	Rearrangement	YCB + Others	✓	✗
RB2 [16]	Real	Pouring, Scooping, Zipping, Insertion	Others	✓	✗
Box and Blocks Test [17]	Real	Pick-and-Place	Blocks	✓	✗
SceneReplica (Ours)	Real	Pick-and-Place	YCB (clutter)	✓	✓



NIST Assembly board



FurnitureBench



GRASPA

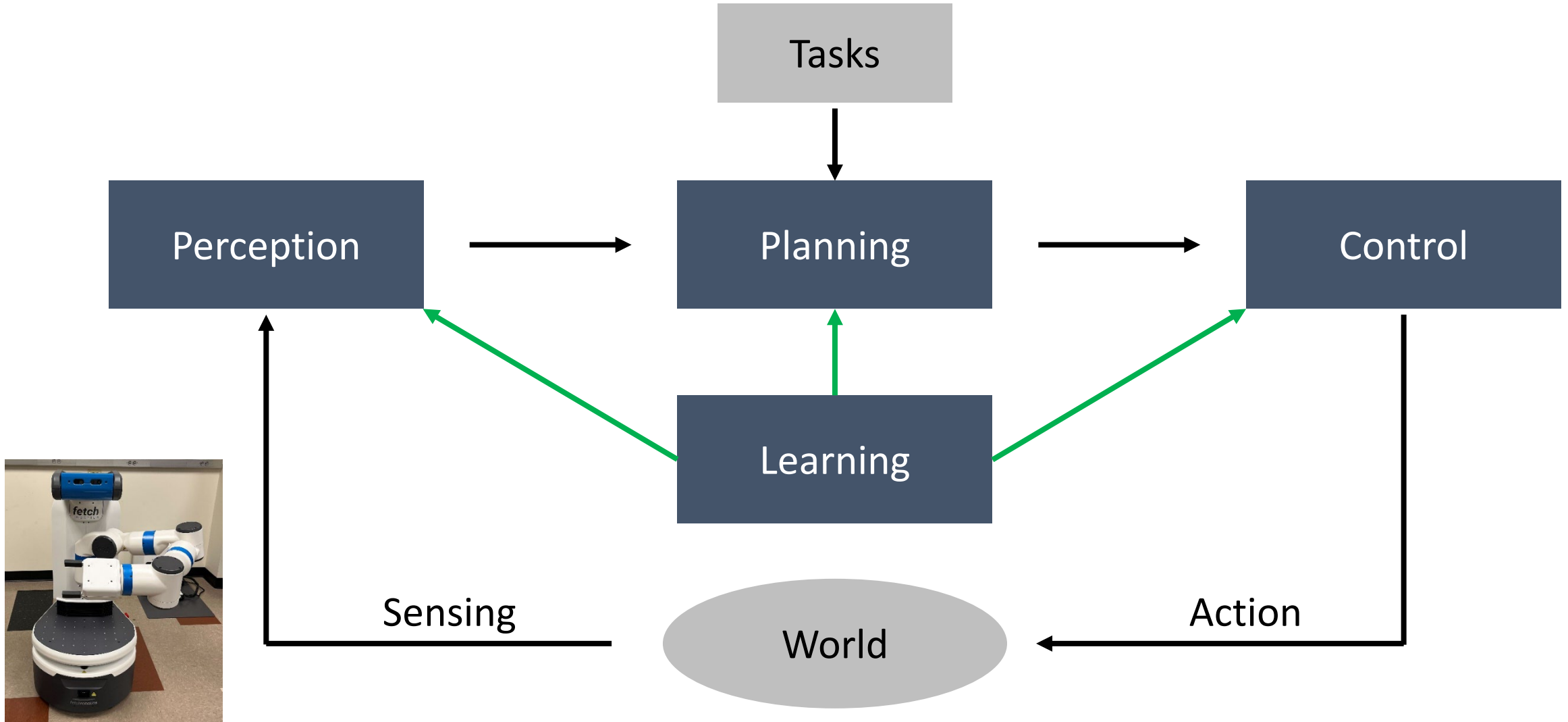


SceneReplica

SceneReplica: Benchmarking Real-World Robot Manipulation by Creating Reproducible Scenes
 Ninad Khargonkar*, Sai Haneesh Allu*, Yangxiao Lu, Jishnu Jaykumar P, Balakrishnan Prabhakaran, Yu Xiang (*equal contribution)
 In International Conference on Robotics and Automation (ICRA), 2024.

The Perception, Planning and Control Loop

Good Old Fashioned Engineering (GOFE)



SceneReplica Benchmark

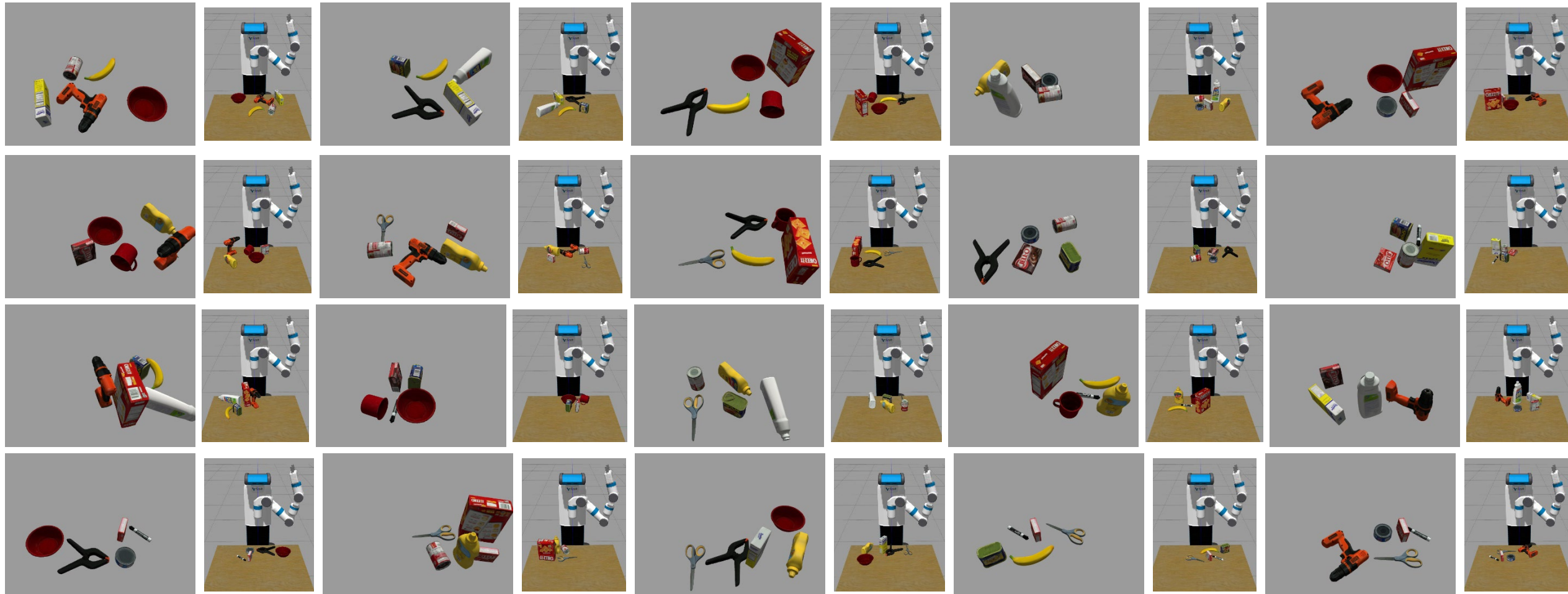
20 Scenes



Ninad Khargonkar

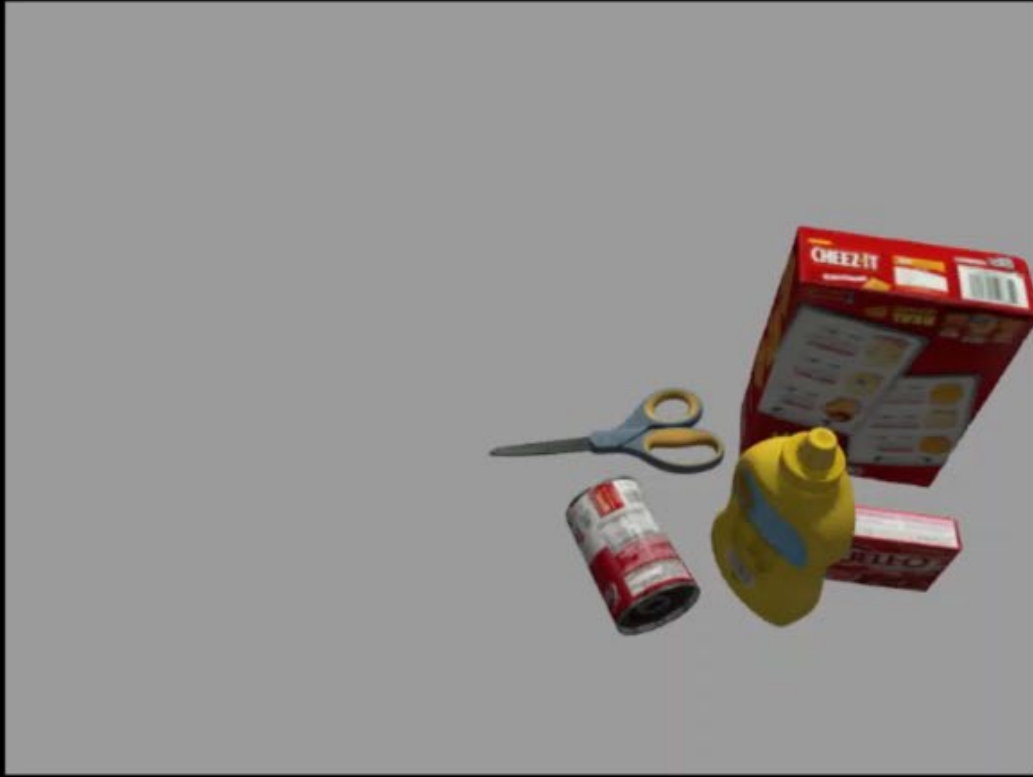


Sai Haneesh Allu



SceneReplica: <https://irvlutd.github.io/SceneReplica/>

Real-World Scene Setup



Reference Image

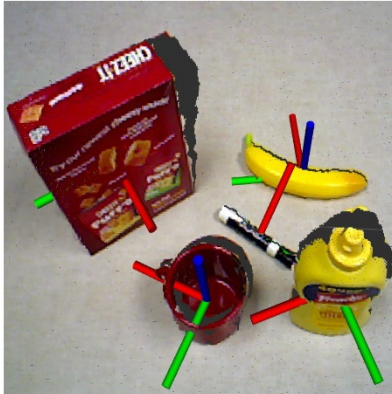


Real World Setup

Model-based Grasping vs Model-free Grasping



Input real world scene



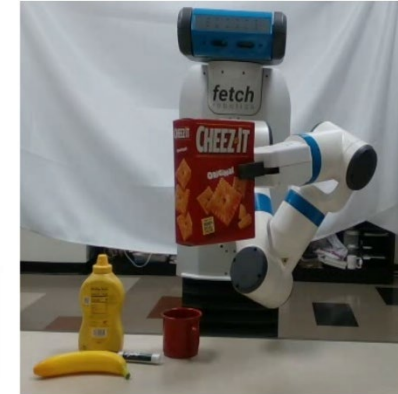
6D Pose Estimation



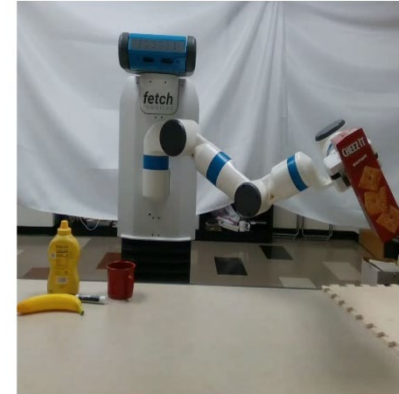
Offline Grasp Database



Motion Planning Setup



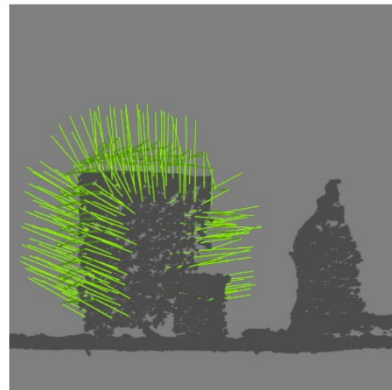
Grasping & Lifting



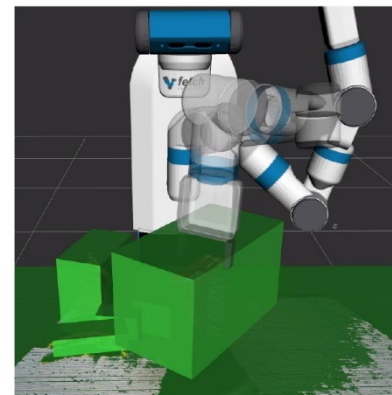
Moving arm for Dropoff



Unseen Object Segmentation



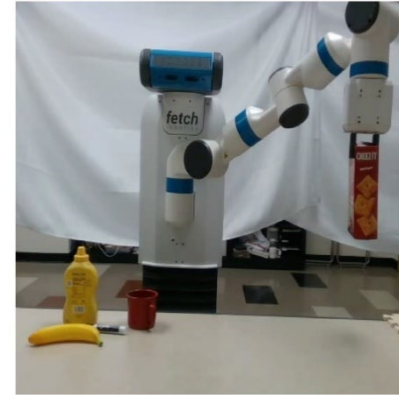
Model-free Grasp Planning



Motion Planning Setup



Grasping & Lifting

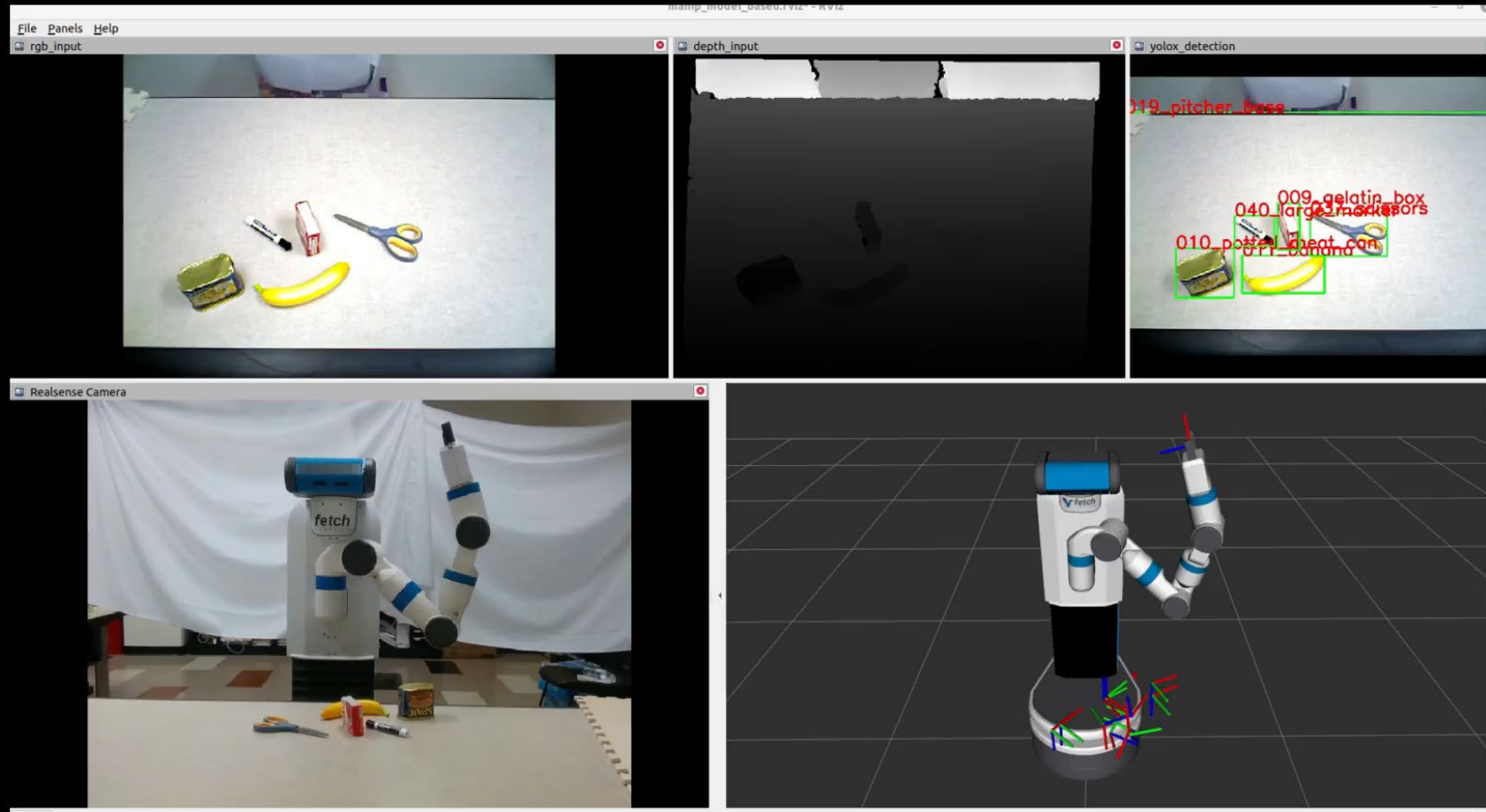


Moving arm for Dropoff

Model-based Grasping Example

[8X] SceneReplica Benchmark

GDRNPP | Graspit + Top Down | MoveIt



Realsense Capture Scene: 148 | Order: Random Rviz Capture

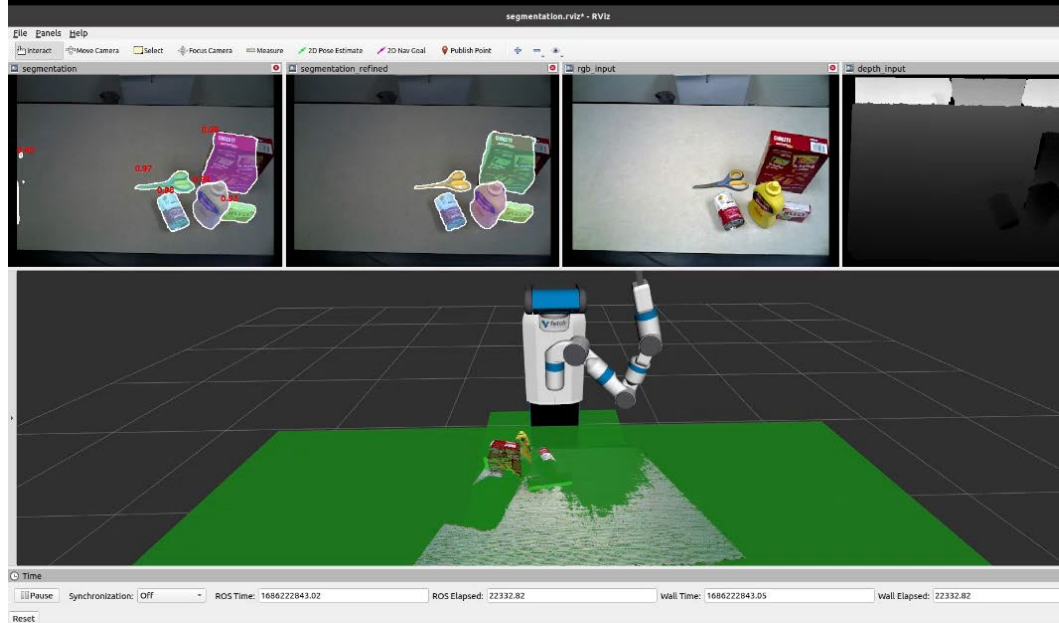
Model-free Grasping Example

8X

SceneReplica Benchmark

MSMFormer | Contact GraspNet + Top Down | MoveIt

Scene: 130 | Order: Random



Rviz Capture



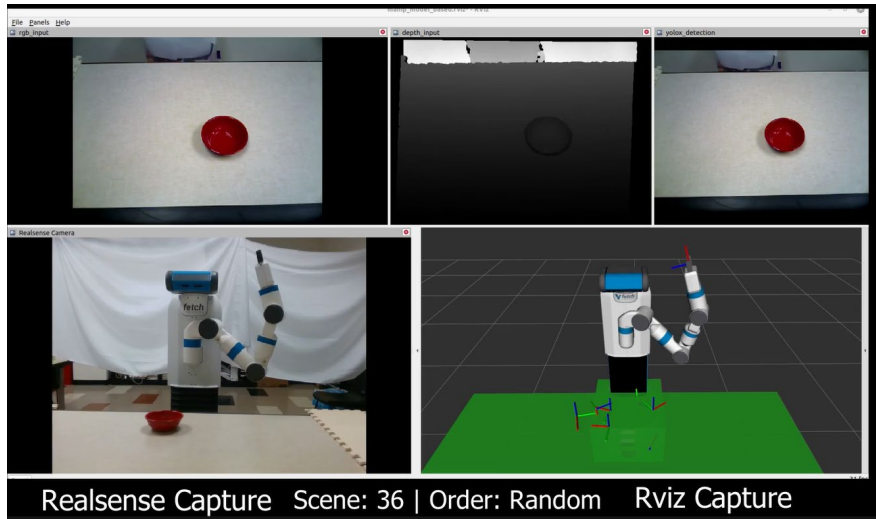
Realsense Capture

Current Leaderboard

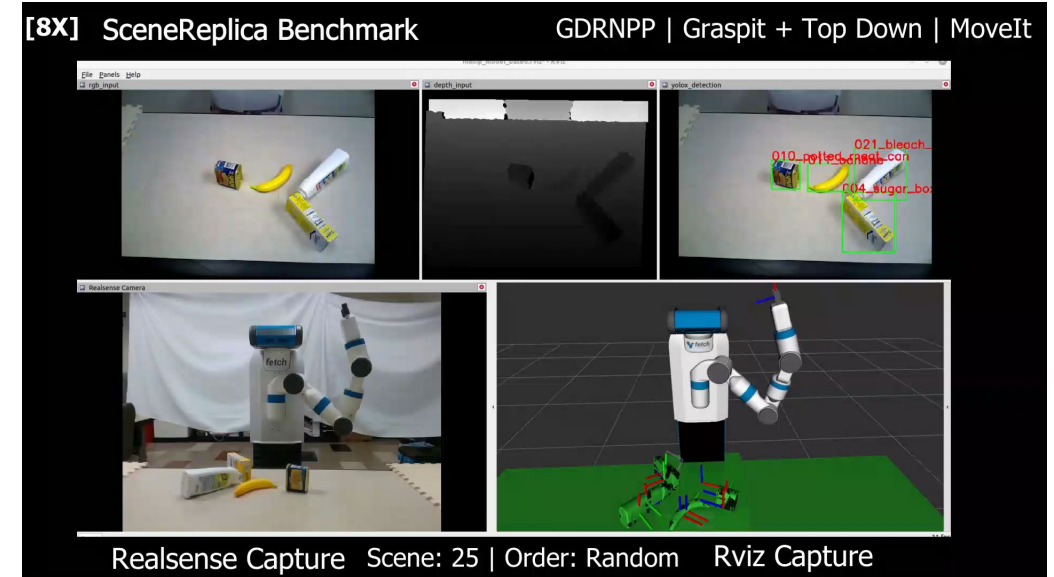
#	Perception	Grasp Planning	Motion Planning	Control	Ordering	Grasping Type	Pick & Place Success 🏆	Grasping Success	Videos
11	GDRNPP [9]	RFP [12] + Top-Down	OMPL [3]	MoveIt	Near-to-Far	Model-Based	70/100	73/100	🔗
10	MSMFormer [8]	Contact-graspnet [7] + Top-Down	GTO [11]	MoveIt	Near-to-Far	Model-Free	65/100	71/100	🔗
7	MSMFormer [8]	Contact-graspnet [7] + Top-Down	OMPL [3]	MoveIt	Fixed Random	Model-Free	61/100	70/100	🔗
3	GDRNPP [9]	GraspIt! [2] + Top-Down	OMPL [3]	MoveIt	Near-to-Far	Model-Based	66/100	69/100	🔗
7	MSMFormer [8]	Contact-graspnet [7] + Top-Down	OMPL [3]	MoveIt	Near-to-Far	Model-Free	57/100	65/100	🔗
3	GDRNPP [9]	GraspIt! [2] + Top-Down	OMPL [3]	MoveIt	Fixed Random	Model-Based	62/100	64/100	🔗
5	UCN [5]	Contact-graspnet [7] + Top-Down	OMPL [3]	MoveIt	Fixed Random	Model-Free	60/100	64/100	🔗

<https://irvlutd.github.io/SceneReplica/>

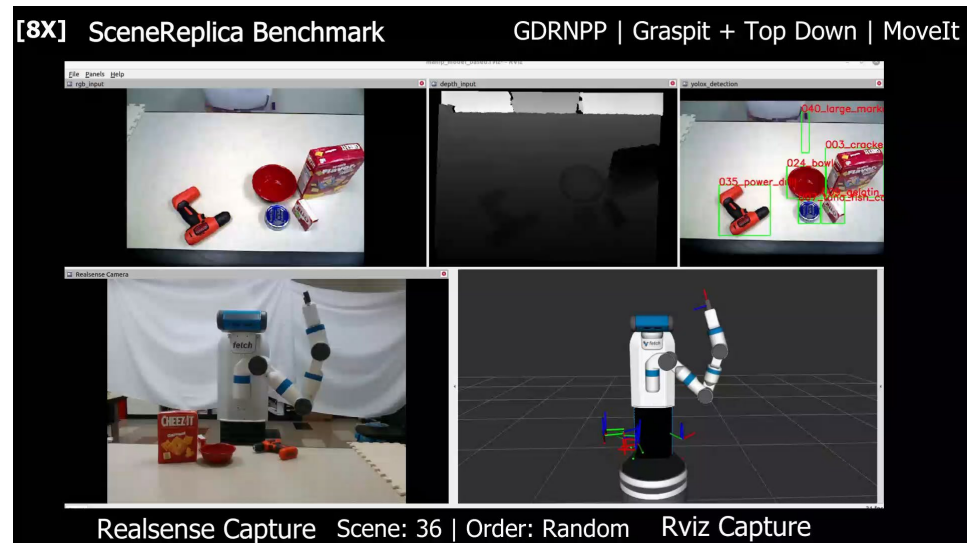
Failure Analysis (GDRNPP + Grasplt! + Top-Down)



Object Detection Error



Grasp Planning Error



Pose Estimation Error

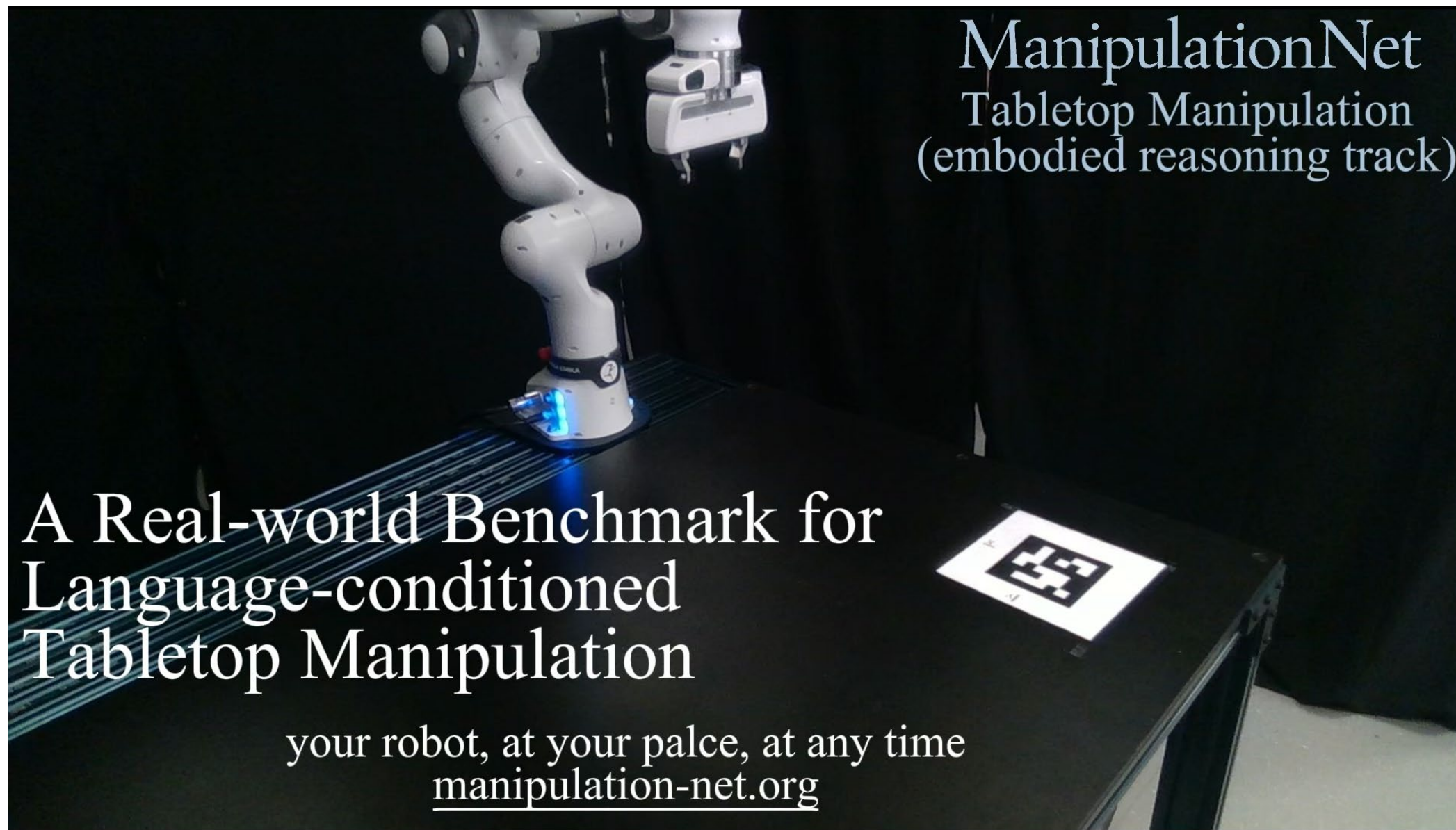
Object	Method 3			
	S	P _{EF}	P _{LF}	EF
003 cracker box	3	2	1	-
004 sugar box	5	-	-	-
005 tomato soup can	5	1	-	1
006 mustard bottle	7	-	-	-
007 tuna fish can	1	5	-	-
008 pudding box	5	-	-	-
009 gelatin box	6	-	1	-
010 potted meat can	7	-	-	-
011 banana	6	-	1	-
021 bleach cleanser	3	1	-	1
024 bowl	2	4	1	-
025 mug	4	-	1	-
037 scissors	4	3	-	-
035 power drill	1	3	2	1
040 large marker	2	4	-	-
052 extra large clamp	5	1	-	-
ALL	66	24	7	3

P_{EF}: #perception failure

P_{LF}: #planning failure

EF: #execution failure

SceneReplica Integrated into ManipulationNet



ManipulationNet Team



A Global Infrastructure for Comparable Manipulation Research Supported By
U.S. NATIONAL INSTITUTE OF STANDARDS AND TECHNOLOGY

Founding Committee



Kaiyu Hang
General Chair, Rice University



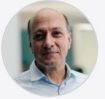
Kenneth Kimble
General Chair, U.S. National
Institute of Standards and
Technology



Yiting Chen
Technical Chair, Rice University



Edward H. Adelson
Massachusetts Institute of
Technology



Tamim Asfour
Karlsruhe Institute of Technology



Ken Goldberg
UC Berkeley



Danica Kragic
KTH Royal Institute of Technology



Xiang Li
Tsinghua University



Yunzhu Li
Columbia University



Aaron Prather
ASTM International



Nancy Pollard
Carnegie Mellon University



Maximo A. Roa-Garzon
German Aerospace Center (DLR)



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U.S. National Institute of Standards
and Technology



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**Podshara
Chanrungrameekul**
Rice University



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Tsinghua University



Kaifeng Zhang
Columbia University



Ziyang Chen
UC Berkeley



Yash Chitambar
UC Berkeley



Shuo Sha
Columbia University

Senior Advisory Board



Hui Li
Autodesk



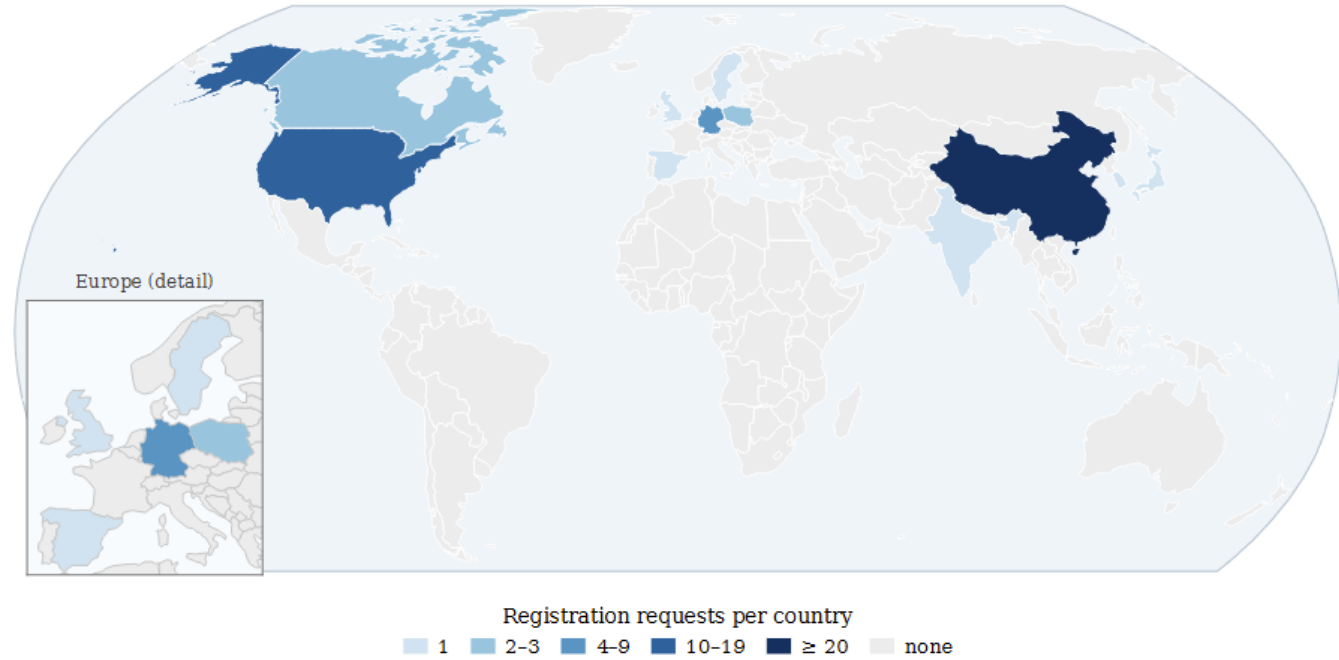
Sachin Chitta
Autodesk

<https://manipulation-net.org/>

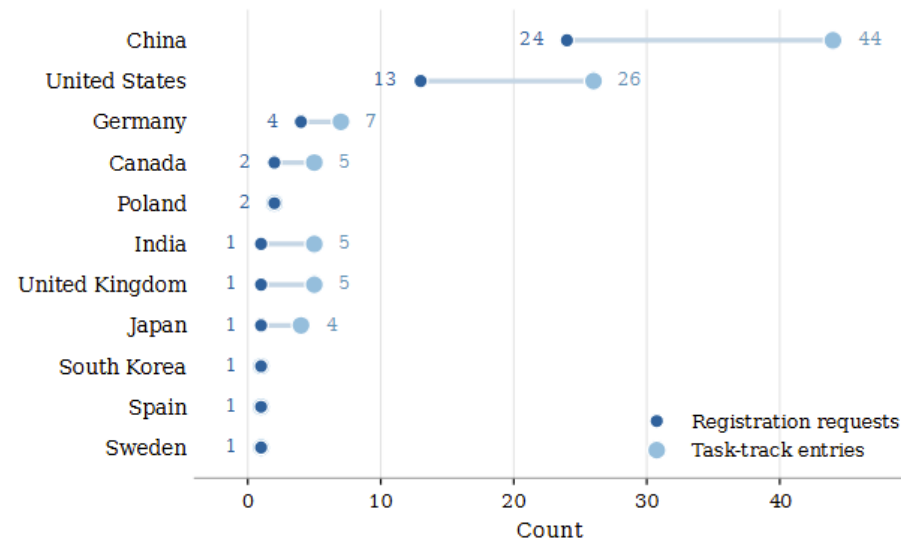
Progress

- 3 continents
- 10+ countries
- 100+ task registrations

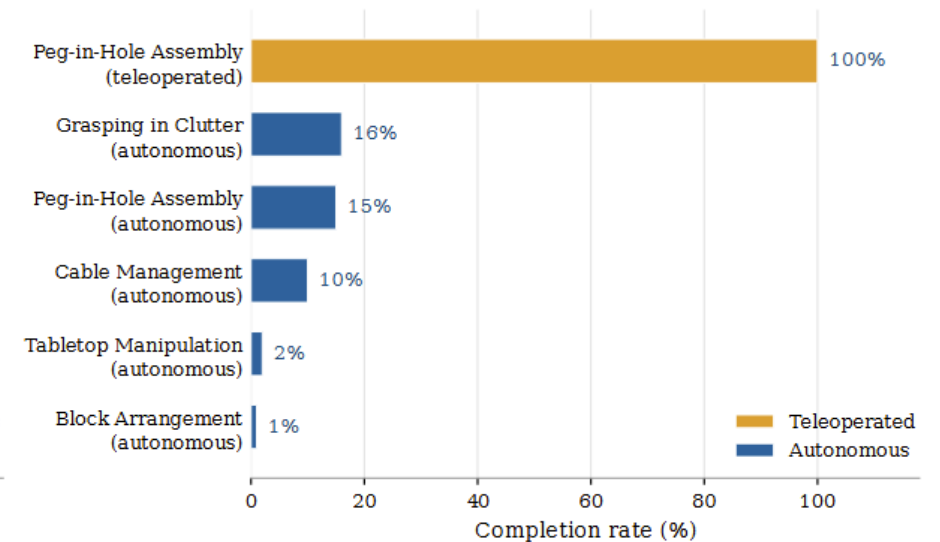
(a) Registration requests by country



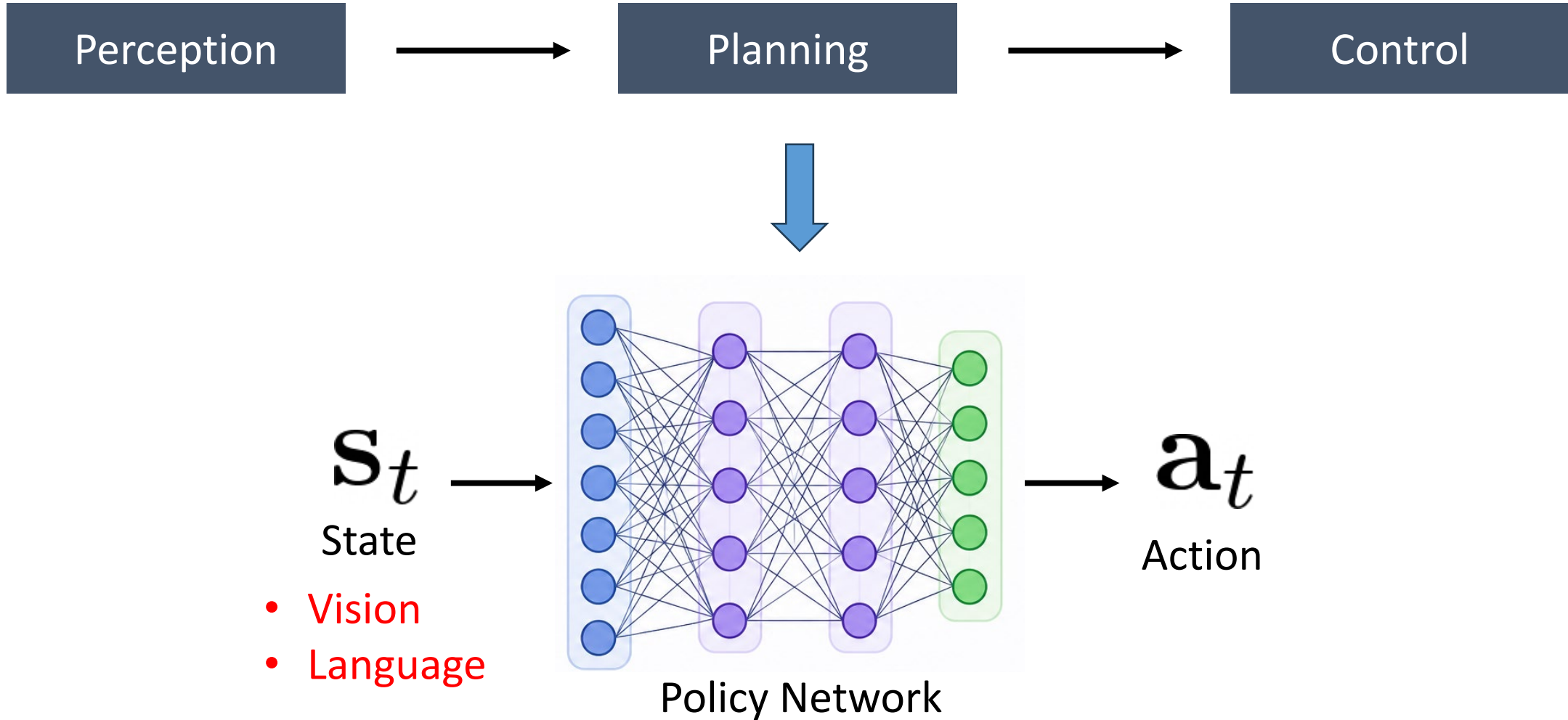
(b) Requests and task-track entries



(c) Task completion rate



End-to-End Manipulation



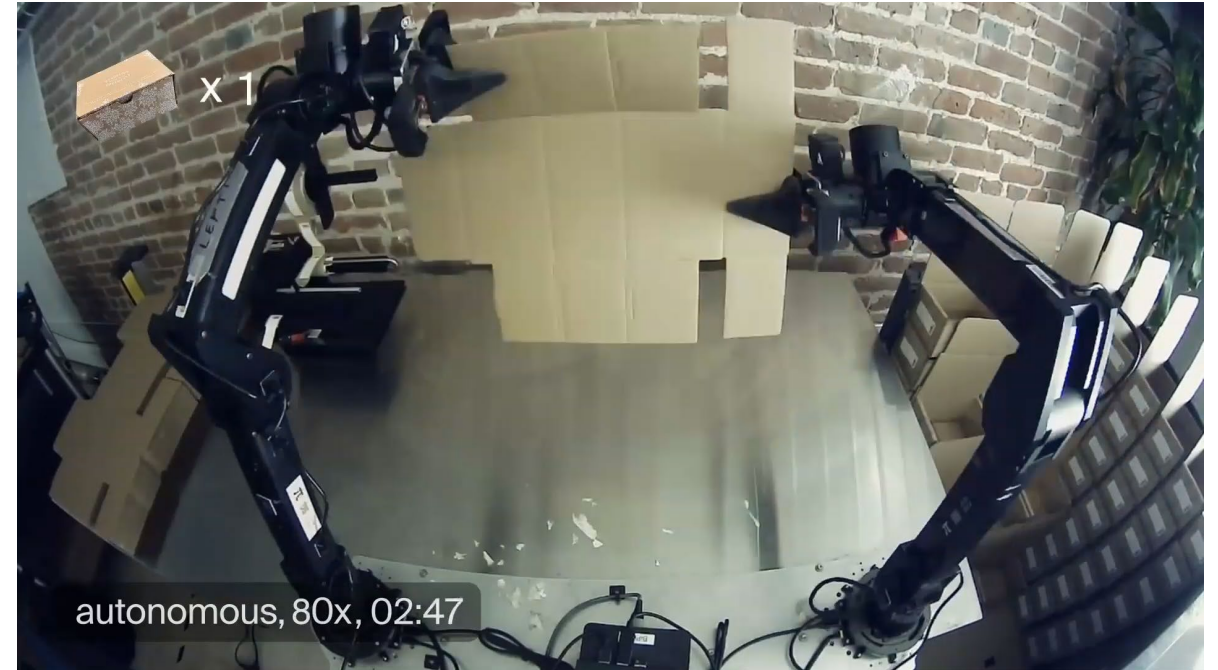
Vision-Language-Action (VLA) Models

Some Recent Breakthroughs



Sunday Robotics

<https://www.sunday.ai/blog/act-2-preview>



Physical Intelligence

<https://www.pi.website/blog/pistar06>

How Good are these VLA Models?

- We need benchmarking



We have seen some impressive robot manipulation policies recently. This is great, but for these results to be convincing (and practical) we should insist on **generalization** across 1. Object location (within some range) 2. Different instances of an object category 3. Background clutter. Authors should present experiments which demonstrate the range of variation which can be handled. Far too often the policy doesn't even generalize across all instances of an object category! Legged locomotion policies were convincing only when they worked across different terrain as in ashish-kmr.github.io/rma-legged-rob... We need to do the same for manipulanda.

VLA-REPLICA: A Low-Cost, Reproducible Benchmark for Real-World Evaluation of Vision-Language-Action Models



Alex S. Huang*



Jiahui Zhang*



Shiqing Tang



Yu Xiang

* Equal Contribution

Intelligent Robotics and Vision Lab at the University of Texas at Dallas

<https://irvlutd.github.io/VLAReplica/>

In NeurIPS Track on Datasets and Benchmarks, 2026.

VLA-Replica Benchmark

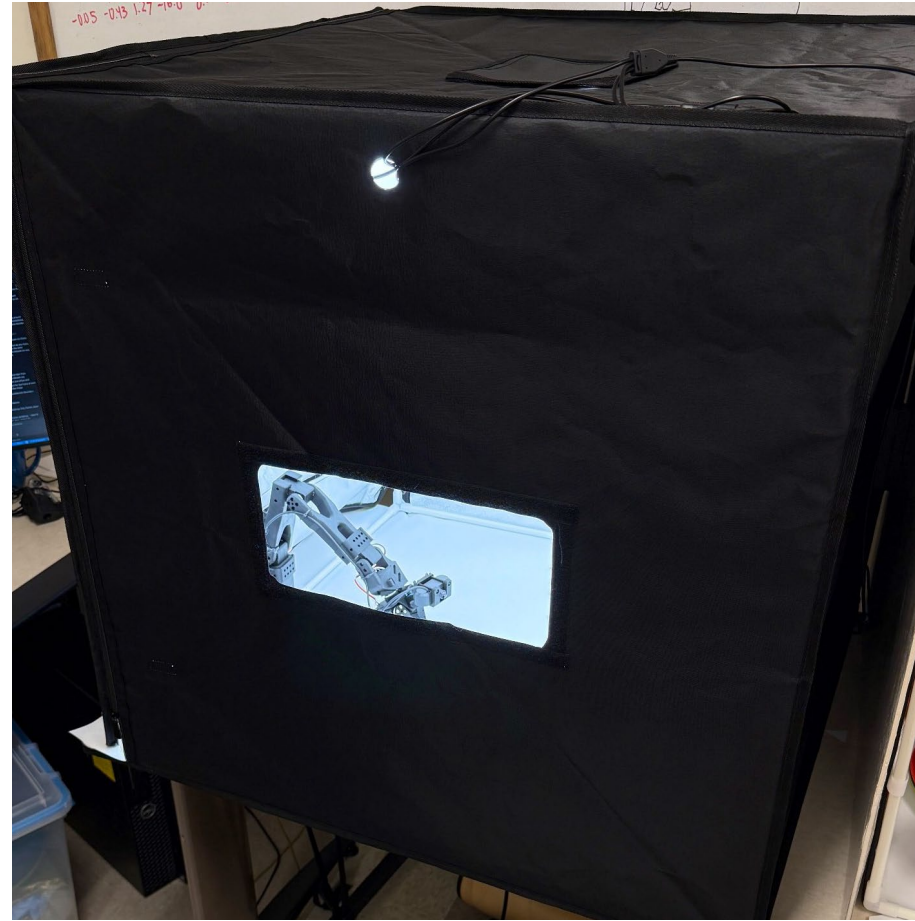
Low-cost Components



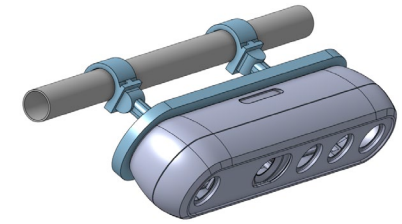
SO-101 Arm ~\$200



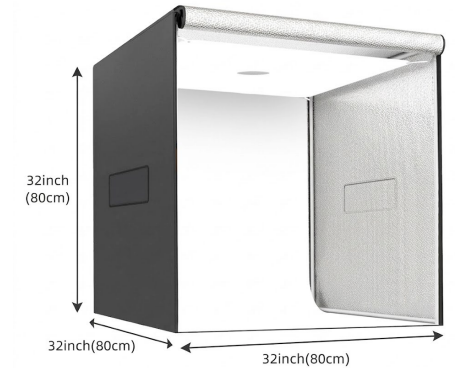
Webcam \$13.98



Real Set-up

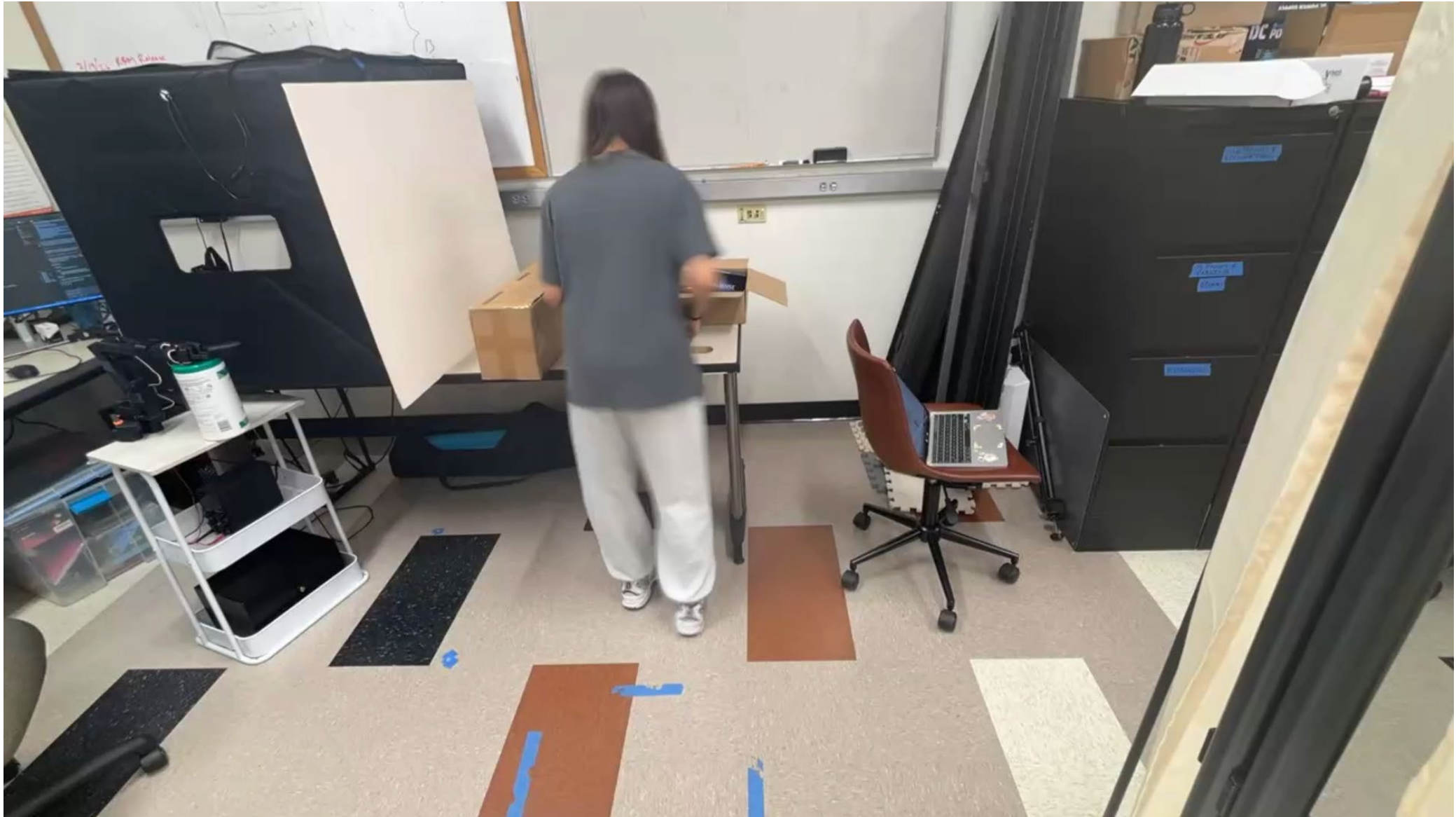


D455 Camera ~\$425



Light Box \$152.99

VLA-Replica Benchmark



VLA-Replica Benchmark

Camera Calibration

2X

Using reference images for camera calibration



Top Camera Calibration

VLA-Replica Benchmark

Task Evaluation

2X

Task 1: Fold the pink towel in half.

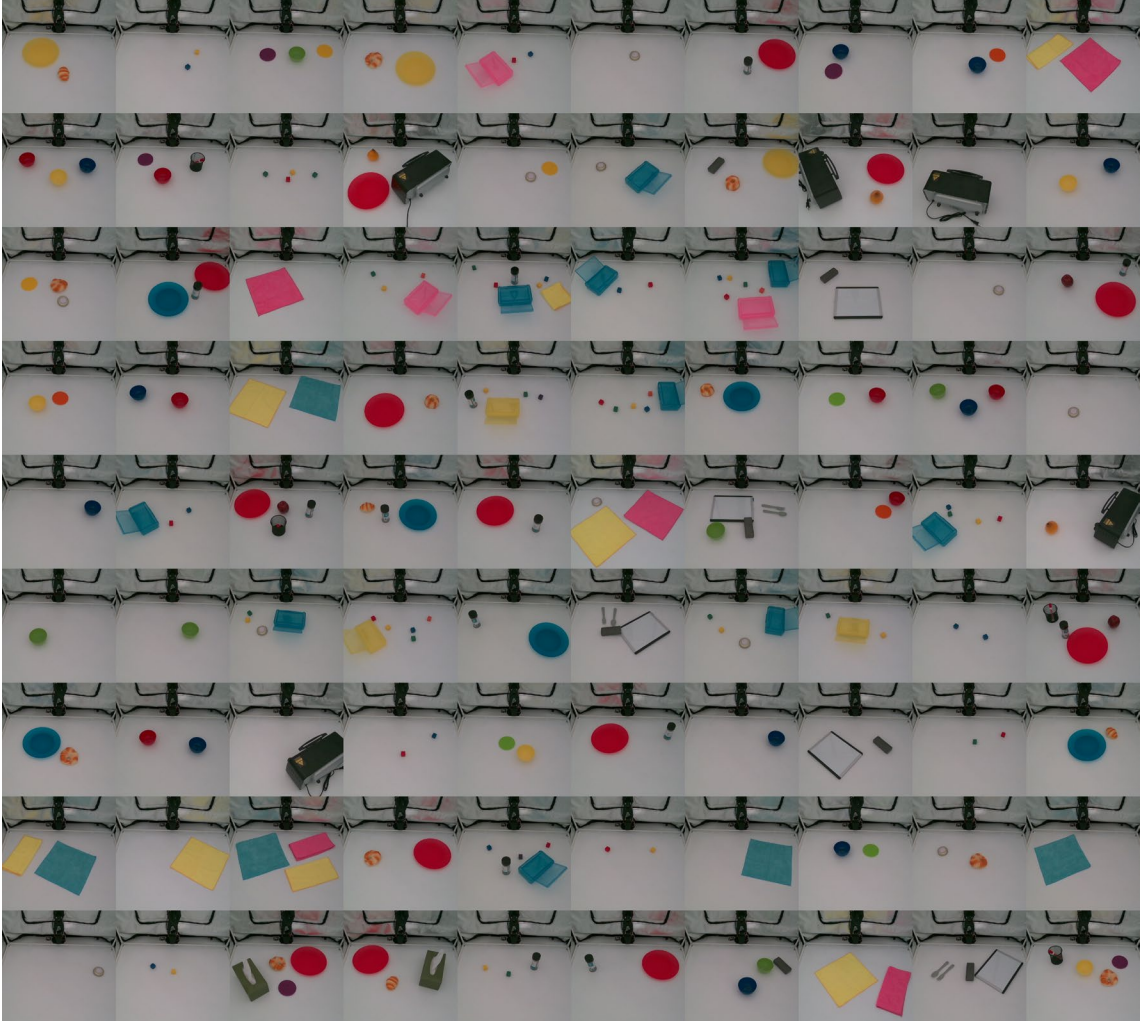


Reference Placement



Align placement with reference frame

VLA-Replica Benchmark: Tasks



We provide:

- 500 demos across 10 tasks
- 10 In-distribution tasks
- 8 Out-of-distribution tasks
- 5 reference frames per task

VLA-Replica Benchmark: Tasks

Task #	Task Type	Training & ID Eval Task Example	OOD Eval Task Example
1	Pick-and-Place	Put bread in red/blue plate	Put bread on yellow plate
2		Put red bowl on purple coaster	Put green bowl on yellow coaster
3		Stack blue cube on red cube	Stack green cube on green cube
4		Put all 2 or 3 blocks in blue box	Put all 3 or 4 blocks in pink box
5	Object Interaction	Fold pink/yellow towel in half	Fold blue towel in half
6		Open oven	N/A
7		Clean whiteboard with eraser	N/A
8	Memory	Pour pepper 1,2 or 3 times to red plate	Pour pepper 4 or 5 times to blue plate
9		Lift green/red/blue bowl 1 or 3 times	Lift yellow bowl 2, 4 or 5 times
10		Press button 1 or 3 times	Press button 2, 4 or 5 times

VLA-Replica Benchmark: Leaderboard



VLA-Replica-ID



VLA-Replica-OOD

3 runs eval with std

Rank	Method	Average	Put bread on plate	Put bowl on coaster	Stack block on block	Put all blocks into box	Fold towel	Open oven	Erase whiteboard	Shake pepper times
1	$\pi_{0.5}$ [6]	0.53 ± 0.05	0.80 ± 0.20	0.87 ± 0.12	0.33 ± 0.12	0.47 ± 0.12	1.00 ± 0.00	0.53 ± 0.12	0.33 ± 0.12	0.33 ± 0.12
2	GR00T N1.7 [8]	0.42 ± 0.06	0.80 ± 0.00	0.87 ± 0.12	0.00 ± 0.00	0.29 ± 0.08	1.00 ± 0.00	0.47 ± 0.31	0.40 ± 0.20	0.20 ± 0.12
3	MolmoAct2-SO100_101 [7]	0.39 ± 0.06	0.93 ± 0.12	0.73 ± 0.12	0.27 ± 0.12	0.27 ± 0.12	0.73 ± 0.12	0.13 ± 0.23	0.53 ± 0.12	0.13 ± 0.12
4	π_0 [5]	0.32 ± 0.03	0.67 ± 0.12	0.67 ± 0.31	0.00 ± 0.00	0.00 ± 0.00	0.67 ± 0.23	0.20 ± 0.00	0.33 ± 0.12	0.27 ± 0.12
5	SmolVLA [3]	0.29 ± 0.04	0.80 ± 0.20	0.20 ± 0.20	0.27 ± 0.12	0.00 ± 0.00	0.73 ± 0.12	0.33 ± 0.12	0.20 ± 0.00	0.00 ± 0.12
6	ACT [1]	0.21 ± 0.08	0.33 ± 0.12	0.07 ± 0.12	0.00 ± 0.00	0.00 ± 0.00	0.60 ± 0.20	0.33 ± 0.12	0.20 ± 0.00	0.27 ± 0.12
7	X-VLA [4]	0.13 ± 0.02	0.27 ± 0.12	0.13 ± 0.12	0.00 ± 0.00	0.00 ± 0.00	0.53 ± 0.12	0.07 ± 0.12	0.00 ± 0.00	0.20 ± 0.12
8	DiT-D [2]	0.11 ± 0.05	0.20 ± 0.20	0.13 ± 0.12	0.00 ± 0.00	0.07 ± 0.12	0.20 ± 0.00	0.20 ± 0.35	0.20 ± 0.00	0.07 ± 0.12
9	DiT-F [2]	0.07 ± 0.05	0.13 ± 0.23	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.00	0.27 ± 0.12	0.13 ± 0.23	0.13 ± 0.12	0.00 ± 0.12

VLA-Replica Benchmark: Leaderboard



VLA-Replica-ID



VLA-Replica-OOD

3 runs eval with std

Rank	Method	Average	Put bread on plate	Put bowl on coaster	Stack block on block	Put all blocks into box	Fold towel	Shake pepper n times	Lift bowl n times	Press button times
1	MolmoAct2-SO100_101 7	0.47 ± 0.05	0.87 ± 0.12	0.87 ± 0.12	0.60 ± 0.00	0.20 ± 0.20	0.87 ± 0.12	0.13 ± 0.12	0.13 ± 0.12	0.07 ± 0.0
2	$\pi_{0.5}$ 6	0.39 ± 0.10	0.93 ± 0.12	0.60 ± 0.20	0.20 ± 0.20	0.27 ± 0.12	0.80 ± 0.00	0.27 ± 0.23	0.00 ± 0.00	0.07 ± 0.0
3	GR00T N1.7 8	0.35 ± 0.00	0.93 ± 0.12	0.67 ± 0.12	0.00 ± 0.00	0.13 ± 0.12	1.00 ± 0.00	0.00 ± 0.00	0.07 ± 0.12	0.00 ± 0.0
4	π_0 5	0.30 ± 0.02	0.73 ± 0.12	0.67 ± 0.12	0.20 ± 0.20	0.00 ± 0.00	0.53 ± 0.06	0.22 ± 0.03	0.00 ± 0.00	0.07 ± 0.0
5	SmolVLA 3	0.27 ± 0.04	0.73 ± 0.12	0.33 ± 0.12	0.20 ± 0.20	0.13 ± 0.12	0.47 ± 0.12	0.00 ± 0.00	0.20 ± 0.00	0.07 ± 0.0
6	ACT 1	0.08 ± 0.01	0.40 ± 0.20	0.27 ± 0.12	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.0
7	DiT-D 2	0.08 ± 0.04	0.07 ± 0.12	0.33 ± 0.23	0.00 ± 0.00	0.00 ± 0.00	0.07 ± 0.12	0.07 ± 0.12	0.00 ± 0.00	0.07 ± 0.0
8	X-VLA 4	0.05 ± 0.04	0.33 ± 0.31	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.00	0.07 ± 0.12	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.0
9	DiT-F 2	0.04 ± 0.01	0.13 ± 0.12	0.07 ± 0.12	0.00 ± 0.00	0.07 ± 0.12	0.07 ± 0.12	0.00 ± 0.00	0.00 ± 0.00	0.00 ± 0.0

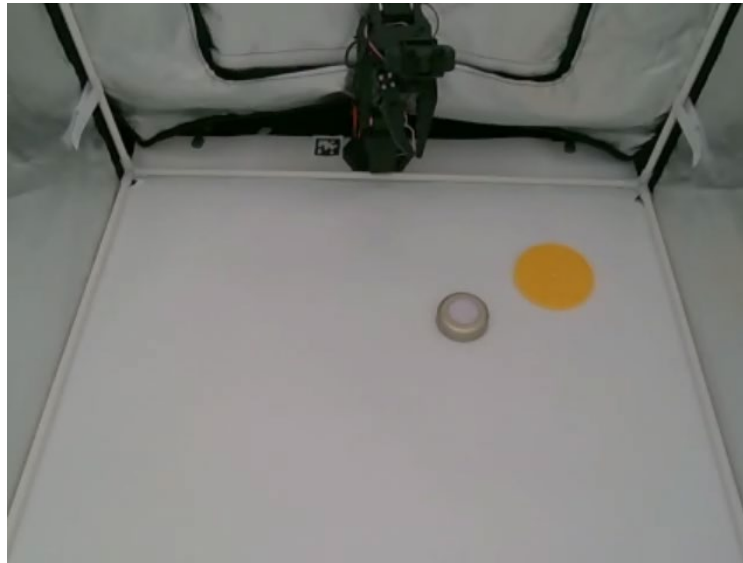
<https://irvlutd.github.io/VLAReplica/>

VLA-Replica Benchmark: Pi0.5 Behaviors



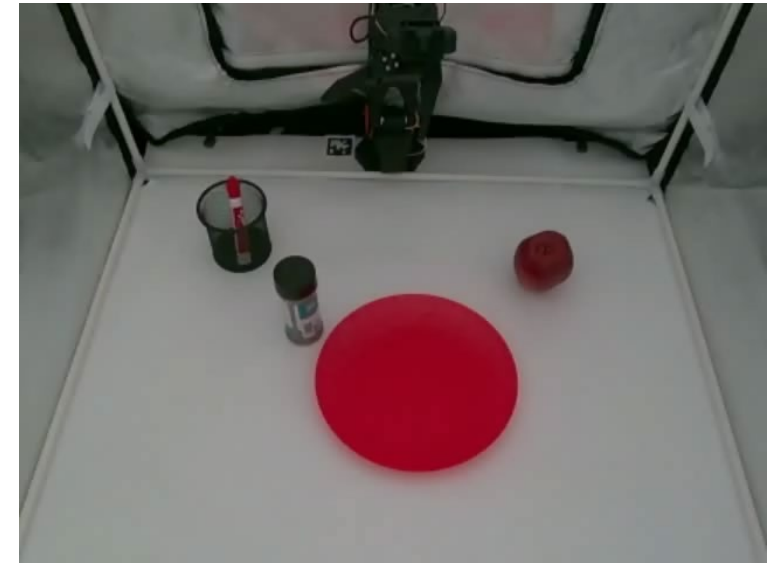
Stack the **yellow** block on the **blue** block

Precise manipulation
is challenging



Press the button **three** times

Memory task is challenging

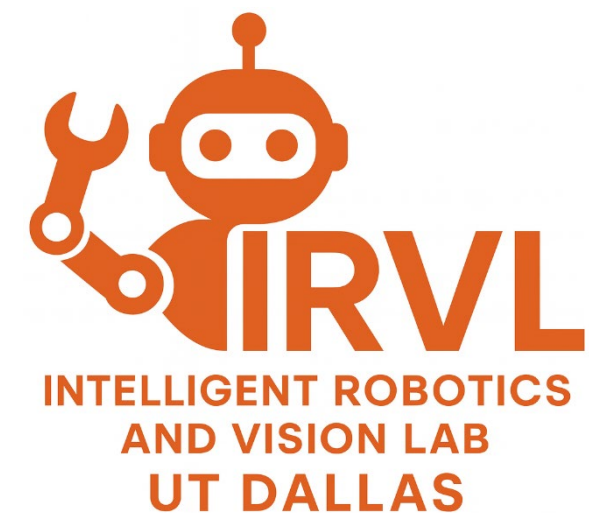


Pour **1** shake of pepper into the **red** plate

Failure recovery

Discussion: Integrating Components Together

- Modular pipelines
 - Perception-planning-control (ROS, ROS2)
 - Interpretability and debugging
 - Easier safety and constraint enforcement
 - Error accumulation
 - High latency usually, hard for closed-loop execution
- End-to-end manipulation
 - Training-testing (LeRobot)
 - Closed-loop manipulation
 - Failure recovery
 - Better for contact-rich dexterity
 - Poor interpretability and generalization
 - Hard to debug (More data? Better model?)



SONY



PENG



<https://labs.utdallas.edu/irvl/>

Assisted by
Mr. Garvin R. Solomon

Thank you!