

CSE 571: AI-Robotics

Time: 2026-02-19 10:00-11:20am

Location: CSE2 271

Deep Perception for Manipulation

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UNIVERSITY *of* WASHINGTON



Stanford
University

Myth and truth about robot perception

For the following statement, tell us whether you think it is true or false, **and why**.

2



Statement:

Robot Perception == Computer Vision

3

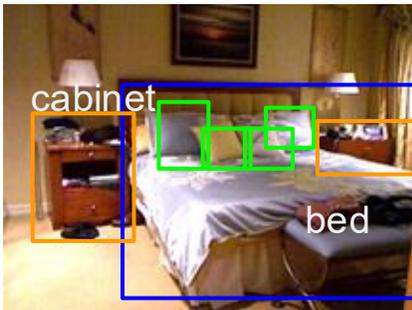


True or false, and why.

Statement:

Robot Perception == Computer Vision

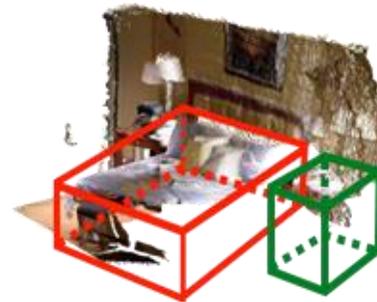
Not true. Robot need more information: 3D, physical properties ...



Object detection

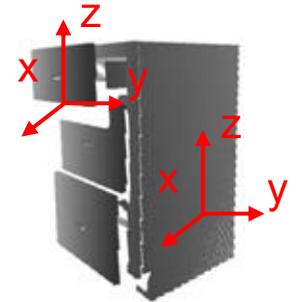


How far is the cabinet?



3D object detection

How to open the cabinet?

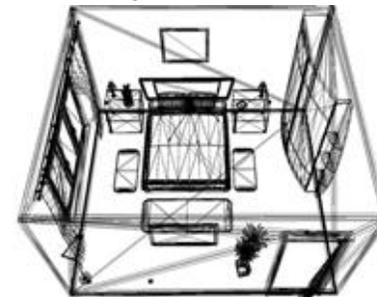


Object articulation



Image segmentation

How to get there?



3D reconstruction

How much force to use?



Friction, mass

Statement:

Robot perception is harder than computer vision?

5



True or false, and why.

Statement:

Robot perception is harder than computer vision?

Sometimes, yes.

Need to output more information (previous slide)

Sensitive to speed

Sensitive to error

Statement:

Robot perception is harder than computer vision?

However, sometimes, robot can use action to simplify perception



Statement:

Robot perception is harder than computer vision?

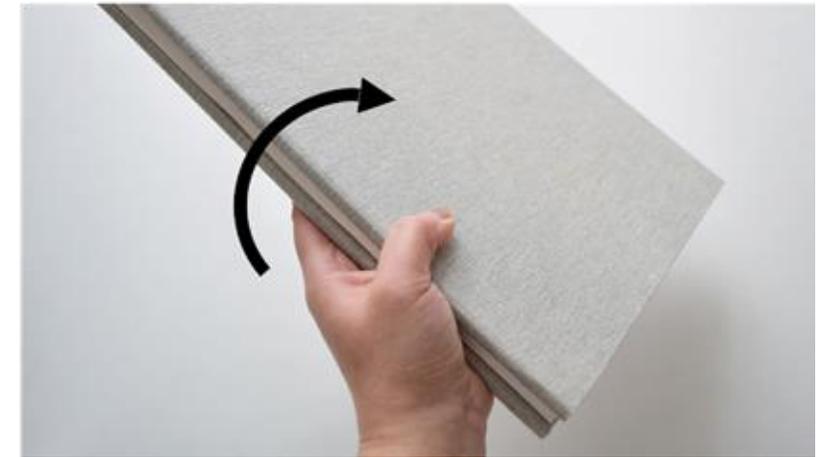
Not necessarily. Sometimes, robot can use action to simplify perception



Action Dipping
Information Temperature
Planing Swim



Pushing
 Weight
 Lift up the box



Flipping
 Title
 Read the book

Statement:

Perception is only about visual data

9

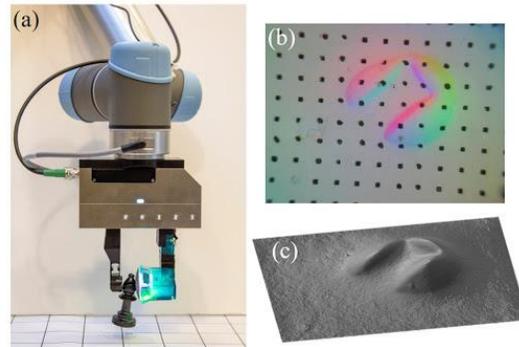


True or false, and why.

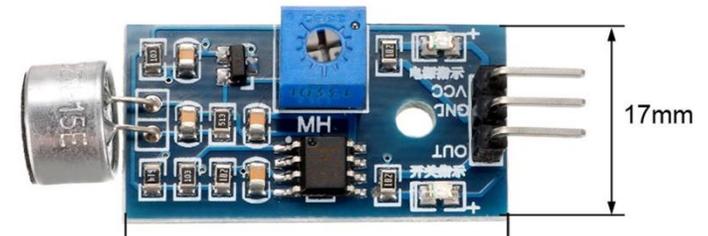
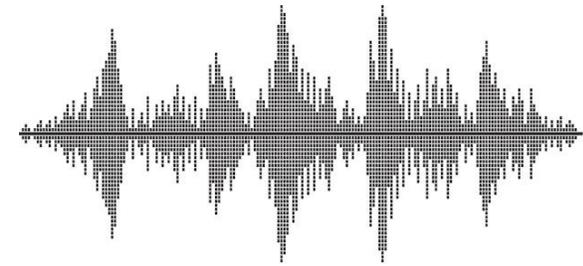
Statement:

Perception is only about visual data

Not true. There are many other sensory modalities a robot can use.



Tactile sensor



Acoustic sensor

Why we need different sensors?

Provide complimentary information – sense different signal

Provide redundant information – improve robustness

Help to develop a rich and coherent representation of the word

-- The association between different modality in our experience let us learned a share representation that let us infer other modality when they are absent



Statement:

Perception is a module in a robotics system

12



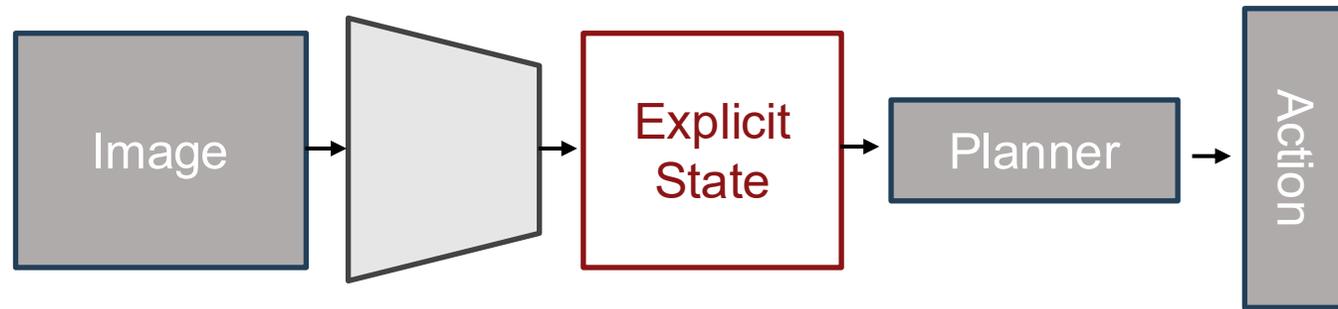
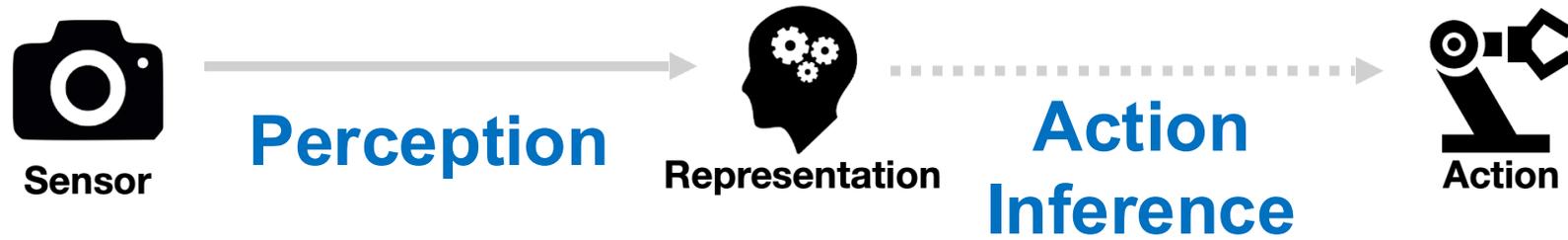
True or false, and why.

Statement:

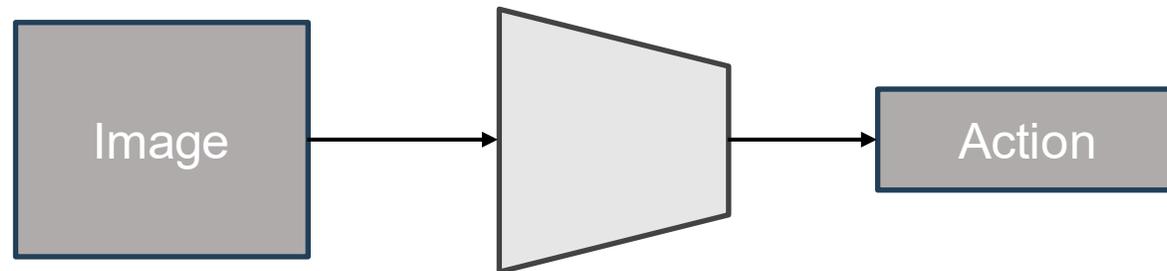
Perception is a module in a robotics system

Yes, and No

Role of perception: (different designs)



Modular design: Perception converts sensor input into a representation of the world that could be used for planning



End-to-end design: Visuomotor policy builds an implicit state representation to predict action

Statement:

Perception is independent of tasks/objectives

15

True or false, and why.



Statement:

Perception is independent of tasks/objectives

No

Objects tell us where
to carve the world

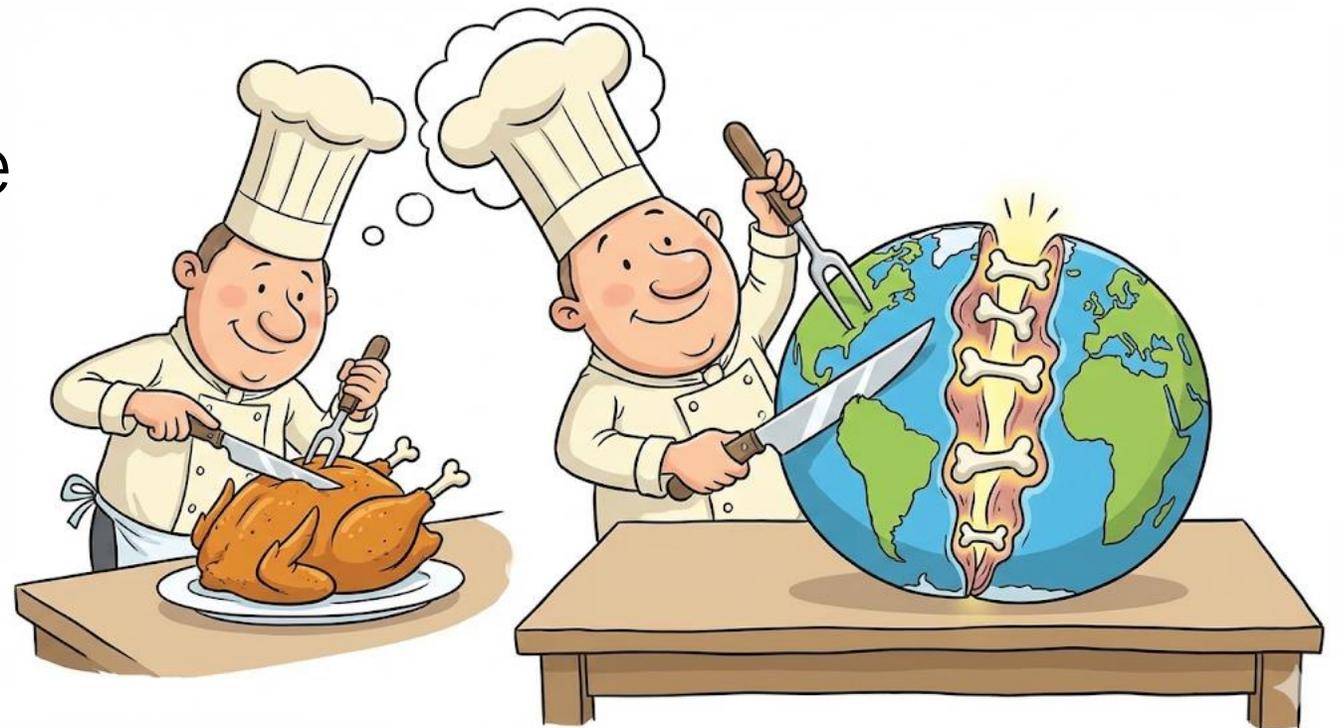


Image credit: Nano Banana Pro

What is an “object” here?



What is an “object” here?

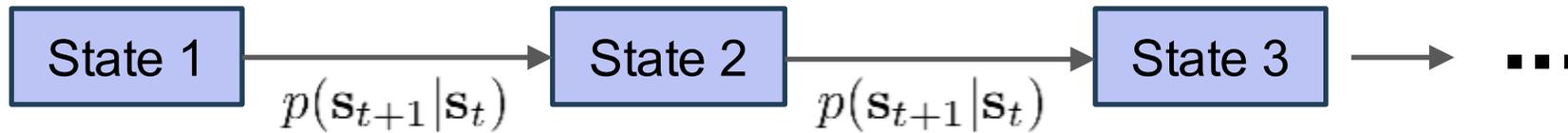


Markov Chain

The most basic probabilistic model of a dynamical system
 No actions yet, only states



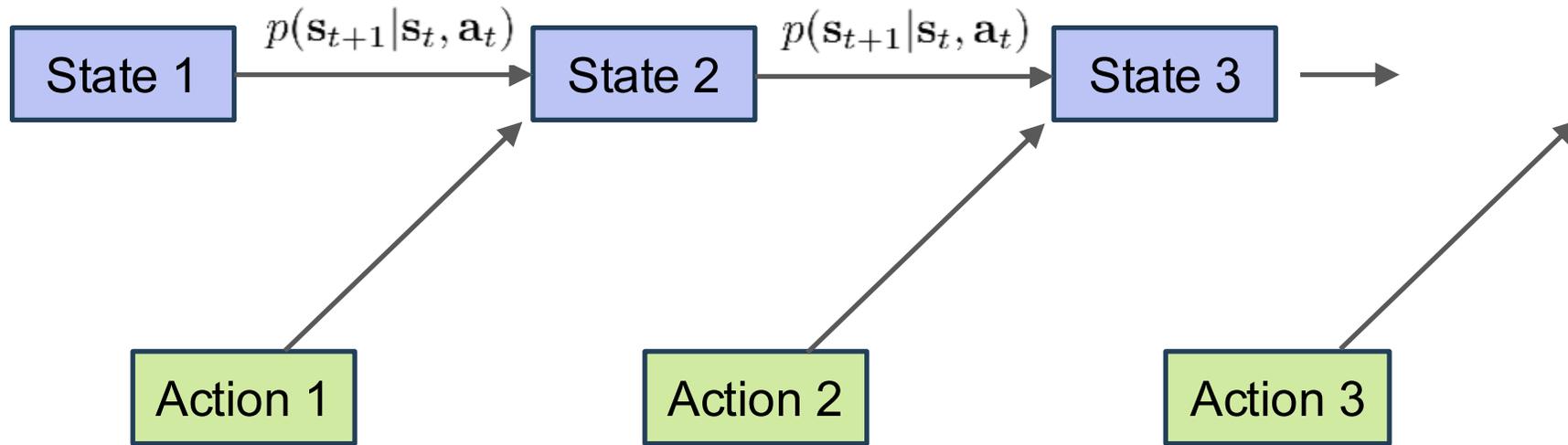
Andrey Markov



Markov property

$$\mathbf{s}_{t+1} \perp \mathbf{s}_{t-1} | \mathbf{s}_t$$

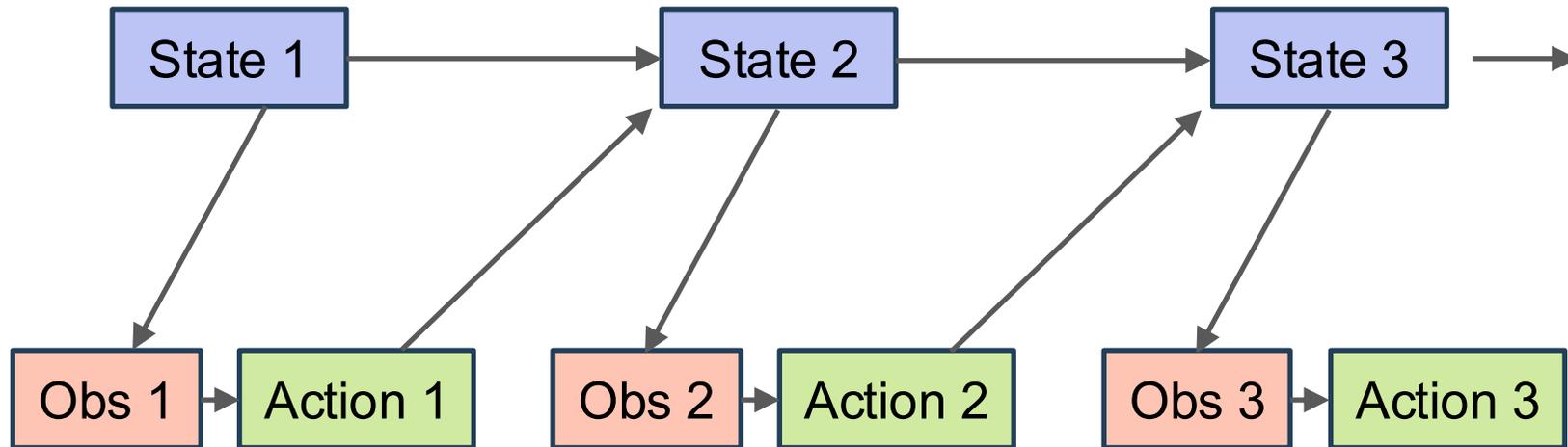
Markov Decision Process (MDP)



Richard Bellman

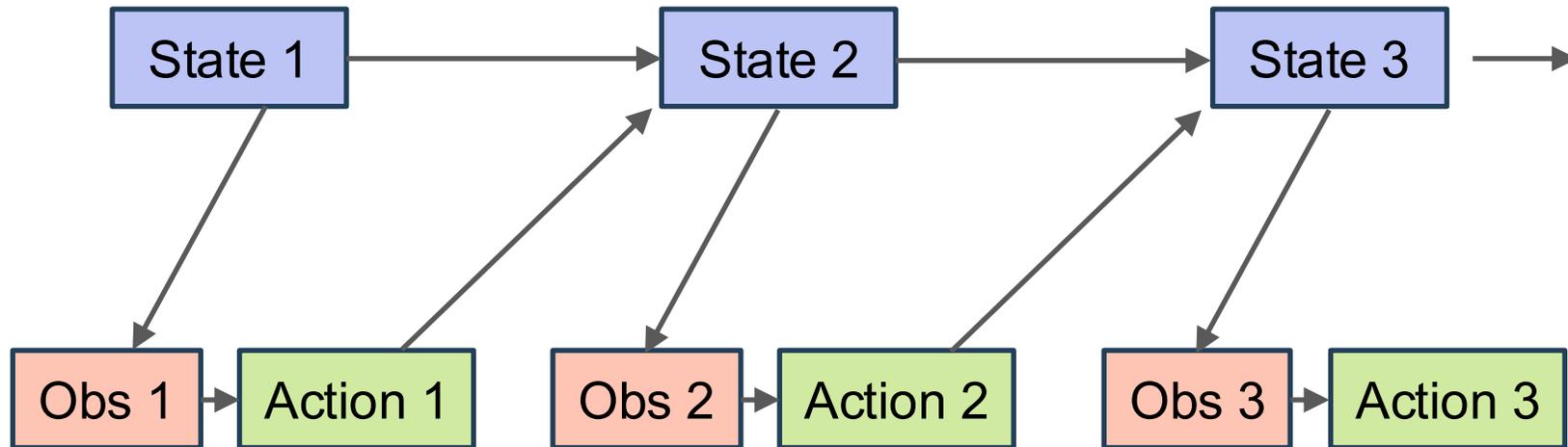
Probabilistic model for sequential decision making
 Also assumes a given reward function $\mathcal{S} \times \mathcal{A} \rightarrow \mathbb{R}$

Partially Observable Markov Decision Process (POMDP)



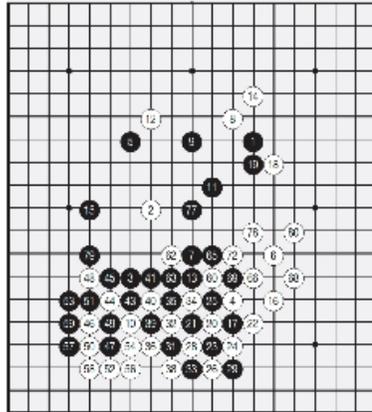
This is the regime real-world robots operate in, where we have access only to observations.

Partially Observable Markov Decision Process (POMDP)



What robot perception does: $\Pr(s_t = s \mid o_{1:t}, a_{1:t-1})$.

What is a state? What is a representation?



3^{361} states?

Game of Go

- an exponentially large number of states?
- infeasible to enumerate, memorize, or search



$256^{3 \times 500 \times 500}$?

Images

Image space has exponentially more states than Go.

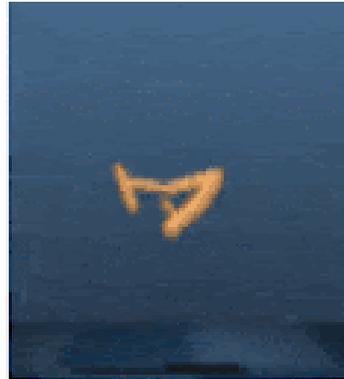
What is a state? What is a representation?



Sparse Cartpole



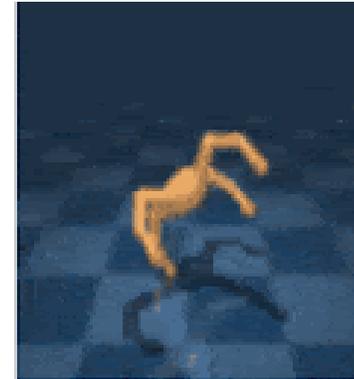
Acrobot Swingup



Hopper Hop



Walker Run



Quadruped Run

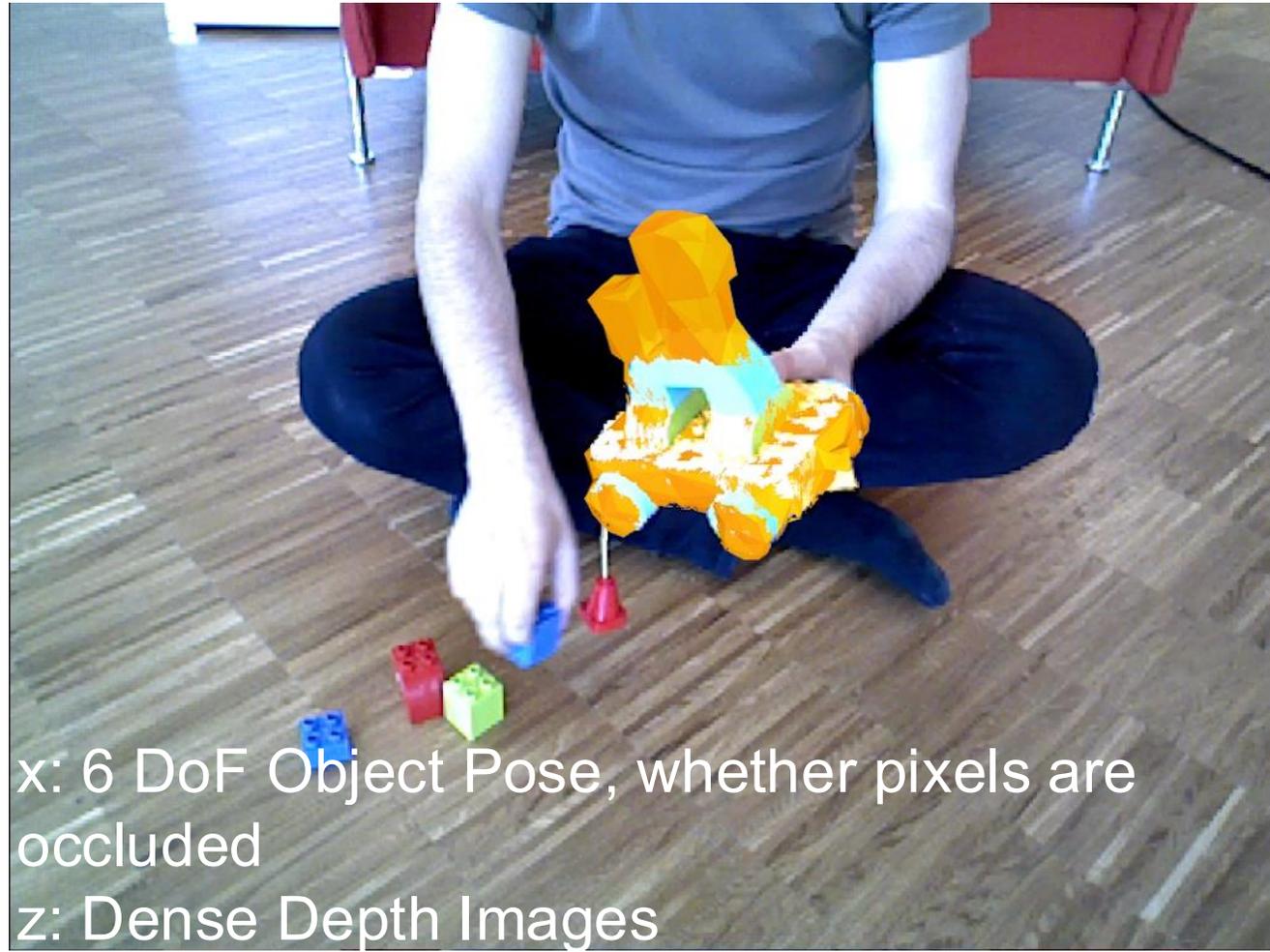
DeepMind Control Suite. Tassa et al. 2018

Representations for Autonomous Driving



Image adapted from NuScenes by Motional. [nuscenes.org](https://www.nuscenes.org)

Representations for Manipulation



Manuel Wüthrich et al. "Probabilistic Object Tracking using a Depth Camera", IROS 2013

What is a state? What is a representation?



Pose?



Meaning in English

“the way that someone or something is shown or described:”

“a sign, picture, model, etc. of something”

- Cambridge Dictionary

Representations in Cognitive Science

Symbolic View

“[...] a hypothetical internal cognitive symbol that represents external reality” (Morgan '14)

“[...] a formal system for making explicit certain entities or types of information [...]” (Marr '10)

“[...] intermediaries between the observing subject and the objects, processes or other entities observed in the external world. These intermediaries [...] represent to the mind the objects of that world.” (Wikipedia - Mental Representations - Representationalism)

Embodied View

“... actions are directly triggered by stimuli in the environment without the need for internal representations” (Gibson '66, Zech '19 on Embodied Cognition)

“... actions are represented by their anticipated effect, that is, action representations essentially entail a mental model of a needed future environmental state” (Jeannerod '06, Zech '19)

Representations in Machine Learning

“Features”, “A good representation is also one that is useful as input to a supervised predictor.” (Bengio '14)

“create a representation of the data to provide the model with a useful vantage point into the data's key qualities. [...] to train a model, you must choose the set of features that best represent the data.” (Google Crash Course of ML Concepts)

“ [...] world models, internal models of how the world works.”; “(1) estimate missing information about the state of the world not provided by perception, (2) predict plausible future states of the world.” (YLC '22)

Requirements for Good Representations

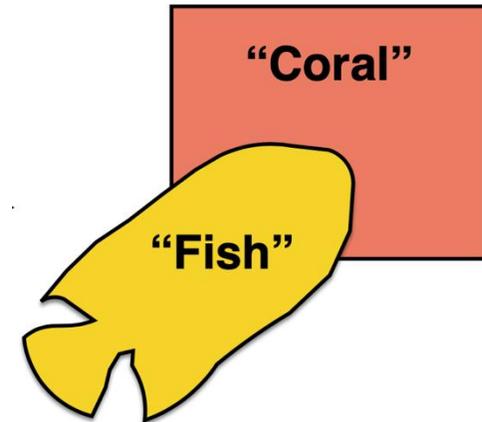
“Solving a problem simply means representing it so as to make the solution transparent.”

Herbert A. Simon, *Sciences of the Artificial*



Requirements for Good Representations

- Compact (minimal)
- Explanatory (sufficient)
- Disentangled (independent factors)
- Hierarchical (feature reuse)
- Generalizes over many tasks



[See “Representation Learning”, Bengio 2013, for more commentary]

Case Studies of State Representations in Manipulation

- Simulator State (and Real2Sim2Real)
- Object-Centric Descriptors
- Affordances
- Keypoints
- Latent
- Hierarchical/Hybrid

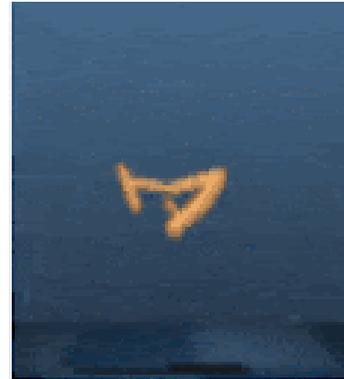
Simulator State



Sparse Cartpole



Acrobot Swingup



Hopper Hop



Walker Run



Quadruped Run

DeepMind Control Suite. Tassa et al. 2018

Simulator State



URDF

What:

Unified Robot Description Format Kinematic and basic physics description of a robot

How:

- XML format
- Tags: link or joint
- Kinematic tree structure
- Order in the file does not matter



Real2Sim2Real with Simulator State

3D scene reconstruction



Real-to-sim transfer of policies

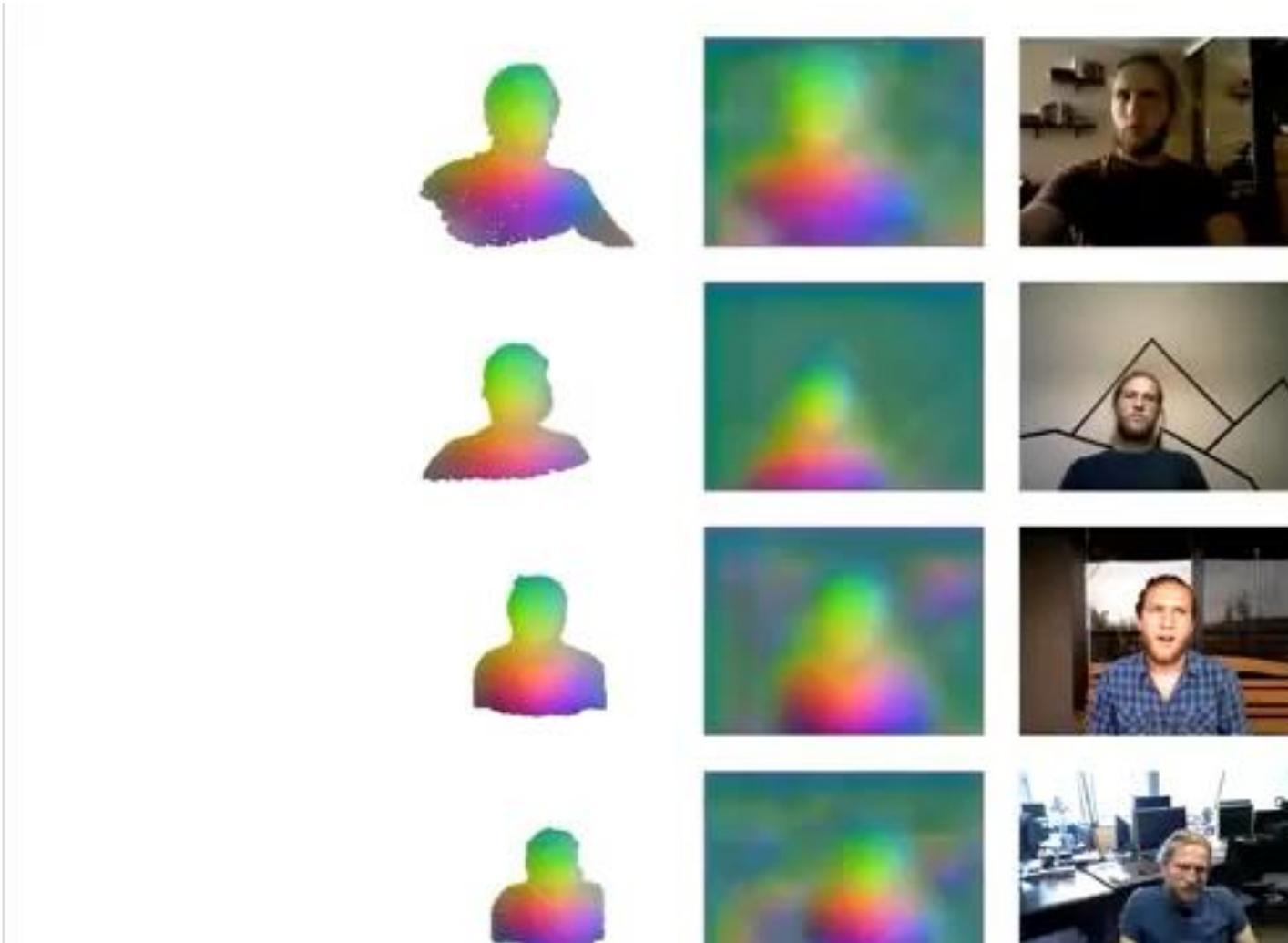
RL fine-tuning in sim



Robust policy in the real world



Object-Centric Descriptors



Schmidt et al., ICRA 2017

Object-Centric Descriptors

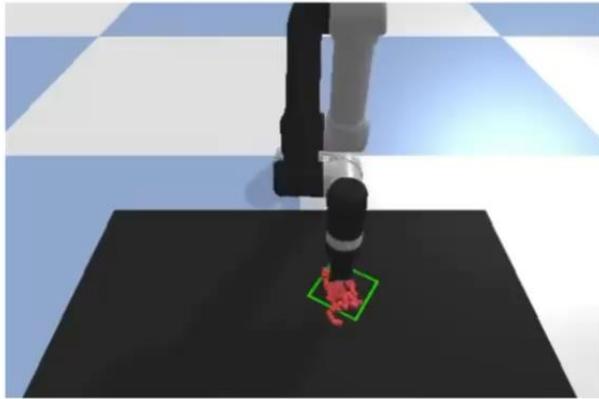
Dense Object Nets: Learning Dense Visual
Object Descriptors By and For Robotic Manipulation

Peter R. Florence*, Lucas Manuelli*, Russ Tedrake

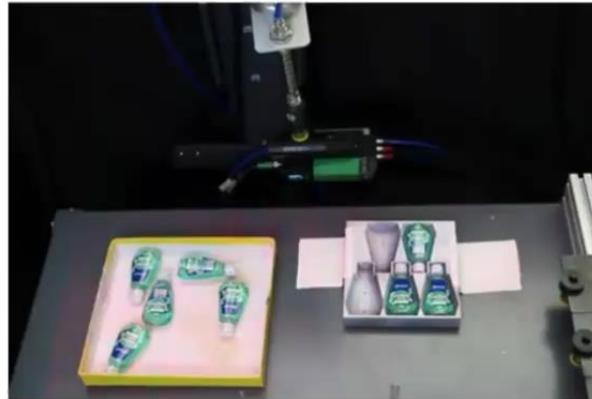
CSAIL (Computer Science and Artificial Intelligence Laboratory)
Massachusetts Institute of Technology

Florence et al., RA-L 2020 (Best Paper Award)

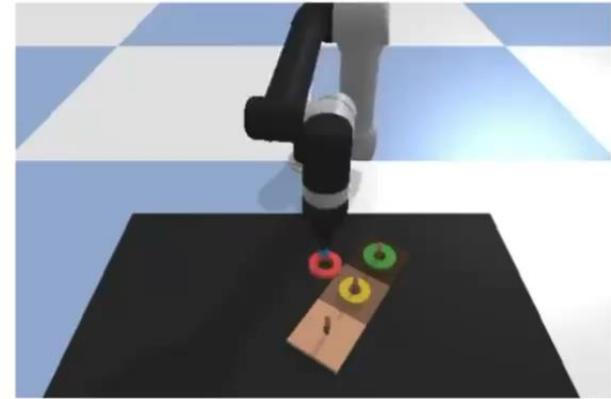
Affordances



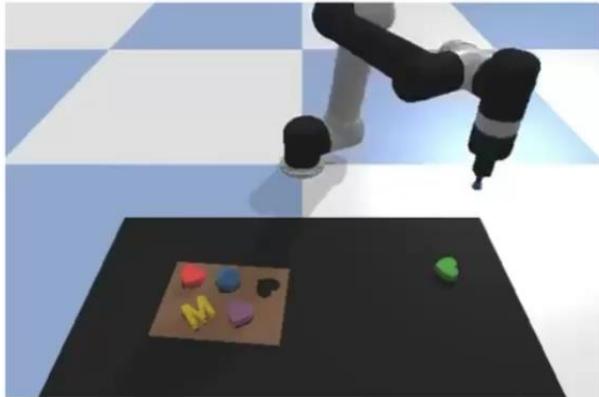
Pushing piles with closed-loop feedback



Production pick-and-place w/o 3D models



Multi-step sequential tasks



Pick-and-place with unseen objects



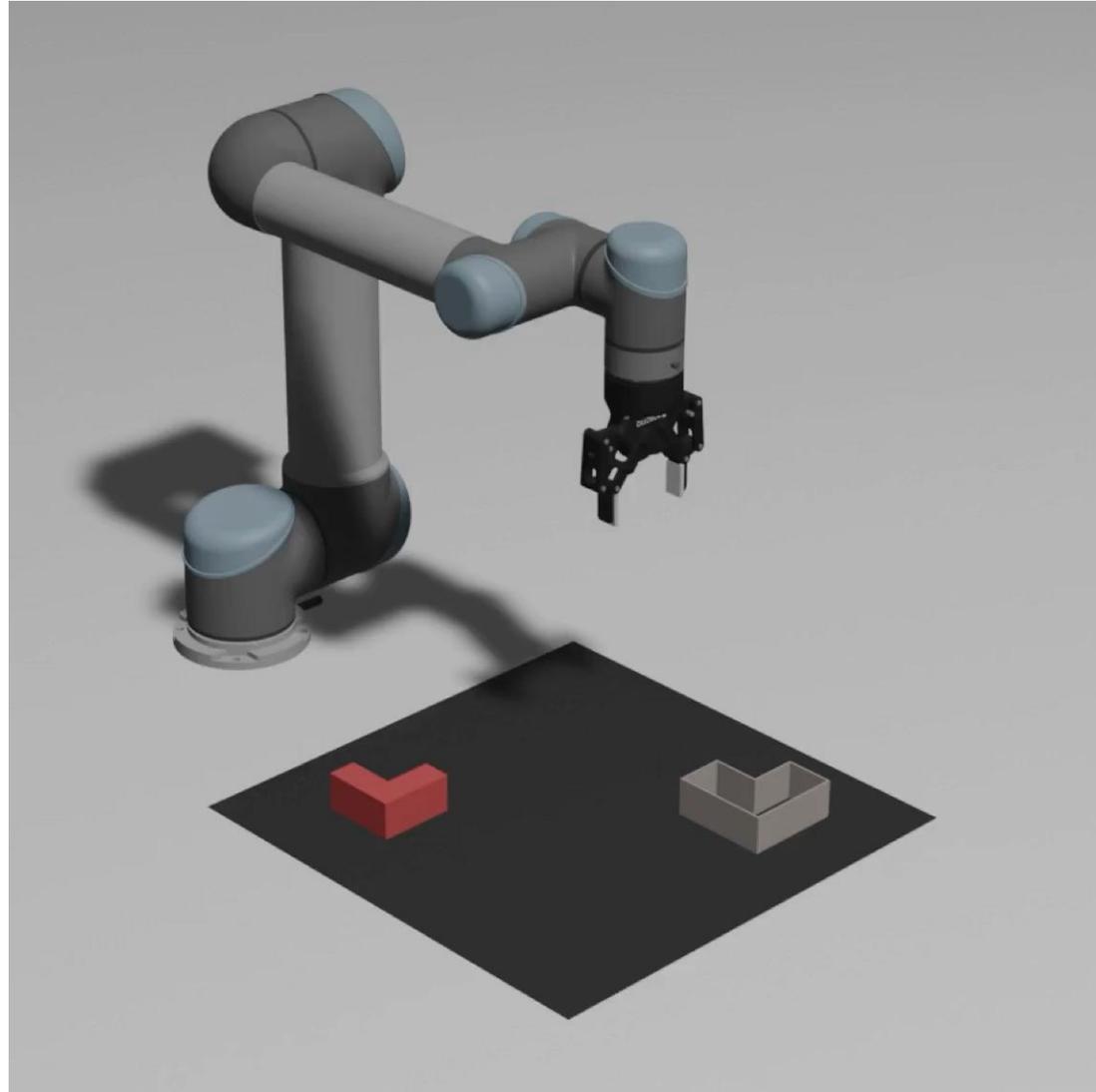
Learning to push piles on real robots



Pick-conditioned placing from 10 examples

Zeng et al., CoRL 2020 (Best Paper Presentation Award Finalist)

Affordances



Zeng et al., CoRL 2020 (Best Paper Presentation Award Finalist)

Affordances



Tang et al., ICRA 2025 (Best Paper Award Finalist)

Keypoints

kPAM: KeyPoint Affordances for Category-Level Robotic Manipulation

Lucas Manuelli*, Wei Gao*, Peter Florence, Russ Tedrake

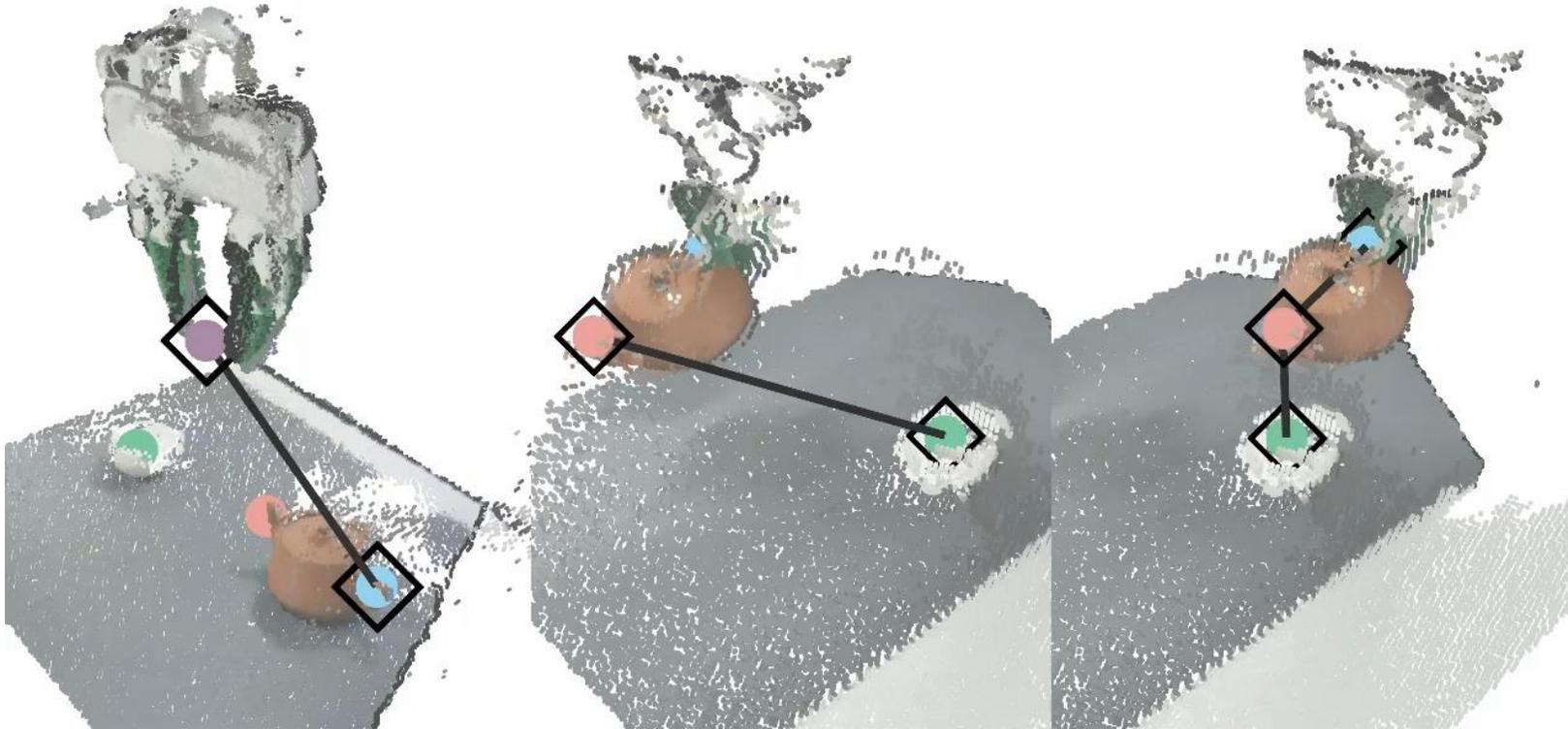
CSAIL (Computer Science and Artificial Intelligence Laboratory)
Massachusetts Institute of Technology

Manuelli et al., ISRR 2019

Keypoints

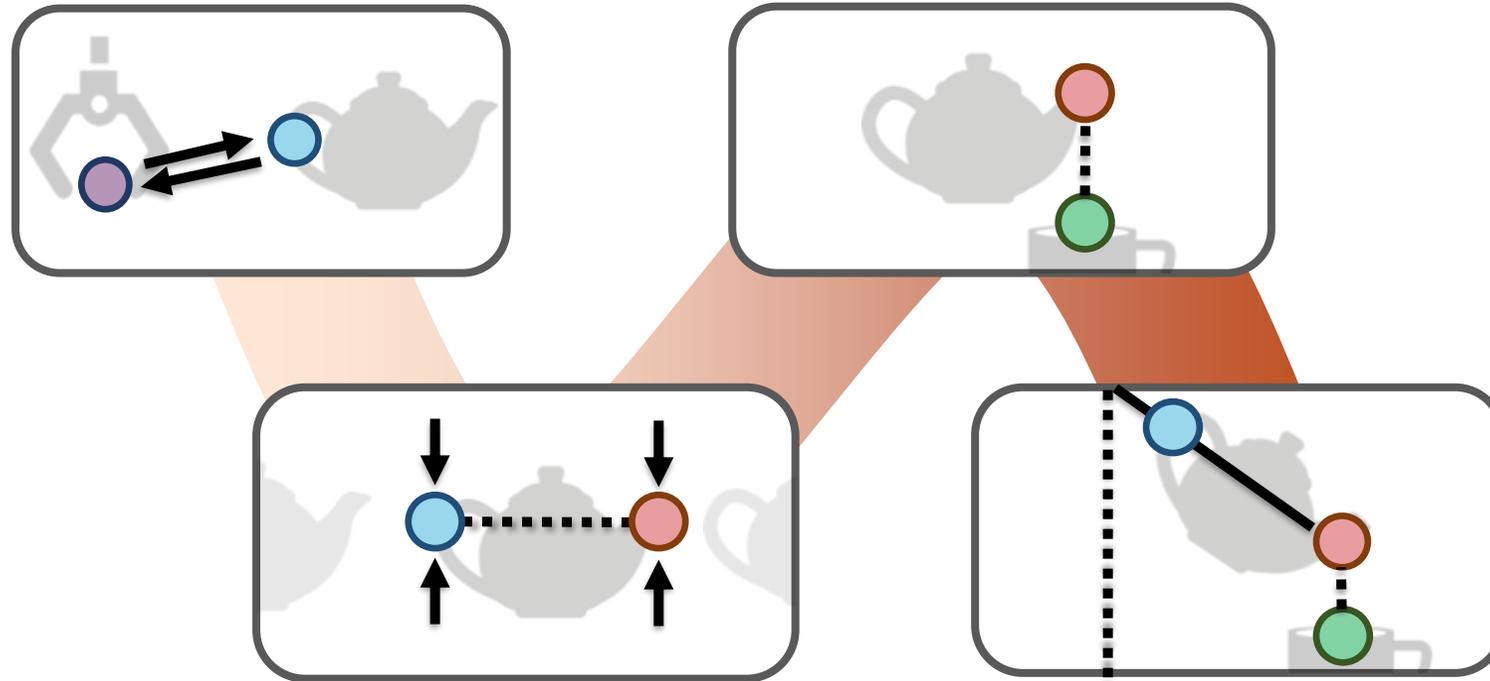
`cost = norm(...)`

`cost = arccos(...)`



Huang et al., CoRL 2024 (Best Paper Award at CoRL LEAP)

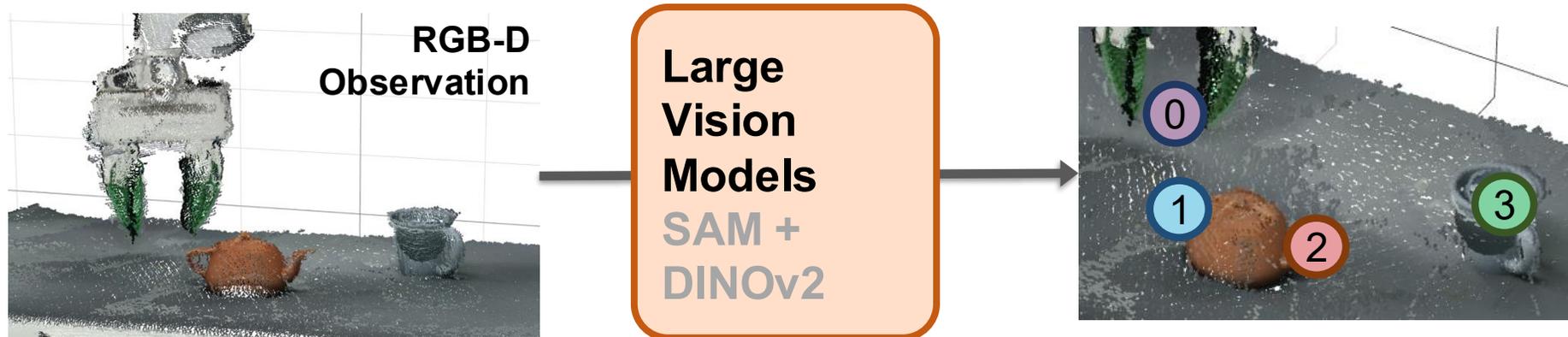
Keypoints



Huang et al., CoRL 2024 (Best Paper Award at CoRL LEAP)

Keypoints

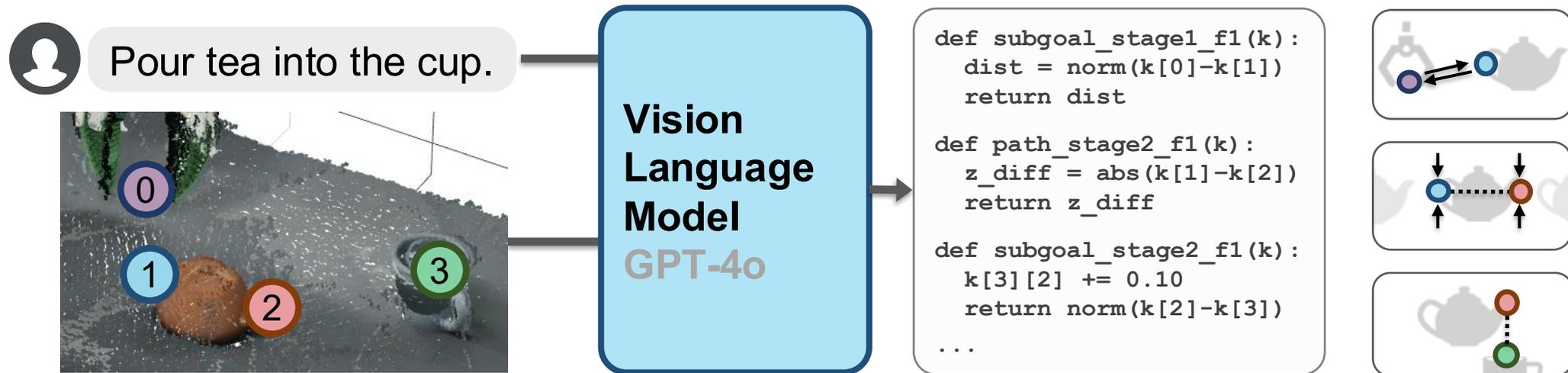
- **Step 1:** Obtain a set of semantically meaningful keypoints in the scene



Huang et al., CoRL 2024 (Best Paper Award at CoRL LEAP)

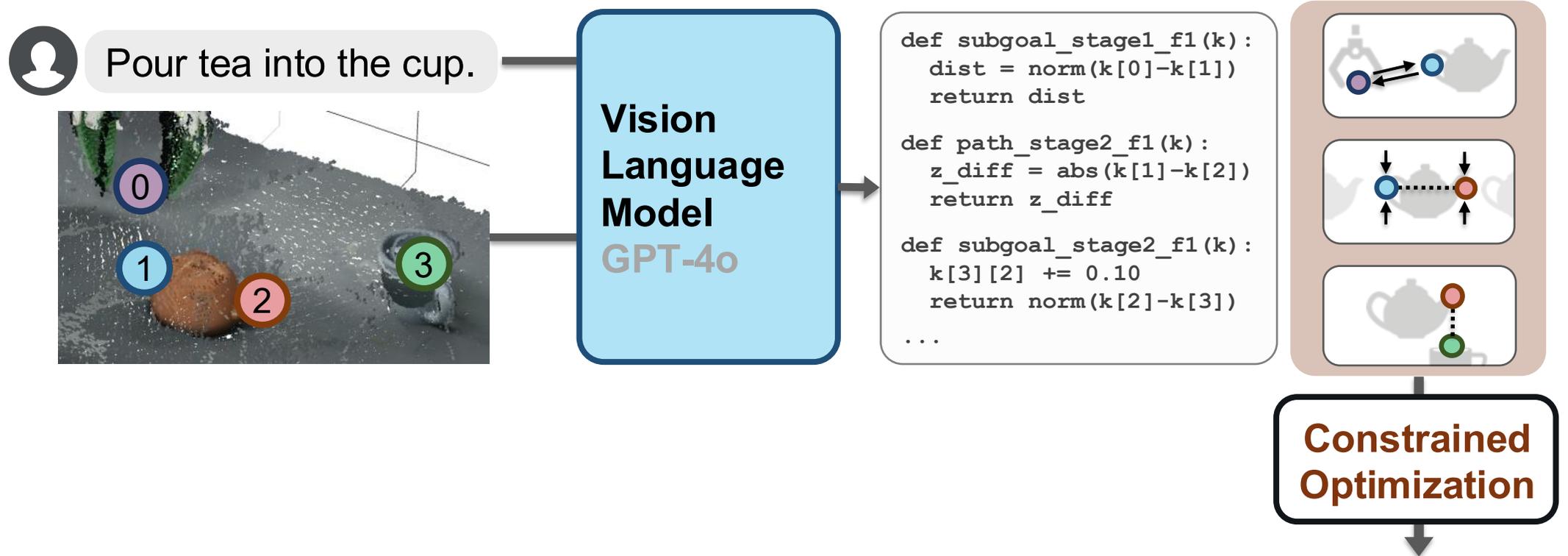
Keypoints

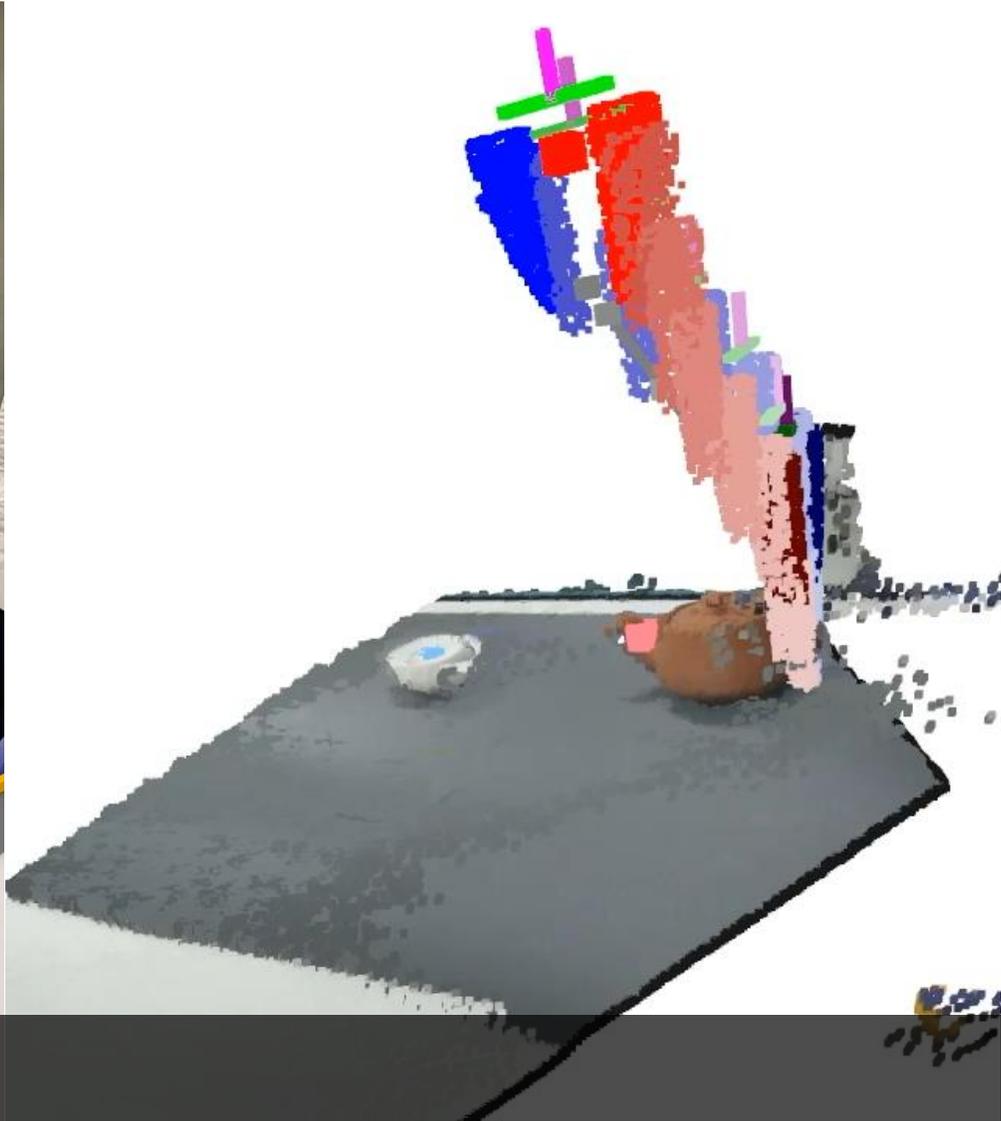
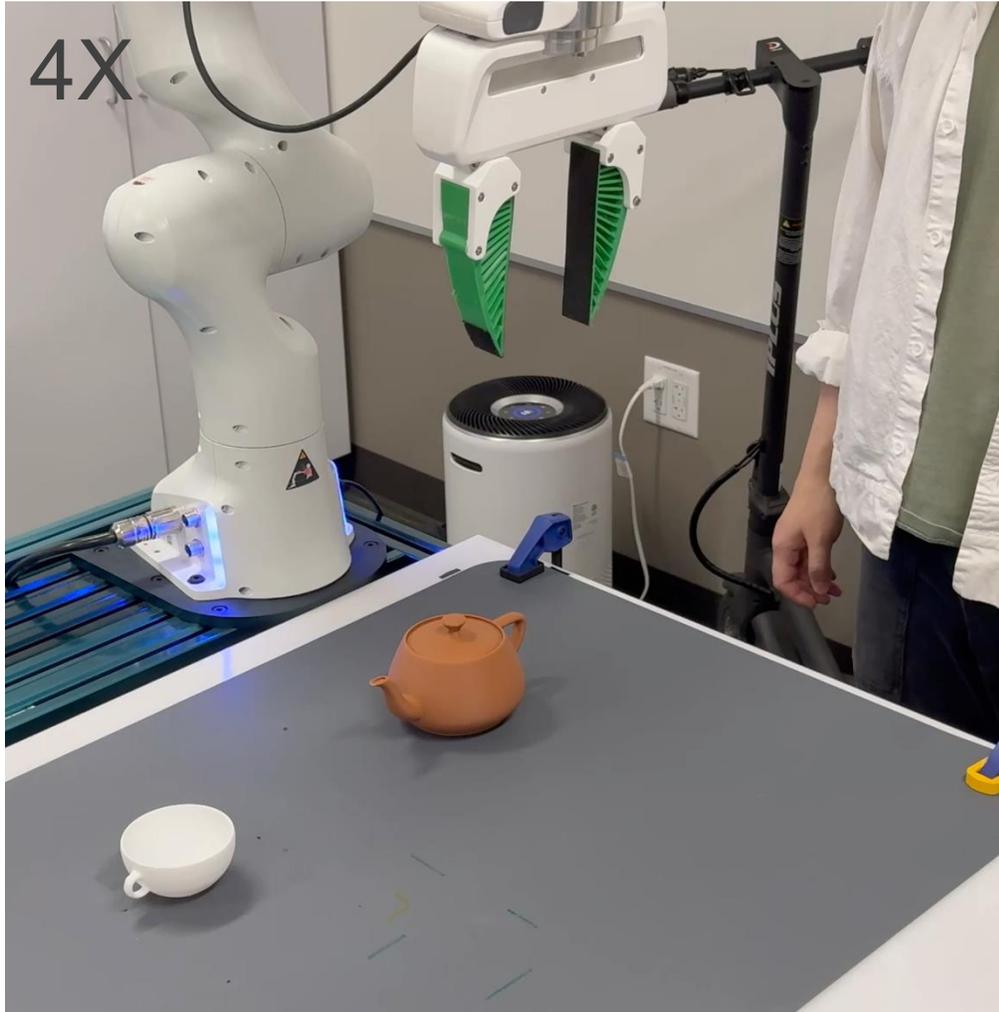
- **Step 1:** Obtain a set of semantically meaningful keypoints in the scene
- **Step 2:** Visually prompt VLM to write keypoint-based constraint code.



Keypoints

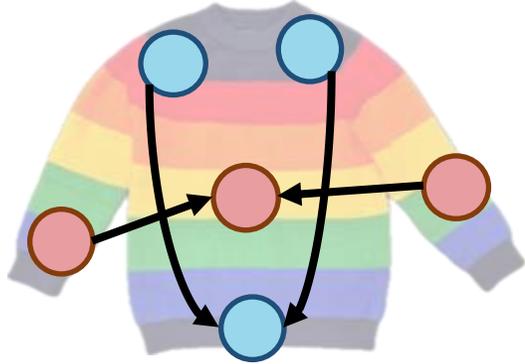
- **Step 1:** Obtain a set of semantically meaningful keypoints in the scene
- **Step 2:** Visually prompt VLM to write keypoint-based constraint code.
- **Step 3:** Perform constrained optimization to obtain robot actions.



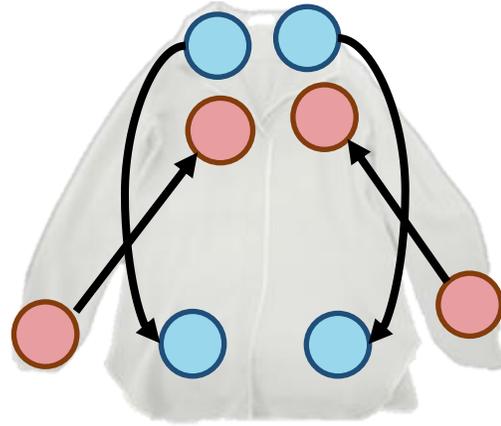


Closed-Loop Replanning at 10 Hz

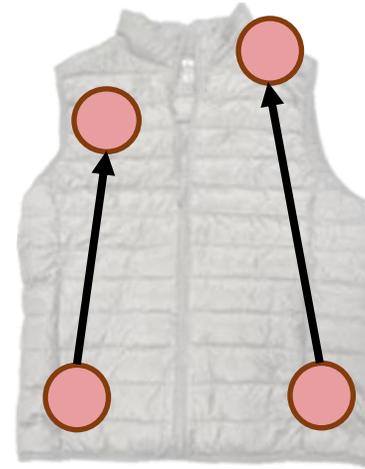
sweater



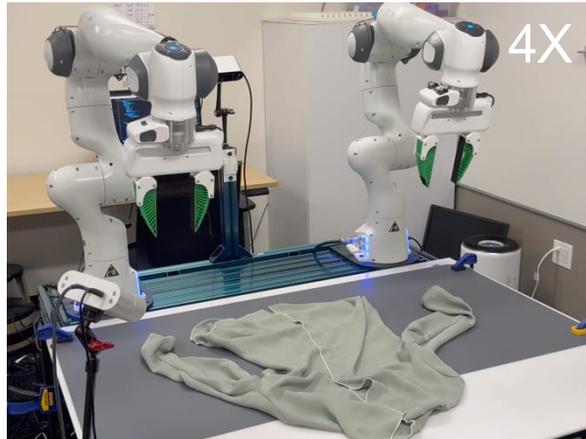
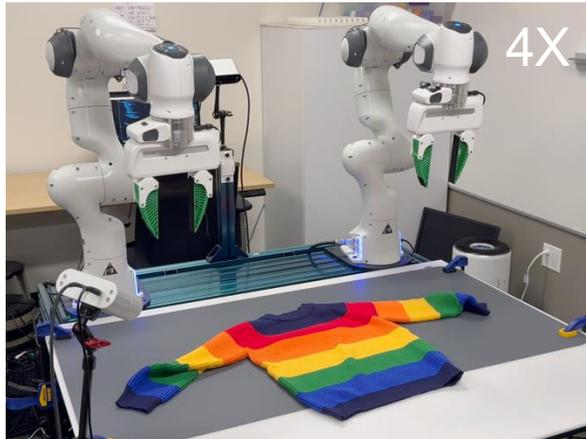
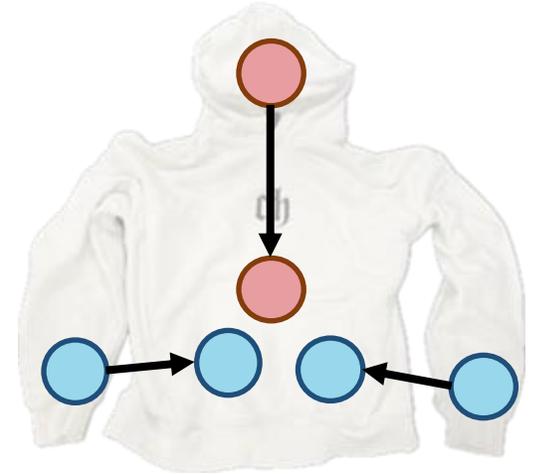
shirt



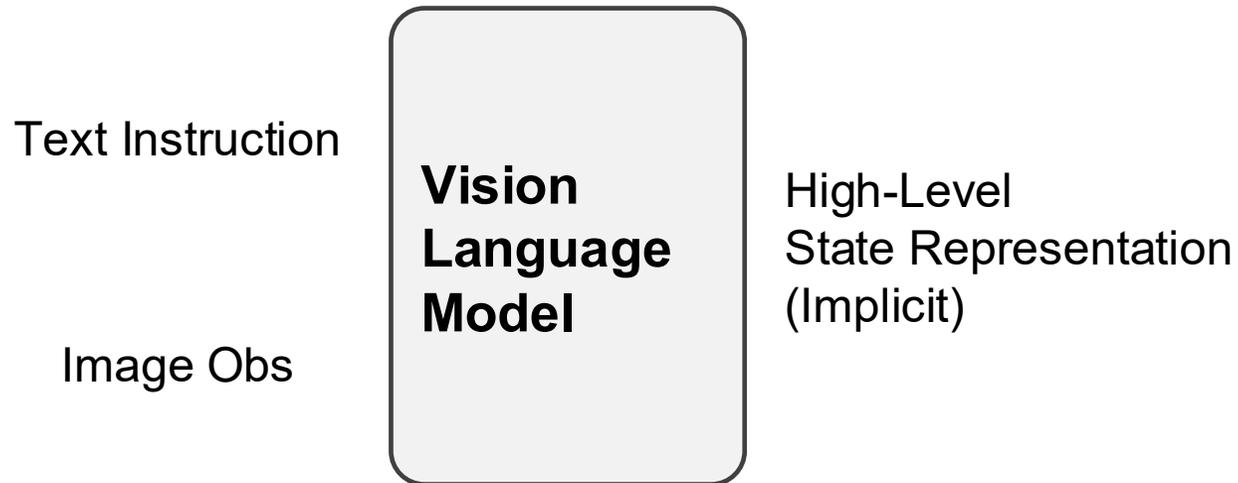
vest



hoodie



Hierarchical/Hybrid State Representation



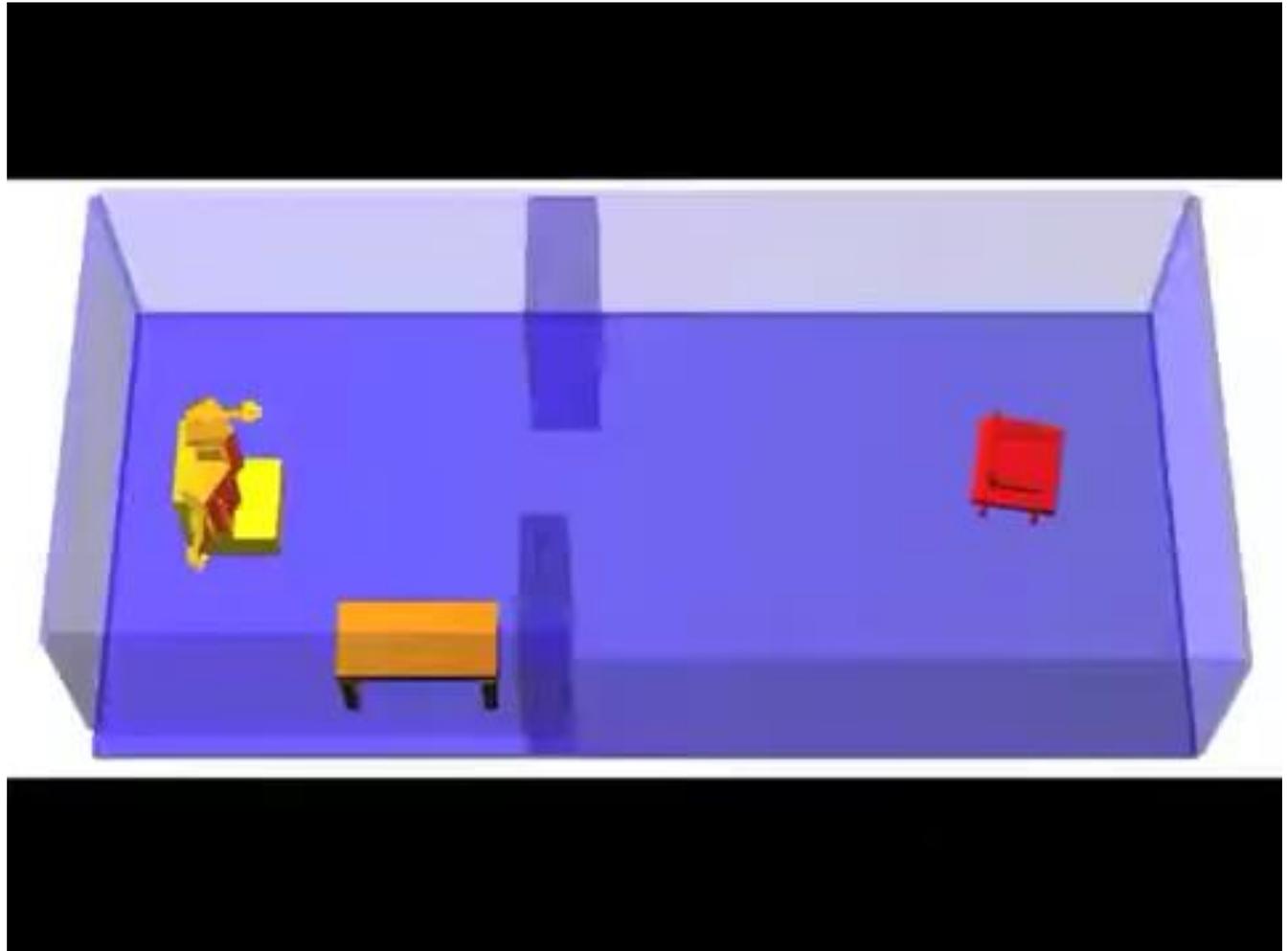
Low-Level State Representation (Keypoints + Point Cloud)

Hierarchical/Hybrid State Representation

Task and Motion Planning

Discrete state: predicates
(e.g., “grasped” or “not grasped”)

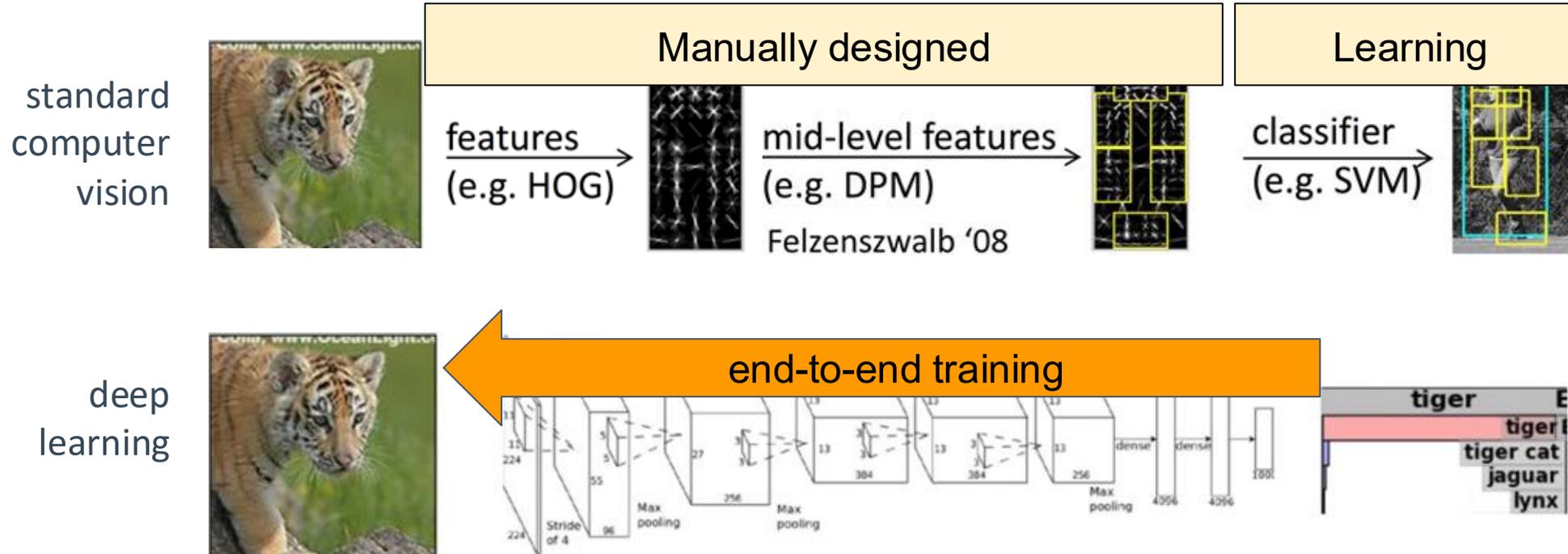
Continuous state: collision
geometry of the robot and the
scene



Leviñ et al., IROS 2013

Latent State Representation

Why Latent State Representation?

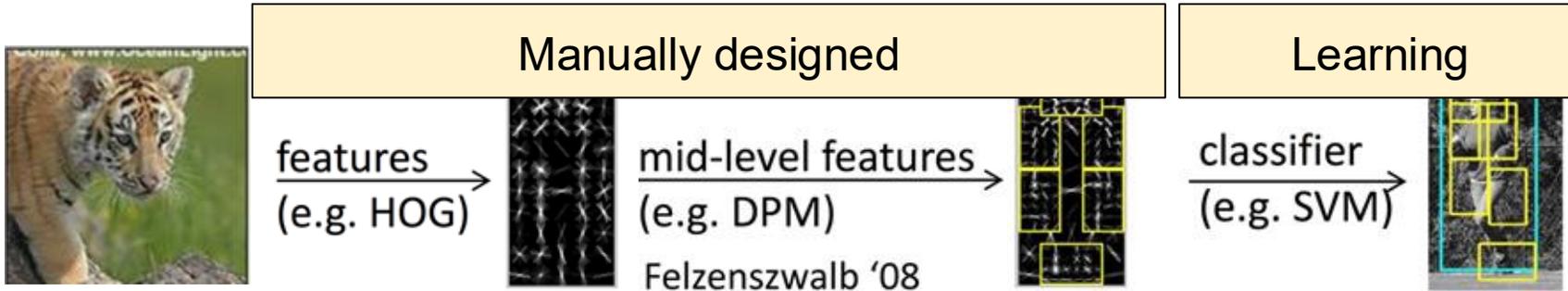


Benefit:

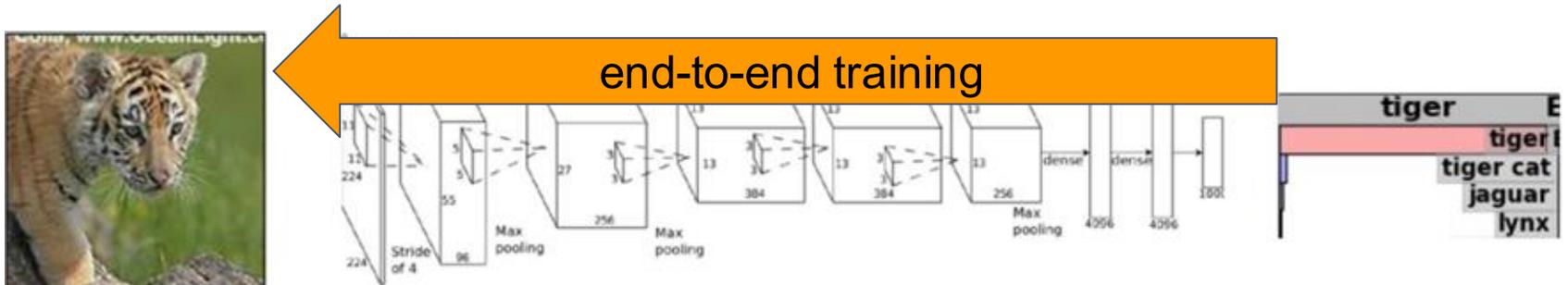
- No manual feature engineering
- Optimize the features for the target task (diff. task requires diff. representations)

Latent State Representation in Policies

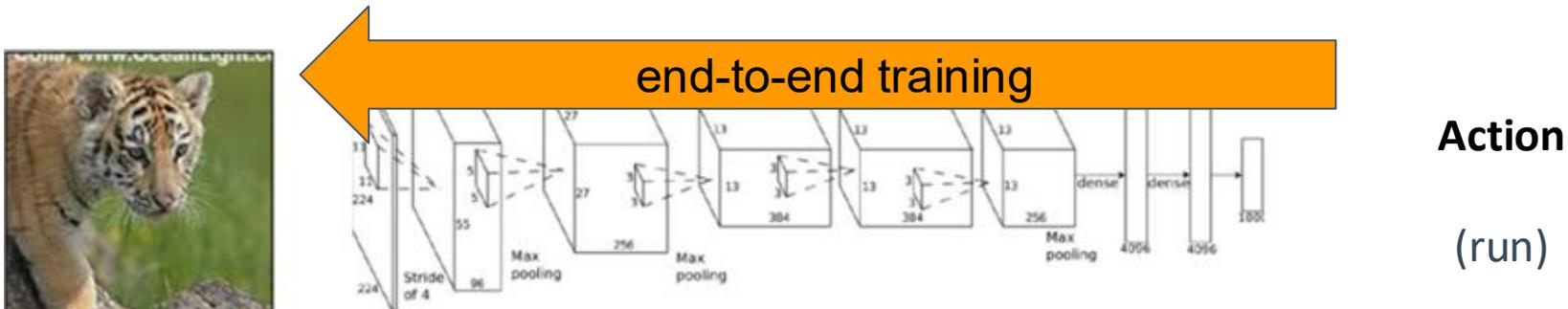
standard computer vision



deep learning for computer vision



deep learning for robots

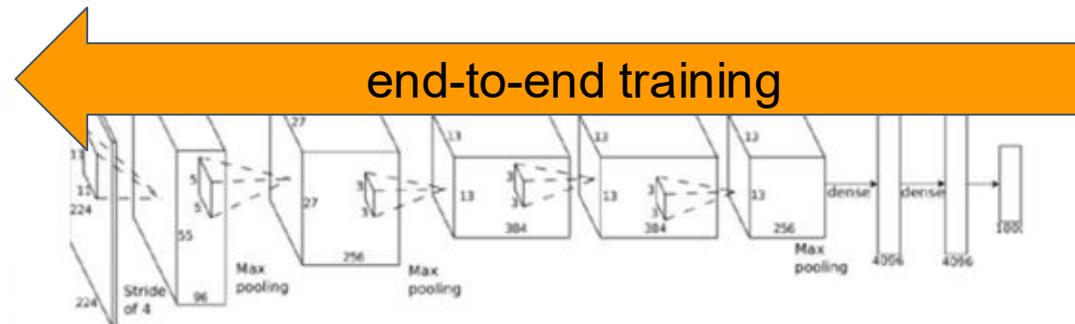


Latent State Representation in Policies

Benefit (hopefully ...):

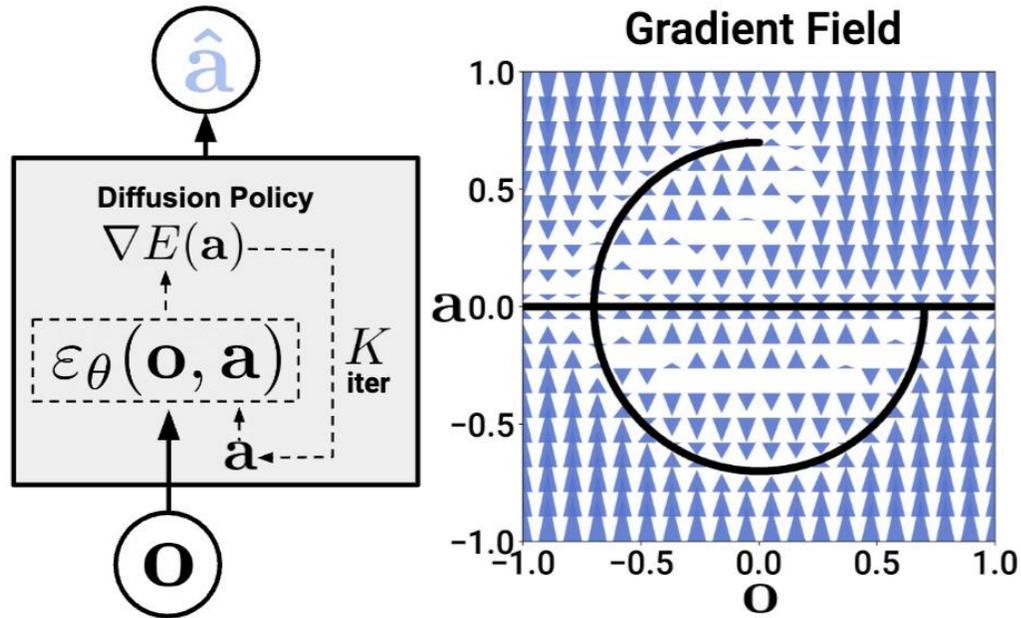
- Less manual engineering on the intermediate representation between perception and planning (Object pose? Grasp pose? Object shape?)
- Optimize representation for the target task (diff. task requires diff. representation)
- More robust to perception error

deep learning for robots

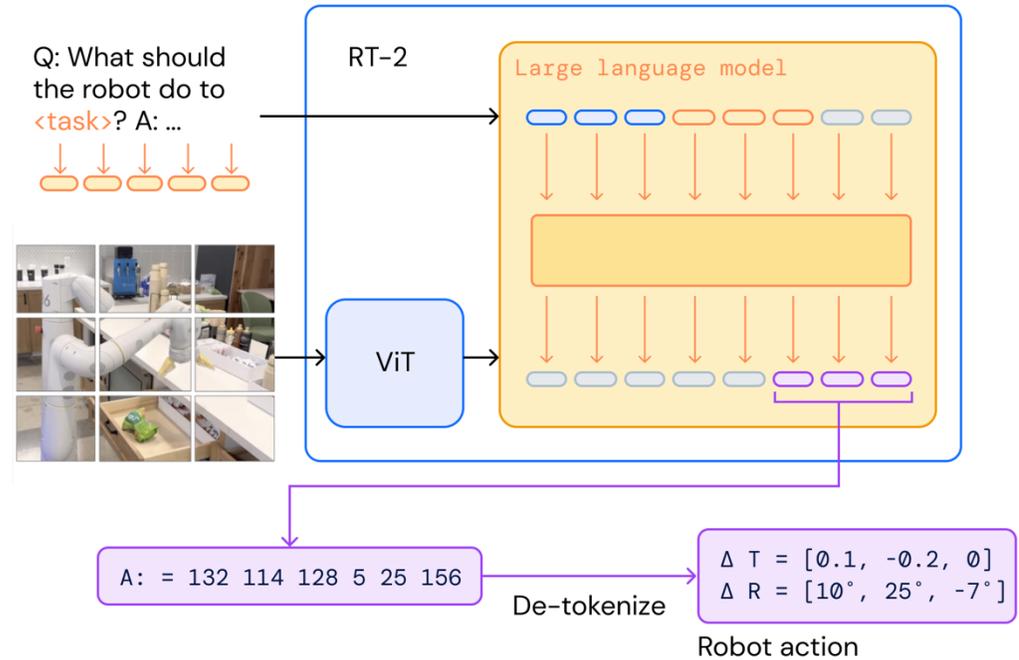


Action
(run)

Latent State Representation in Policies



Diffusion Policy
[Chi et al., RSS 2023]



VLA Models
[Brohan et al., CoRL 2023]

Latent State Representation in Policies



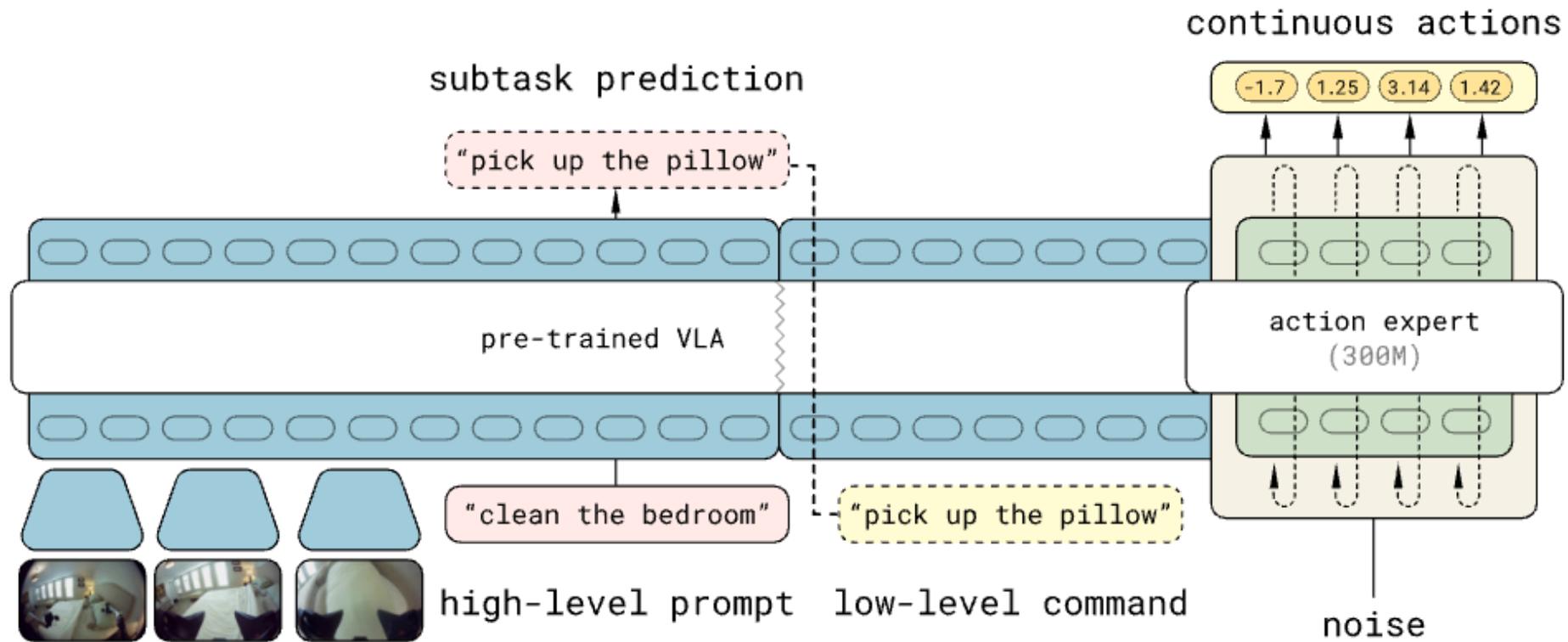
$\pi_{0.5}$
 [Physical Intelligence,
 2025]

Research Frontier

- Perceptual representations for end-to-end policies
- Multimodal perception (Perception beyond vision)
 - Touch, Audio, etc
- Interactive and active perception
 - “We see in order to move; we move in order to see.” – William Gibson
- Perception for world modeling
 - Beyond “perceive to act”, also answer the “what if” queries based on actions

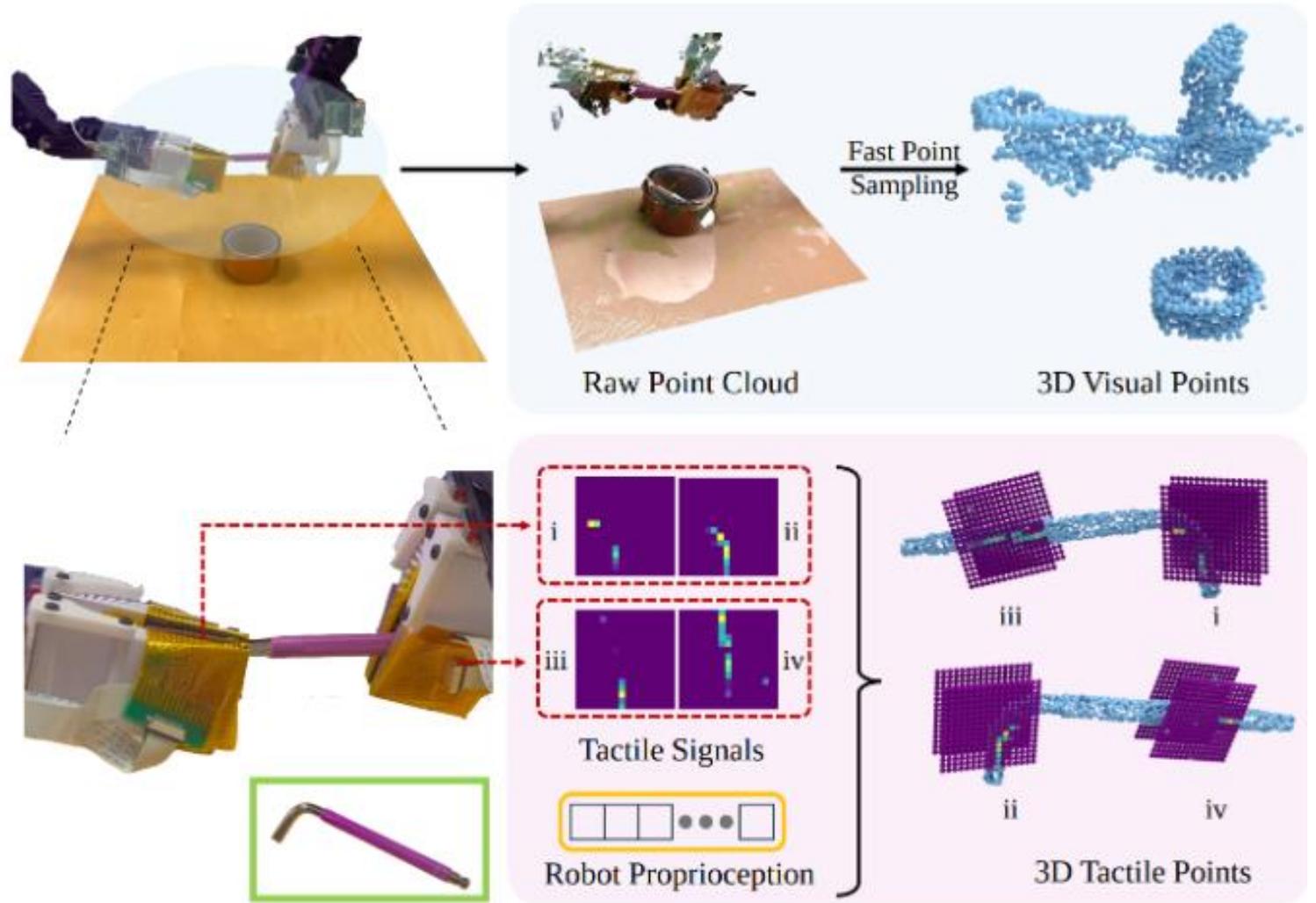
Perceptual Representations for End-to-End Policies

- What should be the pre-trained representation?
- What properties should it have?
- Is vision-language encoder the best choice?



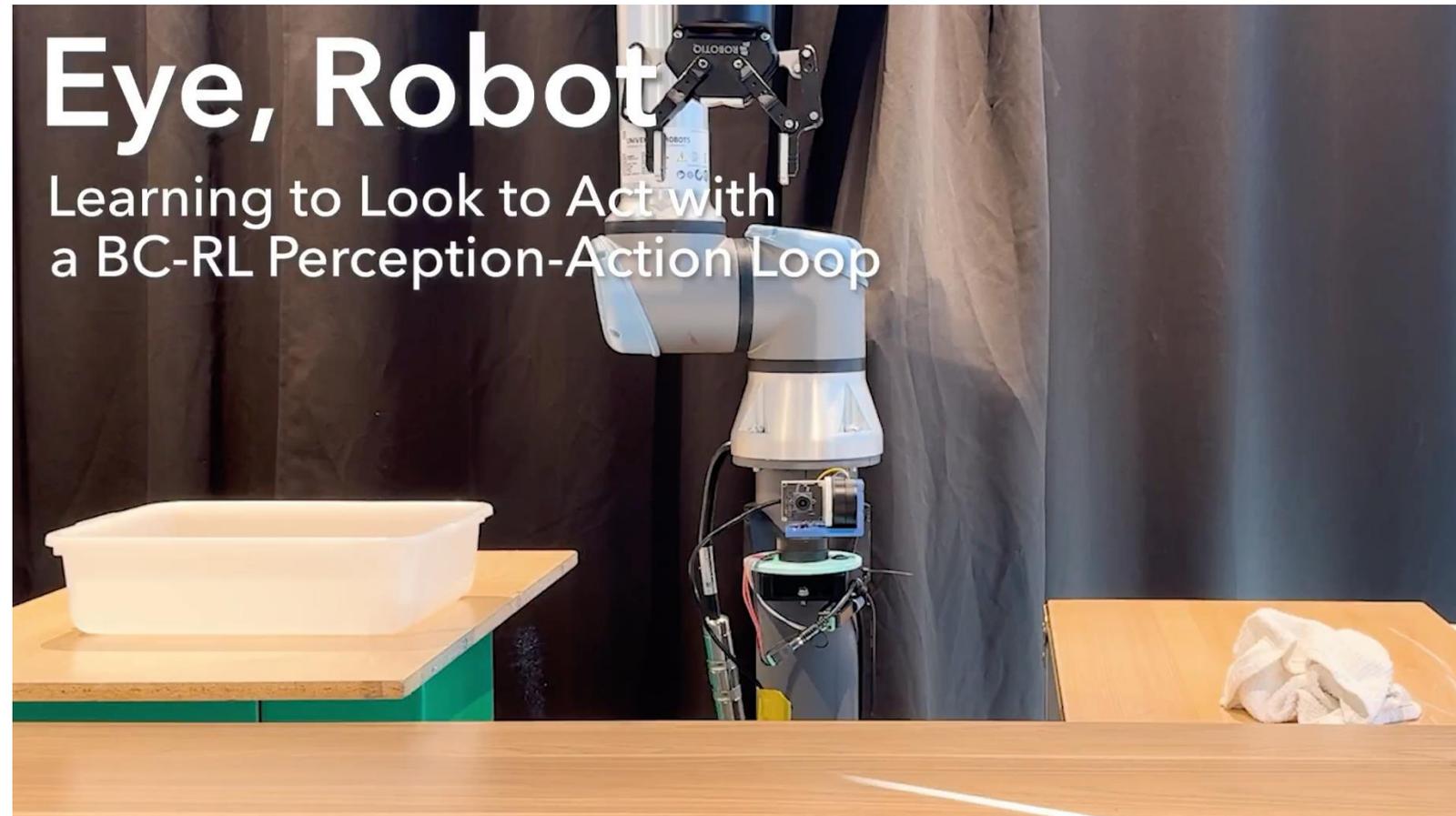
Multimodal Perception

- What is the role of modalities other than vision?
- How to effectively perform sensor fusion?



Interactive and Active Perception

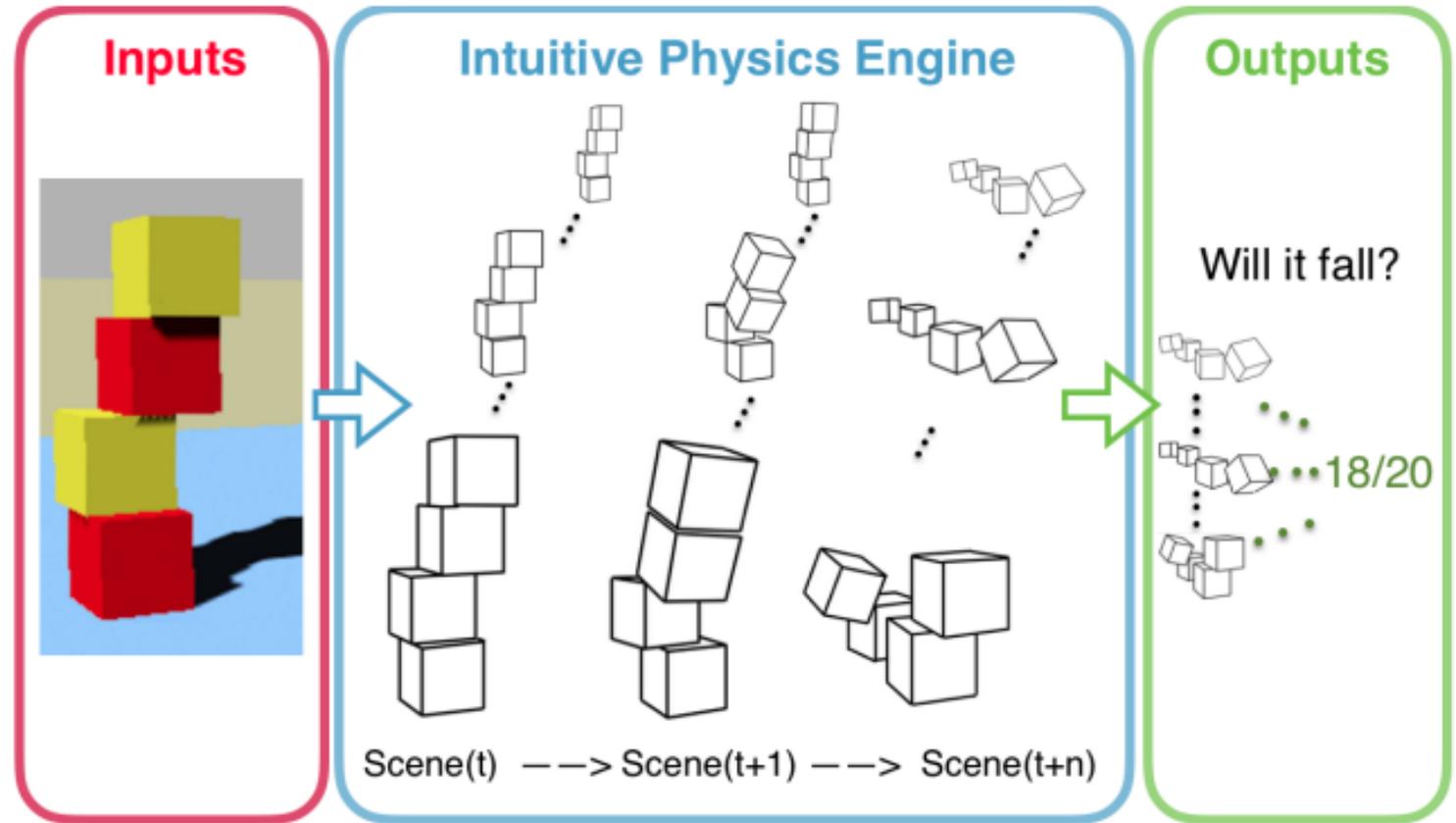
- Humans perceive actively to act with eye gaze, but robots often have fixed cameras.
- How should robots perceive to act and act to perceive?



Eye, Robot. Kerr et al., CoRL 2025

Perception for World Modeling

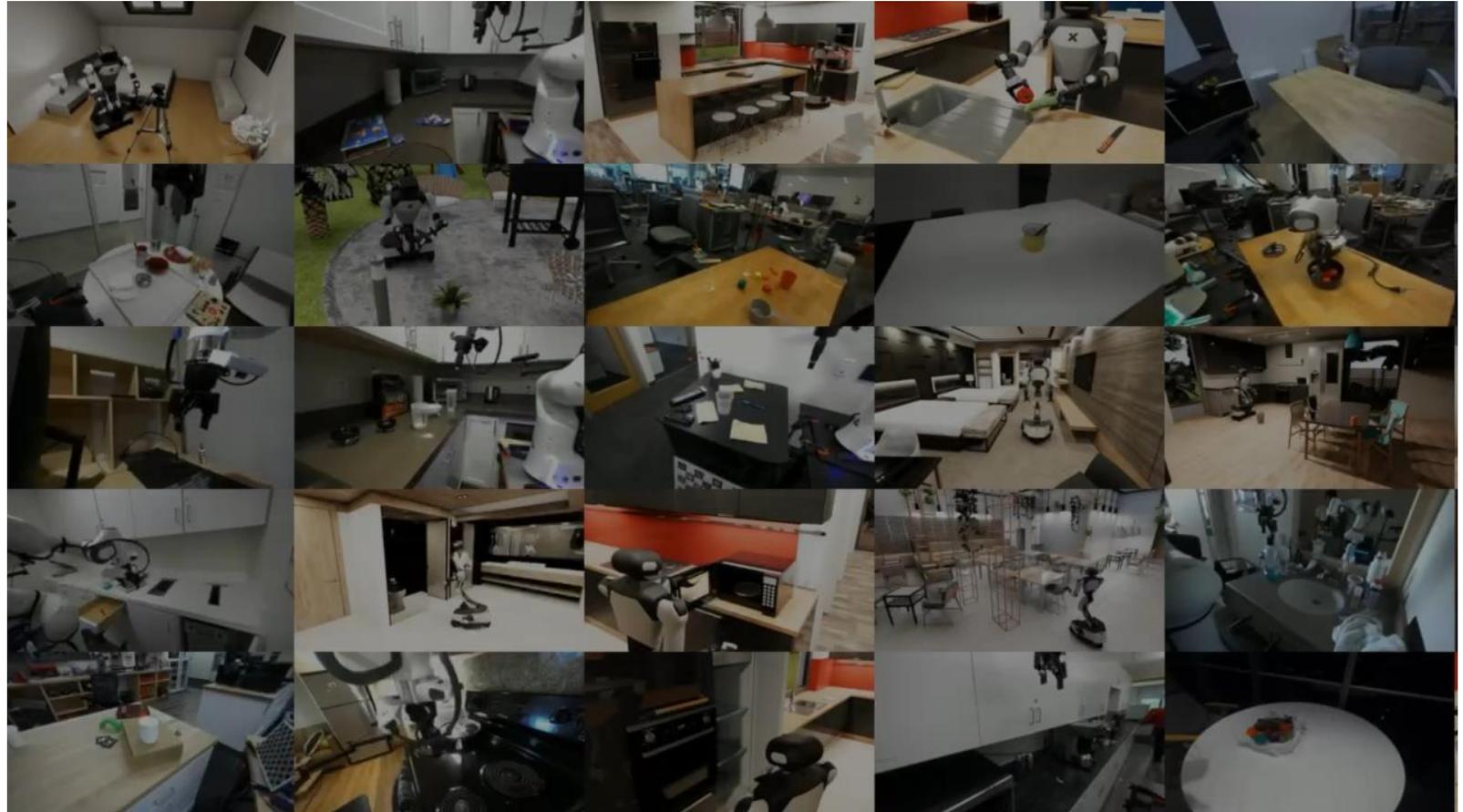
- Humans possess “intuitive physics” commonsense that supports counterfactual reasoning of “what if” queries
- Critical for purposeful manipulation (System 2)
- How can robots achieve the same?



Zhang et al., CogSci 2016

Perception for World Modeling

- Humans possess “intuitive physics” commonsense that supports counterfactual reasoning of “what if” queries
- Critical for purposeful manipulation (System 2)
- How can robots achieve the same?



Huang et al., 2026