

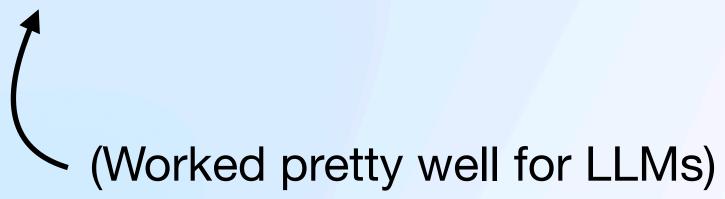


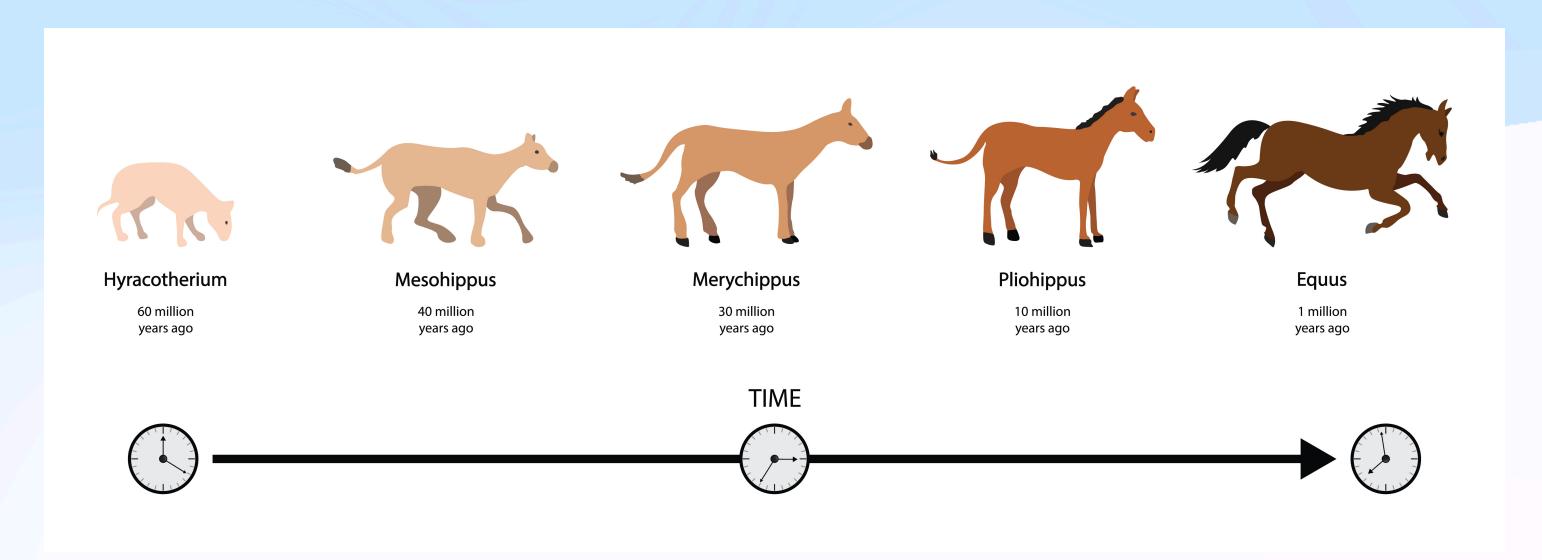
#### For robots, creationism or evolutionism?

Option 1: Create the perfect robot in a lab

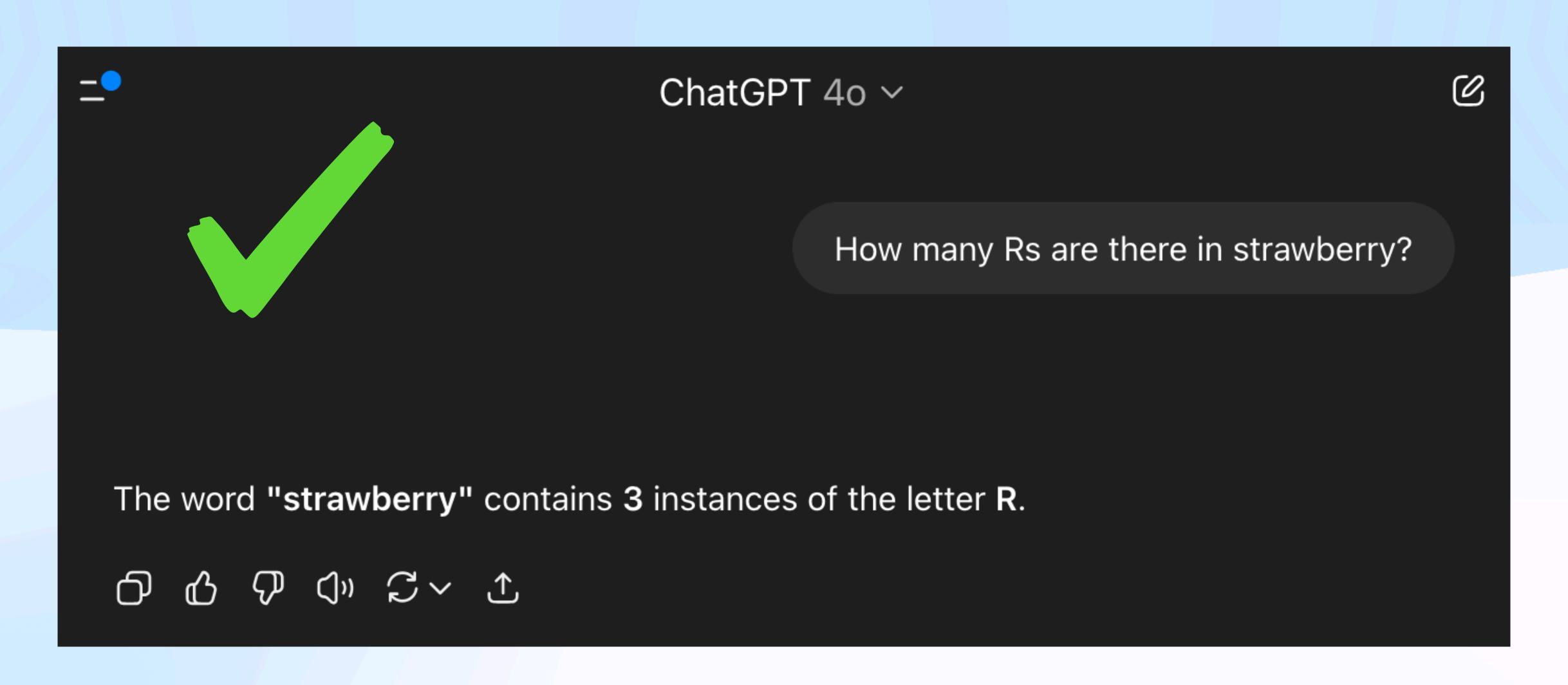




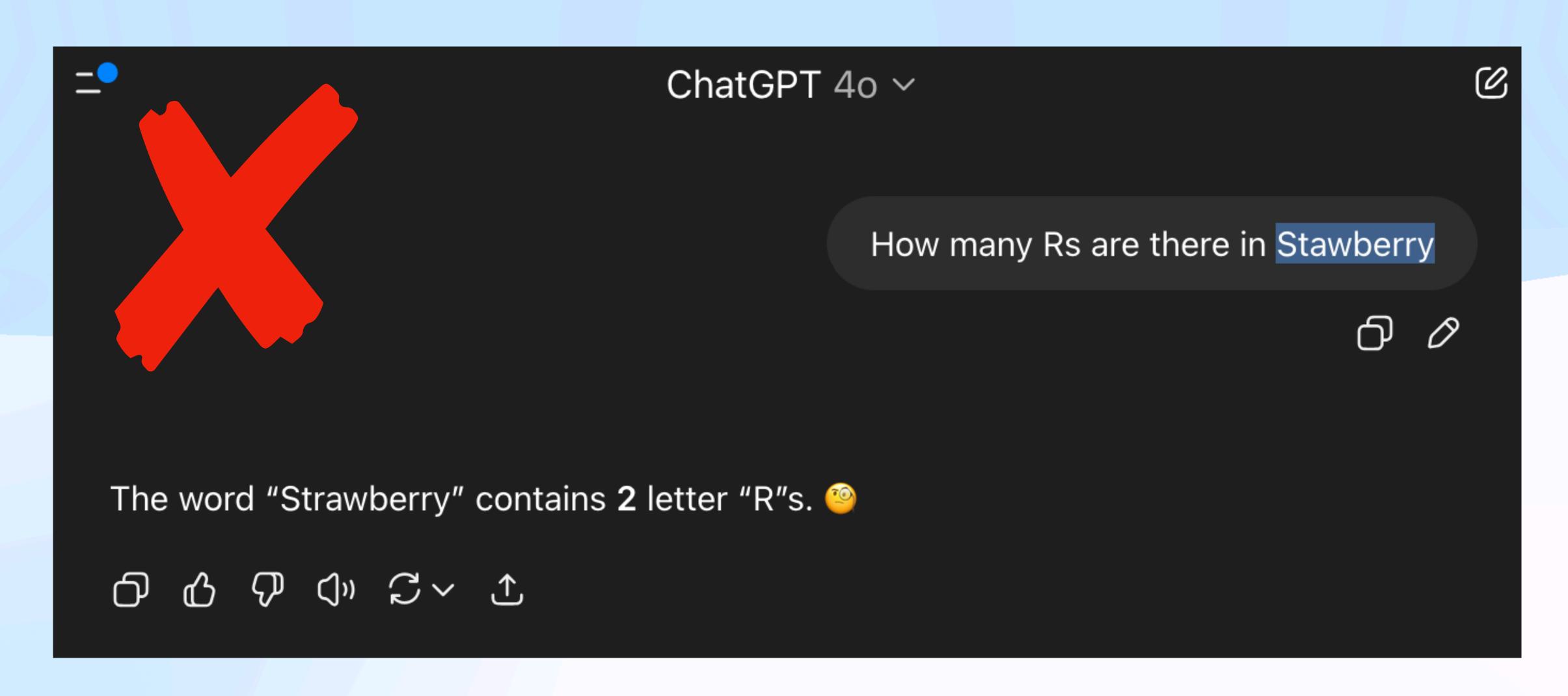




#### Our best methods are still (mostly) interpolation



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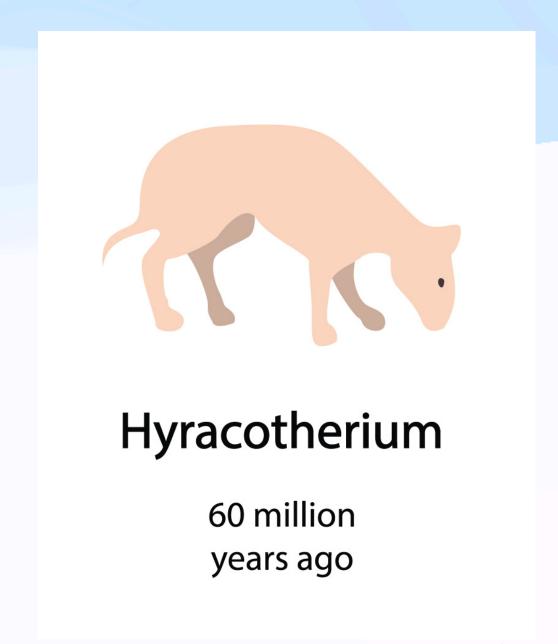


#### For robots, creationism or evolutionism?

Option 1: Create the perfect robot in a lab



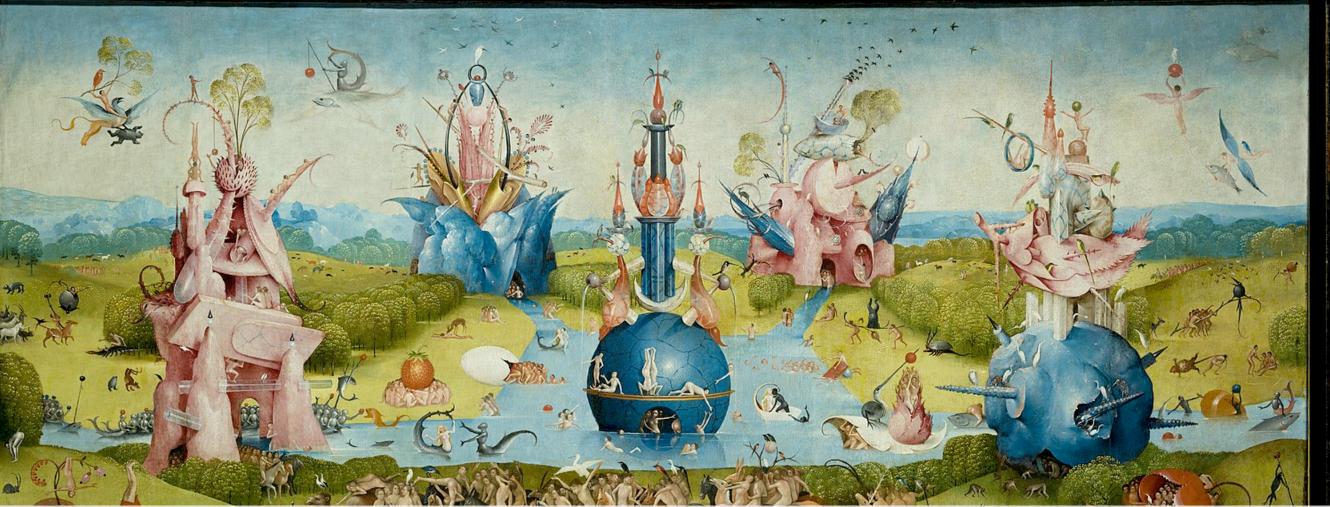
Option 2: Create imperfect robots that get better over iterations



What is this "first imperfect robot"?

Many robots + maybe a few billion \$\$\$







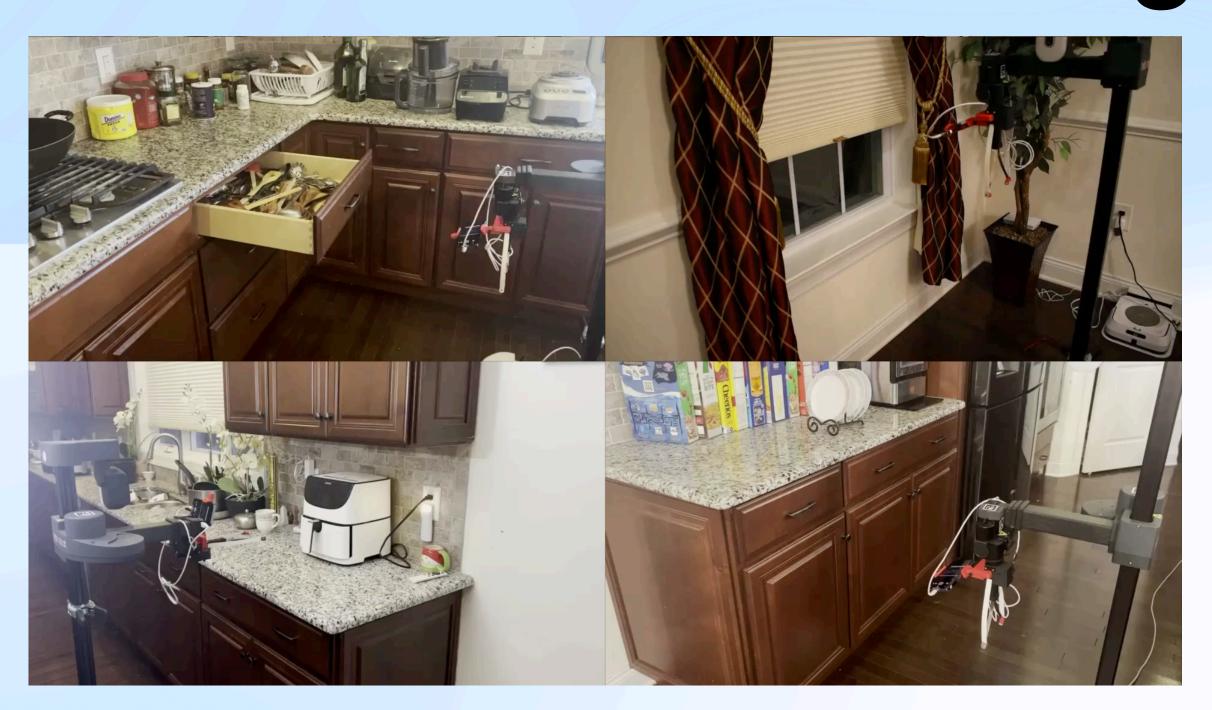
# How can interpolating robots generalize to the wide open world?





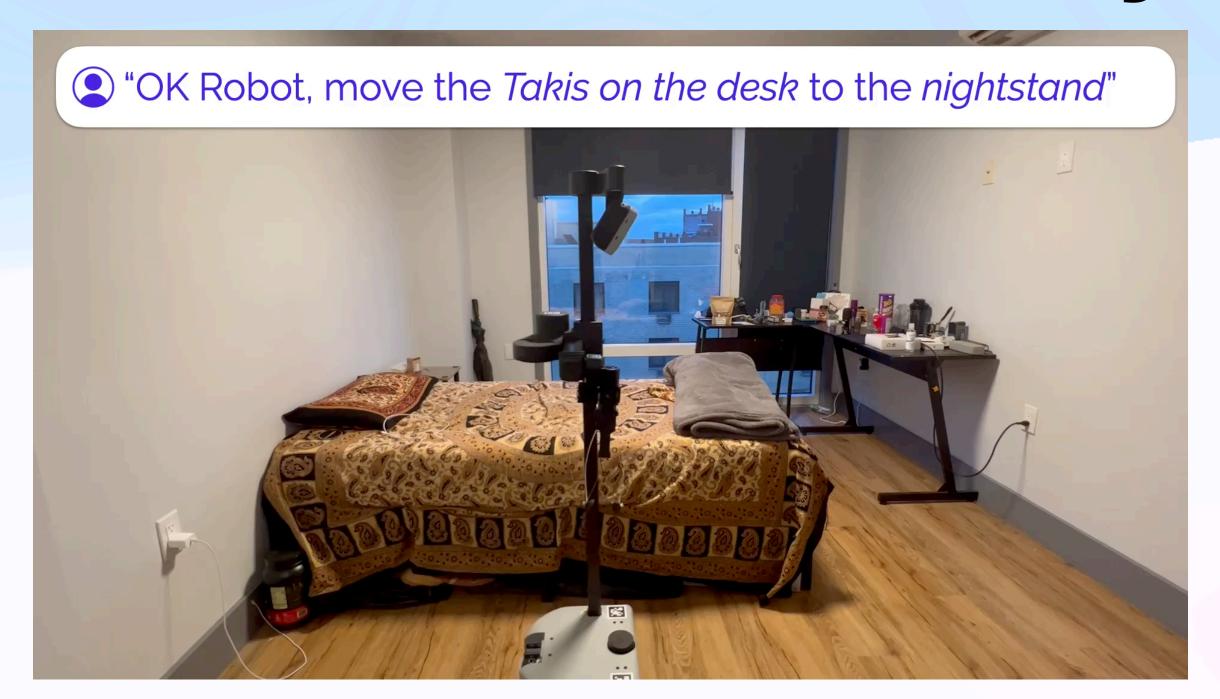


# Interpolating in space with embodiment matching

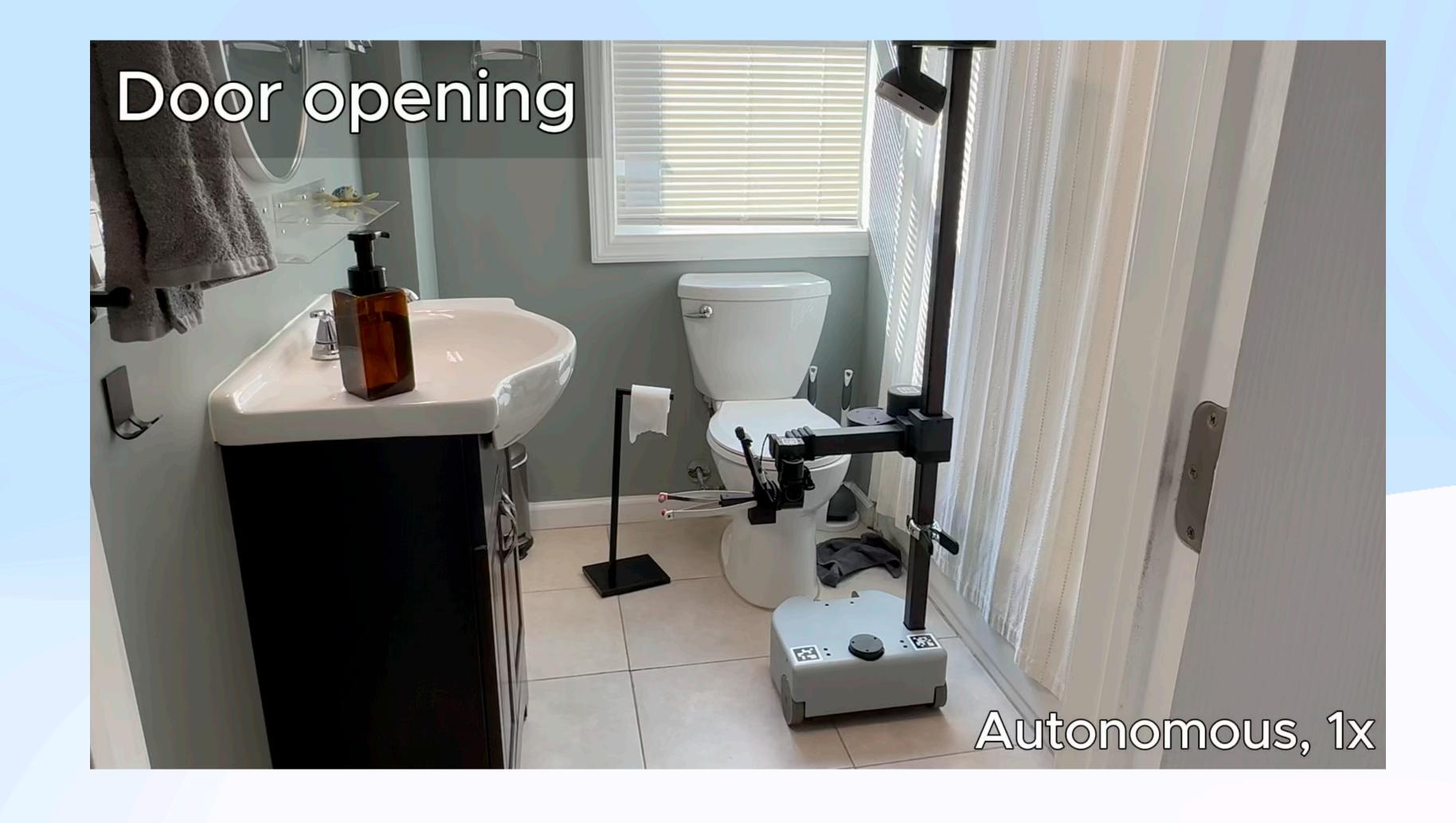


dobb-e.com robotutilitymodels.com

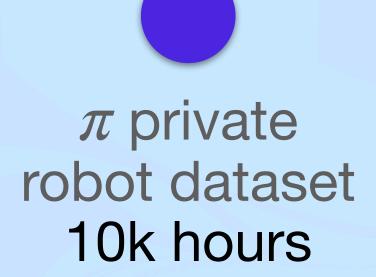
# Interpolating in time with semantic memory



ok-robot.github.io
dynamem.github.io



### The challenge of learning to manipulate anywhere



Large public robot dataset (OXE)
4k hours

GPT-2 475k hours

Llama 3
790,000k
hours

#### Dataset sizes to scale

Assuming 238 words/minute, 1.33 token/word

Thanks @kvablack



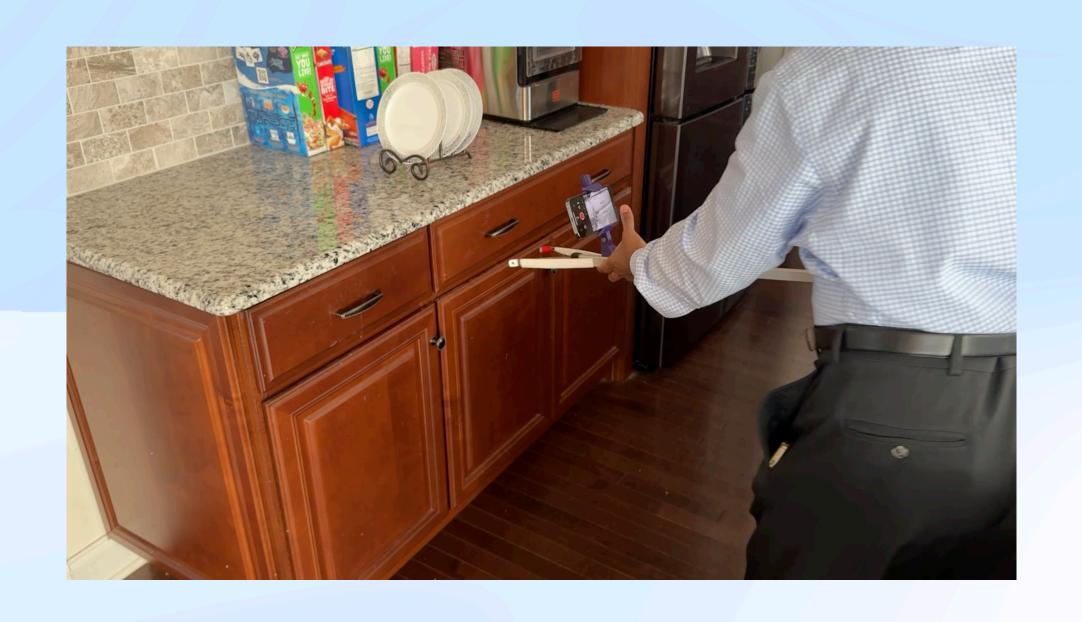




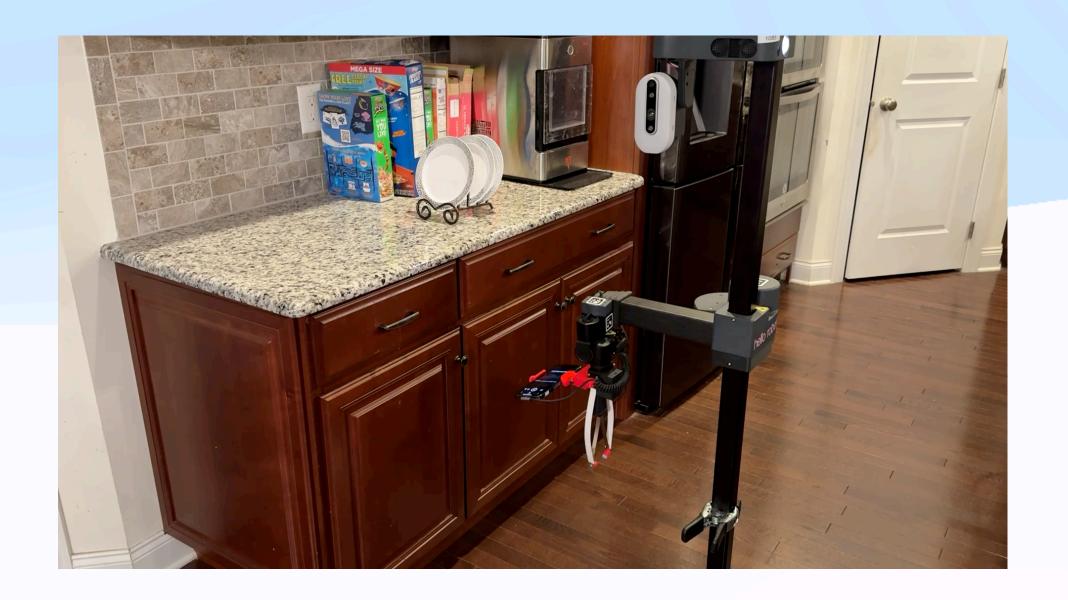
Hello Robot: Stretch end-effector

Our data collection tool

#### Interpolation = efficiency







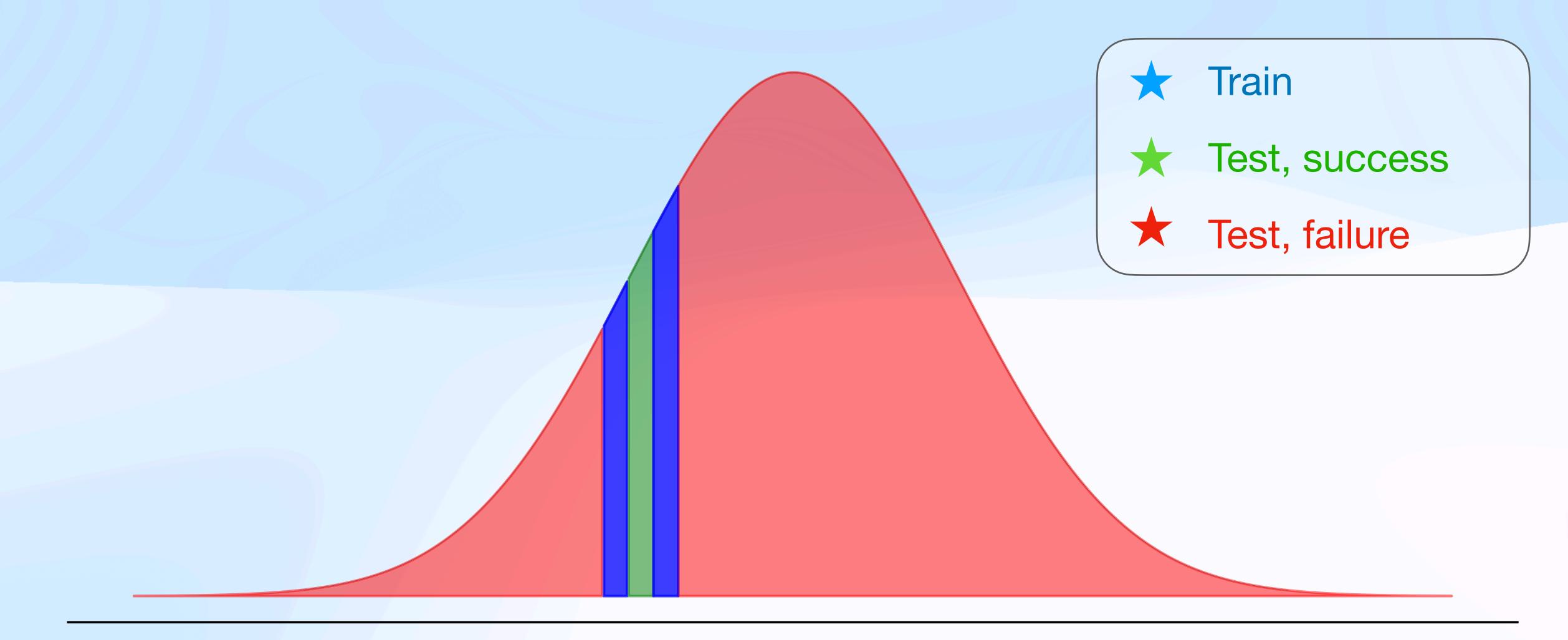
Collect 24 demos 5 minutes

Fine-tune model 15 minutes

Deploy!



#### It's all still interpolation...

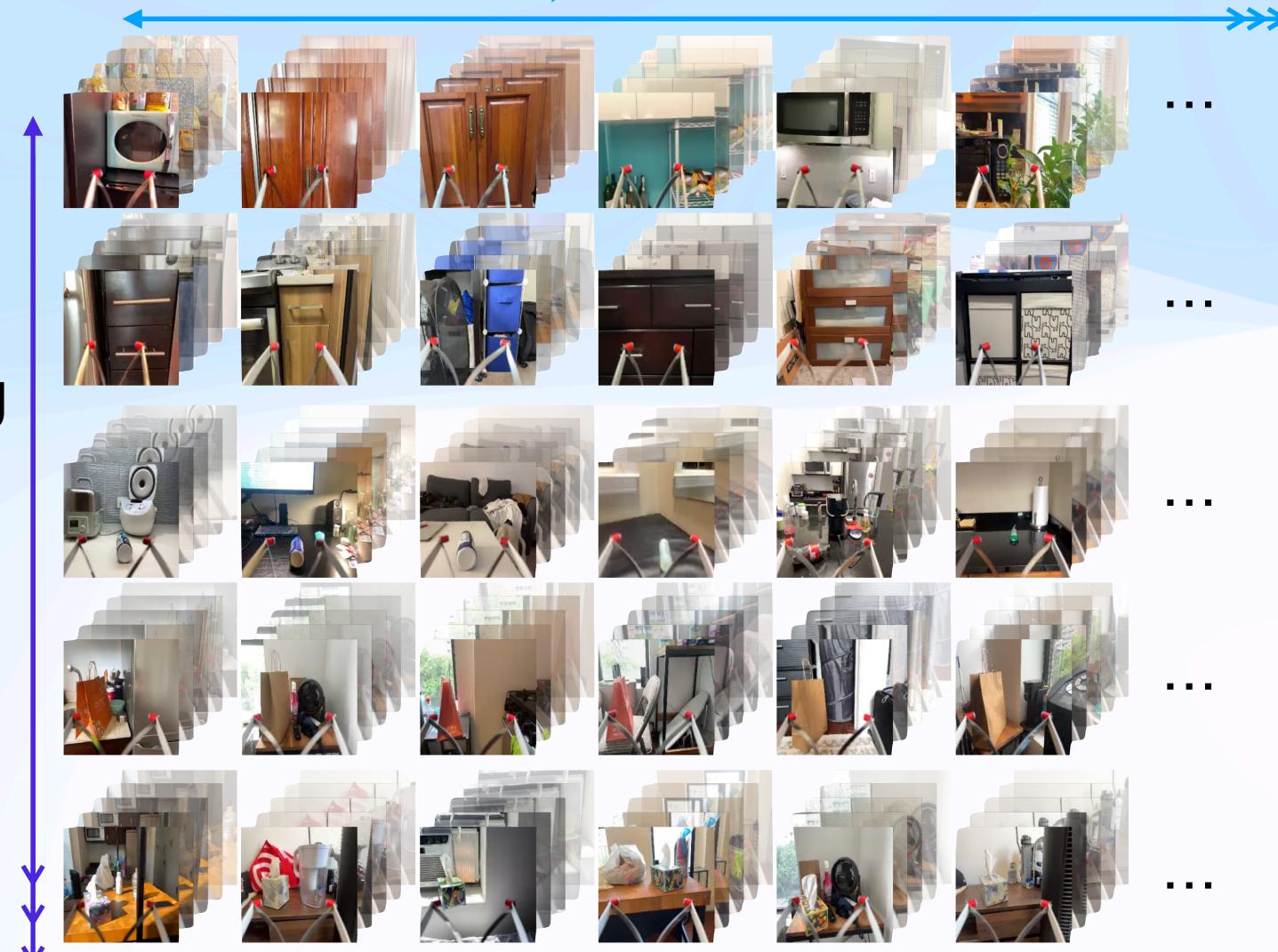


#### But if we focus on quality & diversity...

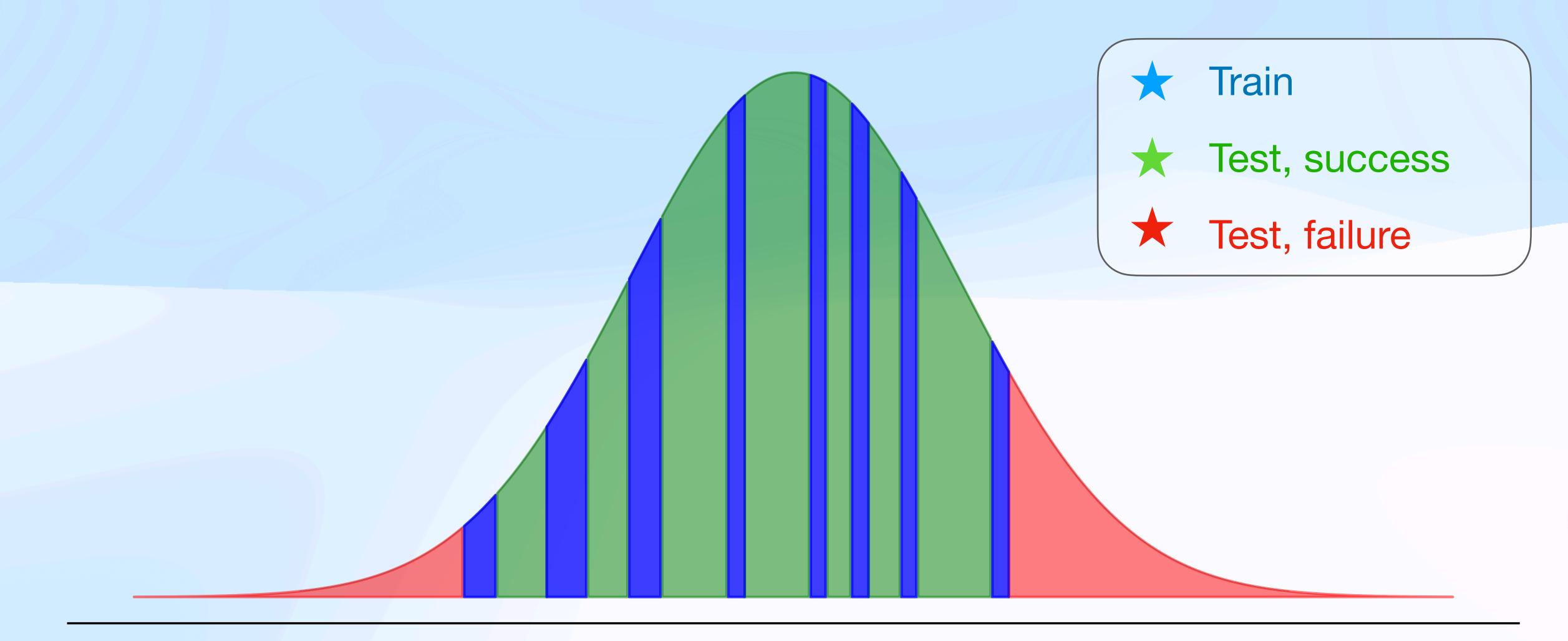
30+ homes, 300+ environments

#### Tasks

Door opening
Drawer opening
Reorientation
Object pickup



#### Interpolation is okay!



#### "Utility models" in novel environments

Right out of the box, no fine tuning!



- Same policy, everywhere, no fine-tuning!
- Same recipe for many single task policies
- 30M params runs on robot

#### Live demo in CVPR 2025



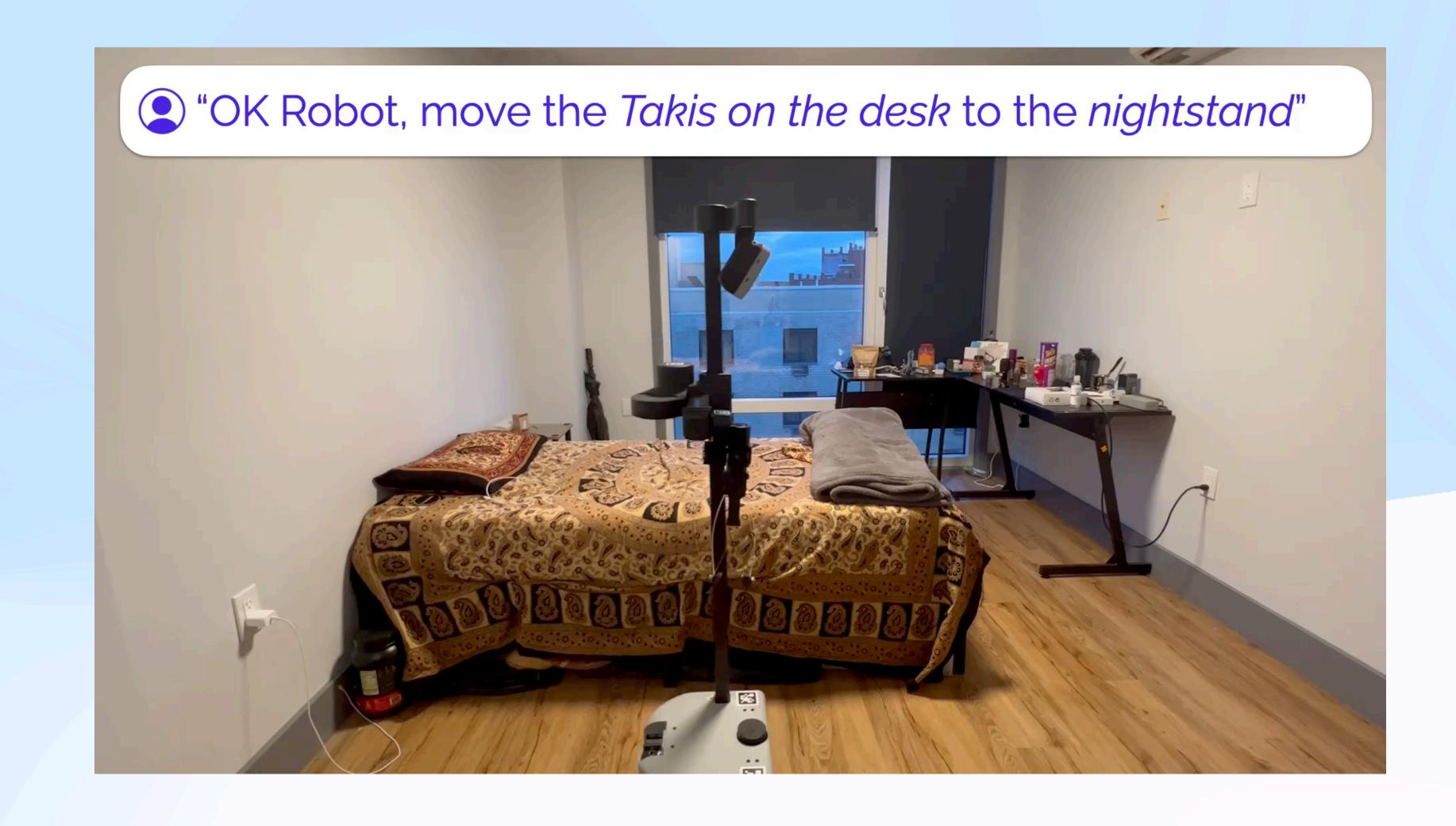
Robot Utility Models



robotutilitymodels.com



Haritheja Etukuru



## The challenge of long-horizon mobile manipulation

#### What passes the "interpolation" smell test?

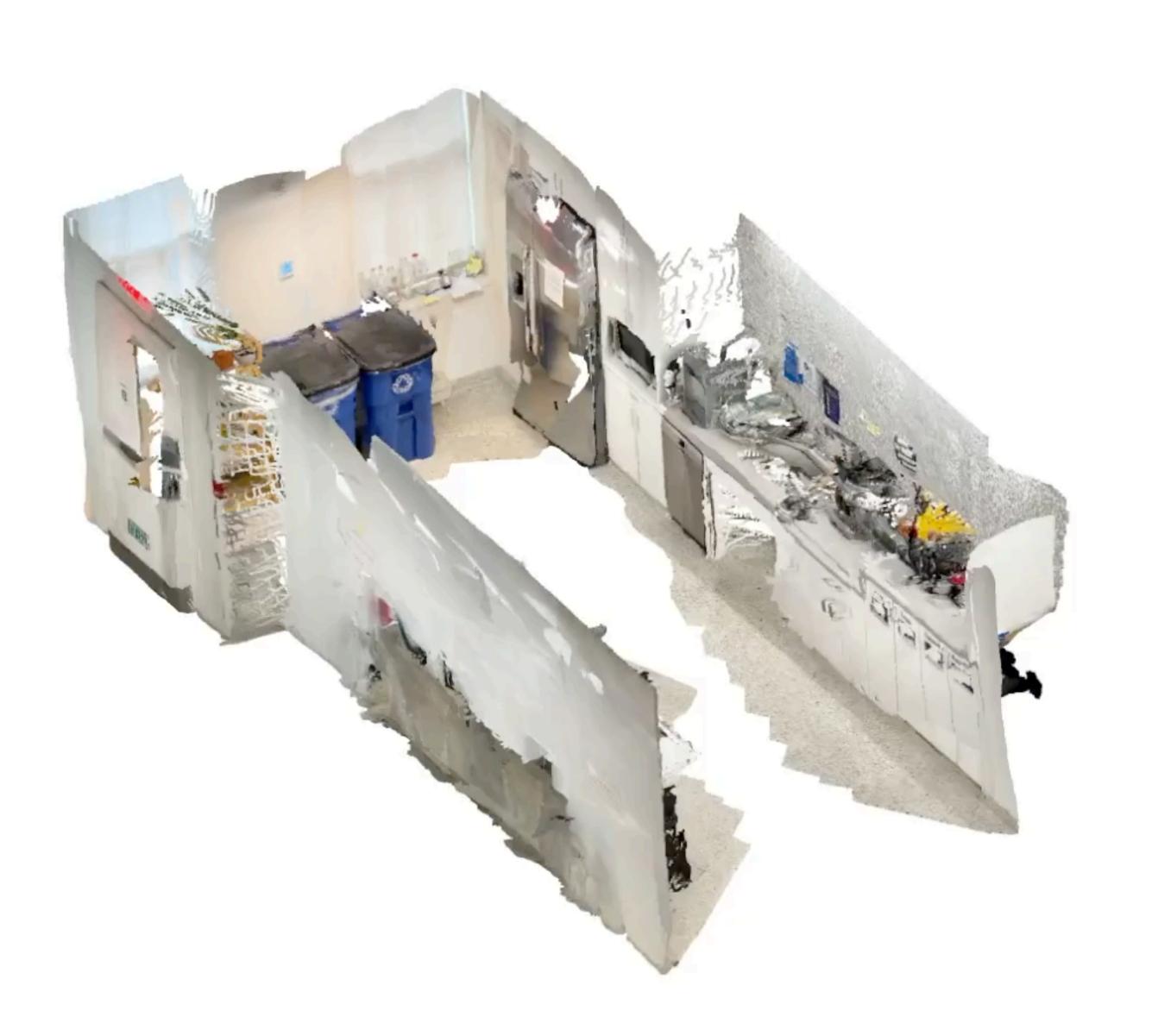
#### Perspective 1: One large model that does everything

 Large, open interpolation space, needs plenty of coverage

#### Perspective 2: Many small models that do one thing, well

 Needs "something" to tie them together over time and space

Memory!



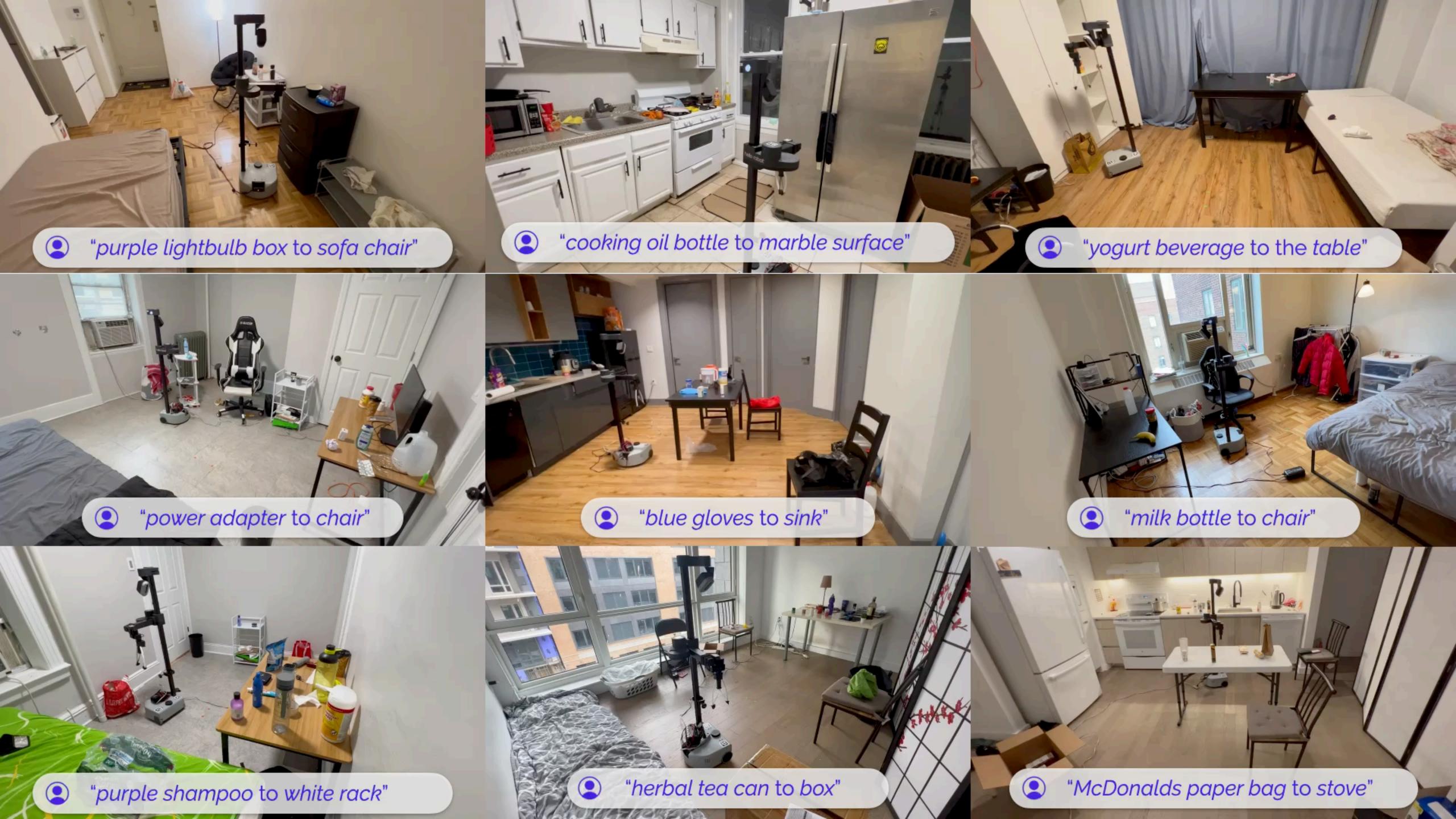
### Spatio-semantic memory representation

Answering "what" and "where"

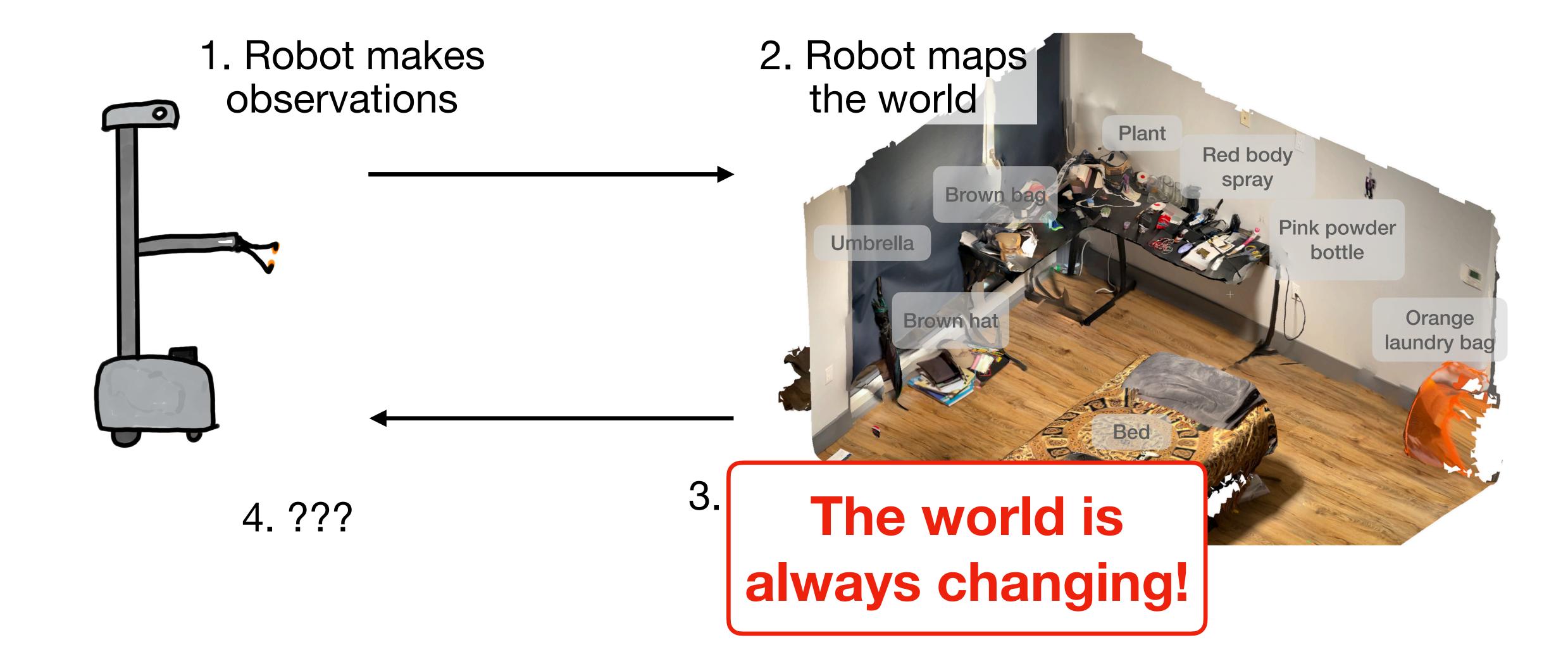
By mapping every point in space to a latent representation

 $x \mapsto z_x \in \mathbb{R}^d$ , which has semantic meaning





#### But robot memory can't keep up with time!





#### Live robot memory defined by four factors

DynaMem: Online Spatio-semantic Memory for Open-world Mobile Manipulation

Factor 1: What do you store?

Factor 2: How do you retrieve?

Factor 3: How do you add?

Factor 4: What do you delete?

### Factor 1: What do you store?

Discrete class labels

Pre 2022

+ abstract latent vectors (CLIP etc.)

Post-2022 CLIP-Fields, OK-Robot + actual images or text

Post-2024, DynaMem

Lookup table

+ Embedding models

+ Large vision/ language models

Factor 2: How do you retrieve?

#### Factor 3: How do you add?

Neural field/ gaussian splat

Scene graph

Voxel maps

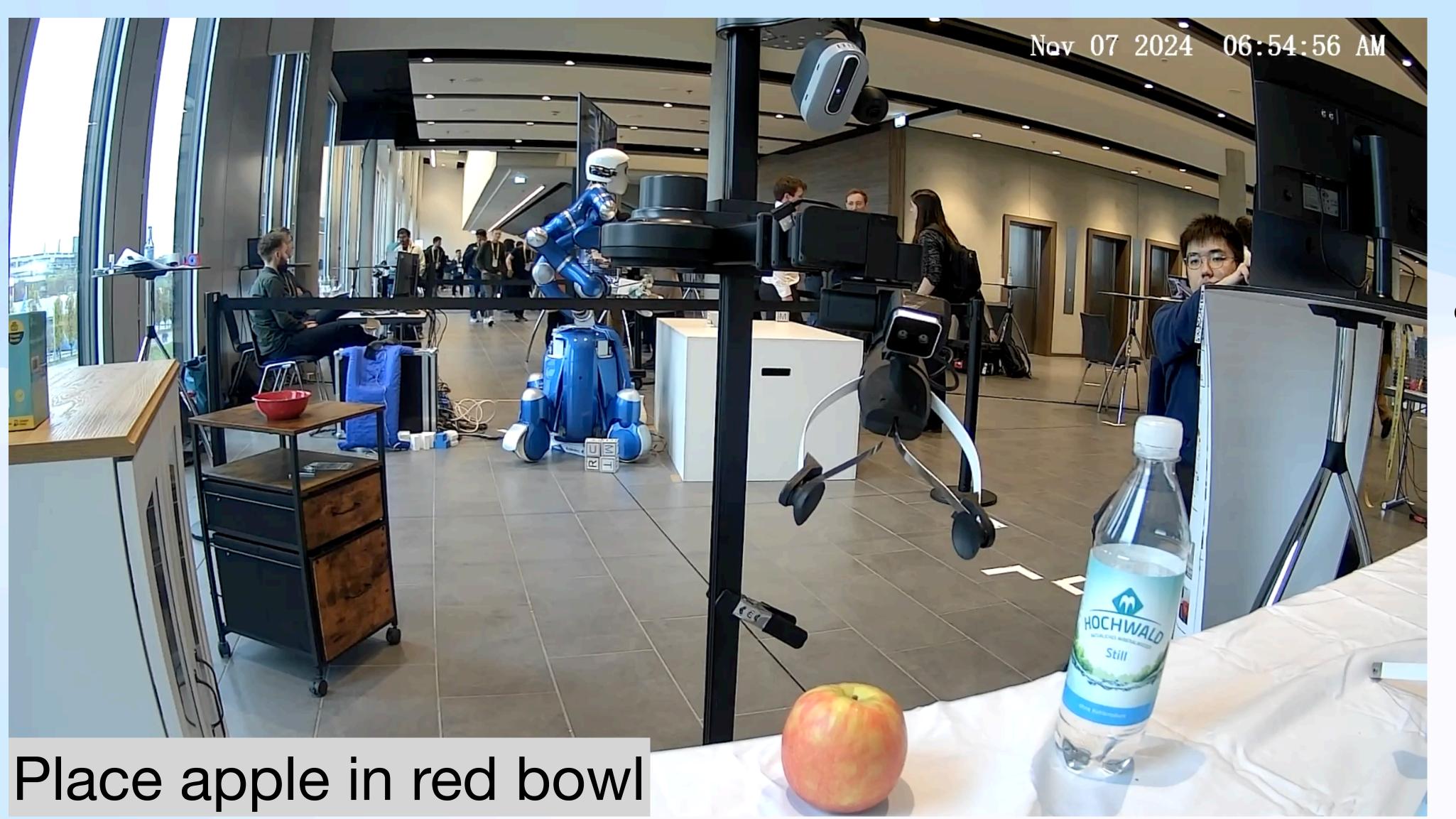
Re-train everything!

Do reliable deduplication!

Use raycasting

Factor 4: How do you delete?

#### Live demo in CoRL 2024



DynaMem



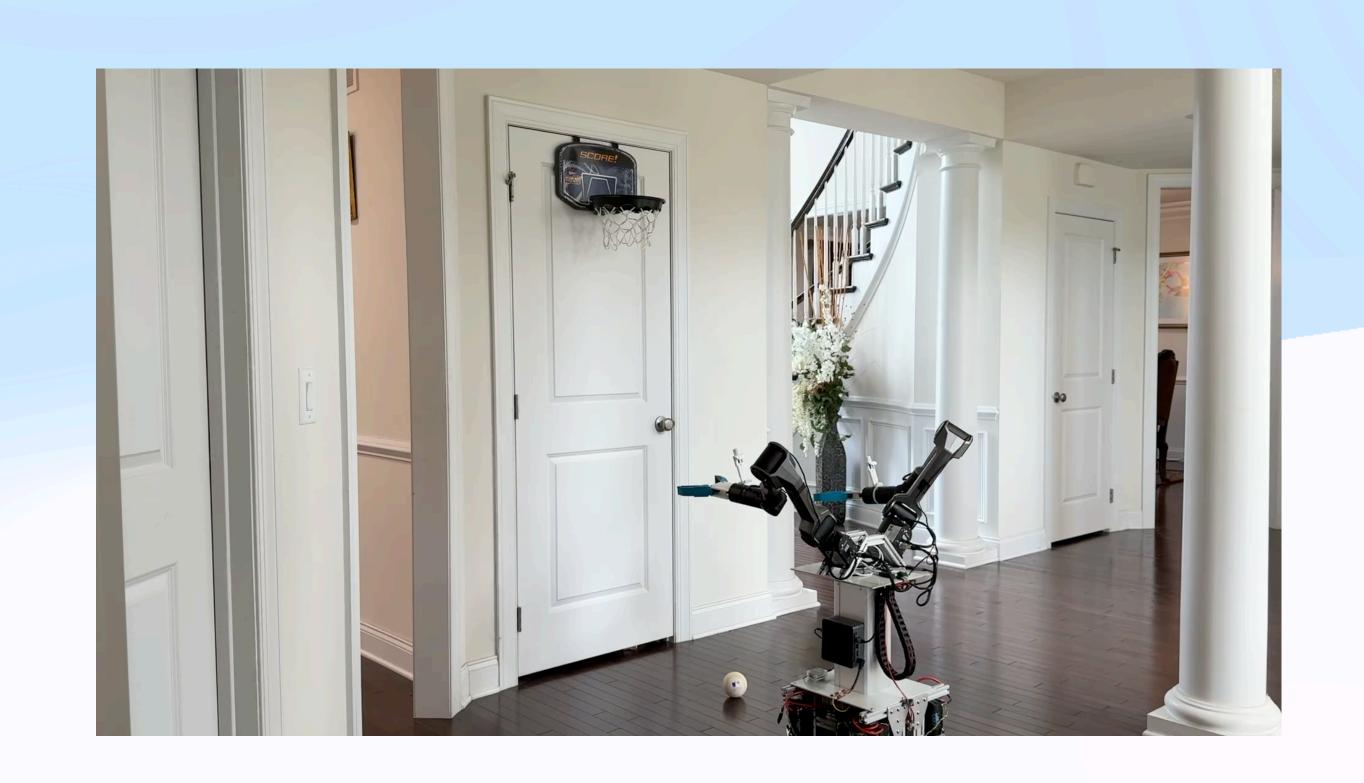
dynamem.github.io



Peiqi Liu

#### Takeaways from the search for general robots

- Interpolation is okay if done with quality and diversity.
- Structure, when done right, will reduce complexity and not increase it.
- Human <-> robot domain gap is inevitable, but we can be smart and work around it.



See more works at <a href="https://mahis.life">https://mahis.life</a> or email me at <a href="notmahi@meta.com">notmahi@meta.com</a>!