

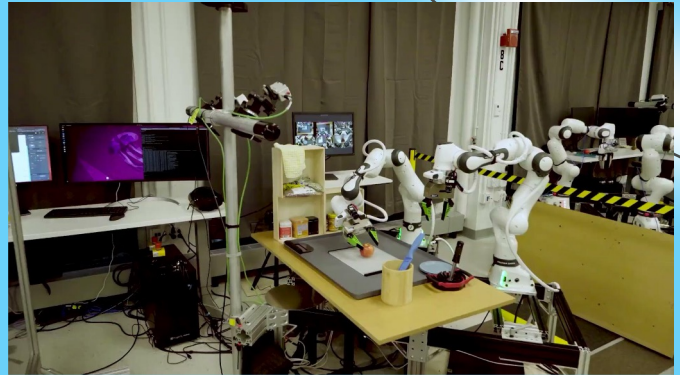
Training Large Behavioral Models for AI Robot Use

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PAPER OVERVIEW

The paper aims to establish LBMs as a reliable system for general purpose physical embodied AI systems.



A Careful Examination of Large Behavior Models for Multitask Dexterous Manipulation

Large Behavior Models Team, Toyota Research Institute



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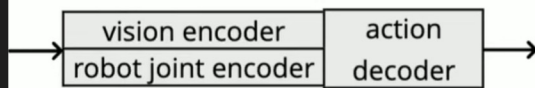
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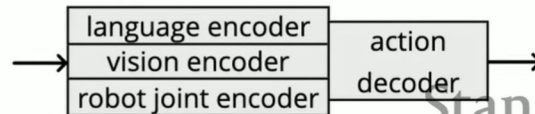
Diffusion Policy (DP) is "single-task"

- Train on ~200 demonstrations
- \Rightarrow visuomotor policy



LBM is "multitask"

- Train on all robot data + internet data
- \Rightarrow *language-conditioned* visuomotor policy



Stanford

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Ramen Corpus

1700 Hours of Robot Training

High quality internal data from real robot teleoperation sessions

Heterogeneous Data

Various tasks, environments, objects, and platforms

Simulations

Over 47,000 full simulations created and utilized to train models

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Platform

Franka Research 3 7DoF Robot Arm

Scene Cameras

Wrist Camera

Adapted finray-style fingers

Evaluation Settings

Nominal (Real)

Object-Centric Distribution Shift (Real)

Station Distribution Shift (Real)

Example Initial Conditions for PutKiwiInCenterOfTable

Example Evaluation Tasks

PutKiwiInCenterOfTable

PushCoasterToMug

BikeRotorInstall

SetupBreakfastTable

CutAppleInSlices

PutBellPepperInBin

PutSpatulaOnPlateFromTable

PlaceCupByCoaster

PutBananaOnSaucer

PutContainersOnPlate

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Ramen Dataset Breakdown

OXE-Ramen

- Curated set of online OpenX-Embodiment datasets
- 1,150 hours of robot data
- Made to correspond with TRI format

TRI-Ramen

- high-quality data collected internally at TRI
- 545 hours of robot data

Fine tuning %

?: Fraction of available data used for fine tuning for a particular dataset

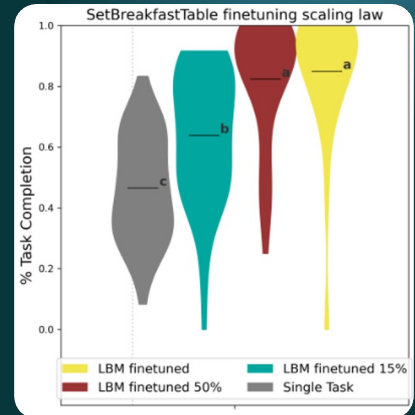


Figure 5: Violin plot of various datasets in action for a particular task.

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Ensuring Evaluation Fairness & Metrics

A/B Testing

- Tester doesn't know which robot is using which training
- Blind testing

Randomization

- Policy order was randomized

Task Completion

Robot evaluator:

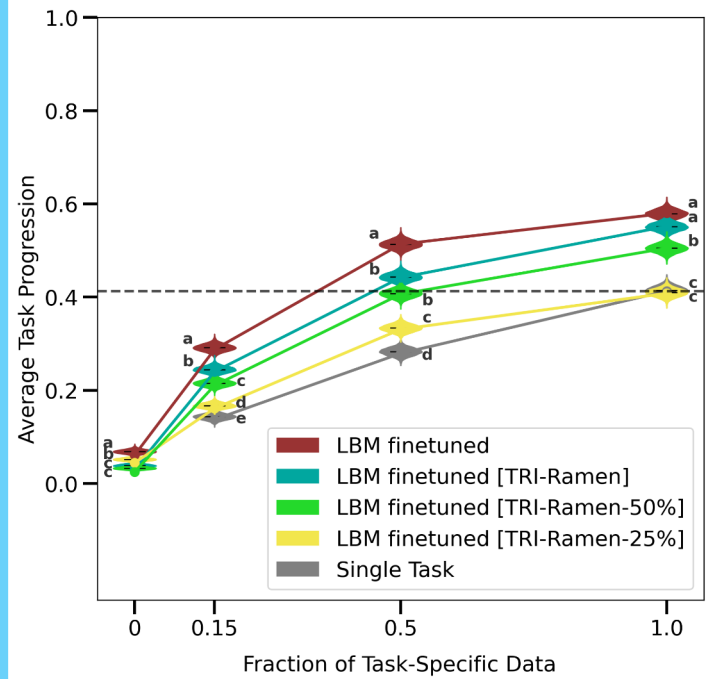
- Set of milestones per task
- Milestones achieved are counted up

Initial Condition Overlay



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Scaling and Efficiency



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Limitation of LBM Capabilities

Non-fine tuned LBM

- Does not always outperform from-scratch single task training

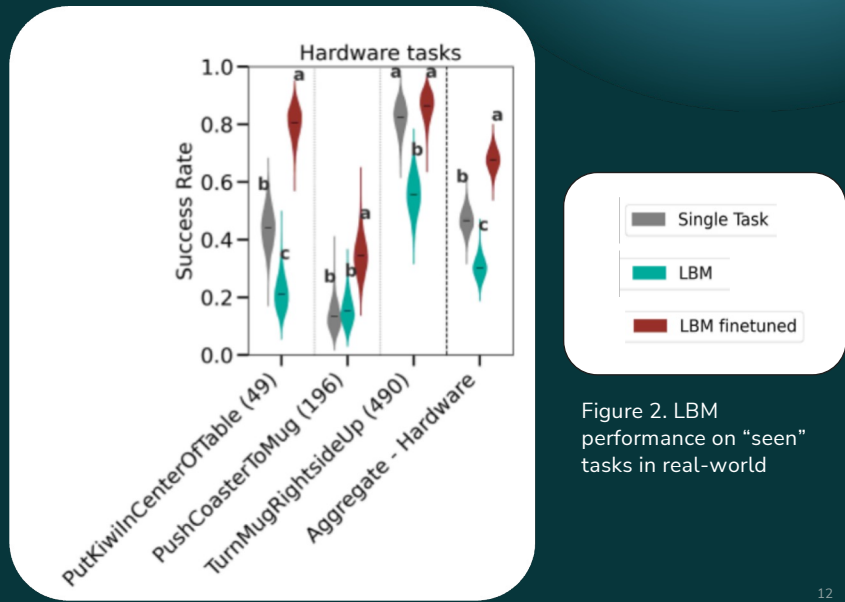


Figure 2. LBM performance on "seen" tasks in real-world

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The Opportunity

- New levels of dexterity (manipulating cloth, liquids, etc)
- Programmed via imprecise natural language and/or a few demonstrations
- “Common-sense” for physical intelligence, leading to open-world robustness
- AI
- Household Robotics

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Sources

<https://bostondynamics.com/blog/large-behavior-models-atlas-find-new-footing/>
<https://toyotaresearchinstitute.github.io/lbml/>

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Questions?

Confidential

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