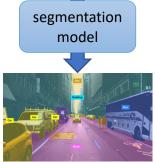
### **Robotic Learning with Large Datasets:** Towards General-Purpose Models for All Robots

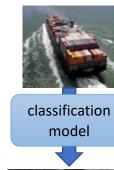
Sergey Levine UC Berkeley



### How AI used to work





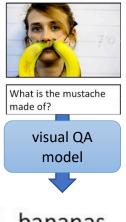


container ship container ship lifeboat amphibian fireboat drilling platform



captioning model

A group of people shopping at an outdoor market.

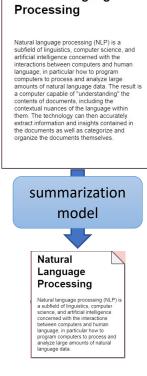


bananas

"Horrible services. The room was dirty and unpleasant. Not worth the money."

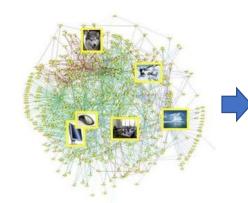
sentiment model

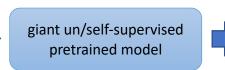
NEGATIVE



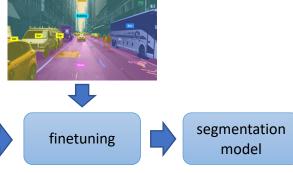
Natural Language

### How AI works now





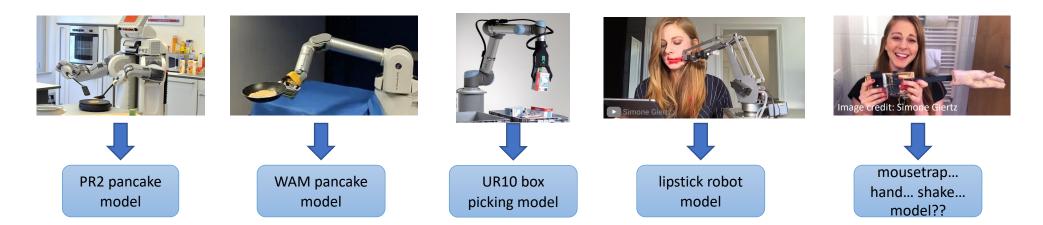
"foundation model"



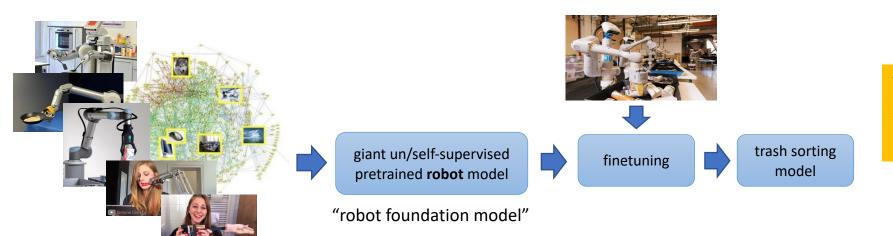


Dreiss et al., PaLM-E: An Embodied Multimodal Language Model. 2023.

### How robotic learning works now

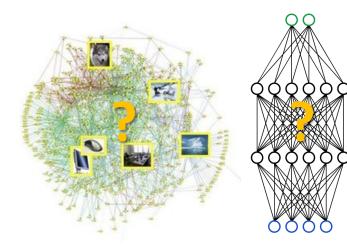


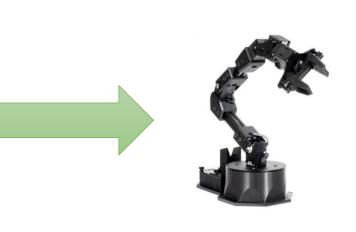
### How robotic learning will work in the future



What would it take to build such a general-purpose robot model?

### How does this look for robots?







#### The data needs to...

- Be diverse enough to support my (new) robot
- Cover a range of environments
- Cover a range of behaviors

#### The model needs to ...

- > Be able to train on all this data
- Be flexible enough to repurpose to new tasks
- Perform at least some tasks in zero shot?





**Bridge Data:** <u>https://rail-berkeley.github.io/bridgedata/</u> Led by Homer Walke, Frederik Ebert, with many others





## Pretraining on bridge data with offline RL (PTR)

Offline RL Pretraining on Bridge Dataset

Put Sushi in Pot 10 domains 100 tasks put corn in bowl 12k demos Put Eggplant on Plate Target dataset batch-mixing bridge and target data  $1 - |\tau|$ Put Eggplant on Plate 2. Fine-Tune on Mix of Bridge Data and Target Data 1. Pre-train on bridge data

Offline RL Fine-tuning on Target Data + Bridge Data

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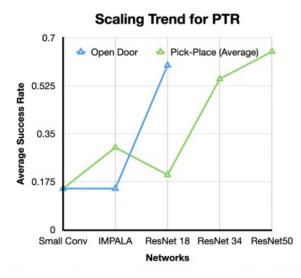
Aviral Kumar, Frederik Ebert, Yanlai Yang, Anikait Singh, Chelsea Finn, Sergey Levine. Pre-Training for Robots: Offline RL Enables Learning New Tasks from a Handful of Trials. 2022.

## PTR (Pre-Training for Robots) Results



learning **entirely new** skills after pretraining on the bridge dataset



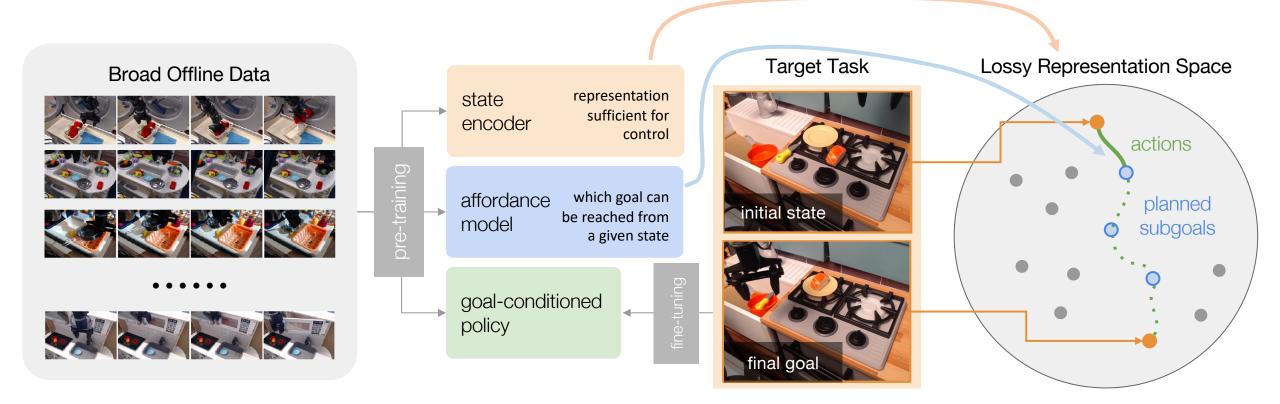


		B	C finetuning		Joint tr	aining	Target	data only	Pre-trai	n. rep. + BC finetune
Task	<b>PTR (Ours)</b>	BC (fine.)	Autoreg. BC	BeT	COG	BC	CQL	BC	R3M	MAE
Take croissant from metal bowl	7/10	3/10	5/10	1/10	4/10	4/10	0/10	1/10	1/10	3/10
Put sweet potato on plate	7/20	1/20	1/20	0/20	0/20	0/20	0/20	0/20	0/20	1/20
Place knife in pot	4/10	2/10	2/10	0/10	1/10	3/10	3/10	0/10	0/10	0/10
Put cucumber in pot	5/10	0/10	1/10	0/10	2/10	1/10	0/10	0/10	0/10	0/10

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Aviral Kumar, Frederik Ebert, Yanlai Yang, Anikait Singh, Chelsea Finn, Sergey Levine. Pre-Training for Robots: Offline RL Enables Learning New Tasks from a Handful of Trials. 2022.

### An unsupervised subgoal planning framework



Fang, Yin, Nair, Walke, Yan, Levine. Generalization with Lossy Affordances: Leveraging Broad Offline Data for Learning Visuomotor Tasks. 2022.

### Task 1: Move colander onto stove and drop object into colander

Model-Free 0.0% Success Rate



FLAP (Target Data Only) **25.0%** Success Rate



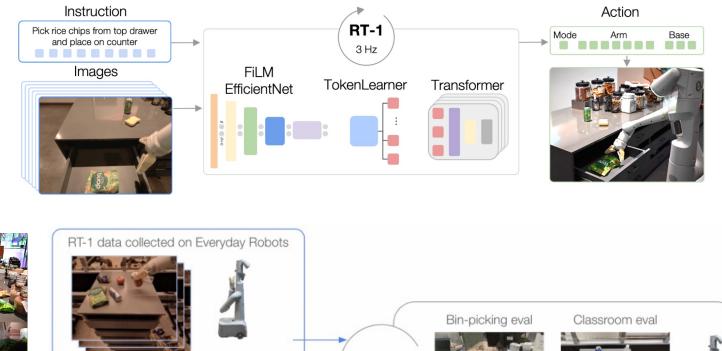
#### 75.0% Success Rate



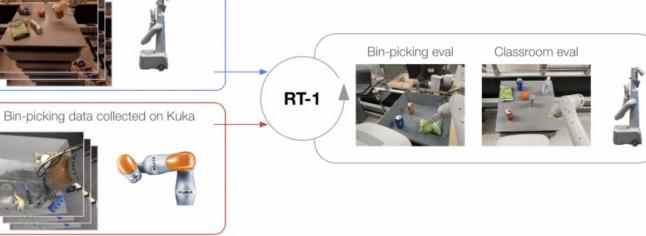
FLAP (Broad Data)

### Massively multi-task policies: RT-1







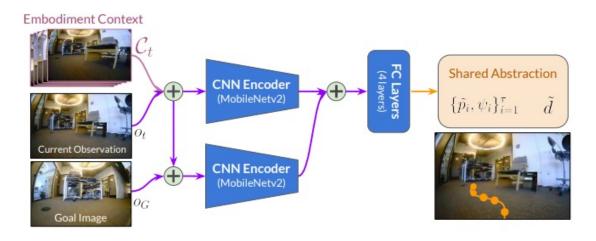


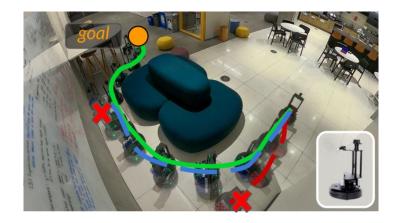
# One policy to control many robots?

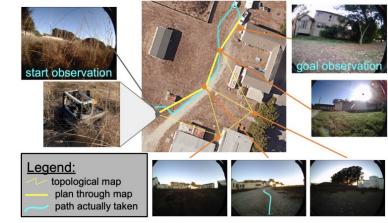
### the question

Can we create a dataset and model that can generalize in **zero shot** to control entirely new robots?

### the setup







Shah\*, Sridhar\*, Bhorkar, Hirose, Levine. GNM: A General Navigation Model to Drive Any Robot. 2022.

## One policy to control many robots?

	Dataset	Platform	Speed	Amt.	Environment
1	GoStanford [26]	TurtleBot2	0.5m/s	14h	office
2	RECON [32]	Jackal	1m/s	25h	off-road
3	CoryHall [35]	RC Car	1.2m/s	2h	hallways
4	Berkeley [33]	Jackal	2m/s	4h	suburban
5	SCAND-S [36]	Spot	1.5m/s	8h	sidewalks
6	SCAND-J [36]	Jackal	2m/s	1h	sidewalks
7	Seattle [37]	Warthog	5m/s	1h	off-road
8	TartanDrive [38]	ATV	10m/s	5h	off-road
	Ours			60h	



**RC-Car** (Kahn et al. 2018)



Spot (Karnan et al. 2022)



TurtleBot (Hirose et al. 2019)

Warthog

(Shaban et al. 2021)

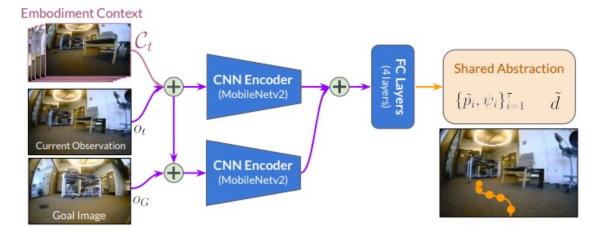


Jackal

(Shah et al. 2021, 2022)



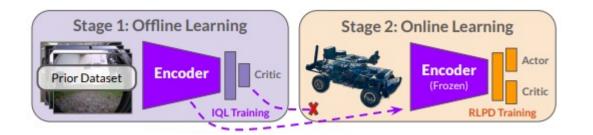
(Triest et al. 2022)

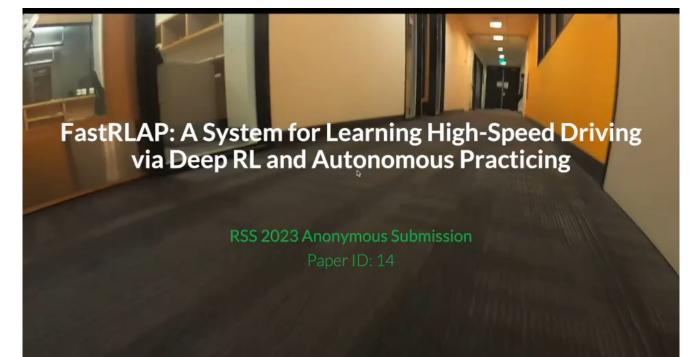




Shah\*, Sridhar\*, Bhorkar, Hirose, Levine. GNM: A General Navigation Model to Drive Any Robot. 2022.

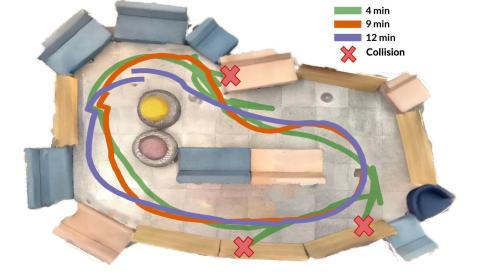
### Now make it go fast!

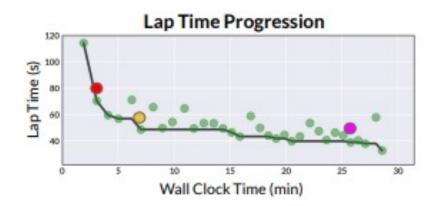




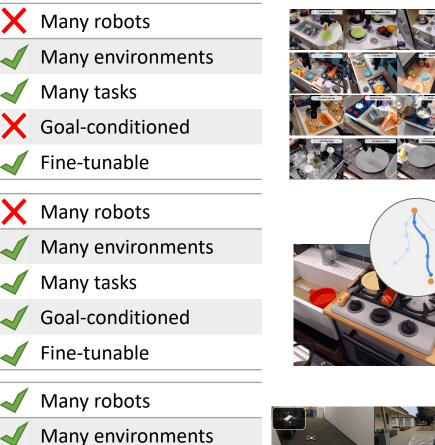
All Videos at 1x (Real-Time)

Stachowicz, Bhorkar, Shah, Kostrikov, Levine. FastRLAP: A System for Learning High-Speed Driving via Deep RL and Autonomous Practicing. 2023.





# How far have we gotten?



Bridge data & pre-training for robots (PTR)

Bridge data & goal-conditioned RL (FLAP)



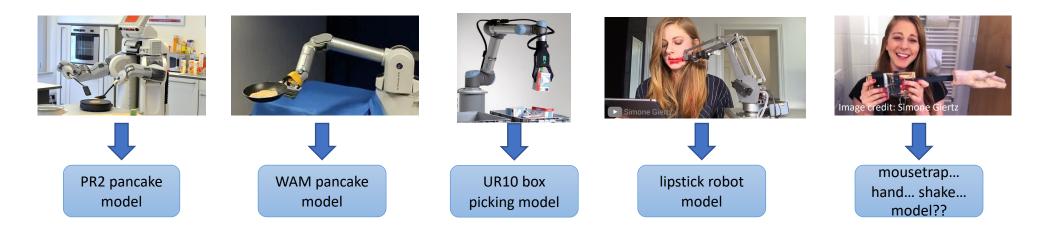


**Goal-conditioned** Fine-tunable

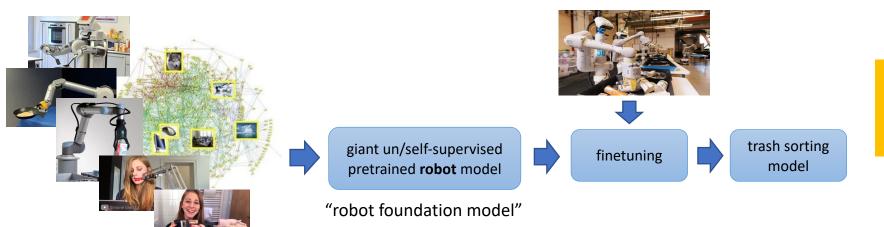


General navigation models (GNM, FastRLAP)

### How robotic learning works now



### How robotic learning will work in the future



What would it take to build such a general-purpose robot model?



RAIL Robotic AI & Learning Lab website: <u>http://rail.eecs.berkeley.edu</u> source code: <u>http://rail.eecs.berkeley.edu/code.html</u>