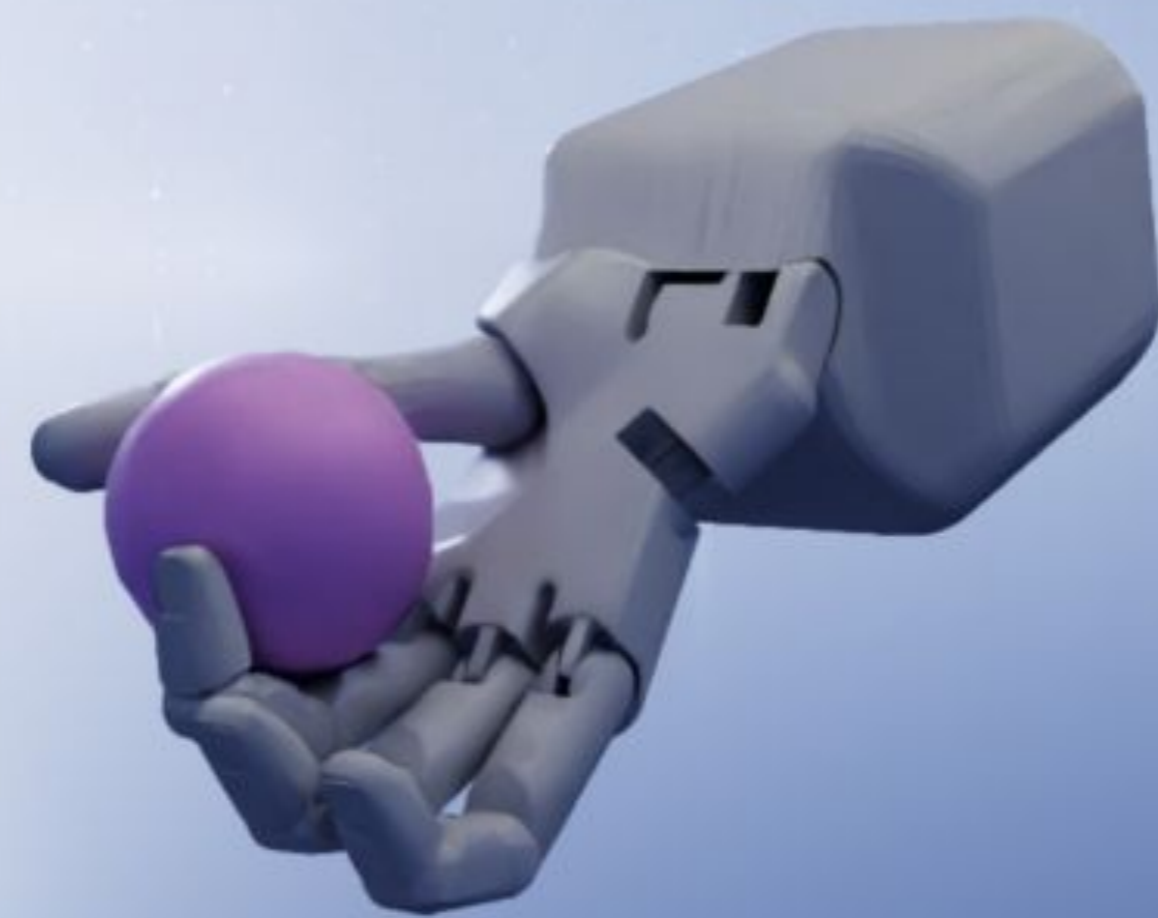


# roto 2.0: The Robot Tactile Olympiad

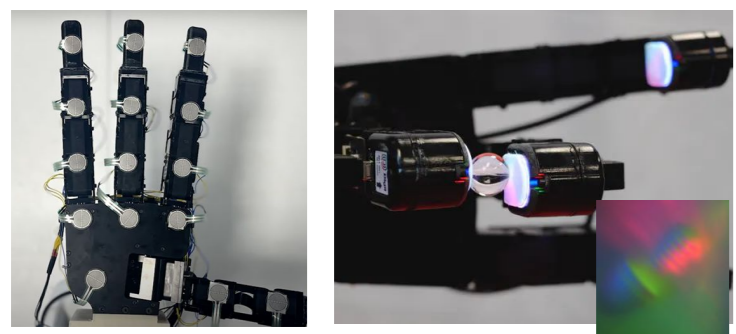
Elle Miller<sup>1</sup>, Jayaram Reddy<sup>2</sup>, Ayush Deshmukh<sup>1</sup>, Trevor McInroe<sup>1</sup>, David Abel<sup>1</sup>, Oisin Mac Aodha<sup>1</sup>, Sethu Vijayakumar<sup>1</sup>



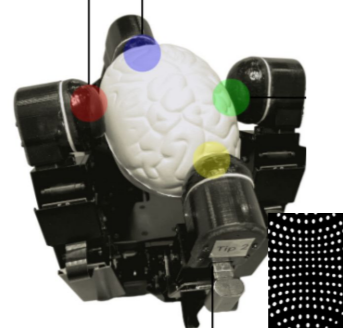
## Why is tactile RL so hard???



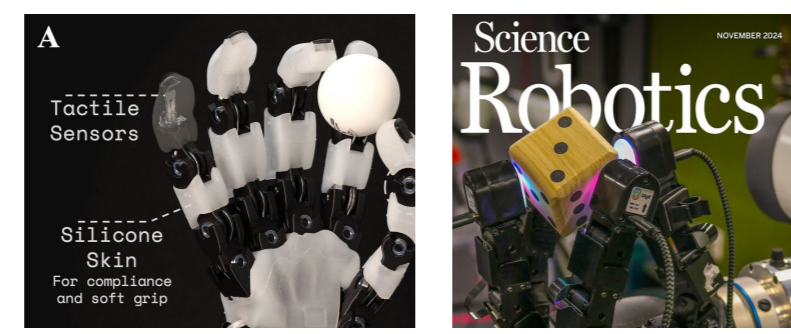
To further complicate things...



Most labs work on unique combinations of sensors and robots, hard to cross-validate



Most results are for in-hand reorientation, leaving the utility of tactile feedback across broader tasks unknown



Lack of standardised tactile-RL benchmarks across tasks and robot morphologies

## New benchmark: roto 2.0

- **5 robot embodiments: 4x hands + 1x arm** (Allegro Hand, ORCA Hand, Shadow Dexterous Hand, Shadow Dexterous Hand Lite, Franka)
- **3 tactile-diverse tasks (and counting):** Find an object, ball bouncing, and Baoding ball rotation
- **! Hyperparameter optimisation !** (essential but often missing component)
- We provide **well tuned baselines** (sweep with 40 trials) that achieve state-of-the-art speeds in sim

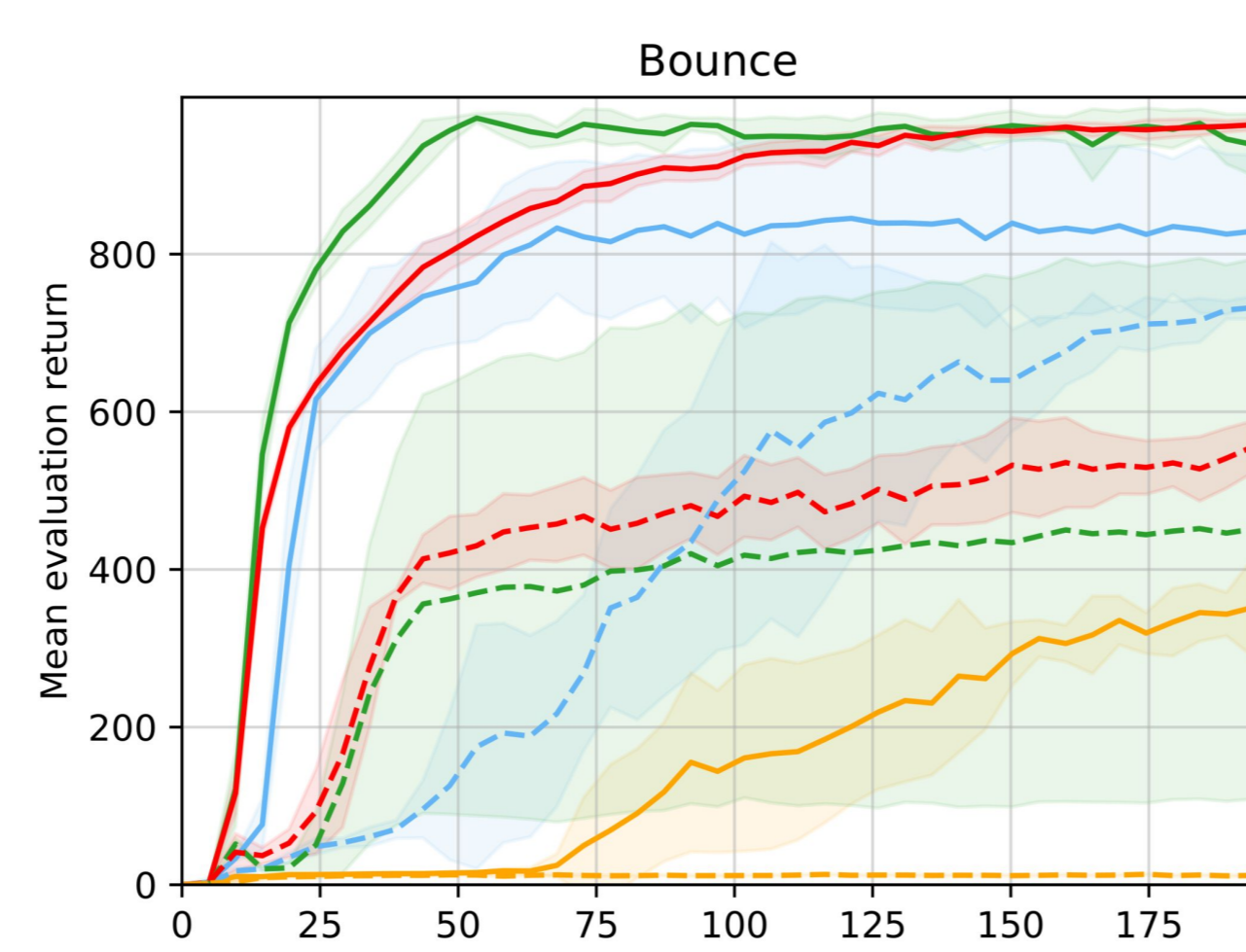
*By open-sourcing our environments and tuned baselines, we hope to reduce the tactile-RL barrier to entry and enable researchers to prioritise fundamental algorithmic challenges over tedious RL tuning.*

## Breakthrough blind performance

Agents: **Blind** (tactile/proprrioception only) & **State-based** (+ object pos/vel)

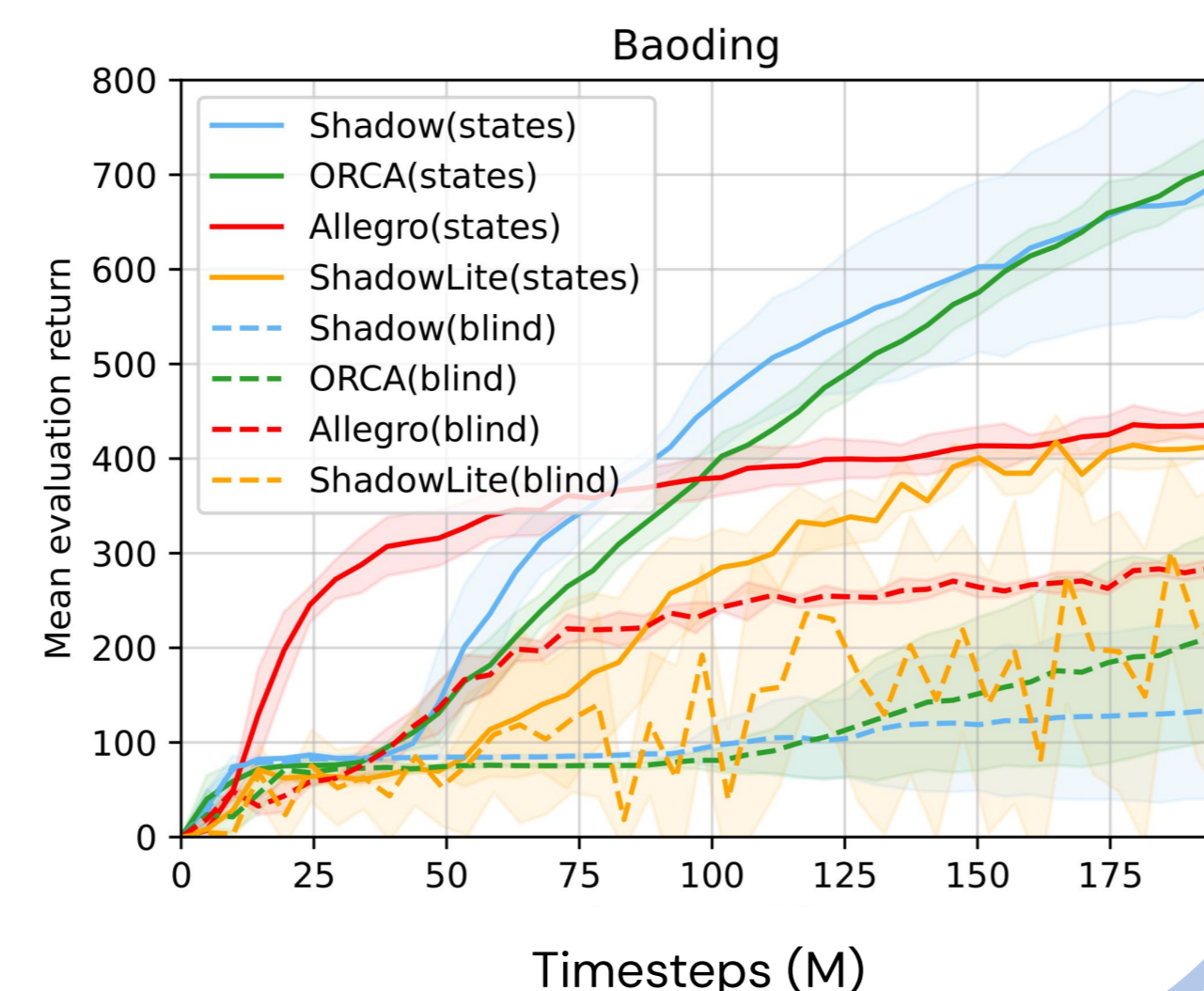
### Bounce

- State-based agents approach the **100-bounce maximum in 10s**; blind agents reach up to **80 bounces** by 200M
- Performance is consistent across morphologies despite unique strategies (except Shadow Lite)



### Baoding

- Large gap in success; state-based hits **35 rotations in 10s**, lucky seeds for blind agents reach max **13 rotations**
- High variance and failure in blind seeds highlight the difficulty of **tactile feature extraction** [1]



## Promising research directions

- **More complex tasks:** While blind policies succeed in simple tasks like bouncing, complex & multi-object manipulation remains an open frontier
- **Richer tactile info & sim2real:** Binary contacts enable sim2real transfer [2]. Future work should focus on bridging the gap for continuous pressure and shear forces
- **Feature extraction:** Developing novel ML methodologies with inductive biases tailored to process tactile signals [1]
- **Scaling sensing area:** Expanding tactile coverage beyond fingertips to the entire hand and humanoid body [3]

We strongly welcome community contributions to roto to accelerate the arrival of the "locomotion moment" for robotic touch.





RoTO is an open-source Reinforcement Learning benchmark environment designed to standardise and promote future research in tactile-based manipulation.

Python  31  6

Star roto on GitHub! 

[github.com/elle-miller/roto](https://github.com/elle-miller/roto)

[1] E. Miller, T. McInroe, D. Abel, O. Mac Aodha, and S. Vijayakumar, "Enhancing tactile-based reinforcement learning for robotic control," in NeurIPS, 2025.

[2] Y. Yuan, H. Che, Y. Qin, B. Huang, Z.-H. Yin, K.-W. Lee, Y. Wu, S.C. Lim, and X. Wang, "Robot synesthesia: In-hand manipulation with visuotactile sensing," in ICRA, 2024.

[3] C. Sferrazza, D.-M. Huang, X. Lin, Y. Lee, and P. Abbeel, "Humanoid-bench: Simulated humanoid benchmark for whole-body locomotion and manipulation," 2024.