Scalable Collaborative Human–Robot Learning (SCHooL) Ken Goldberg, Pieter Abbeel, Anca Dragan, Stuart Russell, UC Berkeley NSF NRI 2.0: Award 1734633: Sept 2017 - Aug 2021



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ERKELEY ARTIFICIAL INTELLIGENCE RESEARCH





- Imitation Learning
- Inverse Reinforcement Learning (IRL)
 - Pioneered by co-PI Russel (1998)
- HRI Optimizing Legibility (Dragan, 2013)
- Cloud Robotics (Kuffner 2015, Kehoe, Abbeel, Goldberg, 2017)

• Apprenticeship Learning (Abbeel & Ng, 2004)

- Learning from Demonstrations

Cooperative IRL

- A CIRL is a 2-player cooperative Markov game $\langle S, (A^H, A^R), P, R_\theta \rangle$
- Human and robot take simultaneous actions, get same reward parametrized by heta
- Human preference R_{θ} initially unknown to the robot
- This incentivizes the human to teach and the robot to learn this preference
- Both agents can reason about the robot's belief state, making it a sufficient state representation (together with the environment state)
- Example: human signals which objects should not be decluttered by replacing them in the environment when the robot removes them

4 Research Objectives

- 1. Extend CIRL Formal Framework:
- 3. Learning Hierarchical Task and Reward Structure
- 4. Bidirectional / Active Human–Robot Communication





$\tau^{\mathbf{H}} \leftarrow \operatorname{argmax} \phi(\tau)^{\top} \theta - \eta ||\phi_{\theta} - \phi(\tau)||^2$

2. Distributed Sensing, Reward Models using Deep Learning



Integrative Application: **Surface Decluttering** To increase productivity and safety in homes, machine shops, warehouses, offices, and retail stores.









Local Severs (eg. 5G):













Robust Task-Based Grasping as a Service

Jingyi Song¹, Ajay Tanwani¹, Jeffrey Ichnowski¹, Michael Danielczuk¹, Kate Sanders¹, Jackson Chui¹, Juan A. Ojea², Ken Goldberg¹

	No Task specified	Lift	Squee Trigg
Task Model			
Grasps			

tasks described on the first row, resulting filtered grasps showed on the third row.

	No Task	Inspect top
	specified	bottom surfa
Task Model		4
$r \in [0.0, 0.25)$		
$r \in [0.25, 0.75]$		
$r \in (0.75, 1.0]$		-



Fig. 3: Task-directed grasping for spray bottle. The stay-out zones of the spray bottle object displayed on the second row model the





Preference learning along multiple criteria: A game-theoretic perspective

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Multi-Principal Assistance Games: Definition and Collegial Mechanisms

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in Interactive Robot Imitation Learning

Daniel Seita¹, Daniel S. Brown¹, Ken Goldberg¹



Artificial Intelligence

A Modern Approach

THIRDEDITION

Stuart J. Russell Peter Norvig

Broader Impacts

"The most important book I have read in guite some time." -Daniel Kahneman, author of THINKING, FAST AND SLOW

Human Compatible

ARTIFICIAL INTELLIGENCE AND THE PROBLEM OF CONTROL

Stuart Russell

by Blooma Goldberg, Ken Goldberg, and Ashley Chase illustrated by Dave Clegg

I'm Blooma. I love art, science, math, dancing, basketball, and roller skating, but my favorite activity is inventing new things... especially robots! And I'm super excited to tell you how I figured out the best way to train a robot.

How to Train Your Robot

How to Train Your

ブルーマ・ゴールドバーグ ケン・ゴールドバーグ アシュリー・チェイス作

How to Train Your Robot

video based on the book: How To Train Your Robot by B. Goldberg, K. Goldberg, and A. Chase, illustrated by D. Clegg (with support from NSF and UC Berkeley's Lawrence Hall of Science)

Info and Resource Page: (Google Doc) https://bit.ly/How-To-Train-Your-Robot-Info

To view subtitled Spanish, Japanese, Hindi, and Chinese (Simplified), just click the small blue CC (closed caption) button on the right.

15 min Video Nov 2020 Subtitled in Spanish, Japanese, Hindi, and simplified Chinese

https://vimeopro.com/citrisproductions/how-to-train-your-robot

