Understanding Human Teaching Modalities in Reinforcement Learning Environments

A Preliminary Report

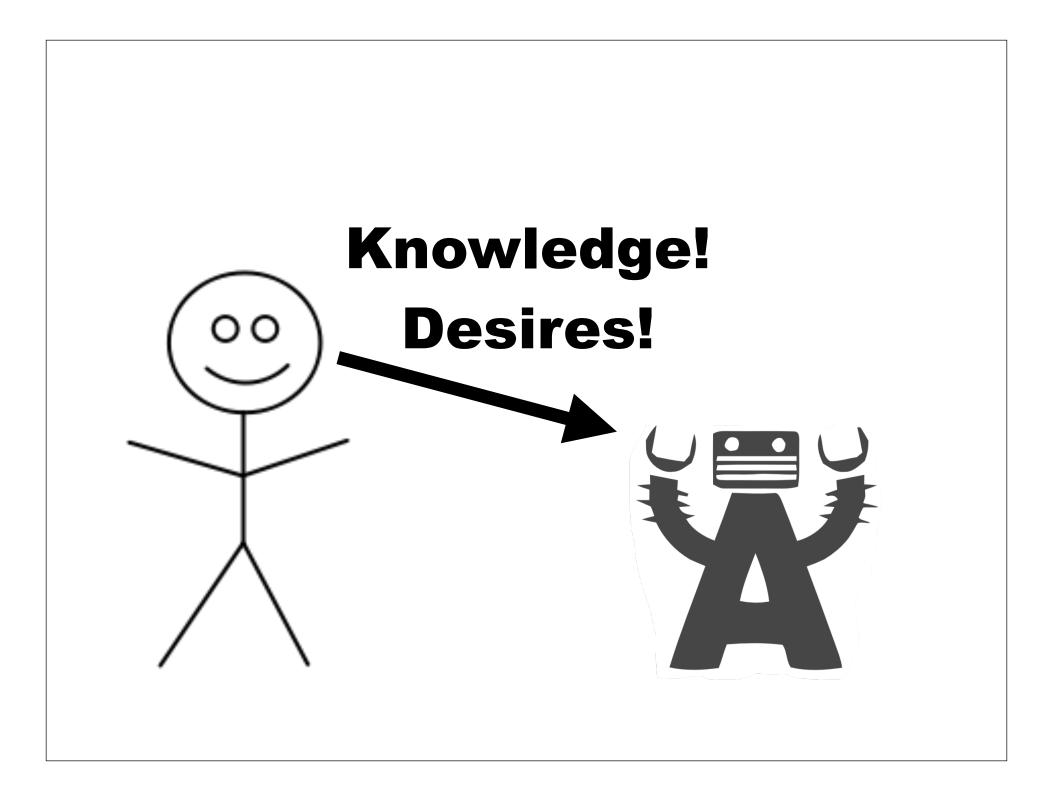
Slides available on the Program page of the ALIHT website.

W. Bradley Knox and Peter Stone

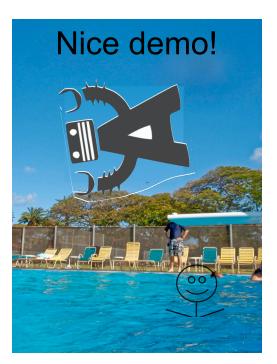


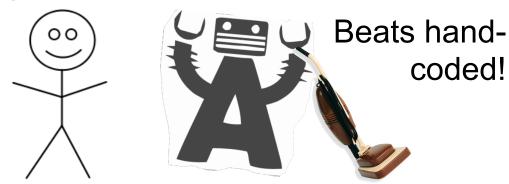
Matthew E. Taylor

LAFAYETTE COLLEGE

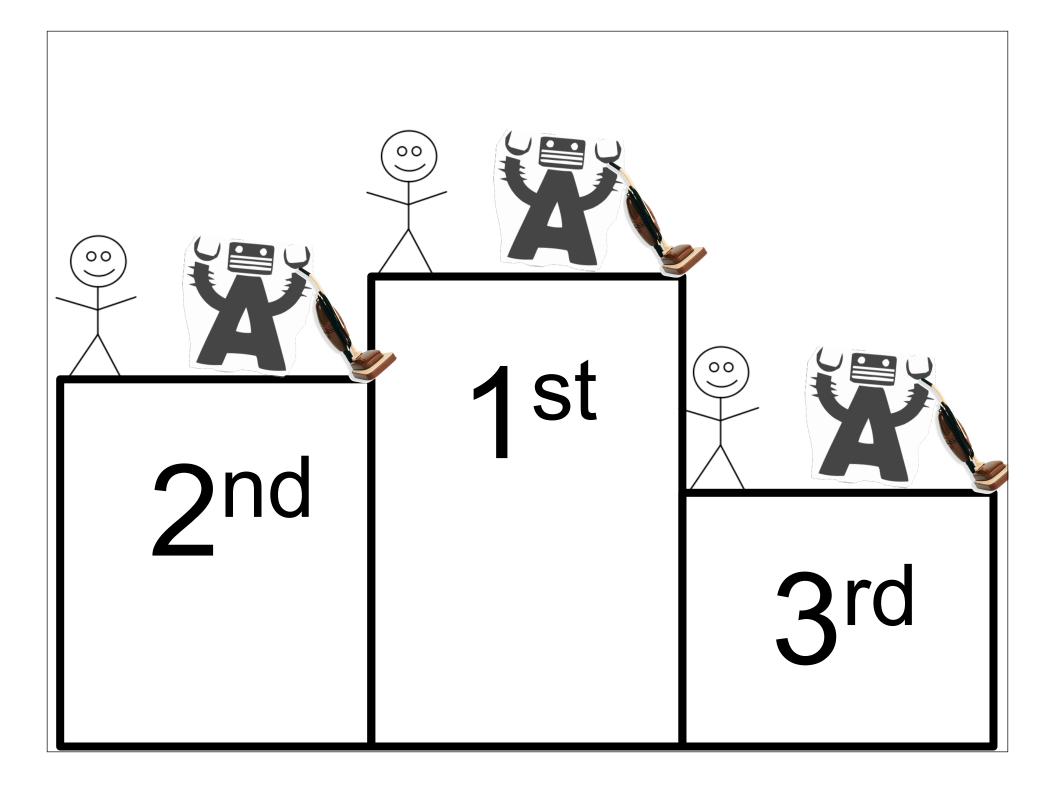


Current state of interactive learning evaluation



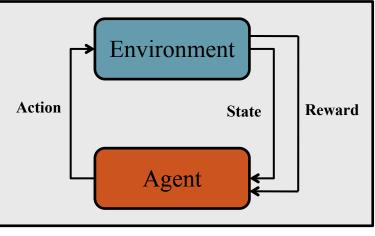






Reinforcement learning tasks

- Learn from limited feedback
- Delayed reward
- Very general
 - Possibly slow learning
- Human end-user cannot determine correct behavior



Learning from demonstration (LfD)

- Goal: reproduce behavior / policy
 - generalizing effectively to unseen situations
- Argall, Chernova, Veloso and Browning. A Survey of Robot Learning from Demonstration. RAS, 2009.





Grollman & Jenkins





Nicolescu & Matarić

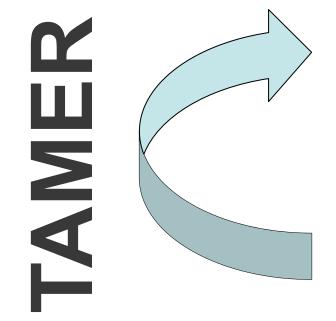
Argall, Browning & Veloso

Lockerd & Breazeal

TAMER

Key insight: trainer evaluates behavior using a model of its long-term quality

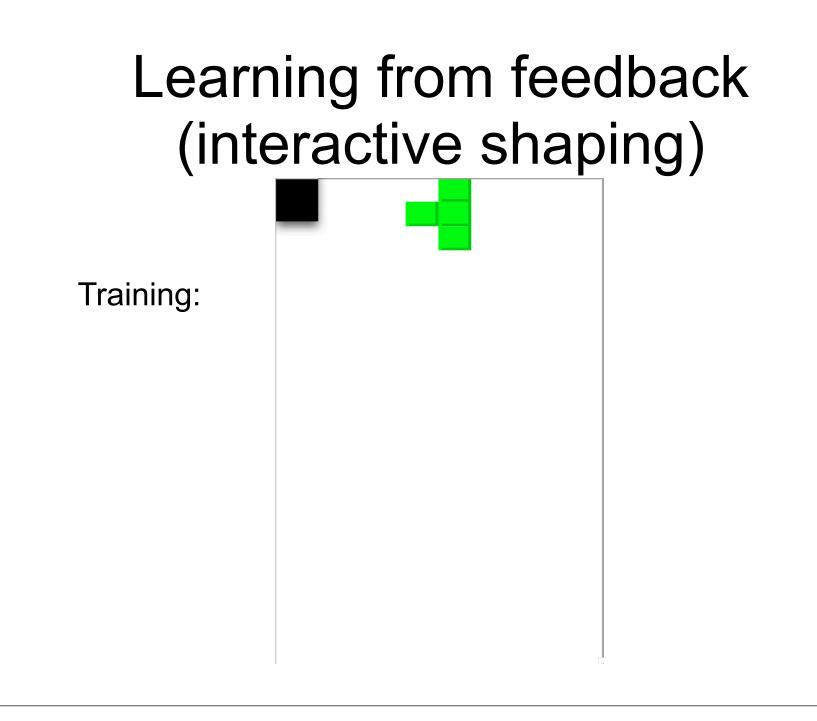
Knox and Stone, K-CAP 2009



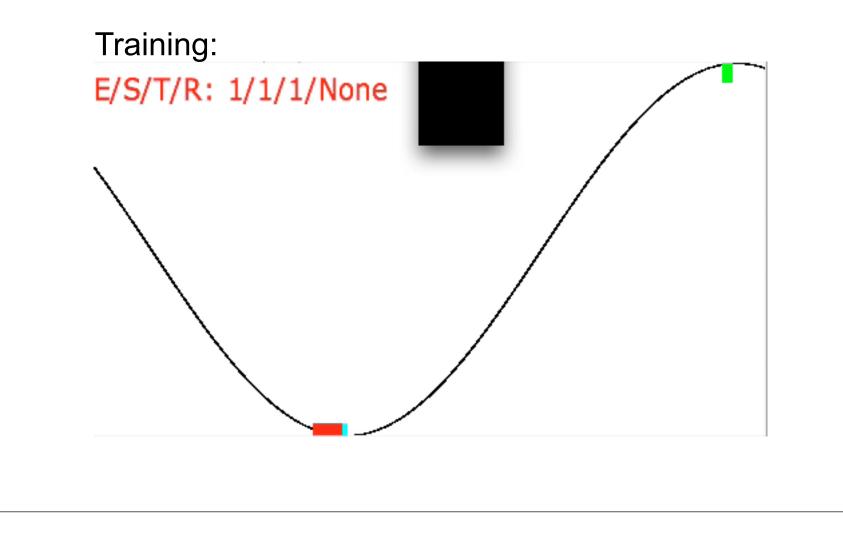
Learn a model of human reinforcement $H: S \times A \rightarrow \mathbb{R}$

Directly exploit the model to determine action If greedy: $action = argmax_a \hat{H}(s, a)$

Knox and Stone, K-CAP 2009



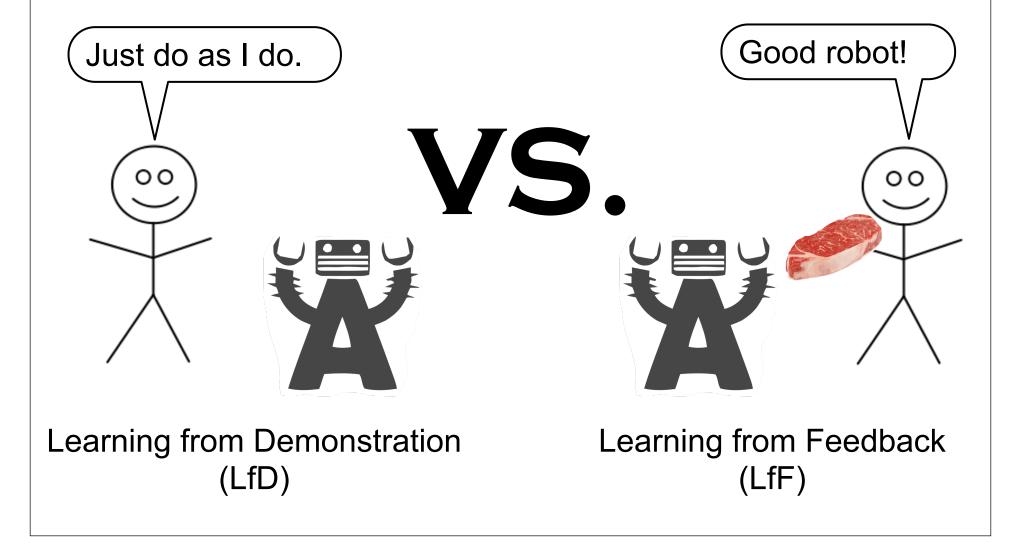
After training:



LfD and LfF vs. RL

- Noisy
- Limited by human ability
- Requires human's time
- Faster learning
- Empowers humans to define task

And out come the contendas!!



Demonstration more specifically points to the correct action

Interface

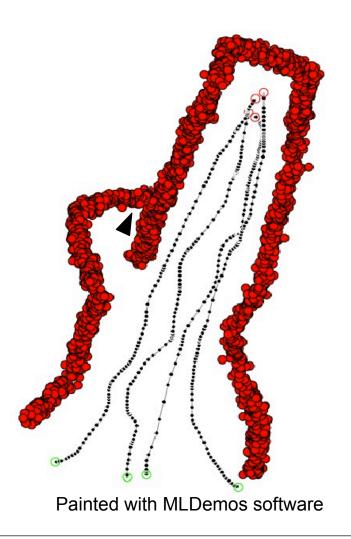
- LfD interface may be familiar to video game players
- LfF interface is simpler and task-independent



Expression of learned model during training: LfF? yes.

LfD? generally no.

- LfD better initial training performance
- LfF can observe and address model's weaknesses
- LfF training and testing performance match up better



Task expertise

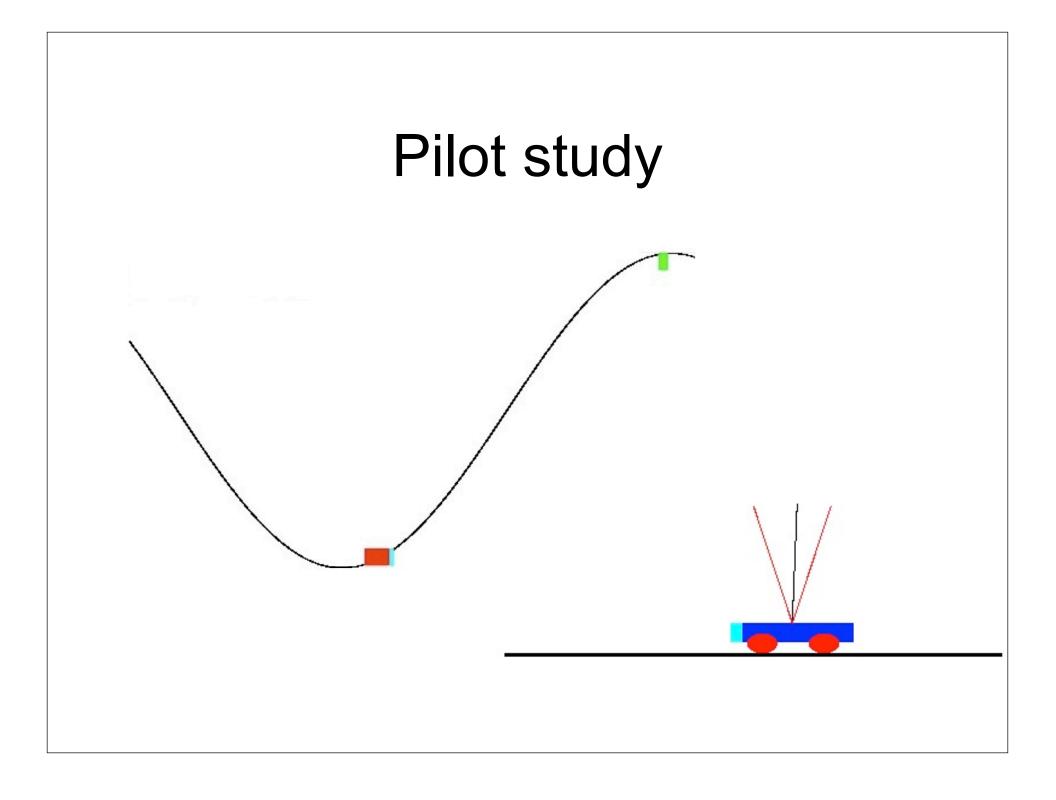
- LfF easier to judge than to control
- Easier for human to increase expertise while training with LfD



Cognitive load - less for LfF

General hypothesis

LfD generally performs better, but situation-dependent



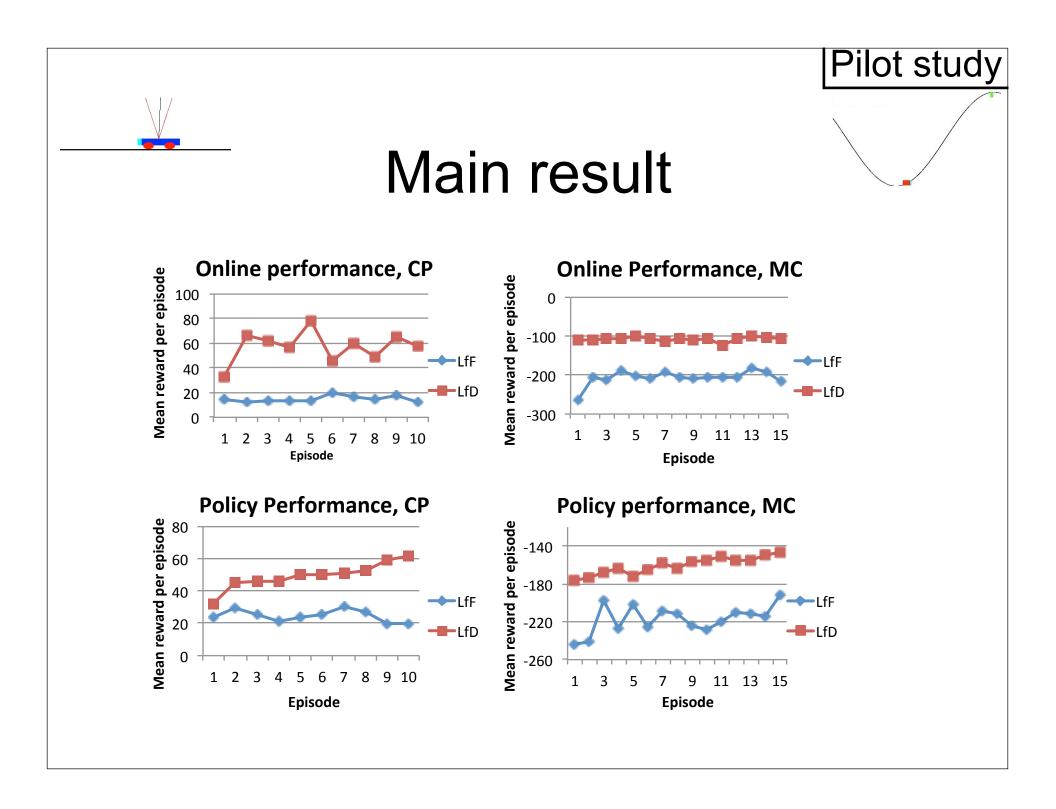
16 undergraduates

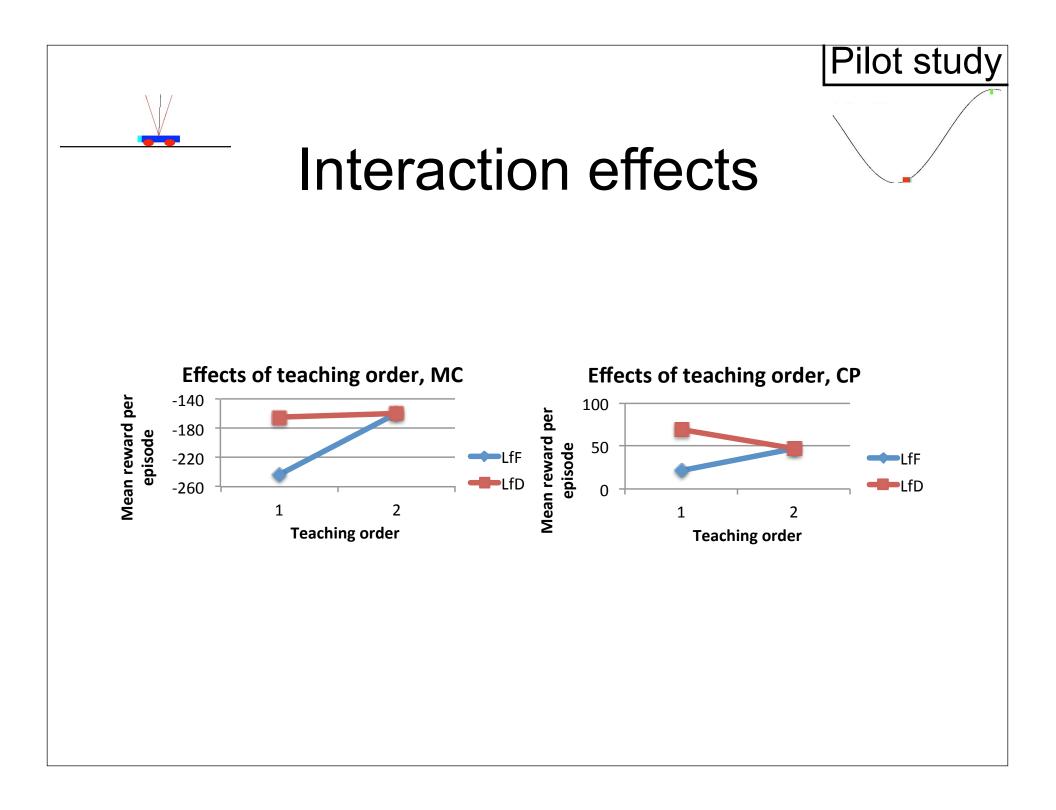
Cart Pole first, then Mountain Car

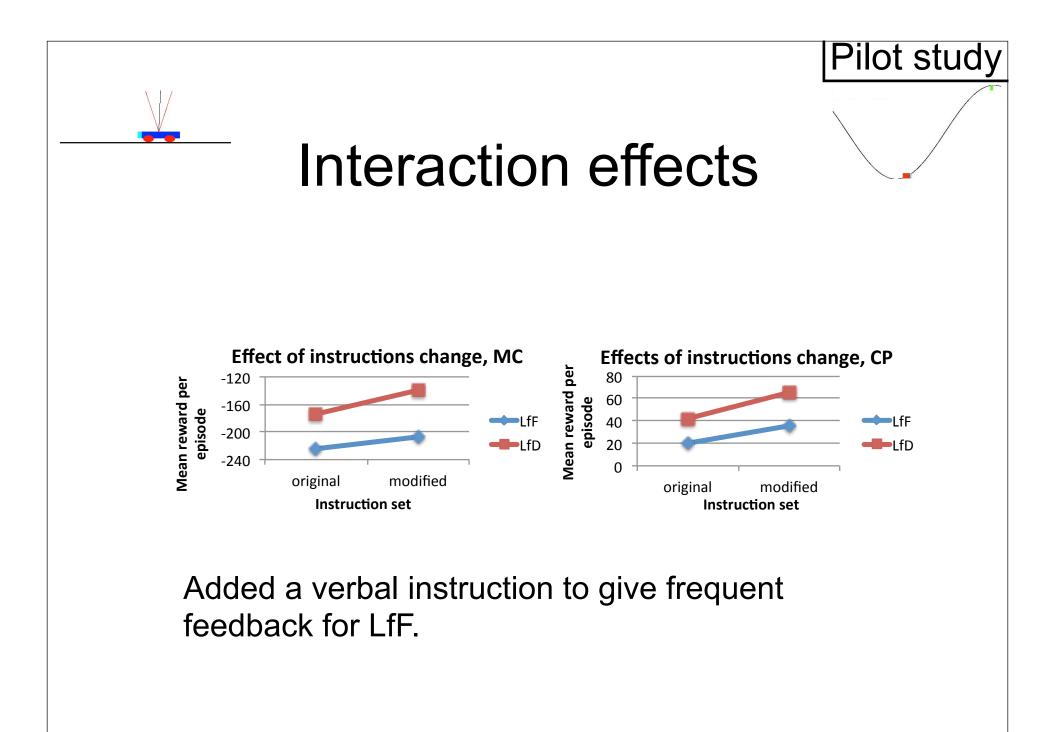
- Practice and test rounds
- Randomized: LfF or LfD first
 - Unbalanced result: LfF was first for 87.5% of CP and 69% of MC

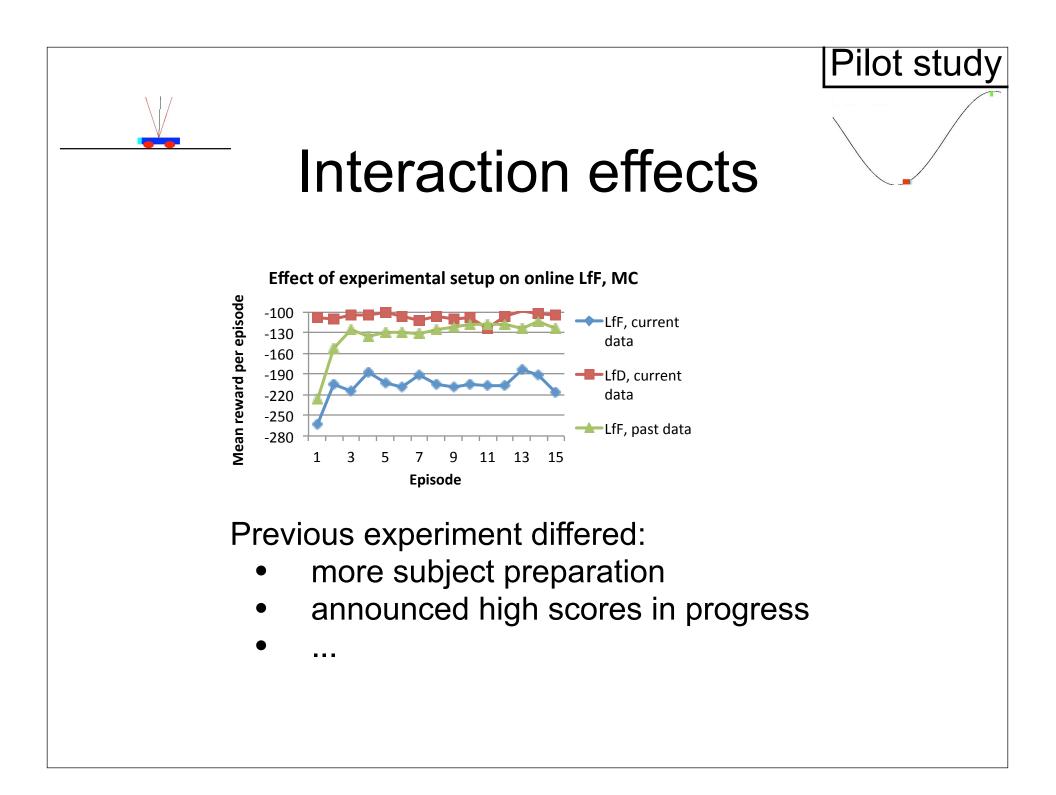
Keyboard interface

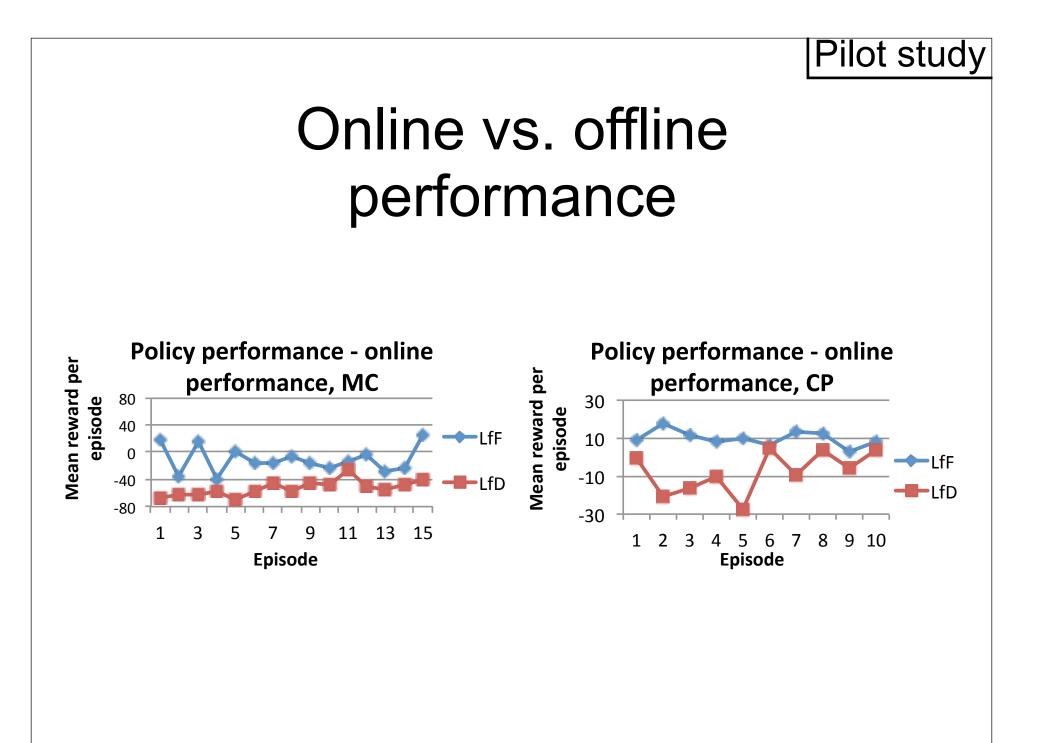
- LfD: j, k, l
- LfF: z, /











Tentative takeaways from performance comparisons

LfD was better in our experiments.

But both were sensitive to the experimental setup.

Tentative takeaways from performance comparisons

Subjects need more preparation for LfF.

- With *zero* task expertise, LfD still allows learning on the job
- LfF vs. LfD interfaces

Tentative takeaways from performance comparisons

LfD's offline, learned performance is generally worse than its training samples.

LfF's offline, learned performance is generally as good or better than during training.

To conclude,

Results

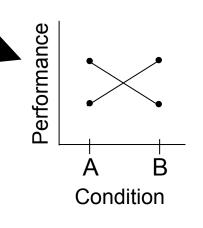
- LfD was better.
- But performance was situational.
- LfF needed more subject preparation.
- LfF models compared better to training performance.

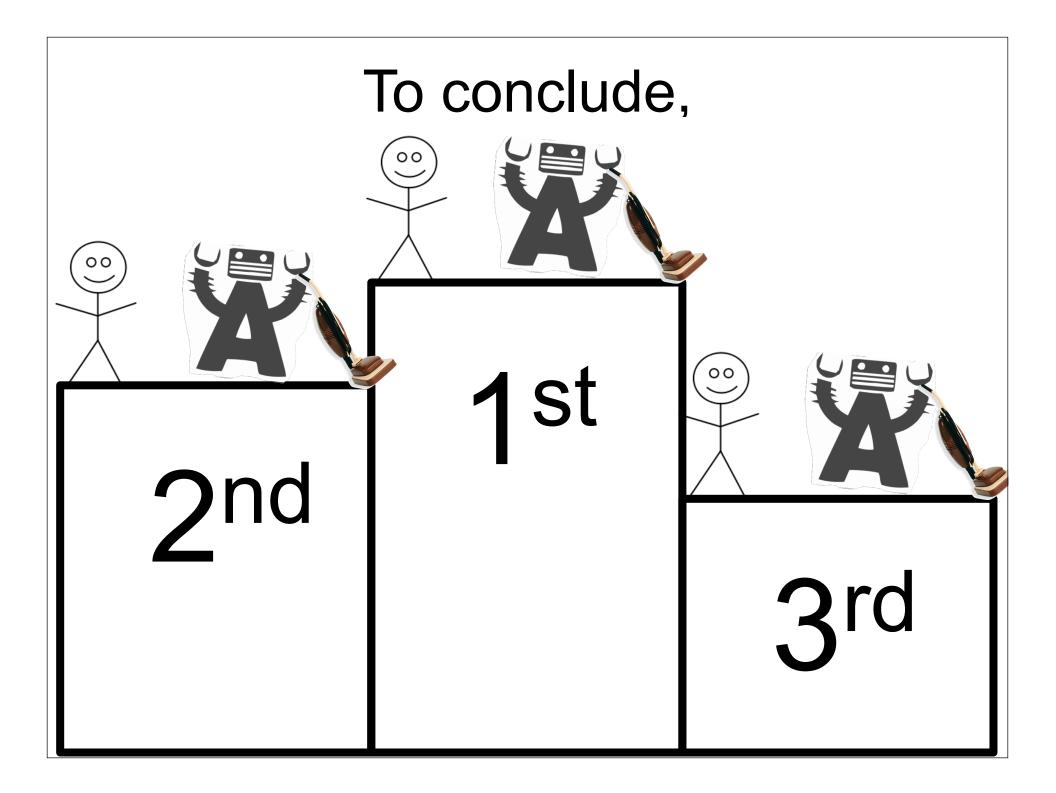
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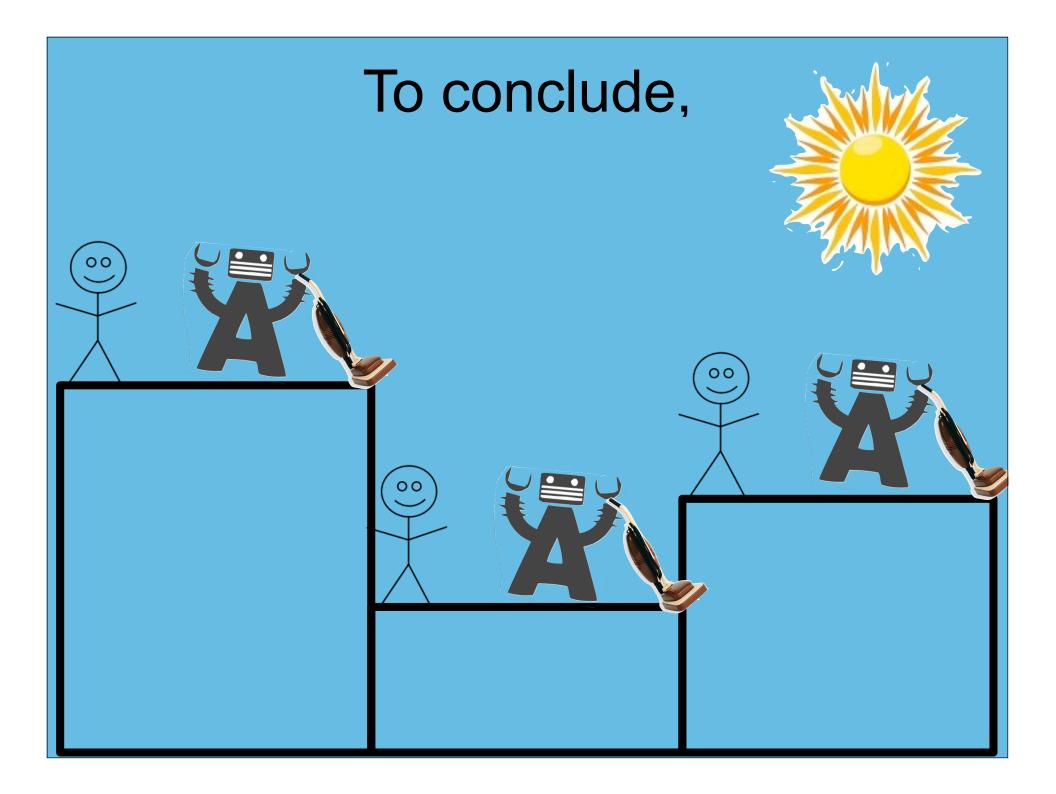
Near future work

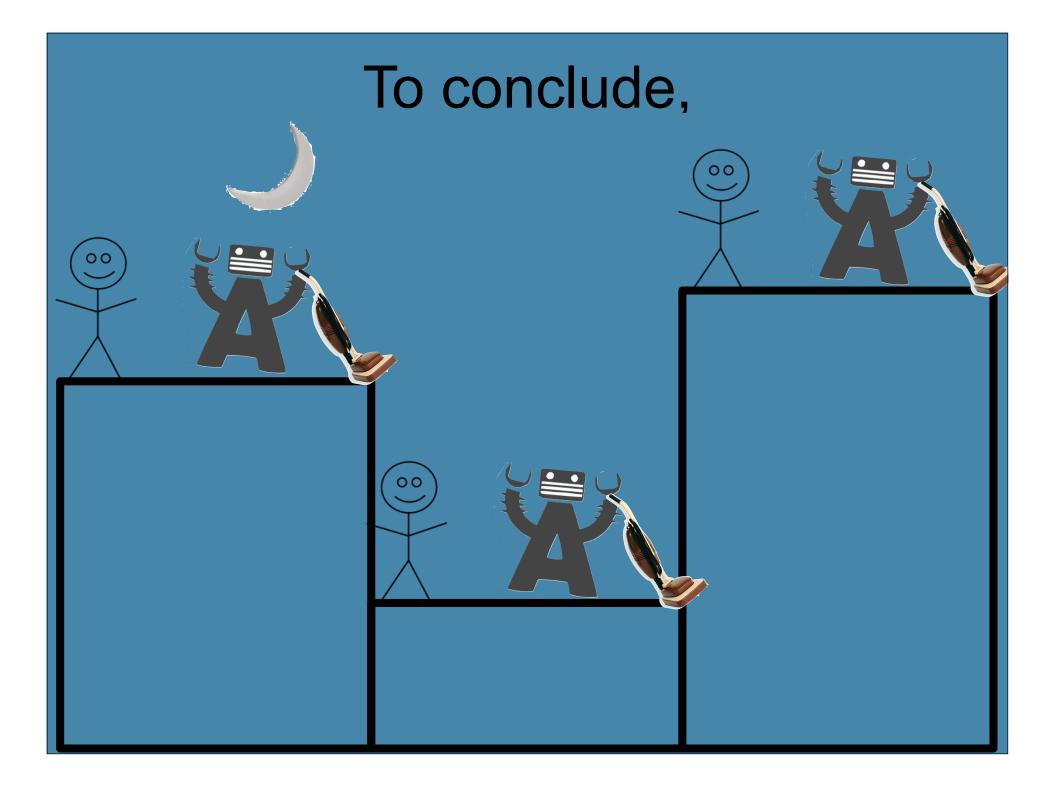
- More subjects
- More balanced conditions
- More interesting manipulations (e.g., model representation and control interface quality)
- Aim for crossover interactions,











To conclude, 8 A WINNER IS YOU