



Matthew E. Taylor (Matt)

Reinforcement Learning in the Real World: Challenges and Opportunities for Human Collaboration

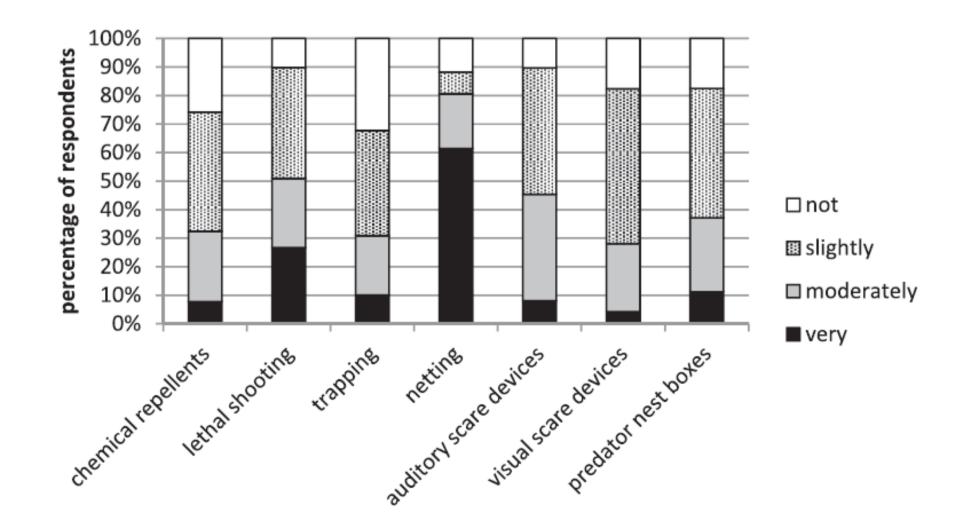




The Intelligent Robot Learning Laboratory

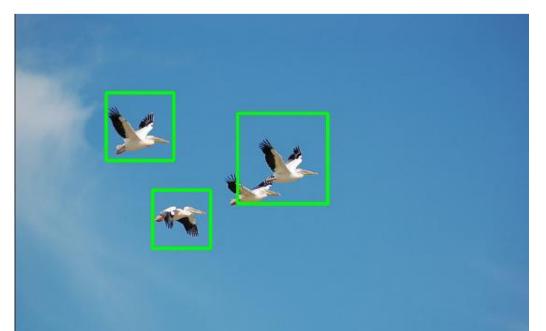
Bird Deterrence

- High-value agriculture
- Annual bird damage in WA ~\$80 million!
- Low-tech approaches
- We measured bird damage with and without UAVs









Bin Dog

- Controlling a novel "bin-dog"
 robot to autonomously traverse
 an apple orchard, ~800 lbs
- Coordinate with human pickers
- Mixed-reality testing



Reinforcement Learning (RL)

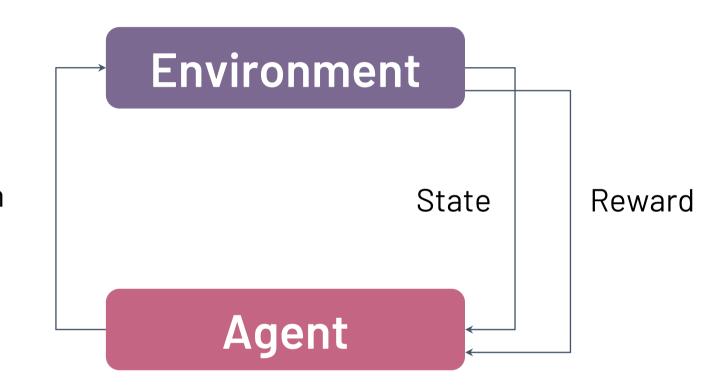
No labels: agent never told right or wrong

Agent interacts with environment (simulator or real world)

Action

Typically can gather data, possibly at cost, by interacting with environment

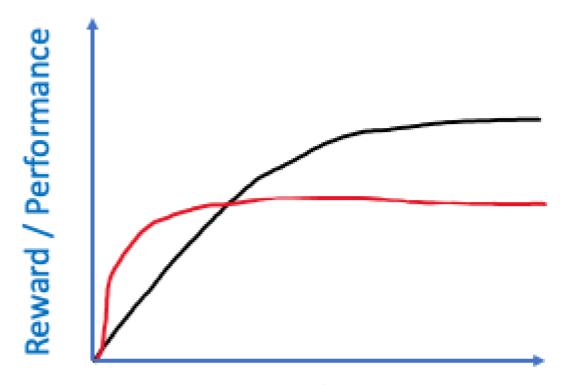
Learns via exploring vs. exploiting



RL Goals

Learn to maximize real-valued reward signal

- With maximal final performance
- With little data
- Reducing human effort
- Discovering novel solutions
- Handling non-stationary environments

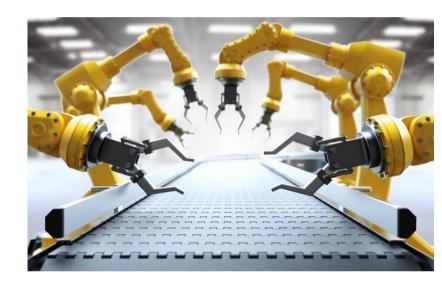


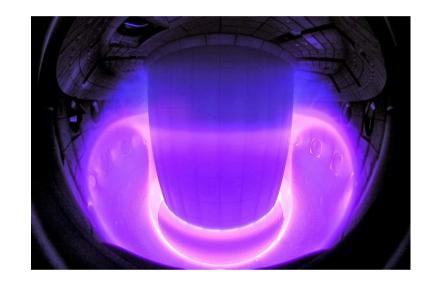
Time / Data

RL Applications

(Un)Supervised learning performs well for many real-world applications







Dota







AlphaGO





Fusion: Tokamak Stock Trading Plasmas



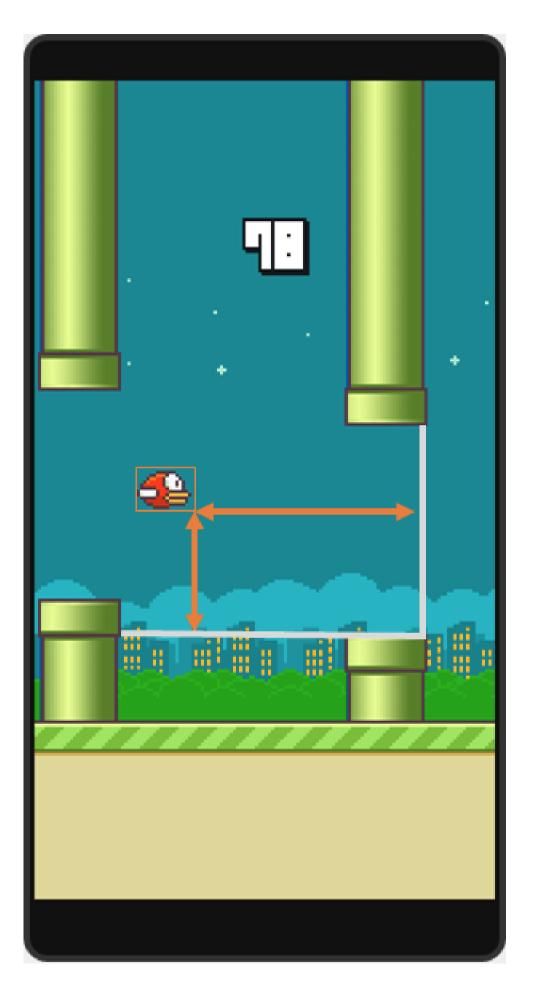
<u>https://www.youtube.com/watch?v=0Jw4HTWvGdY</u>

Example 1: Flappy Bird

Transition function: Controlled by game Action?

Reward?

State representation?



http://sarvagyavaish.github.io/FlappyBirdRL/

Example 2: Aiden

Optimal Order Execution

State: Info about stock & market

Actions: Do nothing, buy/sell a little, buy/sell a lot

Rewards: Based on VWAP (Volume-weighted average price)

Transition function: Stock market (real or simulated data)



https://www.borealisai.com/en/applying-ai/aiden/

Ideas for Sequential Decision Processes?

Advantages of AI & Autonomy



Advantages of AI & Autonomy

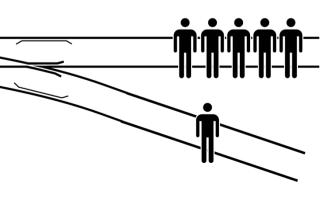


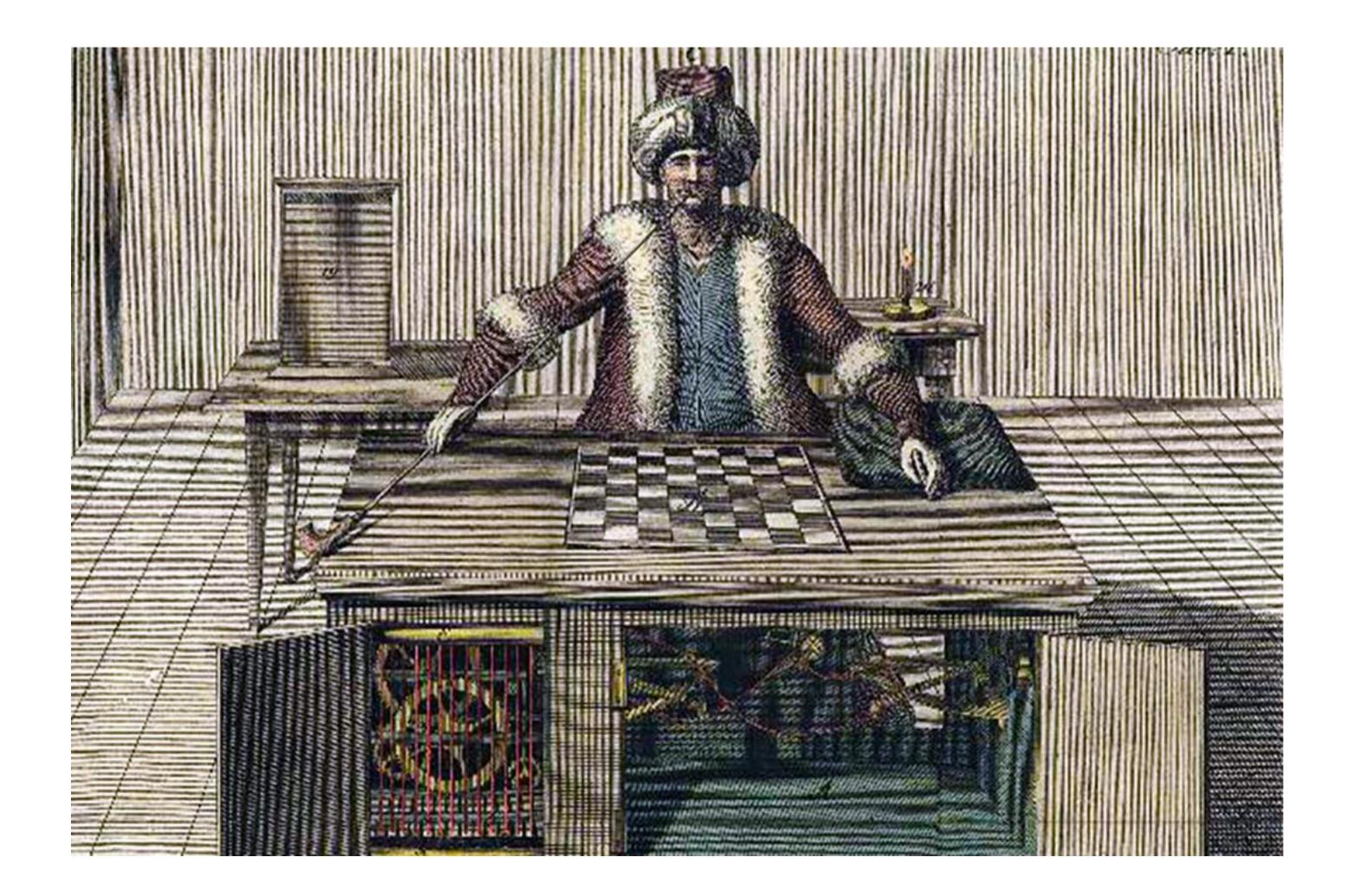


Potential Disadvantages













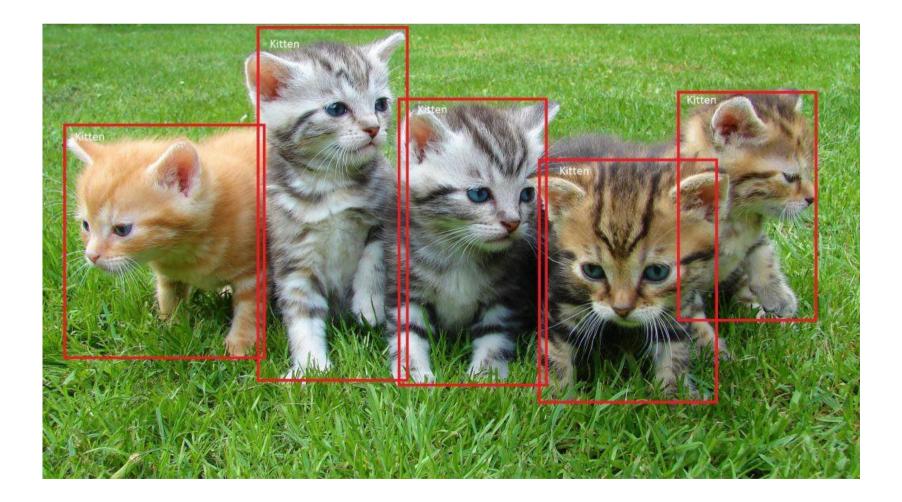
- O Cat
- Not Cat
- O Maybe/NotSure



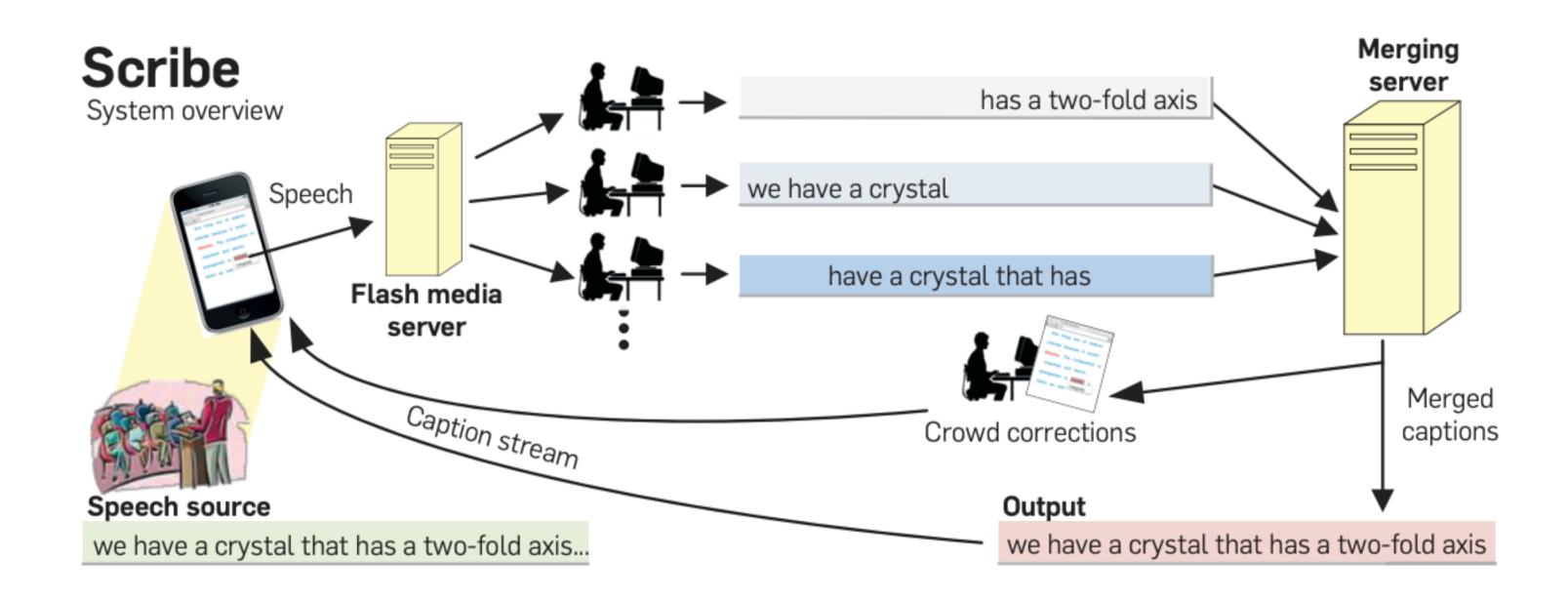
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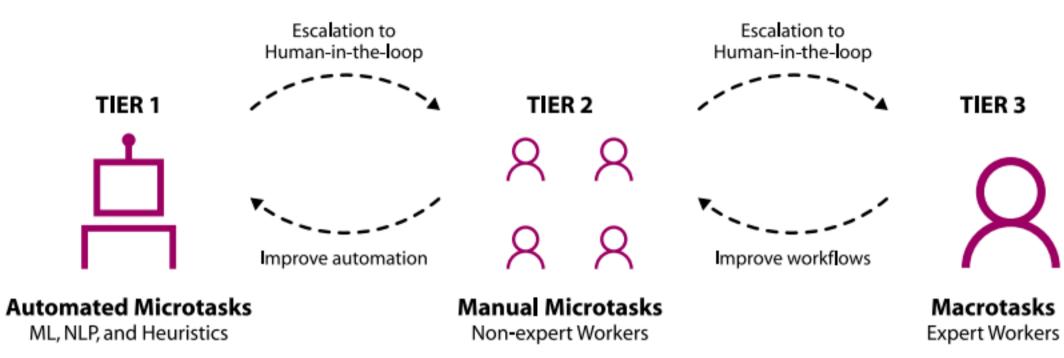
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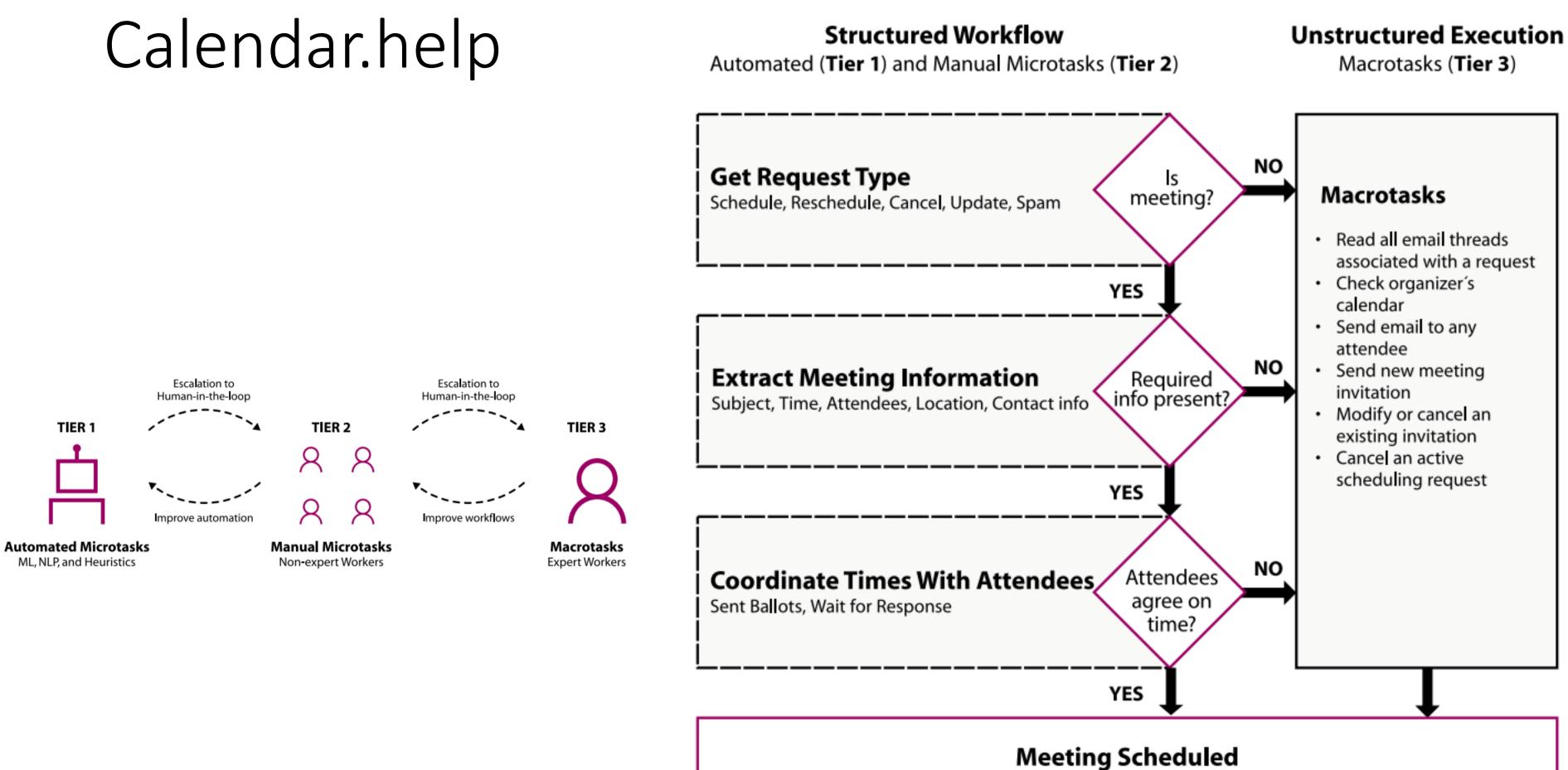
Legion & Scribe



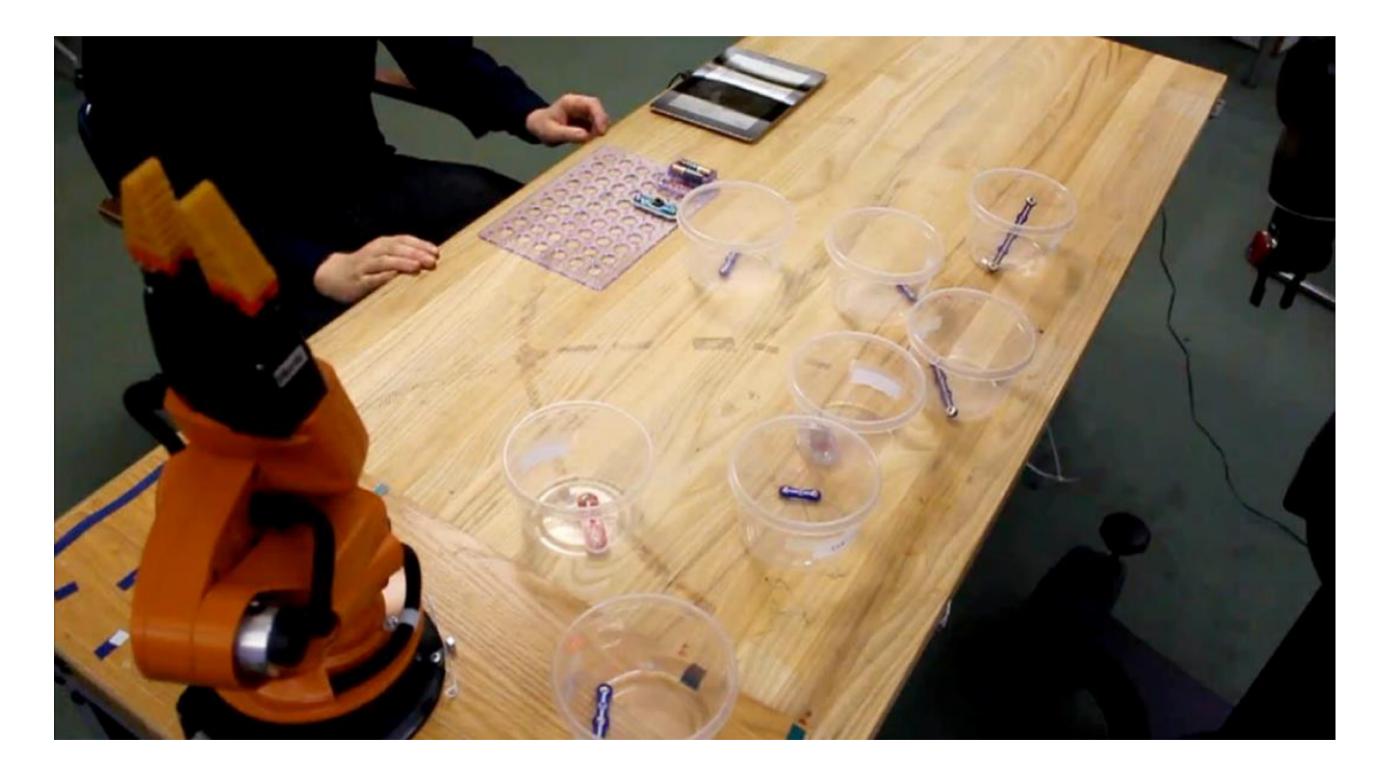
Calendar.help



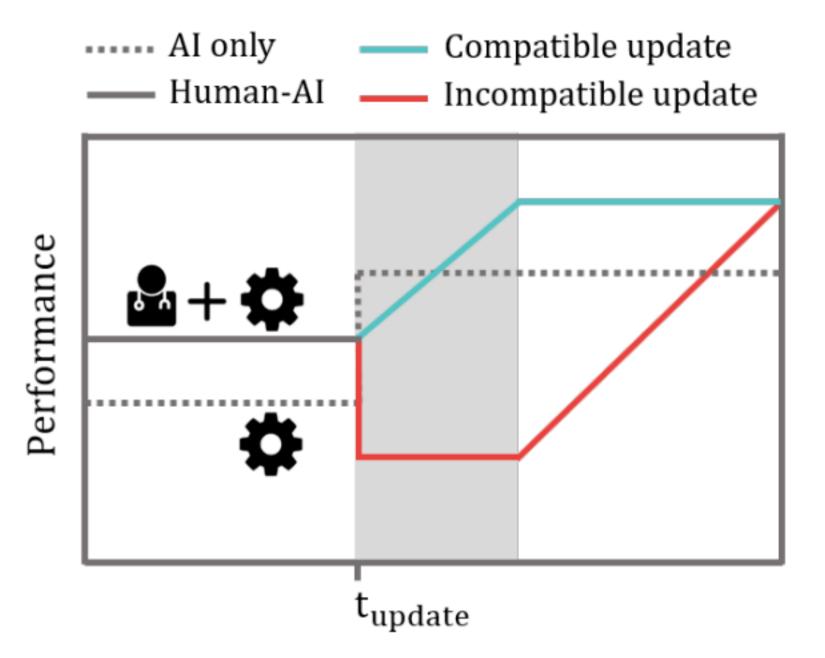




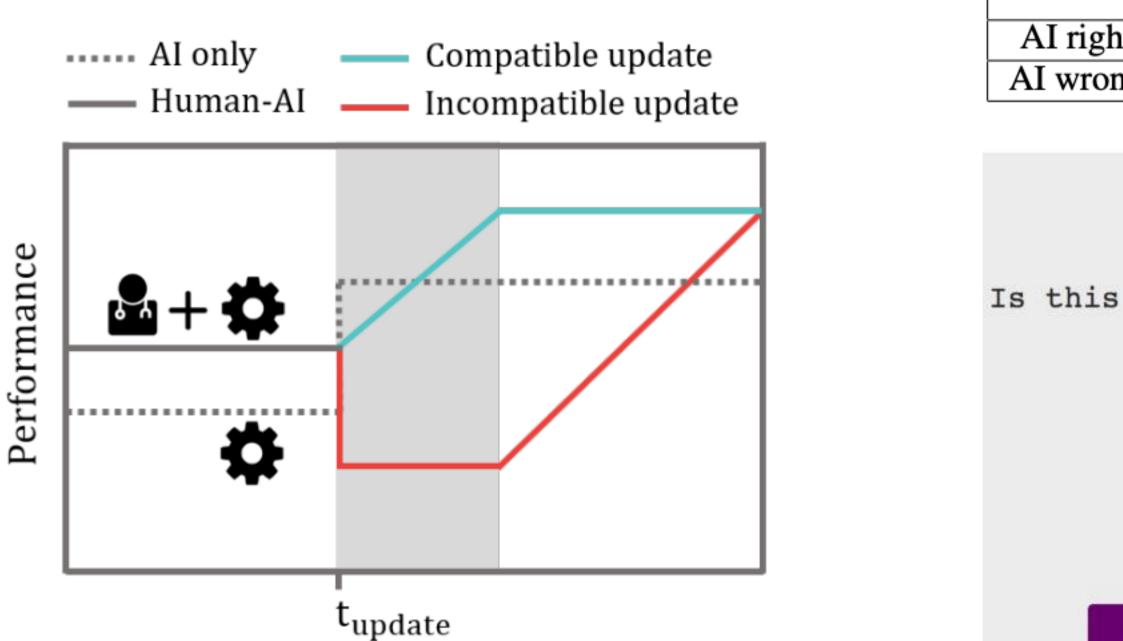
Robotics Teamwork



Human-Al Teams: Performance/Compatibility Tradeoff



Human-Al Teams: Performance/Compatibility Tradeoff



	Accept	Compute
ht	\$0.04	0
ng	-\$0.16	0



Is this object defective?

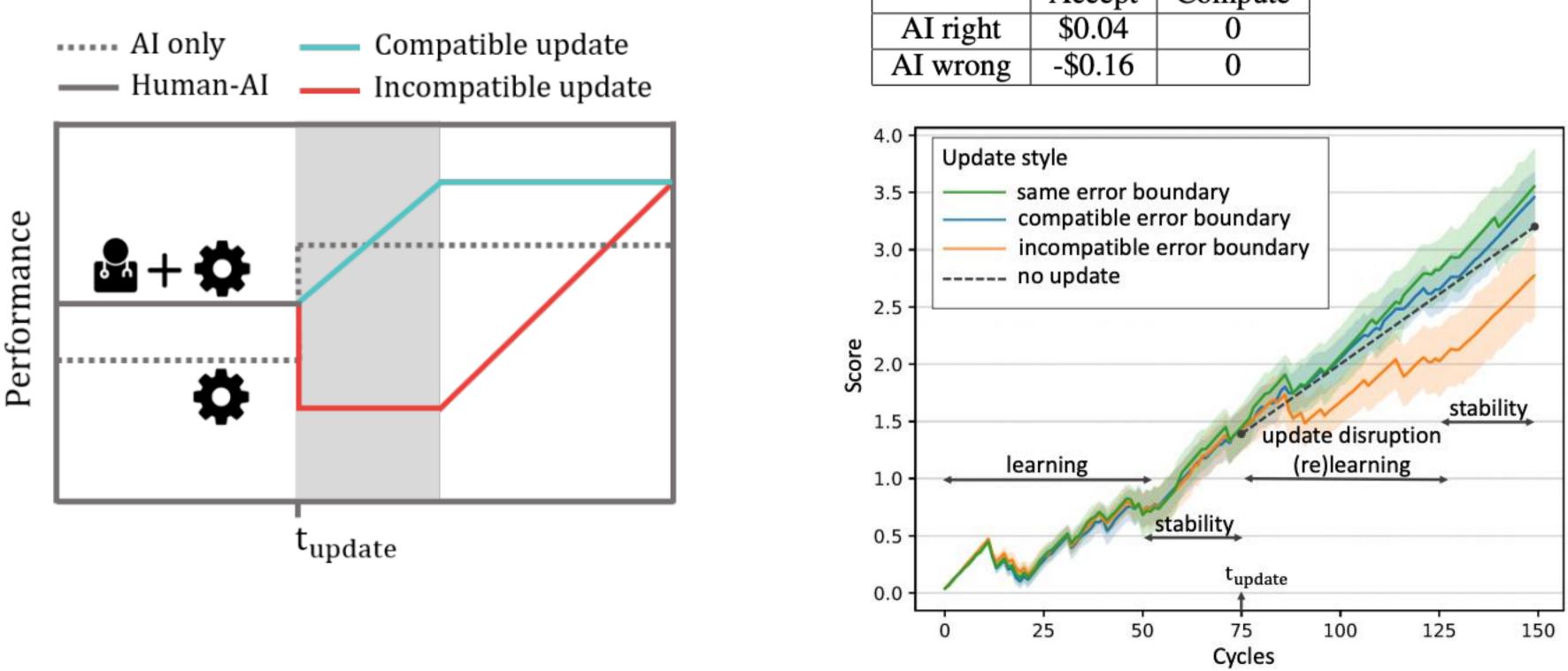


Features in the object:

Feature Value

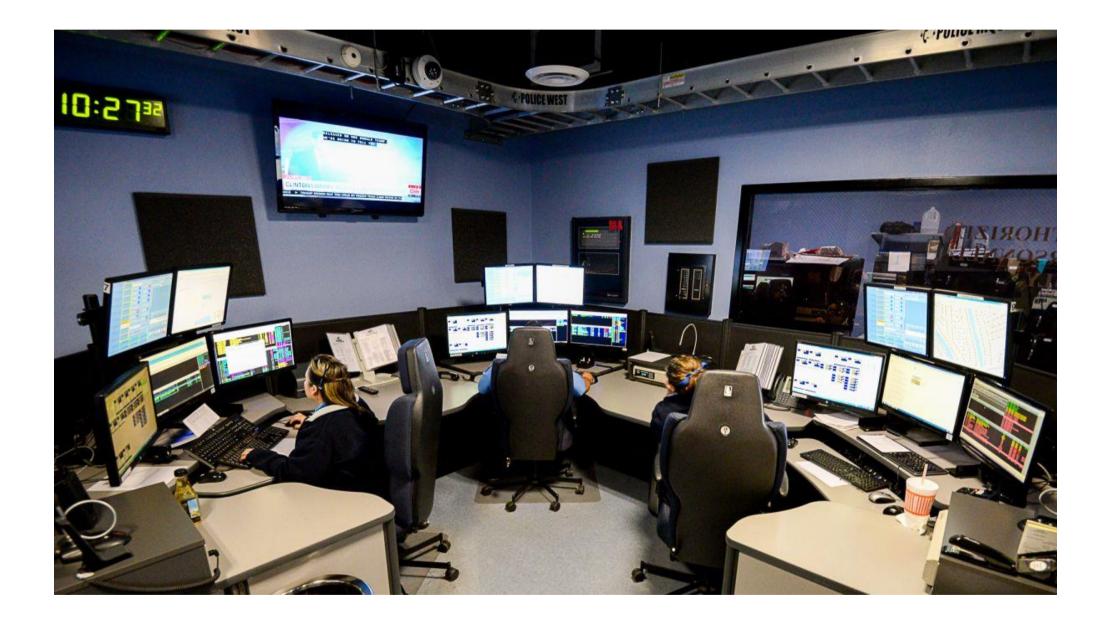
color	blue
shape	circle
size	small

Human-Al Teams: Performance/Compatibility Tradeoff



	Accept	Compute
ht	\$0.04	0
ng	-\$0.16	0

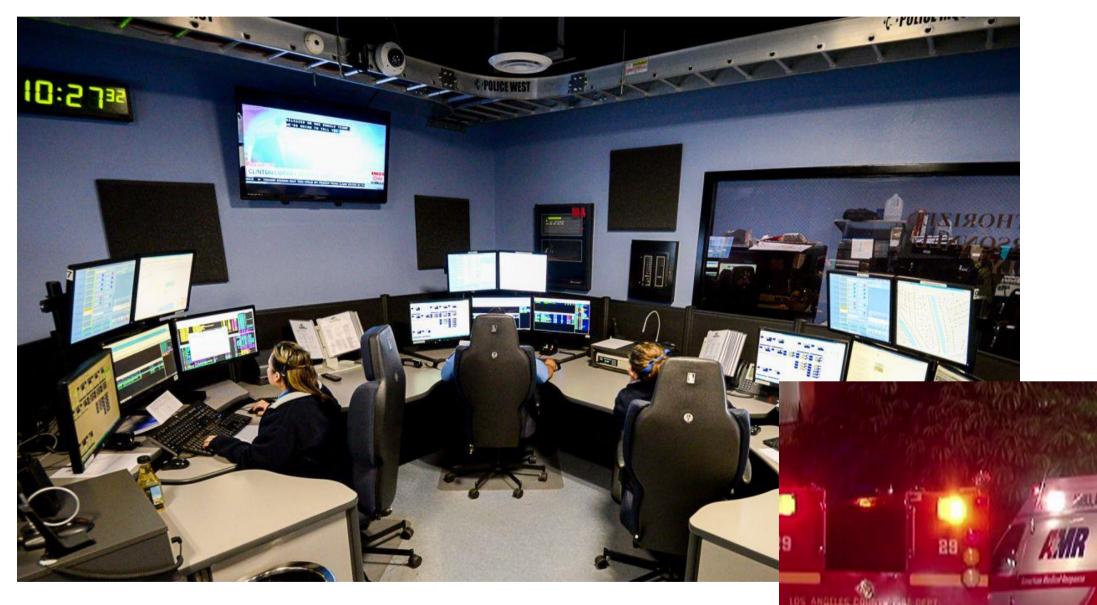
Emergency Vehicle Dispatching







Emergency Vehicle Dispatching





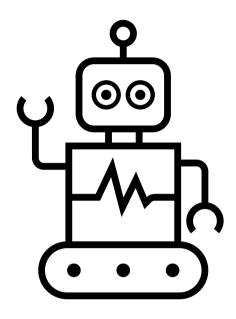




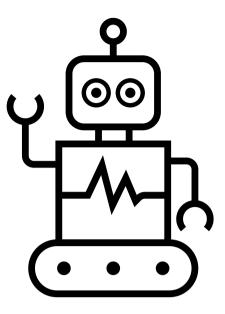
Question 1: What are the right RL problems?

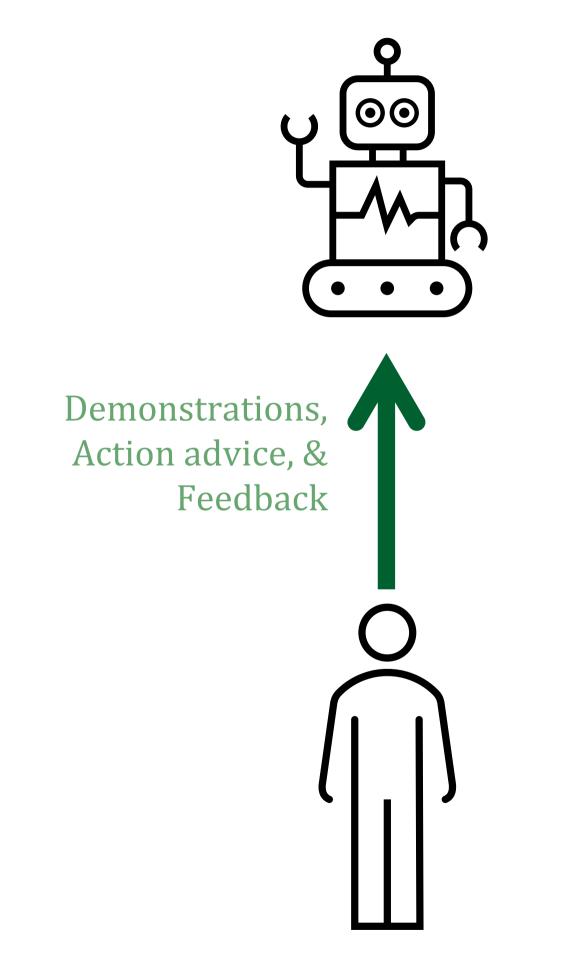
- Sequential decision task
- High impact
- Robust to exploration

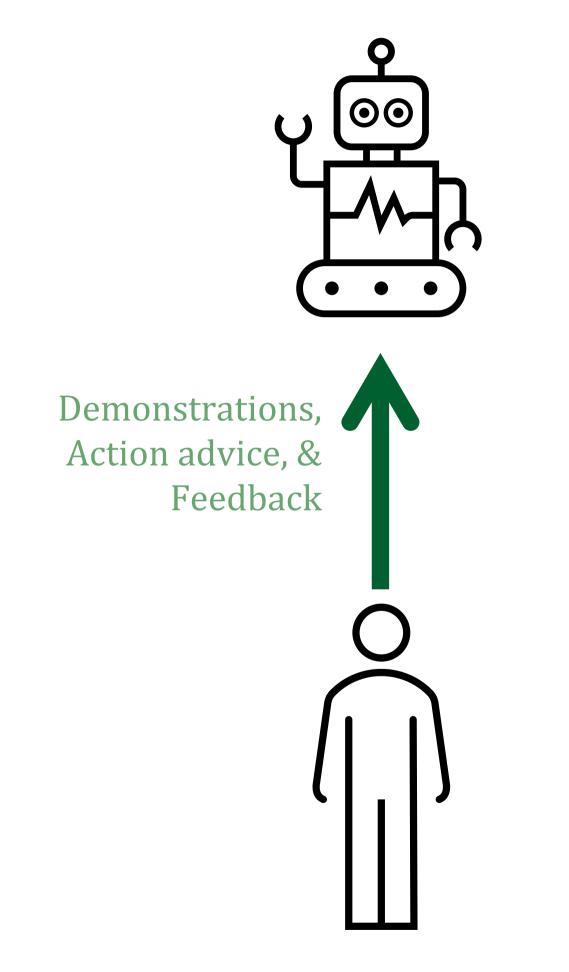


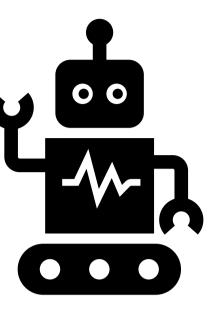


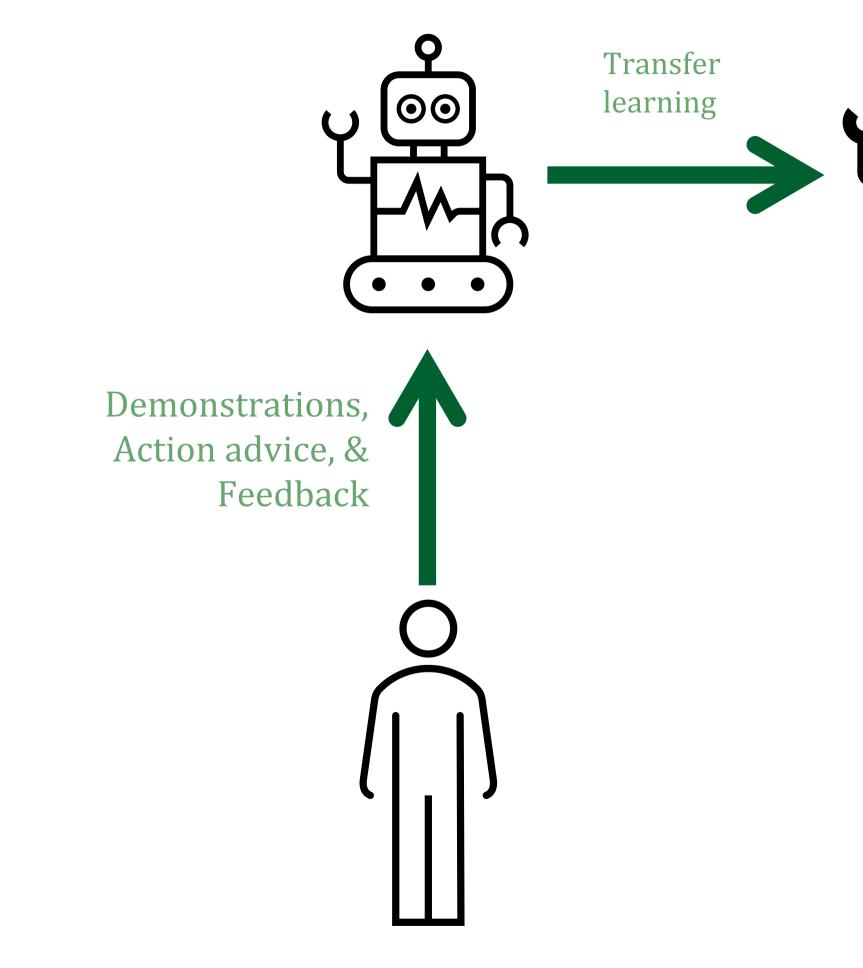


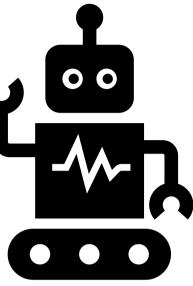


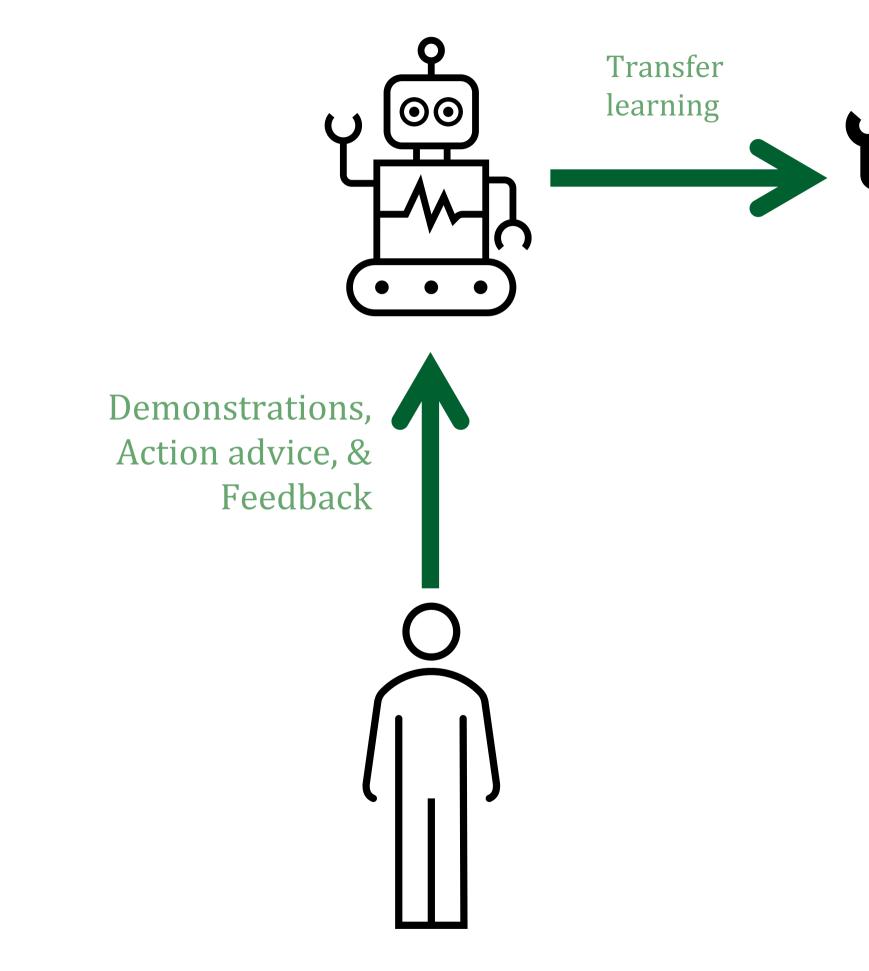


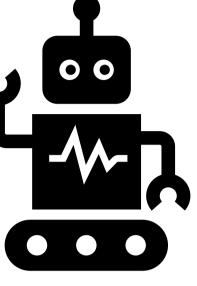




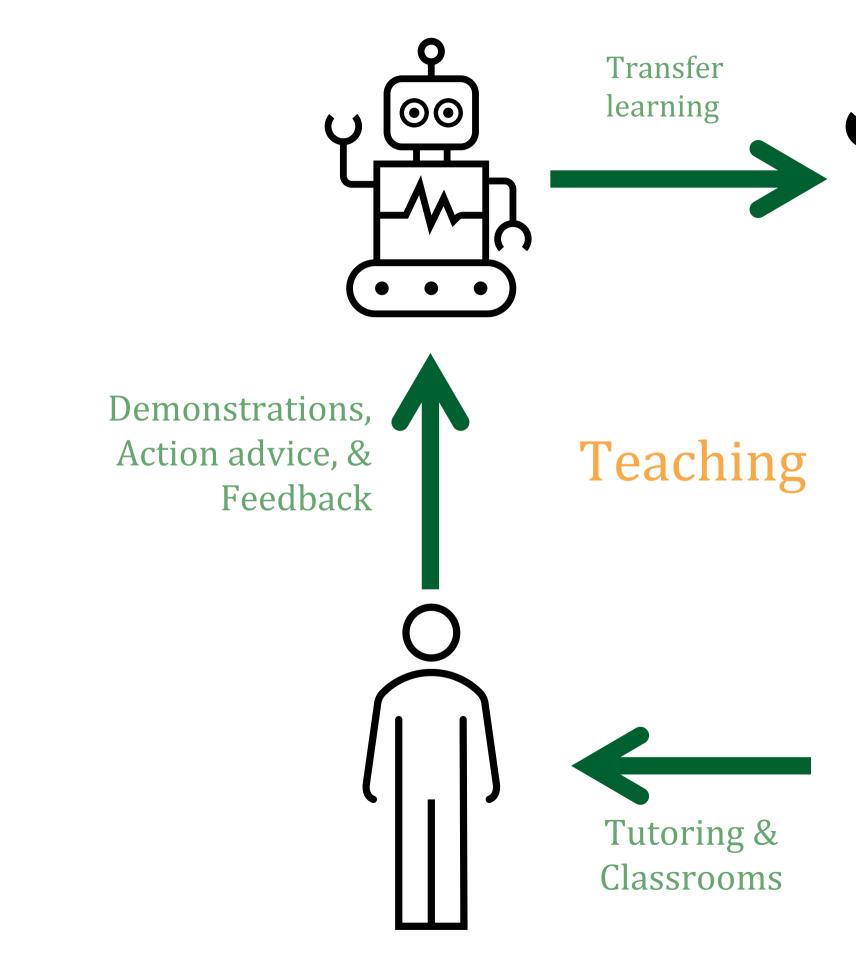


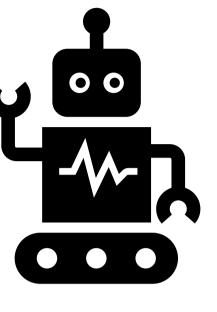




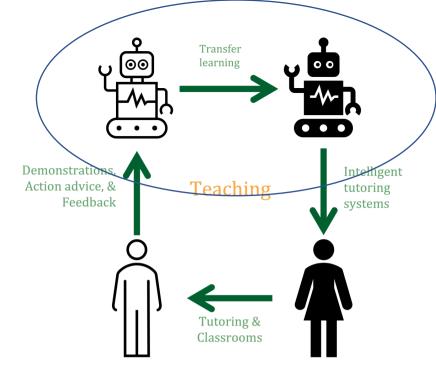


Intelligent tutoring systems

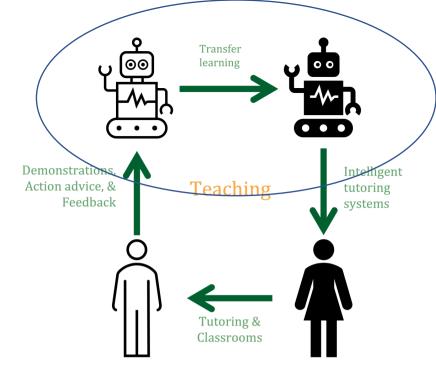




Intelligent tutoring systems



Transfer Learning



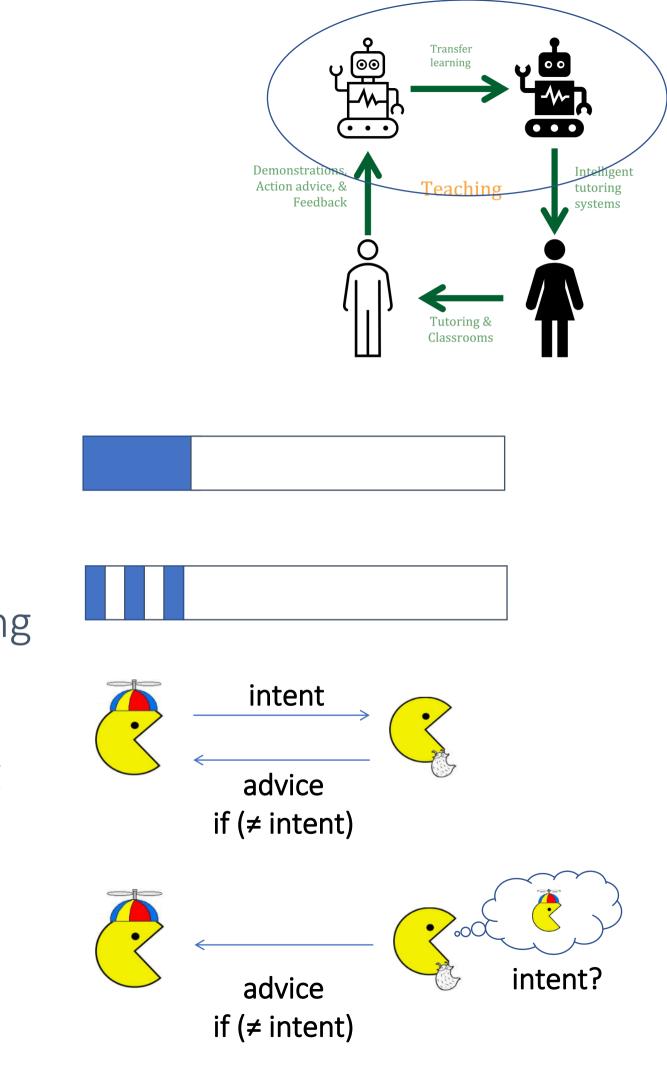
- Transfer Learning
- HAT \leftarrow Teacher Decides

Early advising

Importance advising

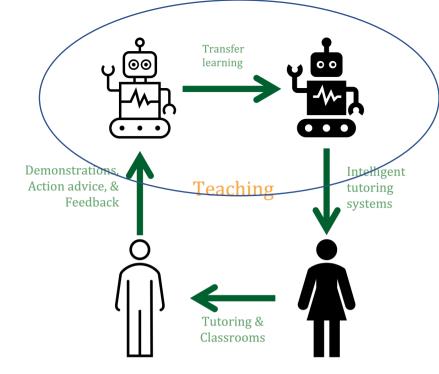
Mistake correcting

Predictive advising



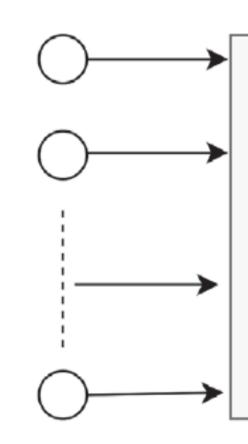
- Transfer Learning
- HAT ← Teacher Decides
- LeCTR (Omidshafiei+, 2018) ← Teacher Learns

Advising Reward Name	Description
JVG: Joint Value Gain	Task-level value V
QTR: Q-Teaching Reward	Teacher's estimate
LG: Loss Gain	Student's task-level
LGG: Loss Gradient Gain	Student's task-level
TDG: TD Gain	Student's temporal
VEG: Value Estimation Gain	Student's value esti

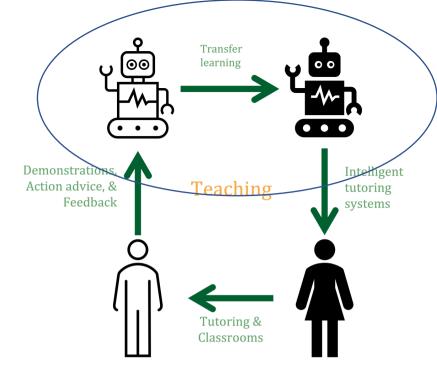


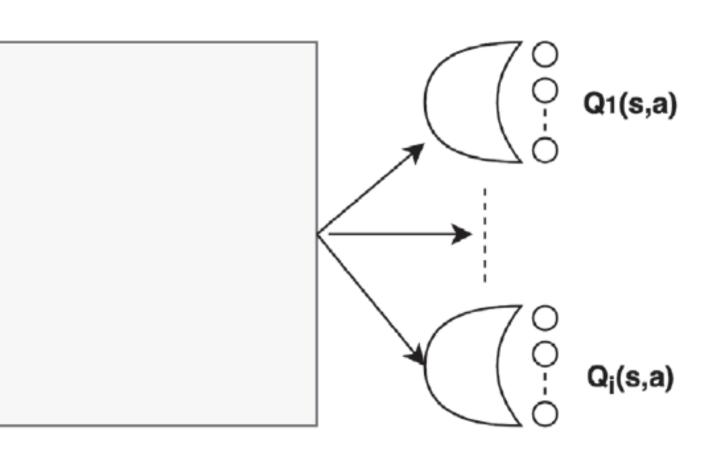
 $(s; \boldsymbol{\theta})$ improvement after learning of best vs. intended student action el loss $\mathcal{L}(\theta^i)$ reduction el policy gradient magnitude difference (TD) error δ^i reduction timate $\hat{V}(\theta^i)$ gain above threshold τ

- Transfer Learning
- HAT ← Teacher Decides
- LeCTR (Omidshafiei+, 2018) ← Teacher Learns
- RCMP → Student Requests



Inputs

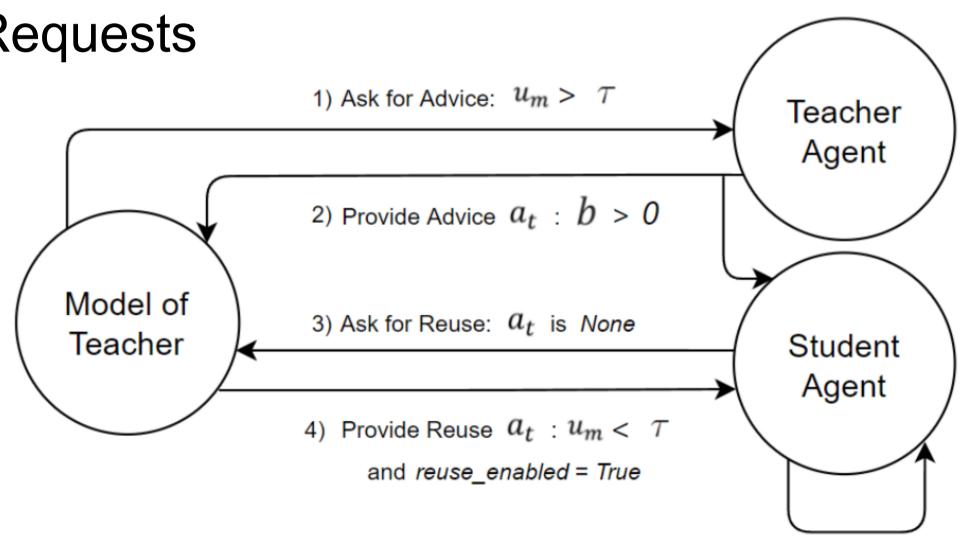


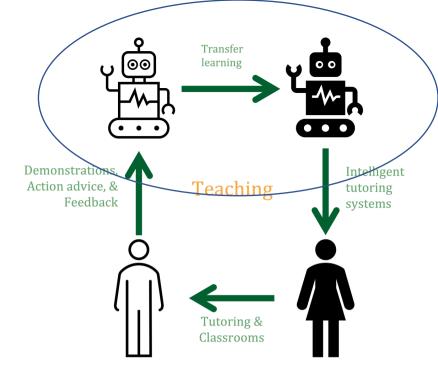


Hidden Layers

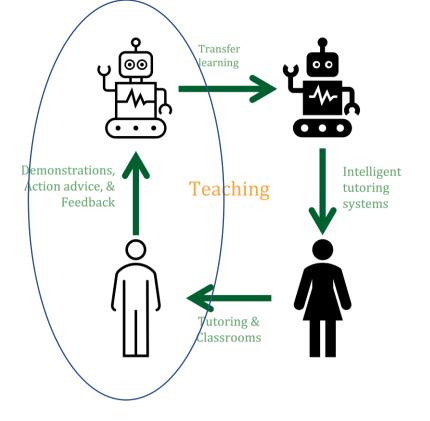
Heads

- Transfer Learning
- HAT ← Teacher Decides
- LeCTR (Omidshafiei+, 2018) ← Teacher Learns
- RCMP → Student Requests
- AIR (ilhan+, 2021) \rightarrow Student Requests



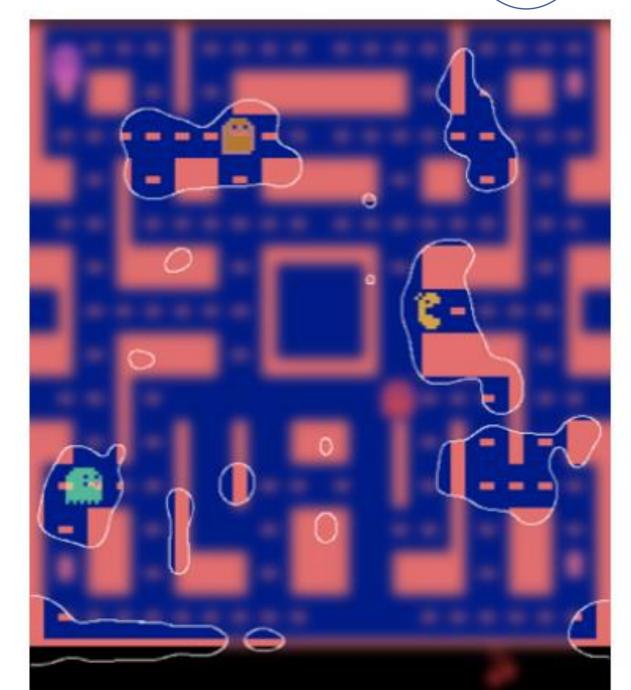


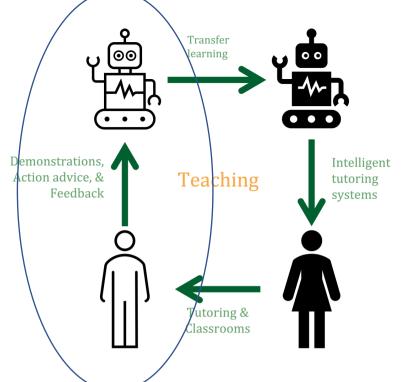
5) Follow self: a_t is None



Agent can also provide Explanations

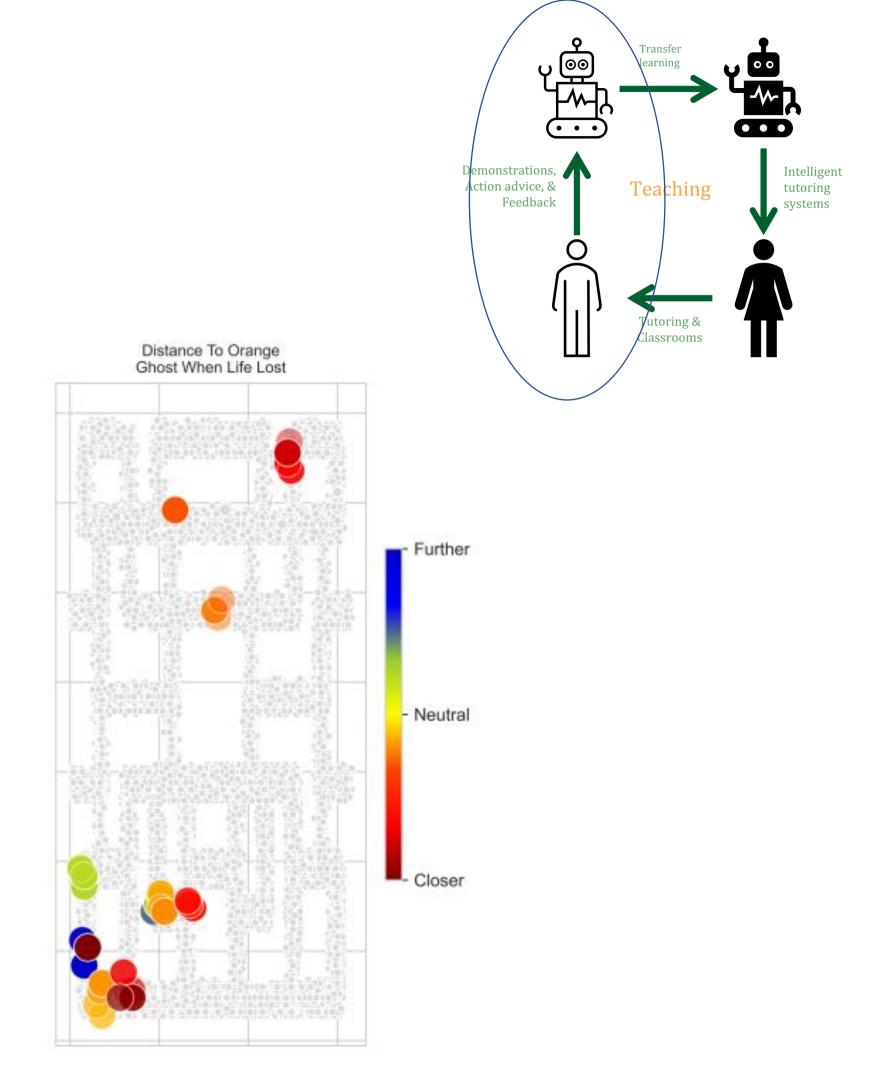
- "Why did you do A?"
- "Why didn't you do B?"
- "What would happen if you did C?"
 (Davis Pierson+, 2022)





Agent can also provide Explanations

- "Why did you do A?"
- "Why didn't you do B?"
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 (Davis Pierson+, 2022)



Agent can also provide Explanations

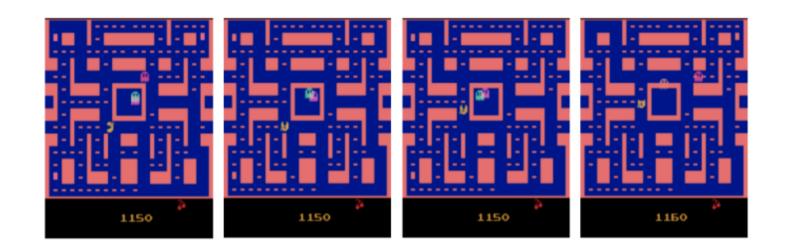
- "Why did you do A?"
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question

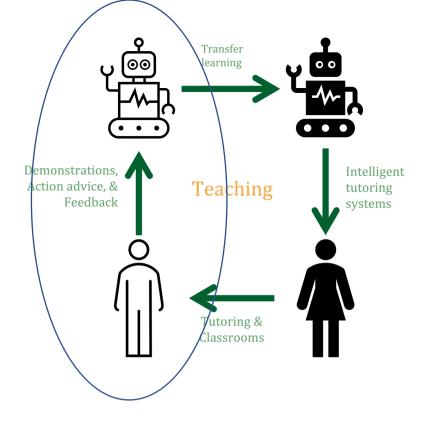
3/4

If you were in a competition where your agent needed to STAY ALIVE THE LONGEST, which AI agent would you select to play on your behalf?





auestion

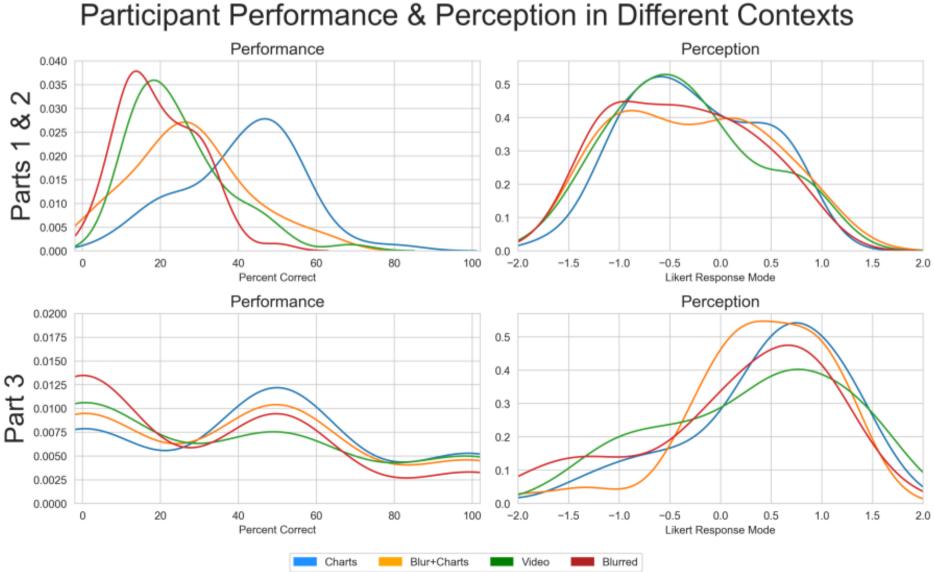


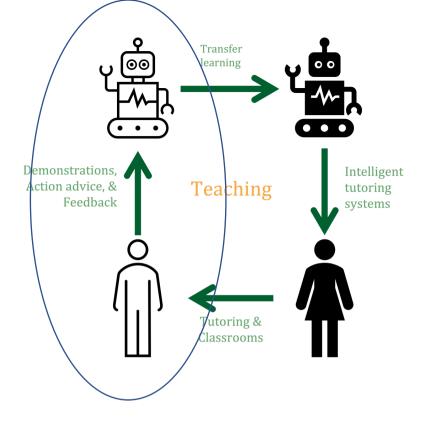
What will Agent Thompson do next in the situation described in the 4 1/13 images above?

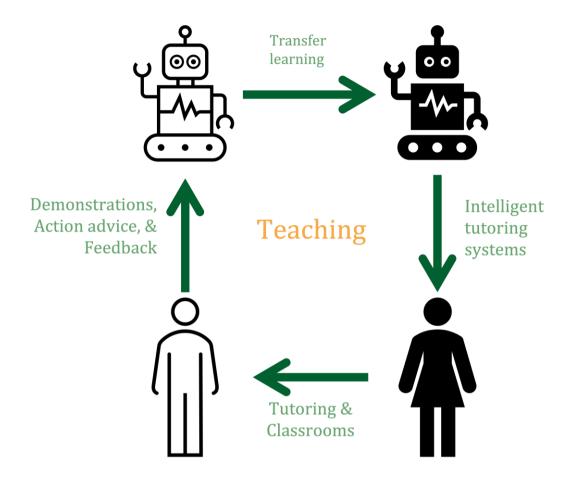


Agent can also provide Explanations

- "Why did you do A?"
- "Why didn't you do B?"
- "What would happen if you did C?" (Davis Pierson+, 2022)







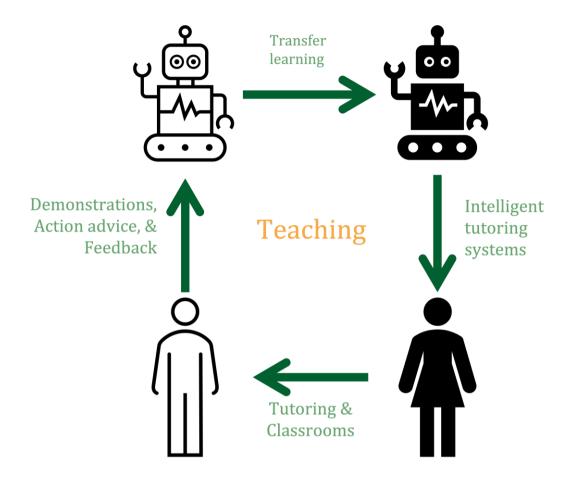
Other open questions

Human-agent teaming

Multi-agent systems Multi-agent, multi-human systems

Explainability Accountability Safety

My claim: Bar is low!

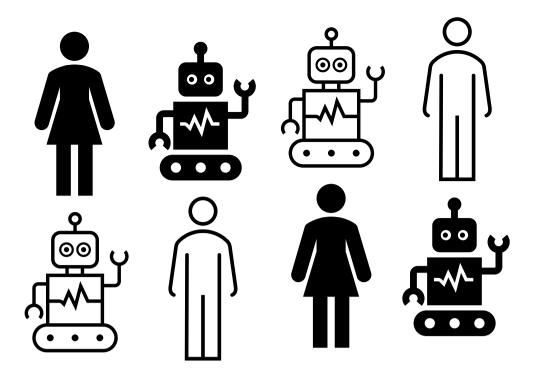


Sequential decision tasks?

Good task for machine intelligence?

Bootstrap off human knowledge?

Human-in-the-loop system?



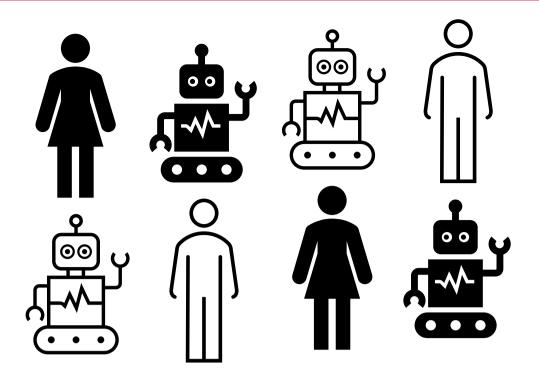
Multi-agent, Multi-human Teaming

cogment

First released in 2019. Version 1.0 released in June 2021. Continuous updates released since, including Cogment Verse, a series of actors and environments to make Cogment faster and easier to use.

The **first platform** to allow the design, training, and deployment of complex **intelligence ecosystems**, mixing **humans and artificial agents** of various kinds

It orchestrates heterogeneous ML & non-ML agents with real-time human interaction.





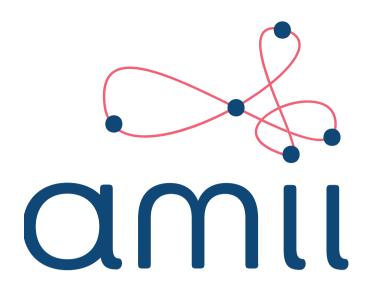




Thank you!

http://irll.ca

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