Generic-to-Specific Reasoning and Learning for Ad Hoc Teamwork

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February 26, 2025



¹ Funded in part by US ONR Science of Autonomy Award N00014-20-1-2390.

Collaboration without Prior Coordination

Example-FA

Example-HFO

Example-VH

- Ad hoc Teamwork: collaborate with other agents "on the fly".
- Limited prior knowledge of other agents/robots; fully/partially observable state but no (limited) communication.
- Need to reason with, learn, and understand behavior of other agents.

Existing Work and Our Approach

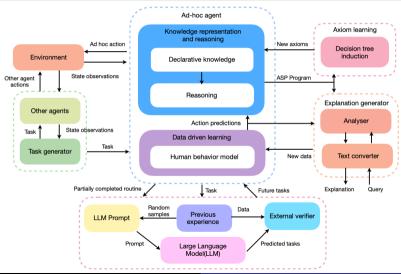
- Encoded protocols/plays, probabilistic sampling-based methods.
- State of the art: sequential decision making with data-driven component.
 - Deep network-based models; behavior of other agents or agent "types".
 - Optimize choices using long experience history.
 - Resource-hungry, opaque, domain knowledge not leveraged.
- Cognitive systems-inspired approach; joint reasoning, learning, control; different
 abstractions and processes. How to effectively leverage reasoning (domain-specific
 knowledge), interactive learning (behavior models), and LLM (generic prediction)?

Hasra Dodampegama and Mohan Sridharan. Reasoning and Explanation Generation in Ad hoc Collaboration between Humans and Embodied AI. In International Conference on Logic Programming and Nonmonotonic Reasoning (LPNMR), October 2024.

Hasra Dodampegama and Mohan Sridharan. Back to the Future: Toward a Hybrid Framework for Ad hoc Teamwork. In the AAAI International Conference on AI (AAAI), February 2023.

Reuth Mirsky, Ignacio Carlucho, Arrasy Rahman, Elliot Fosong, William Macke, Mohan Sridharan, Peter Stone, Stefano V Albrecht. A Survey of Ad Hoc Teamwork: Definitions, Methods, and Open Problems. European Conference on Multiagent Systems (EUMAS), 2022.

Architecture Overview



Contributions

- Non-monotonic logical reasoning, probabilistic reasoning, interactive learning, and foundation models inform and guide each other.
- Better performance than methods considered state of the art, using orders of magnitude fewer resources.
- Step-wise refinement: simplifies design; confidence and scalability.
- Ecological rationality for reliable and efficient reasoning and learning.
- Relational descriptions as on-demand explanation of decisions.
- Separation of concerns: common methodology, automation.

Execution Example: Reasoning + Learning

human agent1

agent3 agent4

That's all folks!

SMPT talk: 1615 February 27 (Rm. 125).

PhD, Postdoctoral research positions available.

