CSE3000 Research project: Agent Failure, Trust Repair, and Fluency in Human-AI Team

~ Impact of Opportunistic Interdependence Relationship in a Human-Agent Team ~

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1. Introduction

Human Autonomous Teams (HATs) combine capabilities to perform tasks more efficiently.

Trust recovery after **trust violations** is important to maintain a high trust level, which is crucial to team performance.

Past studies found:

- Communicating **uncertainty** in advice **mitigates** trust loss following trust violation [1].
- Expressing **regret/**providing **explanations** in apology is an effective trust repair strategy [2].

Collaborative fluency: measurement of coordination and meshing of actions in a team[3].

Interdependence relationship: set of complementary relationships that parties rely on to manage joint activities [4].

2. Research Question

How do **opportunistic (soft) interdependence relationships** affect

1) **trust violation** and **trust repair**
2) **collaboration fluency**

compared against independence (baseline) condition?

Why soft interdependence?

- It is necessary to achieve true teamwork [5].
- Successful teams tend to manage soft interdependence well [5].

3. Methodology

MATRX was used to conduct a user study.

Objective: to collaborate with an AI agent (RescueBot) and to rescue victims in different areas.

Figure 1: God view of the environment

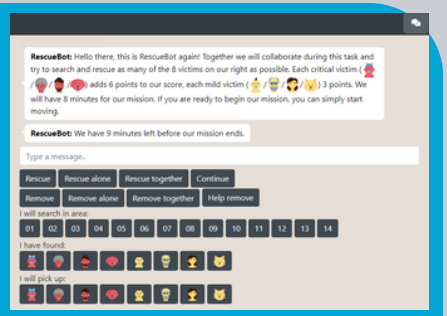

Figure 2: Messaging functionality

Procedure:

During the game, three extreme rains arrived. Getting hit by rain led to reduction in playing time/score. Before each rain, weather forecasting message was sent by RescueBot:

- ***1st advice:*** at 2 minutes mark. **Correctly** predicts the storm.
- **2nd advice**: at 4 minutes mark. **Incorrectly** predicts light rain, leading to trust violation. Trust repair follows.
- ***3rd advice:*** at 6 minutes mark. **Correctly** predicts the storm.

Trust was measured after the three advice/feedback with a questionnaire based on the trust scale for XAI context [6]. Collaboration fluency was measured with existing questionnaire [3] and objective metrics.

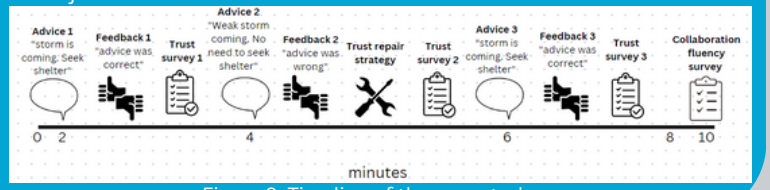

Figure 3: Timeline of the user study

4. Results and discussion

Trust:

- Significantly higher trust for opportunistic condition before trust violation (T1)

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Influenced by the sense of team structure

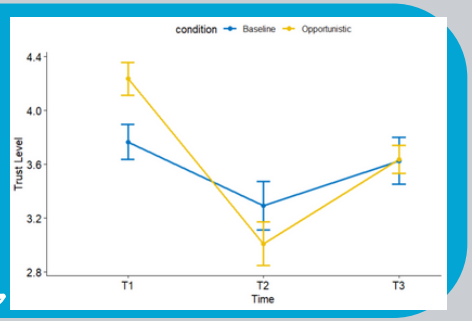

Figure 4: Effect of opportunistic interdependence (yellow) and baseline (blue) on trust at different times

- Opportunistic experienced significant trust violation/recovery

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Interdependence supports continuous calibration of trust

Collaboration Fluency:

- No significant difference based on questionnaire
- Objective metrics showed a difference in ratio of robot idle time but nothing significant

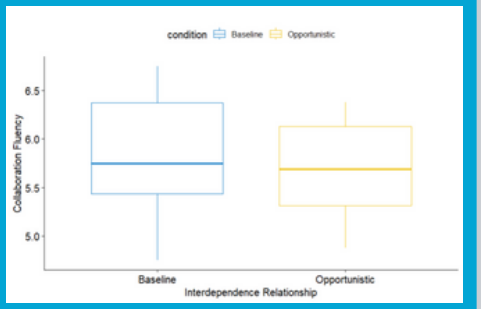

Figure 5: Analyzing the effect of opportunistic interdependence (yellow) and baseline (blue) on fluency

Limitations in the experiment:

- Arrow key press speed
- Difference in trust and fluency between subjects who got punished by rain/who did not etc

5. Conclusions and future work

- Trust was higher for opportunistic due to team structure
- Trust violation/recovery more affected by opportunistic due to role of interdependence to support continuous exploration of trust
- No significant result in terms of fluency

Possible next step:

Investigate a type of trust repair strategy that is effective for the teams with opportunistic interdependence in particular.

References

[1] E. Kox, L. Siegling, and J. Kerstholt, "Trust development in military and civilian human–agent teams: The effect of social-cognitive recovery strategies," International journal of social robotics, vol. 14, pp. 1323– 1338, July 2022. Funding Information: This material is based upon work supported by the Dutch Ministry of Defense's exploratory research program. Publisher Copyright: © 2022, The Author(s)
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[3]Guy Hoffman. "Evaluating Fluency in HumanâRobot Collaboration". In: IEEE Transactions on Human-Machine Systems 49.3 (2019), pp. 209–218. doi: 10.1109/THMS.2019.2904558.
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