



# U.S. ARMY COMBAT CAPABILITIES DEVELOPMENT COMMAND – ARMY RESEARCH LABORATORY

## Transparent Communications for Effective Human-Autonomy Teaming

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APPROVED FOR  
PUBLIC RELEASE



# CHALLENGES



- Human interaction with increasingly sophisticated systems capable of complex reasoning
  - Challenges that need to be resolved in order to achieve assured autonomy and effective human-autonomy teaming (International Organization for Standardization, 2020; Topcu et al., 2020)
  - Top three requirements for high-stake AI systems: **transparency**, traceability, and human control (European Commission 2020)
  - 9 principles: lawful; purposeful and performance-driven; accurate, reliable, and effective; safe, secure, and resilient; understandable; responsible and traceable; regularly monitored; **transparent**; and accountable (US Executive Order on “Promoting the Use of Trustworthy AI in the Federal Government”; Dec 2020)
  - Geoffrey Hinton (2018 Turing Award recipient): “What we need is for neural nets now to begin to be able to **explain** reasoning”
  
- Military Context
  - Six barriers to human trust in autonomous systems, with ‘low **observability**, predictability, directability, and auditability’ as well as ‘low **mutual understanding** of common goals’ being among the key issues (Defense Science Board’s *Summer Study on Autonomy*, 2016)





# TRANSPARENCY

- Human trust and joint human-system performance
  - Humans interacting with highly automated systems encounter multiple challenges: understanding the current system state, comprehending reasons for its current behavior, and projecting what its next behavior will be (Endsley 1995; Sarter & Woods, 1995).
  - In order to support effective human-autonomy teaming and joint decision making, the human and the machine agent need to understand each other's intent, reasoning, and expected outcomes.
    - Information that the human has but the machine does not have access to (e.g., intelligence reports)
    - Adding or removing constraints
  - Making AI's output more transparent in order to maximize the joint performance of the human-machine team (Matheny et al., 2019; Sanneman & Shah, 2020)
    - DARPA's eXplainable AI (XAI) Program
    - NSF Program on Fairness in Artificial Intelligence (FAI) in Collaboration with Amazon (2020-2023)





# TRANSPARENCY RESEARCH



## ■ **Transparency Frameworks**

- Situation awareness-based Agent Transparency (SAT) (Chen et al. 2014, 2018)
- Human-Robot Transparency (Lyons et al. 2014)
- Coactive System: Observability, Predictability, Directability (Johnson et al. 2014)

## ■ **Human-Robot Interaction**

- Small ground robots (Olatunji et al. 2020; Pynadath et al. 2018; Selkowitz et al. 2017; Wright et al. 2020)
- Multiagent management via an intelligent planning agent (Bhaskara et al. 2021; Mercado et al. 2016; Stowers et al. 2020; Vered et al. 2020)
- Robotic swarms (Hepworth et al. 2020; Roundtree et al. 2019; Roundtree et al, 2020)

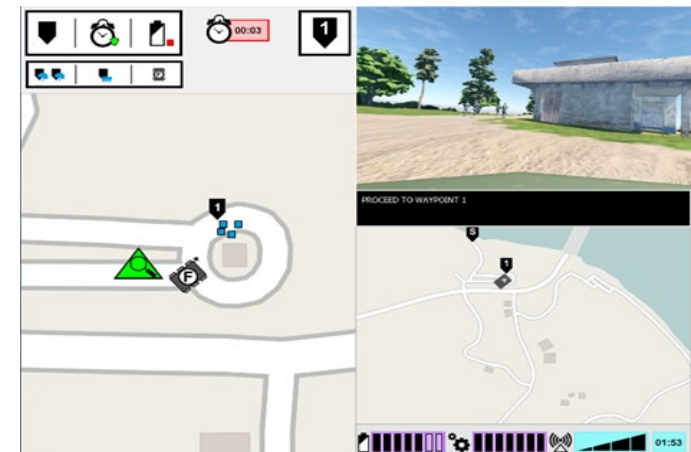
## ■ **Automated/Autonomous Driving** (Krause et al., 2020; Kunze et al. 2019)

## ■ **Aviation**

- Emergency landing planning agent (Lyons et al. 2016)
- Workload management agent in a helicopter cockpit environment (Roth et al. 2020)

## ■ **Explainable AI** (Chien et al., in press; Holder & Wang, 2021; Miller, 2019; Sanneman and Shah, 2020)

## ■ **Individual and Cultural Differences** (Matthews et al. 2020; Chien et al. 2020)





# SA-based Agent Transparency (SAT)

- Definition of Agent Transparency: “A quality of an interface pertaining to its abilities to afford an operator’s comprehension of an intelligent agent’s intent, performance, future plans, and reasoning process” (Chen et al., 2014)
- Focus on operator task performance and trust calibration

## Level 1

- **Purpose**
  - **Desire** (Goal selection)
- **Process**
  - **Intentions** (Planning/Execution)
  - Progress
- **Performance**

**What’s going on and what is the agent trying to achieve?**

## Level 2

- Reasoning process  
(**Belief**)(**Purpose**)
  - Environmental & other constraints/affordances

**Why is the agent doing it?**

Situation Awareness (SA) (Endsley, 1995)  
BDI Agent Framework (Rao & Georgeff, 1995)  
Trust calibration (Lee & See, 2004)

## Level 3

- Projection to Future/End State
- Potential limitations
  - Uncertainty; Likelihood of error
- History of Performance

**What should the operator expect to happen?**

Chen, J.Y.C. et al. (2014). *Situation Awareness-based Agent Transparency* (ARL-TR-6905).

Chen, J.Y.C. et al. (2018). Situation awareness-based agent transparency and human-autonomy teaming effectiveness. *Theoretical Issues in Ergonomics Science*, 19(3), 259-282.

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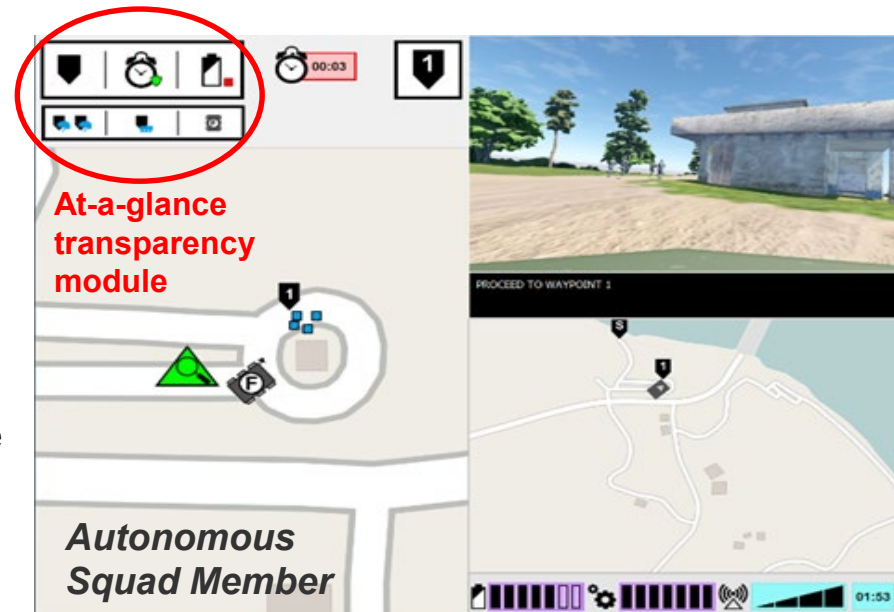




# SAT-BASED RESEARCH

## Human-Robot Interaction

- Human interaction with a small ground robot
  - Robotic support of an infantry squad (Autonomous Squad Member) (Selkowitz et al. 2017; Wright et al. 2020)
  - Robotic support for threat detection (Pynadath et al. 2018)
  - Older adults' interaction with an assistive robot (Olatunji et al. 2020)



**Robot:** I have finished surveying the Auto Parts Store. I think the place is safe. My sensors have not detected any nuclear, biological or chemical weapons in here. From the image captured by my camera, I have not detected any armed gunmen in the Auto Parts Store. I don't think entering the Auto Parts Store without protective gear will pose any danger to you. Without the protective gear, you will be able to search the building a little faster.

Enter without protective gear

Put on protective gear

Pynadath, D. et al. (2018). Transparency communication for machine learning in human-automation interaction. In: *Human and Machine Learning. Human-Computer Interaction Series* (ed. J. Zhou and F. Chen) Springer, Cham.

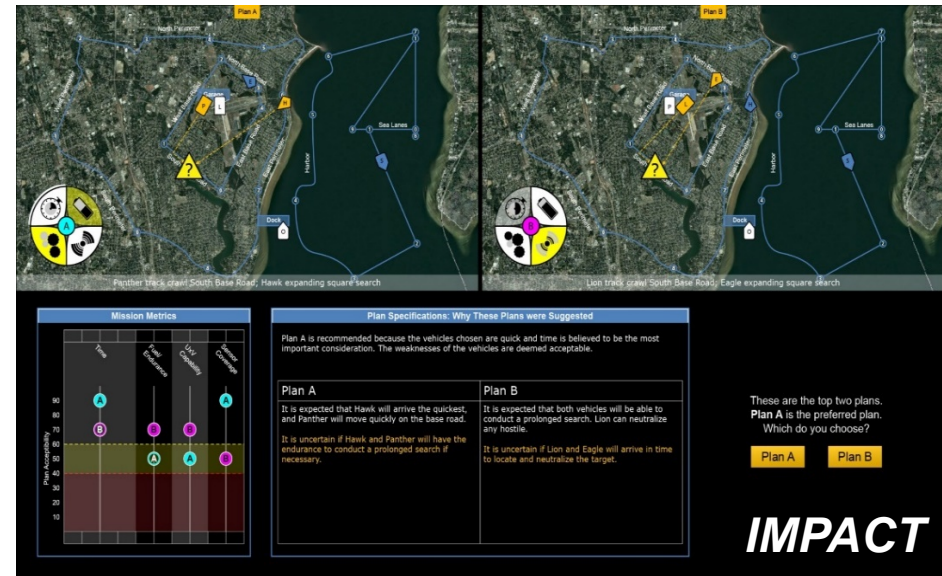


# SAT-BASED RESEARCH



## Multiagent Management

- Mission planning involving autonomous aerial and ground vehicles
  - IMPACT (Mercado et al. 2016; Stowers et al. 2020)
  - Defense S&T Group of Australia (Bhaskara et al. 2021)
- Workload-adaptive cognitive agent in helicopter cockpit environments (Roth et al. 2020)



## Human-Swarm Interaction

- Roundtree et al. (2019) identify key challenges associated with applying transparency design principles to achieve the three levels of SAT.
  - Design guidelines on transparent human-swarm interface visualizations (Roundtree et al. 2020).





# SAT-BASED RESEARCH



## Explainable AI

- **Planetary Rover**
- XAI design and evaluation framework (similar to SAT) based on human users' informational needs related to their situation awareness in the human-agent tasking environments (Sanneman & Shah, 2020)
- **Fake News Detector**
- SAT-based explainer for a fake news detection system (Chien et al. in press)
- **“Junior Cyber Analyst”**
- SAT-based HMI for a “junior cyber analyst” AI agent that works with the “senior” human analysts to identify cyber threats (Holder & Wang, 2021)

## Planetary rover example

- **Level 1 XAI:**
  - Engineer** - terrain information, current battery level (inputs); current path plan and next stopping point/time (plan); next science action (decision/action)
  - Scientist** - next science action (action/decision); inputted image of rock for science analysis (input); rock classification (output)
- **Level 2 XAI:**
  - Engineer** - terrain map with rover path costs including untraverseable areas with infinite cost (policy information - costs); battery usage for current path (constraints); list of possible science actions and associated rewards (policy information - rewards); battery usage for each science action (constraints)
  - Scientist** - list of possible science actions and associated rewards (policy information - rewards); list of semantic features, such as color, contributing to the rock classification (feature information); sensitivity to light given inputs (sensitivity information)
- **Level 3 XAI:**
  - Engineer** - map of maximum traverseable distance given current battery level (continued action); remaining battery level after each possible science activity (continued action)
  - Scientist** - predicted rock classification under different lighting conditions (changed inputs)

Sanneman, L., & Shah, J. A. (2020). A Situation Awareness-Based Framework for Design and Evaluation of Explainable AI. In *Proc. EXTRAAMAS 2020*.

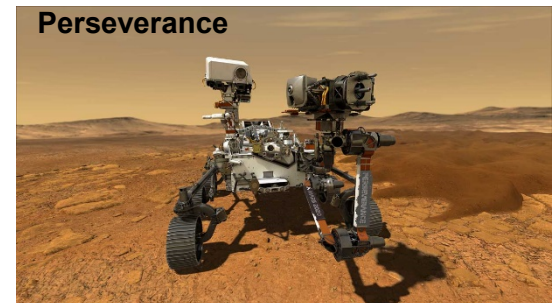


Image: NASA/JPL-Caltech





# FINDINGS OF SAT-BASED RESEARCH



## Operator Performance

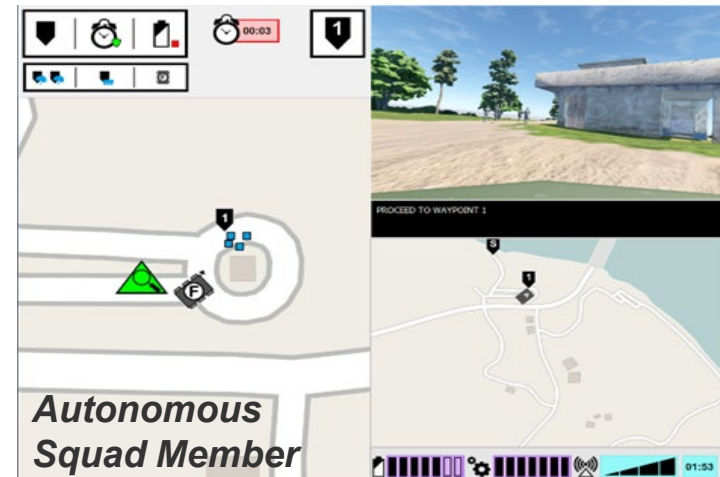
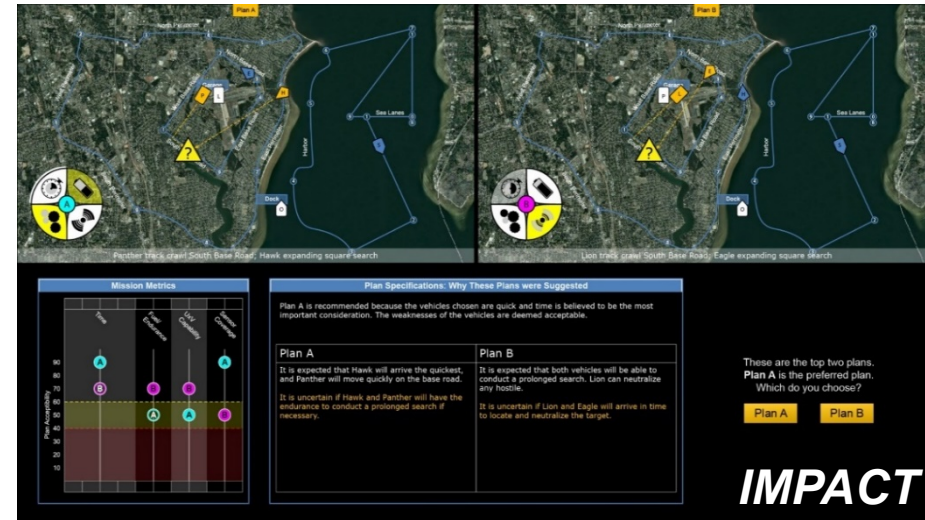
- Significant improvements as agent transparency (AT) increases
  - More effective trust calibration

## Operator Workload

- No sig. increases as AT increases

## Operator's Perception of Agent

- Operator's perceived *trust* tends to increase as AT increases
  - Factors such as agent reliability may impact trust more
  - Effects of uncertainty information on trust not always consistent
    - Level 3 (projection) information may lead to over-trust in an unreliable agent (Bhaskara et al., 2021)
- Operator's perceived *humanness* (anthropomorphism and intelligence) increases as AT increases



*Increasing AT => better operator performance without increases in workload*



# INDIVIDUAL & CULTURAL DIFFERENCES



## *Individual Differences*

- Individual differences in attitudes toward robots (e.g. unreasonable expectations of robot capability or negative attitudes toward humanlike robots) may impact humans' mental models of robots' task performance, which in turn, may affect their trust calibration and SA (Matthews et al. 2020)
  - Transparent interface design suggestions based on the SAT framework
  - Transparency content should be compatible with the operator's mental model by highlighting appropriate aspects of robots' capabilities



## *Cultural Differences*

- Effects of cultural differences on human-agent interaction in the context of multiagent management (Chien et al. 2020)
  - Three distinct cultural backgrounds (based on the Cultural Syndromes Theory) were assessed in the experiment: United States (Dignity), Taiwan (Face), and Turkey (Honor).
  - Transparency had an impact on operator's interaction with the planning agent (i.e. compliance with agent's recommendations), but the effects of agent transparency were significantly influenced by participants' culture. For example, Face culture participants had a higher tendency to accept recommendations from an opaque agent.
  - When transitioning autonomy technologies from one culture to another, user interface modifications and training interventions may be required



# CHALLENGES & FUTURE RESEARCH



- Architecture of transparent interfaces and info requirements
  - System users vs. evaluators
  - Operator vs. scientists
- Transparent interfaces (all 3 levels of SAT) for systems that continue to learn and evolve
  - How to convey newly-acquired capabilities?
- Real-time generation of transparent/explainable content
  - Modalities of transparent/explainable interfaces
  - Effects on operator workload (Kunze et al. 2019; Skraaning & Jamieson, 2021)





# REFERENCES

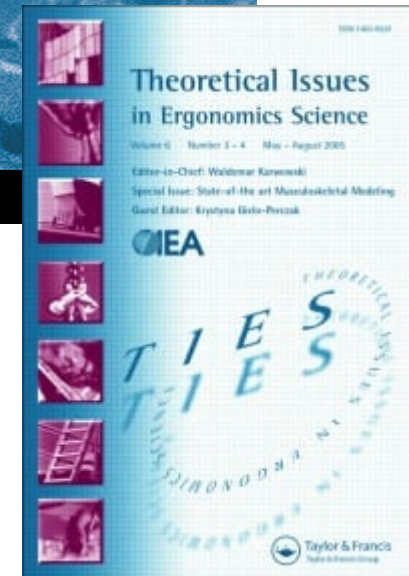
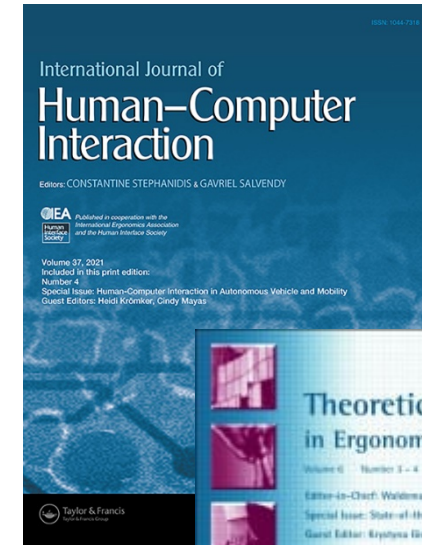


## References

- ❑ Chen, J.Y.C., Lakhmani, S., Stowers, K., Selkowitz, A., Wright, J., & Barnes, M. (2018). Situation awareness-based agent transparency and human-autonomy teaming effectiveness. *Theoretical Issues in Ergonomics Science*, 19(3), 259-282.
- ❑ Chen, J.Y.C. et al. (2014). *Situation Awareness-based Agent Transparency* (ARL-TR-6905).
- ❑ Chen, J.Y.C., & Barnes, M.J. (2014). Human-agent teaming for multi-robot control: A review of human factors issues. *IEEE Transactions on Human-Machine Systems*, 44(1), 13-29.

## Additional Resources

- ❑ *International Journal of Human-Computer Interaction*: Special Issue on “Transparent Human-Agent Communications” (2022)
- ❑ *IEEE Transactions on Human-Machine Systems*: Special Issue on “Agent and System Transparency” (2020)
- ❑ *Theoretical Issues in Ergonomics Science*: Thematic Issue on “Human-Autonomy Teaming” (2018)



## IEEE TRANSACTIONS ON HUMAN-MACHINE SYSTEMS

A PUBLICATION OF THE IEEE SYSTEMS, MAN, AND CYBERNETICS SOCIETY