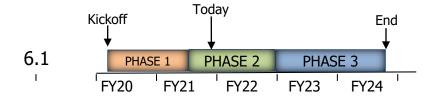
# Artificial Social Intelligence for Successful Teams (ASIST)

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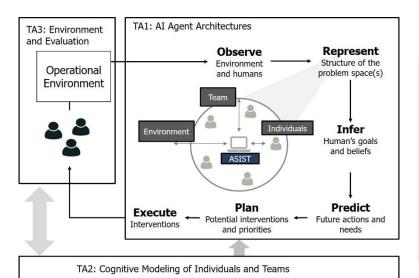
Human AI Teaming Workshop
July 29, 2021





### Artificial Social Intelligence for Successful Teams (ASIST)

ASIST will develop AI theory & systems that demonstrate machine social skills needed to infer the goals and beliefs of human partners, predict what they will need, and offer context aware interventions in order to act as adaptable and resilient AI teammates



#### Measures

# fNIRS Eye tracking: pupil diameter gaze Facial Satisficing Workload Spatial ability Demographics Goals Task knowledge

expression • Strategy
Competency test score

Testbed message bus events

### Challenges:

- Application of Machine Theory of Mind and Teams to identify team problems and automate coaching to improve team function and performance
- Modularization of Agent Social Intelligence (ASI) agents with asynchronous sensing and that are robust, adaptable, safe, and effective

#### Approach:

- Develop ASI agent as coaches to optimize team function
- Infer mental states, predict behaviors, and offer guidance that improves decisions and coordination between members
- Operate in increasingly complex and specialized environments
- Adapt to unexpected perturbations
- Contribute to a revolution in cognitive modeling with shared situational awareness for effective human-machine teaming
- Engage with the warfighter as a collaborator: *capable of understanding* the attitudinal, behavioral, and cognitive components of teamwork

### Accomplishments:

- Created and prototyped experimental design and procedures
- Conducted full-scale evaluations
- Developed and released testbed; adapted for use in other DARPA programs



Minecraft USAR Testbed Environment



### Applications of Symbiotic Teams – Mission Relevance

Force Application: Design and test teams & systems in complex tactical missions



Coaching, Training and Education: Maintenance and repair,
Tactical field care, Adaptive training



Cyber Protection Teams: network security, cyber systems engineering



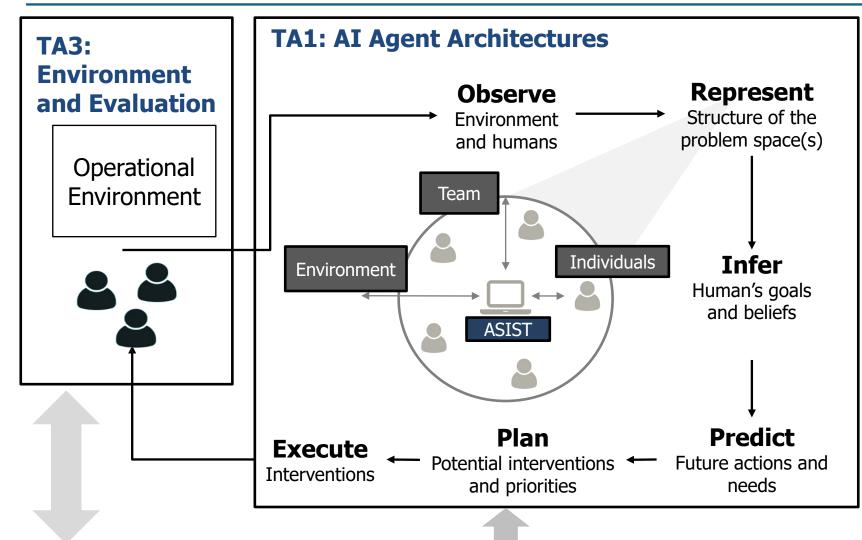
Search And Rescue/Urban Raid: First responder, Hostage rescue



ASIST will allow AI to be a dynamic team partner (collaborative and interdependent) to understand mission objectives and intervene for effective teaming



# Program Structure



### **TA2: Cognitive Modeling of Individuals and Teams**

### ASIST agents will:

- Provide artificial coaching to optimize team function
- Operate in increasingly complex and specialized environments
- Adapt to unexpected perturbations
- Contribute to a revolution in cognitive modeling for human-machine teaming

### Change in Program Phases:

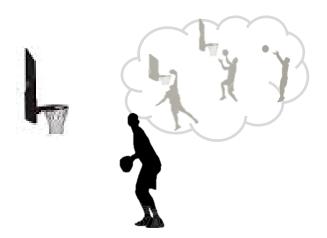
- First year success in programmatic and integration enabled faster pivot to more complex teaming....more profound results
- Moving from passive to active to generalization



# What makes humans good teammates?

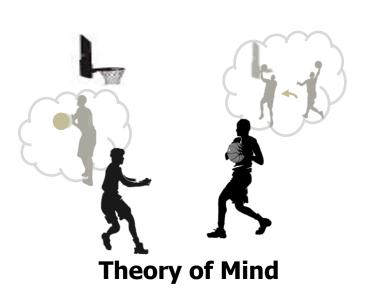
# **Mental Models** of Environment

Humans can build robust mental models of their environment



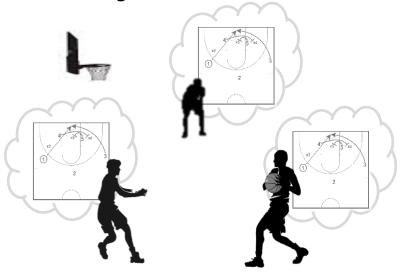
# Mental Models of Others

Humans can infer, from observed actions and context, the mental states of other humans



# **Shared Mental Models**

To perform in teams, humans use experience and training to align their Mental Models

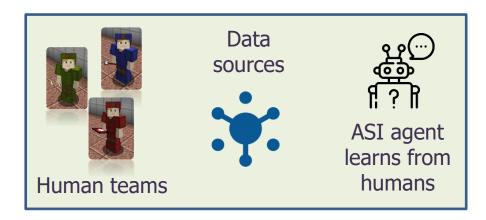


### **Social Intelligence**



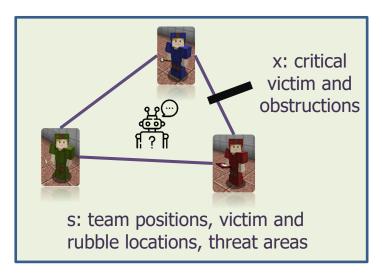
# Inferring Team Shared Mental Model (SMM)

# Observable (s) and latent (x) features influence team performance



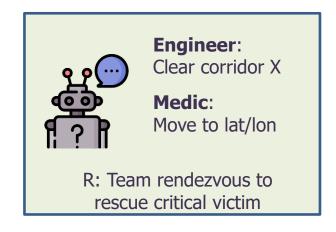
Learn team SMM (model human teams)

# Latent features can differ among team members



Infer lack of team SMM (model misalignment)

# Decide when and how to intervene



Intervene (correct errors or improve team performance)

Hard problem to sense all the observable and latent features necessary to create Team SMM



### Challenges for agents that collaborate with humans for effective HMT

Learn independent of well-defined specifications (train interactively from human feedback)



**Techniques to build human-level AI** 

Agents that understand & are understood by humans

More traditional AI systems use predefined reward functions or labeled data to train an algorithm

- Learn human mental models to better coordinate with them
- Learn true signal derived from human behaviors or explicit feedback (NL, demonstrations, preferences, etc.)
- Human feedback as policy or reward shaping

Satisfying literal specification, but not achieving intended outcome ... just reward is often not great for evaluation



Source: (Amodei & Clark, 2016)



# How it is done today? Model Theory of Mind (ToM) of human participants and team

### **Current methods of ToM and limitations**

Challenge	Approach	Limitations and difficulties modeling human psychology			
Mental Models of Environment	Decision theory	Rely on assumptions of rationality that people constantly violate			
	Cognitive architectures	Agents not equipped with a general cross-domain knowledge (only used in narrow tasks); does not support multiple conflicting goals			
	Formal systems	Not able to deal with complex knowledge structures of humans; representations insensitive to distinctions among conflicting goals			
Mental Models of Others	Game theory	Rely on concepts of equilibria that people rarely achieve in an unstructured social setting			
	Bayesian	Performs Bayesian inference over beliefs and desires simultaneously; does not support multiple conflicting goals			
	Neural Network	Model single agent (not human) in simple environment; high training cost; not scalable to complex domains			
Team Shared Mental Model	New research in real world human-AI teaming (e.g. learning from human teams, modeling teams with no explicit reward specification)				

- Baker, Chris, Jara-Ettinger, Julian, Saxe, Rebecca, and Tenenbaum, Joshua. Rational quantitative attribution of beliefs, desires and percepts in human mentalizing. Nature Human Behaviour, 1(4):0064, 2017
- Rabinowitz, N. C., Perbet, F., Song, H. F., Zhang, C., Eslami, S. M. A., & Botvinick, M. (2018). **Machine Theory of Mind**. ArXiv:1802.07740

### **ASIST models of ToM and Theory of Teams**

Mental Models of Environment

Learn and use abstractions in complex open world environments AI approaches to learning environment and human team mental models:

- Instance-based learning theory
- Cooperative inverse reinforcement learning
- Imitation learning
- Learning by demonstration
- Story-based Bayesian inference
- Etc...

**ASIST** 

Infer human mental states even in ambiguous and new situations

Theoreticallygrounded cognitive frameworks for effective teaming

Mental Models of Others (Theory of Mind)

Team Shared Mental Models (Theory of Teams)

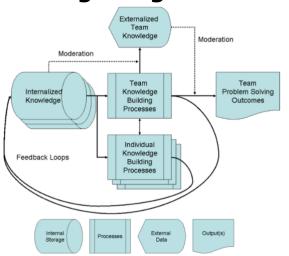
- Nguyen, T., Gonzalez, C., Cognitive Machine Theory of Mind, 42nd Annual Meeting of the Cognitive Science Society (CogSci 2020), pp. 2560-2566
- Jain, V., Jena, R., Li, H., Gupta, T., Hughes, D., Lewis, M., & Sycara, K., "Predicting Human Strategies in Simulated Search and Rescue Task," NeuRIPS, AI+HADR Workshop, 2020.



### Agent Social Intelligence (ASI) enables Superhuman Teams

### Apply variety of research dimensions to human-AI symbiosis

### **Integrating Theories of Teams for ASI**



**Macro-Cognition in Teams** 

- · Shared Mental Models
- Macro-Cognition in Teams
- Charnov's Marginal Value Theorem (MVT) and Optimal Foraging Theory
- Trust to Strategy Alignment
- Collective Intelligence
- Etc...

Test hypotheses derived from grounded social science theory (pre-registration for rigorous and reproducible research), produce useful modular **analytics** to improve ASI

ASI agent capabilities to observe human & team behaviors, infer their mental states, predict future behaviors, and intervene ipelines Observe behavior **ASI Architecture** & state Open Generalizable Modular Infer & Assess Scalable effects Predict Sensors **Feedback ASI Agents**  Robust Adaptable I.D. False Safe Intervene & Weak Effective beliefs

**Result:** Machine ToM, along with operationalized theories from team science enable automated identification of risks to team and mission and coaching to improve team function and performance



### Testbed with operational complexity in a controlled environment

- DoD-Owned Testbed & Experimentation
  - Automated, replicable HSR, flexible, containerized, cloud-based
  - Supports real-time AI agent interaction with human
  - Distributed team environment with robust message scheme
  - Evaluations of increasing complexity and scale, common baseline measures, objective and subjective data
- Studies inform focus of future studies
  - Study 1: Individual, passive ASI
    - Focused on broad territory of AI-human interaction; learning and baselining
      - Infer the training condition (state) of the player
  - Study 2: Team, passive ASI
    - Focused on measures relevant to Theory of Mind; 68 human teams (204 participants)
      - Infer participant mental model
      - Predict participant performance and action
  - Study 3: Team, active ASI
    - Focus on effective interventions
- ASI Agents
  - Understand human social intelligence in a team context
  - Predict what is needed and intervene as an effective partner
  - Handle known and unknown perturbations in task, team, mission, or environment for fast adaptation and team resilience

"No battle plan survives first contact with the enemy"
-- Helmuth von Moltke

Agent orchestration, execution, and analytic results (complexity, accuracy, speed of inference)

Minecraft USAR Testbed Environment



Measure	Study 1	Study 2			
Complexity of mission/testbed	1 player, 3 training conditions	team of 3; 2 planning conditions; variation in knowledge (maps, blocks)			
Volume of messages/data	200GB/2000 files	>500GB/3000 files			
Common metrics	2: team peformance, mental model	4: team performance, mental models (2), ToM			
ASI capabilities pre- registered (TA1)	24	30 (>50)			
Hypotheses pre-registered (TA2)	47	58 (>100)			
Number of analytic agents	2	5 (>20)			

Study 3-6 will increase team and task complexity and produce agents that are more general, modular, scalable, useful, and trustworthy



# Modeling a USAR team collaboration task and experiment

**Application**: Computer simulations of USAR collaborative task inspired by real world scenarios

**Team:** 3-member team with role selection and skill

Search Specialist (Searcher)



Medical Specialist (Medic)



Heavy Equipment Specialist (Engineer)



Sledgehammer

**Environment:** Minecraft representation of the USAR open world environment

Regular and Critical Victims









**Task**: Leverage team roles to find and save regular and critical victims; recover from perturbations (e.g. threats) in the environment

Goal: Save as many victims as possible in 15 min

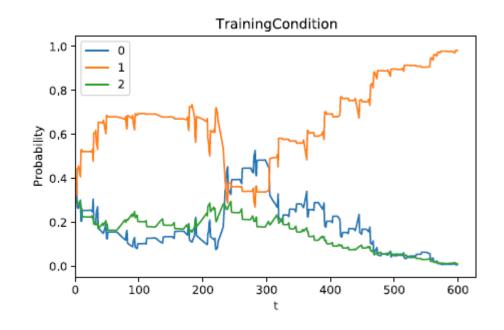
**Sub-goal**: Support team member under threat



### Agent accurately infers mental state of human

- Training induces a complex cognitive state (goals, strategy, beliefs, etc.) that drives a player's behavior (choices, reactions, etc.), making this as close as possible to measuring a true "Theory of Mind"
- Human players were given 1 of 3 training conditions at start of play, including information on tools and strategies
- Agents continuously update inference of training conditions over the course of a mission
- Agents inferred training conditions from data (80% accuracy)

Agent infers mental states based only on observable behaviors



Predict Actions and Performance (e.g. triage of a victim, navigation):

Accuracy of action predictions ranged from 74.7-99.2%

Infer Mental Models (e.g. knowledge condition):

69-80% accuracy



# **Program Metrics**

Measure	Phase 1 Passive ASI		Phase 2 Active ASI		Phase 3 Generalizations	
Level of complexity  Levels of team complexity	Single Human Multiple Humans		Multiple Humans, new mission		Multiple Humans, Non- Human Teammate, New environments/missions	
Levels of mission complexity	Simple	Medium	Medium	Complex	Complex	Complex
Direct: Accuracy of state/action prediction Time to generate initial inference Predict team performance ToM: Infer training/information		ole to human pach)	Comparable to human (Coach)		Comparable to human (Coach)	
Survey measure: Usefulness, Trust			>75% of users: •Use Agent •Report trusting agent		>95% of users: •Use Agent •Report trusting agent	
Adaptation time Resilience time Coordination after perturbation (friction)	_		30% reduction compared to human team		60% reduction compared to human team	
# integrated	>20		>40		>60	
# used by TA1 agents			75%		90%	
# hypotheses tested/published	>50		>100		>200	

#### **Mission Complexity:**

- Victim types and location by region and room
- Number and location of blockages
- Threat rooms
- Team planning conditions

#### **Common Metrics:**

- Predict final score/trial (team performance)
- Infer map type (mental model)
- Infer block meaning (mental model)
- Predict action given false belief (ToM)

With Successful ASI, machines can be effective partners with humans



# **DARPA** ASIST Summary

- ASIST will deliver advanced AI agents that can infer individual and team mental states, identify risks to team cohesion, and effectively intervene to optimize team performance
- Exciting area of fundamental AI theory and application...
  - to autonomously learn rich models of humans and teams
  - enable highly effective teams in complex multi-agent tasks
  - with potential for wide range of DoD applications (search and rescue, information operations, cyber teams, planning, training, force application)
  - ASI is a critical building block to achieve AGI

Creating ASI agents for highly effective human machine teaming



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