

Unobtrusive Measurement and Autonomous Estimation of Human Internal Cognitive States

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Traditional methods interrupt primary task

- Post-hoc surveys (NASA TLX, Modified Bedford Workload Scale, trust questionnaires, SART)
- System freezes (SAGAT)
- Probes and queries by experimenter (SPAM)



Embedded measures and psychophysiological signals can inform <u>real-time</u> estimates of operators' trust, mental workload, and situation awareness



Primary task: station docking simulation

Docking Status

100

Unobtrusive methods and validation					
	Embedded	Psycho- physiological	Gold standard for validation		
Trust	 Time to accept or reject "trust action" recommendation 	EEG [1]EDA sensingfNIRS	 Propensity to Trust Scale [2] Overall Trust Scale [3] 		
Mental workload	 Secondary workload task responses [4] Time spent monitoring secondary workload task 	 EEG [5] Eye tracking [5] fNIRS 	 Modified Bedford Workload Scale NASA TLX 		



Situation awareness embedded measure: verbal callouts at intervals for RCS fuel level and distance

Mental workload and trust embedded measures: two choice visual secondary task and autonomous system trust recommendation



Subjects set an electric potential level for their vehicle to try and match that of the station, based on comparing an



150

CAPTURED

Data Link

Predicted Station Potential:

Situation awareness	 Tertiary task callout response time/accuracy [4] 	EEGEye trackingfNIRS	• SART



QQ



Challenges

1. Are "unobtrusive" measures truly unobtrusive?

- Secondary workload measures will have some influence on primary task performance
- Simultaneous tasks can confound each other, and are contrived/don't always have a real life analogue
- Biomedical sensors can inhibit motion and limit operational use
- 2. Internal cognitive states are impossible to know for certain
 - Difficulty of "validating" unobtrusive measures when the gold standard measures themselves are not direct measurements
 - Psychophysiological signals may provide more objective measurements, but a baseline is needed to define signal artifacts
- 3. How can an absolute value of trust be quantified?
 - Influenced by different subject's predispositions to trust autonomous systems

Future work for autonomous estimation

Estimation methods (Kalman filtering, neural networks) will combine and weight real-time unobtrusive measurements with predictions from computational models, generating estimates of cognitive states for an adaptive human-automation interface. We hypothesize that if an adaptive interface can know its operator's cognitive states, it can adapt to best aid the operator.



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- Easier for an operator to report changes in trust rather than an absolute level
- "Blind" trust tasks vs. "informed" trust tasks
- Trust should be measured in a way that can be easily used by an adaptive system

