



Developing an Autonomous Task Manager for Intelligence, Surveillance, and Reconnaissance Human-Machine Teams

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Integrity ***** Service ***** Excellence

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- Workload consists of multiple tasks and long shifts
- Work within Human-Machine Teams (HMT) to Process, Exploitation and Dissemination (PED) Essential Elements of Information (EEIs) to customer down range
- Multitasking environment may require automation to improve overall performance (accuracy and efficiency)
 - Reduce repetitive "busy work"
 Examples:
 Copy & Pasting
 Target Detection





2



Autonomous Manager (AM) is a "new agent" within the PED cell. Through simulation, AM currently:

- Intelligently and dynamically parses task allocation in realtime based on agent performance and workload
- Simulates performance with varying prior uncertainty
- Incorporates physiological indicators of human workload (e.g. heart rate)







(a) Multi-INT dashboard of four tasks (e.g. T1, T2, T3 and T4)(b) Example of task allocation based on performance criteria and workload





AM Simulation Performance





Mean Performance

- AM Parser performs 11.37% better than Human baseline
- Optimal Performance performs
 5.46% better than AM Parser

<u>Performance Improvement (Δ)</u>

Under which task conditions do we find greatest and lowest degree of improvement







AM with Nonstationary **Performance**

1400

1200

1000

600

400

200

0

0

0.1

0.2

0.3

0.4

Optimal AM Parser Human

0.5

Performance (Accuracy)

0.6

0.7

requency 800







Mean Performance of Optimal and AM Parser

0.8

0.9

- Stationary: 5.46%

Performance Across Nonstationary

Configurations (3 changes)

- Nonstationary (1 change): 5.52%
- Nonstationary (3 changes) : 6.12%
- Nonstationary (6 changes) : 7.03%





Nonstationary Performance











Performance Improvement (Δ)

- Stationary: 11.37%
- Nonstationary (1 change): 11.20%
- Nonstationary (3 changes) : 10.79%
- Nonstationary (6 changes) : 10.57%





Conclusions



- Initial simulation shows:
 - Performance improvement
 - Robust to dynamically changing performance conditions
 - Can be improved with more sophisticated models of workload and more flexible performance thresholds







- Add more physiological indicators of human workload
- Extend AM beyond static thresholds
 - Physiological & Performance
- Model hierarchical dependences between tasks
 - Single HMT, Teams of HMTS
- Develop task environment
 - Modify AM to real-time parsing of physiological and performance





Questions?





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Nonstationary Summary Statistics						
Condition	Optimal Mean (SD)	AM Parser (SD)	Mean Difference	t	р	Effect Size
Nonstationary (1 Change)	66.81% (11.92%)	61.29% (9.45%)	5.52%	58.16	0.000	0.513
Nonstationary (3 Change)	66.97% (11.99%)	60.85% (6.64%)	6.12%	56.07	0.000	0.631
Nonstationary (6 Change)	67.04% (11.91%)	60.01% (5.63%)	7.03%	57.19	0.000	0.755
Condition	Optimal Mean (SD)	AM Parser (SD)	Mean Difference	t	р	Effect Size
Nonstationary (1 Change)	50.09% (14.53%)	61.29% (9.45%)	11.20%	80.67	0.000	0.914
Nonstationary (3 Change)	50.06% (14.62%)	60.85% (6.64%)	10.79%	74.59	0.000	0.948
Nonstationary (6 Change)	49.75% (14.77%)	60.01% (5.63%)	10.26%	67.63	0.000	0.918

