

Accounting for Fatigue in Systems Design and Operations: Issues and Opportunities

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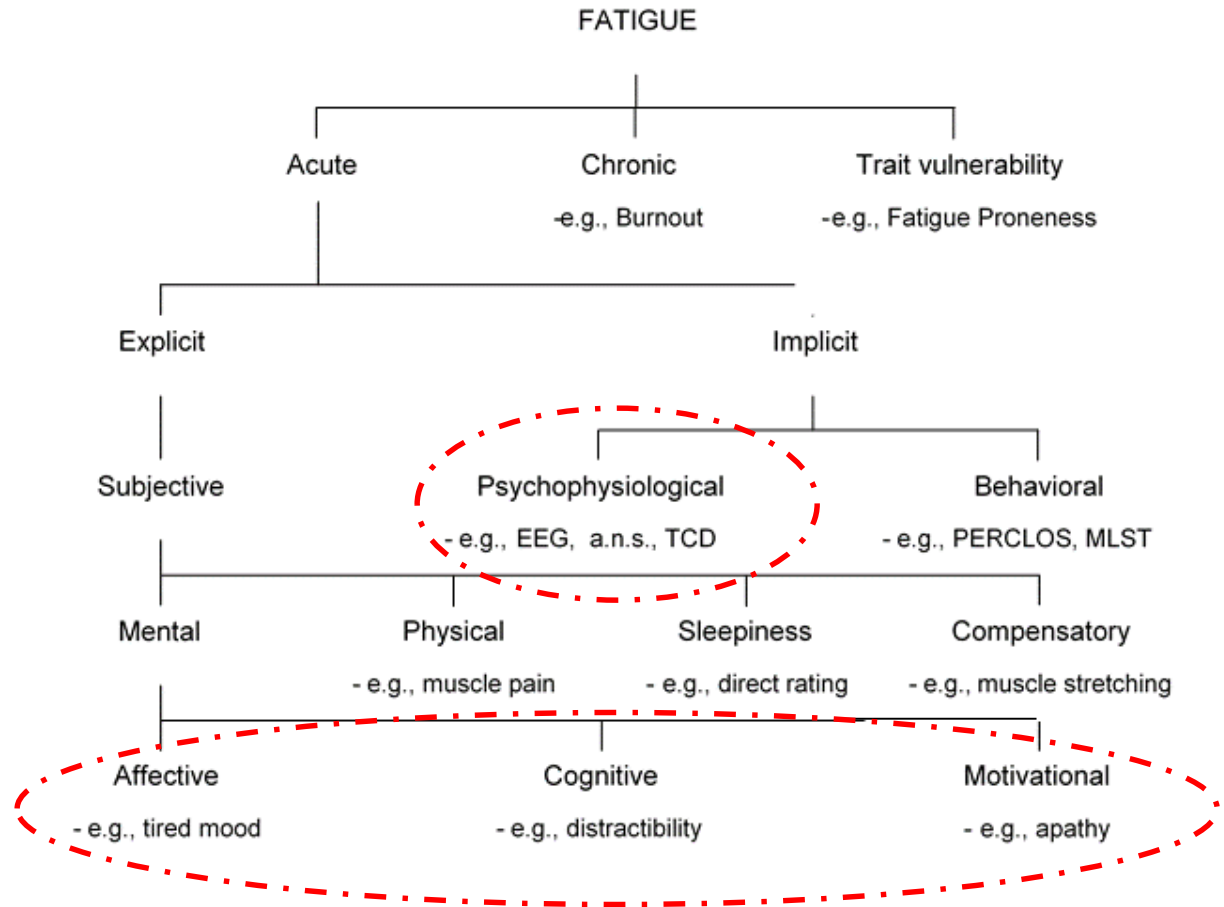
Overview

- Rationale for multivariate assessment strategy
 - Fatigue, workload and stress all have multiple components
- Components of acute subjective stress and fatigue
 - 3-D model of task stress (Matthews et al., 2002)
 - Application to performance prediction in UAV simulation
- Combining psychophysiology and subjective assessment
 - Multiple facets of workload response
 - Physiological and subjective predictors of performance in UGV simulation
- Implications



Facets of Fatigue

- Taxonomy of dimensions of fatigue (Matthews et al., 2012)
- Focus on acute, task-induced fatigue and individual differences in performance



Stress and Fatigue in Autonomous Vehicle Operation

- Task factors
 - Overload and underload
 - Poor interface design
 - Temporal: prolonged monitoring during ISR operations (vigilance)
- Increasing autonomy
 - Vehicle as 'team-mate'
 - Automation monitoring as a major operator function



Subjective Stress: Three Factor Model

(Matthews et al., 2002; Matthews, 2016)

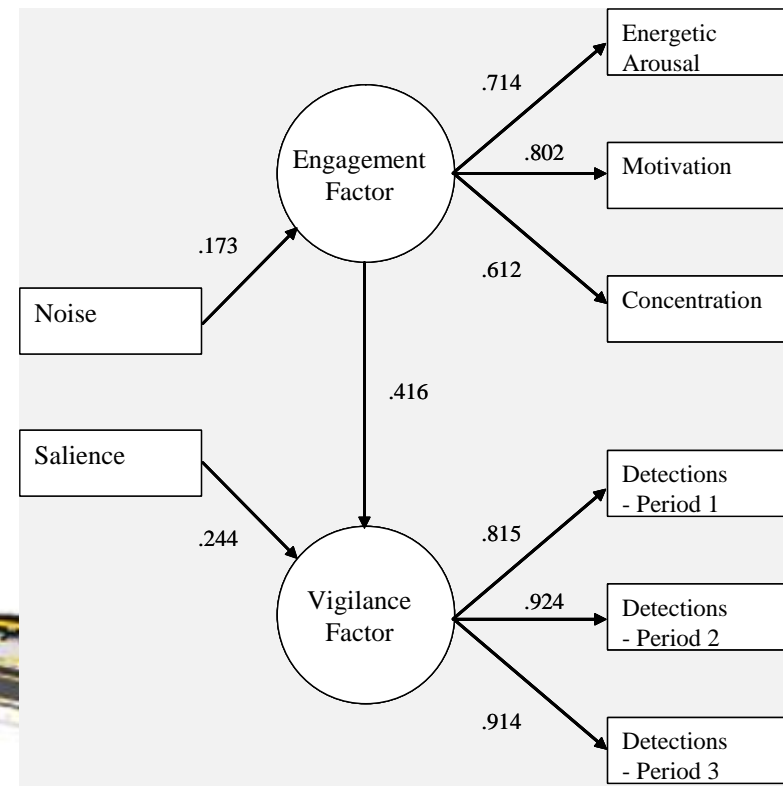
	Task Engagement	Distress	Worry
<i>Principal scales</i>	Energetic arousal Motivation (Intrinsic) Motivation (Success) Concentration	Tense arousal Low hedonic tone Low confidence	Self-consciousness Low self-esteem Cog. Interference (task-related) Cog. Interference (personal)

- General framework for understanding stress and fatigue in performance contexts
- Measurement with Dundee Stress State Questionnaire (DSSQ)
- Fatigue as low task engagement: tiredness, apathy, distractibility
- Only partial overlap with physiological metrics



Fatigue as Task Disengagement

- Task engagement predicts performance of demanding visual attentional tasks in multiple studies (e.g., Shaw et al., 2010)
 - e.g., vigilance, visual search, change detection, facial processing
 - Task engagement as a marker for attentional resource availability
 - Predicts across fatiguing and non-fatiguing tasks
- Engagement as a mediator of stressor effects
 - Effects of jet engine noise on vigilance (Helton et al., 2009)
 - Effects of cold infection on vigilance Matthews et al. (2016)



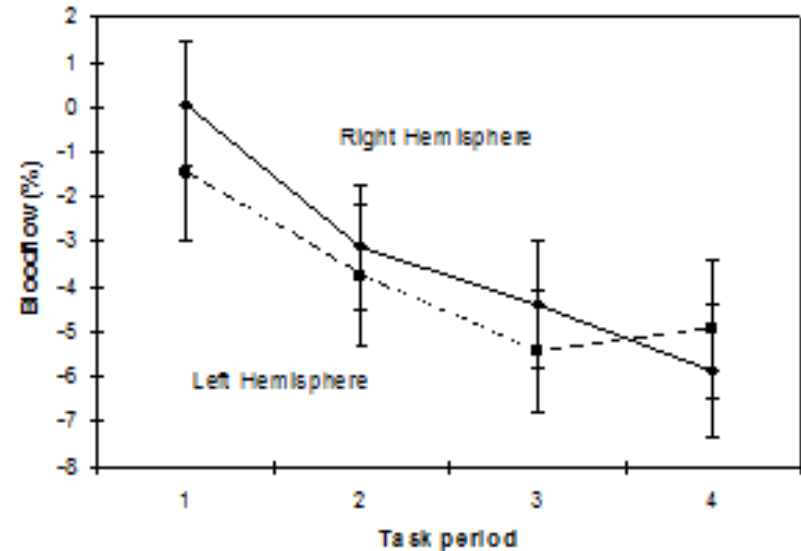
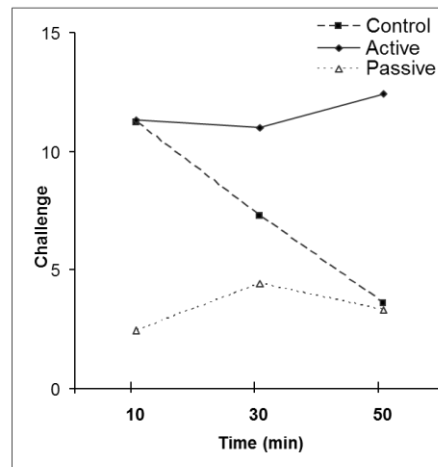
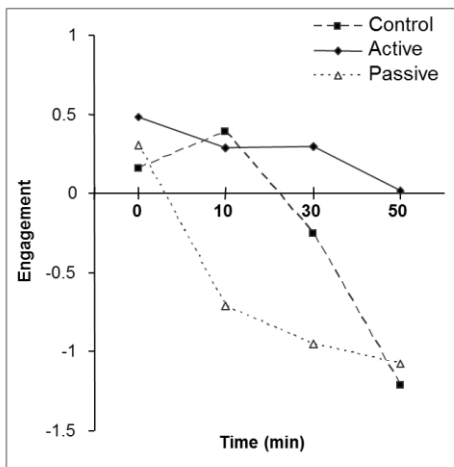
Workload Assessment: Subjective vs. Physio

- Workload is expressed by complex patterns of physiological response
- Dissociated from subjective workload

Study	Task or Manipulation	NASA-TLX effect (points)	Metrics for elevated workload (expected)	Metrics for reduced workload (unexpected)
UGV simulation	Change detection (vs. threat detection)	16.1	Lower HRV, higher rSO ₂ (fNIR), higher frontal theta, higher ICA	
UGV simulation	Dual-task (vs. single task)	10.0	Shorter fixation duration	
NPP simulation	Detection (vs. other tasks)	4.9	Higher rSO ₂ (fNIR)	Lower beta, gamma Lower heart rate Higher alpha
UAV simulation	Cognitive load (vs. low cog. load)	16.7	Higher beta Higher gamma	Higher HRV

Multiple Levels of Driver Fatigue

- Loss of task engagement is accompanied by:
 - Declining cerebral bloodflow velocity (CBFV)
 - Changes in cognitive processing (appraisal and coping)
 - Loss of performance
 - Example data from driving (Saxby et al., 2013; Reinerman et al., 2008)



Performance Prediction: Latent Factor Model

(Matthews et al., 2010)

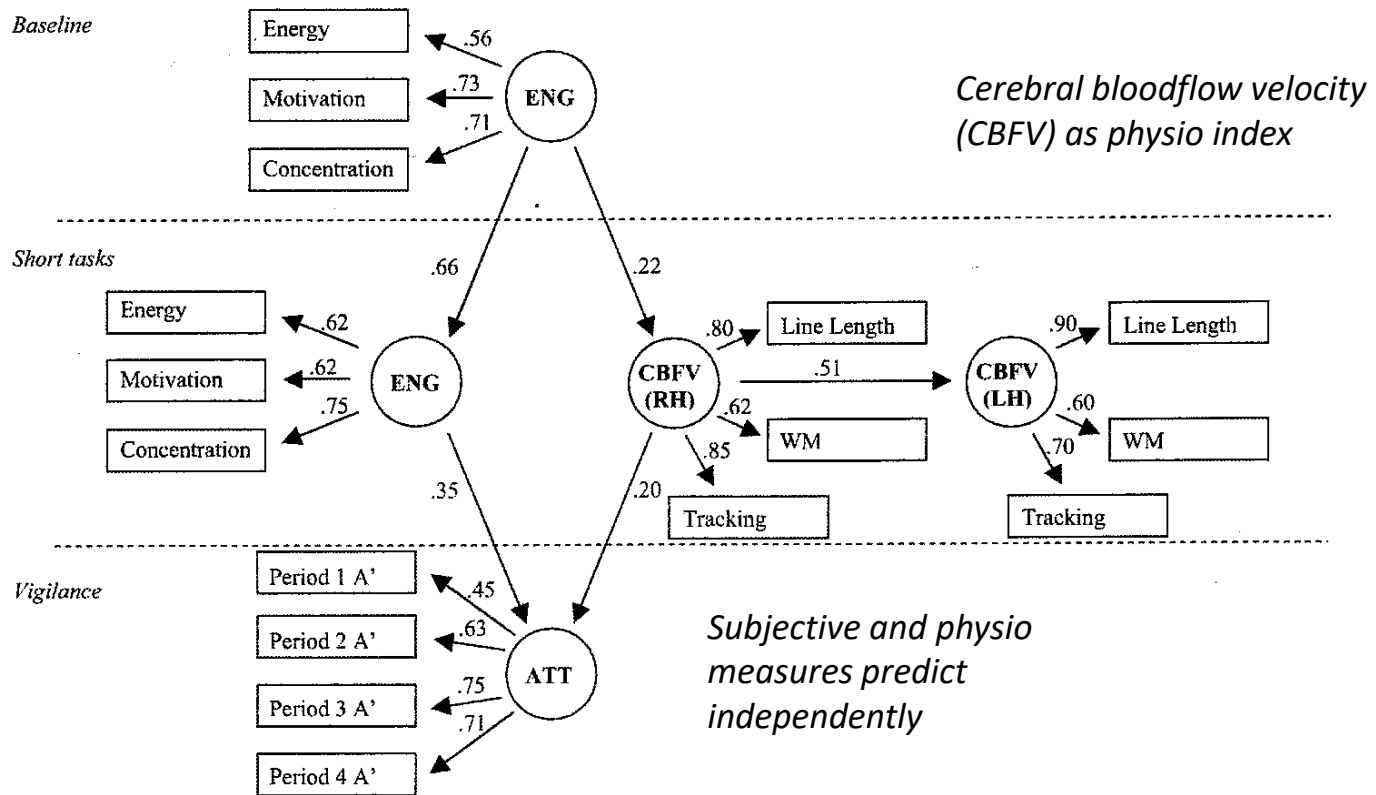
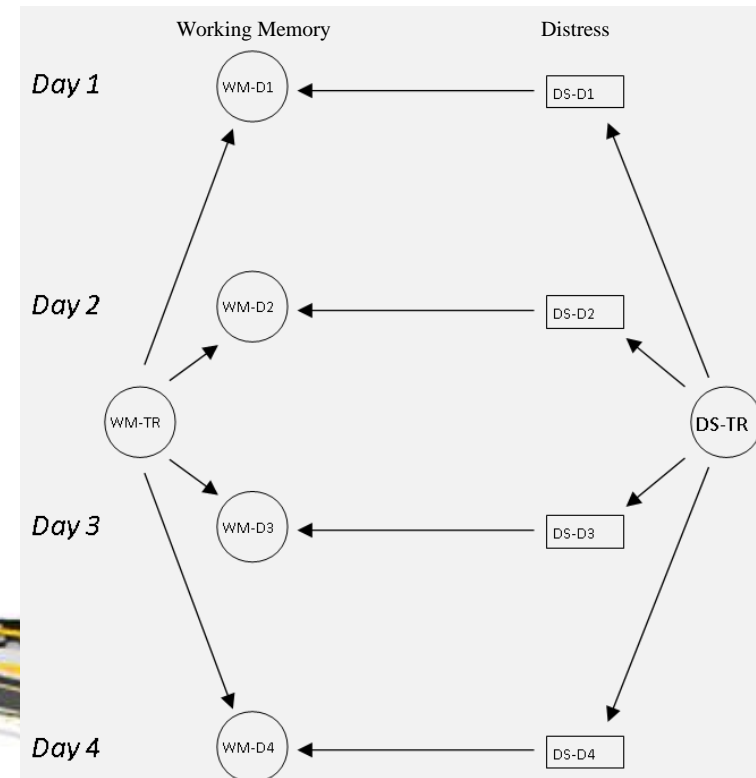


Figure 4. Latent factor model fitted to data from the sensory vigilance task. Eng = Engagement; Att = Sustained Attention; WM = Working Memory.

Distress as a Response to Fatigue

- High workload tasks reliably induce distress
 - Response to vigilance typically combines low task engagement and high distress
 - Distress correlates at $\sim .5$ with NASA TLX workload
- Distress correlates negatively with performance requiring multi-tasking and divided attention
 - e.g., working memory (OSPAN)
- Multivariate modeling of distress (Matthews & Campbell, 2010)
 - State variation matches working memory variation



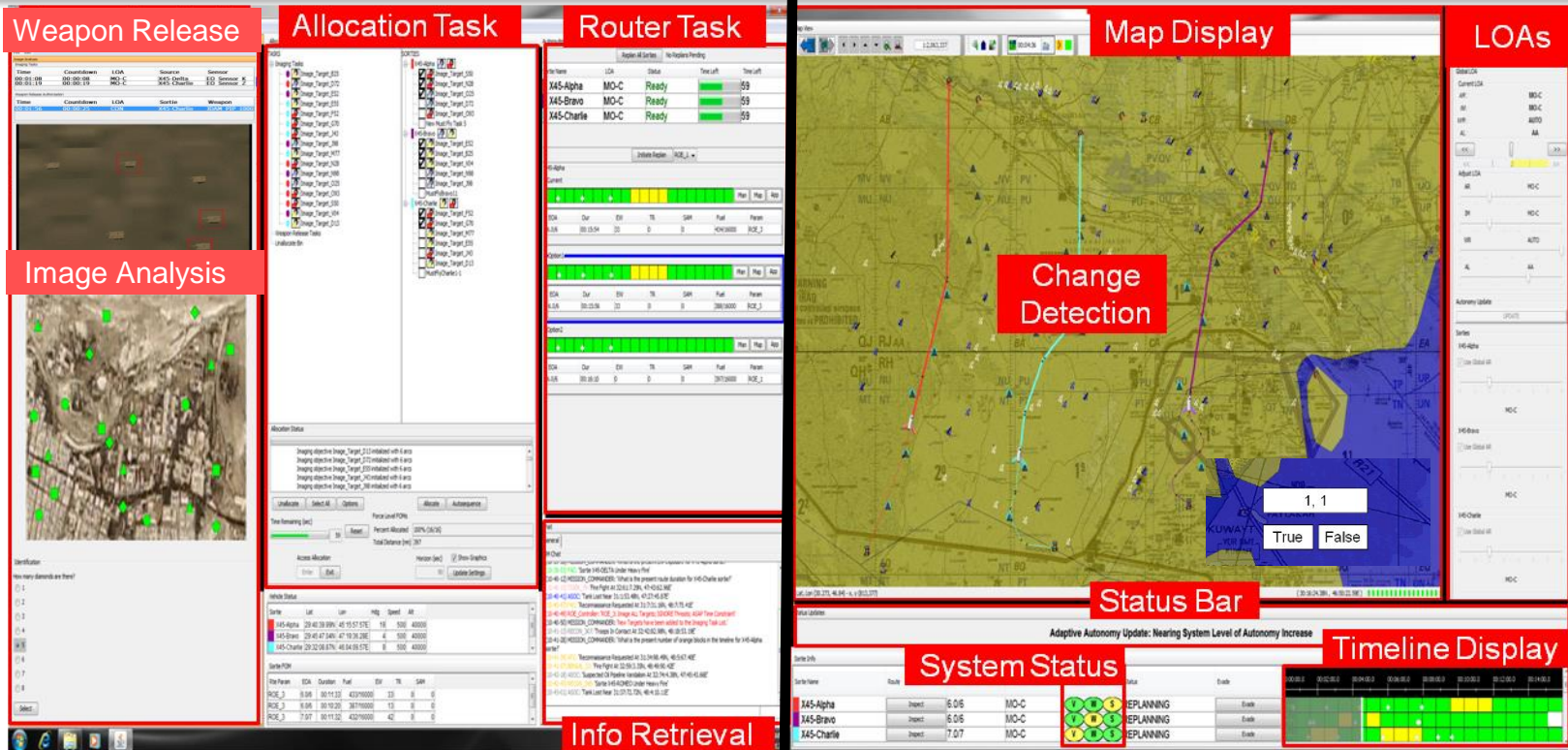
Multi-UAV Simulation: Adaptive Levels of Autonomy (Lin et al., 2015)

- Collaboration with AFRL (Gloria Calhoun, Greg Funke)
- UCF funding from AFOSR Trust and Influence program
- Aims
 - Investigate impact of task load on performance, subjective stress, and reliance on automation
 - Investigate impact of automation characteristics (LOA, reliability)
 - Investigate predictors of performance
 - Manipulations of task load

Method

- Participants
 - 101 UCF students (43 men, 58 women); mean age = 18.95 years
- Design
 - Manipulations of task load (and level of automation)
- Procedure
 - Pre-test questionnaires, including stress state
 - Training (about 30 min)
 - Main task (60 min)
 - Post-task workload and stress

Adaptive Levels of Autonomy (ALOA) Simulation



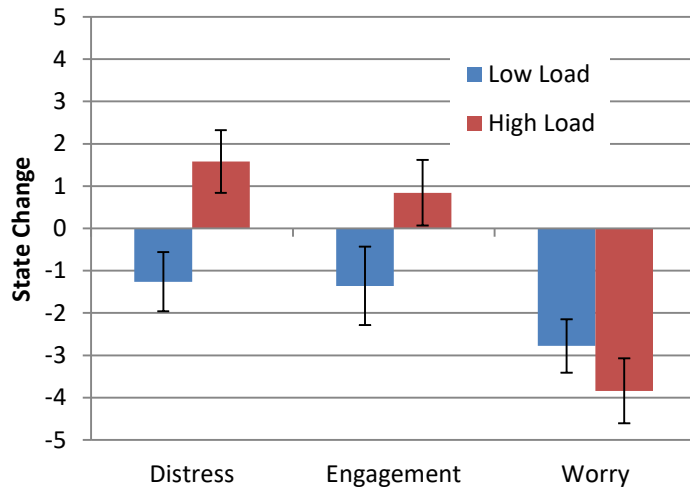
The screenshot displays the ALOA simulation interface with several key components highlighted:

- Weapon Release:** A table showing mission parameters for different weapons.
- Image Analysis:** A satellite-style image with green markers indicating detected targets.
- Allocation Task:** A list of tasks and their status, including 'Weapon Release Task' and 'Allocation Task'.
- Router Task:** A table showing task status for '145-Alpha', '145-Bravo', and '145-Charlie'.
- Map Display:** A topographic map showing mission routes and a 'Change Detection' area.
- LOAs (Levels of Autonomy):** A control panel for setting autonomy levels for various entities.
- Status Bar:** A bar at the bottom of the map display showing 'Adaptive Autonomy Update: Nearing System Level of Autonomy Increase'.
- System Status:** A table showing the status of '145-Alpha', '145-Bravo', and '145-Charlie'.
- Info Retrieval:** A section at the bottom left providing detailed information about the system.
- Timeline Display:** A Gantt-style chart showing the execution timeline of various tasks.

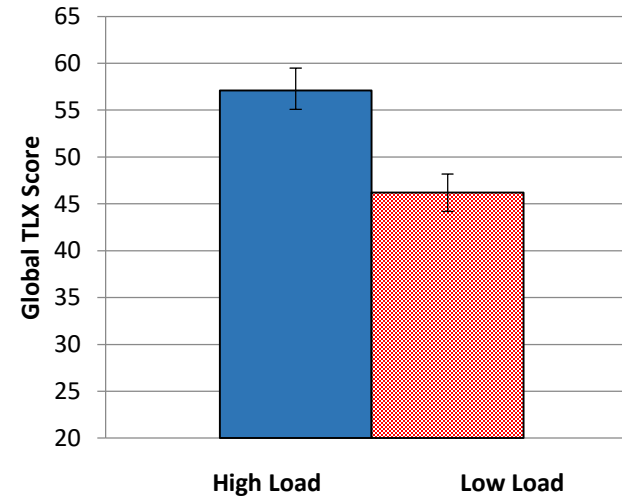
- Multiple sub-tasks on two displays (Calhoun et al., 2011)
- ISR (signal detection) tasks embedded for primary performance assessment
- Automation manipulated for signal detection (weapon release, image analysis)
- Selected tasks used to manipulate workload

Results: Task Load and Stress

- Higher distress and workload (TLX) under high task load
 - Manipulation working as intended



N.B. Post-task – pre-task scores



Stress State and WR/IM Performance - In High Task Load Condition

	Image Analysis			Weapon Release		
	Distress	Engagement	Worry	Distress	Engagement	Worry
Accuracy	-.33*	.14	-.29*	-.41**	.20	-.17
Reliance	-.16	.04	-.24	-.09	-.01	-.20
Neglect	.33*	-.41**	.25	.41**	-.31*	.18
*P<.05, **P<.01						

- Distress most damaging element of state
 - Due to multi-tasking requirement
- Low task engagement (fatigue) associated with neglect

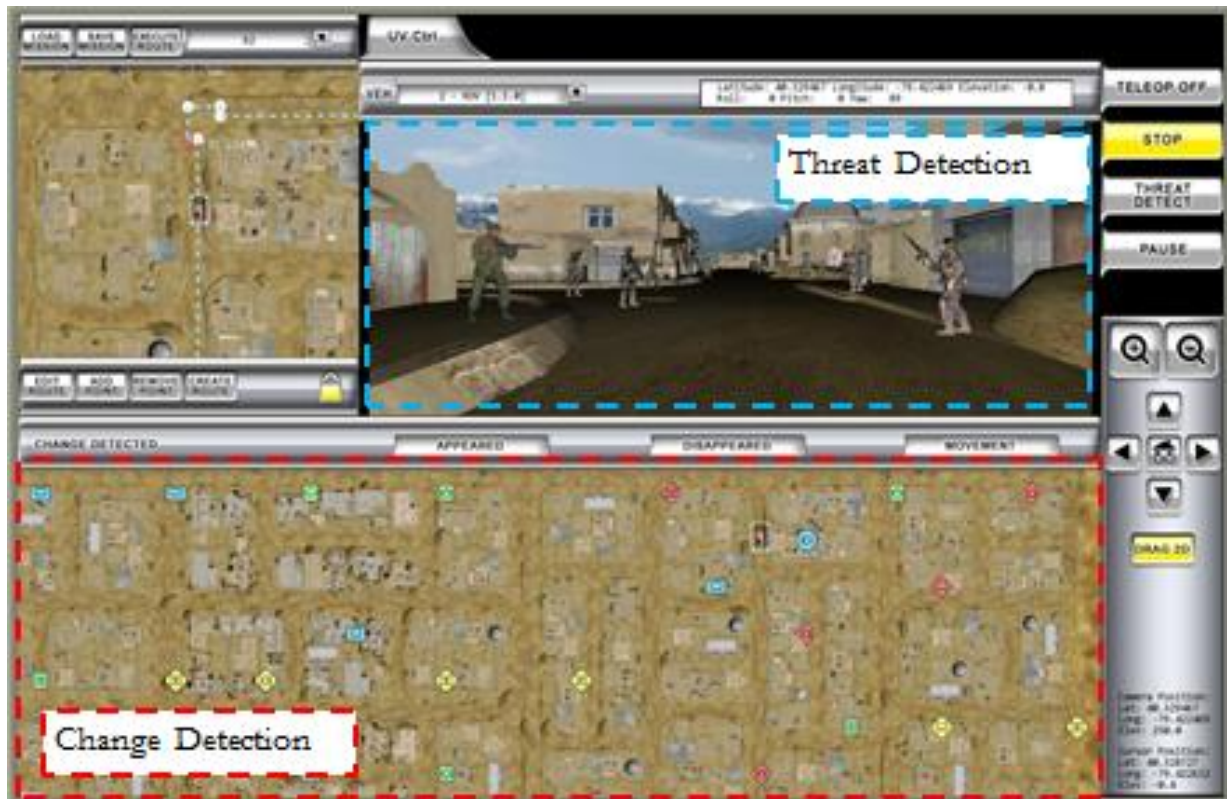
Stress and Workload in UGVs

(Matthews et al., 2015; in press)

- MIX Sim: Remote operation of UGV for ISR
- Physio: Measurement of multiple workload responses
- Aims
 - Is there a unitary physiological workload/stress response?
 - How does stress response correlate with attention and performance?

UGV: MIX Simulation

- MIX testbed: Simulation of OCU of UGV (Taylor et al., 2013)



Task type:

- Change Detection (CD) is higher workload than Threat Detection (TD)

Dual-tasking:

- Dual vs. single task performance

Event rate:

- Calibrated for each task

Method

- Participants: 85 M, 66 F. Mean age = 19.57
- Design
 - Within-subjects: completion of four task scenarios varying in workload
 - Three event rates, varied within scenario

Scenario 1	Scenario 2	Scenario 3	Scenario 4
Change Detection	Change Detection + Threat Detection at constant rate	Threat Detection	Threat Detection + Change Detection at constant rate

- Psychophysiology: 5 min baseline + continuous monitoring
- Workload: NASA-TLX after each task condition (x12)
- Stress state: DSSQ after each task condition (x12)

Physio: Workload Metrics

- Simultaneous recording of multiple metrics

Response system	Metrics
Electrocardiogram (ECG: Mulder, 1992)	Mean heart rate (HR) Heart rate variability (HRV)
Electroencephalogram (EEG: Borghini et al., 2012; Gevins & Smith, 2003)	Frontal theta Alpha Beta
Transcranial Doppler Sonography (TCD: Warm, Tripp, Matthews & Helton, 2012)	Cerebral bloodflow velocity (CBFV) in medial cerebral arteries (bilateral)
Functional Near Infra-Red (fNIR: Warm et al., 2012)	Regional cerebral oxygen saturation (rSO ₂) in prefrontal cortex (bilateral)
Oculometric indices (Jacob & Karn, 2003; Marshall, 2002)	Duration of fixations Pupillometric Index of cognitive activity (ICA)
Subjective (Hart & Staveland, 1988)	NASA-TLX overall score

Results: Metric Correlations

	ECG IBI	ECG HRV	EEG Theta	EEG Alpha	EEG Beta	CBFV Left	CBFV Right	rSO ₂ Left	rSO ₂ Right	Eye Fix. Duration	Eye ICA	NASA-TLX
ECG IBI	-	.53**	-.29**	-.14	-.24**	.01	-.15	-.01	.01	.10	.06	-.02
ECG HRV		-	-.09	-.16	.12	.01	-.09	-.02	-.01	-.06	.05	.11
EEG Theta			-	.68**	.65**	.14	.16	-.13	-.13	.05	-.02	.14
EEG Alpha				-	.36**	.18*	.21*	-.10	-.11	-.01	-.04	.19*
EEG Beta					-	.17	.15	-.03	-.06	.06	.09	.05
CBFV Left						-	.61**	-.10	.00	-.02	.02	-.01
CBFV Right							-	.03	-.01	.01	.00	-.08
rSO ₂ Left								-	.68**	.06	-.08	-.06
rSO ₂ Right									-	.02	.03	-.01
Eye Fix. Duration										-	-.34**	-.02
Eye ICA											-	.03
p < .05, **p < .01.												

Results: Metric Correlations

- Correlations across sensor systems mostly zero
 - ECG IBI vs. EEG theta ($r = -.29, p < .01$) and beta ($r = -.24, p < .01$)
 - EEG alpha vs. left ($r = .18, p < .05$) and right ($r = .21, p < .05$) CBFV.
- Five factors defined by sensor system:

	Factor				
	1 (EEG)	2 (fNIR)	3 (TCD)	4 (Eye)	5 (ECG)
ECG IBI	-.20	.02	.02	.06	.80
ECG HRV	.14	-.02	-.03	-.08	.91
EEG Theta	.90	-.06	-.03	.04	-.10
EEG Alpha	.83	-.02	.05	.05	.23
EEG Beta	.73	.08	.03	-.08	-.18
CBFV Left	-.01	-.04	.91	-.03	.05
CBFV Right	.02	.03	.89	.03	-.07
rSO ₂ Left	.01	.90	-.03	.07	-.02
rSO ₂ Right	-.01	.91	.02	-.06	.03
Eye Fixation Duration	.06	.04	-.01	.81	.04
Eye ICA	.04	.03	-.01	-.82	.06

Correlates of Performance

- All measures averaged across condition
- ECG, EEG and eye fixation duration as physio correlates
- Low distress and worry, high engagement as DSSQ correlates
- Regression models: Physio and subjective state predict independently

	ECG	EEG		Eye	DSSQ		
	HRV	Alpha	Beta	Fix. Dur.	Dist.	Eng.	Worry
Change Detection	-.258**	-.149	-.259**	.262**	-.392**	.451**	-.302**
Threat Detection	-.074	-.172*	-.154	.161	-.214**	.280**	-.247**
Workload (TLX)	.112	.144	.190**	-.024	.611**	-.167*	.080

Note: *p < .05, **p < .01

UGV Study: Conclusions

- No unitary physio response
- Physio and subjective responses dissociate
- No single metric adequately captures response
- Multiple predictors of performance
 - Subjective state and physio independently predictive
- Multivariate assessment needed for system evaluation

Implications

- Evaluating system design requires multiple metrics for workload, stress and fatigue
- Physio and subjective assessments are both diagnostic of performance issues
 - Neither easily substitutes for the other
 - Dimensions critical for performance vary across domains and vary with cognitive demands
- Physio analogues for task engagement critical for operator diagnostic monitoring
- Evaluation of operator response pattern may guide selection and training