



Measuring Attention, Working Memory, and Visual Perception to Reduce the Risk of Injuries

Presented by:
Behzad Esmaeili, Ph.D.

SARMAD

Safety, Risk Management, And Decision-making Research Group

Overview of Research Group

Mission Statement: We believe that optimal decisions in the construction industry emerge out of scientific evidence and empirical data. To realize such a paradigm shift, SARMAD:

- (1) Promotes original, unbiased, and innovative research that addresses the global challenges in today's construction industry;
- (2) Integrates multiple scientific perspectives to unearth transformative and breakthrough discoveries; and
- (3) Develops state-of-the-art decision-support systems, tools, and applications that enable agile and rapid industrial transformation

Vision Statement: SARMAD envisions a construction industry that is: safe for its workers; sustainable for the environment; ethical for our society; creates value for its customers; on time, on budget, and productive for its project managers; collaborative for its stakeholders; and profitable for its constructors.



Motto: *Set the standard in quality and innovation!*

Construction Safety, Risk Management, & Decision Making

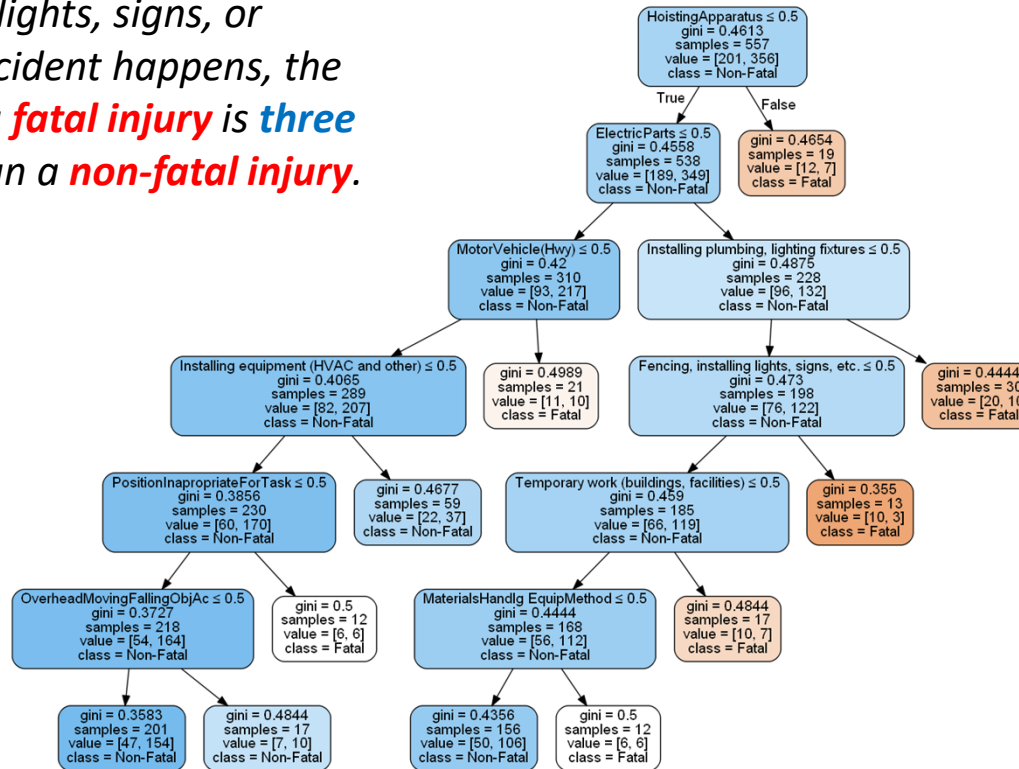
- Data Analytics
- Human Error
- Ergonomics
- Cyber-Physical-Human Systems
- Visualization and Training
- Project Performance

1. Data Analytics

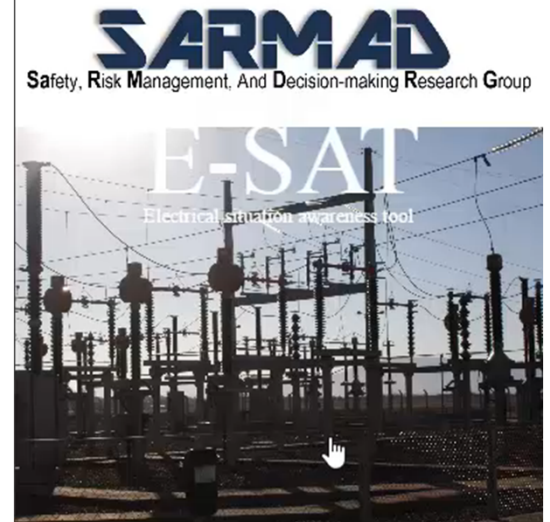
Data Analytics

Accident scenario:

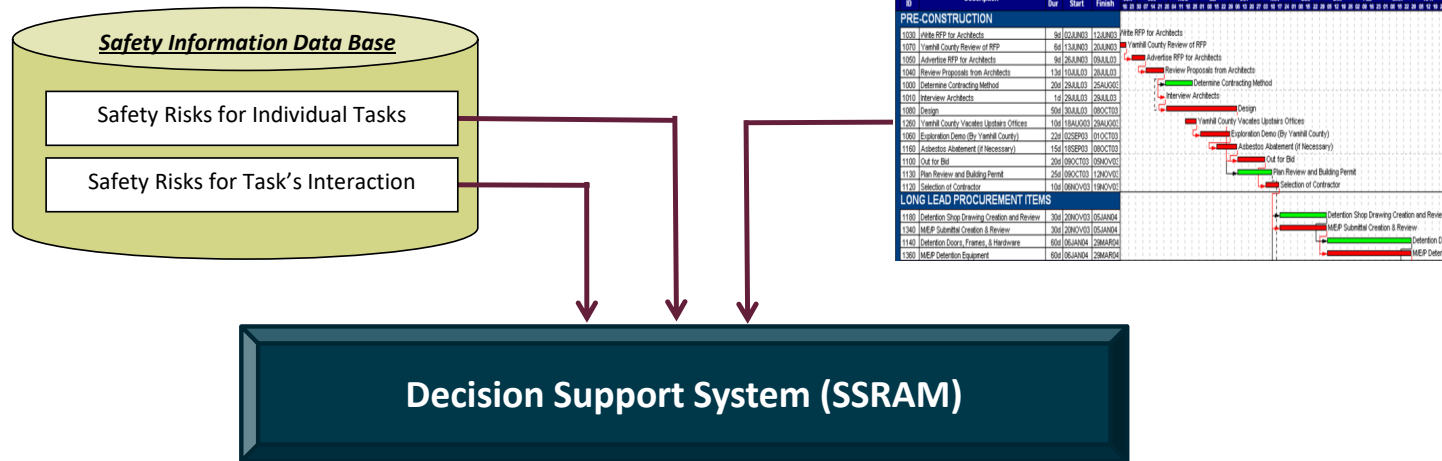
If an employee is working on electric parts to install lights, signs, or fences, if an accident happens, the probability of a **fatal injury** is **three times** more than a **non-fatal injury**.



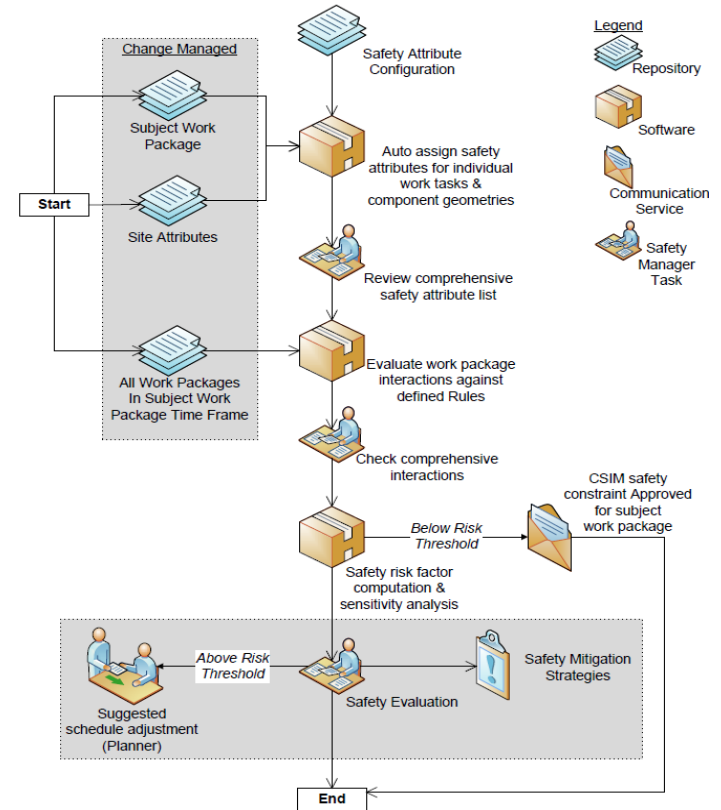
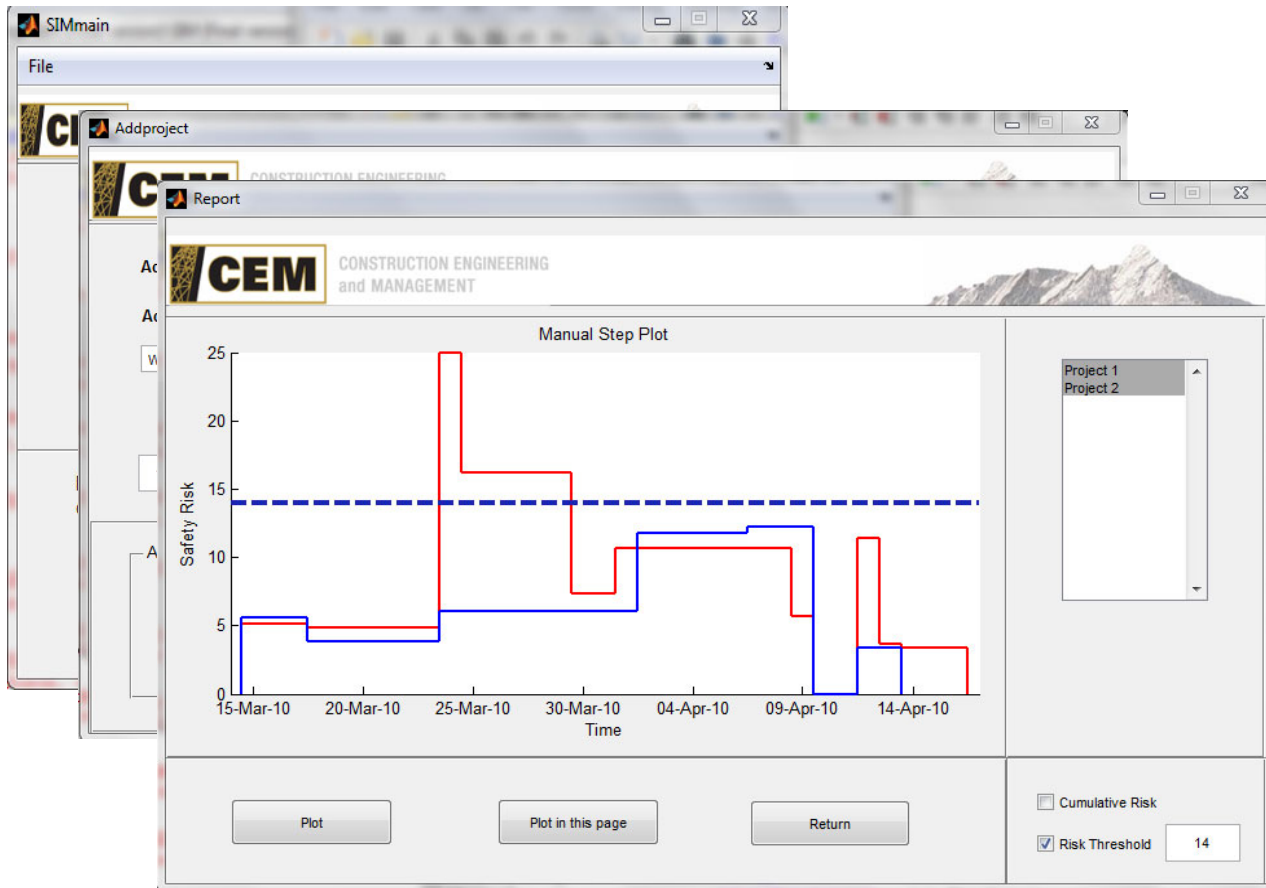
Fatal 
 Non-fatal 

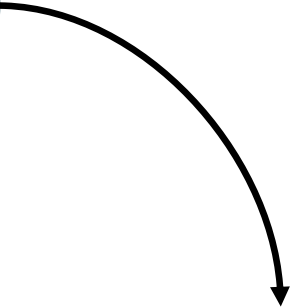
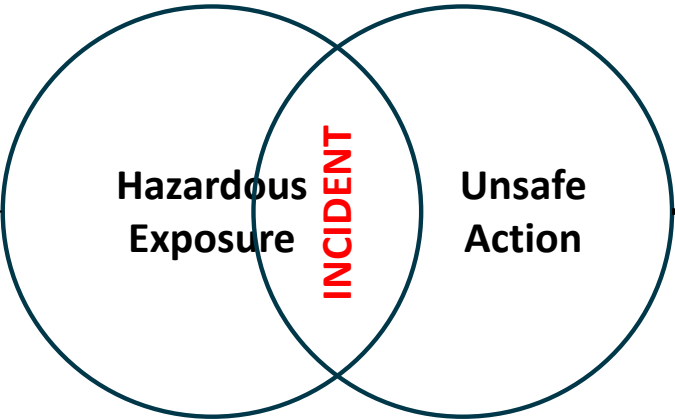


Attribute-Based Risk Management



Attribute-Based Risk Management





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2. Human Error

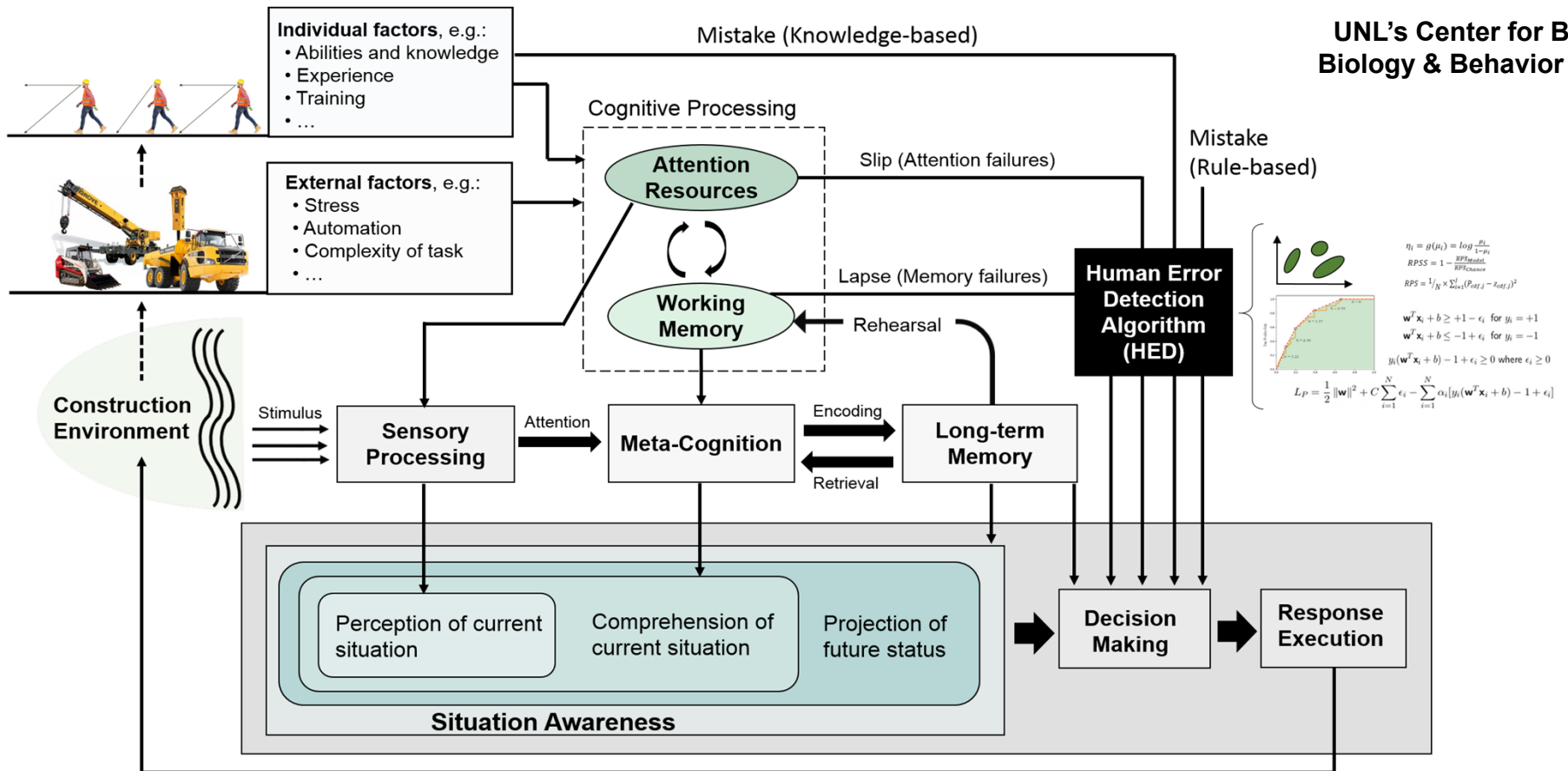
Problem Statement



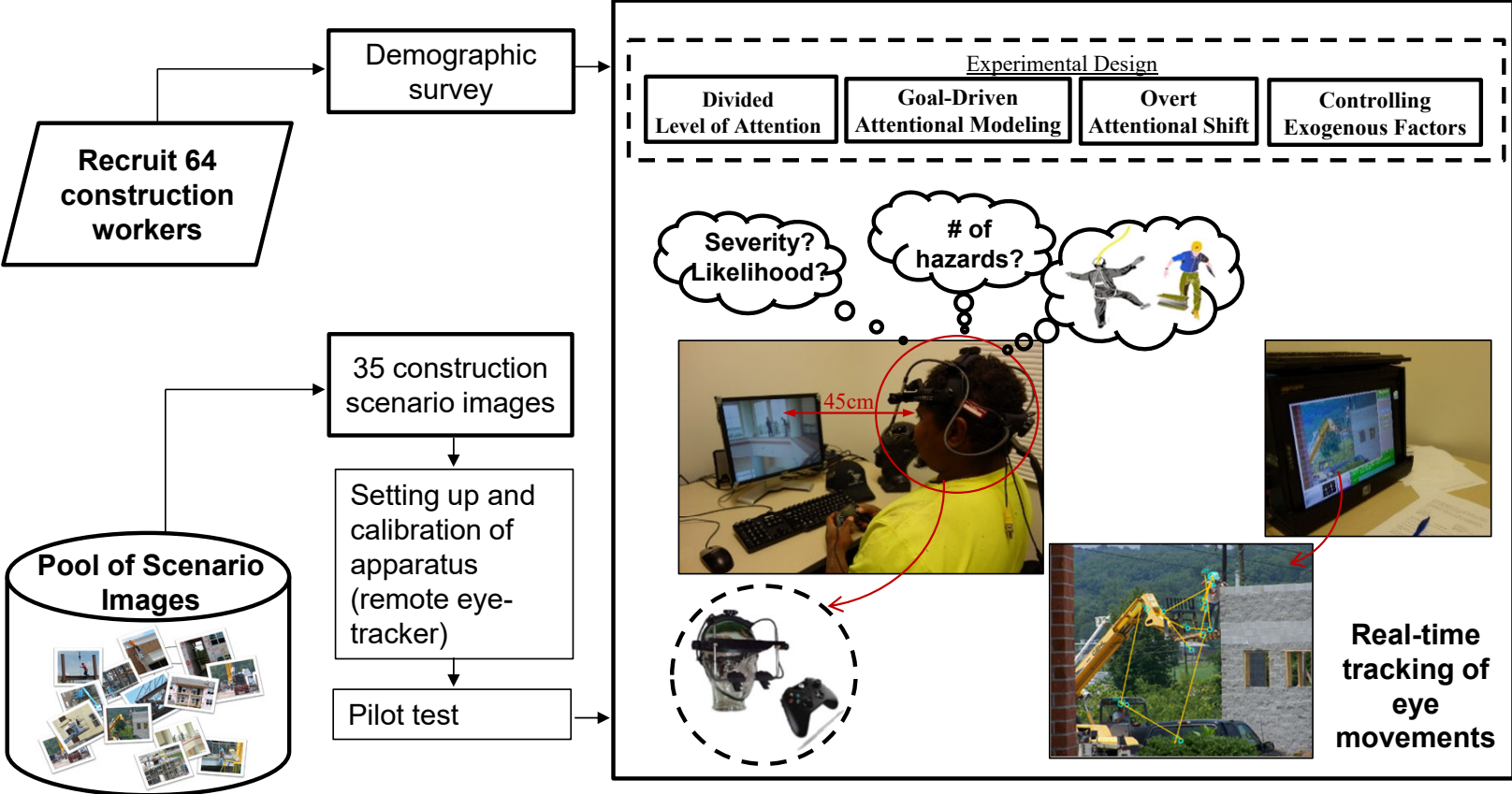
Human Error Detection (HED) Framework



UNL's Center for Brain, Biology & Behavior (CB3)



Attention and Human Error

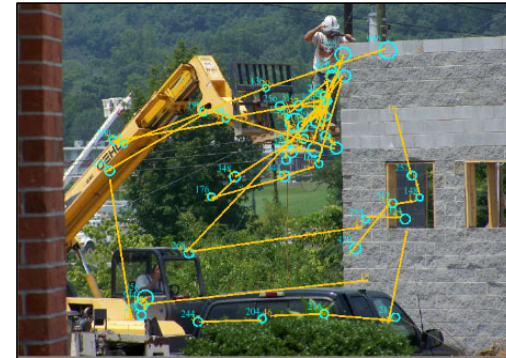


Safety Training: *Receiving traditional safety training has minimal impact on workers' attentiveness to hazards.*



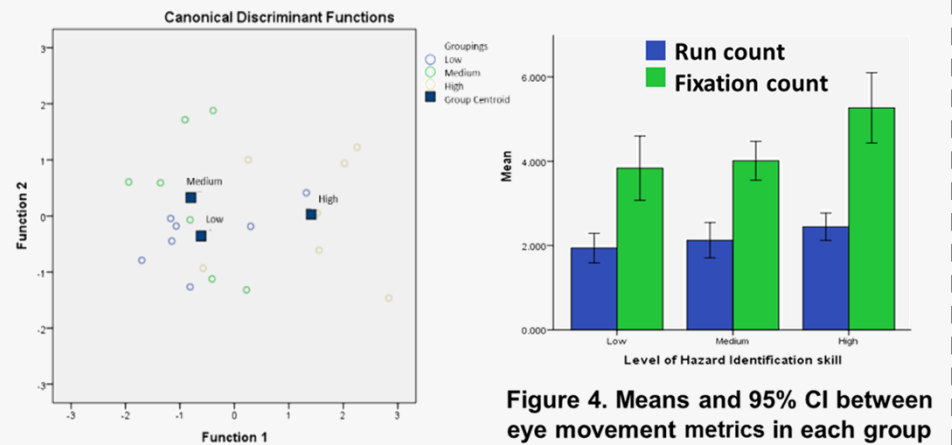
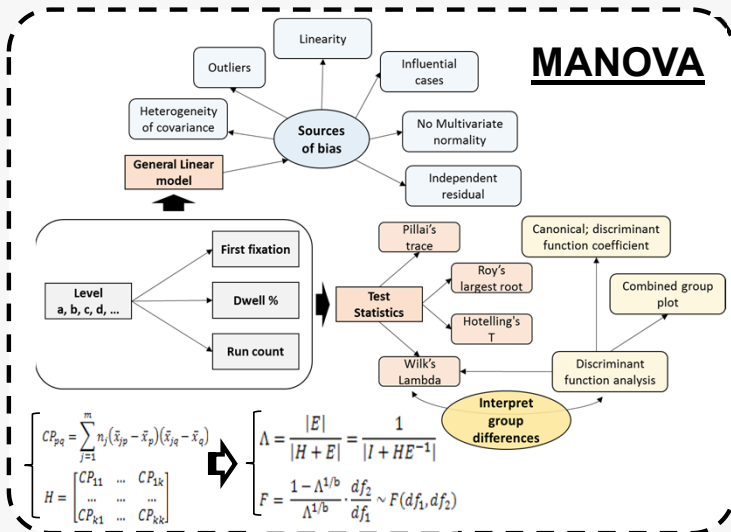
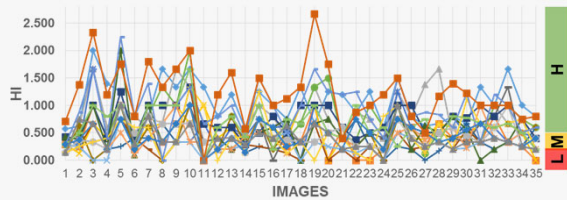
Attentional distribution of groups: with (left) vs. without training (right)

Working Experience: *As a worker gets more experience in construction field, s/he will be more aware of safety requirements in jobsite.*



Cognitive process of: G1=<5 yrs (left), G2: >10 yrs (right) of experience

$$HII_j = \frac{H_i}{H_{tot}} \longrightarrow HII_{T,j} = \text{Average}(HII_j)$$



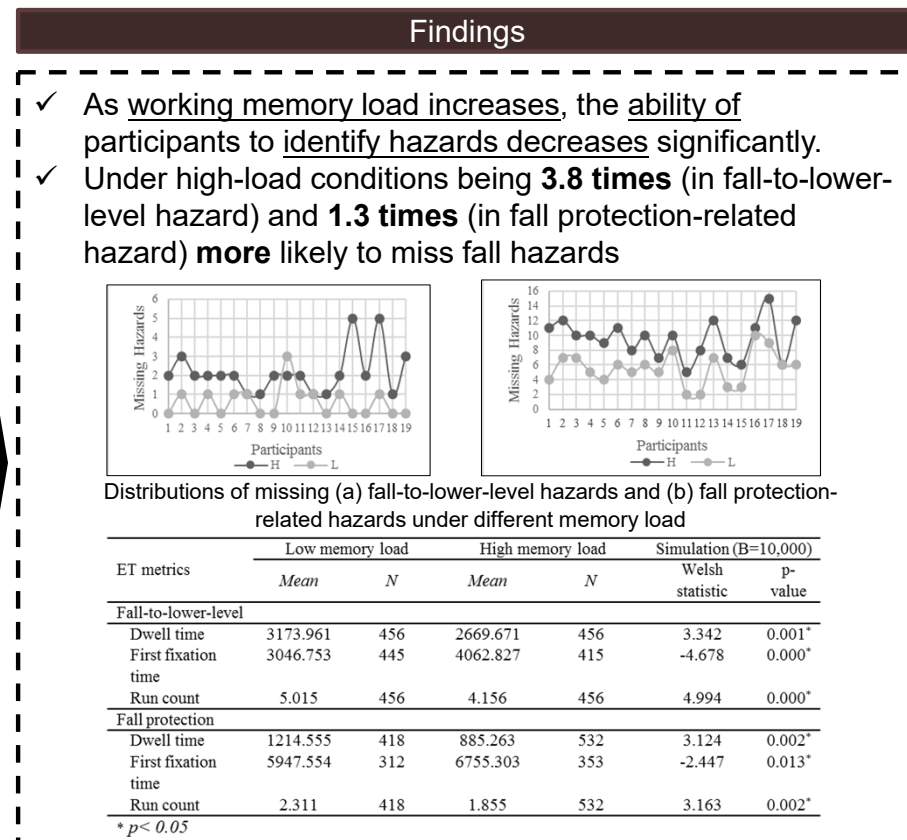
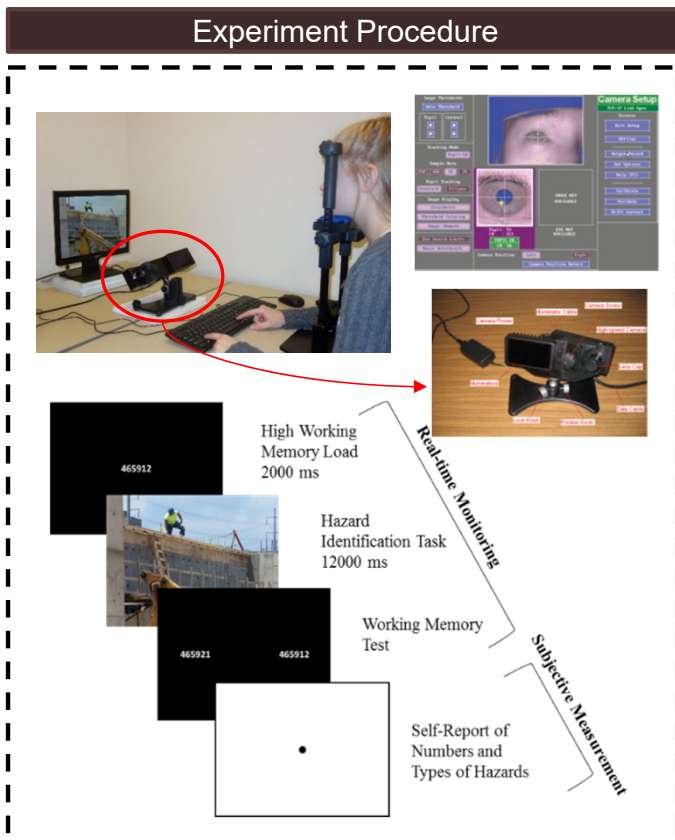
Source of Hazard	Value	F	df	Error df	Sig.
Ladder	0.633	2.622	6	34	0.034*
Fall to Lower Level	0.651	2.738	6	34	0.028*
Fall-Protection	0.714	3.149	6	34	0.015*
Struck-By	0.732	3.268	6	34	0.012*
Housekeeping	0.642	2.682	6	34	0.031*
All Hazard Types	0.622	2.559	6	34	0.037*

*p ≤ 0.05

✓ Using Pillai's trace, worker's level of hazard identification skill has significant effect on their eye movements metrics. $V = 0.622$, $F(3, 34) = 2.559$, $p = 0.037$.

Working Memory and Human Error

- ✓ **Goal:** Measuring the Impact of Working Memory Load on the Safety Performance of Construction Workers



Risk Taking Behavior

Safety-Risk Perception Measurement

Phase I

Statement-based questionnaire to measure risk perception of construction workers

- Risk statements
- 1- Working on ladders
 - 2- Working near an unprotected edge (roof)
 - 3- Working near an unprotected opening
 - 4- Working near a skylight
 - 5- Working on a scaffold
 - 6- Working on structural frames (e.g., steel frames)
 - 7- Working on an aerial platform
 - 8- Working/standing on heavy equipment
 - 9- Working on a slippery surface
 - 10- Working on an unsecured or unstable surface
 - 11- Working on an uneven working surface

Risk perception score = (Frequency Score) x (Severity Score)

Phase II

Scenario-based questionnaire to measure risk perception of construction workers



Scenario Image #1

Scenario #1.1. Assume that one of your colleagues in the worker who is working on the elevated platform as shown in the image. He is inspecting the installation of panel facades and helping to pass a wire through one of the windows. He will continue working in this situation for about 15 minutes.

From the best of your knowledge and experience of construction sites (if it applies), in this situation, what will be the frequency of each of the following outcomes?

Injury Outcomes	Very Unlikely	Unlikely	Neutral	Likely	Very Likely
First aid	0	0	0	0	0
Medical case	0	0	0	0	0
Lost work time	0	0	0	0	0
Permanent disability	0	0	0	0	0
Fatality	0	0	0	0	0

Scenario #1.2. Do you have any concerns about the construction safety issues in this scenario? If yes, please specify.

Scenario #1.3. If you were in your colleague's place, how likely would you stop working in this situation and report the unsafe working condition to your employer?

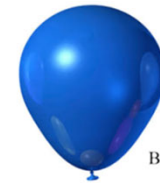
	Very Unlikely	Unlikely	Neutral	Likely	Very Likely
	0	0	0	0	0

Scenario #1.4. Assume that your colleague is doing the same activity, but this time he is asked not to do a safe activity. How would you assess the frequency of each of the following outcomes in this safe situation?

Injury Outcomes	Very Unlikely	Unlikely	Neutral	Likely	Very Likely
First aid	0	0	0	0	0
Medical case	0	0	0	0	0
Lost work time	0	0	0	0	0
Permanent disability	0	0	0	0	0
Fatality	0	0	0	0	0

Risk-Taking Behavior Measurement

Balloon Analogue Risk Task (BART):
Modeling real-world risk behavior through the conceptual frame of balancing the potential for reward versus loss.



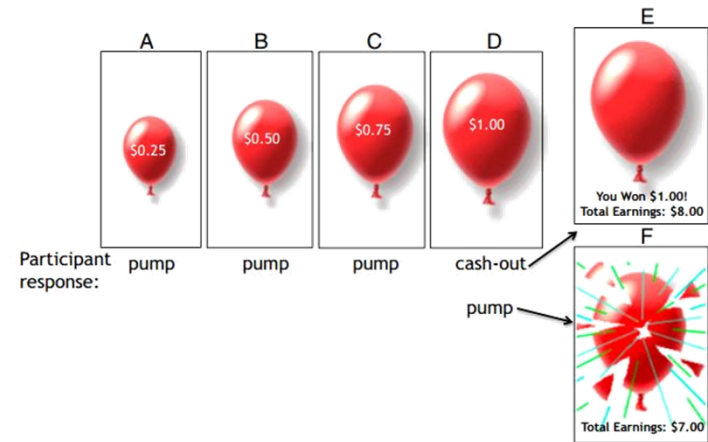
Number of pumps:

Potential earnings:

Balloon #: 10

Collect \$\$\$

Pump the balloon

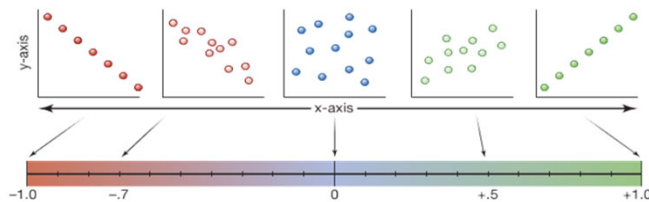


Risk Taking Behavior

Null hypothesis 1 (H_{01}): There is no association between risk perception and the risk-taking behaviors of construction workers.

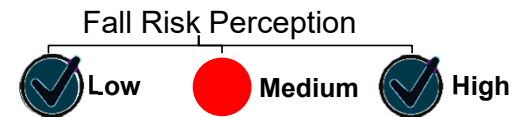
Null hypothesis 2 (H_{02}): There is no significant difference in the risk-taking behaviors of individuals with different risk perception.

Correlational Analysis



- Risk perception is significantly related to risk-taking behaviors ($p\text{-value} = 0.026 < 0.05$).
- Workers who have lower risk perception on construction sites generally engaged in more risk-taking behaviors than those who have higher risk perception.

Permutation Analysis



- Risk-taking behavior score of the subjects with high-risk perception was moderately statistically different from the risk-taking behavior score of the subjects with low-risk perception (Welsh t-statistic = 1.91; $p\text{-value} = 0.08 < 0.1$).

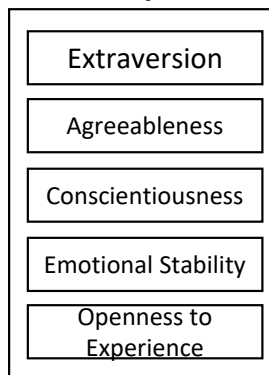
Personality and Human Error

Null hypothesis 1 (H_{01}): There is no association between Big Five personality traits and workers' attentiveness (fixation count or run count) to fall hazards.

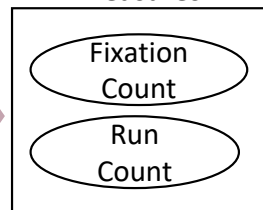
Null hypothesis 2 (H_{02}): Workers' Big Five personality traits have no impact on their attentiveness (fixation count or run count) to fall hazards on a construction site.

Correlational Analysis

Personality Dimensions

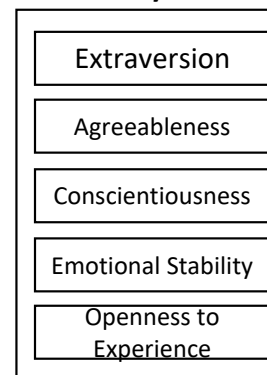


Visual Attention Measures

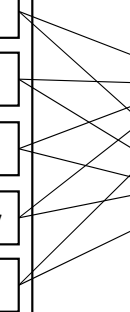
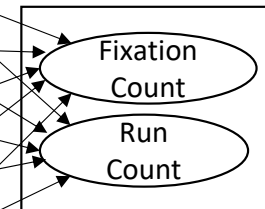


Permutation Analysis

Personality Dimensions



Visual Attention Measures

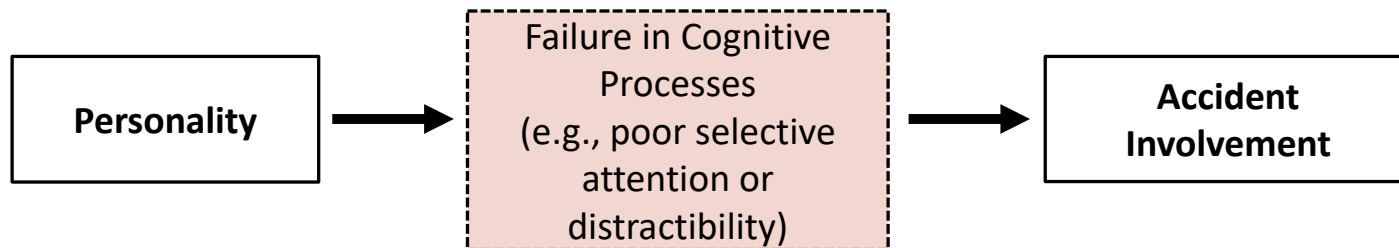


Personality and Human Error

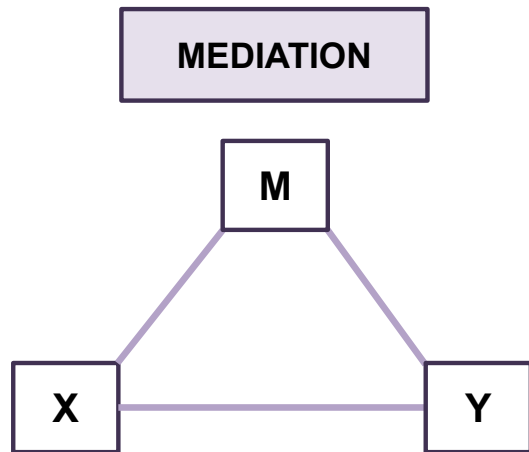
(a) Traditional Approach



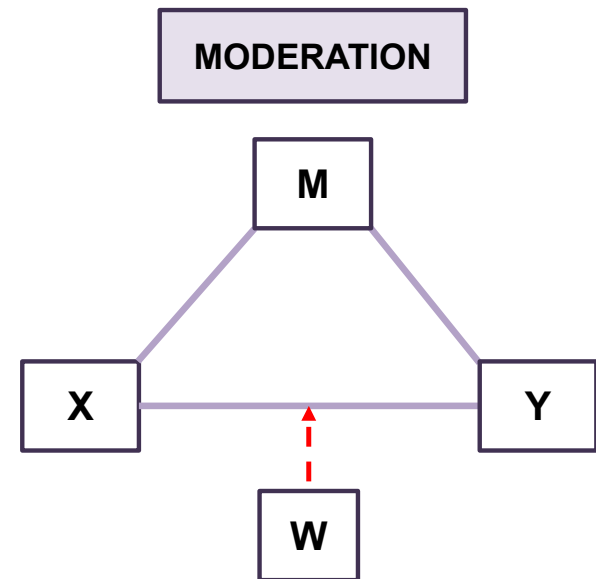
(b) Alternative Approach



Mediation & Moderation

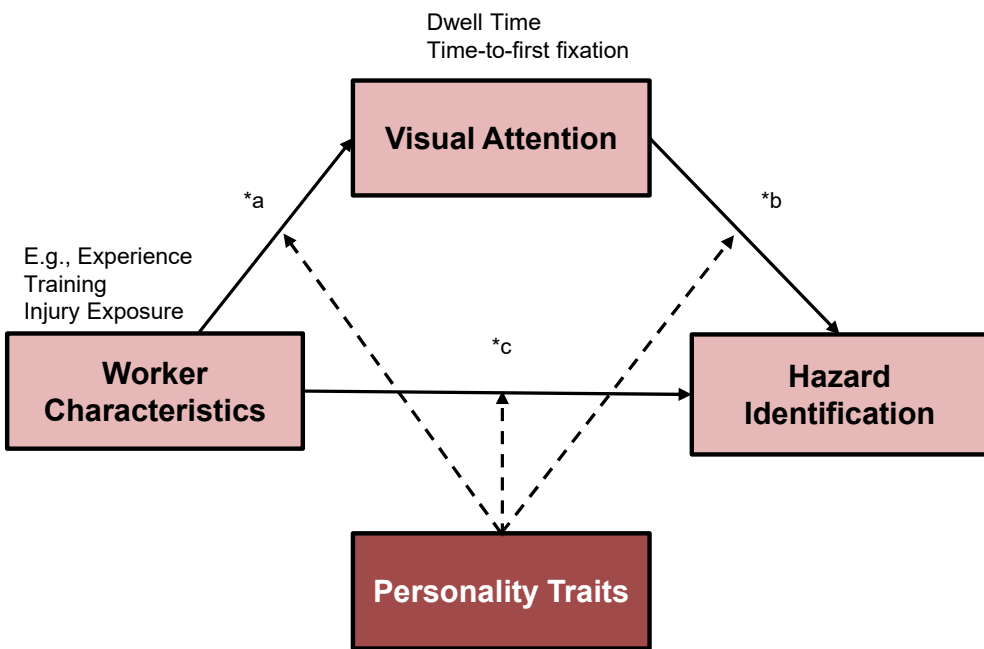


M transmits the effect of X on Y



W influences the strength of association between X and Y

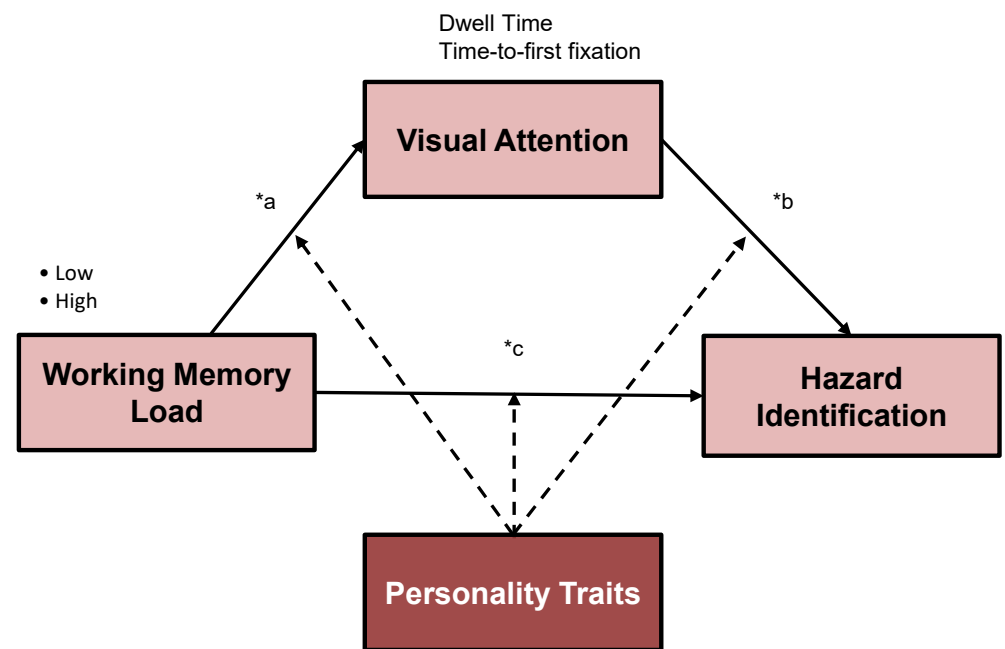
Moderated Mediation



*Paths a, b, c

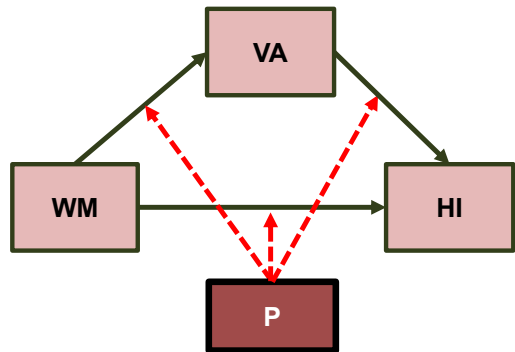
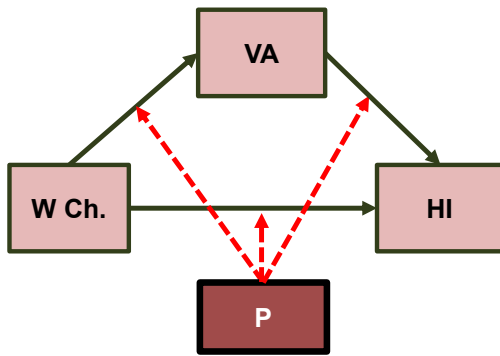
i Each participant has a unique entry of variables

Multilevel Modeling



i Data from multiple participants nested in both low and high WM groups

Mediation & Moderation



PREDICTOR	OUTCOME	MODERATOR	B	P-VALUE	L-CI	U-CI
Work Exp	TTFF	Consc. (L)	0.024	0.009*	0.006	0.041
Work Exp	TTFF	Open. (H)	0.019	0.018*	0.003	0.034
D. Time	HI	Consc. (H)	0.076	0.020*	0.013	0.139
TTFF	HI	Consc. (H)	0.244	0.010*	0.062	0.427

PREDICTOR	OUTCOME	MODERATOR	B	P-VALUE	L-CI	U-CI
WM	TTFF	Agreeableness	-2991.36	0.005*	-4994.38	-988.34
WM	TTFF	Conscientiousness	-3281.87	0.007*	-5584.53	-979.21
TTFF	HI	Agreeableness	-0.151	0.012*	-0.283	-0.046
TTFF	HI	Conscientiousness	-0.165	0.015*	-0.314	-0.049

W Ch.: Worker Characteristics

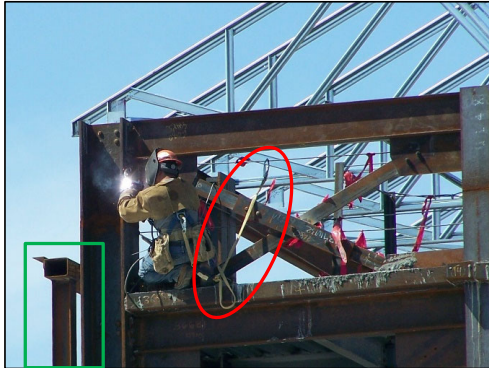
VA: Visual Attention

HI: Hazard Identification

WM: Working Memory



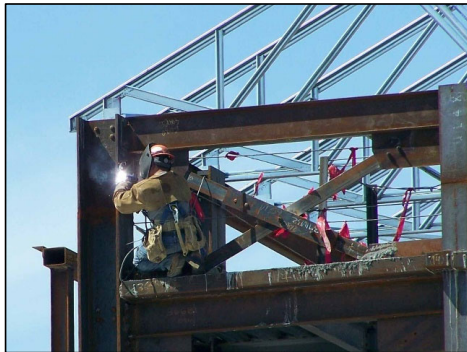
Change Blindness: Examples



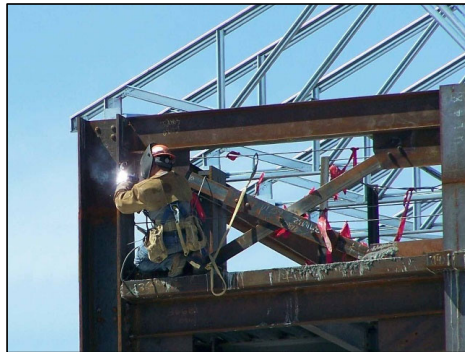
○ Safety relevant
□ Safety irrelevant



○ Safety relevant
□ Safety irrelevant



Safety relevant change



Safety irrelevant change

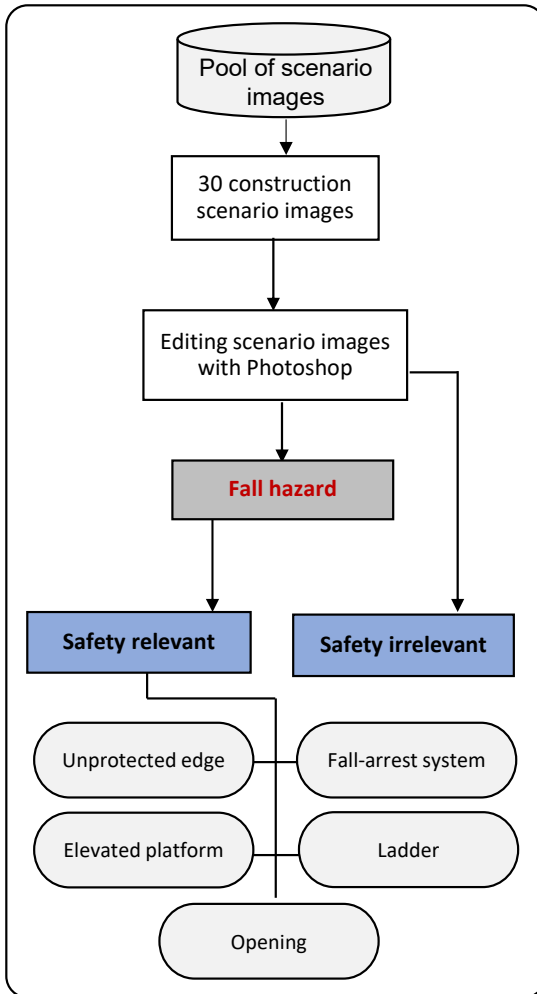


Safety relevant change

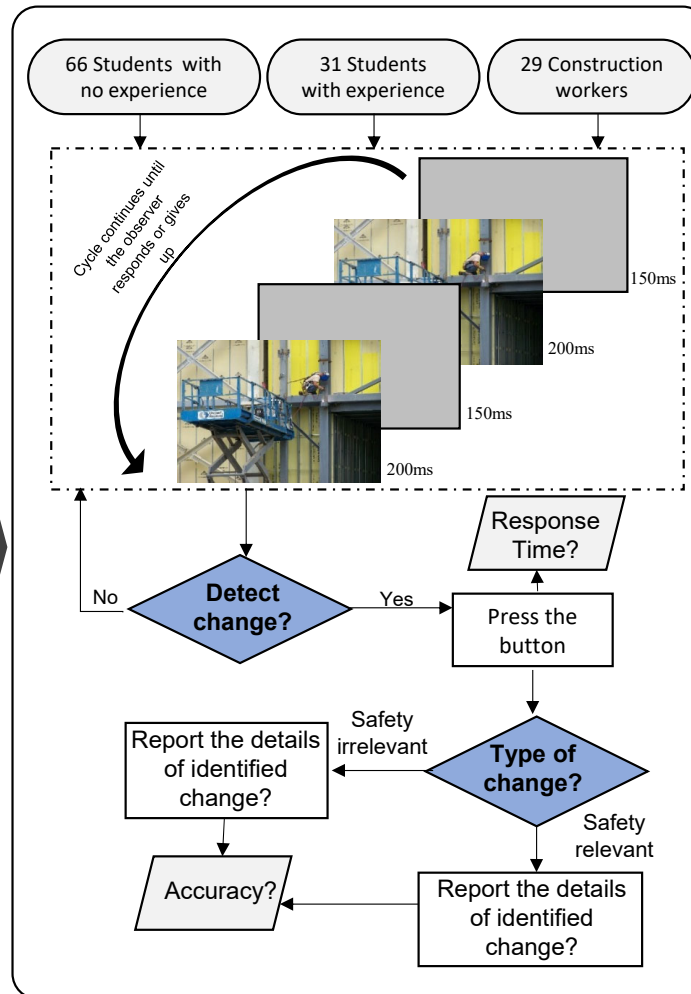


Safety irrelevant change

Planning and Experimental Design

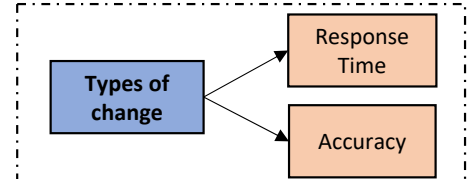


Data Collection

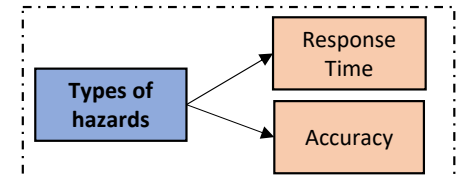


Data Analysis

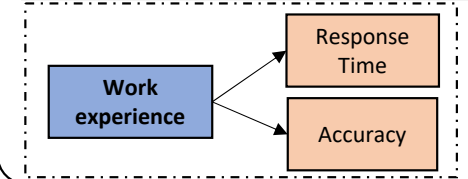
Null Hypothesis I: *Participants' ability to detect various changes at jobsites does not depend on the types of change.*



Null Hypothesis II: *Participants' ability to detect various changes at jobsites does not depend on the fall target types.*



Null Hypothesis III: *Participants' ability to detect various changes at jobsites does not depend on their level of working experience.*



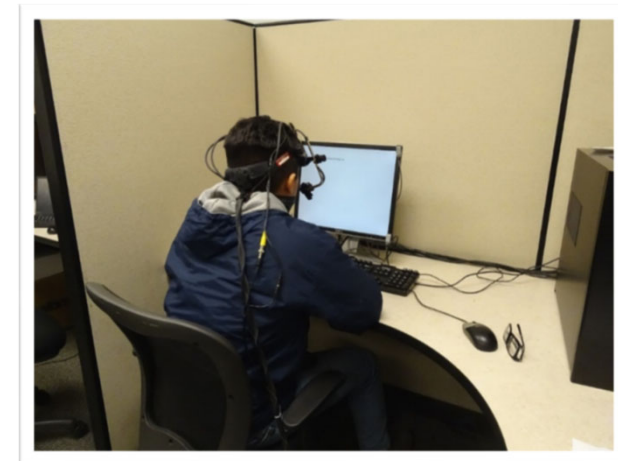
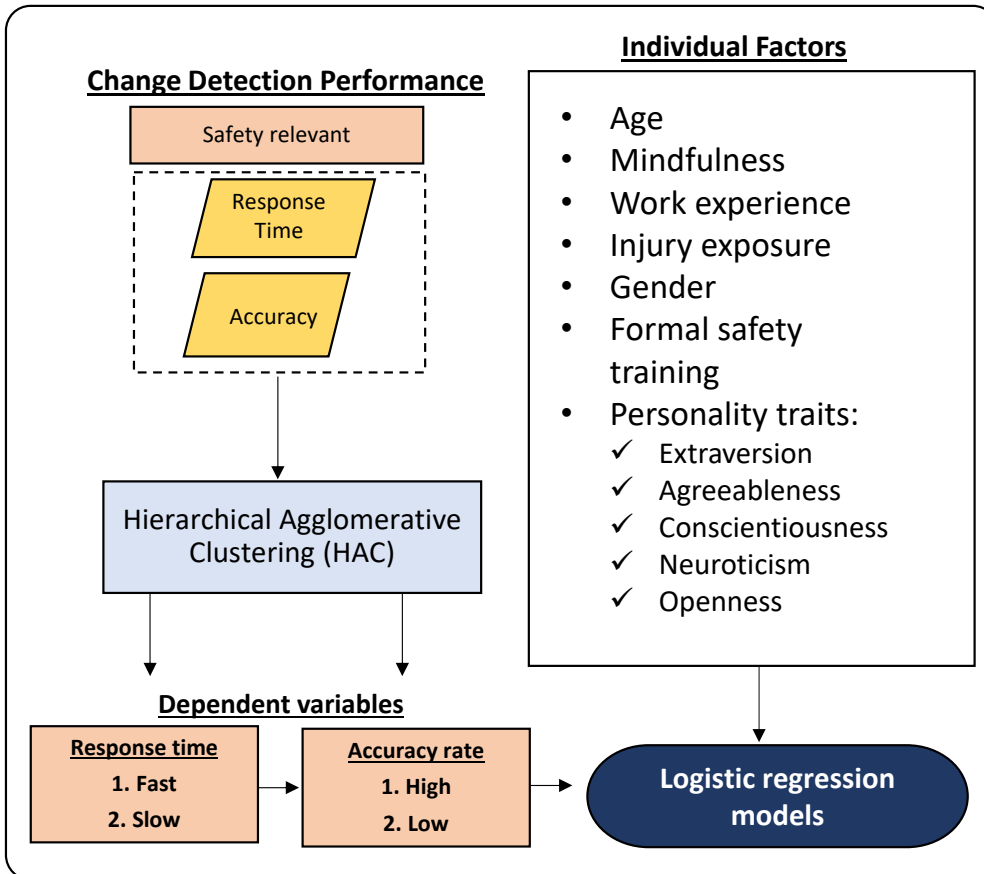
Can you see the change between these two almost identical images?



Null Hypothesis IV: *There is no association between the dependent variables (percent accuracy and mean response time) and the individual factors*

Null Hypothesis V: *There is no difference in attentional allocation (i.e., fixation metrics) of observers towards a type of changes (i.e., safety-relevant versus safety-irrelevant changes)*

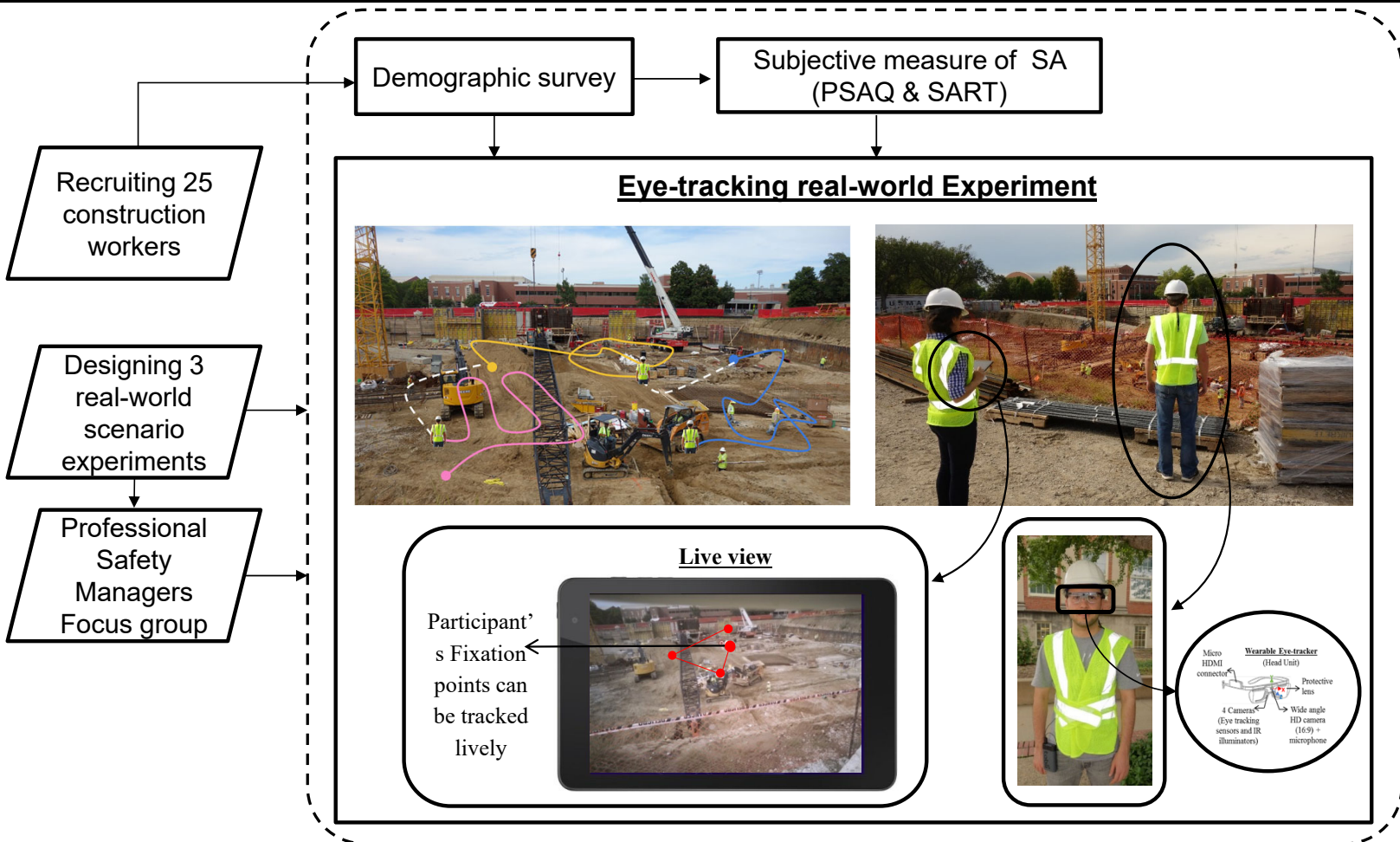
Null Hypothesis VI: *There is no difference in attentional allocation (i.e., fixation metrics) of observers towards the various types of fall hazard changes*



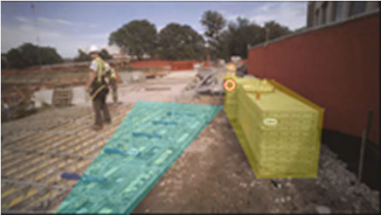
Tobii Pro Glasses 2



Real-time Situation Awareness Using Wearable Technology



Real-time Situation Awareness Using Wearable Technology



Scene 4

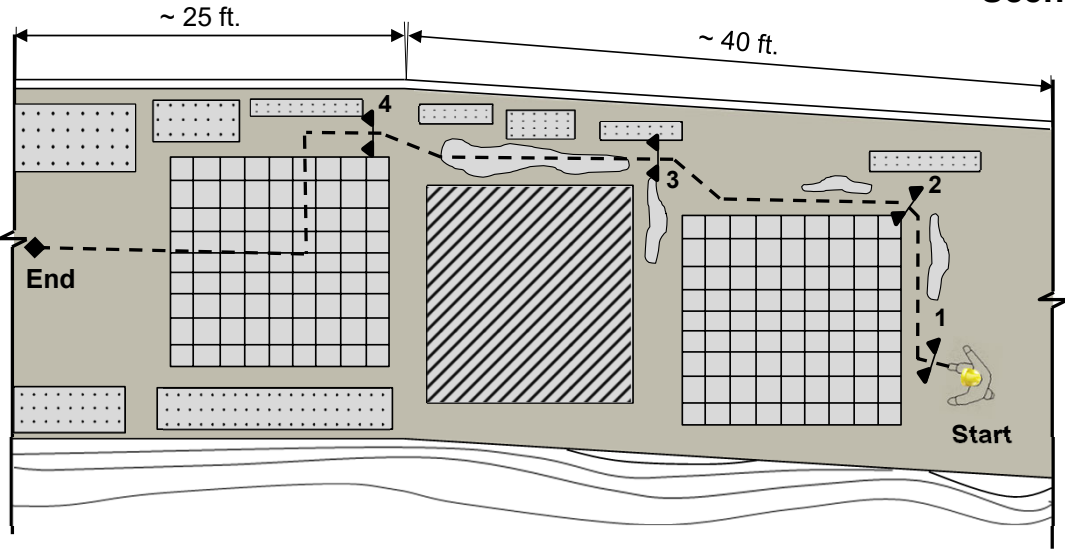


Scene 3



Scene 2

- Guide:**
- Wall formwork
 - Supporting panel
 - Stored materials



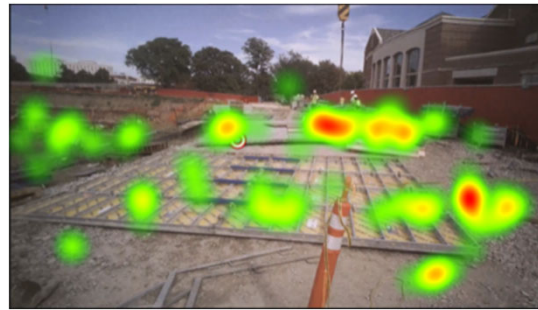
Scene 1

Real-time Situation Awareness Using Wearable Technology

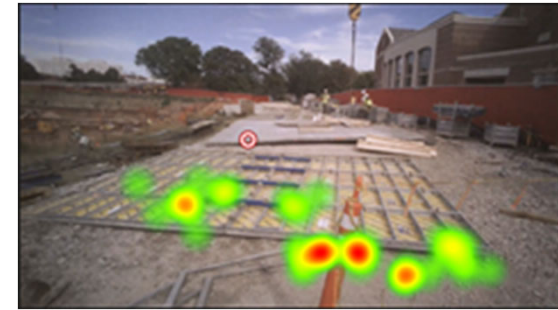
Scene 1



High SA



Low SA



ET metrics	AOI	High SA		Low SA		Permutation results	
		Mean	SD	Mean	SD	Welch's t	p -value
Run count	Scene 1	2.167	1.941	0.000	0.000	2.735	0.10**
	Scene 4	8.000	4.099	4.500	2.887	1.584	0.15
Fixation count	Scene 1	2.000	2.898	0.000	0.000	1.690	0.25
	Scene 4	42.667	29.248	11.000	9.764	2.455	0.05*
Fixation duration	Scene 1	0.144	0.188	0.000	0.000	1.709	0.00*
	Scene 4	0.853	0.585	0.220	0.195	2.457	0.05*
Dwell duration	Scene 1	0.143	0.186	0.000	0.000	1.891	0.10**
	Scene 4	0.857	0.583	0.220	0.195	2.475	0.05*

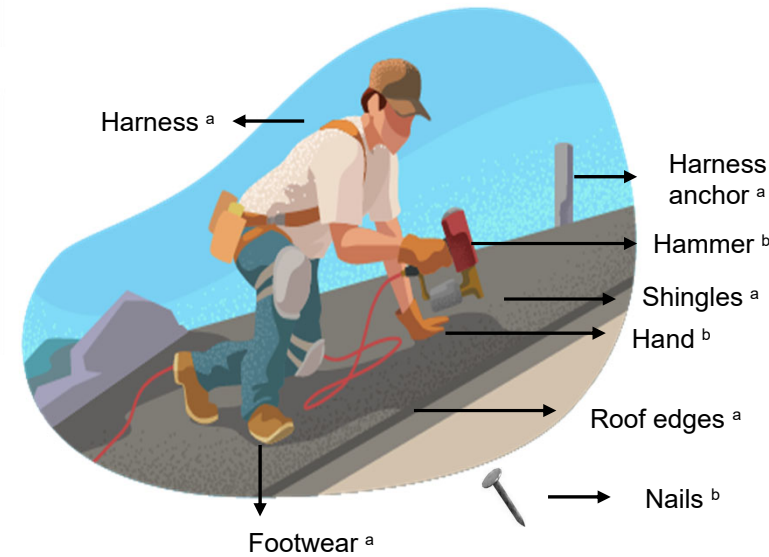
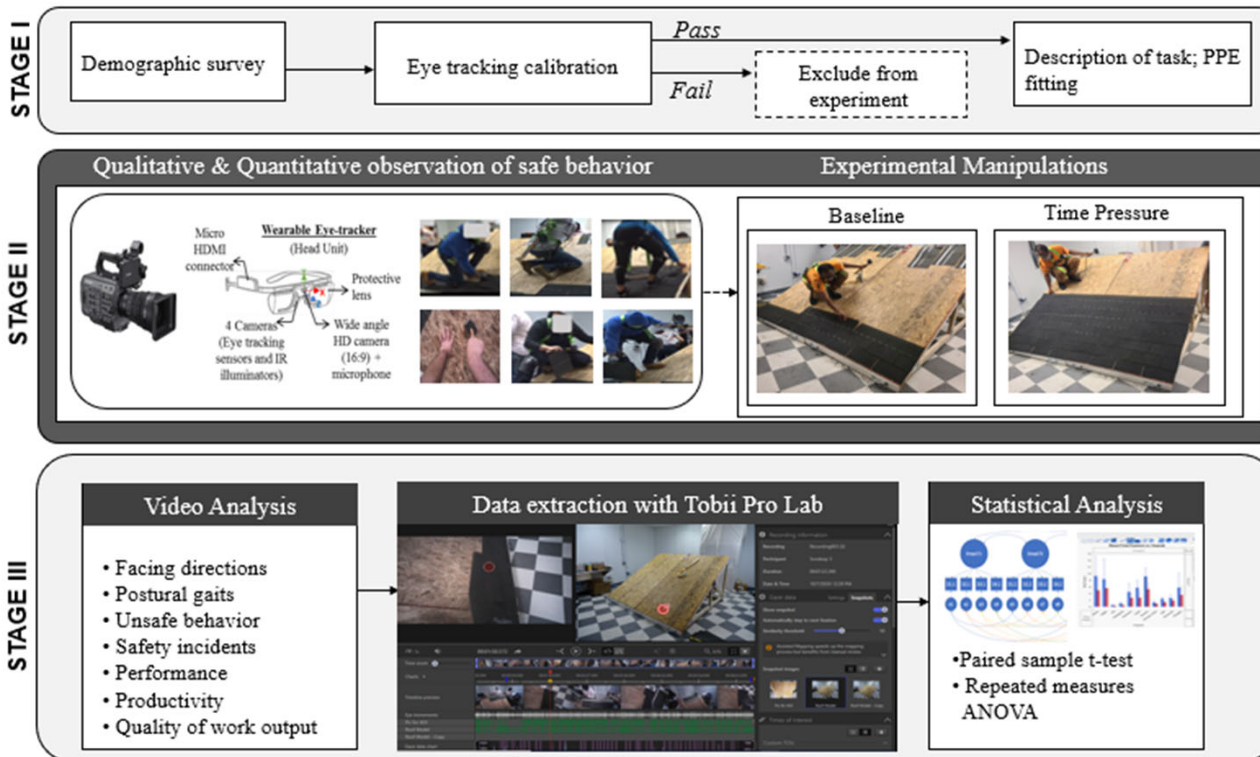
* $p < 0.05$, ** $p < 0.1$

The Influence of Time Pressure on Safety Performance



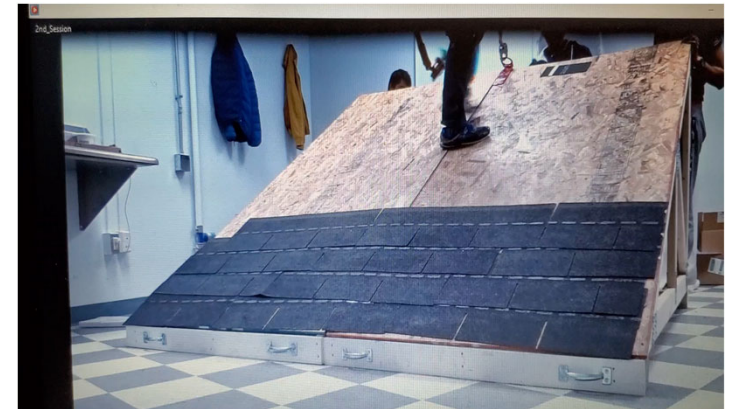
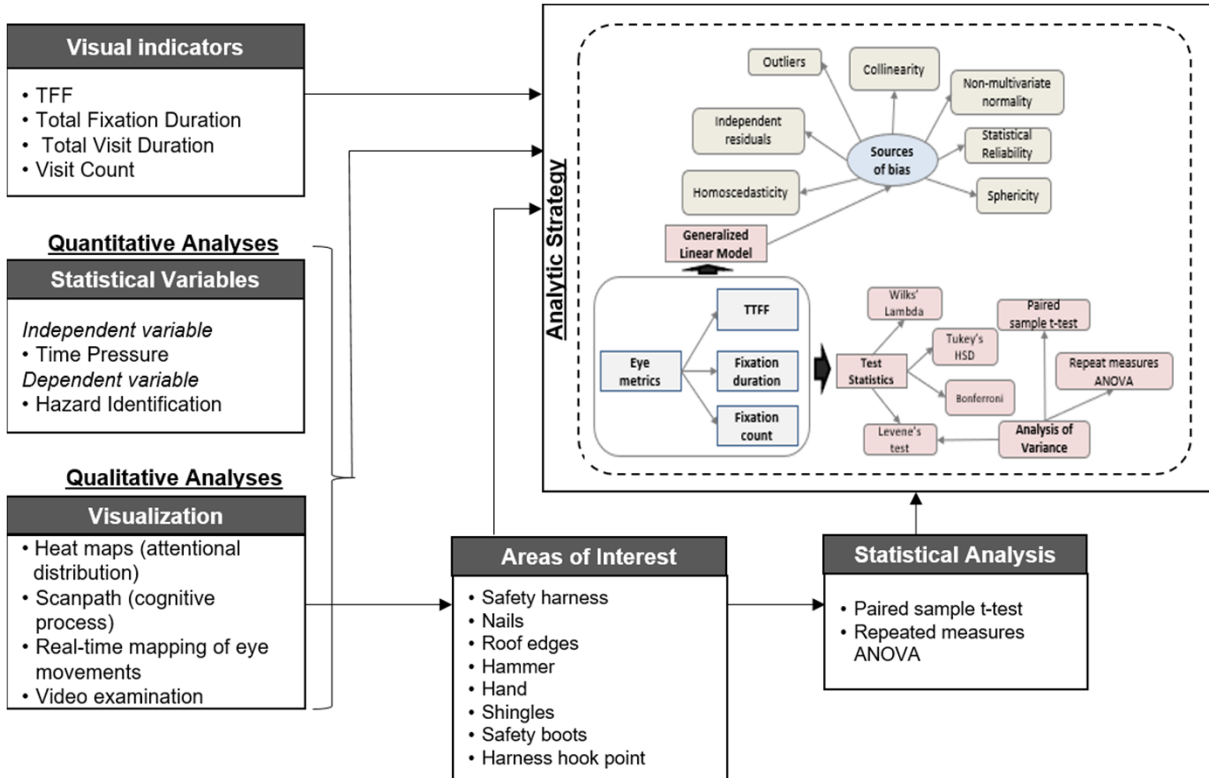
Null Hypothesis I: Time pressure does not impact attention to safety hazards (e.g. slips, trips, falls, hand injuries)

Null Hypothesis II: Time pressure does not impact unsafe behavior of construction workers



i ^a STF AOIs
^b Hand injury AOIs

Statistical Analysis



Attention and Unsafe Behavior

AOIs	TTFF	TFD	TVD	VC
Harness	7.726*	9.604*	16.431*	28.367*
Edges	8.408*	9.197*	15.478*	26.029*
Footwear	13.968*	8.920*	15.043*	27.695*
Shingles	13.853*	6.081*	8.826*	19.661*
Hand	13.884*	5.471*	8.062*	17.038*
Hammer	19.795*	4.418*	6.137*	15.290*
Nails	42.917*	2.183*	3.009*	7.077*
Harness anchor	68.436*	1.128*	1.404*	4.172*

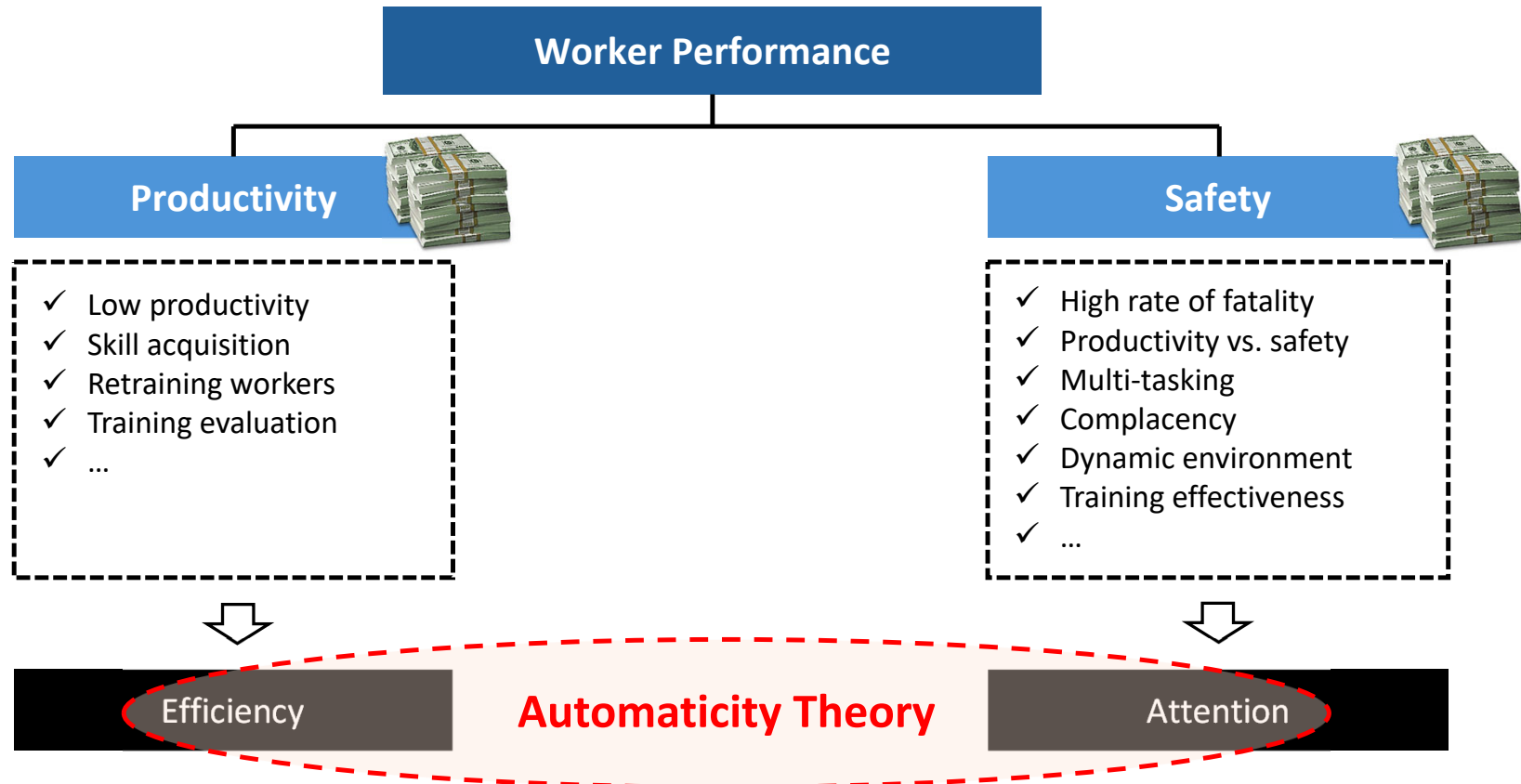


Unsafe behavior: (a) Sitting close to roof edge; (b) Hammer too close to finger; (c) Poor hand coordination; d) Hammer for support at roof tip

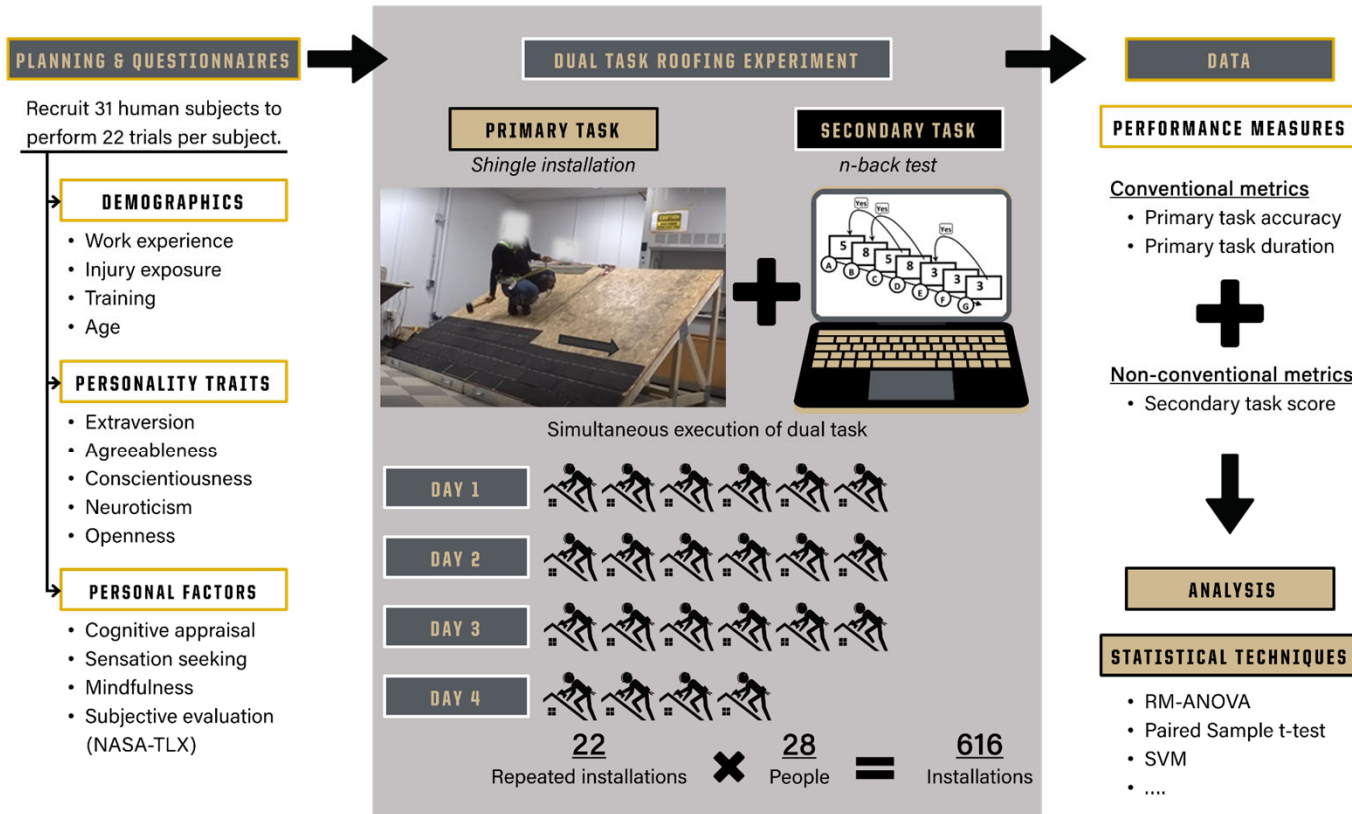


Safety incidents: (a) Near-skid; (b) Slip hazard; (c) Nail puncture hazard; d) Fall off roof

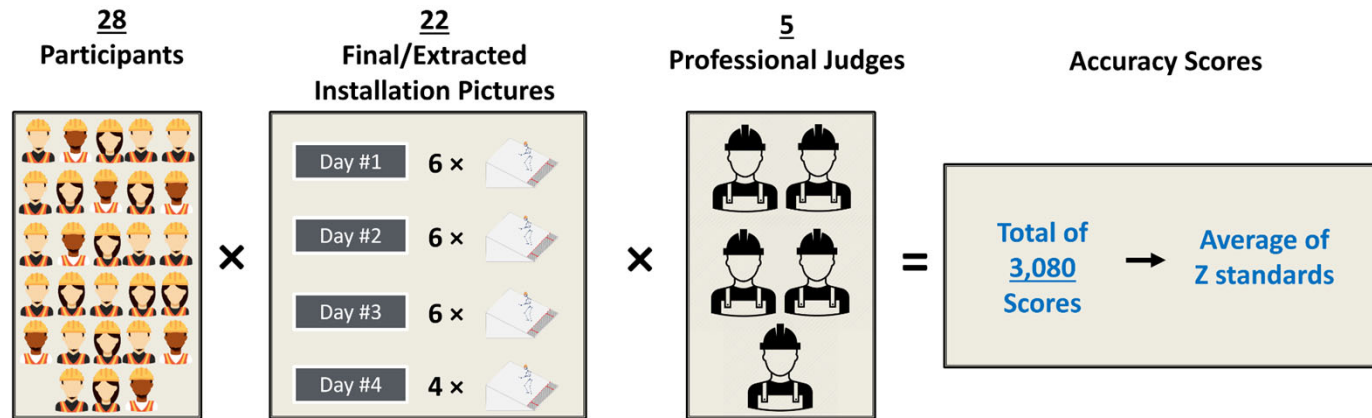
Automaticity



Automaticity



Automaticity



Examples

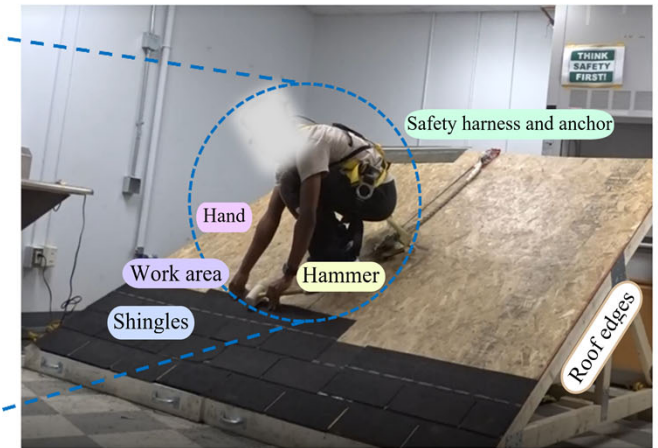
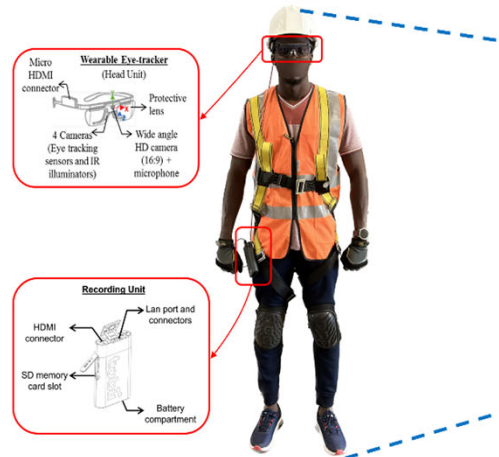
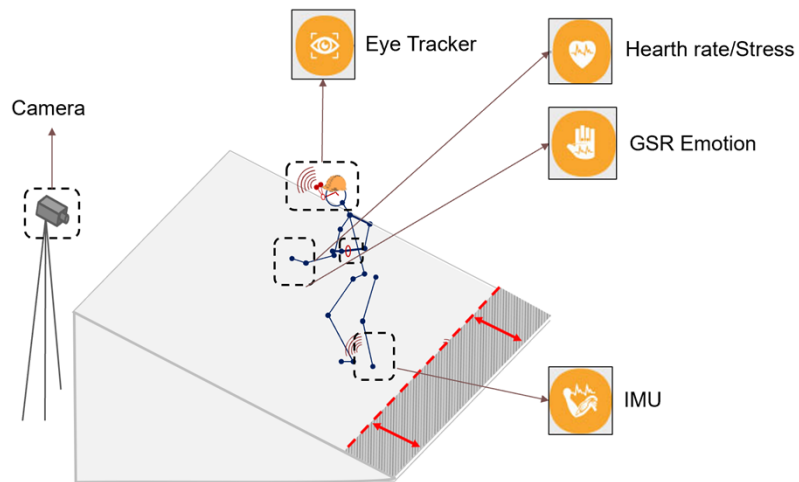
Poor Performance



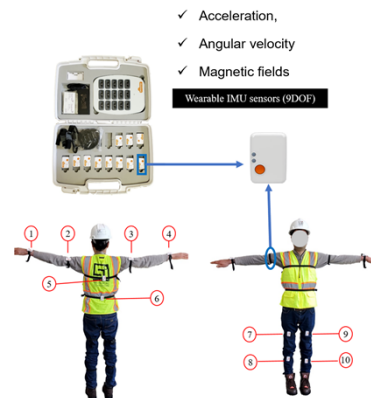
Improved Performance



Automaticity



28 participants x 22 installations = **616 installations**



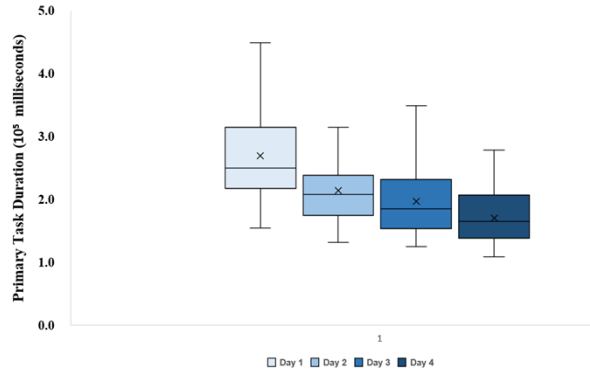
Smart watch

- ✓ Heart rate (HR)
- ✓ Heart rate variability (HRV)
- ✓ Pulse Ox
- ✓ Respiration rate
- ✓ Stress level

Automaticity

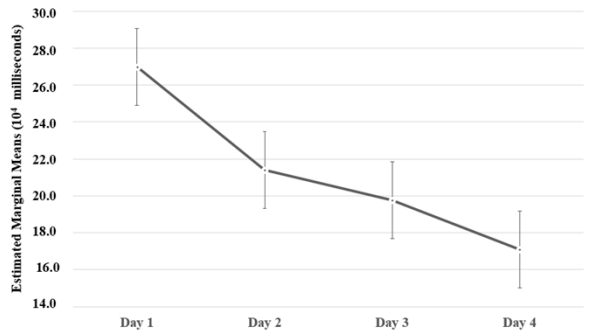
Primary Task

Box Plot of Primary Task Duration



Day 1 Day 2 Day 3 Day 4

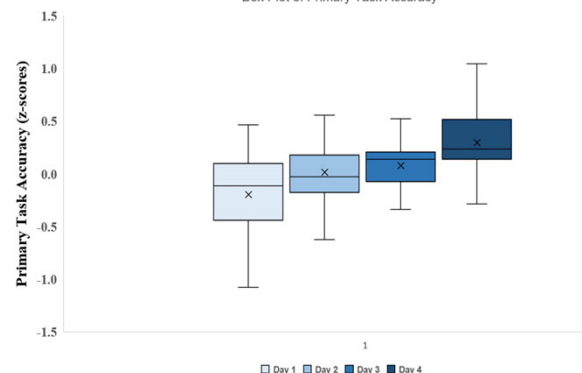
Duration



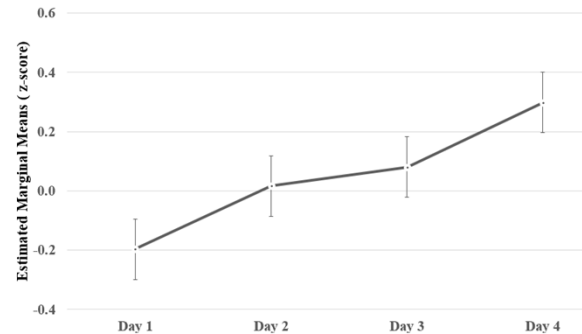
Day 1 Day 2 Day 3 Day 4

Accuracy

Box Plot of Primary Task Accuracy



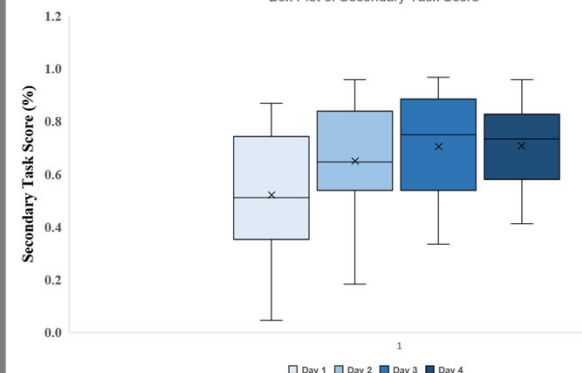
Day 1 Day 2 Day 3 Day 4



Day 1 Day 2 Day 3 Day 4

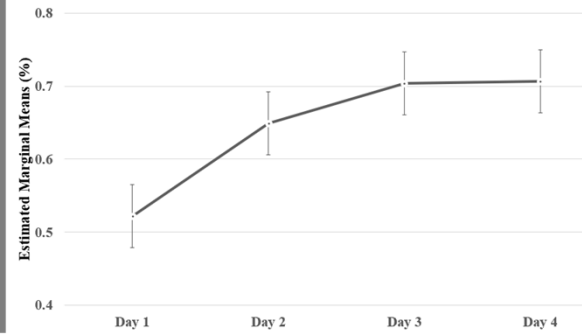
Secondary Task

Box Plot of Secondary Task Score



Day 1 Day 2 Day 3 Day 4

Score



Day 1 Day 2 Day 3 Day 4

Automaticity

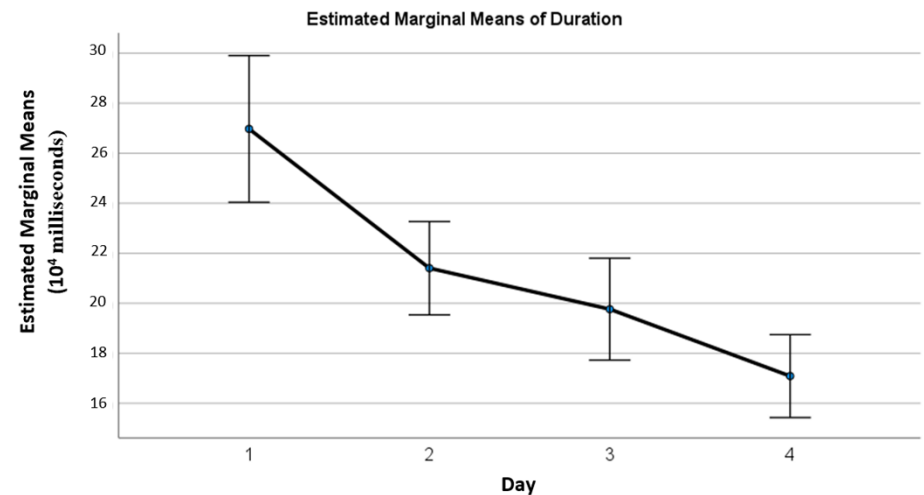
Primary Task Duration

Primary task durations (milliseconds)			
Day	N	Mean (10 ⁴)	Std. Deviation (10 ⁴)
Day 1	28	26.976	7.563
Day 2	28	21.406	4.815
Day 3	28	19.767	5.254
Day 4	28	17.091	4.279

Repeated Measures ANOVA of Primary Task Duration: Multivariate tests

Effect		Value	F	Hypothesis df	Error df	P-value
Day	Pillai's Trace	0.768	27.529 ^b	3.000	25.000	<0.001
	Wilks' Lambda	0.232	27.529 ^b	3.000	25.000	<0.001
	Hotelling's Trace	3.303	27.529 ^b	3.000	25.000	<0.001
	Roy's Largest Root	3.303	27.529 ^b	3.000	25.000	<0.001

* Design: Intercept; Within Subjects Design: Day; b. Exact statistic



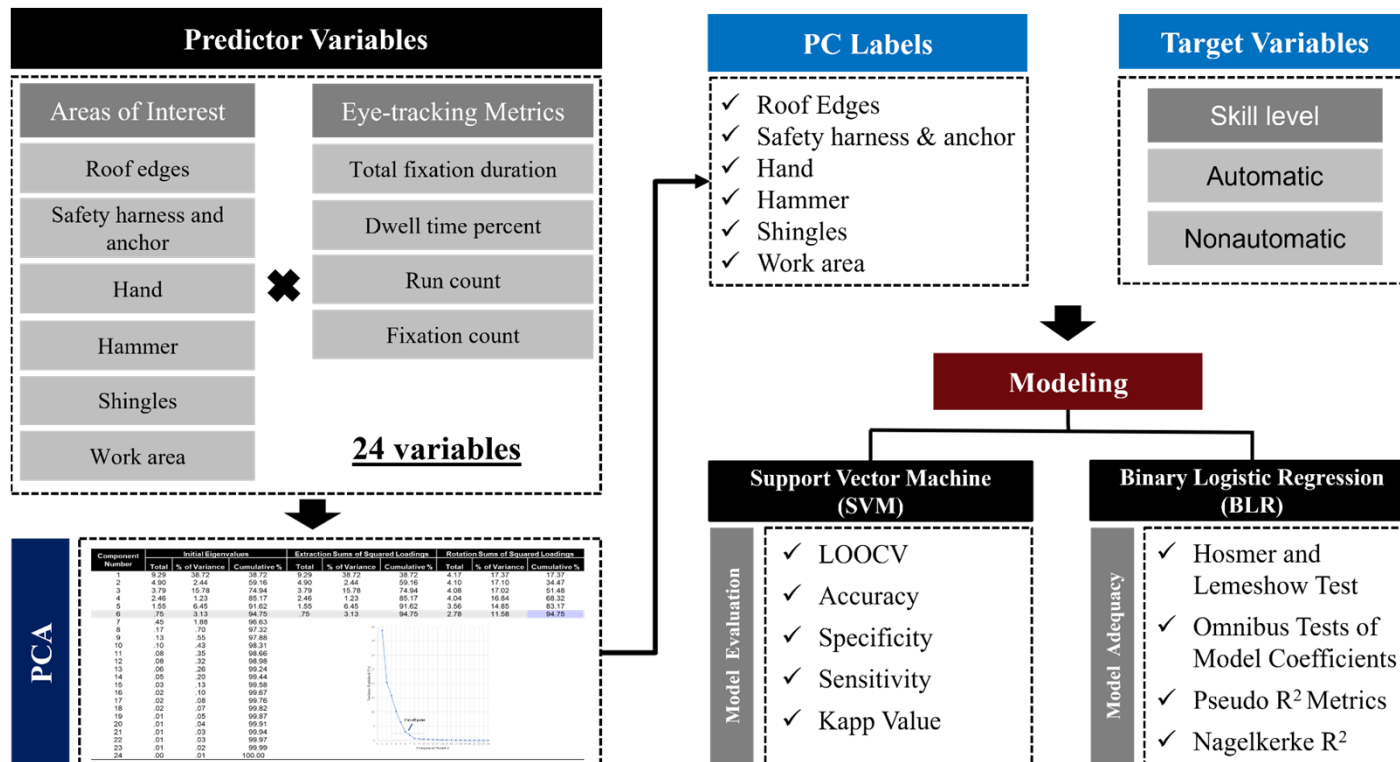
Automaticity

Descriptive Statistics of the Eye-tracking Metrics

AOIs	Statics	Total Fixation Duration		Fixation Count		Dwell Time Percent		Run Count	
		Novice	Expert	Novice	Expert	Novice	Expert	Novice	Expert
Roof Edges	Mean	0.006	0.004	0.037	0.004	0.929	0.528	0.019	0.014
	Std Dev	0.007	0.004	0.040	0.004	1.059	0.623	0.017	0.012
Safety Harness and Anchor	Mean	0.006	0.004	0.040	0.019	0.836	0.457	0.026	0.011
	Std Dev	0.006	0.005	0.036	0.023	0.864	0.583	0.020	0.013
Hand	Mean	0.015	0.014	0.085	0.074	2.117	1.996	0.044	0.043
	Std Dev	0.010	0.015	0.055	0.066	1.545	2.196	0.025	0.033
Hammer	Mean	0.013	0.005	0.075	0.035	1.758	0.677	0.040	0.023
	Std Dev	0.010	0.005	0.055	0.033	1.350	0.676	0.027	0.022
Shingles	Mean	0.067	0.048	0.385	0.314	12.838	9.248	0.117	0.122
	Std Dev	0.036	0.028	0.158	0.161	4.963	4.949	0.036	0.046
Work Area	Mean	0.185	0.125	0.972	0.708	38.704	31.774	0.155	0.134
	Std Dev	0.081	0.066	0.334	0.315	8.247	10.091	0.048	0.049

Automaticity

Data Analysis



Automaticity

Variables in the Proposed BLR Model Equation

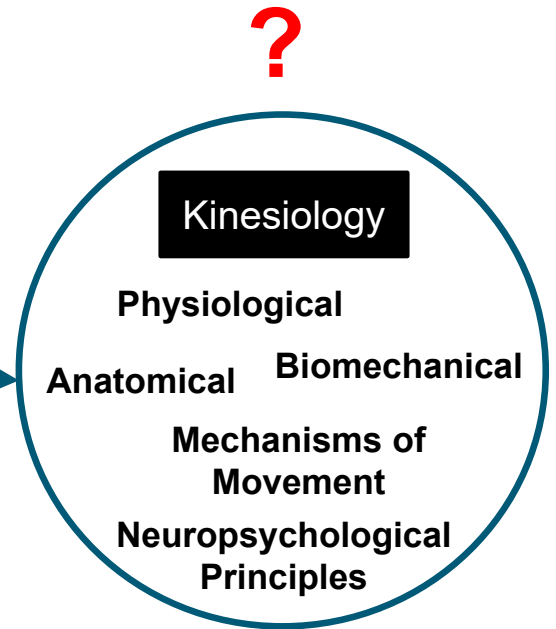
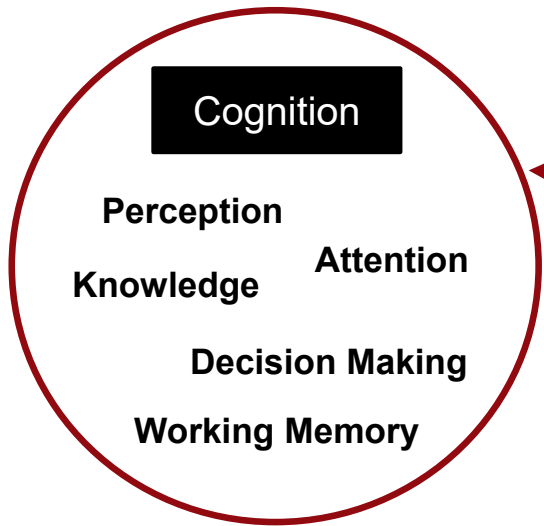
AOIs/PCs	β	S.E.	Wald	df	P-value	Exp(β)	95% C.I. for Exp(β)	
							Lower	Upper
Roof Edges	-.603	.446	1.832	1	.176	.547	.228	1.310
Safety Harness & Anchor	-1.489*	.484	9.472	1	.002	.226*	.087	.582
Hand	.445	.396	1.265	1	.261	1.560	.719	3.389
Hammer	-1.521*	.511	8.847	1	.003	.219*	.080	.595
Shingles	.248	.379	.429	1	.512	1.282	.610	2.695
Work Area	-.864*	.428	4.080	1	.043	.421*	.182	.975
Constant	-.140	.384	.132	1	.716	.870		

* The regression coefficient is significant at the .05 level.

$$\begin{aligned}
 \ln(odds) = & - .140 - .603 \times AOI_{\text{Roof Edges}} - 1.489 \times AOI_{\text{Safety Harness \& Anchor}} \\
 & + .445 \times AOI_{\text{Hand}} - 1.521 \times AOI_{\text{Hammer}} \\
 & + .248 \times AOI_{\text{Shingles}} - .864 \times AOI_{\text{Work Area}}
 \end{aligned}$$

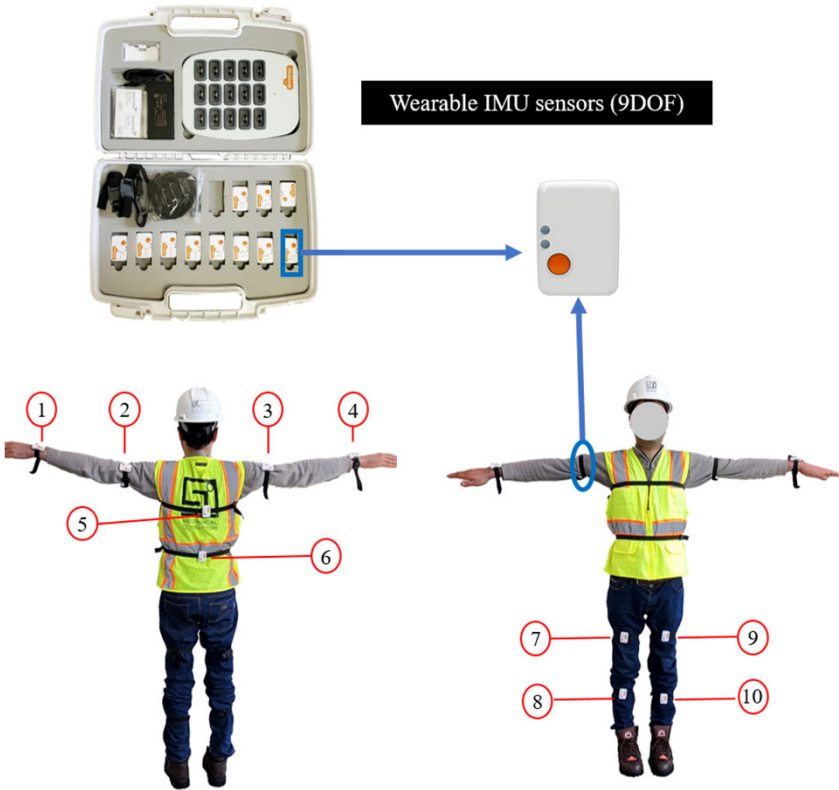
Automaticity

Evaluation Statistics	SVM (LOOCV)	BLR (LOOCV)
Accuracy	.768	.768
95% Confidence Interval	(0.636, 0.870)	(0.636, 0.870)
No Information Rate	0.5	0.5
P-Value [Acc > NIR]	3.667e-05	3.667e-05
Kappa	0.536	0.536
Precision	0.727	0.742
Sensitivity/Recall	0.857	0.821
Specificity	0.679	0.714



3. Ergonomics

Assessing Lower Extremity Kinematics of Roofing Tasks



Level-Surface (0°)

Walking



Squatting



Sloped Surface (30°)

30° Wooden Roof Platform



The Effect of Fatigue on Workers' Safety Performance

Research Questions

- ❖ Is the fatigue pattern different between/within the lumbar flexor and lumbar extensor muscles for cyclical lifting tasks?
- ❖ When fatigue starts? How much recovery they need?

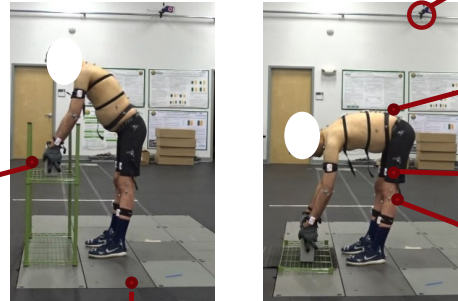


Fatigue protocol



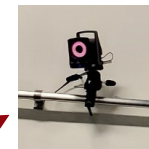
Oldcastle 4-in x 4-in x 16-in
Concrete A-Shaped Fence Block

Lifting Activity



Force plate-Bertec:

- ✓ Ground reaction forces (GRF)
- ✓ Ground reaction moments (GRM)



Motion capture systems



IMU shimmer



EMG Delsys



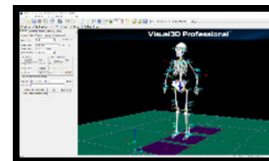
Passive reflective
markers



Camera



Exoskeleton



Kinematic and kinetic outputs:

- Median frequency of lumbar muscles
- Peak compressive force and Shear force at joints.
- Joint angles

Experiment Conditions:

No.	Factors	Level	
1	Lifting height	Below knee	Waist
2	Weight of Loads	Light	Heavy
3	Working pace	Slow	Fast
4	Fatigue	Yes/No	Yes/No

4. Training and Visualization

Visualization for Safety



Capture	Visualization	Augmentation

Select to learn more →
 Select to learn more →
 Timer → **00:41**
 training session 1/10

- Hazard: Fall - Unprotected
- Hazard: Fall - Ladders
- Hazard: Struck-By - Excavator Swing
- Hazard: Struck-By - Excavator Swing

Silica Hazard in Construction (SH-05053-SH8)

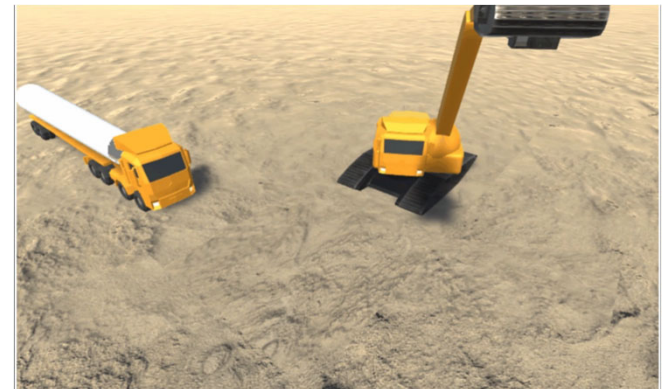
MISSION STATEMENT

THIS IS A 3D SIMULATION TRAINING PROGRAM ON SILICA PROTECTION IN THE CONSTRUCTION INDUSTRY AND AN EDUCATIONAL TOOL TO TEACH USERS ABOUT SAFETY ISSUES ON SILICA PROTECTION.

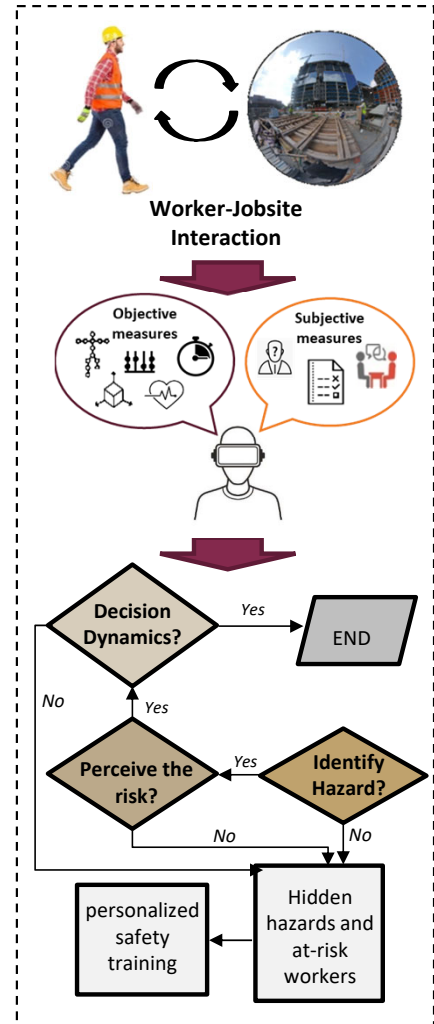
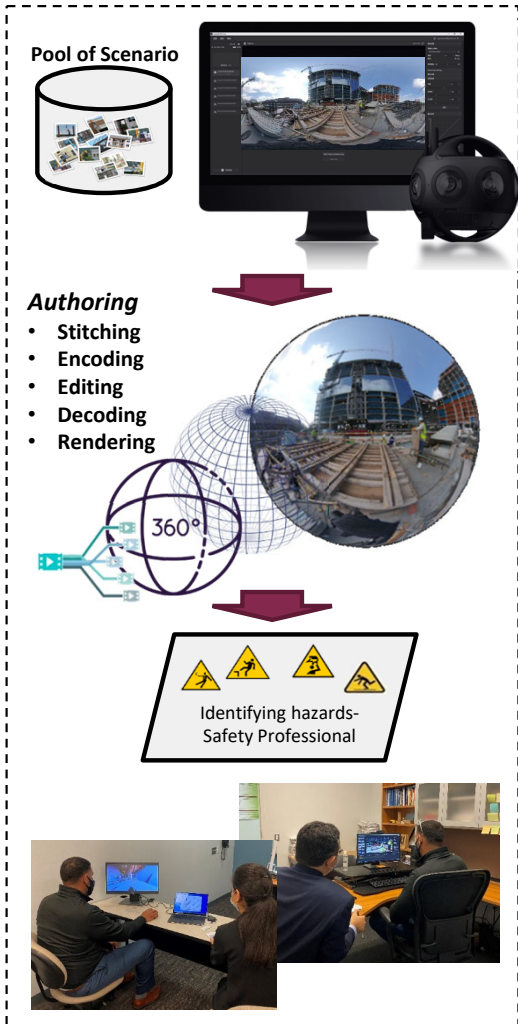
SAFETY SCENARIOS HAVE BEEN CREATED TO BE EXPLORED. A COMPREHENSIVE TEST IS IN EACH SCENARIO TO MEASURE THE OBTAINED KNOWLEDGE ON THAT SPECIFIC SCENARIO.

BEGIN

Wind Tower Construction Safety (SHTG-FY-20-02)

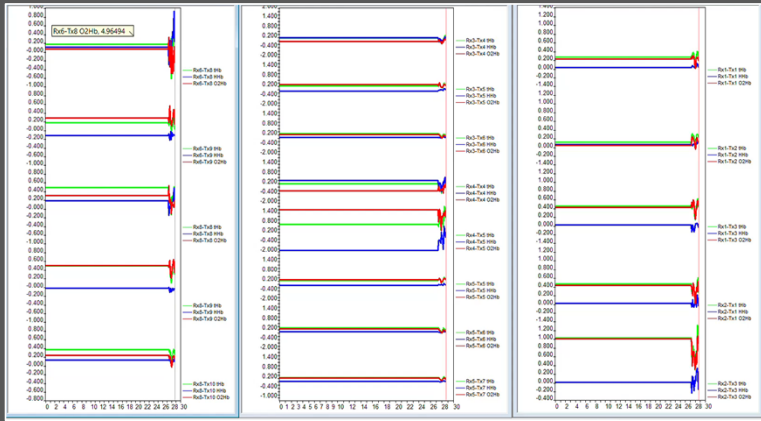


360-degree Dynamic Panorama Augmentation for a Real-Time Safety Warning



360-degree Dynamic Panorama Augmentation for a Real-Time Safety Warning

OxySoft Brite-fNIRS



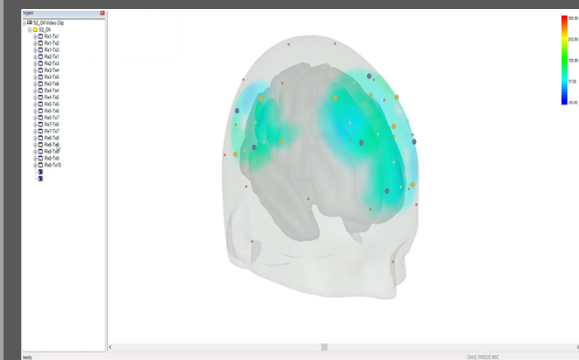
At Jobsite



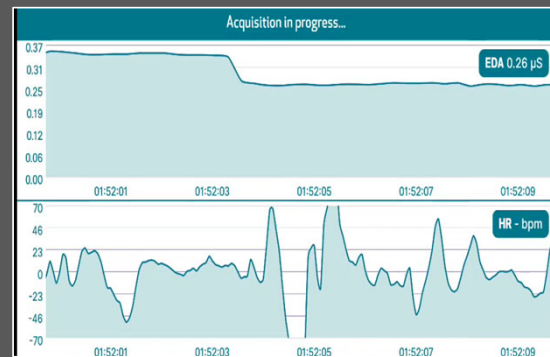
360 Panoramic Augmentation



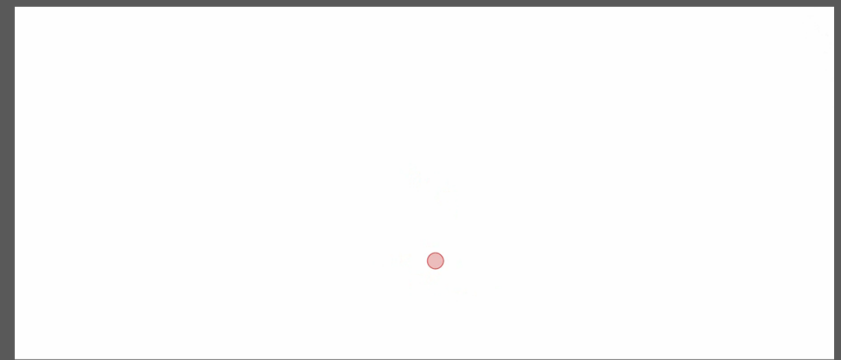
Brain Activation



E4 Empatica (GSR and HR)



Real-Time Attentional Distribution



360-degree Dynamic Panorama Augmentation for a Real-Time Safety Warning

Internal factors (e.g. different abilities of situational awareness)

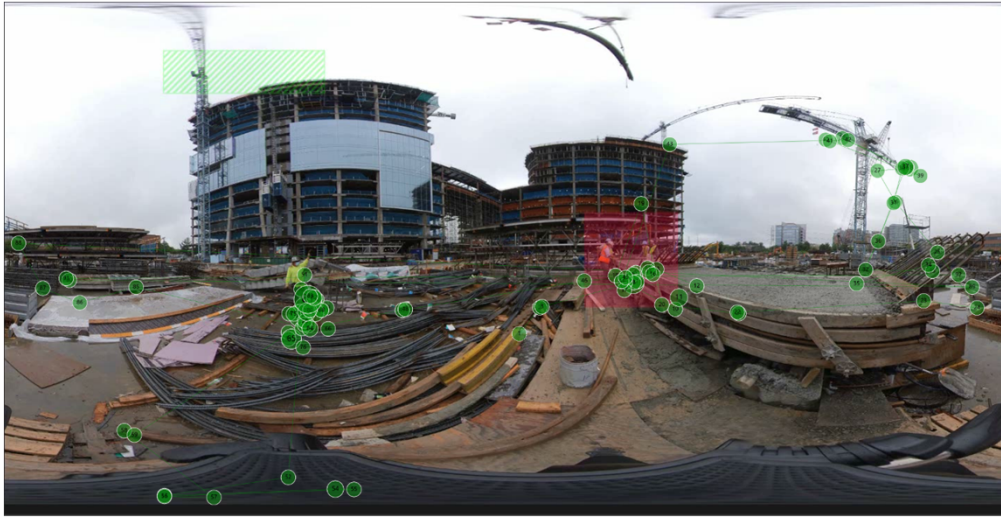


- Lower situational awareness



- Higher situational awareness

360-degree Dynamic Panorama Augmentation for a Real-Time Safety Warning



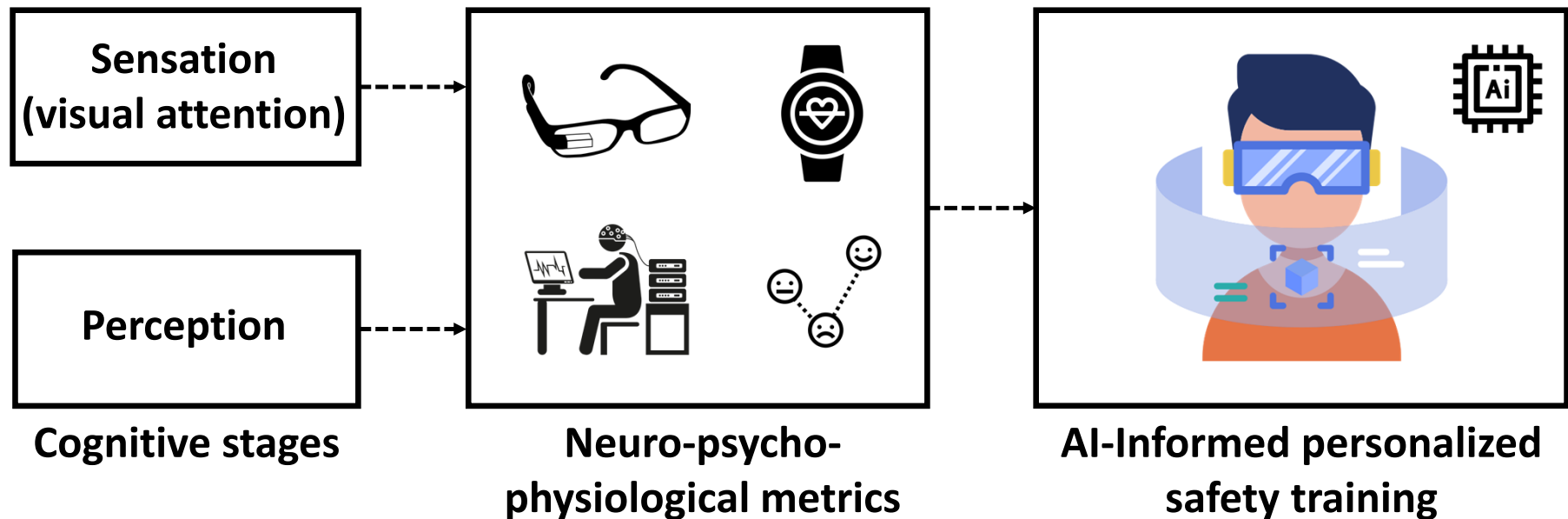
- **Lower situational awareness**



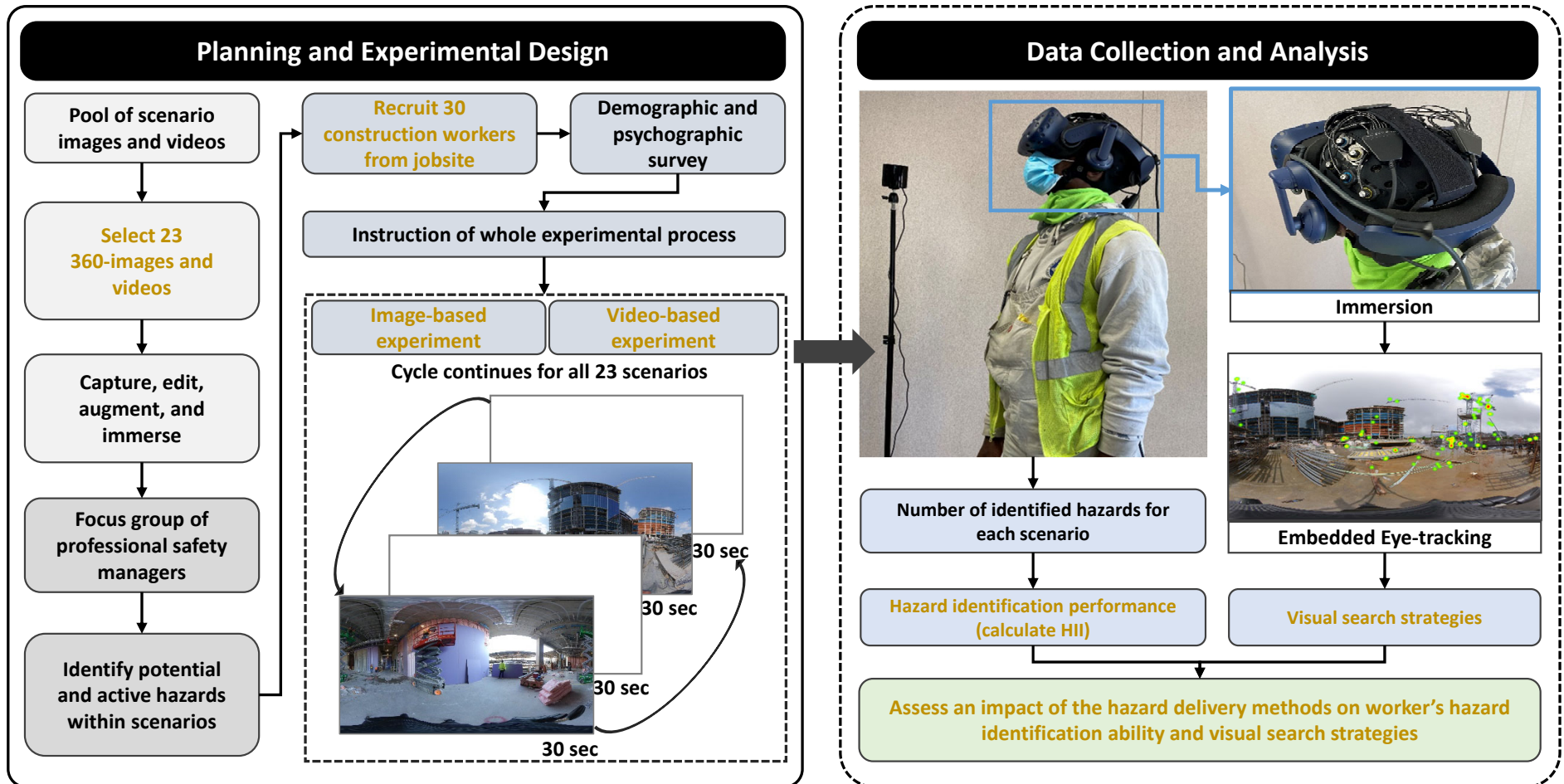
- **Higher situational awareness**

Overall Goal

The overall goal of this research is to translate **non-invasive neuro-psycho-physiological metrics** (e.g., brain signals, emotional responses, eye movements) into **information personalized AI-based training systems**



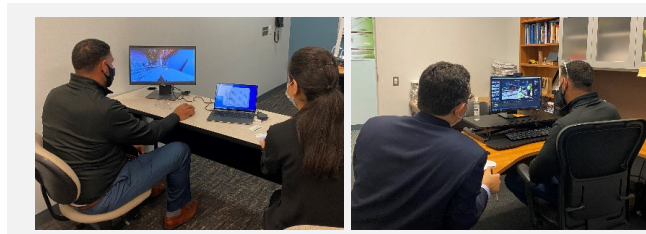
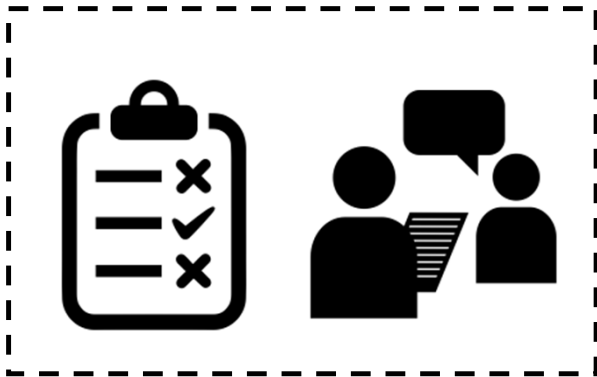
Training Development



Training Development

- Hazard identification index (HII)

- Subjective hazard identification and demographic data



- Identifying hazards by safety professional

- Predefined hazards



- Worker's subjective hazard identification by using surveys and interviews

- Individual hazard identification index (HII) score

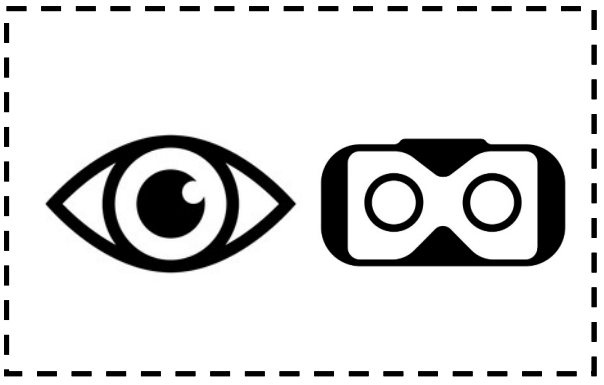


$$\rightarrow 0.5 = \frac{2 = (\text{number of identified hazards})}{4 = (\text{total number of hazards})}$$

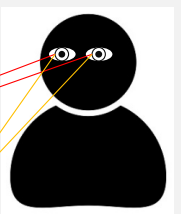
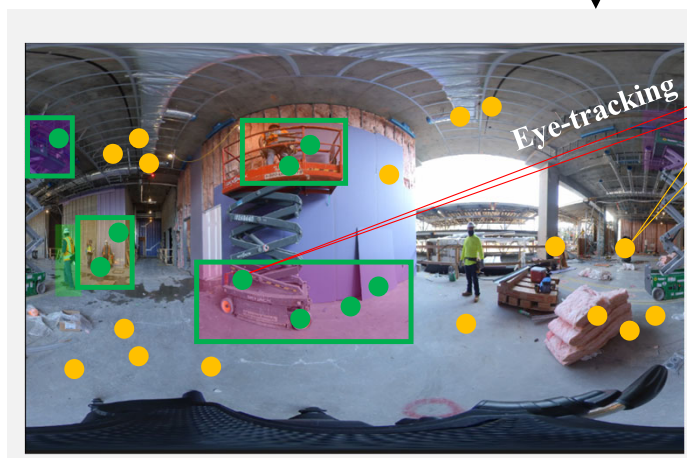
Training Development

Eye-tracking data

- Various eye-tracking metrics (e.g., fixation duration and count)



- Area of Interest (AOI) was marked in each scenario based on predefined hazards

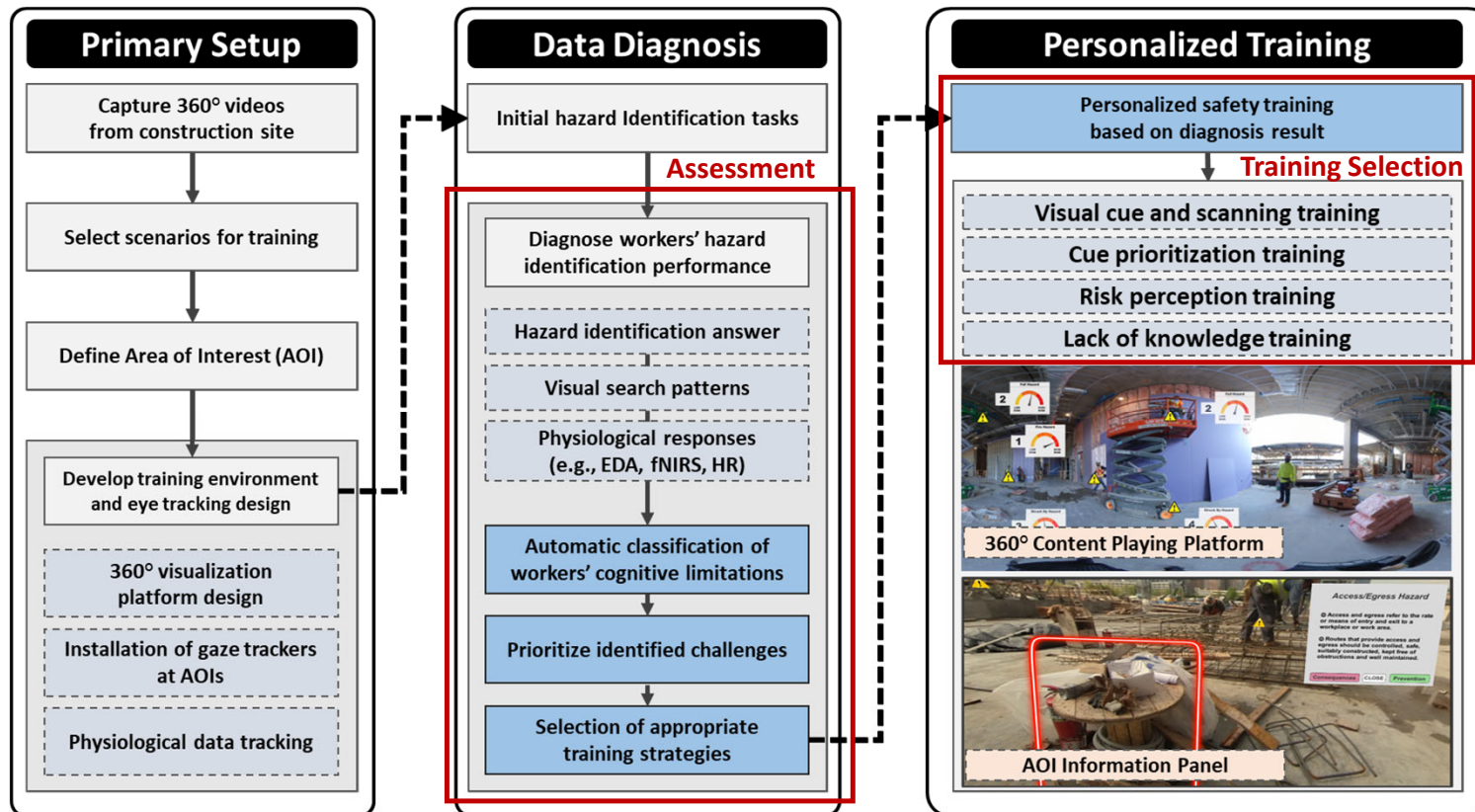


Embedded eye-tracker

- AOIs
- Fixations within AOIs
- Fixations outside of AOIs

Training Development

Framework of the proposed safety training

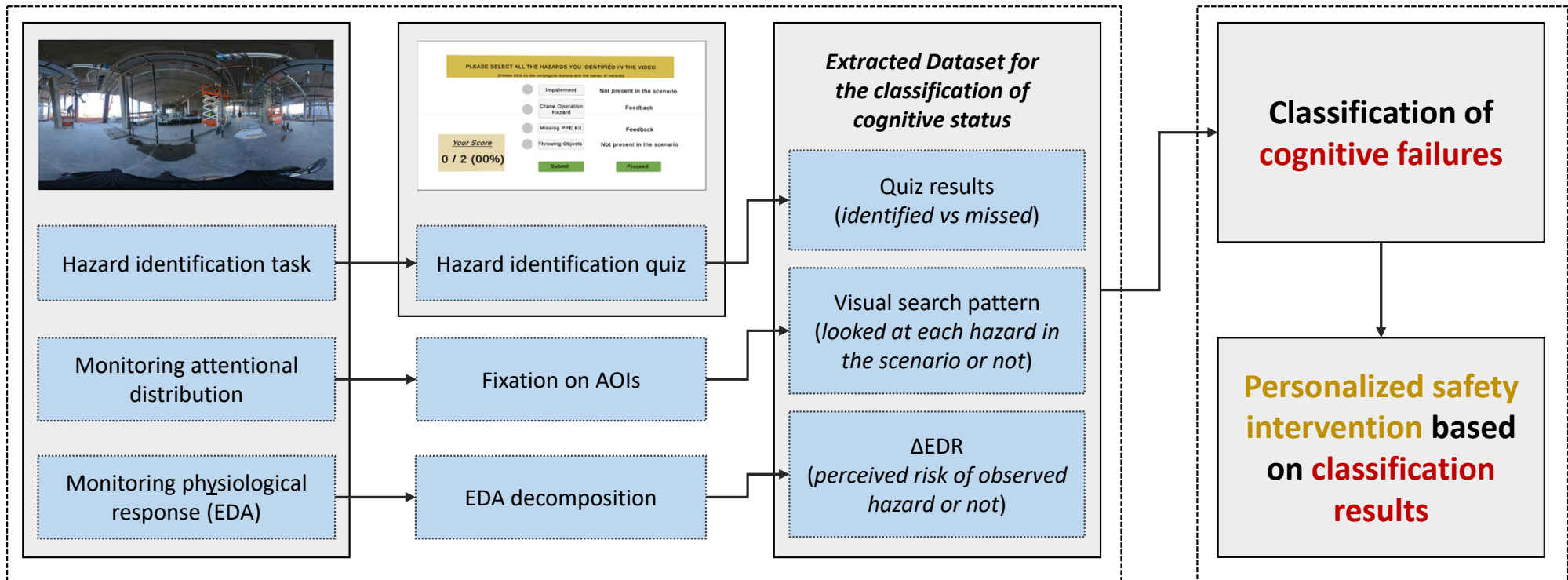


A Windows application incorporating with HTC Vive headset with embedded eye-tracking capability and an Empatica E4 medical wristband

Training Development

- AI system embedded in the personalized training

Multi-modal data collection and preprocessing



Training Development

▪ Cognitive failures and classification mechanisms

Visual attention	Physiological reaction	Subjective report	Cognitive Status
No	No	Miss	Attentional failure <i>(i.e., when hazard is not viewed and not identified)</i>
Yes	No	Miss	Inattentive blindness <i>(i.e., when hazard is viewed and not identified + the risk of hazard is not perceived)</i>
Yes	Yes	Miss	Low-risk perception/High-risk tolerance <i>(i.e., when a hazard is viewed and not identified + the perceived risk associated with the hazard is below the individual's risk tolerance)</i>
Yes	Yes	Identify	Correct hazard identification <i>(i.e., when hazard is viewed and identified + the risk of hazard is perceived)</i>
No	No	Identify	You got lucky <i>(i.e., when hazard is not viewed and but reported as identified)</i>
Yes	No	Identify	Inappropriate risk perception <i>(i.e., when a hazard is viewed, but the risk of hazard is not properly perceived)</i>

Classification of key cognitive failures by combining multi-modal datasets

Training Development

▪ Cognitive failures and recommended interventions

VISUAL CUE SCANNING AND SEARCH STRATEGY TRAINING

Visual Cue Scanning and Search Strategy Training for Missing PPE Kit Hazard:

Our data showed that you missed Missing PPE Kit Hazard, due to ineffective cue scanning and search strategy. In this training, first we will show you how a safety professional will scan the same scenario looking for cues. Then, next training will help you to learn how to prioritize different risks existing in the scenario.

Now, press the PLAY button to start the training.

Visual search strategy training



Aiming to **enhance workers' fundamental visual scanning abilities** by showing an expert's visual search patterns


Training Development

■ Cognitive failures and recommended interventions


LACK OF KNOWLEDGE TRAINING

Lack of Knowledge Training for Missing PPE Kit Hazard:

Our data showed that you may need to learn more about Missing PPE Kit Hazard, its consequences and preventive measures. In the training, please click on the caution buttons to learn more about the hazard. Now press the play button to start the training.



Lack of knowledge training



Aiming to **enhance workers' safety knowledge** by providing relevant information regarding different hazards

Training Development

▪ Cognitive failures and recommended interventions

RISK PERCEPTION TRAINING

Risk Perception Training for Struck By Hazard:

Our data showed that you missed Struck By Hazard, due to low risk perception / high risk tolerance. This training will highlight the importance of identifying this hazard in a timely manner and associated consequences of misperceiving it on an actual construction site.

Now, press the PLAY button to start the training.

Risk perception training

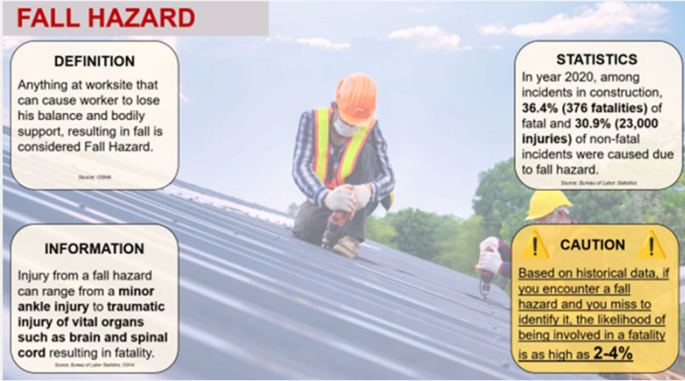
FALL HAZARD

DEFINITION
Anything at worksite that can cause worker to lose his balance and bodily support, resulting in fall is considered Fall Hazard.

STATISTICS
In year 2020, among incidents in construction, **36.4% (376 fatalities)** of fatal and **30.9% (23,000 injuries)** of non-fatal incidents were caused due to fall hazard.

INFORMATION
Injury from a fall hazard can range from a **minor ankle injury to traumatic injury of vital organs such as brain and spinal cord** resulting in fatality.

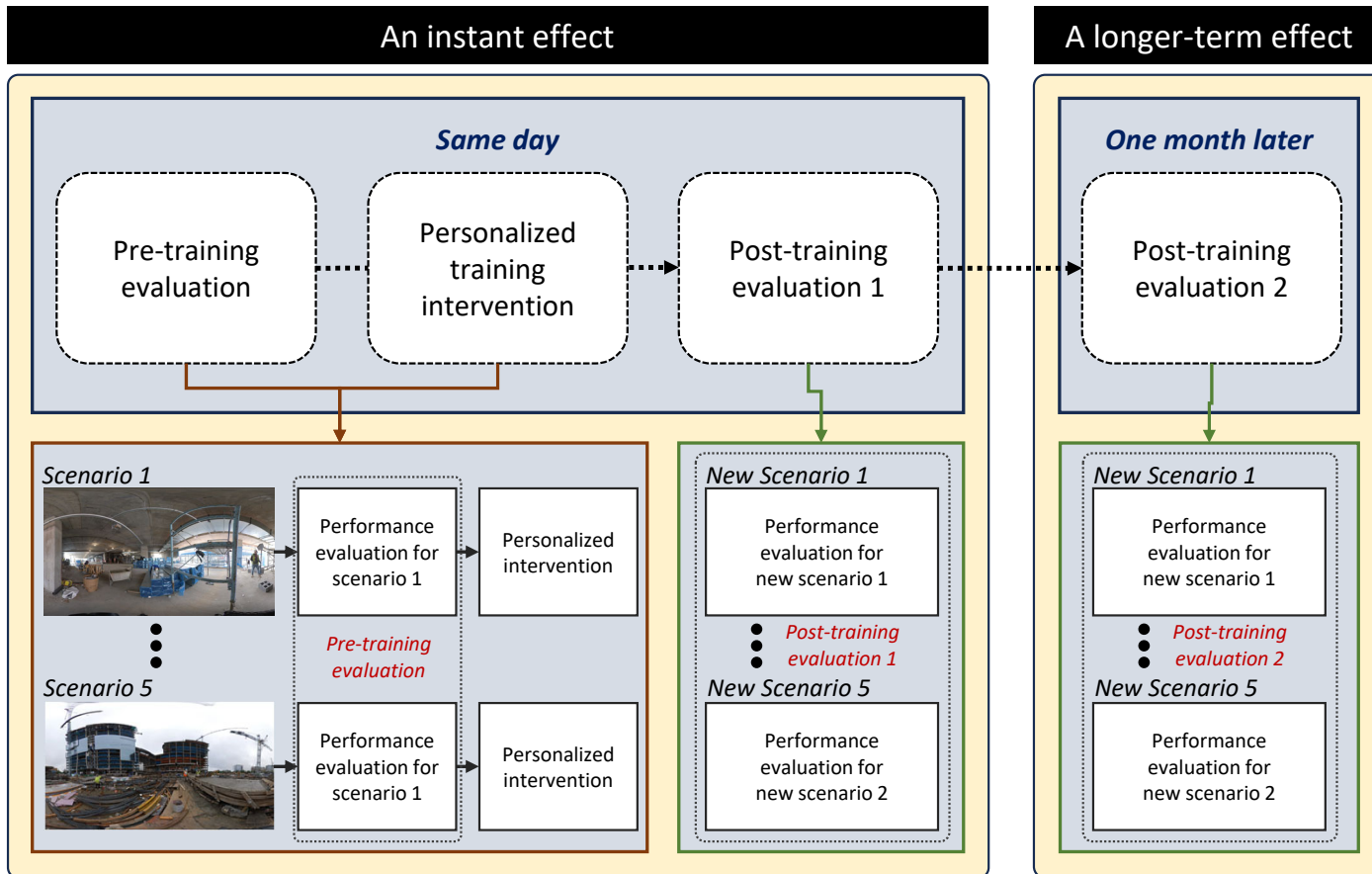
CAUTION
Based on historical data, if you encounter a fall hazard and you miss to identify it, the likelihood of being involved in a fatality is as high as **2-4%**



Aiming to **enhance workers' perceived risk level** by illustrating the severity of potential consequence of various hazard types

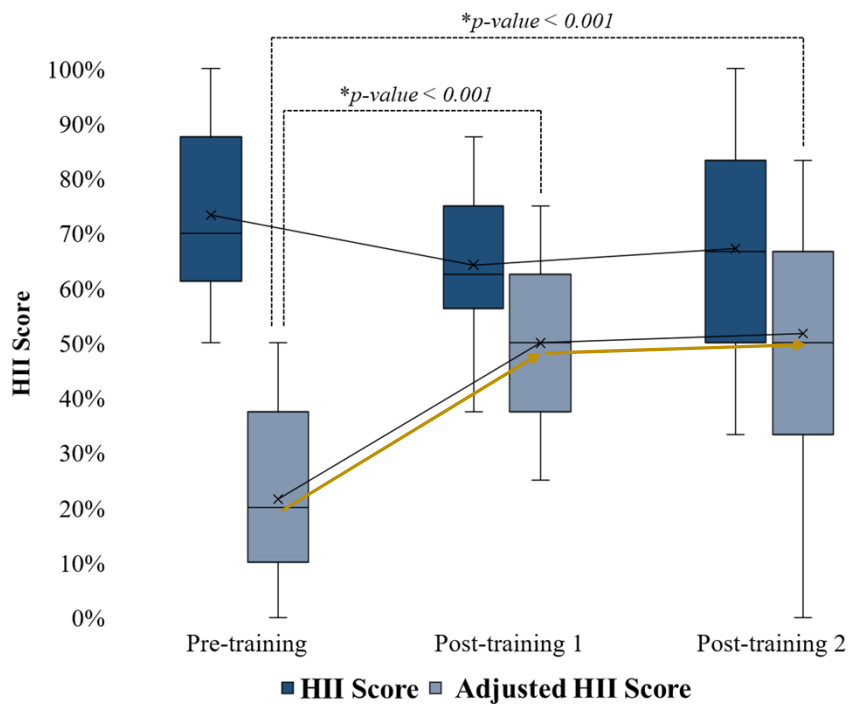
Training Development

Overview of the validation process



Training Development

Improvement in hazard Identification performance



Original HII score:

The ratio of identified hazards to the total number of hazards

vs.

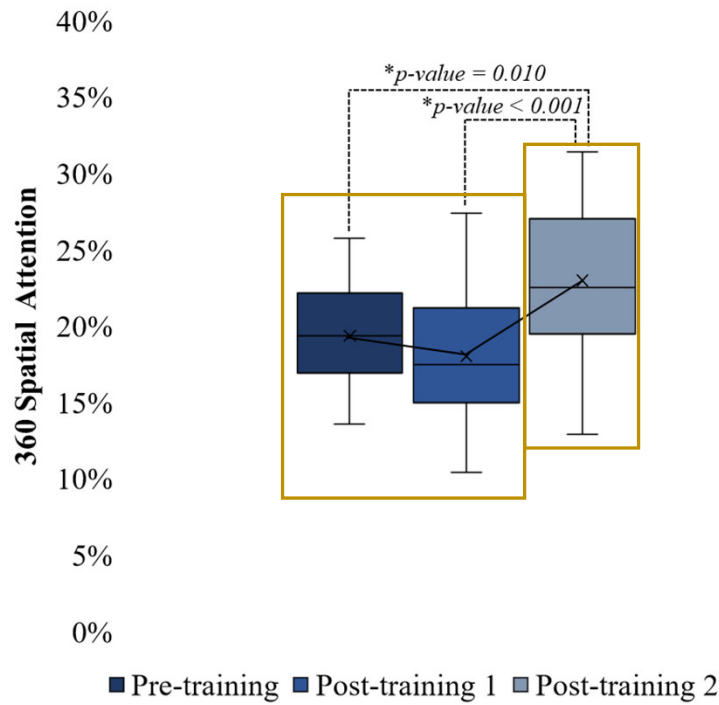
Adjusted HII score:

The ratio of identified hazards (*excluding improper identification*) to the total number of hazards

Adjusted hazard identification scores were significantly increased in post-training sessions

Training Development

- Improvement in visual search strategies



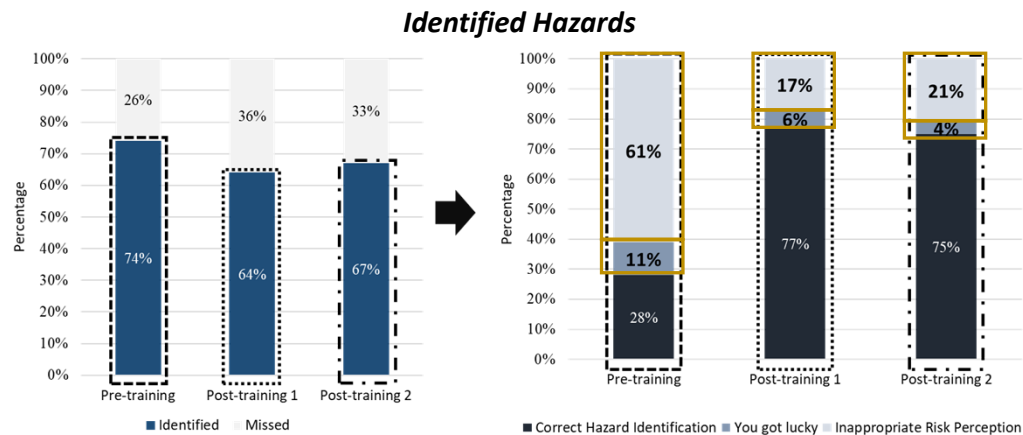
Subjects showed slightly lower 360 spatial attention abilities in Post-training session 1

Virtual reality fatigue from experimental design

Participants' 360 spatial attention abilities were **remarkably improved** in Post-training session 2
(Improvement in visual search skills)

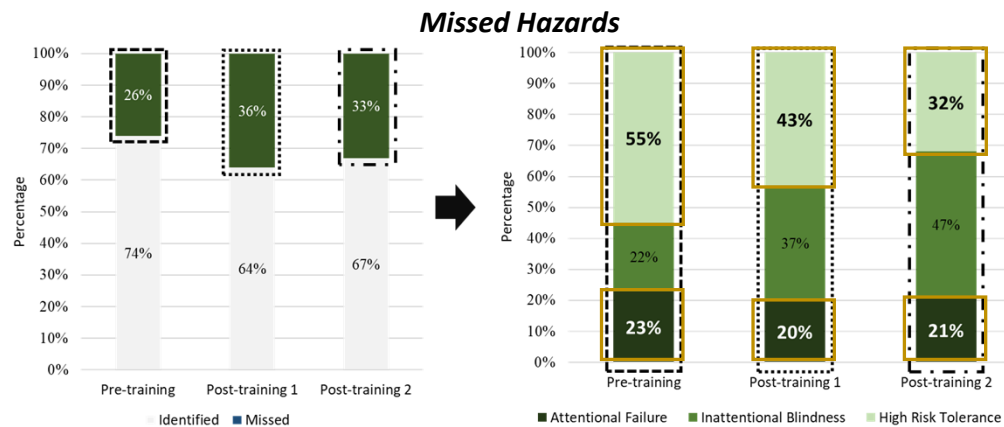
Training Development

Reduction of cognitive failures and improper identification



Perceptual-based errors were significantly reduced in post-training sessions

Skill-based errors were marginally reduced in post-training sessions



Knowledge acquisition

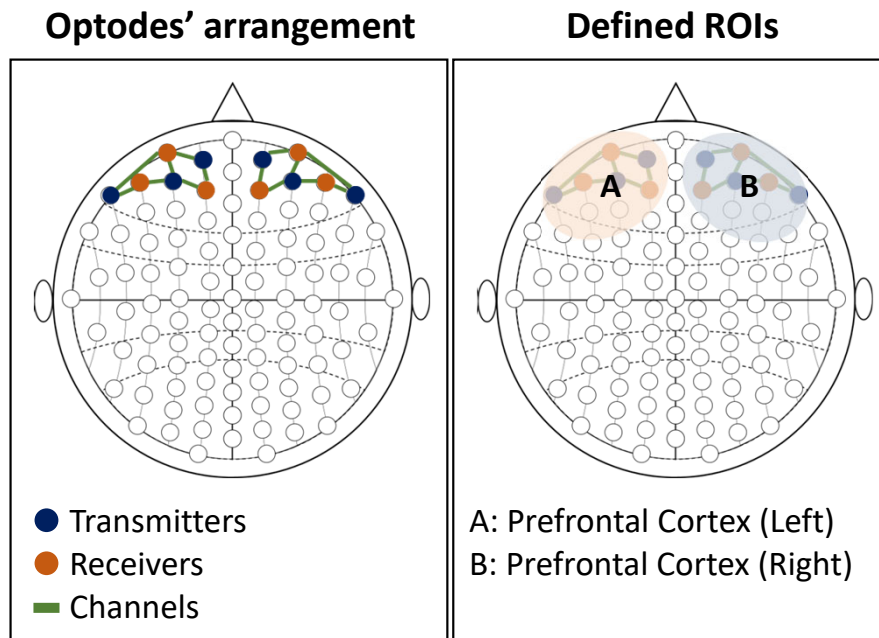
VS.

Skill development

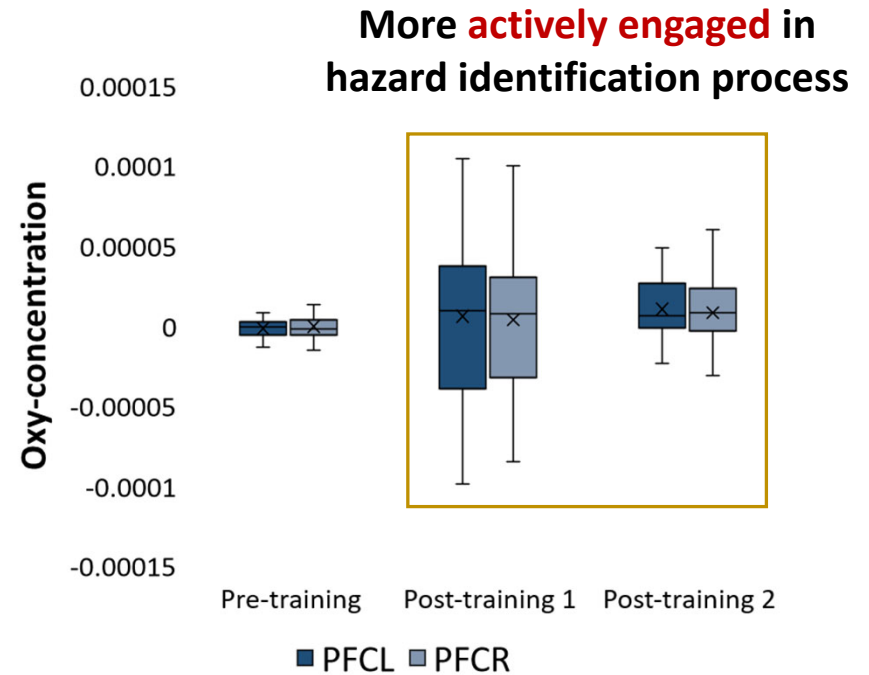
Require repeated training

Training Development

- Changes in brain activation

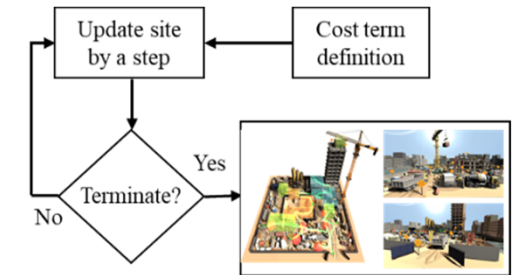


Functionality of PFC: Decision-making / Integrating sensory information

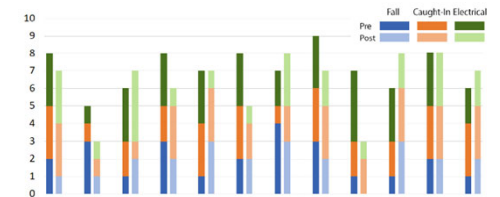


Slightly higher activation of prefrontal cortex in Post training sessions

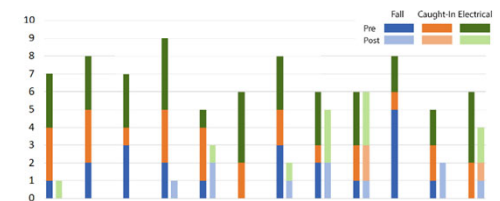
Synthesizing and Personalizing Safety Training Scenarios



$$C_{\text{total}}(R, H) = w_D C_D(R, H) + w_S C_S(R, H) + w_T C_T(R, H),$$



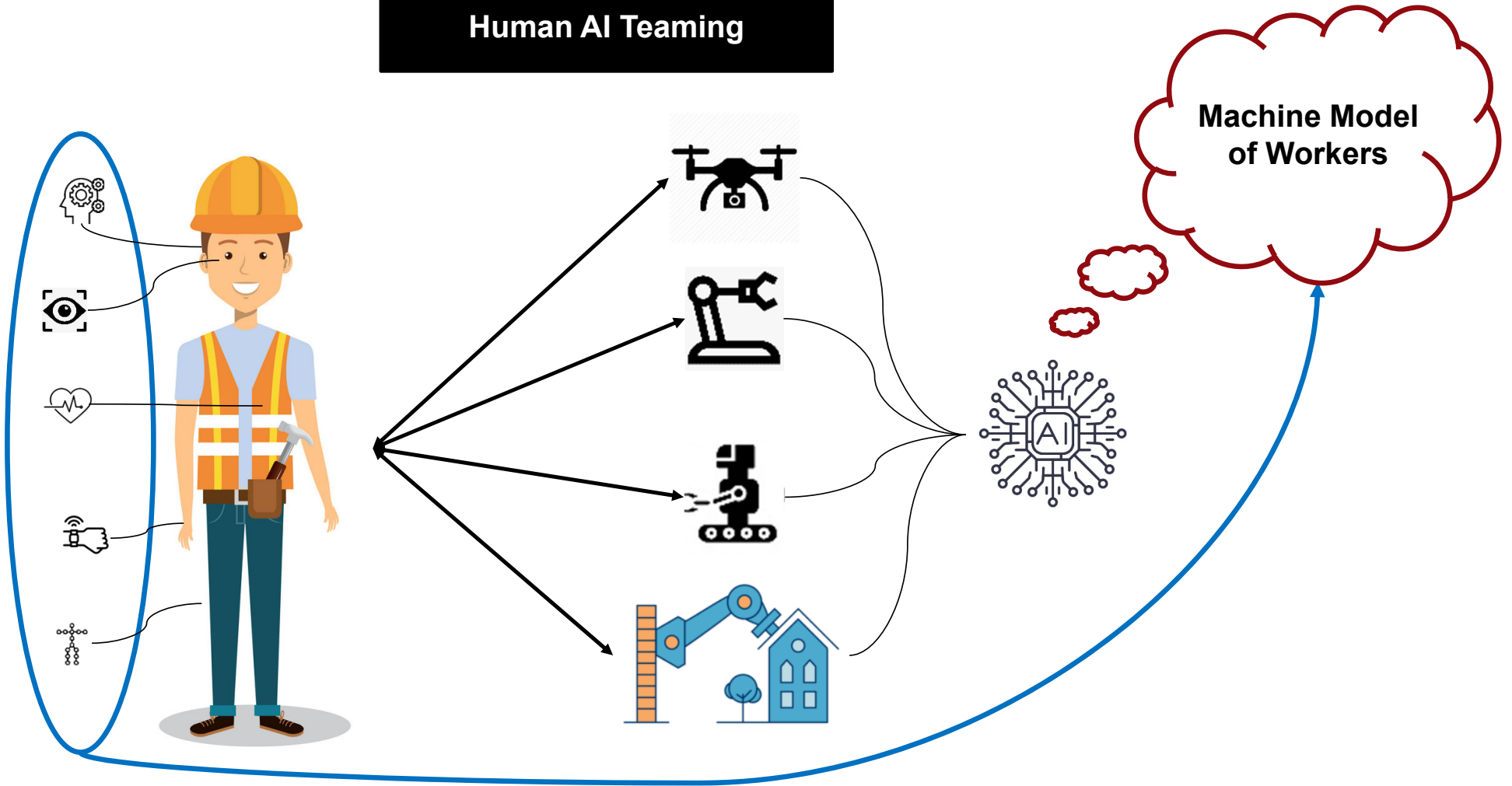
(a) The Control Group.



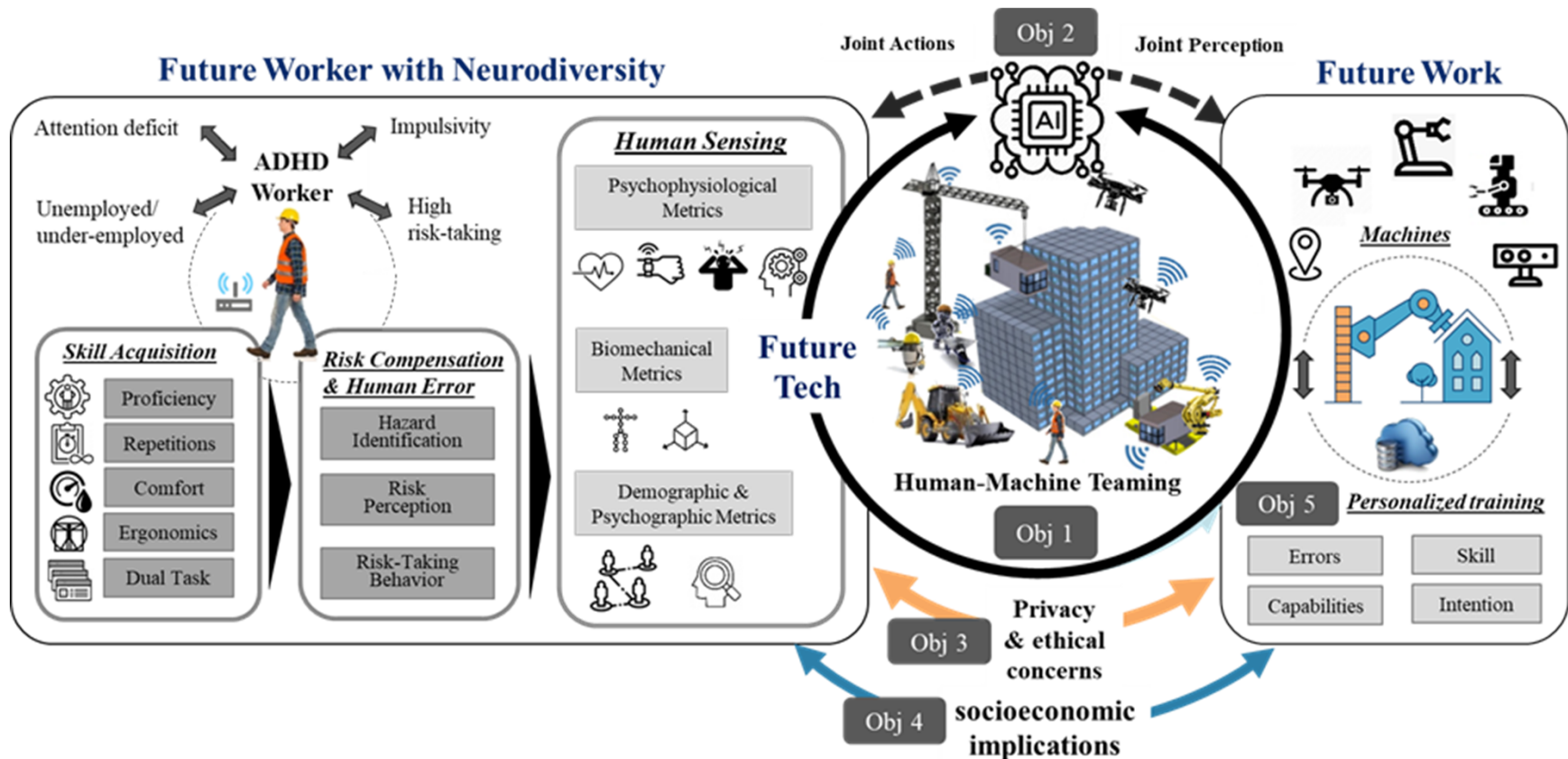
(b) The VR Group.

5. Cyber-Physical-Human Systems

Human AI Teaming



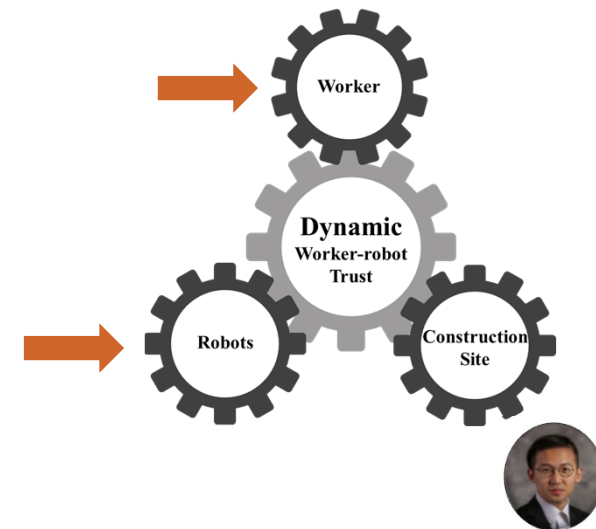
Worker-AI Teaming to Enable ADHD Workforce Participation



Worker-Autonomy Trust

- **Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks**

Chang, W. C., Ryan, S. M., Hasanzadeh, S., & Esmaili, B. (2023, July). Attributing responsibility for performance failure on worker-robot trust in construction collaborative tasks. In EC3 Conference 2023 (Vol. 4). European Council on Computing in Construction. *(Best paper)*



Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

- The **perfection** of autonomous agents still **cannot be guaranteed**, especially in such dynamic and unpredictable workplaces.
- Their unexpected performance (e.g., **failures**) will **decrease human trust** levels.
- In human-robot collaboration, **responsibility attribution for the failures** could be **humans** (*wishful thinking*) or **robots** (*self-serving bias*).
- Two types of **trust transfer** (i.e., **multi-tasks** and **multi-agents**) have been proposed by the literature.

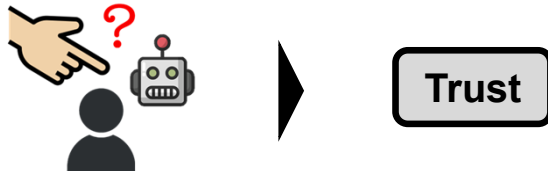


Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

Limitations of current literature:

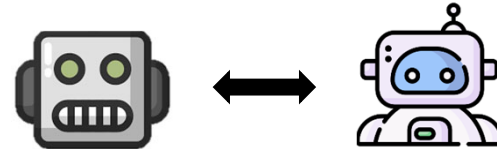
1

*The impact of **responsibility attribution** for failures on **trust***



2

*Limited discussion on **multi-agent trust transfer***



This study aims to investigate the effect of **attributing responsibility for failure** and **trust transfer** on worker-robot trust in the future jobsite



Background

Point of Departure

Methodology

Results

Discussion

Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

Autonomous Agents



Bricklaying Robot

(Lift/drop blocks)



Drones

(inspection, survey, delivery)



AI-assistant

(Provide drone info.)

Data Collection



1 HTC VIVE
(eye-tracking)



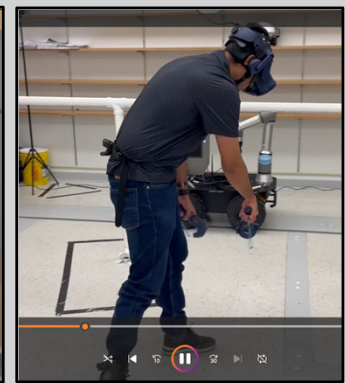
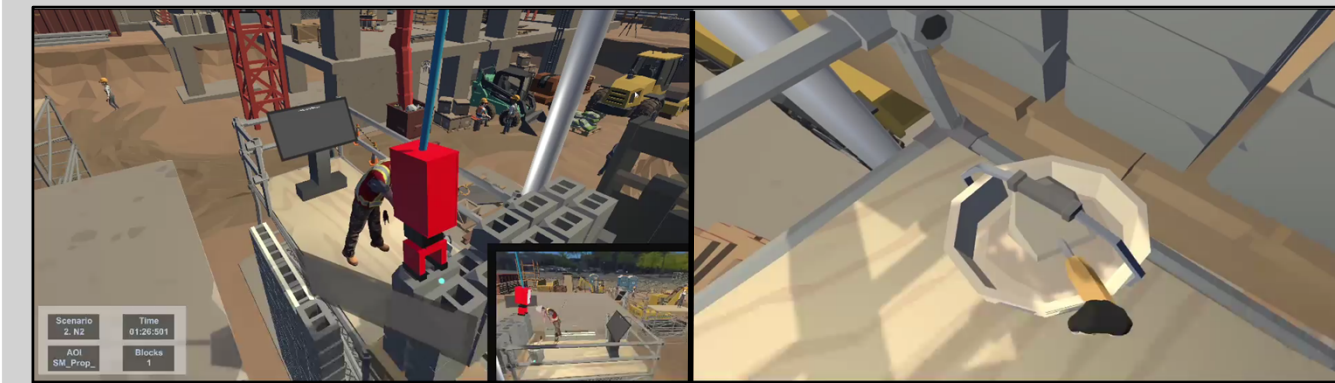
2 Trackers
(motion)



3 Sensors
(Psychophysiological)



Synchronization



Background

Point of Departure

Methodology

Results

Discussion

Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

Trust Survey

Baseline Module

Trust Survey

Error Module

Trust Survey

Post Survey

- To what extent the auto-agent's behavior can be **predicted**?
- To what extent can you **depend** on the auto-agent to do its job?
- What degree of **faith** do you have that the auto-agent will be able to cope with similar situations in the future?
- Overall, how much do you **trust** the auto-agent?

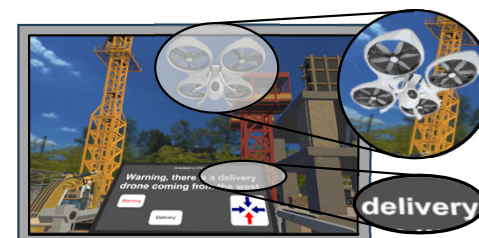
① ~ ⑤



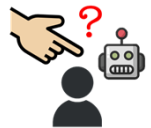
"Robot failure"



Drone failure



AI-assistant failure



Responsibility Attribution

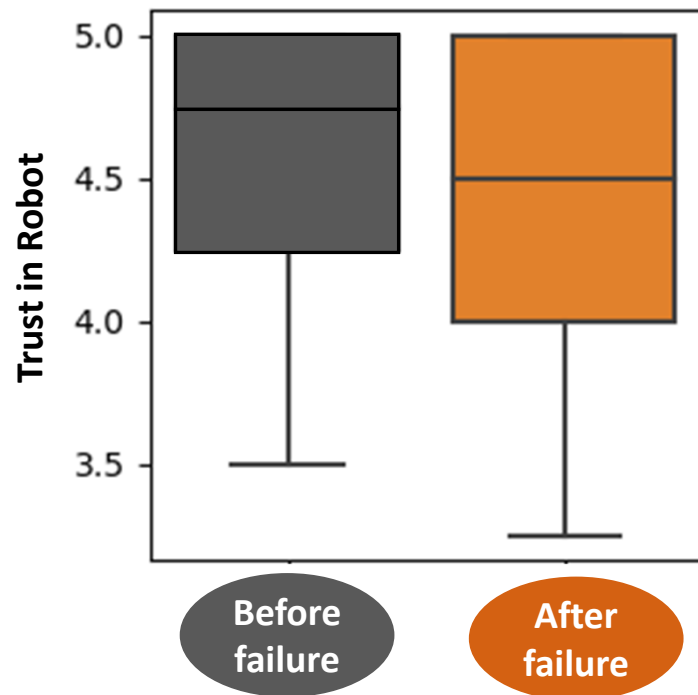
Human

Robot



Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

The impact of **robot failures** on trust changes

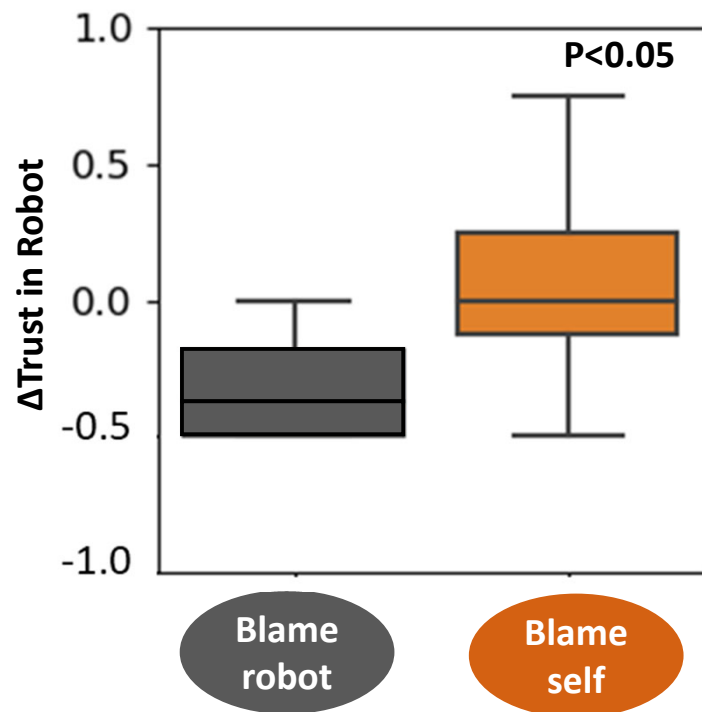


Robot failure did not significantly affect worker trust



Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

The impact of **responsibility** on trust changes



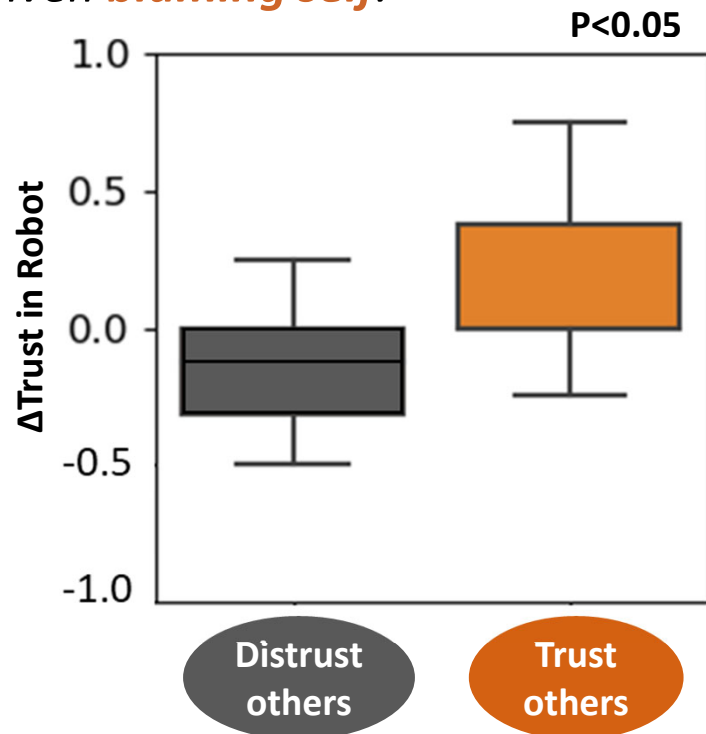
Robot failure did not significantly affect worker trust

Attributing responsibility significantly affected workers' trust changes



Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

Given *blaming self*:



Robot failure did not significantly affect worker trust

Attributing responsibility significantly affected workers' trust changes

Distrust in other agents significantly affected workers' trust change



Attributing Responsibility for Performance Failure on Worker-robot Trust in Collaborative Tasks

Robot failure did not significantly affect worker trust



*Those participants might **overtrust** a faulty auto-agent*



Attributing responsibility significantly affected workers' trust changes



Distrust in other agents significantly affected workers' trust change



Inappropriate** trust transfer might lead to **overtrust** / **undertrust

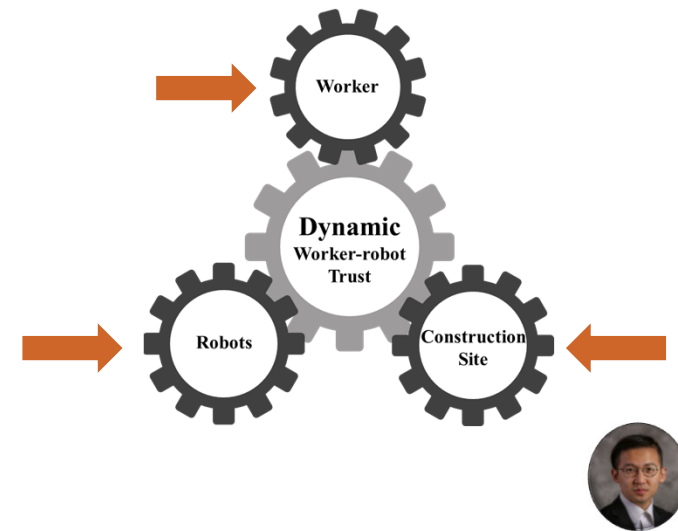


Safety issues



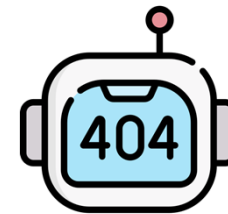
Situational Awareness

- **The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure**



The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

- Workers must **allocate attention** to dynamic objects (e.g., robots) to maintain **situational awareness (SA)**.
- The **decreasing trust**, caused by failures, makes workers **more attentive to robots**.
- **Time pressure** forces workers to focus on the designated tasks and to become **less attentive to robots**.



The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

Limitations of current literature:

1

*Not considering the impacts of **robot-related** and **work-related** factors on trust simultaneously*



This study aims to examine which factor plays a **dominating role** in affecting workers' SA



Background

Point of Departure

Methodology

Results

Discussion

The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

Autonomous Agents



Bricklaying Robot



Drones



AI-assistant

(Dynamic objects)

Data Collection



1 HTC VIVE



2 Trackers



3 fNIRS

(Eye-tracking)



Performance Metrics



Dwell Time



Run Count



Dynamic drones

(Situational Awareness)

(Areas of Interest)

Dwell Time

How much time did participants fixate on AOIs

Run Count

How many times did participants return their attention to AOIs



Background

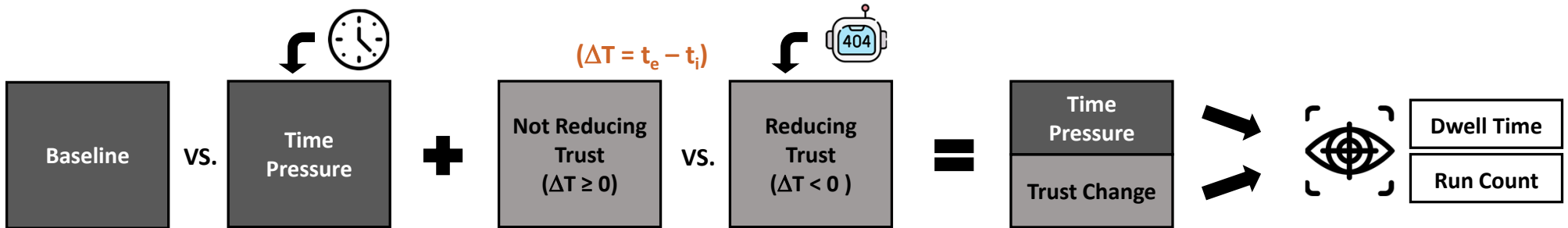
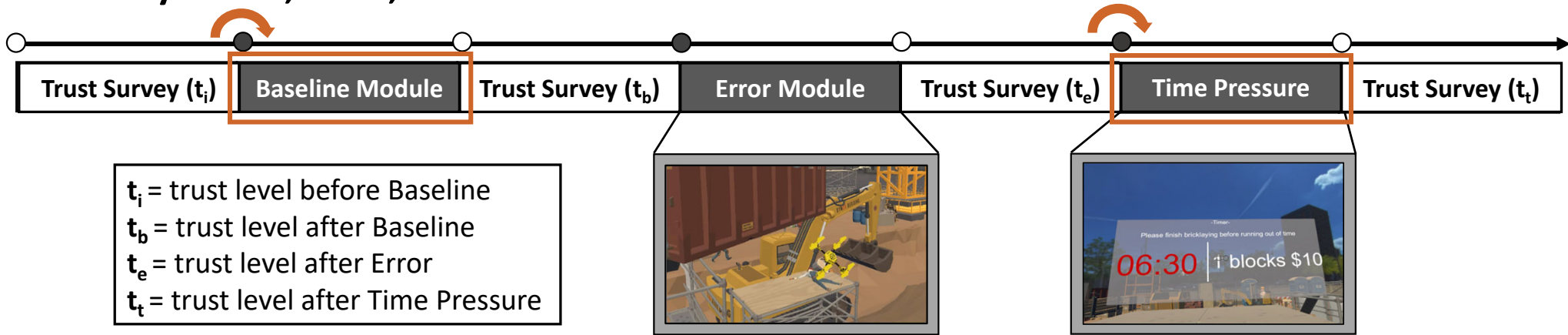
Point of Departure

Methodology

Results

Discussion

The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure



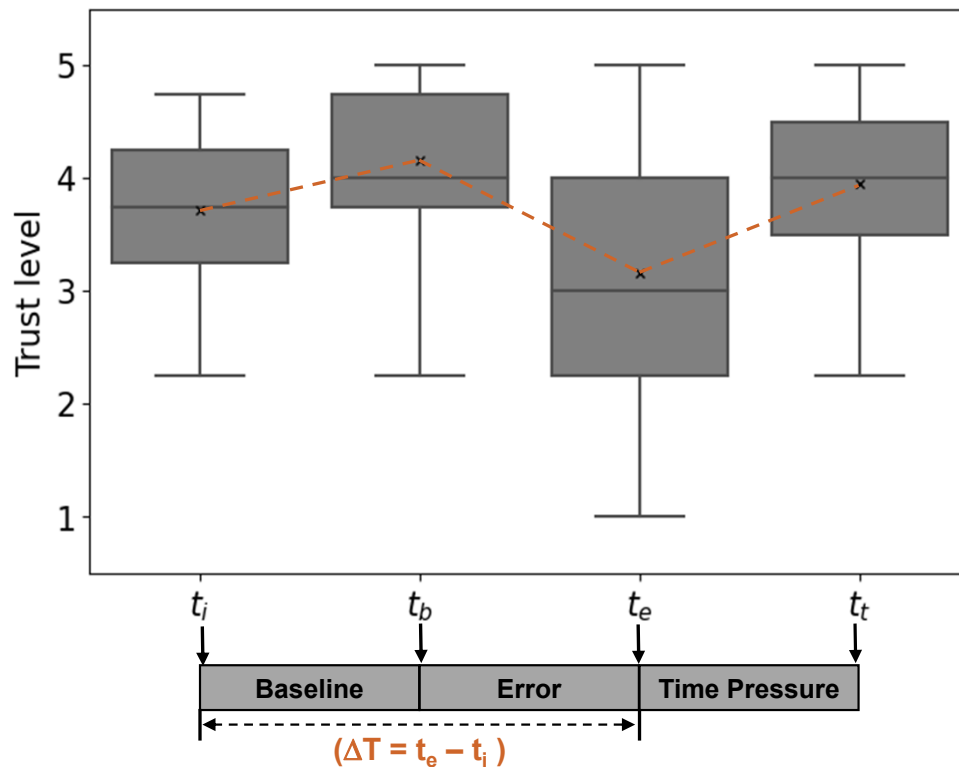
Within-subject variable

Between-subject variable

Mixed-designed ANOVA



The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

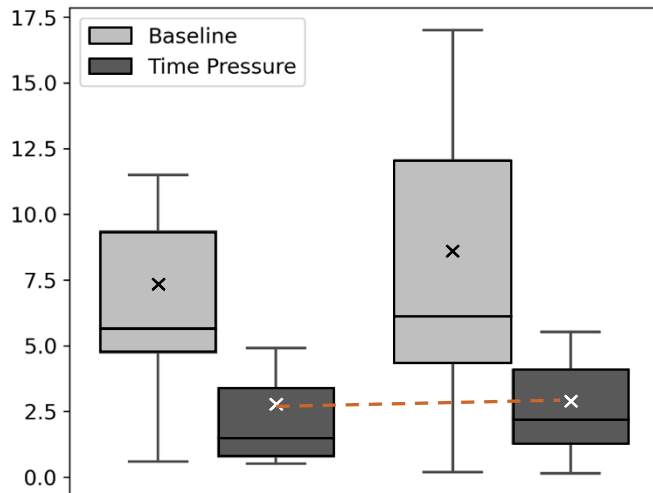
**72%**Reducing Trust
($\Delta T < 0$)

vs.

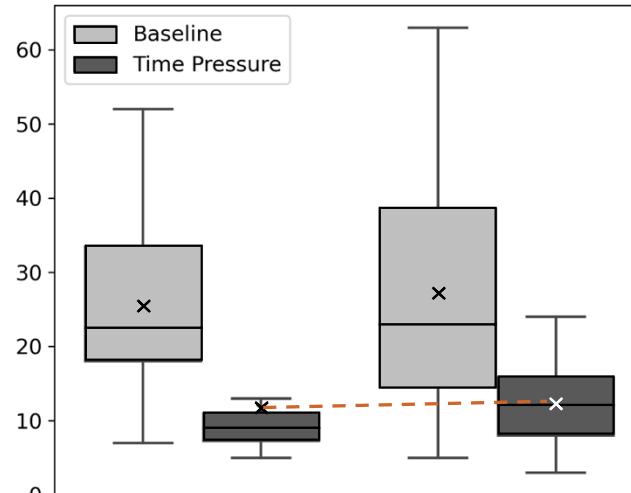
28%Not Reducing Trust
($\Delta T \geq 0$)

The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

Dwell Time



Run Count

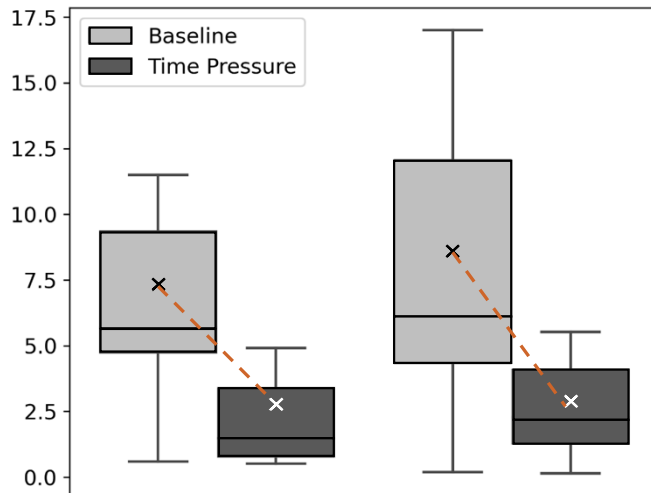


Decreasing trust did not significantly affect the SA

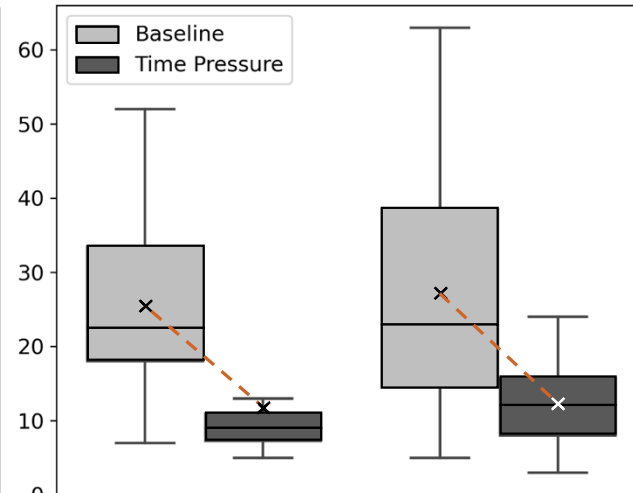


The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

Dwell Time



Run Count



Decreasing trust did not significantly affect the SA

Time pressure significantly affected the SA



The Importance of Situational Awareness (SA) in Future Jobsite: Toward the Effects of Faulty Robot, Trust, and Time Pressure

Decreasing trust did not significantly affect the SA

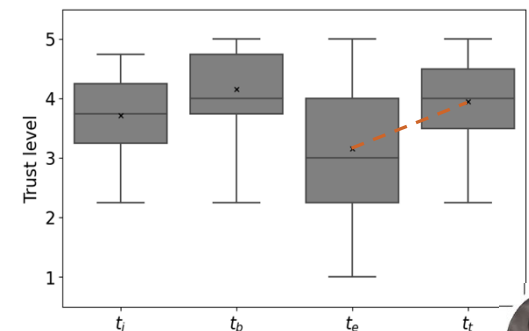
Time pressure significantly affected the SA

Time pressure plays a dominating role in affecting worker SA than trust

Time pressure provokes attentional tunneling and overtrust for workers

SAFETY FIRST

Safety issues



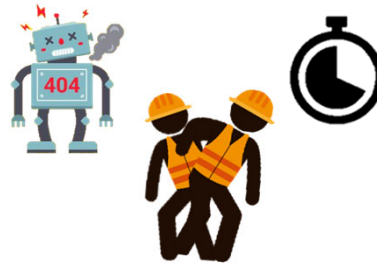
Conclusions

What are the next steps after identifying the following uncertainties?

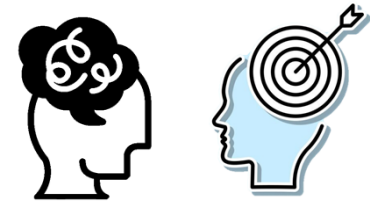
Trust-building

TOO
MUCH

Situational
Awareness



Inclusivity



Human-centric, sustainable, and **resilient** worker-autonomy teaming







ANY QUESTIONS?

Thanks for your Time and Attention!