



# RELIABILITY REQUIREMENTS FOR AUGMENTED REALITY IN VISUAL SEARCH TASKS

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- Augmented reality (AR) technologies have great potential to improve battlefield performance
- Soldiers must process information from an outside source and integrate it into their decision making
- AR that fails to provide correct information, or provides incorrect information, may harm performance
  - e.g., unnoticed failures, distractions, distrust of accurate information, etc.







- Augmented reality-aided target ID will not be perfect (Biros, Daly, & Gunsch, 2004)
- Soldier trust is required for use/adoption of new technology, as distrust = disuse (Parasuraman & Riley, 1997)
- What level of AR accuracy is necessary?
  - ...to improve human performance?
  - …to facilitate trust?







• Errors in the false-alarm prone AR will be more damaging to both objective performance and subjective state than miss-prone AR



• Errors (either type) above distant targets will be more damaging to objective performance than errors above close targets





- Participants asked to spot tanks in 54 consecutive grassland scenes
  - Each scene contained between 0 and 8 targets
- Search task guided by intuitive AR icons:



- Participants assigned to one AR error-type condition: false-alarm prone or miss-prone
- AR reliability varied throughout: {25%, 40%, 55%, 70%, 85%, 100%}
  - Reliability corresponded to number of AR mistakes in a scene



# MISS AND FALSE ALARM EXAMPLES









- We used a simple visual perception task that did not require previous experience (so anyone could participate)
  - Sample should match population on <u>relevant variables</u>
- Total of 184 participants recruited in person and over the Internet
- Internet participants excluded for poor screen resolution (n=32) or for not finishing the task (n=12)







- Useful for describing how participants experience the task
- Similarities/differences between objective and subjective metrics are informative (i.e., not recognizing safety hazards)
- Three self-report measures:
  - Survey on Trust in AR ("How much do you trust the AR to help you?")
  - Overall workload scale from NASA Task Load Index ("How hard was that?")
  - Gas Tank Questionnaire ("How much energy do you have left?")











- Participants missed targets when paired with false alarm-prone AR
  - Visual field cluttered with AR-marked targets: participants missed valid targets
- Miss-prone AR never hurt performance
  - Visual field missing AR-marked targets: participants nonetheless found valid targets







- Distant targets were more difficult to discern: participants missed more of them
- Distant targets magnified the undesirable effects of unreliable AR
  - Distant targets even *more* likely to be missed with false alarm-prone AR (despite always being properly marked)







- Participants incorrectly selected non-targets when paired with false alarm-prone AR
  - Visual field cluttered with AR-marked targets: participants selected invalid targets
- Miss-prone AR never hurt performance
  - Visual field missing AR-marked targets: participants were not tempted to select invalid targets







- Distant targets were more difficult to discern, and elicited more false alarms than close ones
- Distant targets again magnified the effects of unreliable AR
  - Erroneously-marked distant targets were *much* more likely to elicit false-alarms compared to close targets







- Miss-prone AR: participants increased their search time as AR mistakes increased (adaptive)
- False-alarm prone AR: participants reduced their search time as AR mistakes increased (maladaptive)
  - Less diligence/gave-up (but still made more responses)







- Participants reported greater resource drain with miss-prone AR
  - Coupled with longer search time: potentially working harder (i.e., not giving up)
- Trust surveys and self-reported workload were similar between AR types
  - Participants subjectively unaware of differences between conditions







- False alarm-prone AR was more damaging to accuracy
  - Increased both misses and false alarms
- Participants with miss-prone AR could compensate for poor AR by increasing effort
  - False alarm-prone AR overwhelmed participants, who responded with less effort
- Similar trust and workload self-report, despite objective performance differences
  - Danger: in some cases, participants are unaware of when AR may hurt





- We spend a great deal of effort avoiding "misses," but participant misses increased with false alarm-prone AR
- Findings were consistent with prior research: false alarms can be more damaging at the same level of performance, are annoying and distracting
- Participants were unable or unwilling to pay the cognitive cost of working with false alarm-prone AR: more difficult task
- Even with highly motivated soldiers, perseverance may cost greater mental effort and result in inevitable mistakes









- Future AR systems may allow users to adjust detection thresholds (i.e., sensitivity)
  - Knowing that misses and false alarms are not equivalent, how much freedom do we give users?
- Users rated trust similarly, despite differences in objective performance – soldiers may not recognize risk of false alarms, especially if misses are "high cost"
- Potential solution: employ system constraints and/or user training to prevent alert oversaturation and disengagement from false alarms







- Effects of unreliable AR on participant performance were magnified with distance
  - At range, AR is providing more support; human cannot compensate
- AR systems aiding target search will require greater reliability at longer ranges to improve human performance
  - This finding may generalize to more difficult perceptual tasks in general
- Do not automate/augment with insufficient accuracy only automate/augment what you can do well







- AR accuracy required to improve human performance depends on contextual factors (AR error type, target range, etc.)
- In visual search, false alarms are more damaging to performance than misses are
- Disparity between objective performance and subjective responses suggests potential risk
  - "We can't compensate for poor AR if we don't know that it's hurting us"
- AR will rarely be perfect, but <u>it should improve human performance</u> over a "manual performance" baseline





# NVESD Perception Laboratory Augmented Reality Program Overview





- NVESD interests: Sensor feeds, see-through displays
  - Many potential benefits to AR technology for Soldiers (situational awareness, decision-making, communication, etc.)
- Assumption: providing Soldiers with AR information will improve their performance
- Many factors affect quality of AR information
  - Perceivable?
  - Intuitive?
  - Timely?
  - Relevant?
  - Accurate?
- Initial research areas:
  - Visual Search,
  - Target Acquisition,
  - Vehicle Identification,
  - Navigation







# Our simulations currently focus on AR accuracy and human performance:

- How accurate does AR have to be in order to improve performance?
- What are the worst types of errors an AR system can make?

#### These are task-specific and potentially device-specific questions

#### Goals:

- 1) Contribute to general AR usage guidelines
- 2) Adapt our existing simulation capabilities to be able to define sensor- and task-specific AR requirements







# **Target Acquisition and Spatial Error**





#### **Research Objectives:**

- Evaluate AR aid to target acquisition performance and how errors impede performance
- Evaluate the level of AR accuracy necessary to improve target acquisition performance at various ranges



Example imagery depicting a scenario with an AR designation perfectly aligned (left) and misaligned (right)







# Scene Generation in Night Vision Image Generator Software (NVIG):

- Virtual humans arranged in a 60° arc around the sensor, placed closely together (1m apart)
- A single target held an AK-47
- Participants: 18 U.S. Soldiers







- 6 AR Conditions: No AR, Perfect AR, plus 1°, 2°, 3°, and 4° of angular error
- 3 Ranges: "Close," "Intermediate," and "Distant"
- Targets placed randomly within 3° sections, centered at 6°, 9°, 12°, and 15°
  - Target locations were counterbalanced across all AR Condition and Range combinations
- 144 trials, divided in 8 blocks (rest)
- Counterbalanced the 8 blocks of trials by AR Condition and Range













# "INTERMEDIATE" RANGE





















- Participants stayed for 5 days (vehicle identification training, other experiments); two cohorts
- Highly realistic sensor grips simplified controls for optical zoom, "speed boost," and target designation
- Group presentation on experiment and controls
- 27 training trials (3 trials each of No AR, Perfect AR, and 4° of angular error at each range)
- Experiment: breaks as desired between blocks of trials,10 minutes at halfway point
- Length: ~120 minutes







# Significant Main Effects:

- Range
- AR Condition

Compared to No AR:

- All AR conditions improved
  performance
- All AR conditions protected against increased reaction time with increased range
  - Protection decreased w/angular error







Compared to Perfect AR:

- 1° and 2° did not significantly impair performance, but 3° and 4° did
- 3° and 4° also showed greater increases as range increased
- No significant differences at "Close" range for any imperfect AR conditions
- 3° significantly worse at "Distant" range
- 4° significantly worse at "Intermediate" and "Distant" ranges







**Results Summary:** 

- Incremental degradations in AR accuracy produced progressive degradations in performance
- Even imperfect AR was always beneficial in this simulation (won't be true for every task)
- Greater AR accuracy is needed at greater ranges: all AR yielded approximately the same benefit at the "Close" range, but greater error at the "Distant" range yielded deficits compared to perfect AR





# **Vehicle Identification Accuracy**





#### **Research Objectives:**

- Evaluate AR aid to vehicle identification performance and how errors impede performance
- Evaluate the level of AR accuracy necessary to improve vehicle identification performance at various ranges







#### Independent variables

- AR Conditions: 100%, 75%, and 50% reliable AR, No AR pretest and posttest
- 3 Ranges

"Intermediate"

 Time to make a decision: unlimited time vs. 5 second time constraint

**Dependent Variables:** Accuracy & Response time

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- 20 U.S. Army Soldiers trained on infrared vehicle ID prior to experiment
- Scene Generation in NVIG
- Sequence: Training, No AR Pretest, AR Trials, No AR Posttest
- No AR and AR components had both a time-constrained and a timeunlimited portion
  - All participants took both
  - Randomly assigned to always begin with time limit or no time limit
- **Baseline:** 72 Images (3 blocks of 24, 1 block per range)
- **AR Trials:** 216 images (9 blocks of 24, 1 block per range X AR reliability)
- Participants were told to evaluate different ostensible AR systems
  - After each block of images, asked to reset their trust
- Participants asked to take breaks between sections of the test









# ACCURACY WITH AR COMPARED TO UNAIDED PERFORMANCE



## "Close" Range

- Perfect AR approached significant improvement
- Imperfect AR: non-significant

### "Intermediate" Range

- Perfect AR: significant improvement
- Imperfect AR: NS
- Time-constraints caused a greater reduction in perfect AR accuracy

## "Distant" Range

 All AR information is a significant improvement





# ACCURACY WITH IMPERFECT AR COMPARED TO PERFECT AR



#### "Close" Range

- 50% reliable AR caused significant impairments
- 75% reliable AR caused impairments that approached significance

# Intermediate and Distant Ranges

 Both 50% and 75% reliable AR caused significant impairments







#### Significant main effects

- AR Condition
- Distance to target (i.e., range)
- Time constraint

 Substantial shift b/w pretest and posttest





# RESPONSE TIME WITH AR COMPARED TO UNAIDED PERFORMANCE



#### "Close" Range

- 50% and 75% reliable AR significantly slower
- Perfect AR slower, approached significance

### "Intermediate" Range

- 50% reliable: significantly slower
- 75% reliable: slower, approached significance
- Perfect AR: significantly faster
- Time-constraints: perfect AR decrease was not as severe

## "Distant" Range

- 50% reliable significantly slower
- 75% reliable and perfect AR <sup>0</sup>L slower, but not significantly slower





# RESPONSE TIME WITH IMPERFECT AR COMPARED TO PERFECT AR



# "Close" and "Intermediate Range

 Both 50% and 75% reliable AR: significantly slower

### "Distant" Range

- 50% significantly slower
- 75% not significantly different







- Progressive AR error yielded progressive impairments
- Participants were able to use perfect AR effectively
- Greatest benefit w/perfect AR, at "Distant" range, with unlimited time
- Imperfect AR was only clearly beneficial at "Distant" range when participants were clearly struggling
  - 50% reliable AR always slower compared to No AR and perfect AR
  - 75% reliable AR showed many similar impairments, just less severe

### • Trends by range:

- Close: little AR benefit, yet slowed participants down
- Intermediate: Perfect AR clearly beneficial, but time-constraints hurt improvement more than other conditions
- Distant: All AR beneficial, greater reliance on AR

## Time Constraints

- Significantly impaired performance
- Generally affected AR conditions similarly (except "Intermediate w/perfect AR)
- Most benefits to Accuracy usually a relatively small cost of speed





- Experimental design: No AR trials were separate pretest and posttest
  - Designed intentionally to capture learning/practice effects and to reduce length of core experimental trials
  - Became disadvantageous with severe learning effects
  - May overestimate baseline performance compared to other conditions
  - Future iterations: additional initial practice with NVIG simulated imagery and integrate baseline with other trials
- Participants in our study expect AR mistakes
  - May have caused additional skepticism with perfect AR (slower responses)
  - Results may not generalize to unexpected AR errors
- Broad AR reliability intervals (25%)





# **Future Efforts**





- Target Acquisition: target density, clutter, field of view, & field of regard
- Vehicle Identification: Algorithms biased towards threats and other imagery degradations
- Visual Search: Misses and false alarms in a high clutter environment
- Land navigation: imperfect waypoints









- Immersive AR Display simulations conducted in NVESD's mixed-reality Virtual Prototyping Holodeck (VPH)
- Target acquisition in an immersive environment
- Eye-tracking studies examining effectiveness and efficiency of AR Symbology



A person in the VPH (left) and the scene he sees in his VR display (infrared scene rendered by NVIG)



A Soldier gives a hand signal (left) and how he appears to a fellow Soldier in the VPH (Right)



