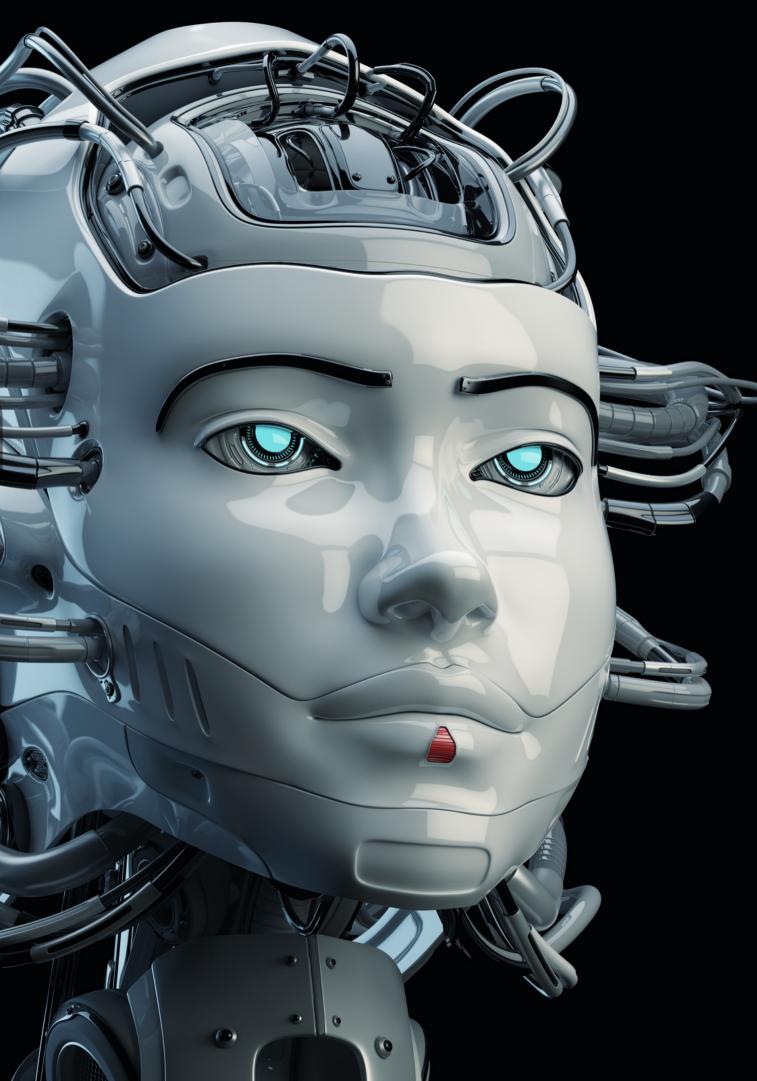
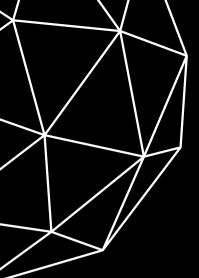
THE UNIVERSITY OF ALABAMA IN HUNTSVILLE

Stakeholder Elicitation For Autonomous Robotic Systems Design For Mission-Critical Environments

NICOLE MOORE





STAKEHOLDER PREFERENCES

Stakeholders

A firm's investors, customers, and employees. (Pirson & Milhotra, 2011)

Preferences

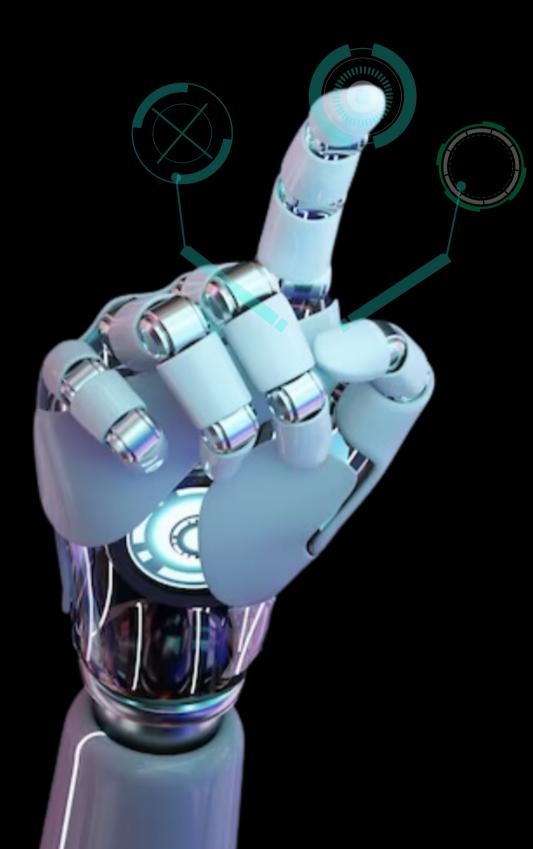
Existing values that are constructed and can be elicited (Tversky & Thaler, 1990)

Preference Formation

• Contextual predispositions • Prior information obtained regarding the choice • Framing of the decision • Prior biases (Desai & Krajbich, 2022) • Personal subjective values • Confidence within the choice (Lee & Daunizeau, 2020)



PREFERENCE ELICITATION METHODS

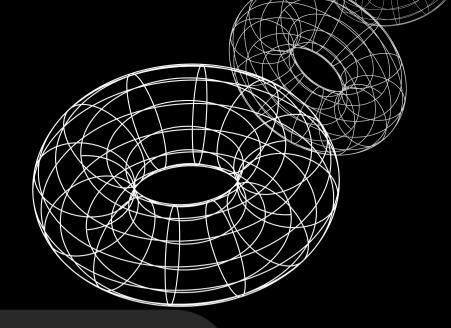


Preference Elicitation

- Process in which a model is developed to identify which preferences the stakeholders favor best. • What is gained (i.e. implicitly/explicitly) through a utility
- function? (Zintgraf et al., 2018)

Methods

- relative comparisons between items
 - scoring or ranking
 - clustering by similarity
 - conducting interviews



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MISSION-CRITICAL ENVIRONMENTS





Mission-Critical Environments

- Critical job functions that impact military agency performance (Weger et al., 2022)
- Can result in:
 - Loss of life
 - Financial loss
 - Serious injury to personnel



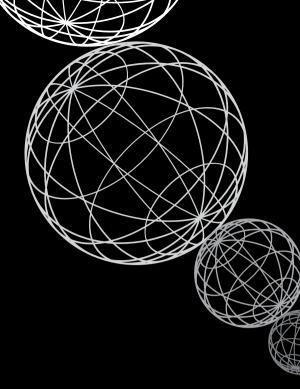
PREFERENCE FAMILIARITY & TRANSPARENCY

Familiarity

- Mere Exposure Effect (Liao et al., 2011)
- Habituation Paradigm (Houston-Price & amp; Nakai, 2004; Zajonc, 2001).

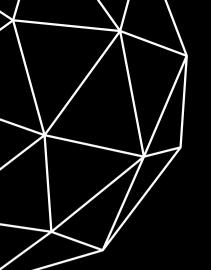
Transparency

- - Transparency
 - Reliability
 - Performance



• Prevalent for autonomous system adoption

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HUMAN ROBOT INTERACTION

Human-Like Features in Robot Design

- Warmth & Competence
- Female Robots
 - Higher on communal dimension (e.g., friendly, polite, affectionate)
- Male Robots
 - Higher on agentic dimension (e.g., assertive, determined, authoritative)

(Carpinella et al., 2017)

Communication Type

- Verbal Communication
- Female Voice
 - Acoustic parameters; emotional prosody (Sokhi et al., 2005)

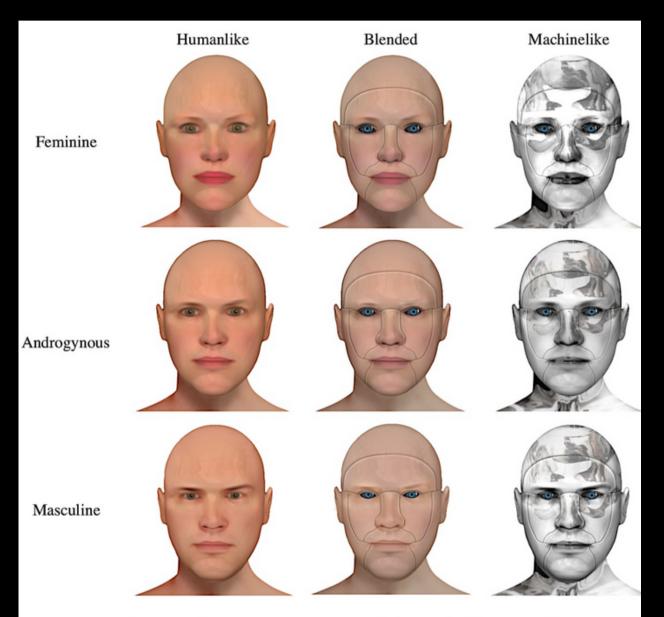
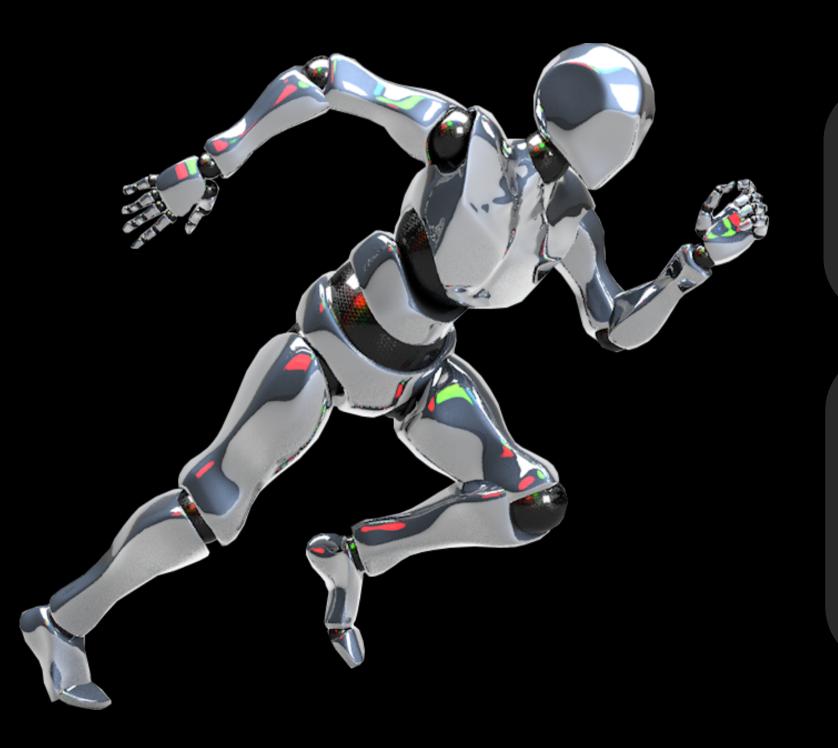


Figure 3. Robot Image Stimuli (Study 4)

(Carpinella et al., 2017)



USER-ROBOT TRAINING METHODS

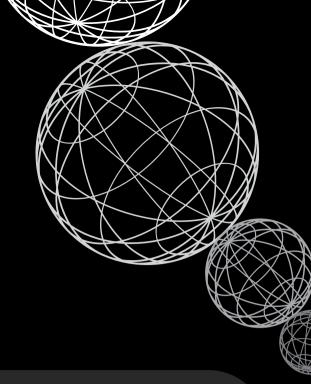


Methods

• Learner-centered approach

Cognitive & Behavioral Components

- Cognitive processes in learning/training:
 - Memory, attention, decision-making, social, and
 - emotional processes
- Accelerated learning methods • Simulation training



• User is active engager (Wiltshire & Fiore, 2014)

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RESEARCH QUESTIONS

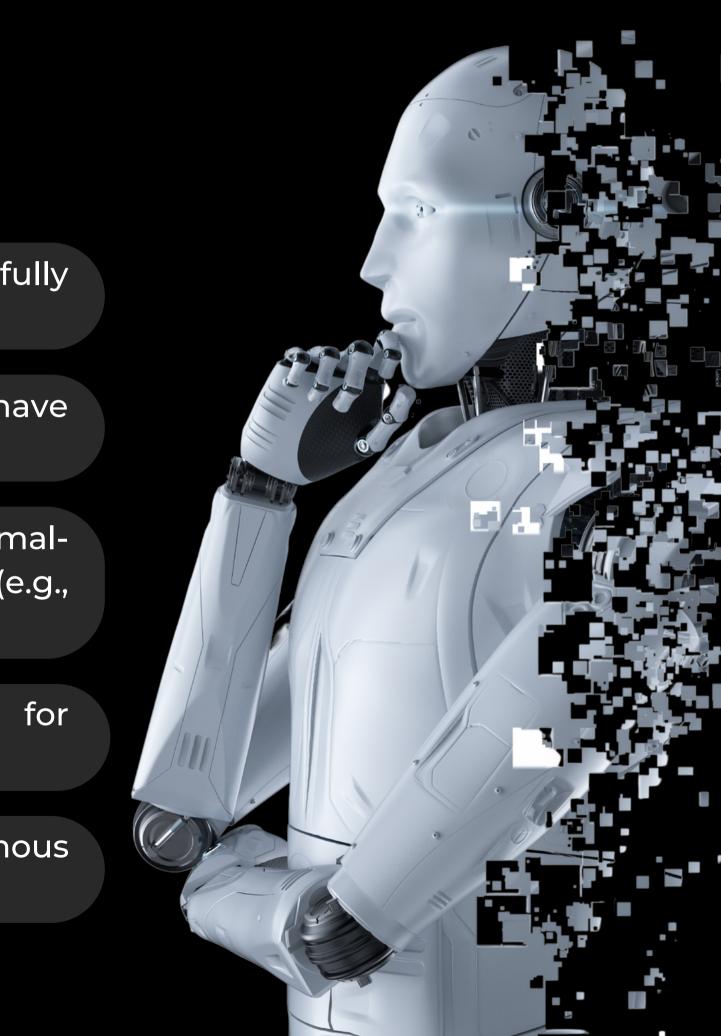
1: What preferences do stakeholders hold regarding the abilities a fully autonomous robot should possess to handle combat situations?

2: What potential barriers or apprehensions would stakeholders have when using an automous robot?

3: Would stakeholders want an autonomous robot to look more animallike (e.g., quadrupedal), human-like (i.e., humanoid), or machine-like (e.g., autonomous mobile robot (AMR))?

4: What are stakeholder preferences regarding vocal tone for autonomous robotic systems?

5: What training methods would stakeholders prefer for autonomous robotic system familiarization?





METHODOLOGY

Design

- Structured Interview w/ Open-ended questions
 - Express preferences in detailed, comprehensive format
 - Elicit unbiased responses



- Google Meets
- Qualtrics
- SPSS

Participants

- *N* = 5 Individuals from various job occupations
 - Dynetics, Naval Postgraduate School, U.S. Navy, and Acomb Ostendorf & Associates (AOA)



- 2020).



Materials

• Private laptop/tablet/ or personal device

Measures

• Interview Questionnaire Structured, open-ended questions • Inductive Content Analysis (Kyngäs et al.,

• Prevalent themes in responses



PROCEDURE



Consent

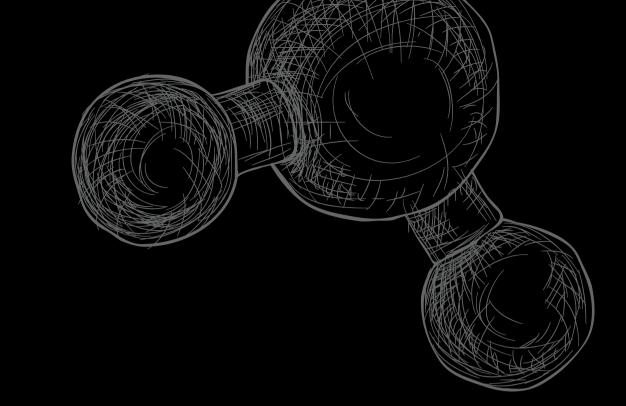
~ 5 minutes

Interview Questions

~ 20 - 30 minutes

Demographic Questions

~ 10 minutes









Debriefing

~ 5 minutes

Release of Participants

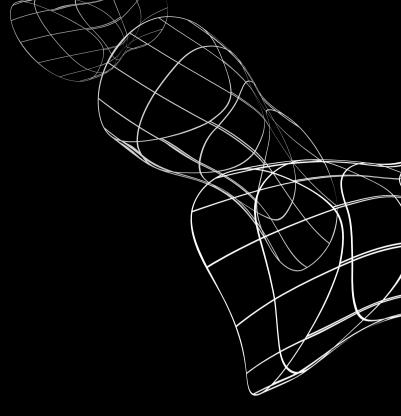
~ 50 minutes total time

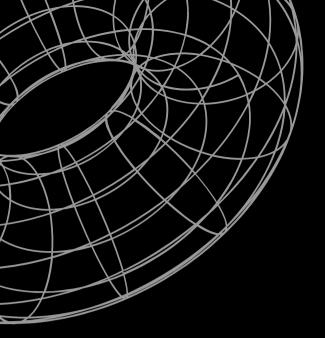


EXAMPLE QUESTION

Would you want an autonomous robot to look more animallike (e.g., quadrupedal), human-like (i.e., humanoid), or machine-like (e.g., autonomous mobile robot (AMR))?





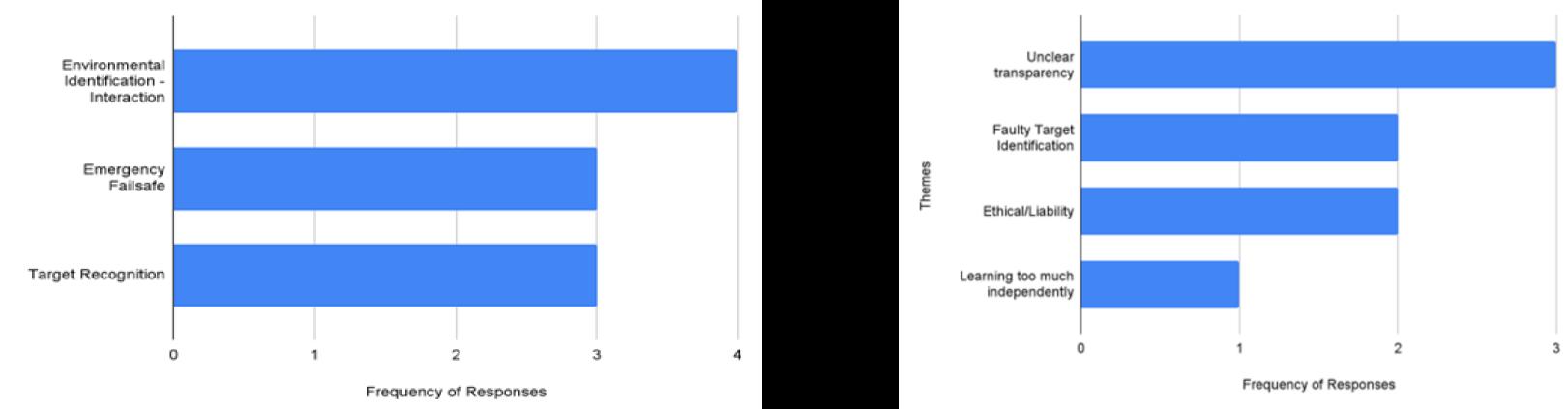


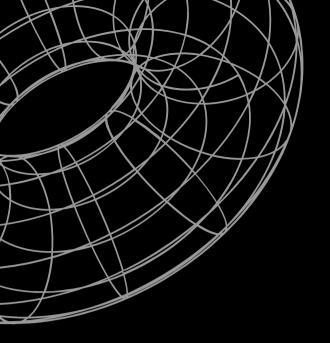


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RESULTS

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RESULTS

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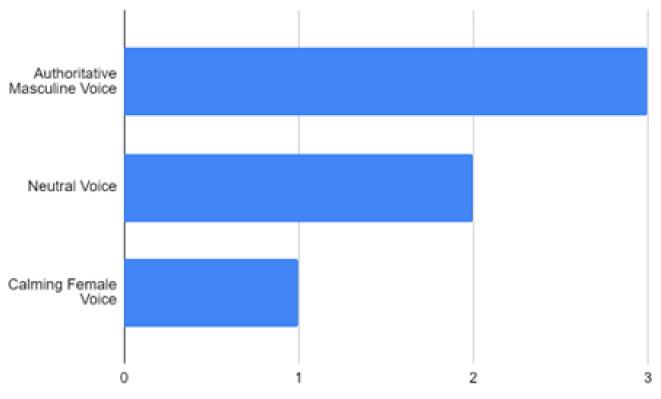
Themes

Simulation Training

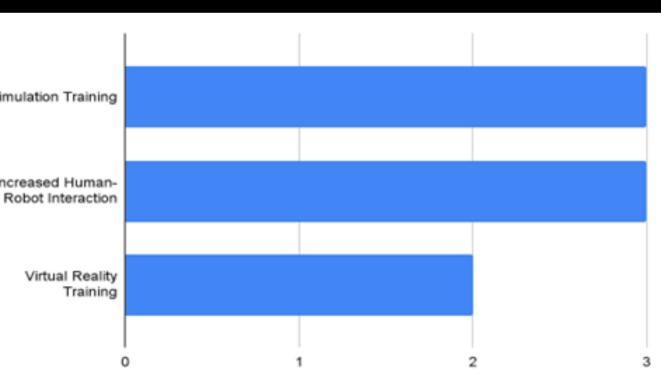
Increased Human-

Themes

Quadrupedal Efficiency Preferred Themes Machinelike Humanoid is uncomfortable 3 2 0 Frequency of Responses



Frequency of Responses



Frequency of Responses

WHY IT MATTERS

Robots in the Workplace

- Growing interest in Industrial & Nonindustrial robots
 - Industrial Robots perform strenuous tasks to relieve humans
 - Nonindustrial Robots assistive (e.g. lifting aid) or non-assistive
 - (e.g., entertainment/companion)
- Robots in professional fields
 - Trust in automated decisions

(Busse et al., 2021)

A Future in Technology

- Healthcare Robots

 Growing demand for caregivers
 - Increased Elderly autonomy with robotic assistance
- Factory/Warehouse Robots

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- Increased efficiency
- Reduced human injury

(Busse et al., 2021)



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QUESTIONS?



INCLUDE | 2022

THANK YOU FOR WATCHING

