

Estimating Attentional States of Older Adults for Remote Caregiving

Thomas M. Kwok¹, Yuanze Huang², Clara Colombatto², Yue Hu¹

¹ Department of Mechanical and Mechatronics Engineering, University of Waterloo, ON, Canada
² School of Psychology, University of Waterloo, ON, Canada

- Remote caregiving robots can support older adults aging in place, but **caregivers often have limited awareness of users' attentional states**.
- **Challenge: Existing attentiveness cues** are primarily derived from **younger adults** and may not generalize well due to **age-related changes** in facial appearance and behavior.
- **Finding: Eye blinks and lower-face movements**, particularly lip parting and chin raising, show promise as **facial indicators of attentiveness in older adults**.

Introduction

- Remote caregiving robots can help older adults age safely and independently at home.
- Caregivers often have limited awareness of older adults' attentiveness due to the limited camera view available in teleoperation systems
- **Main challenge: Existing attentiveness cues have been identified using younger adults.**
- Aging changes facial appearance and behavior, which may reduce the reliability of conventional attentiveness indicators.
- We investigate **which facial cues are associated with attentiveness in older adults** to support future remote caregiving systems.

Which facial cues can be used to estimate older adults' attentiveness for remote caregiving?

Methodology

- **Participants:** Conducted a pilot study with 4 older adults (69.5 ± 5.1 years, 2 males & 2 females)
- **Experiment:** Participants performed a Sustained Attention to Response Task (SART), responding to all digits except the target digit "3".
- A hidden camera recorded participants' facial behavior during the task.
- **Data Processing:** OpenFace 2.0 was used to extract Action Units (AUs) from the video recordings.
- **Data Analysis:** Logistic & linear mixed-effects models were used to identify AUs associated with attentional states.

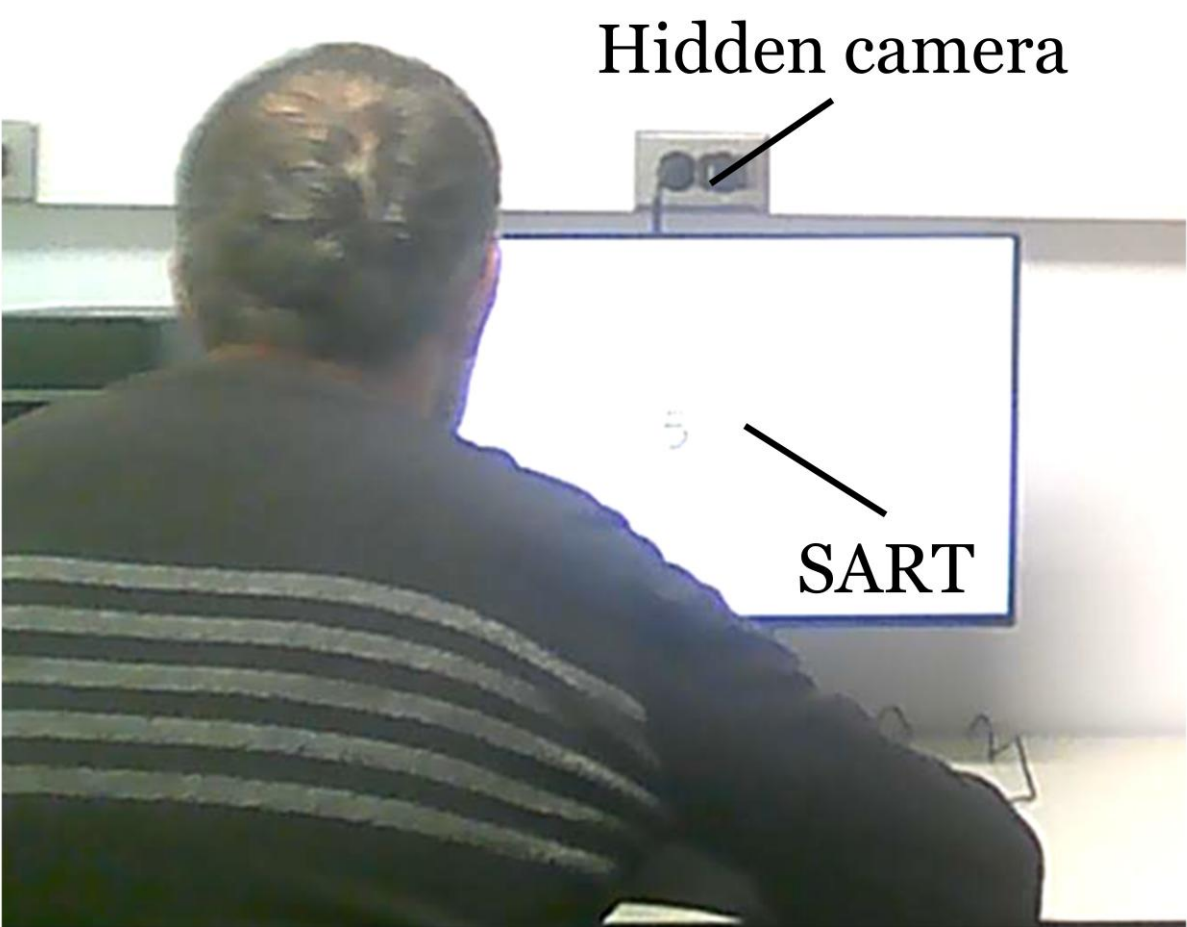
Results & Discussion

- **Results: Eye blinks (AU45), lip parting (AU25), and chin raising (AU17) showed strong and consistent associations with attentiveness.**
- **Potential Impact: Attentiveness estimation could support more effective remote caregiving.**
- **Future Work:** Larger participant cohorts, integration of gaze and head movement features, and real-time deployment in remote caregiving robots.

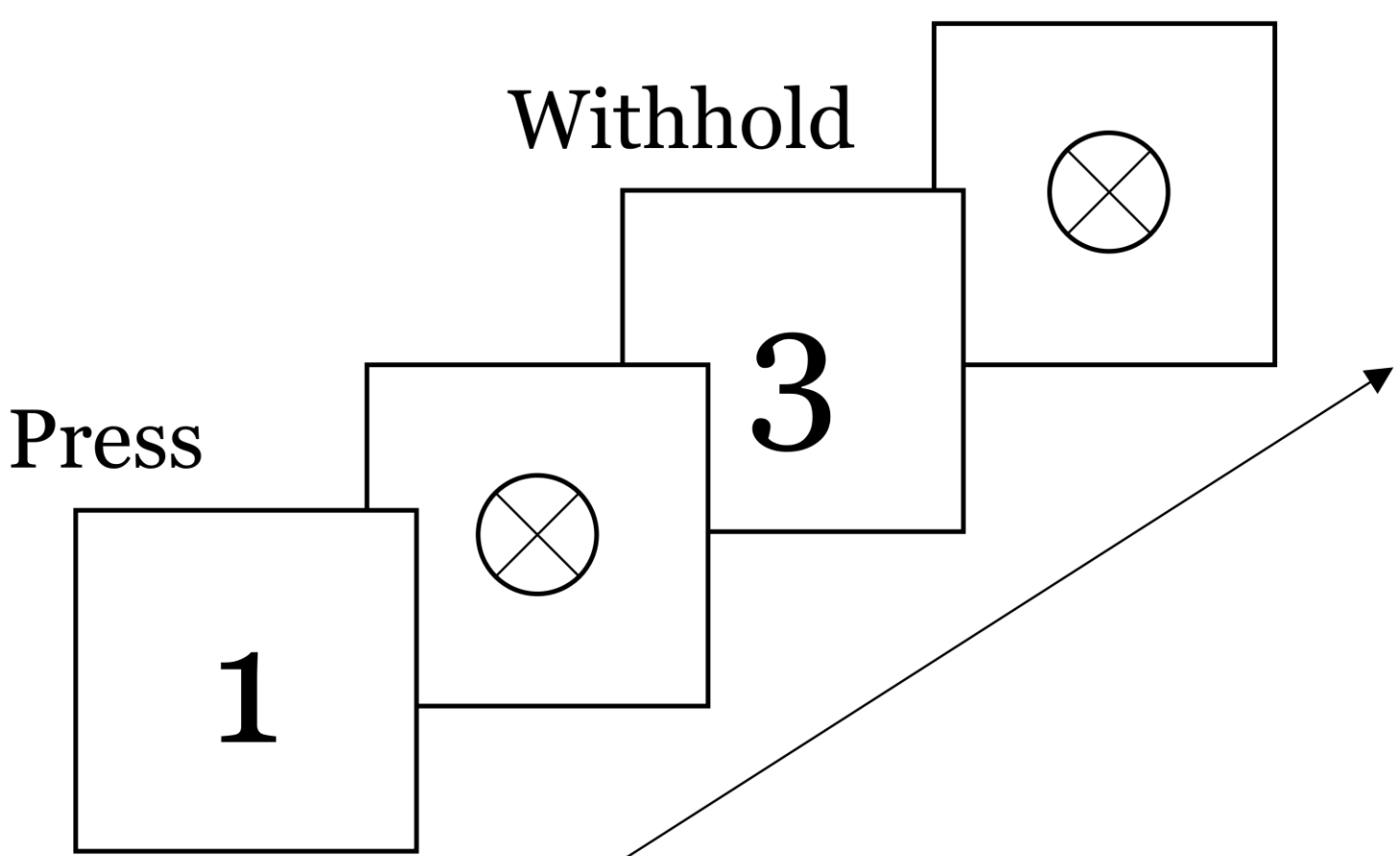
SIGNIFICANT ACTION UNITS ASSOCIATED WITH ATTENTIONAL STATE.

AU*	Estimate	Statistic	p-value#
AU12 _c (Lip Corner Puller)	4.83	OR = 125.28	< 0.001
AU25 _c (Lips Part)	4.44	OR = 84.66	< 0.001
AU45 _c (Blink)	4.09	OR = 59.88	< 0.001
AU17 _c (Chin Raiser)	1.29	OR = 3.64	< 0.01
AU14 _c (Dimpler)	-7.34	OR = 0.001	< 0.001
AU45 _r (Blink)	0.07	<i>t</i> = 4.98	< 0.001
AU26 _r (Jaw Drop)	0.09	<i>t</i> = 4.32	< 0.001
AU25 _r (Lips Part)	0.06	<i>t</i> = 3.81	< 0.01
AU15 _r (Lip Corner Depressor)	0.02	<i>t</i> = 3.63	< 0.01
AU01 _r (Inner Brow Raiser)	0.04	<i>t</i> = 3.50	< 0.01
AU17 _r (Chin Raiser)	0.05	<i>t</i> = 3.26	< 0.01
AU09 _r (Nose wrinkler)	0.008	<i>t</i> = 2.84	< 0.05
AU20 _r (Lip Stretcher)	0.02	<i>t</i> = 2.60	< 0.05
AU23 _r (Lip Tightener)	0.02	<i>t</i> = 2.52	< 0.05

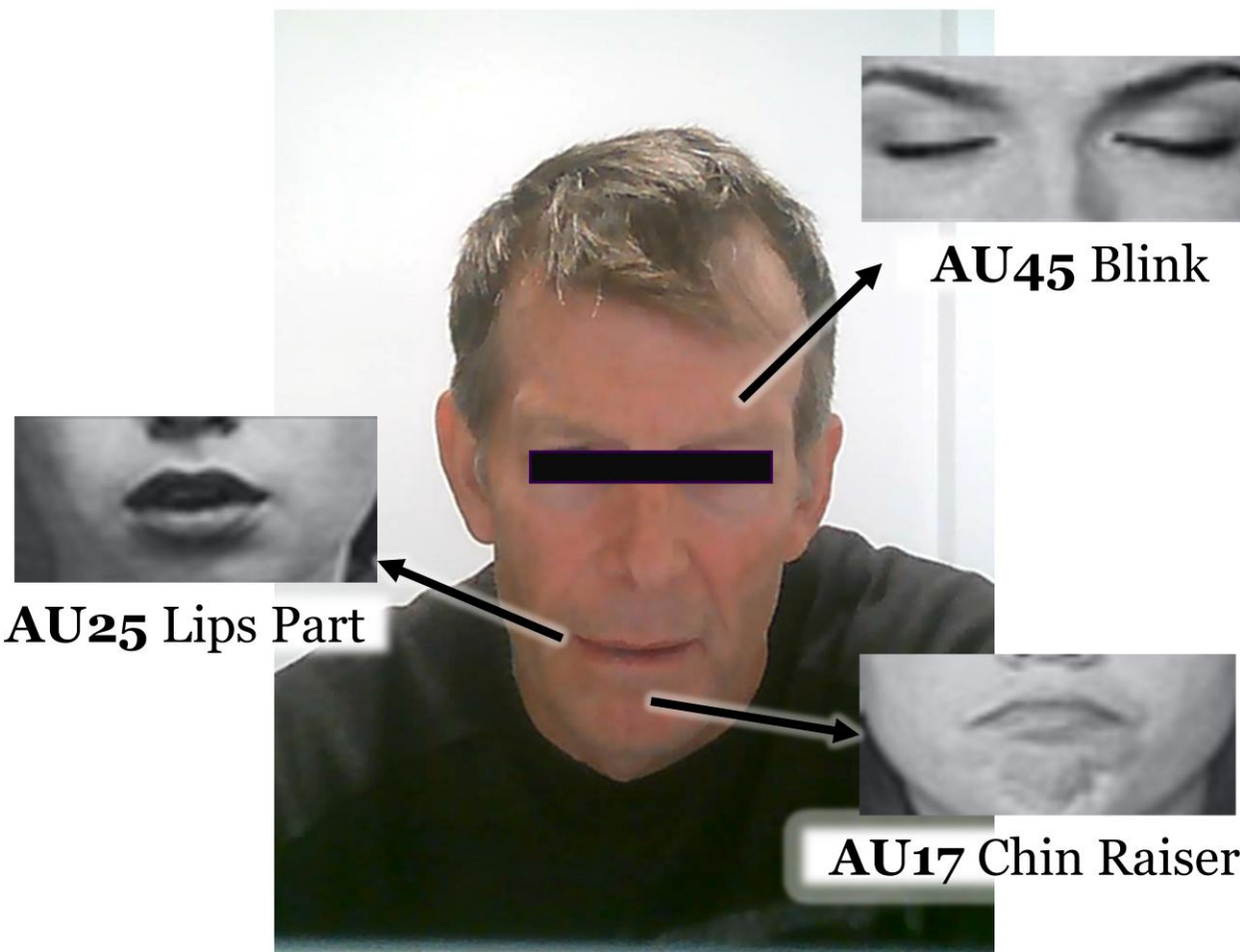
* AU_c denotes Action Unit occurrence, and AU_r denotes Action Unit intensity, analyzed with logistic and linear mixed models, respectively.
p-values were adjusted using false discovery rate (FDR).



Experiment setup



Sustained Attention to Response Task (SART)



Video frame of hidden camera

Feel free to connect with **Thomas** on LinkedIn!



UNIVERSITY OF
WATERLOO

