

Precision Triage in Healthy Aging:

Uniting AI, Clinical Science, and Community
Care to Pre-empt the "Silent Danger" of Frailty

PARTNERS:

**CAPTUREPROOF, NUHS, NTUC HEALTH,
CHISEL & TEMASEK**

TEMPLE UNIVERSITY & ALS HOPE FOUNDATION



MEGHAN CONROY, CEO CAPTUREPROOF
VALENCIA MARCH 10, 2026



20:47

5G 66

00:01,71

Lap

Stop

Lap 1

00:01,72

World Clock

Alarms

Stopwatch

Timers

Disclosure

This work is supported by the Centre for Healthcare Innovation Start-up Enterprise Link (CHISEL) funding. Organised by the Centre of Healthcare Innovation (CHI), presented by Temasek Foundation, and supported by the Ministry of Health Singapore (MOH), CHI Start-Up Enterprise Link (CHISEL) is a cross-cluster led platform that convenes the Singapore public healthcare system to accelerate the adoption of market solutions from start-ups and SMEs.

Specifically Dr. Reshma Merchant at the National University Health System and Dr. Lily Yeo, the Head of Active Ageing Centers for NTUC Health in Singapore;

Dr. Terry Heiman-Patterson, an ALS expert at Temple University, and the ALS Hope Foundation for their partnership on our neurodegenerative tracking



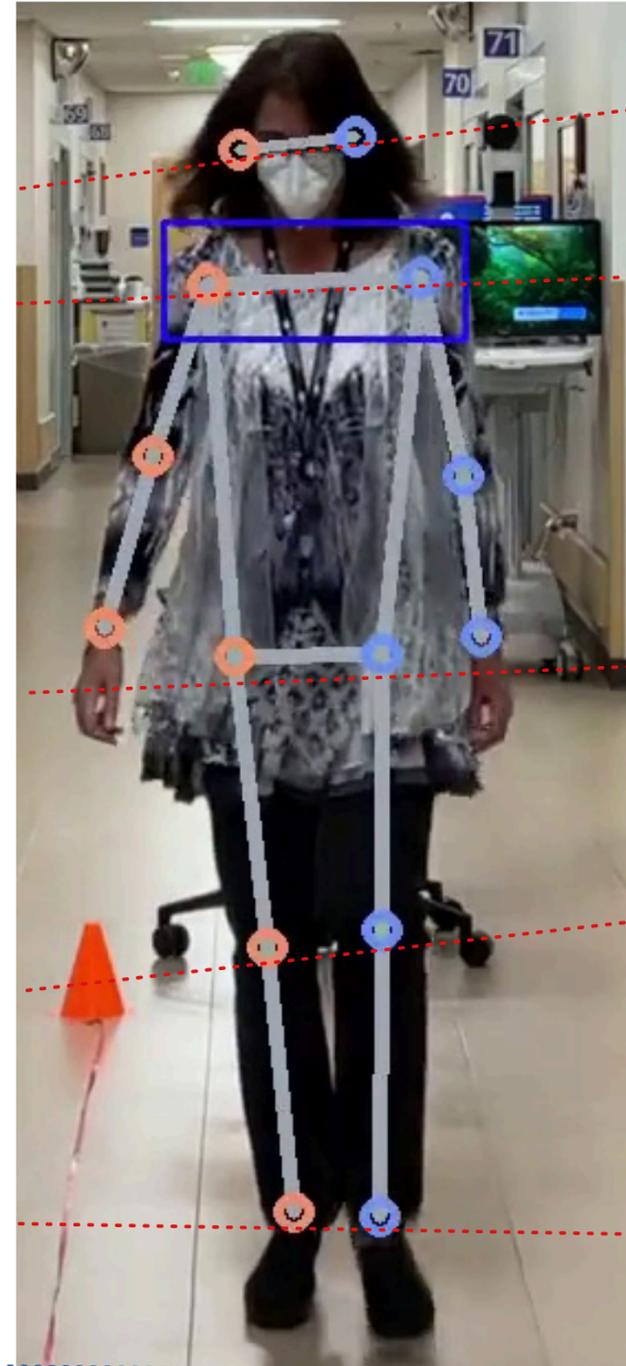
2050: 30% over 65



CAPTURE PROOF



TURNING VISUALS INTO VITAL SIGNS



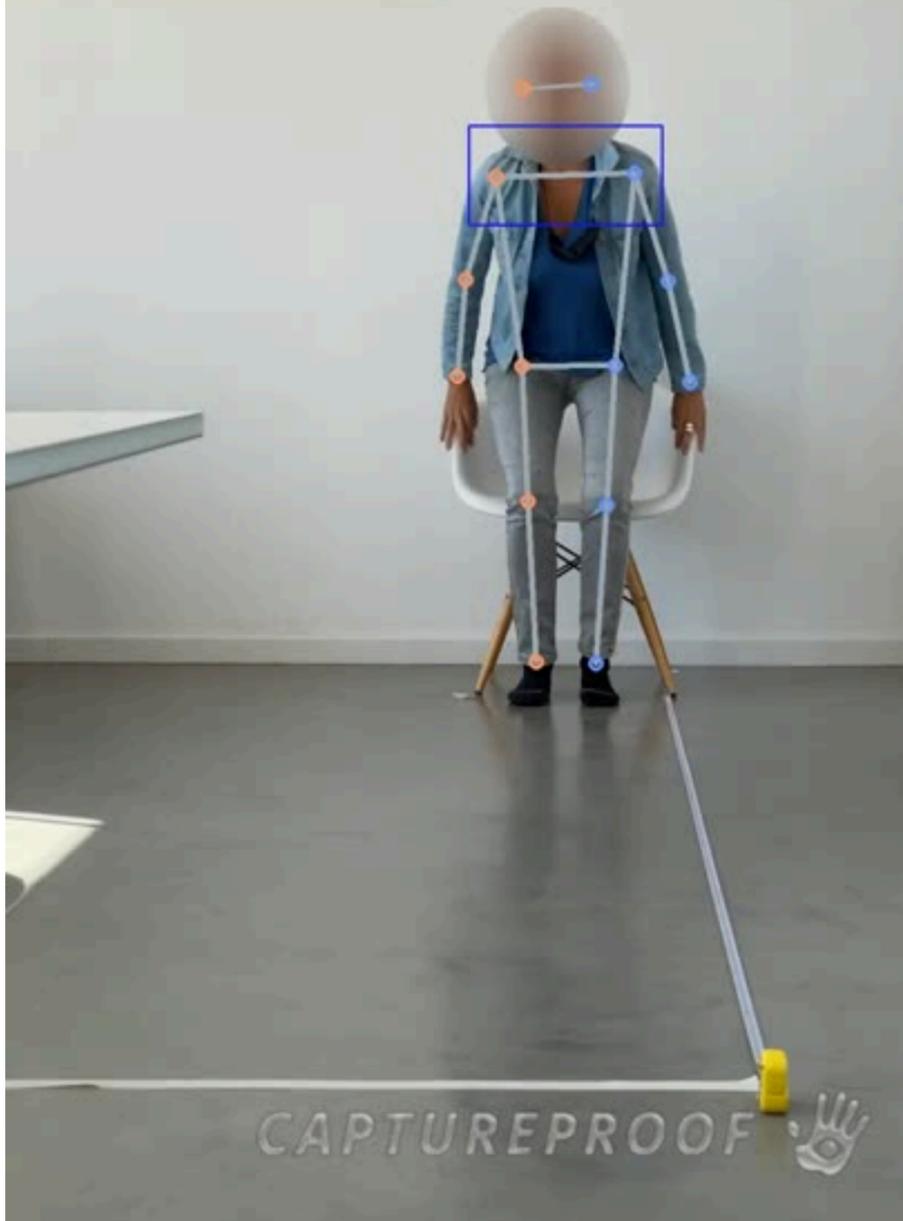
Head

Shoulders

Hips

Knees

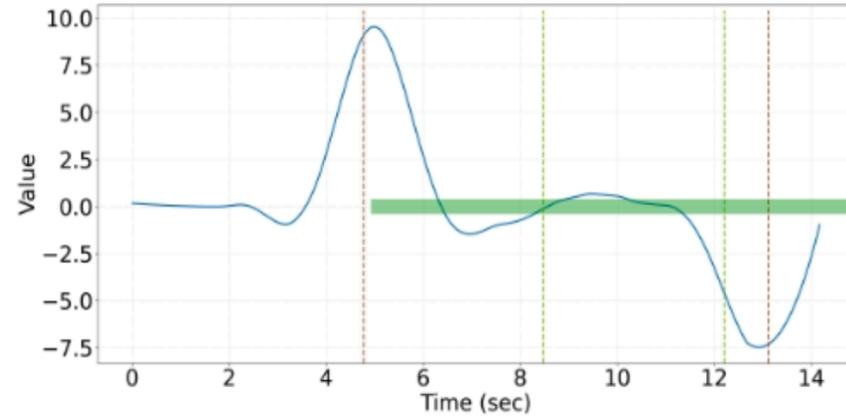
Ankles



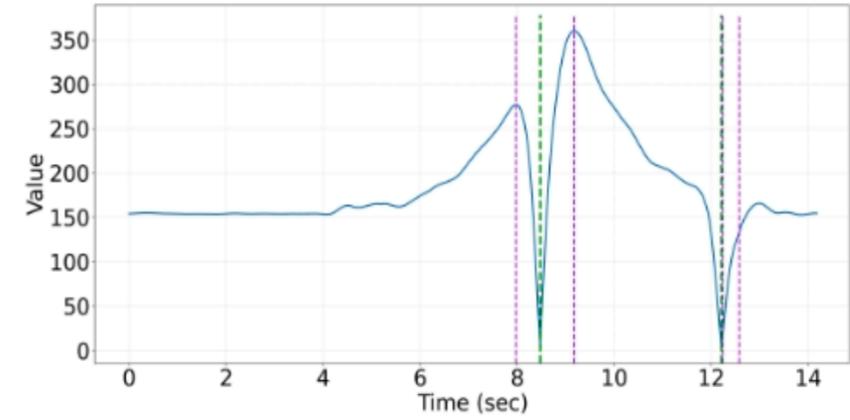
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Total Time 8.3 seconds

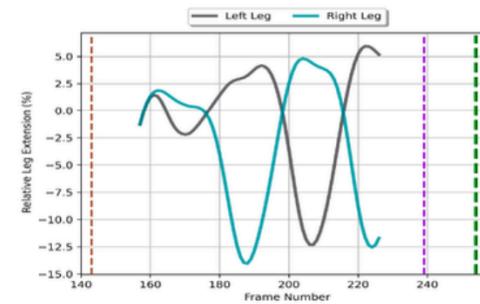
Peaks (height,width) Rise (9.5, 1.6 sec) Sit (7.5, 1.8 sec)



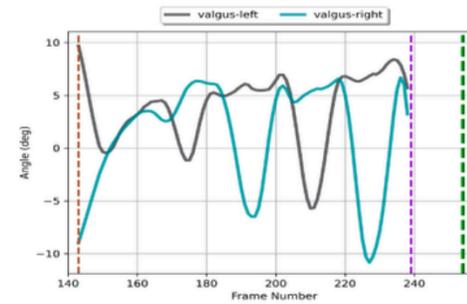
Turn Times 1.2, 0.3 seconds



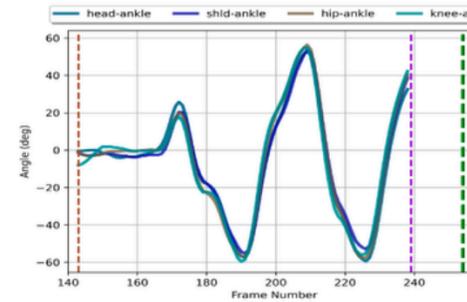
Balance Scan: Cadence and Listing - 1



			Left	Right	Stride trailing %
1	R	0.74	22	19	75
2	L	0.59	18	18	67
3	R	0.57	17	13	***
avrg	L	0.59	18	18	67
avrg	R	0.65	20	16	75
avrg	L&R	0.63	19	17	71

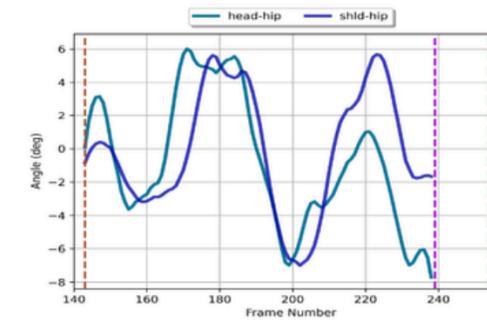


Curve	Average	Max	Min	Std Dev	Median
valgus-left	3.82	9.69	-5.72	3.39	4.93
valgus-right	1.26	6.67	-10.85	4.99	3.10

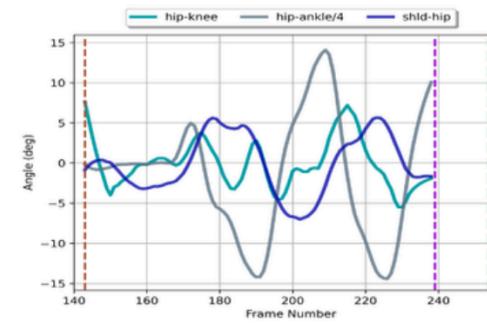


Curve	Average	Max	Min	Std Dev	Median
head-ankle	-7.28	52.95	-59.40	28.92	-2.57
shld-ankle	-6.65	53.54	-55.28	27.53	-2.86
hip-ankle	-6.57	56.31	-57.50	29.83	-1.18
knee-ankle	-6.48	54.89	-59.61	29.55	-0.41

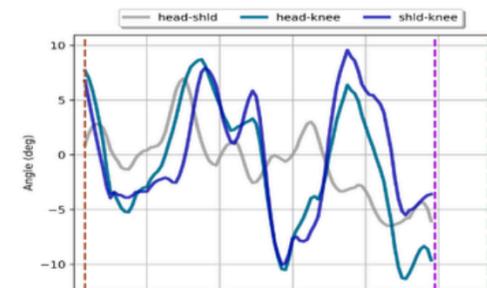
Balance Scan: Listing - 2



Curve	Average	Max	Min	Std Dev	Median
head-hip	-0.71	6.01	-6.99	3.85	-1.34
shld-hip	-0.07	5.67	-7.00	3.62	-0.32



Curve	Average	Max	Min	Std Dev	Median
hip-knee	-0.09	7.62	-5.50	2.99	-0.29
hip-ankle	-1.64	14.08	-14.38	7.46	-0.30
shld-hip	-0.07	5.67	-7.00	3.62	-0.32



Curve	Average	Max	Min	Std Dev	Median
head-shld	-0.63	6.96	-6.47	3.15	-0.27
head-knee	-0.80	8.71	-11.32	5.63	-1.00
shld-knee	-0.16	9.56	-10.02	5.21	-1.35

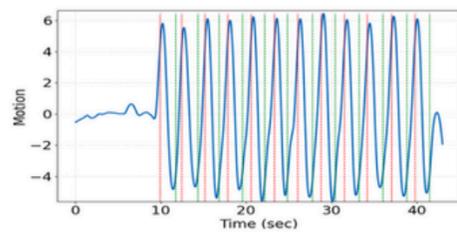
CaptureProof's cutting-edge AI technology is transforming how we assess patients progress.

Balance Scan: Sit and Stand

Fall Risk Calculation: At Risk

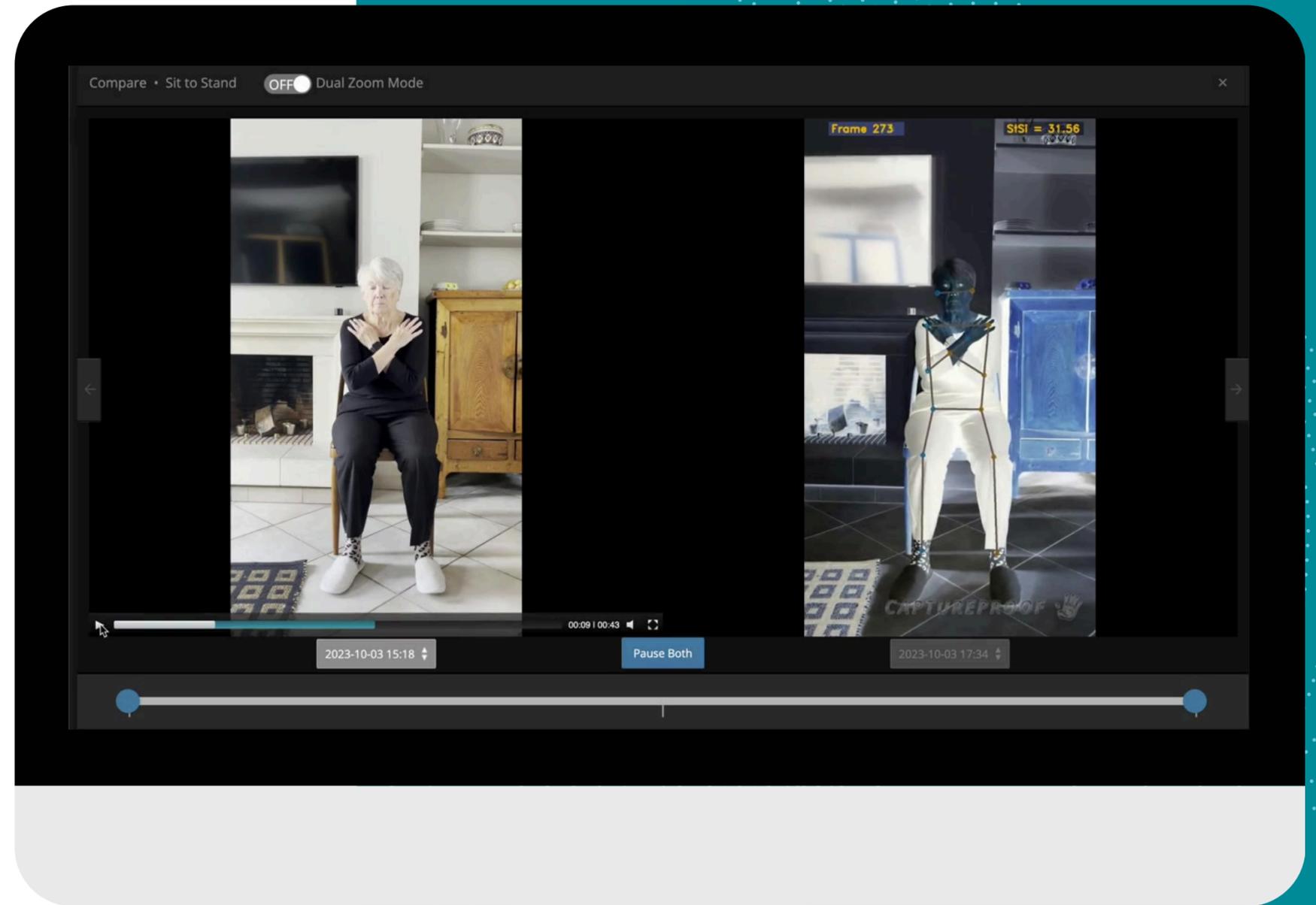
Total Time for 5 Reps 12.24 sec, Number of Full Repetitions 11 in 30 seconds

CDC GUIDELINES: An adult over 65 who takes >12 seconds to complete the Sit-to-Stand Scan is at risk for falling.



metric	value/Avrg sec	Min sec	Max sec
Total Stand to Sit	31.56	12 Reps	
Rise to Sit	1.73	1.50	1.93
Sit to Rise	0.99	0.67	1.23

metric	Average	Min	Max
Rise peak height	6.05	5.55	6.44
Sit peak height	5.16	4.69	5.63



Shared with permission from the patient



In Singapore

falls account for

40%

OF INJURY-RELATED DEATHS

about **ONE-THIRD**
OF OLDER ADULTS



aged 60 and above have
fallen more than once

30% will experience
recurring falls

A fall that is not addressed may lead to
RECURRENT FALLS





Study Design

Deployment Validation Study

Target Population:

200 older adults from Active Aging Centers (AACs).

Eligibility Criteria:

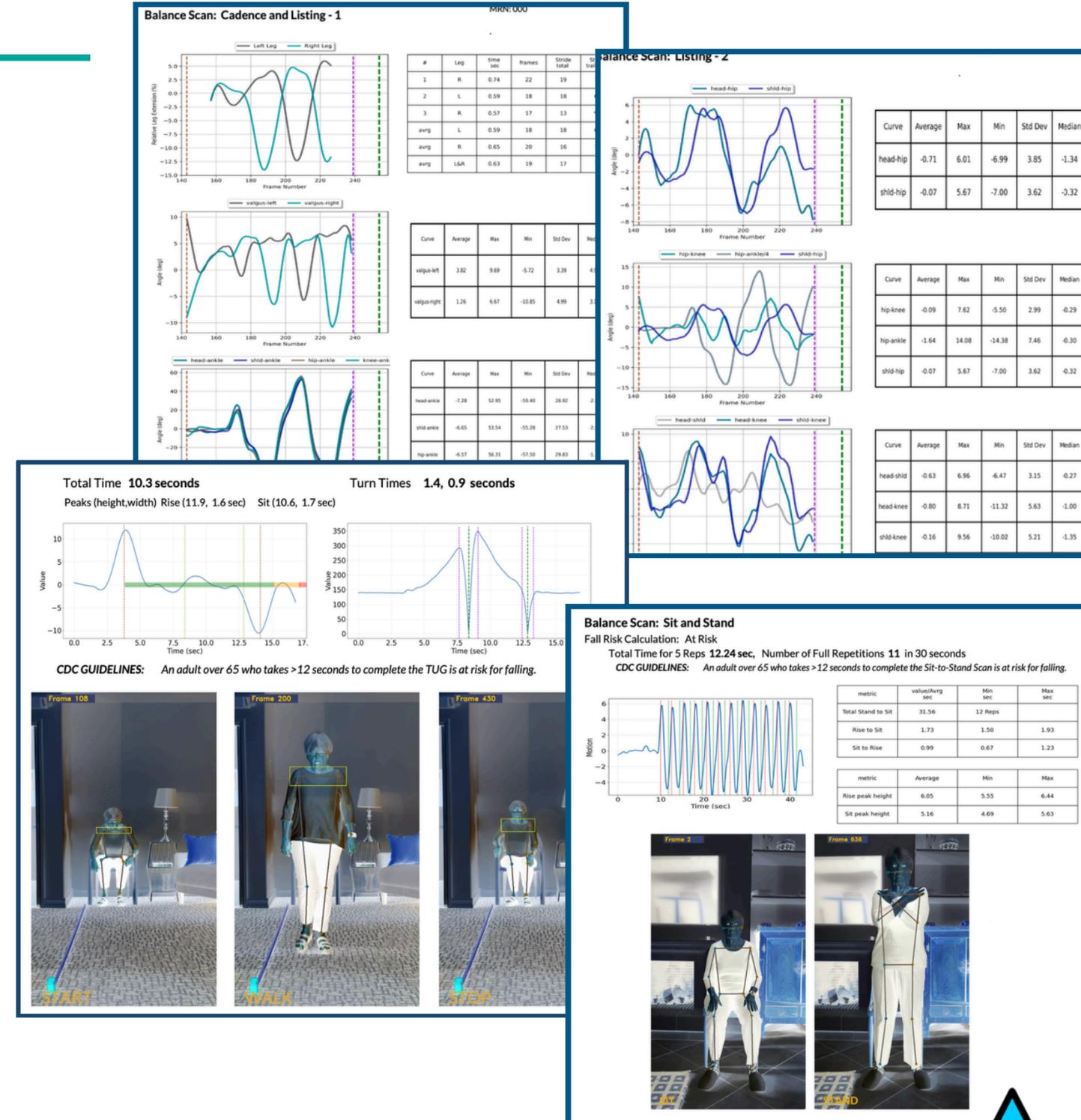
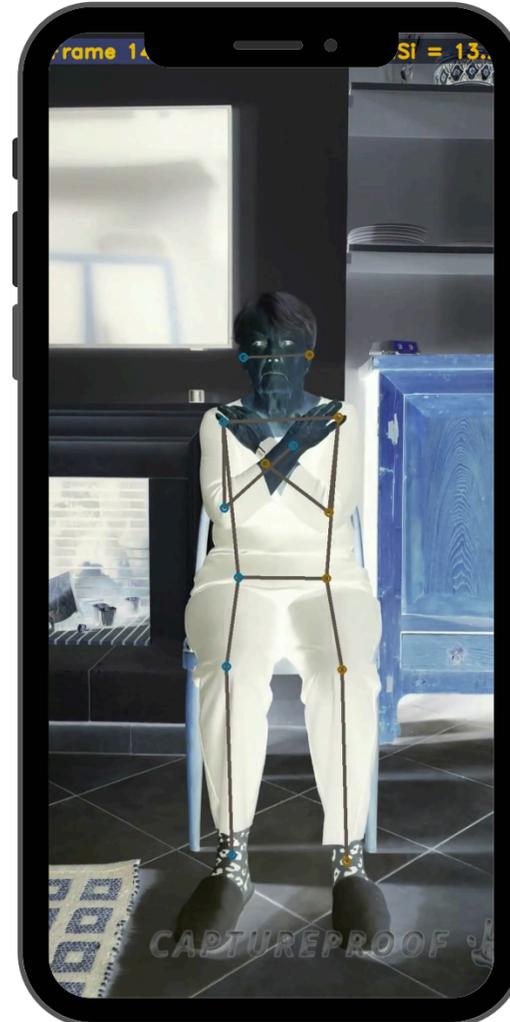
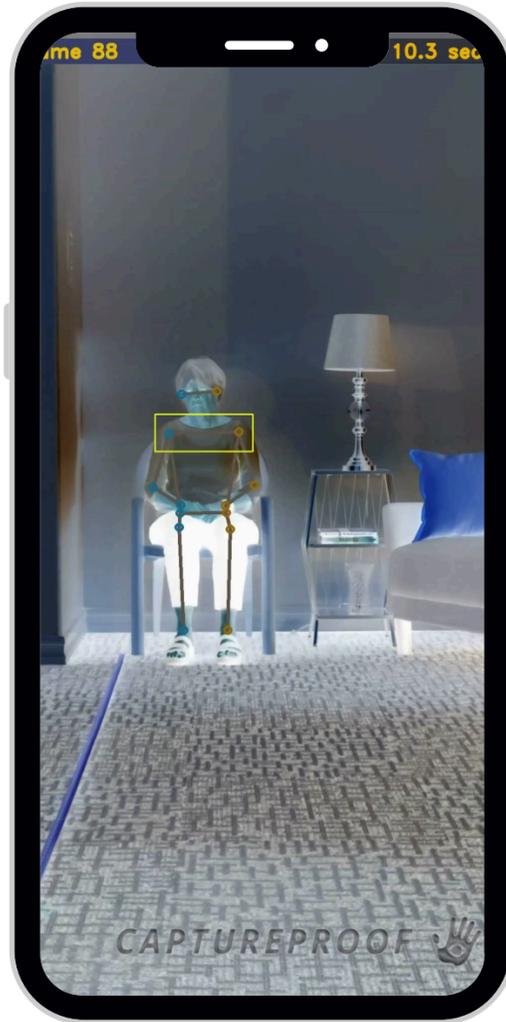
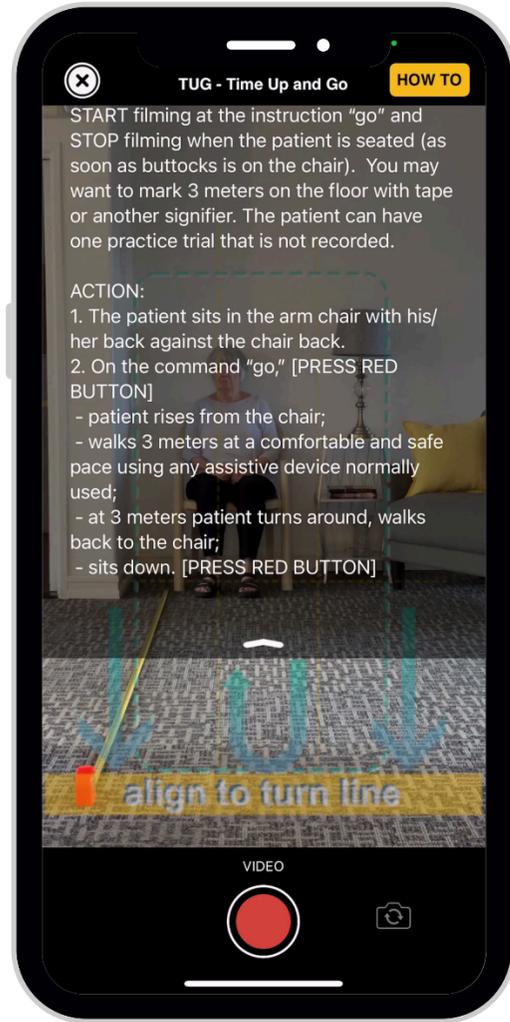
Age ≥ 60 years, ambulatory, able to consent.

Exclusion Criteria:

Unable to walk 10ft, comprehend commands, or participate in mobility assessments.



Deployed Technology



Phenotyping:
 Psychological vs. Medical vs. Mechanical Risk



Enrollment

201 enrolled in initial cohort
July 2025 - February 2026

201 completed 100% of study baseline
No adverse events (serious or otherwise)

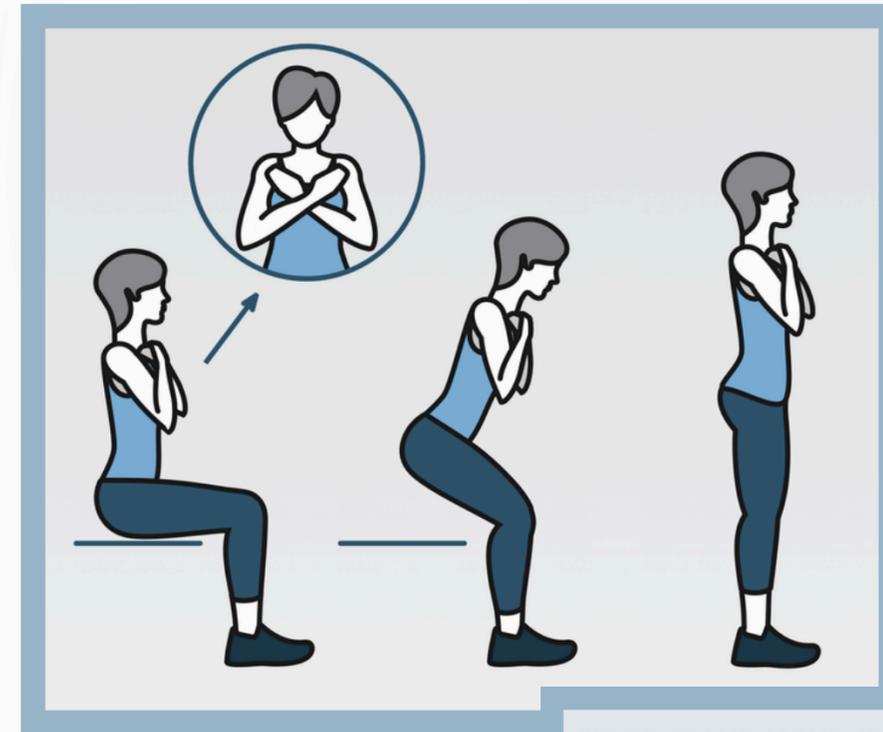
2 No Shows

Withdrawals- 3 cases, the seniors did not want to register with AAC.

Follow up is underway

- 3 month question
- 6 month rescan with SPPB (first 201)

Expansion additional 600 with only CP scan



SIT-TO-STAND MOVEMENT 

What We Expect From You

Sit Still for 3 Seconds
Sit Still for 3 Seconds: Start by sitting comfortably in a chair (one with a backrest is perfect). Keep your back nice and cozy against the chair, and just relax there for 3 seconds. Think of it like a little "ready, set, go" moment!

Stand Up
When you hear "Go" (or if you're firming yourself, just say it in your head), stand up from the chair. You can use your hands on the armrests if that feels better—whatever works for you!

Sit Back Down
Once you're standing, lower yourself back into the chair nice and easy.

Sit Still Again for 3 Seconds
After you're seated, just chill there for another 3 seconds. That's it—you're done!

Why You're Doing This:

This little Sit-to-Stand move is like a superpower test for your legs! It shows us how strong they are and how well they help you get up and down—like when you're popping up to grab a snack or sitting down to read a good book. For us older folks (60 is the new 40, right?), keeping that strength is key to staying independent and zipping around the house. The 3-second pauses at the start and end? Oh, those are CaptureProof's secret sauce! They help the AI (that's the smart computer part) figure out exactly when you're starting and stopping. It's like giving it a clear "hello" and "goodbye" so it can watch your movement super closely and tell us how awesome you're doing. Plus, it makes sure we're all on the same page—no rushing, just nice and steady!

Demographics

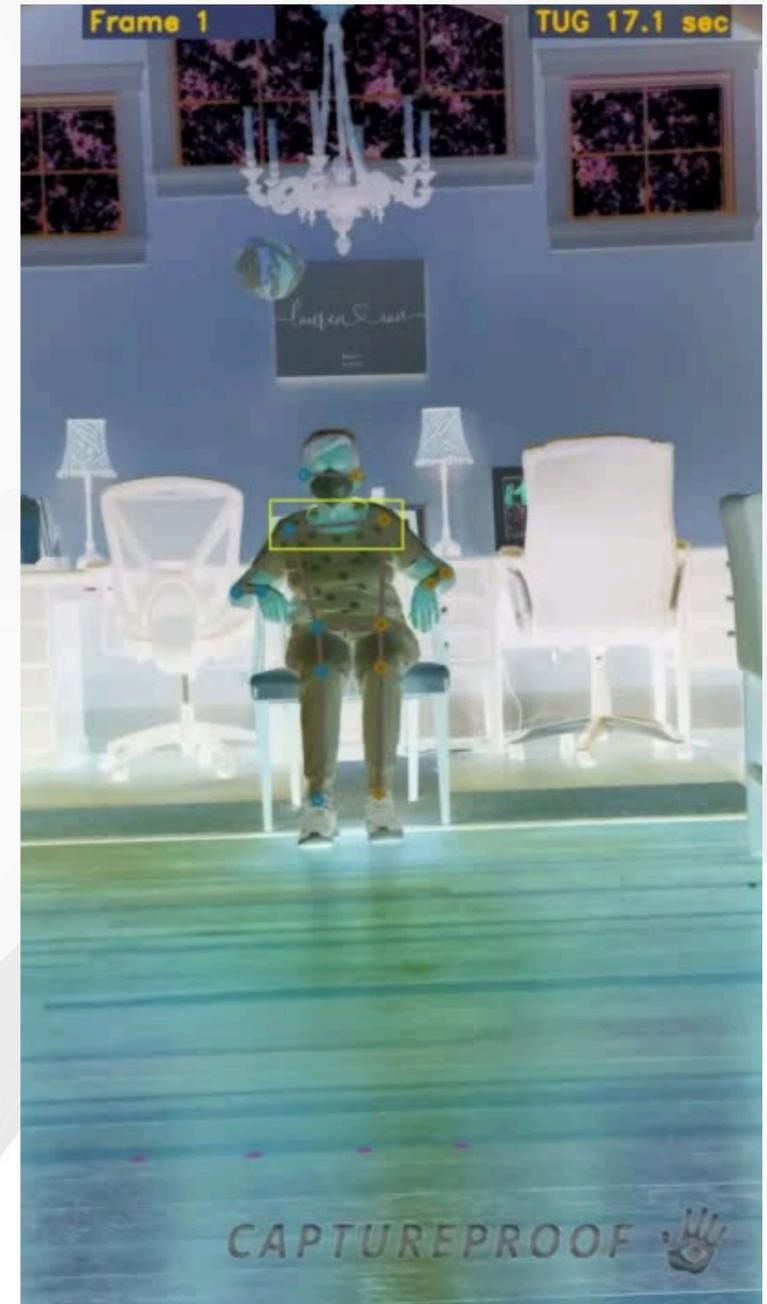
Age: 69.1 years (\pm 5.7).

The youngest participant is 60 and the oldest is 87.

Gender:

55% Female, 45% Male

Body Mass Index (BMI): 25.0 (\pm 4.6)



Methodology - Data Collection

CaptureProof Scan Assessments:

- Timed Up and Go (TUG): Stand, 3 meters and return
- Sit-to-Stand (StS): 30 seconds of motion
- Time, Sway Analysis, Stand and Sit Velocity, Gait Speed, Turning Speed, Cadence Symmetry.

Surveys:

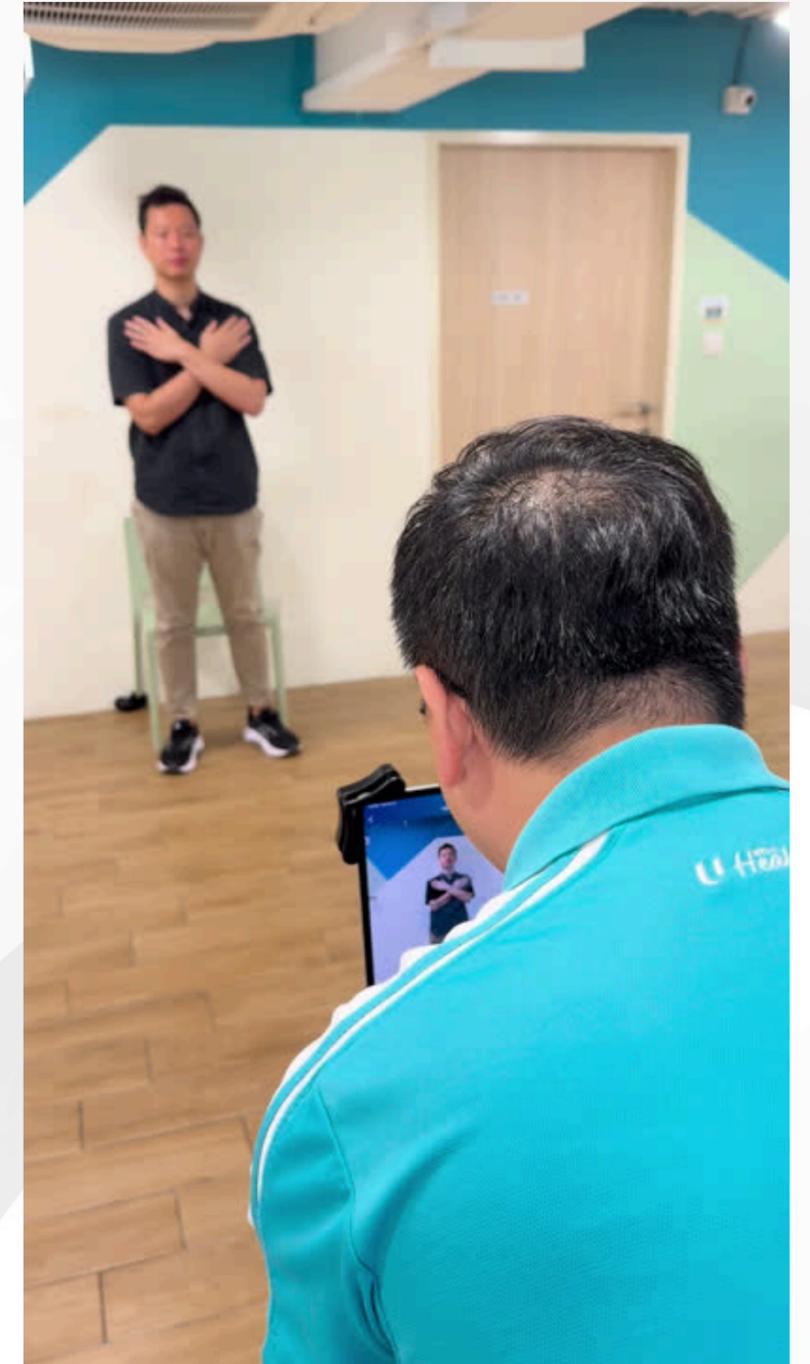
Demographics, STEADI, Cognition, FROP-Com, FRAIL, Fall Efficacy Scale (FES), EQ-5D-5L, NPS, PSSUQ, TFA

Physical Exams:

Muscle Strength, Short Physical Performance Battery (SPPB)

Fall Incidence:

Tracked 1 year before, 3 months after offering - 6 month “rescan & SPPB”



Scans Collected

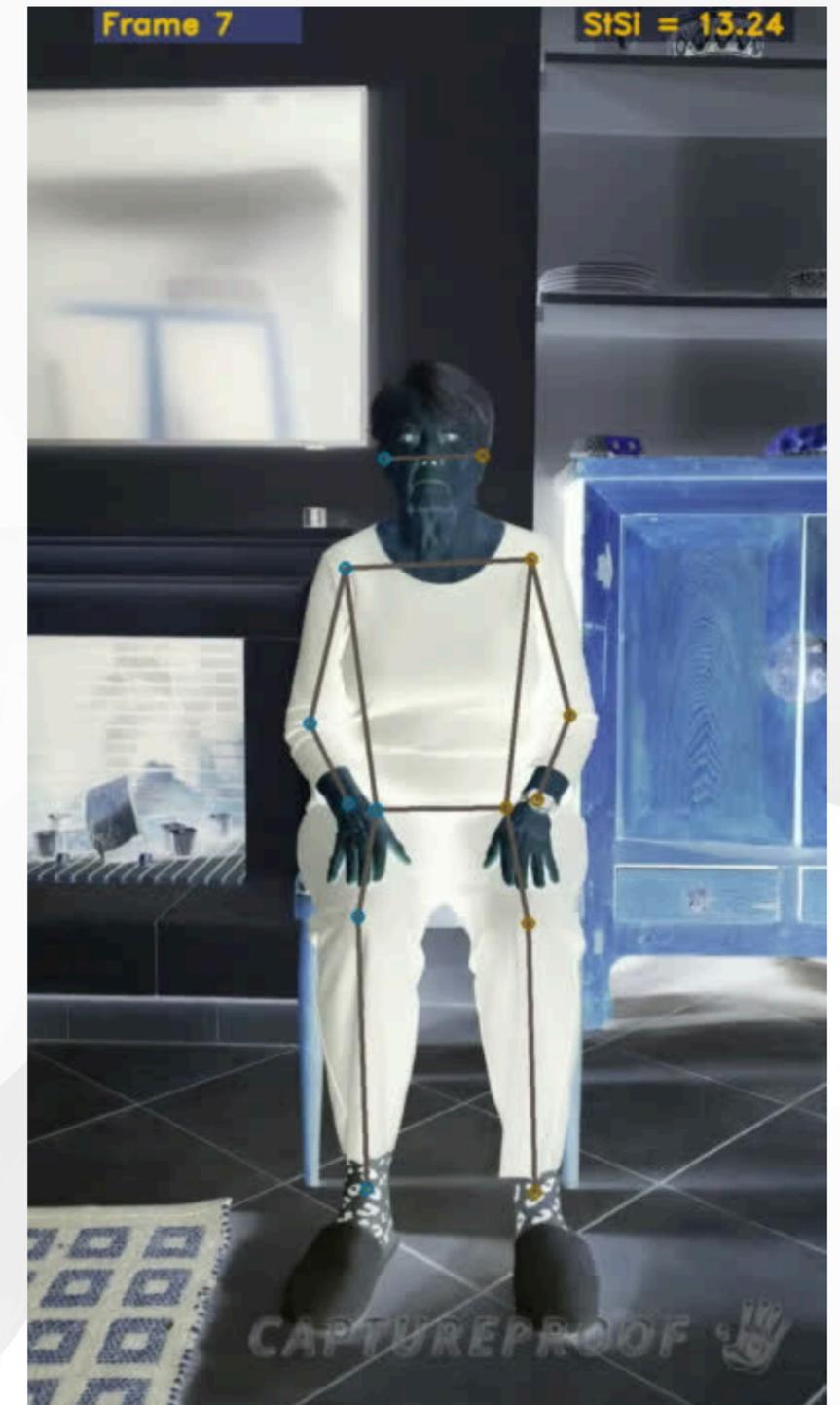
Participant 1-100

- 30 second sit to stand
- Timed up and go
- Scan time 2-3 minutes

Participant 101-201

- 30 second sit to stand
- Timed up and go
- Dual task Timed up and go (20-1 counting)
- Scan time 3-5 minutes

100% Baseline completed (STS + TUG),
99% with *Dual task TUG* completed
72% agree to the 6m follow-up scan (currently on going)



Timeline of Process

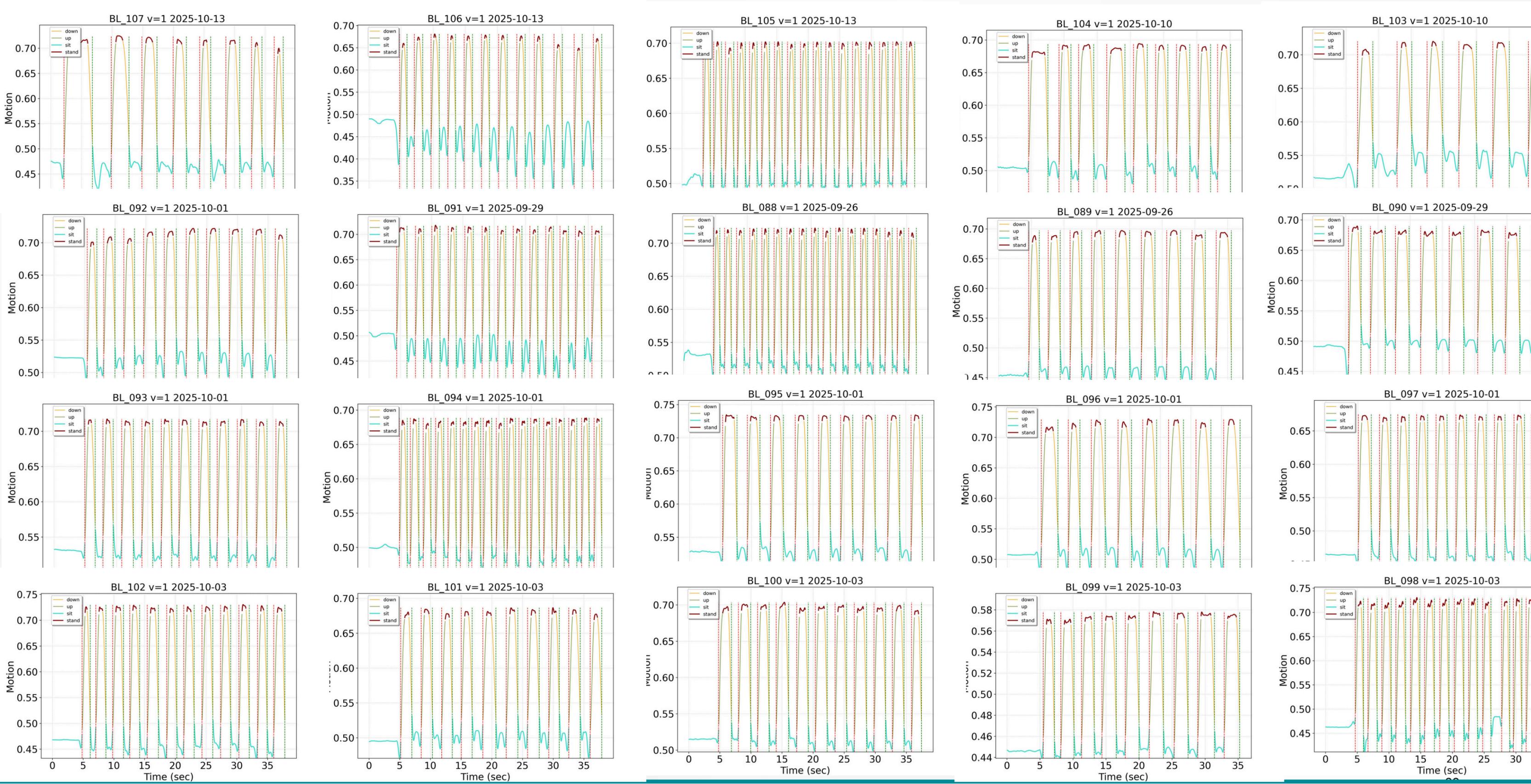
Consent
10 minutes

CP Scan alone
5-15 minutes

SPPB & Surveys
25-35 minutes

Total
~ 1 hour

SILVER WORKFORCE are independent, and first 2 successfully trained additional



Early Data - Sit to Stand

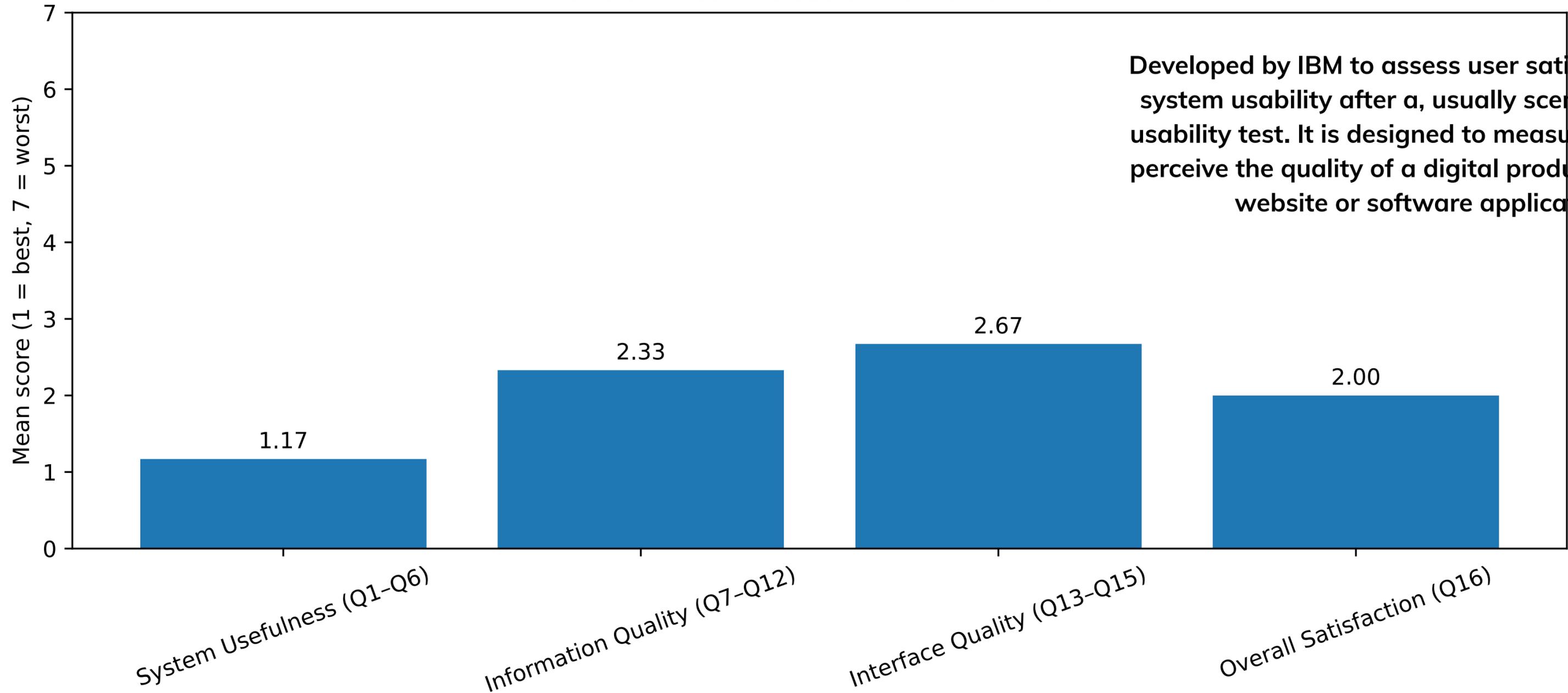
Group #	Group Name	Prevalence	SPPB	CP Rigidity Index
Group 1	True Healthy	40%	10.8	52.8
Group 2*	Unaware Rigid (Silent Danger 1)	17.5%	10.5*	1.5
Group 3	Worried Well	16.5%	10	46.7
Group 4*	Worried Rigid (Silent Danger 2)	10.5%	10*	2.5
Group 5	Accidental Faller	8%	10	59.5
Group 6*	Mechanical Faller	7.5%	9.4*	-2

Phenotypes

1. **True Healthy:** Fall-free, fluid biomechanics.
2. **Unaware Rigid (Silent Danger 1):** Fall-free, rigid biomechanics.
3. **Worried Well:** Safe biomechanics, Restricting activity due to fear of falling.
4. **Worried Rigid (Silent Danger 2):** Afraid of falling and exhibiting the physical stiffness.
5. **Accidental Faller:** Recent fall, healthy biomechanics.
6. **Mechanical Faller:** Recent fall, rigid biomechanics.

Usability summary (PSSUQ)

PSSUQ Subscale Mean Scores (Lower is Better)
Overall mean across all 16 items: 1.94



Developed by IBM to assess user satisfaction with system usability after a, usually scenario-based, usability test. It is designed to measure how users perceive the quality of a digital product, such as a website or software application.

Score

Broader Impact

To advance our combat-frailty initiative in the community, we are strategically expanding CaptureProof beyond its pilot site to create a powerful, integrated network across Active Ageing Centres.

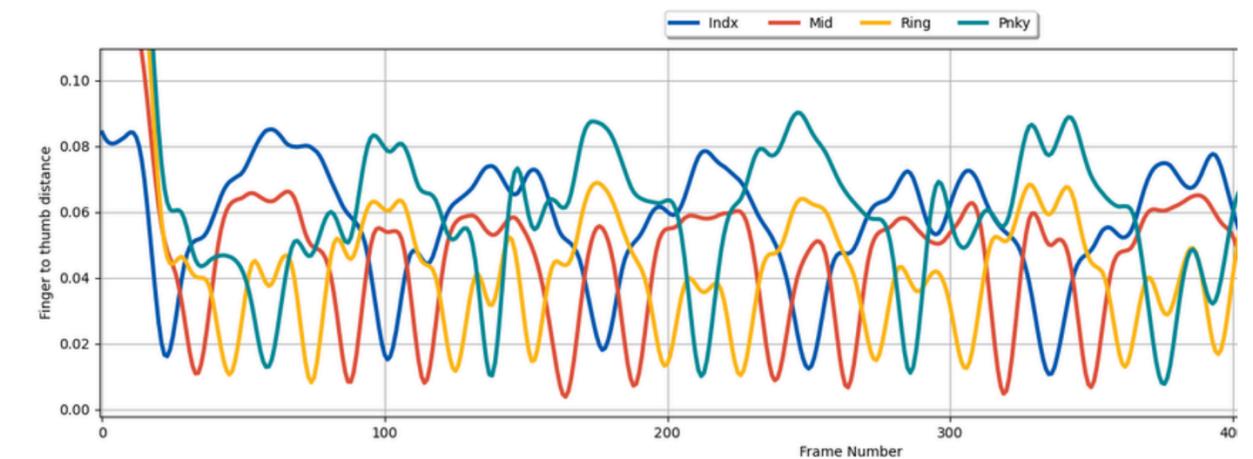
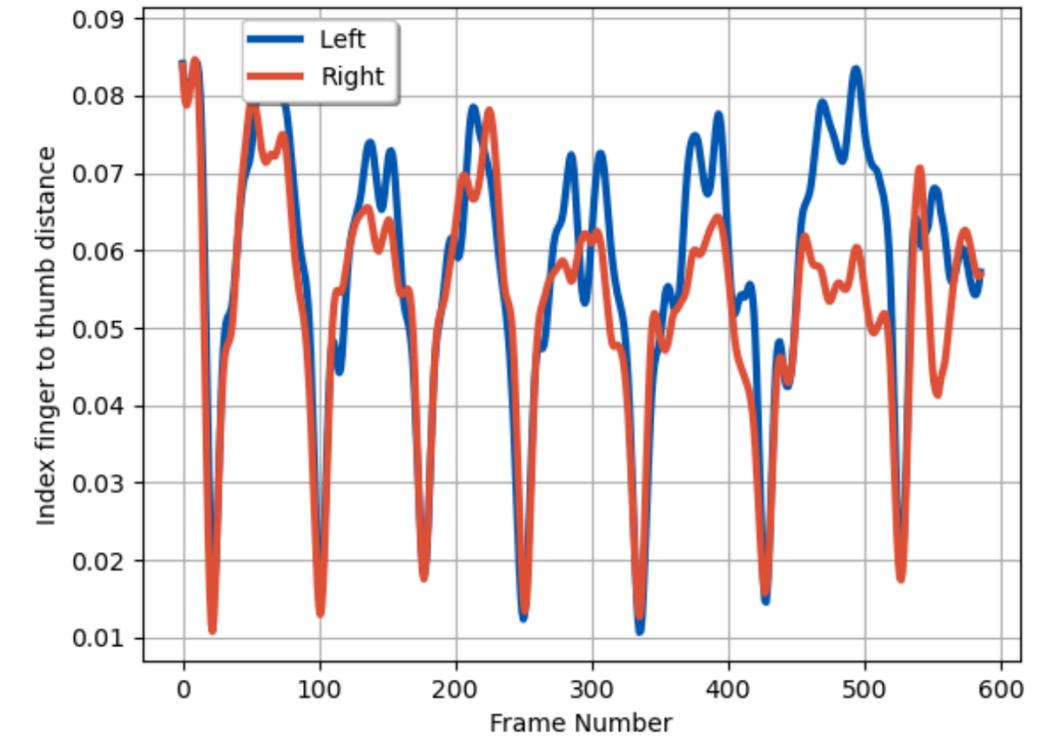
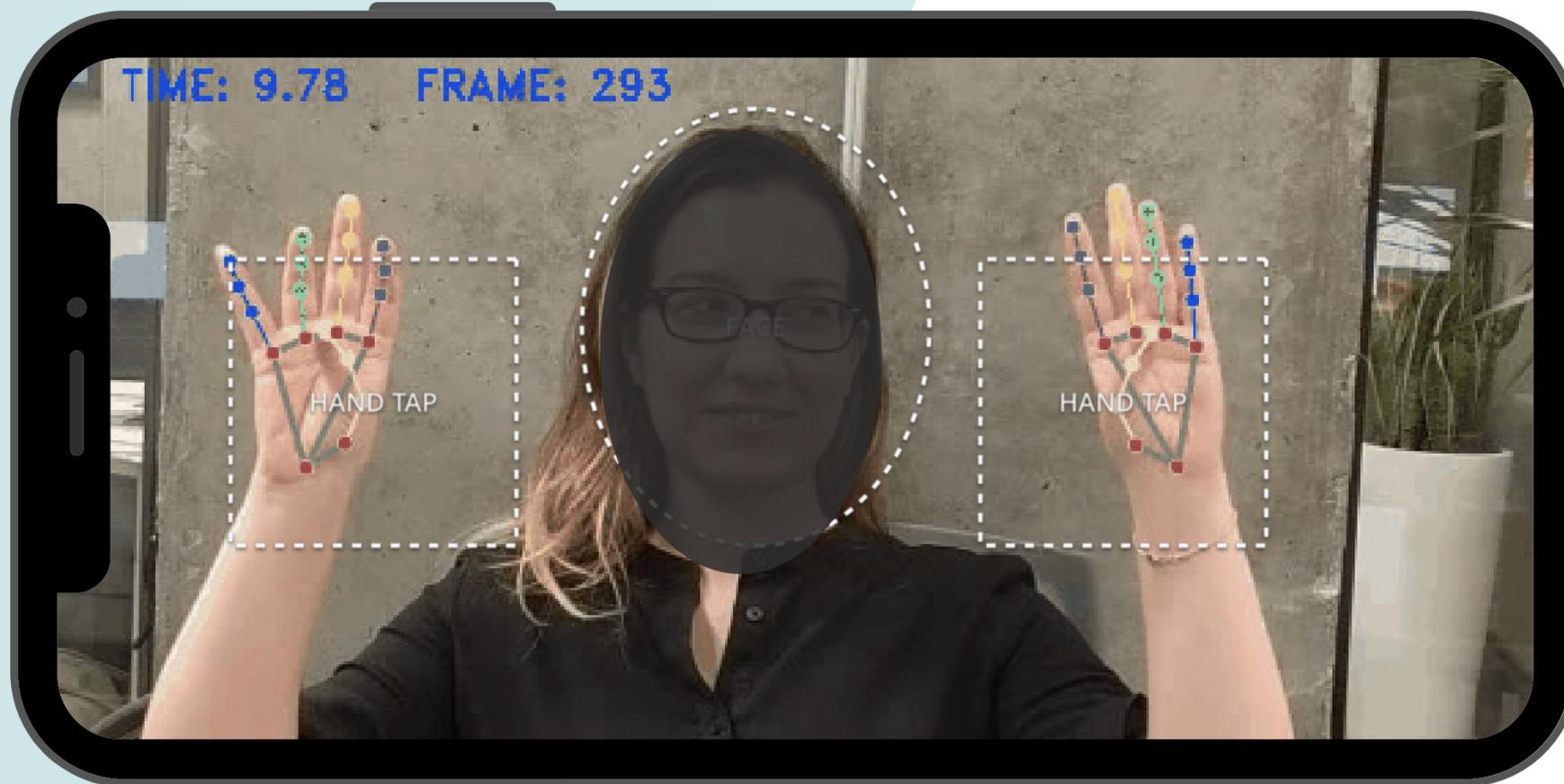
This expansion will maximize our reach to all community-dwelling older adults. The network includes the CaptureProof AI-enabled fall risk assessment—a feature that delivers rapid screening, supports timely referrals and interventions, and promotes crucial fall awareness.

This integrated vision is designed to empower our seniors to remain strong, independent, and safe within their community.

- Lily Yeo
Head, Active Ageing Centre



Beyond Fall Risk – Hand in Motion Finger Taps



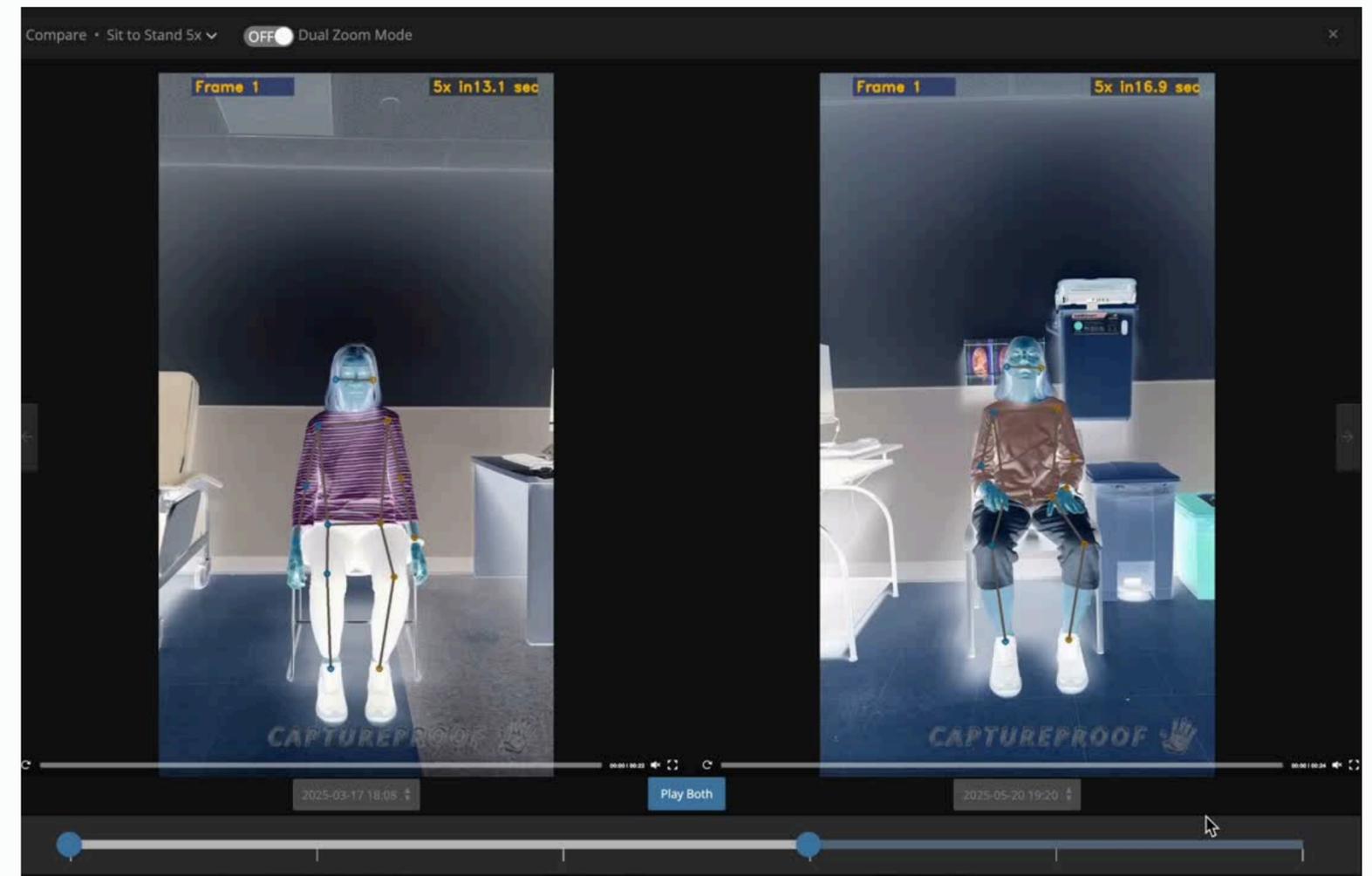
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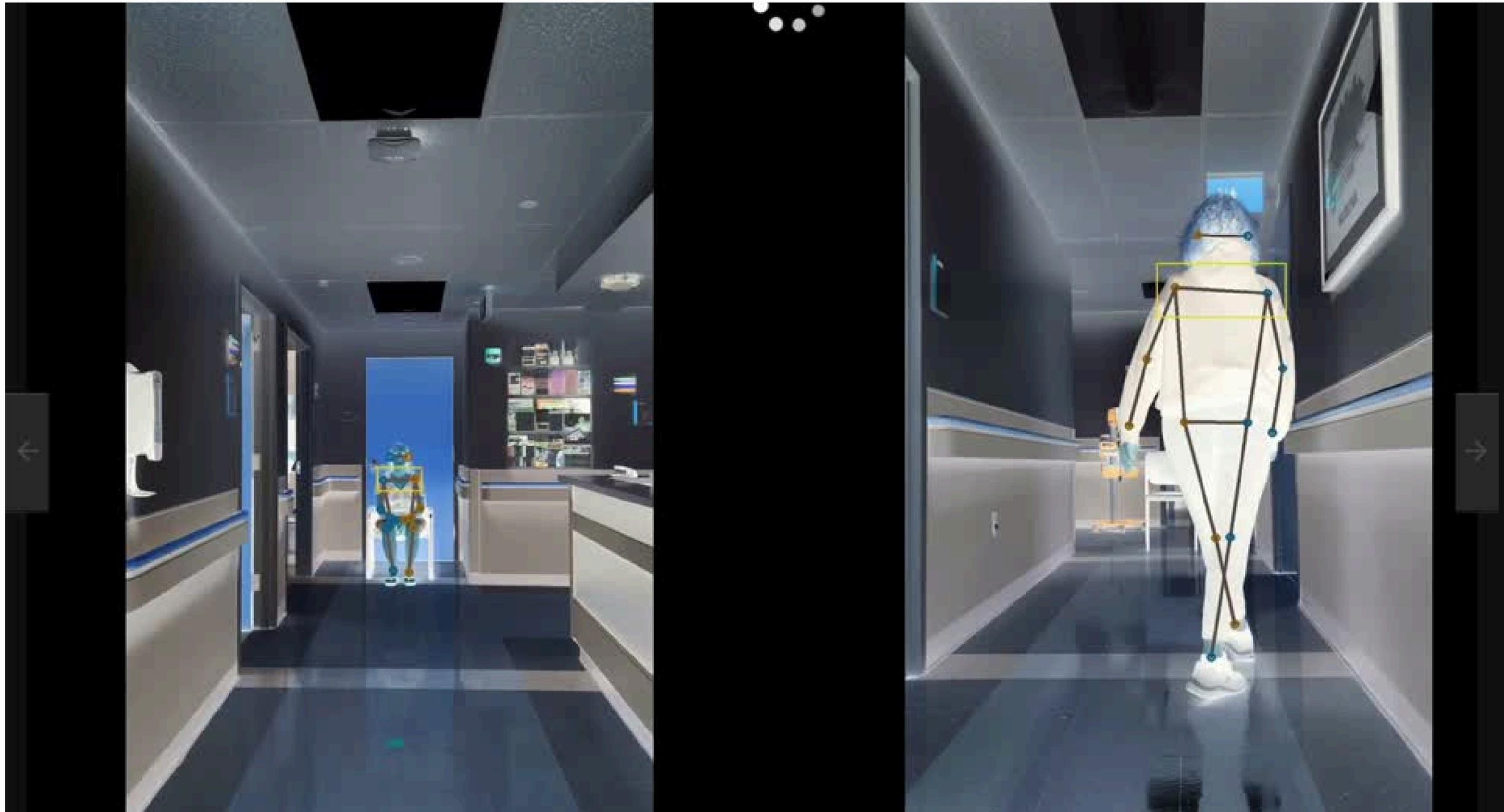
ALS Validation Example

Proven Sensitivity in Neurodegeneration



- **Temple University ALS study:**
 - Compared against ALSFRS-R and ROADS
 - Visual biomarkers detected motor decline weeks to months before scales
- Captured compensatory movement patterns signaling early decline





Beyond the Eye: AI-Enhanced Visual Biomarker Discovery & Tracking for ALS

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Introduction

Amyotrophic Lateral Sclerosis (ALS) is a progressive degenerative neuromuscular disorder causing muscle weakness and impaired mobility. Clinical evaluations including rating scales, such as ALS Functional Rating Scale (ALSF_{RS}-R) and Rasch Overall ALS Disability Score (ROADS), are generally carried out at 3 month intervals. As ALS progresses, attending clinics can become increasingly difficult requiring an increased use of telemedicine. Digital biomarkers offer an innovative solution for more sensitive detection of changes between clinic visits and remote monitoring of persons with ALS (PALS) unable to attend clinic. CaptureProof (CP), a video capture application enhanced by artificial intelligence, may provide this solution through identifying visual biomarkers to remotely monitor ALS progression.

Objectives

The initial goal is to develop and validate a standardized video and photo capture protocol using a smartphone application (CaptureProof) in order to detect and monitor the progression of Amyotrophic Lateral Sclerosis (ALS).

Methods

In order to generate visual biometric markers using CP's AI-Smart Medical Camera, to detect motor involvement in PALS and monitor ALS progression, PALS were recruited from the MDA/ALS clinic at Temple University for this single-arm study. Data collection occurred at Temple University during regular clinic visits or at patients' homes for those unable to attend. Using CaptureProof's smartphone application, participants were recorded performing specific tasks involving facial movement, speech, upper and lower extremity function, balance, and gait (including the Timed Up and Go test). Participants also completed ALSF_{RS}-R and ROADS assessments. CP's proprietary algorithms analyzed the video data, generating biometric markers for each task by evaluating symmetry, fluidity, speed, range, and rate of movements. These markers were compared between PALS and healthy controls, as well as against self-reported ALSF_{RS}-R and ROADS scores. The study was approved by the Institutional Review Board, and informed consent was obtained from all participants.

Initial Results

Our preliminary data set includes 10 PALS (7M, 3F, 48–71 years) and an average ALSF_{RS}-R score of 33.6. Video assessments revealed decreased function in all 9 performing PALS for TUG (9.4–24.7 sec, mean 13.62 ± 5.08 sec vs. norm 8.1 sec), with 5/9 showing normal ALSF_{RS}-R Walking scores, and in 9/9 for 5x SST (12.36–24.5 sec, mean 16.83 ± 4.29 sec vs. norm 11.4 sec), and with 6/9 having normal ROADS gait/stairs scores. Longitudinal data from 1 patient (3 visits) showed increased facial asymmetry (eye closure, lower teeth, pucker lips), reduced max left arm/leg extension, increased max right leg extension, TUG +43.6% to 21.4 sec, and 5x SST +31.5% to 16.3 sec, highlighting CP's precision in tracking progression beyond traditional scales. Audio analysis of the Bamboo Passage in 7 PALS revealed high inhale frequencies (15.0–20.9 breaths/min, mean 17.8 ± 2.2 breaths/min) and reduced words per breath (7.5–8.7, mean 7.8 ± 0.6), indicating respiratory compromise, despite ALSF_{RS}-R scores (21–36) suggesting a milder impairment. Praat analysis confirmed dysarthria (jitter >1%, monotone pitch <50 Hz), with frequent pauses (13–15, ~17% of duration) signaling bulbar decline.

Timed-Up-and-Go: Compared to the normative reference value for 60-69 year-olds (8.1±0.9)¹, data from 10 PALS shows an increased average TUG time (13.6±5.0 sec) among the 9 patients able to perform the test. One patient was unable to complete the TUG due to an inability to walk. All 9 completing subjects had abnormal TUG times (>8.1 sec). On ALSF_{RS}-R walking scores, four patients reported significant dysfunction (score 2, walks with assistance), three reported early ambulation difficulties (score 3), and two reported no difficulties (score 4).

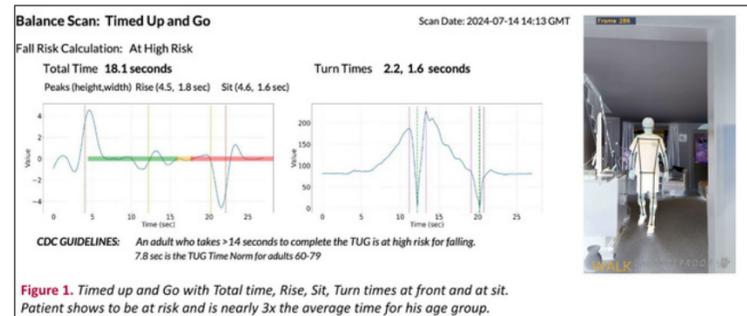
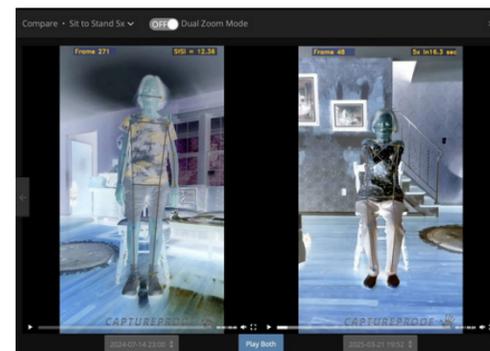


Figure 1. Timed up and Go with Total time, Rise, Sit, Turn times at front and at sit. Patient shows to be at risk and is nearly 3x the average time for his age group.



5-Repetition Sit to Stand: S5-Repetition Sit to Stand: Sit to Stand was performed in 9 participants. Compared to the normative reference value for 60-69 year-olds (11.4 seconds)², data shows increased values in PALS (12.36 - 24.5 sec). ROADS scores associated with gait ("walk around your home") and climbing stairs ("walk up 1 flight of stairs") were normal for both tasks in two participants, normal for walking but abnormal for stairs in four participants, and abnormal for both tasks in two participants, including the individual with the highest sit-to-stand time of 24.5 sec. One patient did not perform the test, due to an inability to stand safely. One patient (seen left) showed a 31.877% increase in Sit to Stand from July 2024 (12.36 seconds) to March 2025 (16.3 seconds).

Finger Taps: Finger taps were recorded for 7 participants. Speed varied between 35.3–150 taps per minute, with tap durations ranging from 0.4–1.7 seconds (estimated from speed). ALSF_{RS}-R scores for handwriting and utensil use were normal (4) for one participant, while six showed varying difficulties (scores 3–1). ROADS scores for "sign name on paper" and "use a knife and fork" were normal (2) for one participant, mixed (2, 1) for one, and abnormal (1) for both tasks in four participants, with the lowest tap speed (35.3–38.2) corresponding to abnormal ROADS scores.

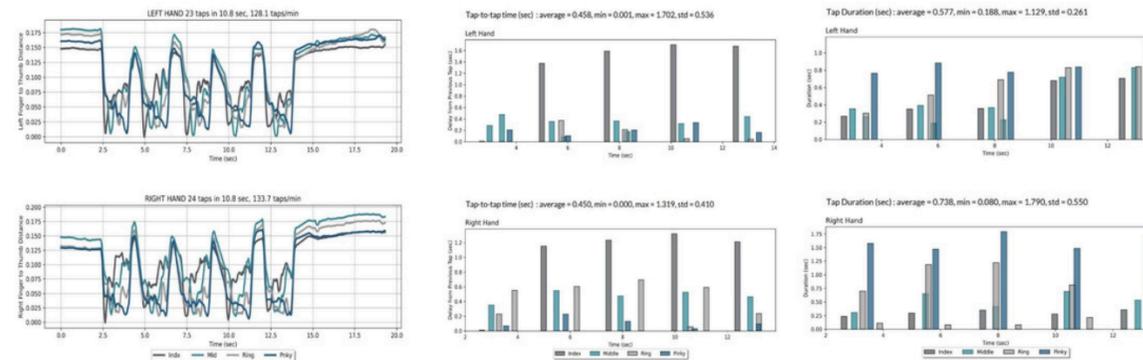


Figure 3. Finger Tap. Patients are instructed to tap each finger to thumb in a forward and reverse sequence.

References:

1. Bohannon RW. Reference values for the timed up and go test: a descriptive meta-analysis. J Geriatr Phys Ther. 2006;29(2):64-8. doi: 10.1519/00139143-200608000-00004. PMID: 16914068.
2. Bohannon RW. Reference values for the five-repetition sit-to-stand test: a descriptive meta-analysis of data from elders. Percept Mot Skills. 2006 Aug;103(1):215-22. doi: 10.2466/pms.103.1.215-222. PMID: 17037663.

Finger Spread: 7/10 PALS Net distances (0.8–41.3 px), spreads (2–19 reps). Initially-proposed metrics (distance from thumb to index finger, spread time) captured and are in development. Analysis of the best biometric and signal are under investigation. Clarity and sensitivity will increase with direct correlation to the sample size.



Figure 4. Finger Spread. Regular motion of fingers. Tremor is apparent at the start and end during the 3 second "hold still" periods.

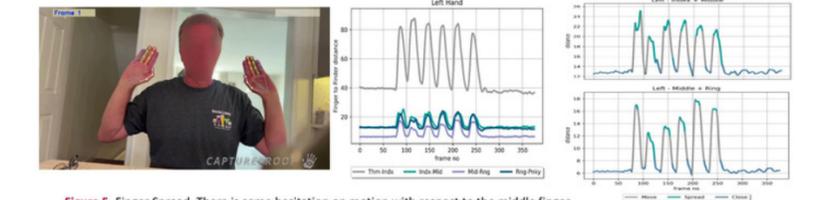


Figure 5. Finger Spread. There is some hesitation on motion with respect to the middle finger.

Facial Movements: Puff (3–16%), asymmetry (0.0004–4.1459%) in 7/10 PALS. F71: lip asymmetry -19.7% to 9.5%, puff declined (Visit 2: 0.201/0.180 to Visit 3: 0.176/0.167 px); M63: 0.0015–4.1459%. Waveform analysis ongoing. **Audio:** 6/10 PALS (128–173.8 WPM, 15–20.9 inhales/min). M63 (173.8 WPM) compensatory; F71stable (141.3 WPM) despite bulbar decline.

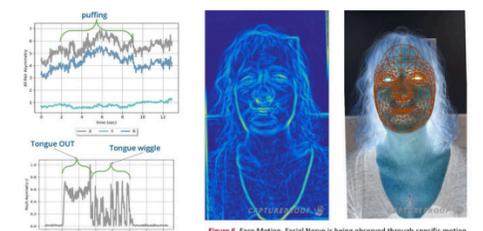


Figure 6. Face Motion. Facial Nerve is being observed through specific motion and identification of symmetry as well as fatigue is being considered.

Discussion and Future Directions

Our data suggests that CaptureProof (CP) accurately captures features associated with ALS motor involvement, enabling detection of functional changes and disease progression. Video assessments of TUG revealed decreased function in all 9 performing subjects (mean 13.62 ± 5.08 sec vs. norm 8.1 sec), despite 5/9 having normal ALSF_{RS}-R Walking scores, suggesting CP's increased sensitivity to motor involvement. This heightened sensitivity, particularly evident in TUG and 5x SST results, highlights the utility of AI-enabled video capture for detecting motor changes, enabling earlier intervention, and tracking ALS progression with precision beyond traditional scales. Audio analysis further supports CP's ability to detect bulbar decline, with high inhale frequencies (mean 17.8 ± 2.2 breaths/min) and dysarthria (jitter >1%) indicating respiratory compromise. Continued data collection will expand our sample size and explore additional motor tasks, biomarker metrics, correlations with ALS rating scales, and longitudinal monitoring of disease progression. The initial set of assessments will be comprehensive, aiming to identify the most sensitive movements for a streamlined test battery, completable in under 10 minutes, to enhance clinical utility. In conclusion, AI-enhanced video assessments provide valuable digital biomarkers for ALS progression, detecting subtle motor and bulbar changes with greater sensitivity than ALSF_{RS}-R and ROADS. From TUG and 5x SST outperforming self-reported scores to longitudinal data showing increased facial asymmetry and motor decline (e.g., TUG +43.6% to 21.4 sec over 3 visits), CP offers a robust tool for early intervention, longitudinal tracking, and personalized care strategies. These findings could inform predictive models and enhance access to specialized ALS care through remote monitoring, with potential applications for other neurodegenerative conditions.

One Capture Layer → Multiple Clinical Performance Metrics

Frailty / Function
fall risk detection,
ageing in place

Neurology
progression + treatment
response

Orthopedics / MSK
rehab trajectory
fine and gross motor

Skin / Wound
high-volume triage;
wounds as subset

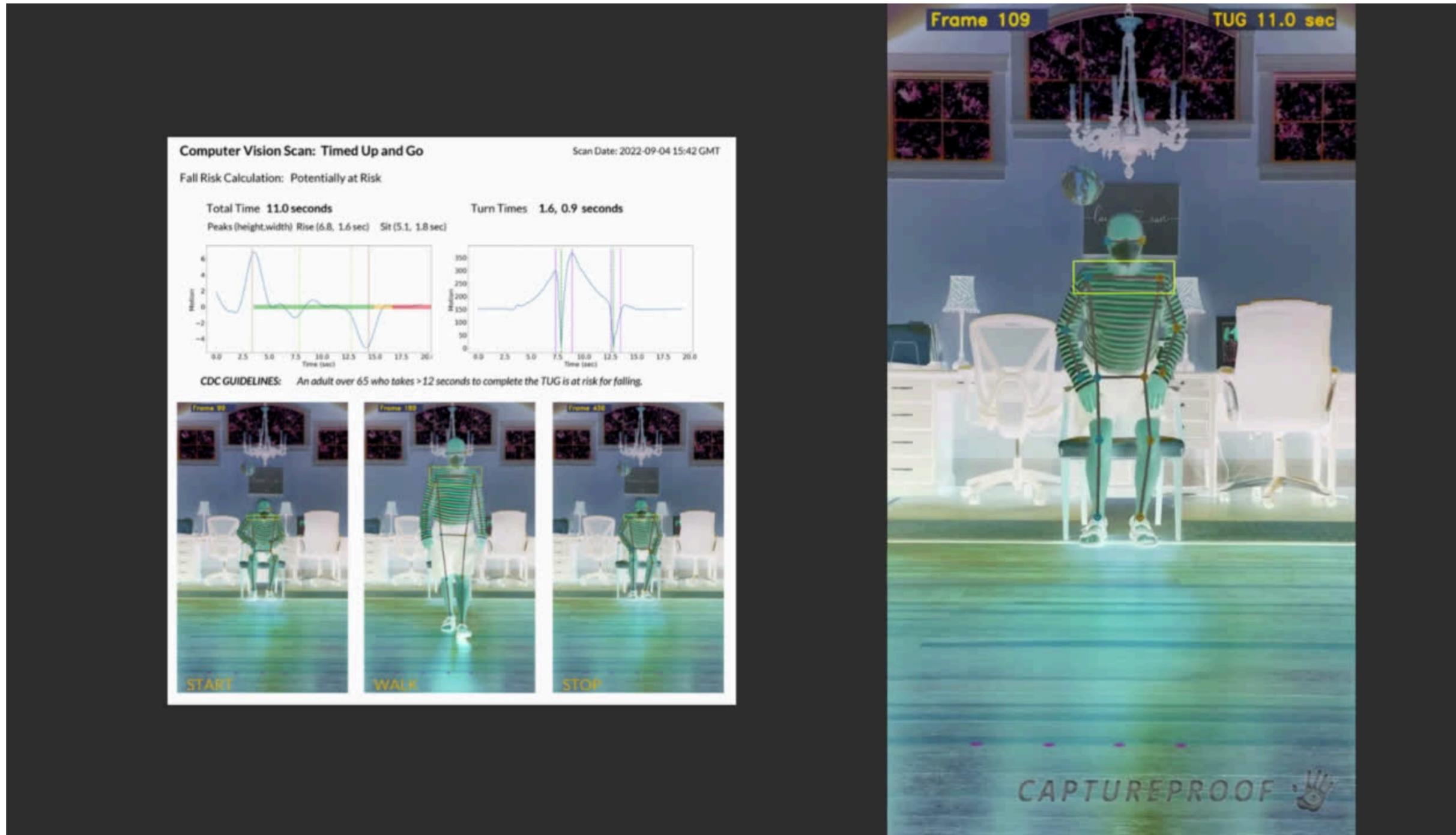


Data Security & Compliance

PDPA-compliant, data encrypted at rest and in transit. CaptureProof prioritizes security and patient privacy.



DISCUSSION, QUESTIONS & FEEDBACK...





"A JOURNEY OF A THOUSAND MILES
BEGINS WITH A SINGLE STEP."

– LAO TZU

Thank you to NUHS, NTUC, NHG,
SingHealth, CHISEL and Temasek

Please reach out to us with questions or suggestions

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