



Empowering Ageing: Tech-driven Tools to Support Function and Well-being

Moderator:

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Speakers



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Using Technology to Address Gaps in Aging

How Wearables and Voice Al Could Modernize Screening, Monitoring and Management

CMS Quality Conference 7/2/2025

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Disclosures

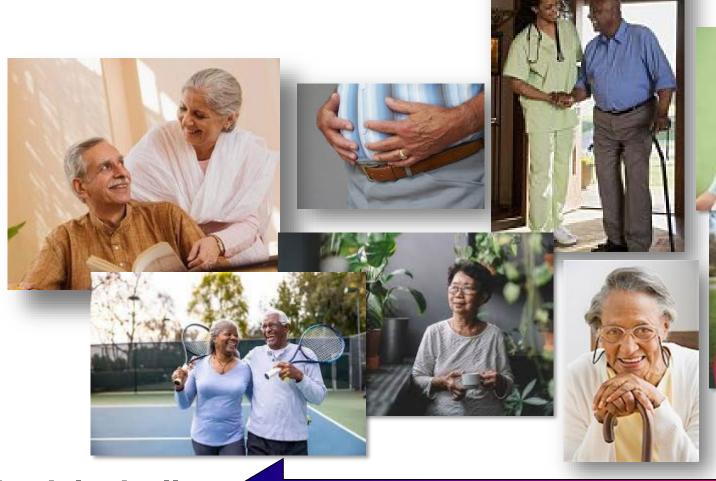
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The University of Chicago and NORC hold Intellectual Property Rights to EngAGE, a technology program which will be discussed today. To date, EngAGE has not been licensed.



80 year old adults....





Physiologically Robust

Frailty

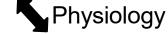
Physiologically Fragile



Frailty Science

External Triggers

Molecular & Genetic



Phenotype

Outcomes



Jeremy Walston, MD
Raymond and Anna
Lublin Professor of
Geriatric Medicine &
Gerontology
Deputy Director,
Division of Geriatric
Medicine and
Gerontology
Co-Director, Biology of
Healthy Aging Program
Johns Hopkins
University

Mitochondrial dysfunction

Epigenetic alterations

Senescence

Autophagy

Altered intercellular communication

Genomic instability

Telomere attrition

Stem cell exhaustion

Deregulated nutrient sensing

Loss of proteostasis

Impaired Stress Response System

Sarcopenia

↑ IL-6, CRP, WBC, Cortisol

Altered Energy Metabolism

Altered Bone Metabolism

RAS

↑ Clotting markers

Glucose intolerance

↓IGF-1, DHEA-S, androgens, estrogens

↓ Micronutrients (total carotenoids, retinol, Vit D, Vit B6, folate)

↓ Immune function

↓ Hemoglobin

↓ Diurnal cortisol variation

↓ Heart rate variability

Weakness / grip strength

Clinical

Unintentional weight loss

Slowness / gait speed

Exhaustion

Low physical activity

Falls / Fracture

Dependency

Social isolation

High Health Care Utilization

Procedural Morbidity & Mortality

Death

Walston 2017, figure adapted with permission





Indications for frailty and physical function evaluation

4Ms Age-Friendly Health Systems Guide

Mobility assessment at all encounters

Frailty Screening Expert Recommendations

- -≥ 70 years
- -≥ 5% weight loss in prior year

Condition and Event Triggers

- Falls evaluation
- Prior to elective procedures
- Diabetes treatment decision making
- Cancer treatment decision making
- Valvular disease treatment decision making
- Medicare Annual Wellness Visit

https://www.johnahartford.org/dissemination-center/view/book-age-friendly-health-systems-a-guide-to-using-the-4ms-while-caring-for-older-adults

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The Successful Aging & Frailty Evaluation (SAFE) ™ Clinic

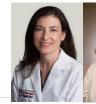






PRIMARY CARE TRANSPLANT UROLOGY ONCOLOGY THORACIC SURGERY

Geriatricians, Geriatrics APP, Social Workers, Physical Therapist, Pharmacist











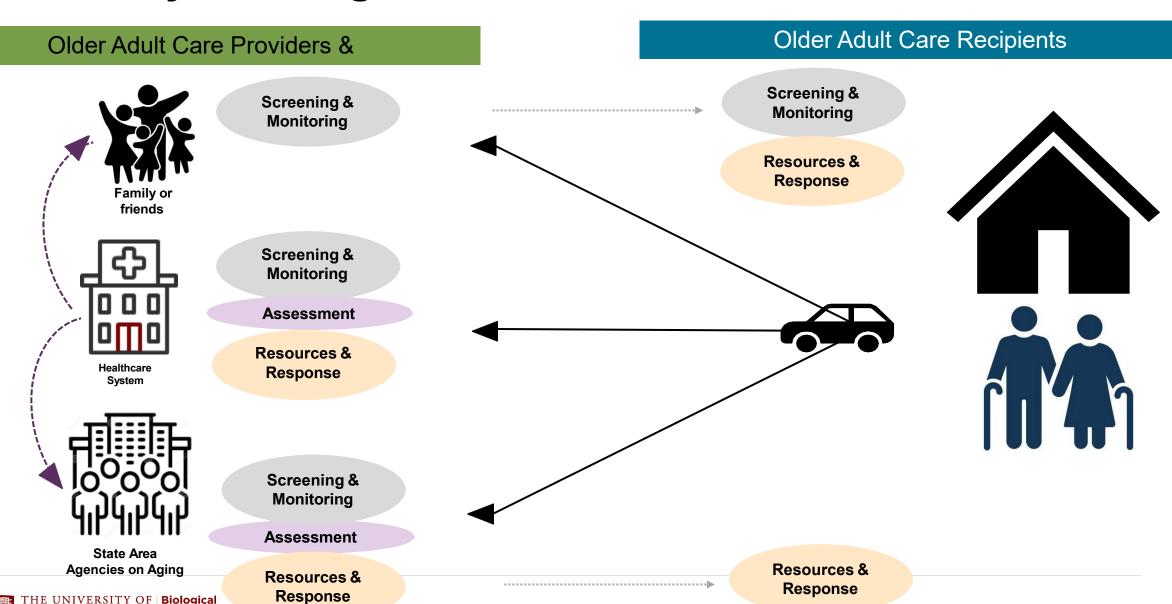




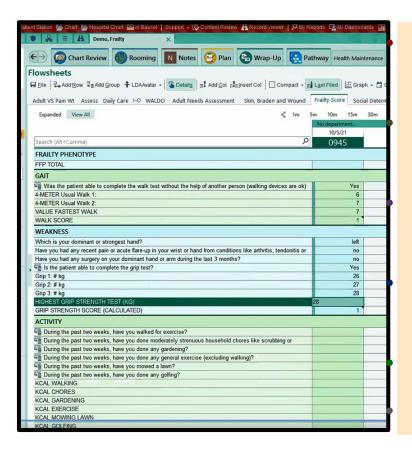




Community-dwelling older adult care & research now...



Frailty and Physical Function Assessment & Management Gaps



Lack of consistent, routine, universally-shared screening of all older adults that trigger early referral to care resources

Time consuming evaluations which prohibits broad adoption despite guidelines

Requires in-person assessments which places an extra burden on people who struggle to reach health services

No mechanism to monitor or incorporate trajectories in risk assessment

Insufficient long-term programming to management frailty
Inefficient coordination of care providers

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Screening for and Monitoring Frailty with Technology?



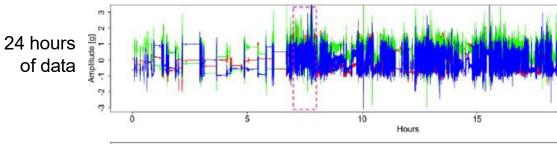


Accelerometers







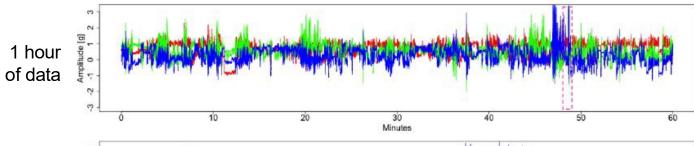








1-mintue of data



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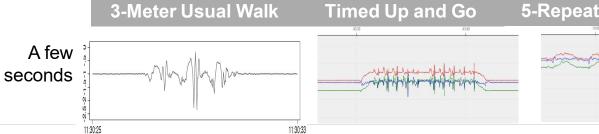
NSHAP BioBox & BioBooklet

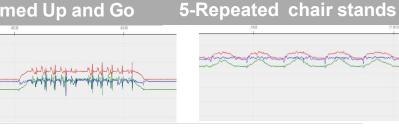




Branch View







Accelerometry activity patterns are complex.

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Shi C et al, npj Aging, 2022

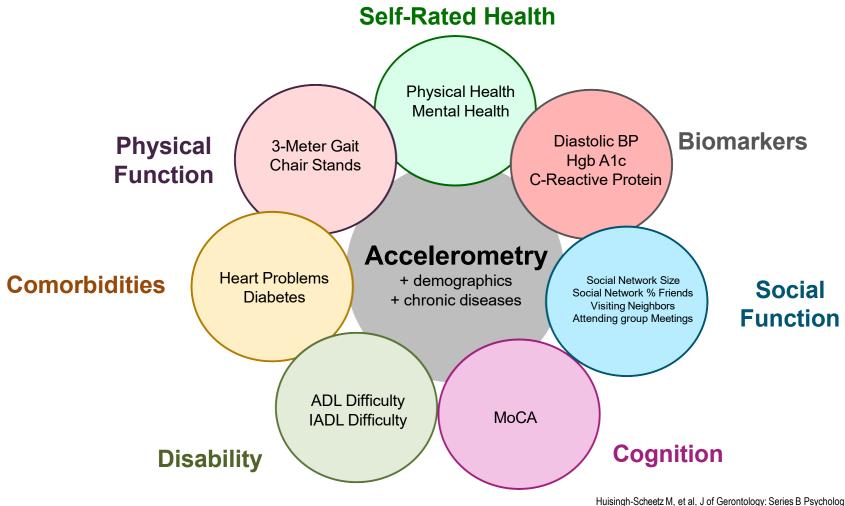
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Higher activity volume = better physical, cognitive and social health





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Rubin D et al, Digital Biomarkers, 2022

Shi C et al, npj Aging, 2022



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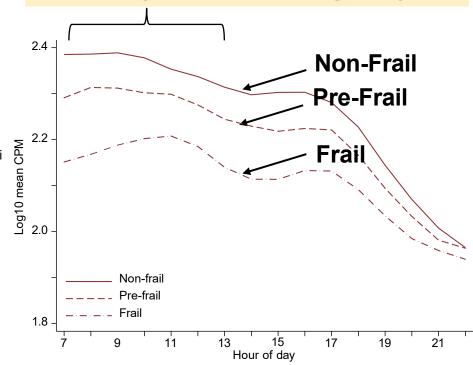


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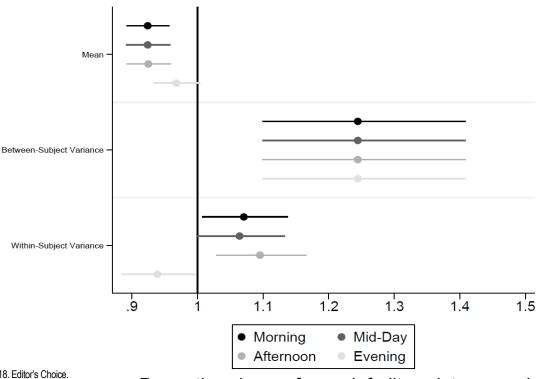
Higher activity volume = better physical function

Linear Regression Outcomes	β (p-value)	N
3-meter walk (seconds)	-0.02 (<0.001)	610
Chair stands (seconds)	-0.02 (0.002)	554

Frailty-related morning delay



Combining morning activity levels & pattern variance are independently associated with frailty



Proportion change for each frailty point across day

Huisingh-Scheetz M, et al, J of the Amer Geriatrics Society, 2016
Huisingh-Scheetz M, et al, J of Gerontology: Series A Biological Sciences and Sciences, 2018. Editor's Choice.
Huisingh-Scheetz M, et al, J of Gerontology: Series A Biological Sciences and Sciences, 2021.

Combined accelerometry features from just 2-3 mornings predicts frailty decline and death, increasing feasibility and scalability.

Survey-Weighted 2-Stage Multivariate Regression Model Predicting 5-Year Mortality & Frailty

N=584 (n=104 died, n=480 alive at 5 years)

	Partially Adjusted Model		Fully Adjusted Model*	
	Logistic	Ordinal	Logistic	Ordinal
	5-Year Mortality*	5-Year Frailty*	5-Year Mortality*	5-Year Frailty*
	OR (p-value)		OR (p-value)	
Mean Morning Hourly Activity (z-score) [△]	0.51 (0.004)	0.81 (0.12)	0.64 (0.04)	0.82 (0.16)
Within-Subject Morning Hourly Activity (z-score) [△]	0.65 (0.04)	0.68 (0.01)	0.64 (0.10)	0.67 (0.01)
Baseline frailty score (0-4)	1.54 (0.002)	2.25 (<0.001)	1.43 (0.04)	1.99 (<0.001)

[△]For each 1 standard deviation

Accelerometry variable correlation: -0.13



^{*}Adjusted for demographics, Charlsons, BMI, MoCA

Machine learning models can use of many accelerometry pattern features to maximize prediction of aging outcomes.

Accelerometry Measures

75th Percentile VMC

75th Percentile of CPM

Harmonic features (98)

Correlation = 0.01 to 0.91

XGBoost machine learning model predicting any decline vs none/improvement on cognitive test using >100 accelerometry pattern features

Sample 1. Local, n=115, 65+, 82% African-American, 7-day hip accelerometry, 1-year follow-up

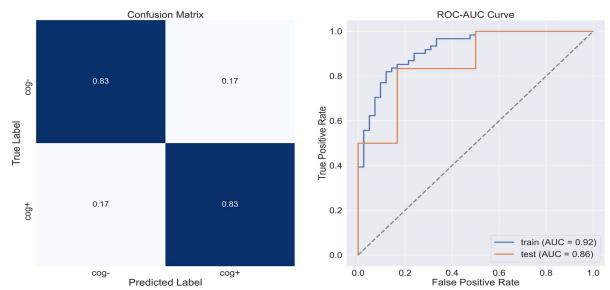


Figure 1. Confusion matrix and ROC-AUC curve summarizing experiments relating 7-day hip accelerometry to 1-year MoCA decline in local sample.

Sample 2. Nationally representative, n=575, 65+, 74% White, 3-day wrist accelerometry, 5-year follow-up.

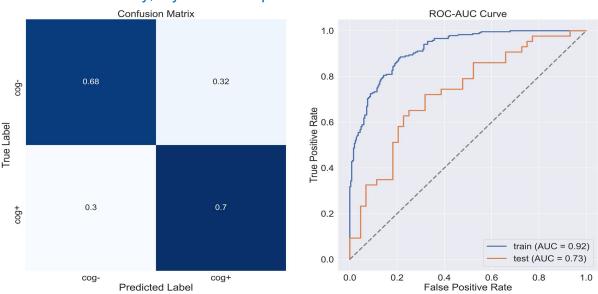


Figure 2. Confusion matrix and ROC-AUC curve summarizing experiments relating 72-hour wrist accelerometry to 5-year MoCA-SA decline.



<u>Detecting and Monitoring Frailty</u>: Combining accelerometry measures to improve detection AND prediction of frailty (prelim results).



Ben Kramer, BA cand Case Western



Andrey Rzhetsky, PhD Professor Department of Medicine and University of Chicago



Yanan Long, PhD UChicago Post-doc



Michelangelo Pagan, BS University of Chicago

Free-Living wrist and hip accelerometry forecast frailty decline among older adult over 1 and 5 years in two distinct observational cohorts. *In progress*.



Accelerometry Measures
Sleep
Total sleep time
Wake Time
Wake after sleep onset
Sleep fragmentation
% Sleep
Sleep variance
Activity
Activity timing
Total activity volume
Peak activity
Average activity level
Mean daily active-to-sedentary transition probability
Time in activity by intensity
Average number of sustained moderate-to-vigorous activity bouts
Activity variance
Gait
Stride amplitude
Cadence
Stride-to-stride variability
Cyclical Patterns
Harmonic features (98)

Detection: frailty phenotype = 0 vs ≥1 points

Method	F ₁ (†)	Precision (†)	Recall (†)	Accuracy (†)	AUROC (†)
Tabular					
CatBoost	0.624	0.690	0.570	0.657	0.657
LightGBM	0.898	0.815	1.000	0.887	0.887
XGBoost	0.804	0.673	1.000	0.757	0.757
DART	0.761	0.697	0.837	0.736	0.736
MLP	0.570	0.933	0.410	0.690	0.690
ResNet	0.840	0.842	0.838	0.840	0.840
Time Series					
MiniRocket-Ridge	0.637	0.614	0.661	0.622	0.622
MiniRocket-Logistic	0.606	0.626	0.586	0.618	0.618

1-Year Prediction: frailty phenotype = worse vs same/better

Method	$F_1(\uparrow)$	Precision (↑)	Recall (↑)	Accuracy (†)	AUROC (↑)
Tabular					
CatBoost	0.807	0.964	0.694	0.834	0.834
LightGBM	0.704	0.891	0.582	0.756	0.756
XGBoost	0.796	0.856	0.743	0.809	0.809
DART	0.752	0.645	0.902	0.703	0.703
MLP	0.812	0.891	0.746	0.828	0.828
ResNet	0.589	0.857	0.448	0.687	0.687
Time Series					
MiniRocket-Ridge	0.680	0.803	0.590	0.722	0.722
MiniRocket-Logistic	0.538	0.658	0.455	0.609	0.609

^{*}Hip accelerometry data; local cohort; adjusted for age, educ, race, income, comorbidities, baseline frailty

Discussion

How do we leverage all noninvasive sensor information to universally screen for aging decline in the free-living environment?

How do we scale scientific findings?

How do we implement data into caregiver, clinical and community workflows?

How can we use the data to trigger and monitor interventions?























Managing Frailty with Technology?





Evidence-Based, In-Home Frailty Treatment



Short Term Medicare Coverage

Long Term Medicare Coverage

Physical & Occupational Therapy

None



Dietician (few indications)

None

Multidisciplinary Care Geriatrics

Social worker

Pharmacy (few indications)

Geriatrics

None

Pharmacy (few indications)



Older Adults Health Technology Industry Landscape

Image by: Keren Etkin

https://www.thegerontechnologist.com/



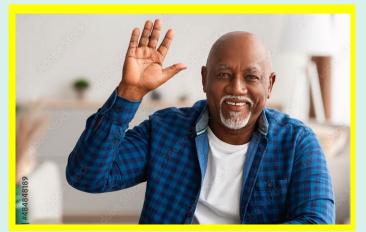




Empowering and connecting for a happier, healthier you



Website









Louise Hawkely, PhD Senior Research Scientist NORC





Older adult user

- Custom Alexa skill
- NIA Go4Life Program Exercises
- Audio & visual instructions
- Pictures
- Music
- Reads encouraging family messages
- Auto adjusts intensity





Care partner user

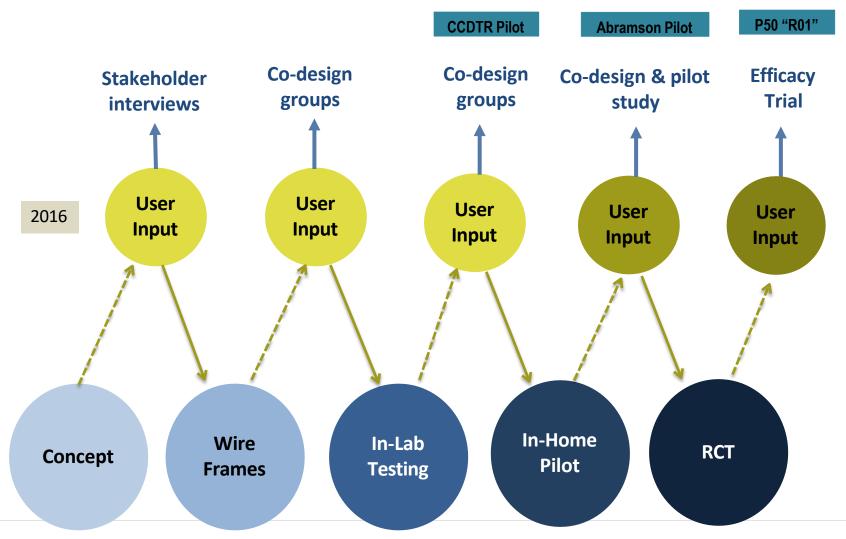
- Daily exercise tracking
- Notices of (in)activity
- Send and receive messages
- •Data from website are stored on HIPPA-approved server.







Participatory design means Co-CREATION



EngAGE 6-month, Double-Blinded, Randomized Controlled Trial

Aim: To test the efficacy of EngAGE on physical & social function versus paper-based exercises.

Intervention arm: EngAGE: Custom voice-activated device application for in-home exercise

Comparison arm: Paper-based in-home exercises

Target group:

- 124 Multimorbid, frail, homebound older adult + care partner dyads.
- 70 dyads randomized to date.



Brandon Foster Technology Support



Brittni Bryant, MPH Program manager



Sylvia Brown, MA Research Staff



Ellen Bloss, MS Research Staff



Wen Wan, PhD Biostatistician

Trial registration: ClinicalTrials.gov NCT05337514

Funding: NIMHD P50MD017349-01, 8199



Prelim blinded outcomes (merged trial arms)

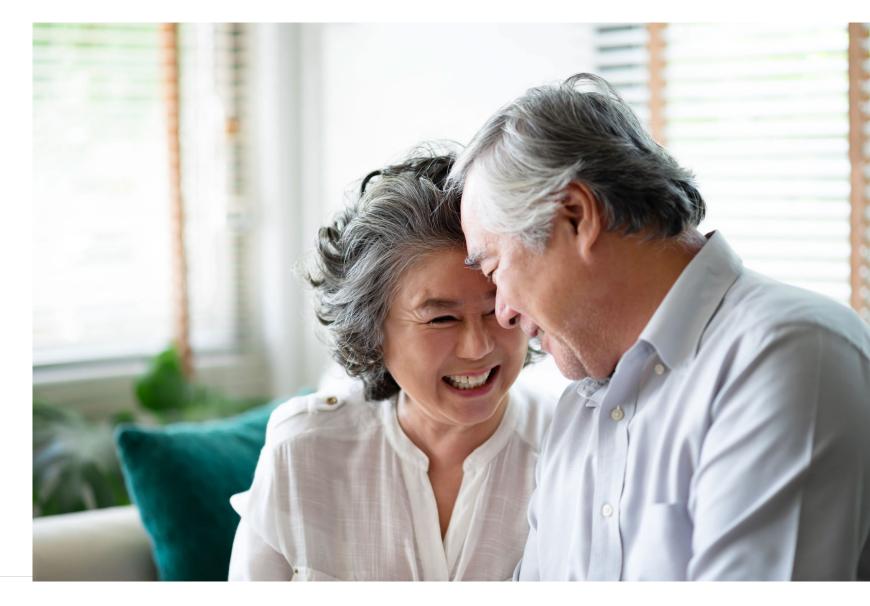
		Visit					
		Baseline	Month 3	Month 6			
		(N=61)	(N=45)	(N=41)			
	Time to complete 5 chair stands						
	Mean (SD)	21.66 (6.74)	20.31 (7.61)	19.82 (6.18)			
et leget 1	Median (Range)	20.0 (11.7 – 46.0)	19.3 (9.3 – 45.1)	19.0 (8.0 – 34.4)			
at least 1	N (N Missing)	61 (0)	37 (8)	38 (3)			
chair							
stand at	Unable to do or >60 seconds.	0 (0.00%)	1 (2.63%)	1 (2.56%)			
baseline	>=16.70 seconds.	48 (78.69%)	22 (57.89%)	25 (64.10%)			
	13.70-16.69 seconds.	11 (18.03%)	9 (23.68%)	9 (23.08%)			
	11.20 -13.69 seconds.	2 (3.28%)	5 (13.16%)	2 (5.13%)			
	< =11.19 seconds.	0 (0.00%)	1 (2.63%)	2 (5.13%)			
	Missing	0	7	2			
		Visit					
		Baseline	Month 3	Month 6			
Unable to		(N=9)	(N=8)	(N=8)			
do at least	SPPB Chair Score:						
1 chair	Unable to do or >60 seconds.	9 (100.00%)	4 (50.00%)	2 (25.00%)			
stand at	>=16.70 seconds.	0 (0.00%)	3 (37.50%)	4 (50.00%)			
baseline	13.70-16.69 seconds.	0 (0.00%)	1 (12.50%)	0 (0.00%)			
	11.20 -13.69 seconds.	0 (0.00%)	0 (0.00%)	1 (12.50%)			
	< =11.19 seconds.	0 (0.00%)	0 (0.00%)	1 (12.50%)			

Discussion...

What is the best strategy to promote tech interventions that are broadly usable and easily adoptable across a range of tech literacy?

How do we use team science to evaluate new and existing tech-based, long-term care programs for healthy aging?

How do we pair monitoring tools with interventions?





Thank you.

megan.huisingh-Scheetz@bsd.uchicago.edu







Precision rehabilitation and emerging technologies

Ryan T. Roemmich, PhD

Director, Center for Movement Studies, Kennedy Krieger Institute

Associate Professor, Department of Physical Medicine and Rehabilitation, Johns Hopkins University School of Medicine

Director, Rehabilitation Precision Medicine Center of Excellence, Johns Hopkins Medicine

CMS Quality Conference

July 2, 2025



What is precision rehabilitation?

Focus on **function**

What is the patient's physical status?

Does the patient have a support system?

Where was the patient discharged after their inpatient stay?

Is there relevant genetic information available?

What is the patient's socioeconomic background?

What is the patient's emotional status?

What is the patient's cognitive status?

Does the patient have reliable transportation?

Does the patient live in proximity to necessary services?





How can we provide the right patient with the right care at the right time?



What is the Rehabilitation Precision Medicine Center of Excellence? Precision at three levels

Healthcare system/operations

- How should rehabilitation be structured and delivered?
- How can we make informed decisions about resource allocation?

Measurement

- How can we develop tools for measurement outside of the clinic?
- How can we develop tools for more granular assessment?

Personalized interventions

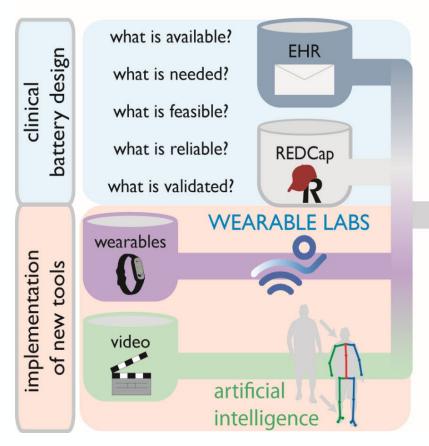
 How can we develop interventions to target patient-specific deficits?





Roadmap for precision rehabilitation

Resource Roadmap for Precision Rehabilitation

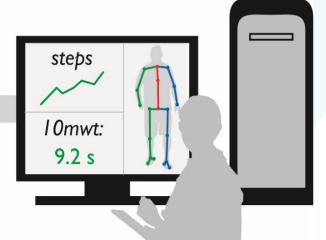


integration and analysis of clinical and real-world data



analytics platform

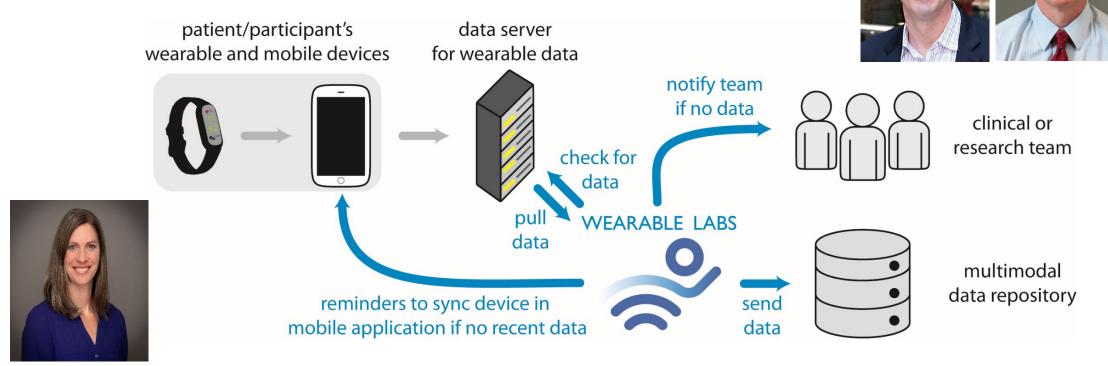
personalized, integrated multimodal data to clinicians







Wearable Labs



Impact of automated data flow and reminders on adherence and resource utilization for remotely monitoring physical activity in individuals with stroke or chronic obstructive pulmonary disease

Margaret A. French, Aparna Balasubramanian, Nadia N. Hansel, Sharon K. Penttinen, Robert Wise, Preeti Raghavan, Stephen T Wegener, Papar T. Roemmich, Pablo A. Celnik
doi: https://doi.org/10.1101/2024.04.15.24305852

This article is a preprint and has not been peer-reviewed [what does this mean?]. It
reports new medical research that has yet to be evaluated and so should not be used to
Ryan T Roemmich 1,5, and Pablo Celnik 2
guide clinical practice.

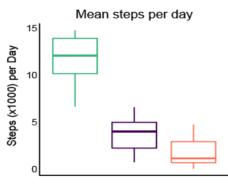
Original Research

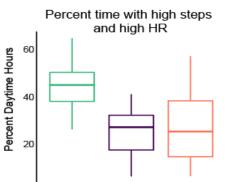
The feasibility of remotely monitoring physical, cognitive, and psychosocial function in individuals with stroke or chronic obstructive pulmonary disease

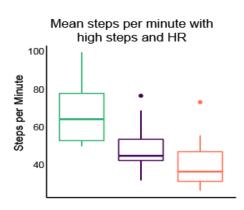
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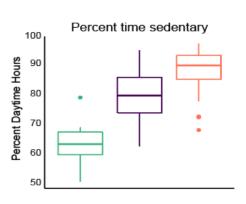
Margaret A French [10] ¹, Eva Keatley [10] ¹, Junyao Li¹, Aparna Balasubramanian², Nadia N Hansel², Robert Wise², Peter Searson^{1,3}, Anil Singh⁴, Preeti Raghavan¹, Stephen Wegener¹, Ryan T Roemmich [10] ^{1,5}, and Pablo Celnik¹

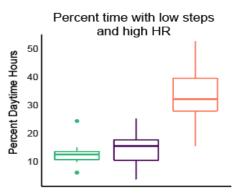
Leveraging wearables in rehabilitation research

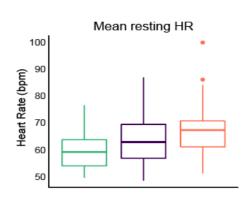


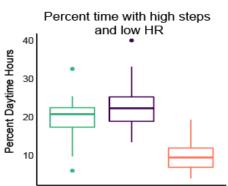












Active Group
Sedentary Group
Deconditioned Group





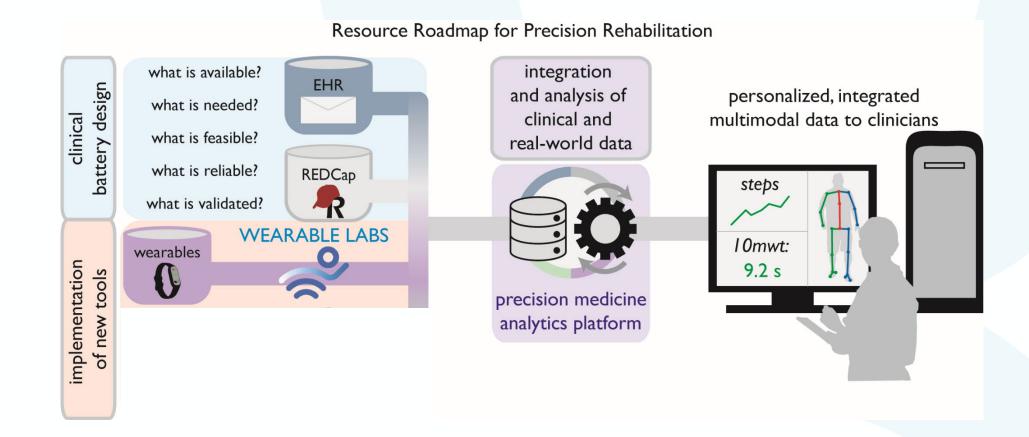






Emily Akrong
Elizabeth Rosenthal

Roadmap for precision rehabilitation...

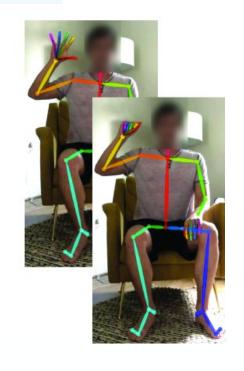






Measurement of human movement

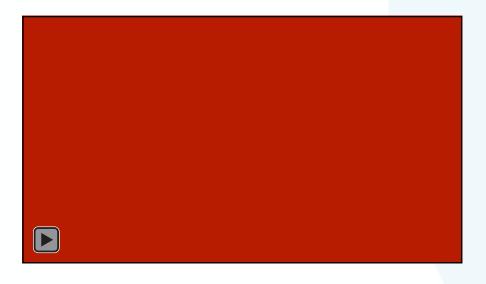


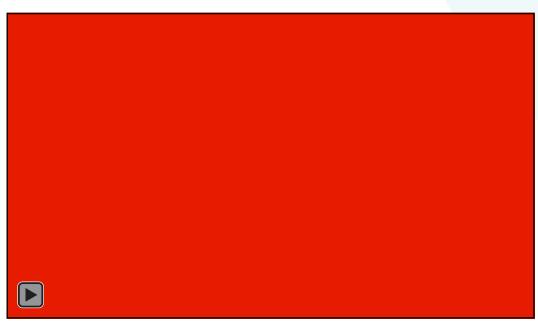




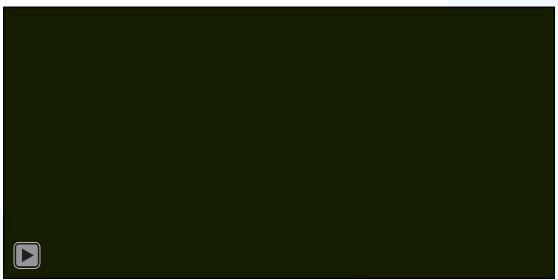


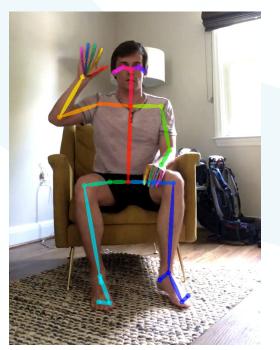


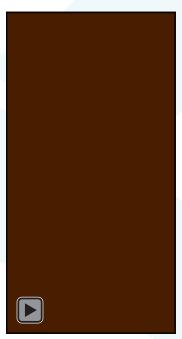










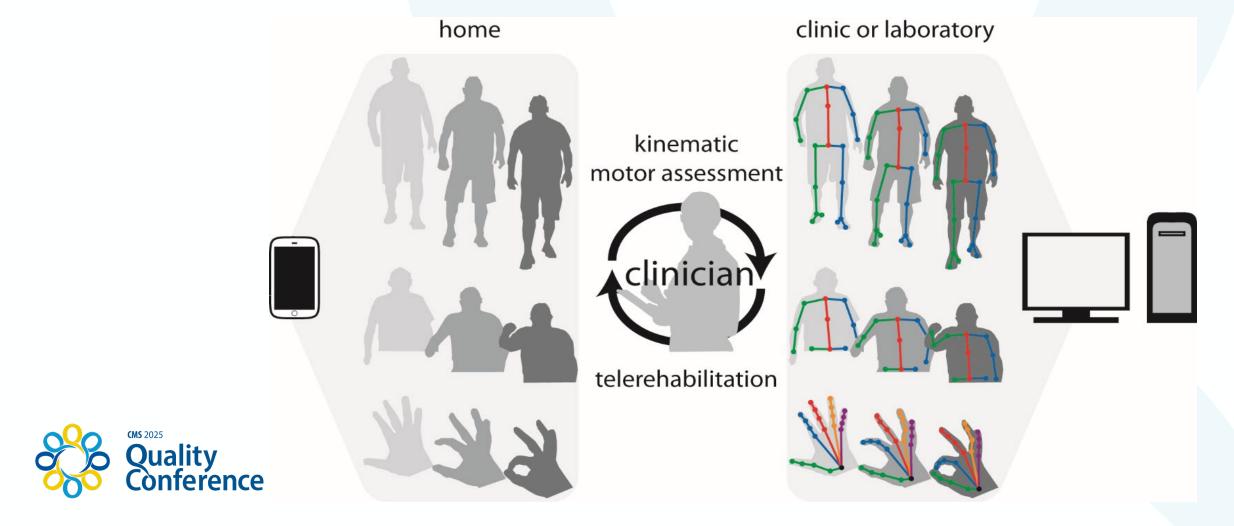




We see considerable potential for use of this technology to address rapidly expanding needs for remote measurement and delivery of telerehabilitation.

The Big Picture Goal:

Clinician-centered use of pose estimation for facilitating remote assessment and telerehabilitation



Interest around KKI/Hopkins (and elsewhere)

Validation of markerless video-based gait analysis using pose estimation in toddlers with and without neurodevelopmental disorders

Jeffrey T. Anderson¹, Jan Stenum², Ryan T. Roemmich^{2,3} and Rujuta B. Wilson^{1,4}*

RESEARCH ARTICLE

Clinical gait analysis using video-based pose estimation: Multiple perspectives, clinical populations, and measuring change

Jan Stenum (5)1,2, Melody M. Hsu1,3, Alexander Y. Pantelyat4, Ryan T. Roemmich (5)1,2-

RESEARCH ARTICLE

Video-based quantification of human movement frequency using pose estimation: A pilot study

Hannah L. Cornman^{1,2,3}, Jan Stenum^{1,2}, Ryan T. Roemmich 6,1,2*

JOURNAL ARTICLE

Accuracy of Video-Based Gait Analysis Using Pose **Estimation During Treadmill Walking Versus** Overground Walking in Persons After Stroke @

Kristen John, BS, Jan Stenum, PhD, Cheng-Chuan Chiang, DO, Margaret A French, PT, DPT, PhD, Christopher Kim, MS, John Manor, DO, ATC, Matthew A Statton, MD, Kendra M Cherry-Allen, PT, DPT, PhD, Ryan T Roemmich, PhD

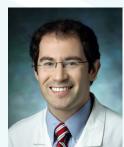




















Two-dimensional video-based analysis of human gait using pose estimation

Opportunities for Improving Motor Assessment and Rehabilitation After Stroke by Leveraging Video-Based

Cherry-Allen, Kendra M. PT, DPT, PhD; French, Margaret A. PT, DPT, PhD; Stenum, Jan PhD; Xu, Jing PhD; Roemmich,

ANALYSIS & PERSPECTIVE

Ryan T. PhD

Pose Estimation

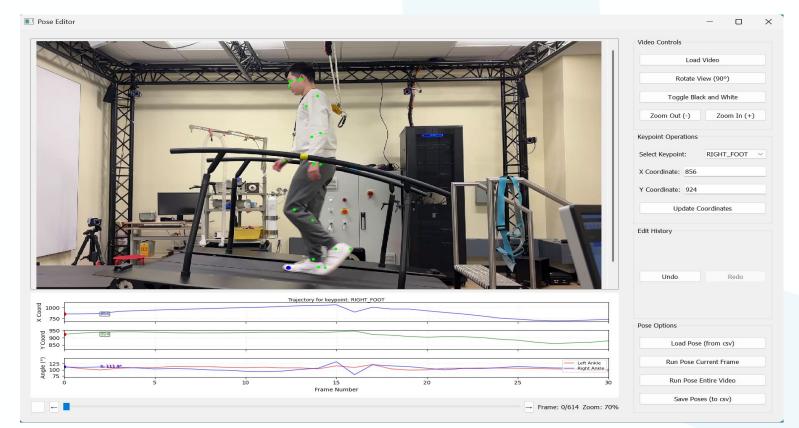


RESEARCH ARTICLE

Applications of Pose Estimation in Human Health and Performance across the Lifespan

by Jan Stenum 1,2 ☑, Kendra M. Cherry-Allen 2 ☑, Connor O. Pyles 3 ☑ ②, Rachel D. Reetzke 4,5 ☑, Michael F. Vignos 3 ☐ and Ryan T. Roemmich 1,2,* ☐ 0

Improving accessibility







JOHNS HOPKINS
TECHNOLOGY VENTURES

https://github.com/JeffZC/p ose-editor/tree/mediapiperr21











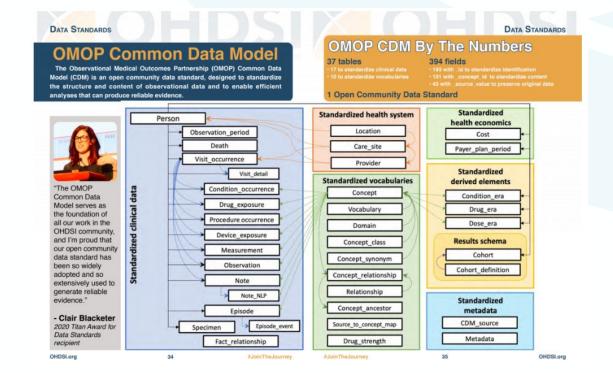
Where are we headed?

Expanding inter-institutional collaboration

Rehabilitation Common Data Model group

Incorporating useful technology

- Beiwe (MGH collaboration)
- NLP (collaborations with Depts of Medicine and Radiology)







Rehabilitation Precision Medicine Center of Excellence Our team 2019-now



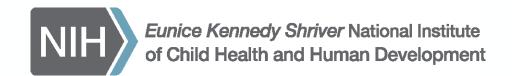
Thank you!



Email: Roemmich@kennedykrieger.org

Funding:
JH AITC (NIH P30AG073104)
NICHD, NIA

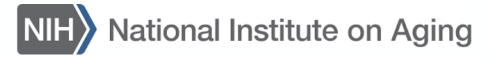
American Parkinson Disease Association
Stanford RESTORE Center (NIH P2CHD101913)
Sheikh Khalifa Stroke Institute (JHM)
Association of Academic Physiatrists
American Heart Association

















Transforming Hearing Through Evidence, Policy & Awareness

Frank Lin, MD PhD

Professor of Otolaryngology & Epidemiology

Johns Hopkins University

Disclosures:

Independent contractor to Apple since 2022 (currently on sabbatical with Apple from 2024-2025)

My presentation today is in my role as a Professor at Johns Hopkins University. I do not represent or speak on behalf of Apple Inc.

Foundational Principles

Everyone's hearing declines with age

Hearing allows us to engage with others & the environment and is foundational to health

Transformative approaches are needed to address hearing loss globally

Hearing Health Transformation Over the Past Decade



Evidence

Demonstrating the impact of hearing & hearing interventions on health



Regulatory policy

Allowing for innovation & accessibility in the hearing technology market



Awareness

of hearing as a life course aspect of health – JHU Hearing Number campaign

Hearing Health Transformation Over the Past Decade-



Evidence

Demonstrating the impact of hearing & hearing interventions on health



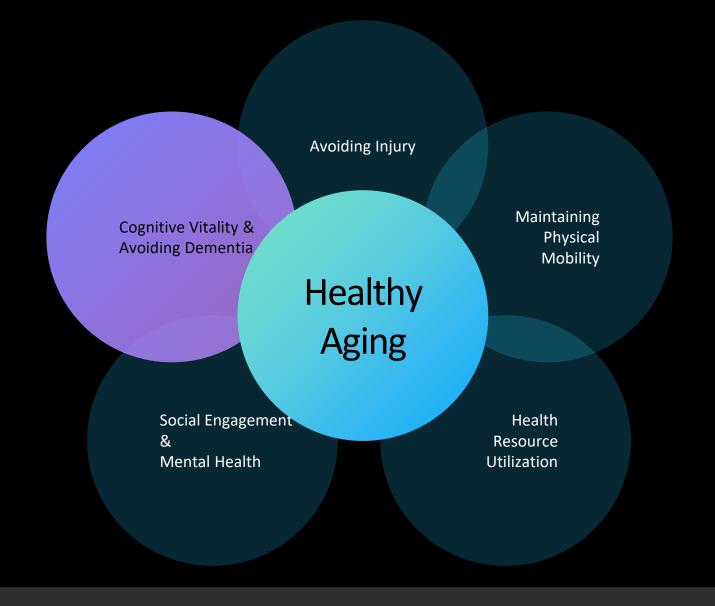
Regulatory policy

Allowing for innovation & accessibility in the hearing technology market

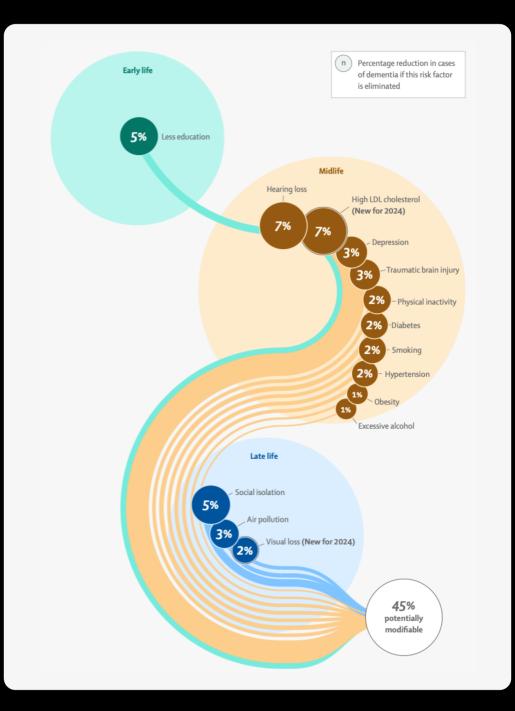


Awareness

of hearing as a life course aspect of health – JHU Hearing Number campaign



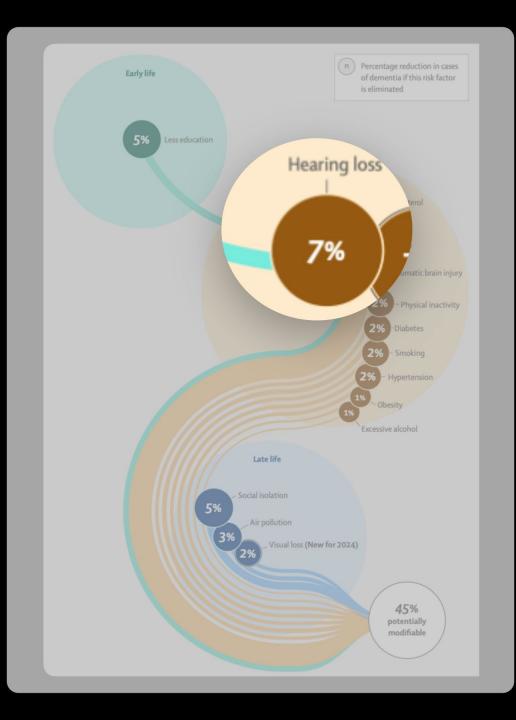
Hearing



Potentially Modifiable Risk Factors for Dementia

2024 Lancet Commission on Dementia Prevention, Intervention & Care

Hearing loss in mid & late life identified as one of the single largest potentially modifiable risk factors for dementia



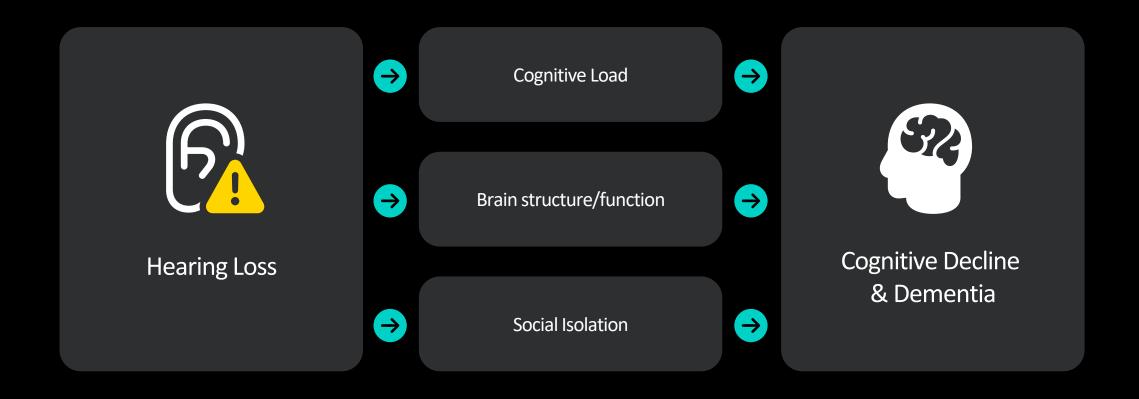
Potentially Modifiable Risk Factors for Dementia

2024 Lancet Commission on Dementia Prevention, Intervention & Care

Hearing loss in mid & late life identified as one of the single largest potentially modifiable risk factors for dementia

Hearing Loss & Dementia

Mechanistic Pathways



Potential role for hearing intervention in reducing risk of cognitive decline & dementia

Reduce the cognitive load of processing degraded sound

Brain structure/function

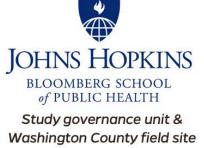
Provide increased brain stimulation

Social Isolation

Improve social engagement

ACHIEVE Collaborative Research Group (2014 - present)













Minneapolis field site

ACHIEVE study



Hearing intervention design



Health education control intervention design



MRI reading center

THE LANCET

Volume 402 · Number 10 404 · Pages 747-824 · September 2-8, 2023

www.thelancet.com

"Based on evidence from the ACHIEVE study, hearing loss might be a particularly important global public health target for dementia prevention efforts."

See Articles page 786

September 2023

Hearing intervention confers a 48% reduction in cognitive loss over 3 years in adults at increased risk of cognitive decline in the ACHIEVE (n = 977) randomized controlled trial

Hearing Health Transformation Over the Past Decade...



Evidence

Demonstrating the impact of hearing & hearing interventions on health



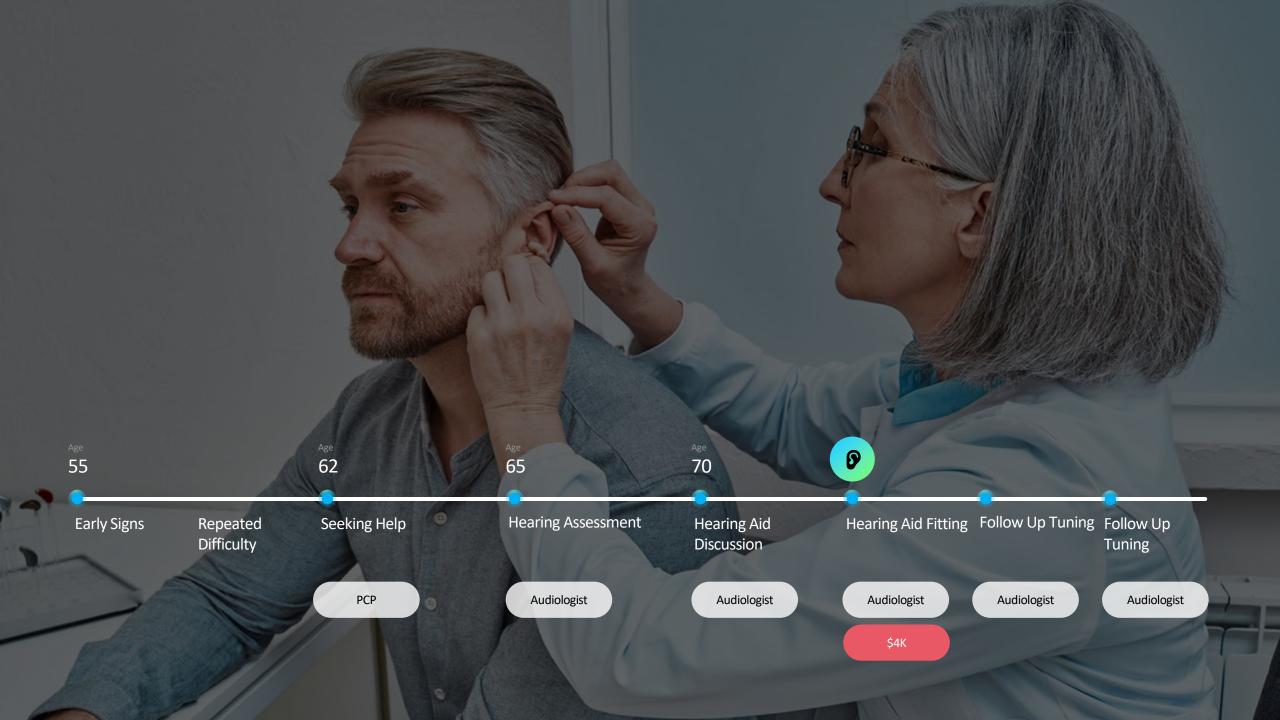
Regulatory policy

Allowing for innovation & accessibility in the hearing technology market



Awareness

of hearing as a life course aspect of health – JHU Hearing Number campaign



Historical Regulatory Policy for Hearing Aids Contributed to this Care Model

Original 1977 FDA regulations for hearing aids (based on technology of that era) precluded hearing aids from being sold OTC directly to consumers

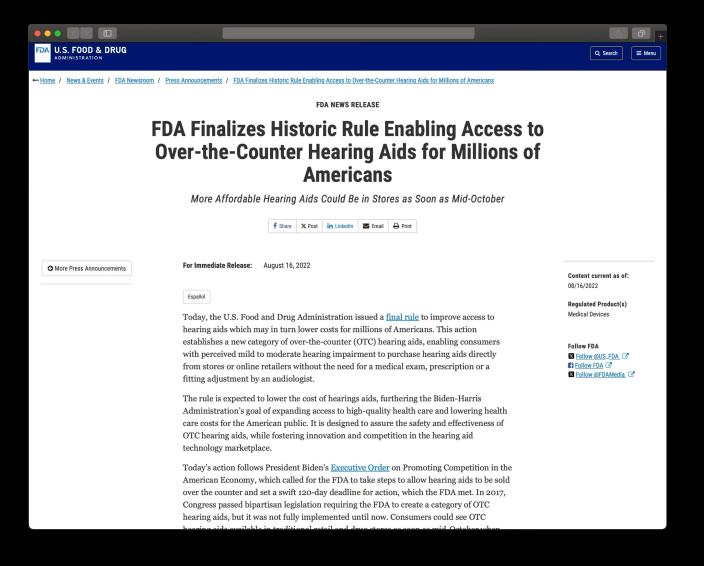
Gatekeeper model – Providers control access to HAs

Barriers to entry for any new hearing aid manufacturer given need to sell through clinical providers

Extensive 'friction' & high costs for consumers to access hearing care

Correcting the Hearing Market Required Changing Outdated Regulatory Policy

Bipartisan efforts in Congress & the White House resulted in passage of the OTC Hearing Act of 2017 and enactment of regulations in 2022



Market Entrants since 2022

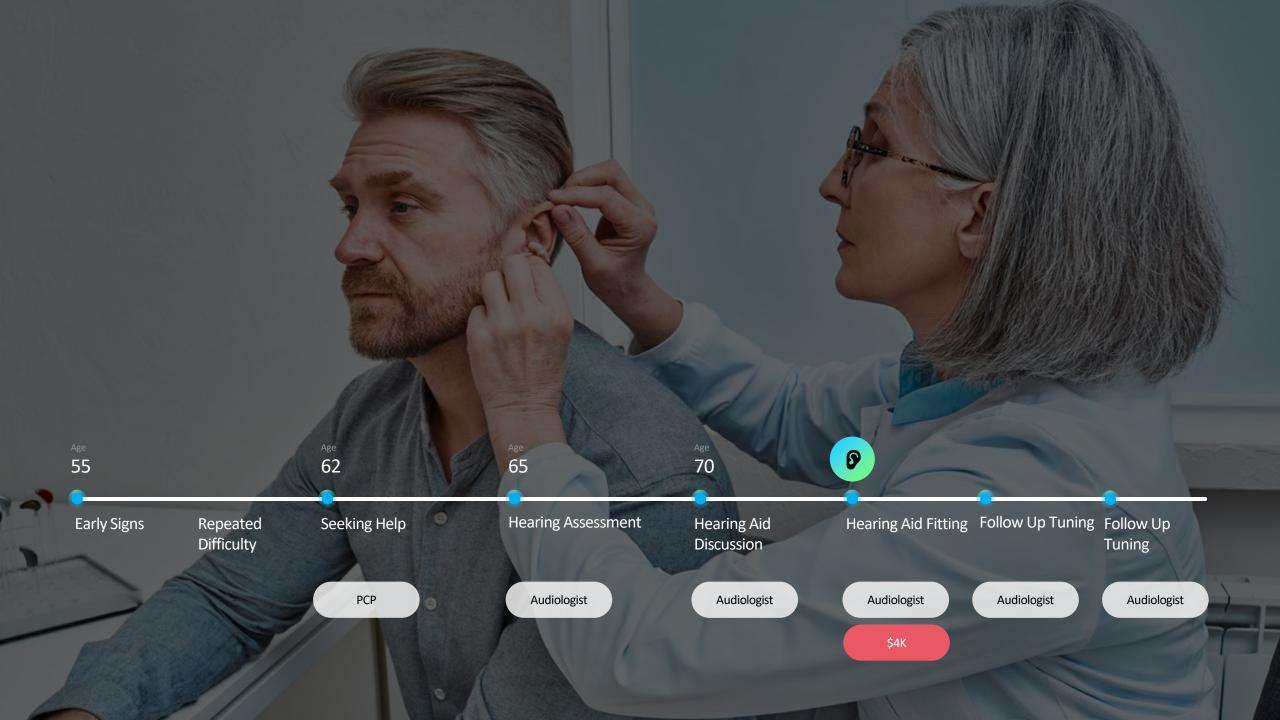


September 2024



Pioneering hearing.

You will soon have access to a Hearing Test,¹ a Hearing Aid feature,¹ and active Hearing Protection² using just AirPods Pro 2 and an iPhone or iPad. It's the world's first all-in-one hearing health experience — and it will be available with a free software update.





Hearing Health Transformation Over the Past Decade...



Evidence

Demonstrating the impact of hearing & hearing interventions on health



Regulatory policy

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Consumer Technology Association™



ANSI/CTA Standard

Four Frequency Pure Tone Average Testing Methodology and Hearing Wellness Reporting Metric for Consumer-Facing Hearing Solutions

ANSI/CTA-2118

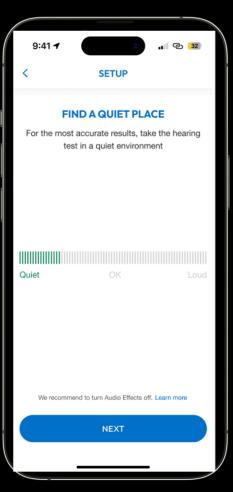
October 2023

January 2025

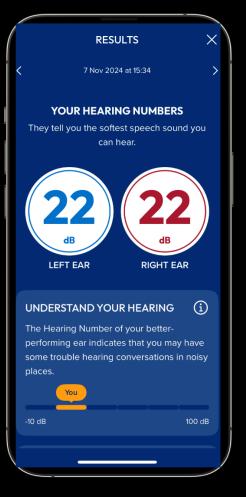
Official Launch of JHU Hearing Number App & Campaign



Test on iOS or Android in multiple languages



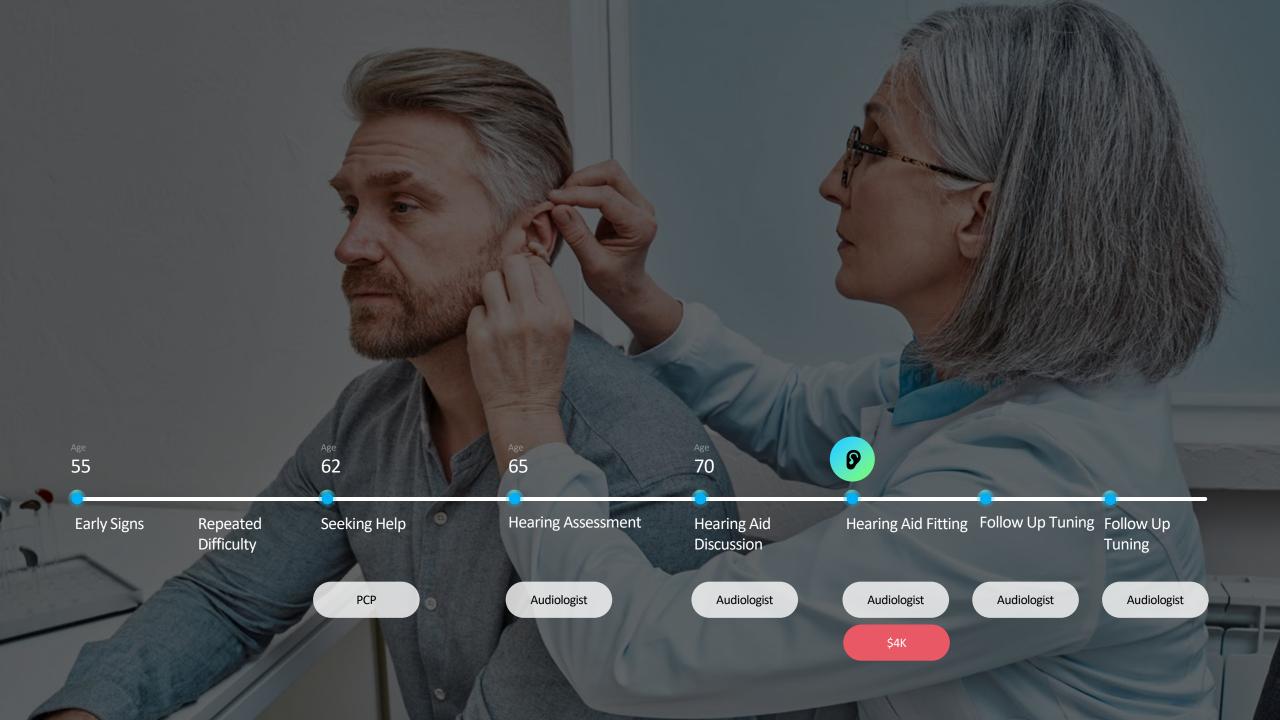
Meets CTA/ANSI PTA4 Standard



Actionable & Shareable Results



Available in 6 Languages





From Late-Life Hearing to Life-Course Hearing SONY **EssilorLuxottica** lexie Jabra GN Hearing Augmentation **Hearing Awareness Hearing Protection** Age 55 20 30

Hearing Health Transformation Over the Past Decade....



Evidence

Demonstrating the impact of hearing & hearing interventions on health



Regulatory policy

Allowing for innovation & accessibility in the hearing technology market



Awareness

of hearing as a life course aspect of health – JHU Hearing Number campaign Improving public health through empowering individuals to act on their hearing health over their life course

Frank Lin, MD PhD

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Johns Hopkins University

flin1@jh.edu

AchieveStudy.org
HearingNumber.org
JHUCochlearCenter.org

