

User-centered design of feedback regarding health-related behaviors derived from wearables: An approach targeting older adults and persons living with neurodegenerative disease DIGITAL HEALTH Volume 9: 1-15 © The Author(s) 2023 Article reuse guidelines: sagepub.com/journals-permissions DOI: 10.1177/20552076231179031 journals.sagepub.com/home/dhj



Karen Van Ooteghem¹, F Elizabeth Godkin¹, Vanessa Thai¹, Kit B Beyer¹, Benjamin F Cornish¹, Kyle S Weber¹, Hannah Bernstein², Soha O Kheiri³, Richard H Swartz^{4,5}, Brian Tan⁶, William E McIlroy¹ and Angela C Roberts^{7,8,9}

Abstract

Objective: There has been tremendous growth in wearable technologies for health monitoring but limited efforts to optimize methods for sharing wearables-derived information with older adults and clinical cohorts. This study aimed to co-develop, design and evaluate a personalized approach for information-sharing regarding daily health-related behaviors captured with wearables.

Methods: A participatory research approach was adopted with: (a) iterative stakeholder, and evidence-led development of feedback reporting; and (b) evaluation in a sample of older adults (n = 15) and persons living with neurodegenerative disease (NDD) (n = 25). Stakeholders included persons with lived experience, healthcare providers, health charity representatives and individuals involved in aging/NDD research. Feedback report information was custom-derived from two limb-mounted inertial measurement units and a mobile electrocardiography device worn by participants for 7–10 days. Mixed methods were used to evaluate reporting 2 weeks following delivery. Data were summarized using descriptive statistics for the group and stratified by cohort and cognitive status.

Results: Participants (n = 40) were 60% female (median 72 (60-87) years). A total of 82.5% found the report easy to read or understand, 80% reported the right amount of information was shared, 90% found the information helpful, 92% shared the information with a family member or friend and 57.5% made a behavior change. Differences emerged in sub-group comparisons. A range of participant profiles existed in terms of interest, uptake and utility.

Conclusions: The reporting approach was generally well-received with perceived value that translated into enhanced self-awareness and self-management of daily health-related behaviors. Future work should examine potential for scale, and the capacity for wearables-derived feedback to influence longer-term behavior change.

Corresponding author:

Creative Commons NonCommercial-NoDerivs CC BY-NC-ND: This article is distributed under the terms of the Creative Commons Attribution-NonCommercial-NoDerivs 4.0 License (https://creativecommons.org/licenses/by-nc-nd/4.0/) which permits non-commercial use, reproduction and distribution of the work as published without adaptation or alteration, without further permission provided the original work is attributed as specified on the SAGE and Open Access page (https://us.sagepub.com/en-us/nam/open-access-at-sage).

¹Department of Kinesiology and Health Sciences, University of Waterloo, Waterloo, ON, Canada

²Department of Nanotechnology Engineering, University of Waterloo, Waterloo, ON, Canada

³Department of Systems Design Engineering, University of Waterloo, Waterloo, ON, Canada

⁴Hurvitz Brain Sciences Program, Sunnybrook Research Institute, University of Toronto, Toronto, ON, Canada

⁵Department of Medicine (Neurology), Sunnybrook Health Sciences Centre, University of Toronto, Toronto, ON, Canada

⁶Rotman Research Institute, Baycrest Health Sciences, Toronto, ON, Canada

⁷School of Communication Sciences and Disorders, Western University, London, ON, Canada

⁸Department of Computer Science, Western University, London, ON, Canada

⁹Canadian Centre for Activity and Aging, Western University, London, ON, Canada

Karen Van Ooteghem, Department of Kinesiology and Health Sciences, University of Waterloo, 200 University Ave, Waterloo, Ontario, Canada, N2L 3G1. Email: kvanooteghem@uwaterloo.ca

Keywords

Wearable electronic devices, feedback, health behavior, self-management, neurodegenerative diseases, aged, communitybased participatory research, stakeholder participation, physical activity, sleep

Submission date: 25 January 2023; Acceptance date: 12 May 2023

Introduction

In the last decade, there has been tremendous growth in the development of wearable technologies ("wearables") for both consumer and research or healthcare applications. Despite some challenges to implementation, the use of wearables for clinical research and care is expanding as digital endpoints are validated.¹⁻³ In older adults and persons living with complex health conditions such as neurodegenerative disease (NDD), objective measures of health captured outside of clinic visits may be particularly relevant for guiding therapeutic development and treatment decisions.⁴⁻⁶ Within the field, sights are set on integrating information derived from wearables into clinical trials and healthcare to capitalize on objective, patient-generated, health information captured in daily life.7,8 These approaches also hold promise for encouraging active patient engagement and shared clinical decision making.⁹ Important strides have been made in wearable device and analytic development in support of physiological and behavioral monitoring during everyday activities,^{2,10} but despite these advances, there have been limited efforts to develop effective methods for sharing information derived from wearables ("feedback") with clinicians and patients, particularly those with complex health conditions. The absence of evidence-informed recommendations addressing "what, when and how" wearables-derived information should be shared remains a critical gap in the field.

In some use cases, information generated from wearables is relatively interpretable, and the utility of the feedback in clinical decision-making is well-established. This is the case for feedback that targets a specific clinical outcome such as information obtained from Holter or continuous glucose monitors.^{11–14} Many wearables, however, have been designed to monitor more general, lifestyle-related, health behaviors such as physical activity, sleep or periods of stress for which evidence in support of the utility of feedback is mixed.^{15–21} Varied results may be linked, in part, to differences in the timing of feedback delivery (e.g. immediate or delayed)^{22–25} or whether feedback is delivered with or without the support of an expert (e.g. healthcare professional and trainer).^{23,26–28} Evaluation outcomes used to examine the utility of feedback (e.g. improved health or positive change in health-related behavior, changes to self-efficacy or motivation)²⁹ may also impact conclusions regarding uptake and effectiveness. For example, older adults in an exercise group who received smartwatch-derived feedback plus cognitive behavioral counselling reduced sedentary behavior more than those in an exercise-only group.³⁰ However, a group of older adults enrolled in a lifestyle intervention program who received immediate feedback from a wrist-worn activity tracker about number of steps and calories burned showed no change in physical activity self-efficacy.³¹

To optimize uptake and effectiveness of wearables-derived feedback for lifestyle-related health behaviors and symptoms of disease, several practical considerations prioritizing the safe, appropriate and meaningful use of feedback must be considered, all of which extend current options afforded by many consumer and commercial devices. Of particular importance for accurate and reliable clinical reporting are including only those wearables-derived endpoints that have been validated for use in the population of interest¹ and which have clear guidelines for interpretation that allow comparisons with similar people^{22,28} and direct appropriate action.³² Commonly used consumer and commercial devices provide limited opportunity to tailor feedback to the individual because of constraints associated with proprietary methods used to derive and share data.³³ These restrictions impact the potential to provide meaningful feedback and are particularly problematic for older adults and clinical cohorts, since a "one size fits all" approach does not consider the impact of unique age- or disease-specific features on reported outcomes. For example, clinical features of Parkinson's disease (PD) such as reduced arm swing, shuffling gait or akinesia can impact established metrics characterizing physical activity and mobility.^{34,35} In these instances, algorithms developed in healthy adults are not likely to provide accurate, reliable or meaningful information to a person living with PD (e.g. Weber et al.³⁶).

Personalizing feedback to ensure relevance to the individual or a sub-group of individuals and developing a communication strategy that supports delivery of information to the target audience are also critically important for successful implementation.^{22–24,28,37} To ensure relevance, assumptions should not be made about what an individual will value, and end users (patients and clinicians) should be engaged in decisions surrounding feedback, so that the information shared is driven by what has been deemed important to personal or clinical decision-making.^{23,27,38} To ensure the safe and appropriate use of feedback by individuals, it is critical to have an effective communication strategy that supports interpretation of the data,²⁵ including contextualizing the results and clearly noting the data's limitations to safeguard against overinterpretation.^{22,26,28,39} Consideration must also be given to feedback methods that will optimize consumption by end users with diverse needs, capabilities and resources (i.e. format, structure, language level, etc.).^{22,25,27,39}

To address the need for accurate, meaningful and digestible information to guide personal and clinical decision making, we designed an iterative co-development and implementation project focused on providing feedback derived from wearables data for older adults and those with cerebrovascular disease (CVD) and NDDs including PD, Alzheimer's disease (AD), mild cognitive impairment (MCI), frontotemporal dementia (FTD) and amyotrophic lateral sclerosis (ALS). We view provision of feedback to these groups to be critical for two reasons: 1) the application of wearable technology is particularly relevant to these cohorts where monitoring health-related behaviors for purposeful action may positively impact health trajectory, health outcomes and overall quality of life;^{35,40–42} and 2) heterogeneity within these groups emphasizes the need for approaches that address individual differences in patient-generated data and individuals' capacity to access, understand and use feedback. The overarching principle of this work was to develop an approach for user feedback that was suitable for older adults and those with complex health conditions, and which did not rely on a particular device or technology. The latter was achieved by using raw data from the wearables to derive validated measures for our populations of interest, rather than relying on standard outputs from proprietary algorithms that can limit confidence in the data and constrain how the data are presented.

The primary objective of this study was to co-design and develop a user-centered approach for reporting outcomes from continuous monitoring with wearables to enhance individuals' self-awareness and self-management activities related to daily health-related behaviors. While this work targeted specific cohorts, the products of this work have broad applicability related to effective wearables datasharing for personal and clinical decision making.

Methods

The study was conducted in two phases (Figure 1). The development phase focused on an iterative co-design and development process with the goal of generating a person-centered feedback report targeting health behaviors, using data derived from wearable sensors. In the second phase of the project, the team evaluated the acceptability, useability and impact of the feedback report in a sample of older adults and persons living with CVD or NDD. The study was conducted within the Ontario Neurodegenerative Disease Research Initiative (ONDRI; ondri.ca), namely the Remote Monitoring in Neurodegenerative Disease study⁴³ (ReMiNDD) and the Health in Aging, Neurodegenerative Diseases and Dementias in Ontario (HANDDS-ONT) study (Figure 1).

Co-design objectives and approach. The goal was to design an end-user accessible feedback report, comprised of evidence-informed metrics regarding modifiable lifestyle risk factors⁴⁴ that could facilitate symptom awareness, self-management and behavioral change. Our co-design process also focused on determining optimized delivery methods for facilitating knowledge uptake and health behavior changes. Implementation and evaluation of the feedback report within the ONDRI studies were intended to solicit feedback from persons with lived experience as a first step toward a future, large-scale, pragmatic implementation trial.

Our approach consisted of iterative cycles of stakeholder input. Each cycle consisted of planning, consolidation, implementation and evaluation stages (Figure 2) where stakeholder perspectives were integrated into report versions following a consolidation process to confirm that proposed design features aligned with project objectives using established criteria (Table 1).

Theoretical framework. Methods were grounded in a participatory research approach⁴⁵ that was further informed by frameworks for health research partnerships^{46–48} emphasizing shared decision making and mutual benefit.⁴⁸ Participatory research uses methods and tools that ensure direct and meaningful collaboration with the people whose interests are the focus of the research. Our approach was tailored to the needs of older adults and emphasized the importance of accessibility and tailored engagement in the planning, relationship-building, co-development and knowledge translation phases of the project.⁴⁷

Stakeholder participants. Co-development of the report included participation by a broad range of NDD stakeholders. In addition to patient end users, stakeholders included those in the position to support changes in health behaviors through individual, health systems and community level actions. Specifically, stakeholders included persons with lived experience (patients and care partners), healthcare providers working across a spectrum of community and tertiary care centers, health charity advocacy/policy representatives, researchers, research funders, research program management and knowledge translation experts. These individuals were brought together under the umbrella of the ONDRI patient and community advisory committee (PCAC).



Figure 1. Timeline for feedback reporting development and evaluation activities.



Figure 2. Feedback reporting co-design and development approach grounded in a model of embedded stakeholder engagement.

Wearable sensor data used to derive feedback report *metrics*. Data included in the personalized reports were collected from ONDRI study participants who wore an array of biosensors 24-hours per day over the course of 7–10 days to monitor their health behaviors. The sensor set was comprised of inertial measurement units mounted on the wrist and ankle (unilaterally or bilaterally, study dependent), as well as a chest-worn sensor containing both a triaxial accelerometer and continuous electrocardiography. The sensor set and wear locations were chosen for their ability to evaluate multiple domains of health relevant to clinical care and self-management of disease in a comprehensive, unified manner.⁴³

Report generation via a purpose-designed analytics pipeline. The metrics included in the feedback report, along with summary data frames used for data review, and the report itself, were generated via a semi-automated, device-agnostic, analytics pipeline for wearables built by our group to produce both single- and multi-sensor outcomes. Currently, the pipeline accepts raw data from any inertial measurement unit or electrocardiographic device that provides access to raw data (termed "device-agnostic"). Briefly, raw sensor data were submitted to a series of quality assurance/quality control (QA/QC) procedures: preprocessed data were automatically checked for signal- and **Table 1.** Decision making criteria for feature inclusion into a health feedback report using data from wearable sensors.

Criteria	Description	Examples
User-focused	Priorities identified by cohort(s) and stakeholders of interest	Do the metrics reflect health domains important to end users and stakeholders? Are they captured in a way that reflects the lived experience of persons who have NDDs?
Scientific fit	Alignment with specific project objectives	Does the specific metric or feature achieve the project objectives and use case?
Evidence informed	Published support for use in cohort(s) of interest, available guidelines to	Is the metric evidence-informed? Is it clinically meaningful? Has it been clinically validated?
	support health changes	Has the evidence base resulted in published guidelines that can be used to anchor feedback?
Technical	Hardware, data pipeline and analytics capabilities	Is it feasible to derive the target metric or feature from the system hardware, sensor array and wear-locations deployed in the clinical/study protocol?
		Have published algorithms been validated for use in the cohort(s) of interest? If not, what is the state of readiness of custom analytics?
Implementation	Presentation, delivery and support	Can the metrics or features be summarized and made available to the user in an easily
		(continued)

Table 1. Continued.

Criteria	Description	Examples
		digestible and supportive way that promotes knowledge uptake and health behavior changes?

sampling integrity, and processed data were screened for device non-wear, multi-sensor synchronization and behavior classification using previously published and custom analytics to derive outcomes of interest based on the target population (e.g. older adults). Any QA/QC screen fails were flagged and resolved, and pipeline-produced summary data frames (Figure 3) were reviewed by trained analysts in the context of participant-specific demographic and clinical information, as well as participant-logged sleep and activity.

When identified, any additional flags were logged, and processing steps reviewed before determining the necessary action. Unresolved instances (i.e. true behavioral data) were addressed with customized feedback that was entered into the report on a per participant basis. Once the data review was complete, a personalized feedback report was generated.

Phase 1: feedback development

Cycle 1–*prototype report*

Planning and consolidation. In Spring 2019, a voluntary survey was made available at an ONDRI community outreach event for an audience comprised persons with lived experience, family care partners and persons from the community seeking information on dementia, CVD and NDDs. The survey asked respondents to identify potential targets for inclusion in a feedback report (see Appendix 1-Initial Survey). Five attendees completed the survey (three NDD, two care partners). Responses were used as a starting point and considered alongside our additional criteria for reporting (see Figure 2 and Table 1) to determine what outcomes to include in a prototype feedback report. To check the face validity of the prototype report, we triangulated its content with our stakeholder group (Figure 4). Stakeholder reflections on the prototype report were integrated as part of the consolidation process (Figure 2).

Wearables data collection and prototype report generation. The prototype version of the feedback report was implemented and evaluated as part of the ONDRI ReMiNDD study. ReMiNDD was conducted between May 2019 and March 2020 and included 39 participants living with CVD or any of the following NDDs as



Figure 3. Sample summary data frame generated for review as part of personalized feedback reporting. Data are included from multiple wearable sensors to integrate information about sleep (purple), gait (green), activity (yellow/orange), and sensor removal (grey), within and across days, for each participant. Each row represents a 24-hour period (days 1–7 stacked), from midnight to midnight. Duration of a specific behavior is represented by the width of the block.



Figure 4. Patient and Community Advisory Committee stakeholder representation. Clinicians include specialty, community-based, and allied health. Twelve members overlapped from prototype to version 1. Prototype development also included input from trial survey respondents with lived experience (n = 5; 3 NDD, 2 care partner) while draft version 1 was informed by ReMiNDD study respondents (n = 13 NDD). NDD: neurodegenerative disease.

diagnosed by a neurologist: PD, AD, MCI, FTD, or ALS (see Godkin et al.⁴³ for participant and protocol details including devices used). Importantly, all devices captured

and logged raw data, which were subscribed to a custom analytics pipeline to derive outcomes related to physical activity, sleep and mobility (i.e. outcomes generated via proprietary software provided by device manufacturers were not included in the report). At the time of enrollment, participants were told that their participation included receipt of a personalized feedback report based on data collected during a week-long period of sensor wear. The ReMiNDD study received ethics approval through the Sunnybrook Research Institute in Toronto, Ontario (REB approval: 007–2019), and all participants provided written, informed consent.

Preliminary implementation and evaluation. Prototype feedback reports were mailed to each ReMiNDD study participant approximately 3 months following completion of the in-home sensor wear portion of the study. Included in the report was a link to an accompanying video that guided participants through reading and interpreting their report. A one-page survey was appended to the back of each feedback report. Thirteen ReMiNDD study participants completed the voluntary survey, which was offered to all participants (n = 39) (see Appendix 1 for Prototype Report Survey and Results).

Cycle 2-version 1 report

Planning and consolidation. Data from Cycle 1 resulted in several changes being made to the prototype feedback report to create a revised version of the report for implementation into a clinical study context. Additional details were layered into the report including supplementary graphs and additional evidence-based measures to address the desire for more information based on Cycle 1 feedback (e.g. inclusion of sedentary behavior and nighttime walking). Changes embedded within the draft report version also addressed the need to account for known processing challenges in our groups of interest by including redundancies in information and ensuring readability by using multi-modal communication supports (e.g. written and visual) and reducing language complexity.^{49,50} All changes were included in draft form for subsequent review by research partners before Cycle 2 implementation.

In Fall 2021, the revised feedback report was submitted to the ONDRI PCAC (Figure 4). A guided focus group process was used to elucidate collective feedback on key areas for report implementation including the volume and type of information, readability for persons living with NDD and those within their circle of care, and overall format, layout and organization of the report. Specific themes that emerged from their feedback included the value of having participants understand why they were receiving feedback and what they might do with the information and led to the addition of content to describe the purpose and potential uses of the report. Further, suggestions from the group surrounding the potential need for participant support with interpretation and a desire to understand how our wearable sensor approach differs from commercial activity trackers led to the addition of a one-on-one virtual visit with each participant, conducted by a trained member of the study team, to review the feedback report (as described below). As in Cycle 1, PCAC feedback was integrated into Version 1 of the feedback report following consolidation processes.

Wearables data collection and version 1 report generation.

The Version 1 feedback report was implemented and evaluated as part of the HANDDS-ONT study (Figure 1); an Ontario-based, adaptive clinical trial⁵¹ in community-dwelling healthy adults (termed here as "controls") and persons living with probable or diagnosed CVD, PD, AD, MCI, FTD or ALS (termed here as neurodegenerative disease or NDD). HANDDS-ONT was designed as a fully remote study connecting clinical data with cellular, protein marker and genetic data, free-living behavioral data measured by wearables and longerterm health outcome data. Individuals were eligible to participate in HANDDS-ONT if: (a) they were 18 years of age or older, with self-reported English proficiency at 7/10 on a modified version of the Language Experience and Proficiency Questionnaire (LEAP-Q);⁵² (b) had internet or telephone access; (c) could attend a community-based laboratory to provide a blood sample; and (d) were under the care of a primary medical care provider and were willing to put forth their provider's name. Participants wore three devices continuously for up to 10 days: an inertial measurement unit on the dominant (or least affected) wrist and ankle (AX6, Axivity

Ltd, UK) and a chest device including combined electrocardiography and a tri-axial accelerometer (Bittium 360, Bittium Faros, FI). Similar to the ReMiNDD study, all devices captured and logged raw data, which were subscribed to a custom analytics pipeline to derive outcomes related to physical activity, sedentary behavior, sleep and mobility. The pipeline also generated the summary data frames for review (Figure 3) and the participant feedback reports. These outcomes were chosen for their relationship to chronic disease and brain health, their importance for self-management of health⁹ and their potential for early detection or monitoring of declines in health.9,44,53-56 Specifically, the report included the following parameters: daily volume and patterns of light and moderatevigorous physical activity and sedentary behavior, daily volume and pattern of stepping, step count accumulated in walking periods of different durations, total sleep time, duration of additional time in bed and patterns of nighttime behavior. See Figure 5 for details regarding the Version 1 report. The HANDDS-ONT study received ethics approval through Clinical Trials Ontario (Project ID: 3684) with Sunnybrook Research Institute in Toronto, Ontario as the Board of Record (REB approval: 2021–1517), and all participants provided written, informed consent.

Phase 2: version 1 report implementation and evaluation

Within the HANDDS-ONT study, feedback reports were returned to study participants in their preferred format (digital, hard copy or both). Given improvements in the automation of our personalized report processing pipeline, the target for turnaround from data offload to participant receipt of the report was 2–4 weeks, depending on format. Participants also received a 45-minute, one-on-one review of their report with trained study personnel via secure video conference. This study visit was structured to include the following:

- 1. Playback of a video, which highlighted key features of the report and oriented participants to the ways in which the data are presented (e.g. introduction to tables and figures).
- 2. Study personnel-led page-by-page review of participant data highlighting key takeaways from the report, including anchoring detected behaviors to personal activity or sleep log recordings (where applicable), and to published guidelines (tailored to age group). To ensure a standardized approach across participants, no specific recommendations related to participants' medical history or disease status were provided. Participants were encouraged to speak with their doctor before making health behavior-related changes and to share the report with others in their circle of care.

Supporting Materials		Data Presentation				
Orientation and Education	Glossary and Guidelines	es High-level Summary		Weekly Summary	Pattern of Behavior	
	<text><text><section-header></section-header></text></text>	What is balance measured Image: control of the second states of non-second second states of non-second second paperal activity Image: control of the second paperal activity	Guidelines II hours or loss per God Tomore and the second	Year Data" Annege: Ethiling or day under: Ethiling or day under: Ethiling or day Ethil 3 metalism out of kill days Annege: 6025 major and Annege: 7025 major and	<figure></figure>	A Pottern of Physical Activity and Sectoratory Balances are associated with the sector of the sector
Report Features	Information Provided					
 Multi-modal communication supports Designed for flexibility (modular) Focus on modifiable behaviors (physical activity, sedentary behavior, sleep, mobility) Glossary and list of guidelines Community resources (not pictured) Layers of information 		 Daily volume of light and moderate-vigorous physical activity and sedentary behavior Periods of sedentary behavior less than and greater than 1 hour Daily step count, step count accumulated in walks of different durations Length of longest walk, number of long walks per day Total sleep time, duration of additional time in bed Patterns of activity, sedentary behavior, sleep, nighttime behavior, and walking (per 24 hours) Periods of sensor removal 				

Figure 5. Sample pages from participant feedback reports. On the left, supporting materials including recipient orientation to tables and graphs, glossary with definitions of outcomes and a summary of published guidelines. On the right, samples showing layers of information from high level summaries through to detailed patterns of behavior. Along the bottom, a description of features and information provided within the report.

 Table 2. HANDDS-ONT participant demographics (survey data as of 20 May 22).

		Control (n = 25)	
Age (years)	72 (60-87)	71 (60-87)	77 (64-84)
Sex (n, % Female)	24 (60%)	19 (76%)	5 (33%)
English proficiency– speaking (0-10) ^a	10 (8-10)	10 (9-10)	9 (8-10)
English proficiency– understanding (0– 10) ^a	10 (8-10)	10 (8-10)	9 (8-10)
Ethnicity (n, % European)	38 (95%)	23 (92%)	15 (100%)
Education (n, % post-secondary)	35 (87.5%)	21 (84%)	14 (93%)
MoCA (0-30) ^a	25 (20-30)	25 (22-30)	25 (20-30)

Data are presented as median range unless otherwise noted. MoCA: Montreal Cognitive Assessment. ^aHigher score indicates greater functioning.

3. Time for the participant to seek clarity or to ask questions about the information contained within the report.

Two weeks after the feedback report review visit, a final study visit was conducted to gather HANDDS-ONT

participant perspectives on the report and probe whether behavior change and/or information sharing had occurred. Data were collected using a mixed methods approach, including both open-ended questions, and a coordinatoradministered survey with questions displayed on-screen. Any additional comments provided by the participant were recorded verbatim within free-text fields (see Appendix 1 for Version 1 Report Survey and details of development).

Data analysis

A convenience sample including the first forty participants aged 60 or over enrolled in the HANDDS-ONT study was used to evaluate version 1 of the feedback report (participation from 09 December 2021 to 20 May 2022). Descriptive statistics were used to summarize the data. Data were summarized as whole group responses and stratified by cohort (i.e. control versus NDD) and cognitive status (COGhigh = MoCA ≥ 26 ,⁵⁷ COGlow = MoCA $\leq 22^{58}$) using clinical data captured within the HANDDS-ONT study. Findings were also supplemented with illustrative examples akin to single subject analysis.

Results

Participants

HANDDS-ONT survey respondents included 25 control participants and 15 participants living with CVD or an NDD, who were sampled from the study's five cohorts as follows: seven CVD (one with MCI), four AD/MCI, and



Figure 6. (A) Summary of feedback report evaluation with respect to readability, volume of information, and helpfulness of information, with rows representing results as follows: Row 1-overall cohort (n = 40); Row 2-neurodegenerative disease (n = 15) versus control (n = 25); and Row 3-COGlow (n = 7) versus COGhigh (n = 19) for cognitive status based on MoCA scores (COGlow \leq 22, COGhigh \geq 26). (B) Summary of results probing impact of feedback (n = 40), specifically behavior change and information sharing (friends or family, and medical practitioner).

four PD. Table 2 summarizes participant demographic and clinical characteristics.

Feedback report evaluation

Results of the evaluation are shown in Figure 6. Most survey respondents (82.5%) found the report "somewhat" or "very easy" to read or understand, and 80% reported "just the right amount" of information/detail. Some differences emerged when data were examined by cohort

(NDD versus control) and by cognitive status (COGlow (n = 7) versus COGhigh (n = 19)), specifically with respect to ease of reading or understanding, which was deemed "somewhat" or "very easy" for a greater proportion of NDD (93%) versus control (76%) participants, and for a slightly greater proportion of COGlow (86%) versus COGhigh (74%) participants. In the context of understanding daily health-related behaviors (i.e. activity, mobility and sleep), 90% found the feedback report "very helpful" or "helpful," and 57.5% of participants made a behavior or

lifestyle change within 2 weeks. When data were examined by cohort or based on low and high cognitive status, all participants living with an NDD or in the COGlow group found the feedback "very helpful" or "helpful" compared to a slightly smaller proportion of control participants (84%) and participants in the COGhigh group (89.5%). Finally, differences emerged between COGhigh and COGlow groups with respect to the proportion who reported a lifestyle or behavior change (68% (n = 13) versus 29% (n = 2), respectively). Examples of lifestyle or behavior changes reported by participants included joining a Pilates class and striving to move more throughout the day (i.e. taking movement breaks), as well as increased awareness of good sleep habits and monitoring of sedentary behavior. Of those participants who did not make a lifestyle change (n = 17), 23.5% (n = 4) did not feel the need to make a change at the time of the survey, 11.8% (n = 2) had no interest in making a change at the time of the survey, 23.5% (n = 4) thought about and took steps towards making a behavior or lifestyle change (e.g. called to inquire about a fitness class) and 41.2% (n = 7) thought about it but did not take steps towards making a behavior or lifestyle change.

When asked about sharing with others in the participant's circle of care, almost all participants (92%) shared their report with a family member or friend, with the proportion of participants who shared with a family member or friend slightly higher within the cohort of persons living with an NDD (100%) compared to the control cohort (88%). Motivation for sharing with a friend or family member included wanting to engage partners in making lifestyle changes, educating and advocating for change in others, and self-advocacy inclusive of helping others understand a symptom or requesting specific support. When probed about sharing with a home care provider, personal support worker, or other paid caregiver, 97.5% (n = 39) of survey respondents reported not having this kind of support person within their circle of care. Within 2 weeks of receiving their feedback report, three survey respondents (all living with NDD) had shared the report with a healthcare provider, with one respondent scheduling an appointment specifically to discuss the report. Of those who did not share with a healthcare provider (n = 37), 35% (n = 13) intended to share at their next visit. Examples of specific motivation for sharing all or part of the report with a healthcare provider included wanting to share with a therapist to better inform symptom management and wanting to share with a family physician to solicit a referral for additional care (e.g. seeking an appointment with an occupational therapist). Reasons for not sharing with a healthcare provider included no concerns emerging from report (n = 14), perceived time constraints of appointments and/or outside of existing health priorities (n = 7), and feelings that information would not be relevant to their care (n = 2).

Although feedback regarding the report was generally positive, a range of profiles did emerge across participants from those who responded favorably across all dimensions (i.e. readability, volume, helpfulness, behavior change and information sharing) to those who reported mixed responses in terms of interest, uptake or utility. For example, one participant living with CVD indicated they found the report very easy to read or understand, and very helpful for understanding their daily health-related behaviors. This participant also made several changes to their behavior because of the feedback report, including increasing their daily activity levels by going up and down the stairs and starting to use a cane to increase balance confidence and facilitate more walking outdoors. This participant also shared the information with several people within their circle of care including a friend and family member, home care provider and multiple healthcare providers. When probed about motivation for sharing with a home care provider, this participant said, "I had things that I needed to discuss with them and couldn't always remember...[the report] gave more clarity for what I wanted to talk about and what I was asking for" and suggested that "[the] home care provider shared this information with her boss to show where I needed more support in particular areas." When probed about motivation for sharing with multiple healthcare providers, they said, "mainly because my memory and my ability to put words to what I want to talk about...the report helped keep my focus in the areas that I wanted to talk about" and "the report helped me get a referral to an occupational therapist...bringing the report to my occupational therapist will help her better understand where to start off." A control group participant indicated they found the report somewhat easy to read or understand and very helpful for understanding their daily health-related behaviors. They made changes to their lifestyle including a commitment to walking four times per week and commented on an increased awareness of the need to break up long sedentary periods with movement. When probed about motivation for sharing with a friend or family member, they said, "[I shared the report] to invite [my] neighbour to help me increase my walking." In contrast, another control participant indicated that they found the report somewhat easy to read or understand, somewhat helpful for understanding their daily health-related behaviors and reported that although they thought about it, they did not take steps towards making any behavior or lifestyle changes. This participant did not share the report with anyone in their circle of care, and when asked why they did not share, they said "[I] did not feel the need to share the report" with a friend or family member. In terms of reasons for not sharing with a medical practitioner, they said "[I] haven't spoken to [my] doctor in 2 years due to COVID but would likely share at my next physical."

Discussion

This study describes the co-design, development and evaluation of an approach to sharing information from wearables for the purpose of improving self-awareness and selfmanagement of health-related behaviors. Feedback included a personalized report delivered in a supportive environment following a week-long period of multi- sensor wear and application of custom analytics to the raw sensor data. Evaluation of the approach suggested the feedback was wellreceived by participants and resulted in enhanced selfawareness and self-management of health, with most participants reporting increased understanding of health-related behaviors, engagement in purposeful actions to change behaviors and/or sharing (or intention to share) information with others in their circles of care.

Critical to the development of the report was ensuring its acceptability and useability. As a group, most respondents found their report easy to read or understand, with an appropriate volume of information, and helpful for understanding daily health-related behaviors and/or making lifestyle changes. It should be noted that embedded supports (1:1 consultation and video supplement) may have influenced participant responses; however, evaluating the impact of each of these features was beyond the scope of this study. Some differences did emerge when participants were stratified by cohort or when subgroups of participants were extracted to examine outcomes with respect to cognitive status. A slightly higher proportion of participants living with an NDD rated the feedback report as helpful compared to control participants, and a larger proportion of respondents in the group with higher cognition scores reported making a behavior or lifestyle change compared to those with lower scores. Lower cognitive ability, however, did not appear to impact participant ratings regarding ease of reading and helpfulness or volume of information, suggesting that reports were accessible to all participants. However, it should be noted that the COGlow group was small, with only seven respondents. Examining individual participants revealed variability in reported uptake (behavior changes and information sharing), even within the same cohort. For example, two participants from the control cohort both indicated that the report was helpful and yet, differed in their intent to make lifestyle changes and interest in sharing with others in their circles of care. This further emphasizes the importance of developing specific methods of feedback reporting to consider the unique needs and abilities of recipients. That said, participants' largely positive responses to personalized feedback suggest that the process used to develop the content and methods for reporting should be considered a template for translating wearables data into meaningful and actionable outcomes. The current study examined short-term changes, or intention to change behavior, to investigate the potential for impact. There is need, however, to

examine the capacity for wearables-derived feedback to influence longer-term changes in health-related behavior. Such work requires examining behavior change in the context of established models (e.g. transtheoretical model of behavior change). Previous work suggests that maintenance of a behavior change (e.g. engagement in a physical activity program) declines over time,⁵⁹ even following interventions that employ multiple behavioral change strategies and span several months (e.g. >6 months).⁶⁰ Wearables-derived feedback may provide a solution to promote long-term behavior change, and future work should aim to understand the impact of feedback on maintenance of behavior when it is delivered longitudinally, at time points when people are most susceptible to reverting their behavior.

Our general approach and the questions used to guide development of the report were grounded in a desire to address the "what and how" of information sharing for outcomes derived from wearable devices, with particular emphasis on: (a) the types of information to include for the specific purpose of improving self-awareness and selfmanagement of health-related behaviors (the "what"); and (b) the method of delivery for older adults and persons living with NDD (the "how"). To this end, the goals of the report (self-awareness, self-management) and the specific needs and perceptions of the target groups were defined early, with intention, and were central to the development process, which included ongoing engagement of persons with lived experience. It should be acknowledged that the convenience sample (first 40 HANDDS-ONT participants) may have biased survey responses to include those from older adults and persons living with NDD who were keenly interested in receiving feedback, given that HANDDS-ONT study recruitment materials noted the inclusion of a personalized health and activity report. Further, although some differences emerged when respondents were stratified, these differences may have been greater, and the ability to interpret and use information may have been impacted if the sample included participants with a larger range of cognitive abilities and physical capacities and/or larger sub-group samples. Finally, since all participants received a 1:1 consultation and video supplement to aid in interpreting their data, it is possible that views on readability, helpfulness and volume of information would have been less favorable if the reports were delivered without these supports.

In addition to examining individual behavior change, the present study examined the potential for impact by probing the extent to which survey respondents shared or planned to share their data with others. Most respondents indicated sharing their report with a family member or friend and several shared or intended to share with a healthcare provider. It is important to note that the 2-week timeline for gathering participant responses specifically impacted the extent to which sharing occurred with healthcare providers,

with participants reporting an intention or plan to share at their next scheduled visit (outside of the study period). Some participants did not deem the information contained within the report important for sharing with a healthcare provider. When probed further, these responses appeared to be driven by "good performance" that was viewed as unnecessary for follow-up or the feeling that their healthcare provider was already aware of the types of findings contained within their report. The former raises the matter of ensuring that outcomes perceived to be positive by end users are consistent with the data and its link to risk, while the latter emphasizes the need to: (a) consider the target audience when making decisions about what information or measures to report; and (b) ensure that potential relevance of the information for clinical decision making is understood by the recipient. Other participants identified real or perceived barriers to sharing with a medical practitioner, including limited access or appointment time constraints, which may further support the need to have clinical stakeholders embedded in the feedback development process when working towards creating information sharing pipelines that are effectively and sustainably integrated into health systems/clinical care. The reasons for sharing, which included a desire to engage partners in making lifestyle changes and educating/advocating for others and self, suggest that information sharing about modifiable health-related behaviors has the potential to empower recipients and to positively impact awareness or behavior of others.

As noted, two important and pragmatic aspects of the development and evaluation process detailed within this paper include the availability of appropriate guidelines to anchor feedback and use of validated measures suitable for the context (e.g. clinical versus free-living behavior) and populations of interest. Guidelines have been established surrounding activity and sleep that have been linked to reduced chronic disease risk and preserved mobility and health in older adults.^{32,61} However, continued work is needed to establish disease-specific guidelines between free-living behaviors and health outcomes in persons living with CVD or NDD that account for specific symptoms and comorbidities. In addition to evidence-informed, disease-specific guidelines, it is also imperative that appropriate analytic methods are used to ensure validity of reported outcomes. For example, despite the ubiquitous use of wrist worn accelerometers to derive physical activity intensity estimates in free-living settings,⁶² this method may be unreliable in older adults or persons living with NDD due to age- and impairment-related changes in movement.³⁶ Instead, an approach that uses multiple body-worn sensors to resolve uncertainty,⁴³ as well as age- or disease-appropriate thresholds for activity intensity classification,63-65 can help to improve real-world activity estimates in these groups. Even where known features of a particular cohort (e.g. muted arm swing in PD) are accounted for, variability across individuals poses an DIGITAL HEALTH

additional challenge that requires person-level consideration of wearables-derived data. Within the current approach, this is addressed using case-based review of data prior to report generation as a quality assurance process.

Finally, there is critical need to address scale and sustainability of methods for providing feedback that accommodate the heterogeneity that exists in older adults and persons living with CVD or NDD (including diversity of background, education, data literacy and experience) and which considers the critical elements needed for successful patient and clinical implementation (i.e. end user education and uptake, data flow and health system integration that ensures privacy, respects patient autonomy, etc.). In this study, methods included 1:1 consultation with trained personnel, which has previously demonstrated success with respect to uptake of information²² and behavior change⁶⁶ but which is time and resource intensive. Consideration should be given to alternate approaches that meet the need for assistance with data interpretation while reducing demands, with sensitivity to the possibility that information contained within the report may cause an individual distress or that a delivery model that provides insufficient interpretive support could lead recipients to make decisions that pose a risk to their health or well-being. Such alternatives may include triaging individuals based on level of support required as determined by relevant health/clinical data, with additional methods including providing professional support on an as-needed basis, leveraging care partner support, and automating where possible, without compromising privacy, clarity of the information or access to resources. The report presented here has been designed to include layers of information and modules that can be inserted or removed as well as a video supplement as methods for managing scope, scale and sustainability.

As advances in this space continue, focus should remain on ensuring end-user needs are met by grounding work in co-design frameworks and processes. In the context of a knowledge-to-action (KTA) framework, the current study addressed several KTA phases including identifying the problem of being able to derive meaning from wearables data and considering barriers and facilitators to adapt traditional metrics and implementation processes for the purpose of positively impacting behavior change in our cohorts of interest.⁶⁷ In addition, where the goal of reporting is the use of wearables' data for specific *clinical* application (i.e. to guide clinical care), this template for development and implementation should be expanded to include clinician stakeholders who engage in co-design of feedback processes optimized for clinical decision making. Finally, unlike "what" and "how," the "when" of information sharing was not systematically investigated in the current study and warrants further attention to examine the optimal time course for delivery of feedback (point-in-time and longitudinal). Such investigations must

consider characteristics of the individual (e.g. disease severity, cognitive capacity, mental health status) and specific targets for intervention (i.e. self-, or clinical management of health or disease), and confront the likelihood that the optimal time course will change for individuals over the course of their disease, in addition to changes in the content and method of delivery.

Conclusion

This study provides a person-centered framework and template for reporting wearables-derived information that considers the needs and perspectives of older adults and persons living with NDD. The reporting system serves as a viable foundation to improve the utility of wearablesderived information in older adults and individuals with complex health conditions, including those with cognitive impairment, and has been built to accommodate continued development. A key feature of the approach is the use of raw device data, which permits tailored analytics to ensure relevant and accurate outputs for different cohorts and use cases. Continued work within our research group will result in a publicly available data processing pipeline for raw data processing and analysis by individuals with limited programming expertise. Continued analyses of data from HANDDS-ONT will enable us to better understand the needs of individuals and to describe the acceptability and utility of the report more richly by assessing responses from participants with a broader range of capabilities and grounding them in a published model of health behavior change. Future work will investigate methods for scaling the approach in a manner that preserves a commitment to personalization, as well as methods for implementing and evaluating feedback longitudinally to support behavior change.

Acknowledgements: The authors wish to thank ReMiNDD and HANDDS-ONT participants and study partners, as well as the Ontario Neurodegenerative Disease Research Initiative (ONDRI) patient and community advisory committee (PCAC) members, ONDRI investigators, ONDRI research coordinators and the ONDRI project management team for their input and administrative support of this project.

Contributorship: KVO, ACR, RHS and WEM conceived the study and were involved in protocol development and gaining ethics approval. KVO, FEG, BT, KBB, ACR and WEM developed the data analytics plan and assisted with data interpretation. VT, FEG, KVO and ACR developed feedback delivery supports and conducted literature review, data collection and data analysis. KBB, BFC and KSW led algorithm and analytics pipeline development and data processing and review for outputs included in feedback reporting. HB, SOK and KBB were involved in development of pipeline for feedback report generation. KVO, FEG and ACR wrote the first draft of the manuscript. All authors critically reviewed and edited the manuscript and approved the final version.

Declaration of conflicting interests: The author(s) declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

Ethical approval: The human research ethics committee of Sunnybrook Research Institute, Toronto, ON, Canada, approved both the ReMiNDD (REB approval: 007-2019) and HANDDS-ONT (REB approval: 2021-1517) studies, the latter which included the study protocol described within this manuscript.

Guarantor: KVO.

Funding: The author(s) disclosed receipt of the following financial support for the research, authorship, and/or publication of this article: This work was supported by the Ontario Brain Institute, an independent non-profit corporation, funded partially by the Ontario government. The opinions, results, and conclusions are those of the authors and no endorsement by the Ontario Brain Institute is intended or should be inferred.

Supplemental material: Supplemental material for this article is available online.

ORCID iDs: Karen Van Ooteghem D https://orcid.org/0000-0002-6743-3332

F Elizabeth Godkin D https://orcid.org/0000-0002-8470-2805 Soha O Kheiri D https://orcid.org/0000-0003-2409-7858

References

- Goldsack JC, Coravos A, Bakker JP, et al. Verification, analytical validation, and clinical validation (V3): the foundation of determining fit-for-purpose for Biometric Monitoring Technologies (BioMeTs). NPJ Digit Med 2020; 3: 55.
- Byrom B, Watson C, Doll H, et al. Selection of and evidentiary considerations for wearable devices and their measurements for use in regulatory decision making: recommendations from the ePRO consortium. *Value Health* 2018; 21: 631–639.
- Tung J, Gage WH, Van Ooteghem K, et al. Wearables at the point of care. In: Godfrey IA and Stuart S (eds) *From A to Z: wearables in modern medicine (1st edition). Exploring use and integration of wearables.* Cambridge, USA: Academic Press, 2021, pp.73–84.
- Patel S, Park H, Bonato P, et al. A review of wearable sensors and systems with application in rehabilitation. *J Neuroeng Rehabil* 2012; 9: 21.
- 5. Espay AJ, Bonato P, Nahab F, et al. Technology in Parkinson's disease: challenges and opportunities. *Mov Disord* 2016; 31: 1272–1282.
- Dorsey ER, Papapetropoulos S, Xiong M, et al. The first frontier: digital biomarkers for neurodegenerative disorders. *Digit Biomark* 2017; 14642: 6–13.
- 7. Dunn J, Runge R and Snyder M. Wearables and the medical revolution. *Per Med* 2018; 15: 429–448.

- Lavallee DC, Lee JR, Austin E, et al. Mhealth and patient generated health data: stakeholder perspectives on opportunities and barriers for transforming healthcare. *Mhealth* 2020; 6: 8.
- Jim HSL, Hoogland AI, Brownstein NC, et al. Innovations in research and clinical care using patient-generated health data. *CA Cancer J Clin* 2020; 70: 182–199.
- Khusainov R, Azzi D, Achumba IE, et al. Real-time human ambulation, activity, and physiological monitoring: taxonomy of issues, techniques, applications, challenges and limitations. *Sensors (Switzerland)* 2013; 13: 12852–12902.
- Zimetbaum P and Goldman A. Ambulatory arrhythmia monitoring: choosing the right device. *Circulation* 2010; 122: 1629–1636.
- Galli A, Ambrosini F and Lombardi F. Holter monitoring and loop recorders: from research to clinical practice. *Arrhythm Electrophysiol Rev* 2016; 5: 136–143.
- Grunberger G, Sherr J, Allende M, et al. American Association of clinical endocrinology clinical practice guideline: the use of advanced technology in the management of persons with diabetes mellitus. *Endocr Pract* 2021; 27: 505–537.
- Li A and Brackenridge A. The role of continuous glucose monitoring in pregnancy. *Obstet Med* 2022; 15: 6–10.
- 15. Brickwood KJ, Watson G, O'brien J, et al. Consumer-based wearable activity trackers increase physical activity participation: systematic review and meta-analysis. *JMIR Mhealth Uhealth* 2019; 7: e11819.
- Kirk MA, Amiri M, Pirbaglou M, et al. Wearable technology and physical activity behavior change in adults with chronic cardiometabolic disease: a systematic review and meta-analysis. *Am J Health Promot* 2019; 33: 778–791.
- Bravata DM, Smith-Spangler C, Sundaram V, et al. Using pedometers to increase physical activity and improve health: a systematic review. *JAMA* 2007; 298: 2296–2304.
- Singh B, Spence RR, Sandler CX, et al. Feasibility and effect of a physical activity counselling session with or without provision of an activity tracker on maintenance of physical activity in women with breast cancer—a randomised controlled trial. J Sci Med Sport 2020; 23: 283–290.
- Nguyen NH, Vallance JK, Buman MP, et al. Effects of a wearable technology-based physical activity intervention on sleep quality in breast cancer survivors: the ACTIVATE trial. J Cancer Surviv 2021; 15: 273–280.
- Browne JD, Boland DM, Baum JT, et al. Lifestyle modification using a wearable biometric ring and guided feedback improve sleep and exercise behaviors: a 12-month randomized, placebocontrolled study. *Front Physiol* 2021; 12: 1–15.
- Tang MSS, Moore K, McGavigan A, et al. Effectiveness of wearable trackers on physical activity in healthy adults: systematic review and meta-analysis of randomized controlled trials. *JMIR Mhealth Uhealth* 2020; 8: e15576.
- Cook S, Mayers S, Goggins K, et al. Assessing research participant preferences for receiving study results. *J Clin Transl Sci* 2020; 4: 243–249.
- Long CR, Stewart MK, Cunningham TV, et al. Health research participants' preferences for receiving research results. *Clin Trials* 2016; 13: 582–591.
- 24. Cox K, Moghaddam N, Bird L, et al. Feedback of trial results to participants: a survey of clinicians' and patients' attitudes and experiences. *Eur J Oncol Nurs* 2011; 15: 124–129.

- 25. Dirks LG, Avey JP, Hiratsuka VY, et al. Disseminating the results of a depression management study in an urban Alaska native health care system. *Am Indian Alsk Native Ment Health Res* 2018; 25: 62–79.
- Fernandez C, Kodish E and Weijer C. Informing study participants of research results: an ethical imperative. *IRB* 2003; 25: 12–19.
- Shalowitz DI and Miller FG. Communicating the results of clinical research to participants: attitudes, practices, and future directions. *PLoS Med* 2008; 5: e91.
- Wood J, Cotton SC and Gillies K. The relative importance of information items and preferred mode of delivery when disseminating results from trials to participants: a mixedmethods study. *Health Expect* 2022; 25: 419–429.
- Cheatham SW, Stull KR, Fantigrassi M, et al. The efficacy of wearable activity tracking technology as part of a weight loss program: a systematic review. *J Sports Med Phys Fitness* 2018; 58: 543–548.
- Roberts LM, Jaeger BC, Baptista LC, et al. Wearable technology to reduce sedentary behavior and CVD risk in older adults: a pilot randomized clinical trial. *Clin Interv Aging* 2019; 14: 1817–1828.
- O'Brien T, Troutman-Jordan M, Hathaway D, et al. Acceptability of wristband activity trackers among community dwelling older adults. *Geriatr Nurs* 2015; 36: S21–S25.
- Ross R, Chaput JP, Giangregorio LM, et al. Canadian 24-hour movement guidelines for adults aged 18-64 years and adults aged 65 years or older: an integration of physical activity, sedentary behaviour, and sleep. *Appl Physiol Nutr Metab* 2020; 45: S57–S102.
- Wright SP, Hall Brown TS, Collier SR, et al. How consumer physical activity monitors could transform human physiology research. *Am J Physiol Regul Integr Comp Physiol* 2017; 312: R358–R367.
- Kim DW, Hassett LM, Nguy V, et al. A comparison of activity monitor data from devices worn on the wrist and the waist in people with Parkinson's disease. *Mov Disord Clin Pract* 2019; 6: 693–699.
- Del Din S, Godfrey A, Mazzà C, et al. Free-living monitoring of Parkinson's disease: lessons from the field. *Mov Disord* 2016; 31: 1293–1313.
- 36. Weber KS, Godkin FE, Cornish BF, et al. Wrist accelerometer estimates of physical activity intensity during walking in older adults and persons living with complex health conditions: Retrospective observational data analysis study. *JMIR Form Res* 2023; 7: e41685. DOI: 10.2196/41685.
- 37. Wilkins CH, Mapes BM, Jerome RN, et al. Understanding what information is valued by research participants, and why. *Health Aff* 2019; 38: 399–407.
- Matias R, Paixão V, Bouça R, et al. A perspective on wearable sensor measurements and data science for Parkinson's disease. *Front Neurol* 2017; 8: –7.
- Purvis RS, Abraham TH, Long CR, et al. Qualitative study of participants' perceptions and preferences regarding research dissemination. *AJOB Empir Bioeth* 2017; 8: 69–74.
- Bonato P. Wearable sensors and systems. From enabling technology to clinical applications. *IEEE Eng Med Biol Mag* 2010; 29: 25–36.
- 41. Del Din S, Kirk C, Yarnall AJ, et al. Body-worn sensors for remote monitoring of Parkinson's disease motor symptoms:

vision, state of the art, and challenges ahead. *J Parkinsons Dis* 2021; 11: S35–S47.

- 42. Erb MK, Karlin DR, Ho BK, et al. Mhealth and wearable technology should replace motor diaries to track motor fluctuations in Parkinson's disease. NPJ Digit Med 2020; 3: 1–20.
- Godkin FE, Turner E, Demnati Y, et al. Feasibility of a continuous, multi-sensor remote health monitoring approach in persons living with neurodegenerative disease. *J Neurol* 2022; 269: 2673–2686.
- Livingston G, Huntley J, Sommerlad A, et al. Dementia prevention, intervention, and care: 2020 report of the Lancet Commission. *Lancet* 2020; 396: 413–446.
- Vaughn LM and Jacquez F. Participatory research methods choice points in the research process. *J Particip Res Methods* 2020; 1: 1.
- 46. Tittlemier BJ, Cooper J, Steliga D, et al. A scoping review to identify and describe the characteristics of theories, models and frameworks of health research partnerships. *Health Res Policy Syst* 2022; 20: 69.
- 47. McNeil H, Elliott J, Huson K, et al. Engaging older adults in healthcare research and planning: a realist synthesis. *Res Involv Engagem* 2016; 2: 10.
- Baquet CR. A model for bidirectional community-academic engagement (CAE): overview of partnered research, capacity enhancement, systems transformation, and public trust in research. *J Health Care Poor Underserved* 2012; 23: 1806– 1824.
- Grober E and Bang S. Sentence comprehension in Alzheimer's disease. *Dev Neuropsychol* 1995; 11: 95–107.
- Kim D, Mishra S, Wang Z, et al. Insidious effects of syntactic complexity: are ads targeting older adults too complex to remember? J Advert 2016; 45: 509–518.
- Dickson D, Johnson J, Bergan R, et al. The master observational trial: a new class of master protocol to advance precision medicine. *Cell* 2020; 180: 9–14.
- 52. Marian V, Blumenfeld HK and Kaushanskaya M. The Language Experience and Proficiency Questionnaire (LEAP-Q) assessing language profiles in bilinguals and multilinguals. J Speech Lang Hear Res 2007; 50: 940–967.
- 53. Warburton DER and Bredin SSD. Health benefits of physical activity: a systematic review of current systematic reviews. *Curr Opin Cardiol* 2017; 32: 541–556.
- Cable J, Schernhammer E, Hanlon EC, et al. Sleep and circadian rhythms: pillars of health-a Keystone Symposia report. *Ann N Y Acad Sci* 2021; 1506: 18–34.

- Phillips C. Lifestyle modulators of neuroplasticity: how physical activity, mental engagement, and diet promote cognitive health during aging. *Neural Plast* 2017; 2017: 3589271.
- Naismith SL, Lewis SJ and Rogers NL. Sleep-wake changes and cognition in neurodegenerative disease. *Prog Brain Res* 2011; 190: 21–52.
- Nasreddine ZS, Phillips NA, Bédirian V, et al. The Montreal Cognitive Assessment, MoCA: a brief screening tool for mild cognitive impairment. J Am Geriatr Soc 2005; 53: 695–699.
- Freitas S, Simões MR, Alves L, et al. Montreal cognitive assessment: validation study for mild cognitive impairment and Alzheimer disease. *Alzheimer Dis Assoc Disord* 2013; 27: 37–43.
- Marcus BH, Forsyth LH, Stone EJ, et al. Physical activity behavior change: issues in adoption and maintenance. *Health Psychol* 2000; 19: 32–41.
- Fjeldsoe B, Neuhaus M, Winkler E, et al. Systematic review of maintenance of behavior change following physical activity and dietary interventions. *Health Psychol* 2011; 30: 99–109.
- 2018 Physical Activity Guidelines Advisory Committee.
 2018 Physical Activity Guidelines Advisory Committee Scientific Report, Washington, DC, USA, 2018.
- Strain T, Wijndaele K, Pearce M, et al. Considerations for the use of consumer-grade wearables and smartphones in population surveillance of physical activity. *J Meas Phys Behav* 2022; 5: 8–14.
- Fraysse F, Post D, Eston R, et al. Physical activity intensity cut-points for wrist-worn GENEActiv in older adults. *Front Sports Act Living* 2021; 2: 579278.
- Jeng B, Cederberg KL, Lai B, et al. Step-rate threshold for physical activity intensity in Parkinson's disease. *Acta Neurol Scand* 2020; 142: 145–150.
- Tudor-Locke C, Mora-Gonzalez J, Ducharme SW, et al. Walking cadence (steps/min) and intensity in 61– 85-year-old adults: the CADENCE-adults study. *Int J Behav Nutr Phys Act* 2021; 18: 1–12.
- 66. Webb TL, Joseph J, Yardley L, et al. Using the internet to promote health behavior change: a systematic review and meta-analysis of the impact of theoretical basis, use of behavior change techniques, and mode of delivery on efficacy. *J Med Internet Res* 2010; 12: e4.
- Graham ID, Logan J, Harrison MB, et al. Lost in knowledge translation: time for a map? *J Contin Educ Health Prof* 2006; 26: 13–24.