

Feasibility of Measuring Smartphone Accelerometry Data during a Weekly Instrumented Up and Go Test after Emergency Department Discharge: A Prospective Observational Cohort Study

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Submitted to: JMIR Aging
on: February 20, 2024

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Abstract

Background: Older adults discharged from the emergency department (ED) face elevated risk of falls and functional decline. Smartphones may be a promising modality to conduct remote patient monitoring of mobility after ED discharge, yet its application in this context remains under-explored.

Objective: To assess the feasibility of using smartphones to measure weekly gait and balance data from older adults over an 11-week period after ED discharge

Methods: This single-center, prospective observational cohort study recruited patients aged 60 and older from an academic ED. Participants downloaded the GaitMate app to their iPhones, which recorded 3-axis accelerometer data during 11 weekly at-home "Up and Go" functional tasks. The study measured adherence to task completion, quality of transmitted accelerometer data, and participants' perceptions of the app's ease of use and safety.

Results: Out of 617 approached patients, 149 consented to participate, and 9 dropped out. Overall, participants completed 62% of study tasks. Data quality was optimal in 35% of submissions. At 3-month follow-up, 83% of respondents found the app easy to use, and 95% felt safe performing the tasks at home. Barriers to adherence included the need for assistance, technical issues with the app, and forgetfulness.

Conclusions: The study demonstrates moderate adherence to smartphone tasks for monitoring mobility among older adults after ED discharge, perceived ease of use and safety of completing at-home gait tasks. Identified barriers highlight the need for improvements in user engagement and technology design. Clinical Trial: NA

(JMIR Preprints 20/02/2024:57601)

DOI: <https://doi.org/10.2196/preprints.57601>

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Original Manuscript

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Conflicts of interest: The authors have no conflicts of interest to declare.

Funding: Department of Emergency Medicine, Stanford University

Word count: 4457

Tables: 1, Figures: 4

Supplemental material: Figures

Author contributions

BS originated the project and drafted the initial manuscript. All authors participated in interpreting the results, contributed to the writing of the manuscript, provided critical feedback to the manuscript, and approved the final manuscript draft for submission.

Abstract

Background: Older adults discharged from the emergency department (ED) face elevated risk of falls and functional decline. Smartphones might enable remote monitoring of mobility after ED discharge, yet their application in this context remains underexplored.

Objective: Assess the feasibility of having older adults provide weekly accelerometer data from an instrumented Up-and-Go (i.e. TUG) test over an 11-week period after ED discharge

Methods: This single-center, prospective observational cohort study recruited patients aged 60 and older from an academic ED. Participants downloaded the GaitMate app to their iPhones, which recorded accelerometer data during 11 weekly at-home "Up and Go" tests. The study measured adherence to TUG test completion, quality of transmitted accelerometer data, and participants' perceptions of the app's usability and safety.

Results: Out of 617 approached patients, 149 consented to participate, and 9 dropped out. Overall, participants completed 62% of TUG tests. Data quality was optimal in 35% of TUG tests. At 3-month follow-up, 83% of respondents found the app easy to use, and 95% felt safe performing the tasks at home. Barriers to adherence included the need for assistance, technical issues with the app, and forgetfulness.

Conclusions: The study demonstrates moderate adherence to smartphone TUG tests for monitoring mobility among older adults after ED discharge yet high usability and safety. Incomplete TUG test data was common, reflecting challenges in the collection of high-quality longitudinal mobility data in older adults. Identified barriers highlight the need for improvements in user engagement and technology design.

Keywords: older adult; falls; prediction; gait; sensors

Background

Each year, millions of older adults are discharged from emergency departments (EDs) across the United States¹. A growing body of evidence indicates that these individuals face high risks of adverse outcomes after ED discharge, including falls² and functional decline³. While guidelines aim to identify those at risk of poor outcomes⁴, existing fall risk screening tools using data at the time of the ED encounter have limited ability to predict which patients will fall².

One way to improve identification of older adults at risk for falls is to incorporate remote patient monitoring (RPM) of mobility into post-discharge care. Mobility, which includes gait and balance functions, requires the integration of sensory input, motor planning, and coordination. Gait alterations and balance issues are common in individuals 65 and older^{5,6} and both significantly increase the risk of falls⁷. RPM of gait and balance in home settings may identify mobility problems that are not readily apparent in controlled settings⁸. Additionally, RPM allows for examination of within-person changes over time, which can improve the discrimination of predictive models⁹. However, the success of any RPM depends heavily on the practicality and usability of the technology for older adults.

Numerous tools exist that allow for RPM of gait and balance, including external sensors (e.g., cameras, force plates) and wearable sensors (e.g. smartphones). Unlike external sensors, which require potentially expensive hardware and installation, wearable sensors are portable, cheaper, and in the case of smartphones, near ubiquitous¹⁰. Smartphones are equipped with inertial measurement units typically composed of an accelerometer and a gyroscope. These sensors enable smartphones to accurately monitor gait mechanics¹¹, which can identify individuals at higher risk of falls¹². Despite these capabilities, the potential of smartphone-based RPM of mobility after ED discharge remains largely unexplored. Describing and understanding the drivers of participants' engagement with RPM in research is necessary to determine the success of future real-world implementation of RPM in clinical services.

This study aimed to assess the feasibility of having older adults provide weekly accelerometer data from an instrumented Up-and-Go test¹³ over an 11-week period after ED discharge. The Up-and-Go test (commonly referred to as the timed Up-and-Go [TUG]) involves an individual getting up from a chair, walking forward, turning, returning to the chair, and sitting. The TUG test was chosen because it is simple, quick, and evaluates several key risk factors, including gait and balance, in a single assessment. Instrumented TUG tests using body-worn sensors can identify distinct gait patterns and balance issues¹⁴ validated against standard kinematic measures¹⁵, and can distinguish between individuals who have experienced falls and those who have not¹². To our knowledge, no prior study has reported on adherence to at-home TUG tests.

The primary focus of this study was on adherence, defined as the degree to which the user followed the program as designed¹⁶, which involved completing weekly at-home instrumented TUG tests. Secondary aims focused on data quality, app usability, safety during at-home functional tasks, and barriers to adherence. Data quality is essential for generating meaningful gait and balance features, and various user-specific factors can negatively impact it^{17,18}. Findings from this study provide foundational information for developing age-friendly RPM technologies, anticipating the increasing demand for improved post-discharge transitional care among older adults in the coming years.

Methods

Study design: This study was a single-center, prospective observational cohort study of ED patients. It was approved by the local institutional review board, and written informed consent was obtained from all participants.

Study Setting and Participants: A convenience sample of patients was recruited by research associates (RAs) from a single academic ED with an annual volume of 100,000 visits. Patients were eligible for participation in the study if they were 60 years of age or older, were to be discharged home, and owned an iPhone. We excluded from the study patients currently living in a nursing

home, patients with limited English proficiency, patients who could not walk unaided (i.e. walking without an assistive aid), and patients without the capacity to provide informed consent. If a participant moved to a nursing home after the time of consent, but during the follow-up period, they remained in the study. Compensation of up to \$90 was offered to study participants. The study is reported in accordance with the Strengthening the Reporting of Observational Studies in Epidemiology (STROBE) Statement: guidelines for reporting observational studies.

Study Procedures:

Overview: In the ED, RAs helped participants download the GaitMate app and led them through the self-report and functional baseline assessments. For 11 weeks after ED discharge, patients were asked to complete a weekly TUG test and to report any falls through the app. At 12 weeks after discharge/enrollment, we attempted to reach all participants by phone to collect data on perceived ease of app usage and safety during home TUG tests.

Onboarding & baseline assessments: In the ED, each participant was guided by an RA to download the GaitMate from the Apple App Store and was assigned a unique ID they used to access the app thereafter. Next, participants were led by the RA through baseline survey questions. Subsequently, participants were presented with an in-app instructional video detailing the TUG test¹⁹. The RA then demonstrated the TUG procedures and helped participants complete the first task in the ED which involves the participant standing up from a seated position, walking two meters, turning around, returning to the chair, and sitting down. To augment the assessment, we provided participants a waist belt equipped with a pouch to securely hold the smartphone during the TUG test. This setup enabled us to collect 3-axis accelerometer data from the phone, positioned near the body's center of mass, thereby allowing us to estimate spatial characteristics of steps. Participants who had difficulty placing the phone in the pouch and rotating to their back were instructed to keep it in the front. See **Supplement** for an example of the waist belt and phone placement. We chose a two-meter walking distance for the TUG instead of the original three meters

given that there is limited space in the ED to perform the TUG and concerns about the unobstructed space in patient's homes. After completing the task, participants removed the phone from the pouch and pressed the "DONE" button, prompting the transmission of the accelerometer data directly to institutional research servers for analysis. See **Supplement** for screenshots of the GaitMate app.

Home-based assessments: Each Sunday at 12pm for 11 weeks following ED discharge, participants received a GaitMate notification prompting them to complete their weekly TUG test. After entering their ID, they were instructed to tap the "Weekly Check-in" button. Participants then viewed an instructional video on the task and were asked to complete a safety checklist. This included verification of having cleared a walking space, having set up a chair to one side, having put on regular footwear and belt pouch, and having someone present to assist if needed. After completing the checklist, participants were asked to tap a "READY" button to start recording data. In addition, participants could log any fall by tapping the "Report a fall" button on the main screen, which would prompt the participant to record (1) date of a fall; (2) time of fall; and (3) injury associated with fall. At-home TUG test completion was monitored by RAs. When a participant missed three weeks in a row, RAs attempted to reach that participant by email once and then phone to probe barriers to completing at-home tasks with open-ended questions. If a participant lost their ID they were provided with contact information of study investigators.

Follow-up phone call: At 11 weeks following ED enrollment, a trained research associate called all participants by phone, making up to 3 attempts before marking the participant as lost-to-follow-up. Follow-up phone calls assessed falls over the study period and whether any occurred during at-home TUG tests, perceived ease of app usage and safety completing at-home TUG tests.

Measures:

Baseline assessments: To understand how ED patients who enrolled differ from those who did not, we collected limited information (i.e. age, sex, chief complaint, illness severity) on all pre-screened

patients. To understand baseline characteristics of our participants, we additionally recorded race, ethnicity, ED chief complaints, and active medical problems.

Functional task completion: The primary outcome was completion of the weekly TUG tests defined as any transmitted accelerometer data for a given week. The secondary outcome was accelerometer data quality, assessed by two RAs independently with the lead author serving as an arbiter when there was disagreement. All data quality assessors have extensive training and experience segmenting and generating gait and balance features using accelerometer data. Each task submission was classified as one of four categories: 1. *Optimal data quality*, defined as unambiguous visual segmentation of the data into the sit-to-stand, walk away, walk back, and stand-to-sit portions of the gait task. 2. *Minimal acceptable data quality*, defined as whether at least three steps during the walk away or walk back segment could be visually identified. 3. *Poor data quality*, defined as cases that did not meet category one or two but where some data was transmitted. 4. *Missing*, if no data was transmitted for active participants. **Figure 1** illustrates accelerometer data in the first three categories.

Follow-up assessments: To assess falls during the study period, we asked: "In the past 3-months, how many times have you fallen?" followed by "Did any of these falls result in injury?" and "Did any of these falls result in the need to seek acute medical care?" Finally, we asked: "Can you recall the situation that led to the fall?" To understand how participants perceived the GaitMate app, we asked: "How much do you agree or disagree with the following statements: 1. Overall, the app was easy to use. 2. I felt safe completing the gait task at home." Response options ranged on a five-point scale from strongly agree to strongly disagree. For simplicity, we collapsed the five-point scale into three categories: agree, disagree, or neither agree nor disagree.

Data Analyses: For our primary analysis, we first calculated the TUG test submission rates by week and by participant. We then categorized participants into low adherence (0% to 49%), moderate

adherence (50% to 99%), or perfect adherence (100%) and used ordered logistic regression models to examine whether the adherence was associated with participant characteristics of age, sex, race, active medical problems, and chief complaint category. In secondary analyses, we calculated the distribution of data quality categories for each week and by participant. To quantify older adult perceptions of GaitMate usability and safety, we calculated the percentage who agreed with the usability and safety statements. To understand barriers to adherence, we described qualitative feedback from participants when they reported difficulty with the app.

Results

Study enrollment and retention: **Figure 2** outlines the flow of patients from enrollment through follow-up. From 12/5/22 to 8/9/23, we identified 1059 ED patients from the medical record who were age 60+. We excluded 433 ED patients after discussion with ED providers, with the majority excluded were either because they were not being discharged to home or could not ambulate unaided. We approached 617 patients for screening. Among those approached, 468 (76%) were not interested in study participation, resulting in recruitment of 149 participants. Common qualitative reasons for non-participation included feeling too sick, too busy, or lack of interest. There were no statistical differences in patient age, sex, or ESI between those who agreed to participate and those who did not. A total of 9 participants (6%) dropped out of the study: four participants prior to completing baseline assessments, four more participant in week 1, and one participant in week 7. Follow-up assessments at week 12 were completed in 125/140 (89%) of retained participants.

Participant characteristics: Mean age of enrolled participants was 72 years, and the majority (61%) were men. Almost half of participants (46%) had fallen in the past year, indicating a fall-vulnerable cohort. The presenting ED complaints were highly varied with only 10% presenting for fall-related care. Participants had the co-morbidity profile expected of older adults, with 65% having high blood pressure, 41% having heart disease, and 52% reporting a past orthopedic surgery.

Task completion rates: **Figure 3** summarizes task completion and data quality over the course of the study. Among active participants, accelerometer data from 62% of weekly TUG tests was transmitted over 11 weeks post-discharge. The completion rates declined from 77% in week one to 57% in week eleven. 23% of participants completed all weeks of TUG tests; 19% completed no weeks of TUG tests; 57% of participants completed at least 6 weeks. Adherence was similar across sex, race/ethnicity, active medical problems, and chief complaint categories.

Data quality: Overall, 35.4% of submitted data was rated as optimal quality. The proportion of TUG tests with optimal quality declined from 46% in week one to 31% in week eleven. An additional 3% of submissions were rated as minimal acceptable quality which remained relatively stable for the duration of the study. Almost exclusively, the submissions classified as poor quality seemed to be from truncated samples (see Figure 1 for an example). There was a high degree of variability across participants, as shown in **Figure 4**. For example, 13% of participants had no weeks with optimal data quality and 13% of participants had all eleven weeks with optimal data quality.

Fall rates: During the study period, 27 participants (21.8%) reported falling at least once and 8 (6.5%) reported more than one fall. 13/27 (48%) of the falls were reported through the GaitMate app whereas the remainder were reported through phone follow-up. Among patients who fell, 13 (48.1%) reported a fall injury and 2 (7.4%) reported needing acute medical care for the fall. None of the falls occurred during the at-home TUG test.

Usability, Safety, and Qualitative feedback: Overall, 99/119 of respondents (83.2%) agreed that the GaitMate app was easy to use; 11/119 (9.2%) disagreed. A total of 114/120 respondents (95%) agreed that they felt safe completing the at-home TUG tests; only 1/120 (0.8%) disagreed. We identified several key barriers to completing at-home TUG tests. Several participants related that they missed TUG test submissions because they did not have someone present with them and so could not fulfill the pre-task safety checklist. Several participants reported that they were busy with

managing medical issues, and several reported losing or forgetting their ID. A few participants relayed that they did not notice the app notification delivered each Sunday or found the “Weekly Check-in” button inactive when they had a TUG test due. Finally, numerous participants stated that they accidentally tapped the “DONE” button when placing the phone in the waist belt pouch, thus prematurely ending the task for the week.

Discussion

This study explored the feasibility of collecting weekly accelerometer data during an instrumented TUG test from older adults after ED discharge using a custom iPhone app and the factors influencing adherence to this RPM technology. Our primary finding is that among a diverse cohort of older ED patients with fall risk, 62% of weekly at-home TUG samples were submitted, with declines over 11 weeks after discharge. We also found that less than half of submitted accelerometer data was of optimal quality. Together, these findings suggest that collecting high-quality longitudinal mobility data in older adults is challenging.

To our knowledge, this is the first report of adherence to at-home functional mobility assessments in community-dwelling older adults. The adherence found in this study is better than most prior studies using app-based remote assessments. For example, a systematic review of 99 studies examining adherence to mHealth apps, most of short duration and few including older adults, found an average adherence of 56%¹⁶. Comparing our findings to other RPM studies is difficult given the lack of reliable reporting on longitudinal engagement in prior work²⁰.

Our findings provide insights into technical and human factors that may have influenced protocol adherence. For the 19% of participants who completed no weekly tasks, there seemed to be issues around technological literacy that could not be overcome in study orientation or follow-up support phone calls. For the 58% with variable adherence, factors included needing assistance to perform tasks, technical issues with the app, forgetfulness, and acute health issues during the monitoring period. The GaitMate app prioritized safety by requiring participants to have another person present during task completion. However, this was problematic for the 27% who lived alone.

Other human factors influencing adherence included lost or forgotten user IDs, missed app notifications, and distracting acute health events. To meet regulatory concerns, we required re-entry of a unique ID for each app use, but several participants reported this as an undesired barrier to completion. Our finding that notifications were often missed suggests alternative modalities like text messaging might improve task completion rates²¹. Technical difficulties like premature TUG test closure due to accidental screen taps could be addressed by removing that design feature or requiring verification before ending. Addressing these barriers is crucial for boosting engagement and effectiveness of RPM interventions.

We also identified human factors affecting data quality, with the majority of poor samples likely due to inadvertent button presses prematurely terminating data logging. Future versions should consider an extra verification step before stopping data capture. Variability in quality may also indicate difficulties properly positioning/securing the smartphone in the waist-worn pouch. These findings highlight the need to carefully balance user experience with optimal data collection in RPM. Alternative wearable technologies like Fitbits or smartwatches could allow easier data gathering, but have other limitations around compliance, charging requirements and costs. Prior work shows older adults have high Fitbit adherence²², but current models lack the high-resolution accelerometry needed for detailed gait/balance analysis. Smartwatches are promising²³ but creating limited by the costs of these devices. Further, reliably extracting high-quality gait metrics from wrist-worn sensors is technically challenging given arm swing uncoupled from leg motion.

Limitations of our study include the reliance on a convenience sample from a single academic ED, which may limit the generalizability of our findings. Our sample was also younger and more male than typical ED fall patients²⁴. Additionally, the study's design did not allow for a comparison of adherence to other types of tasks and app designs, limiting the interpretation to Up and Go tasks with our specific app interface. Future research should aim to address these limitations by incorporating a more diverse participant pool, extending the follow-up period, and designing real-time analytics on gait and balance post-discharge to focus fall prevention efforts. Further exploration into personalized interventions and feedback mechanisms within RPM

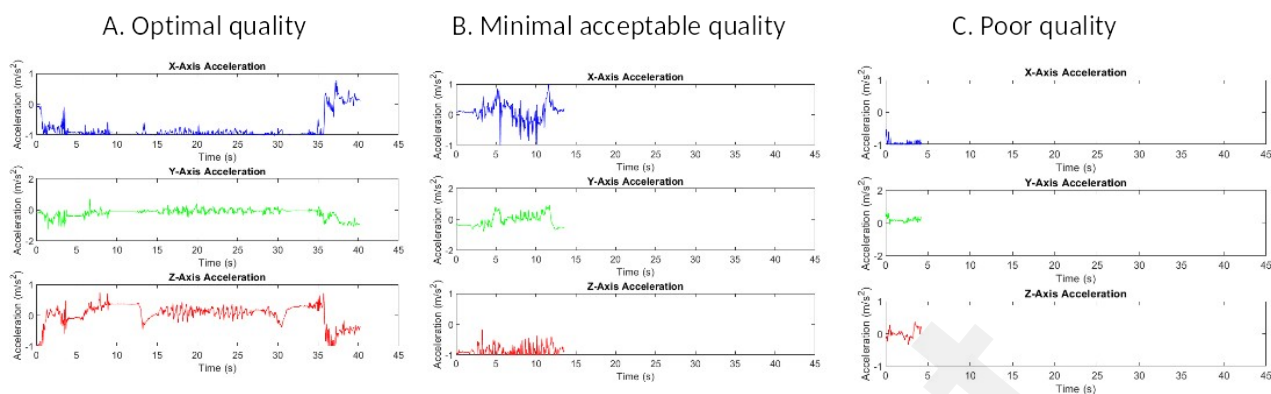
technologies and how to incorporate these into healthcare systems could also enhance patient engagement and adherence.

In conclusion, our study contributes to growing evidence on the potential utility of RPM in post-discharge care²⁵, offering insights into the practical challenges and user experiences for older adults in completing smartphone app-based functional mobility tasks at home. Addressing the human and technological barrier we identified can enable smartphone apps and RPM to play an important role in post-discharge care to reduce fall risk in older patients.

Table 1: Baseline Characteristics

Variable	Enrolled (N=149)	SD/ %
Demographics		
Mean Age	72.3	8.2
Female	58	39%
White, non-Hispanic	98	66%
Fall History		
Any fall in the past year	69	46%
ED Chief complaint category		
Cardiac	27	18%
Respiratory	5	3%
Gastrointestinal	20	13%
Neurological	31	21%
Genitourinary	6	4%
Fall	15	10%
Musculoskeletal	22	15%
Other	23	15%
Medical History		
Cardiac/heart disease	61	41%
Respiratory problems	26	18%
Gastrointestinal problems	47	32%
Vision conditions	29	20%
Endocrine conditions (e.g. diabetes)	51	35%
Motion sickness or vertigo	54	37%
High blood pressure	95	65%
Orthopedic surgeries	77	52%

Figure 1. Examples of Accelerometer Data Quality



Legend: In figure A (Optimal quality), there are clear sit-to-stand, walk away, walk back, and stand-to-sit portions of the gait task as well as displacements indicative of steps. In figure B (Minimal acceptable quality), there is a truncated segment with a grouping of at least three steps. In figure C (Poor quality), there is a severely truncated segment without three consecutive displacements indicative of steps.

Figure 2. Participant screening, enrollment, and follow-up.

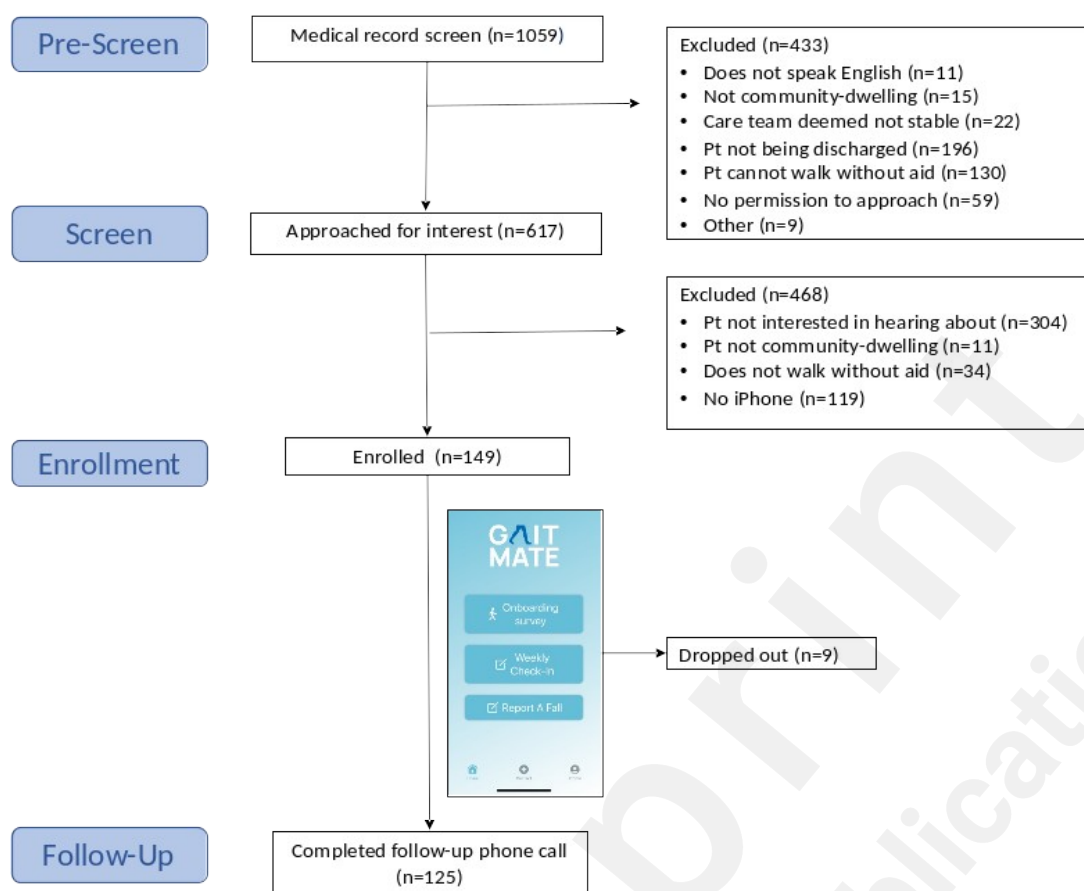


Figure 3. Gait data completeness and quality over study period.

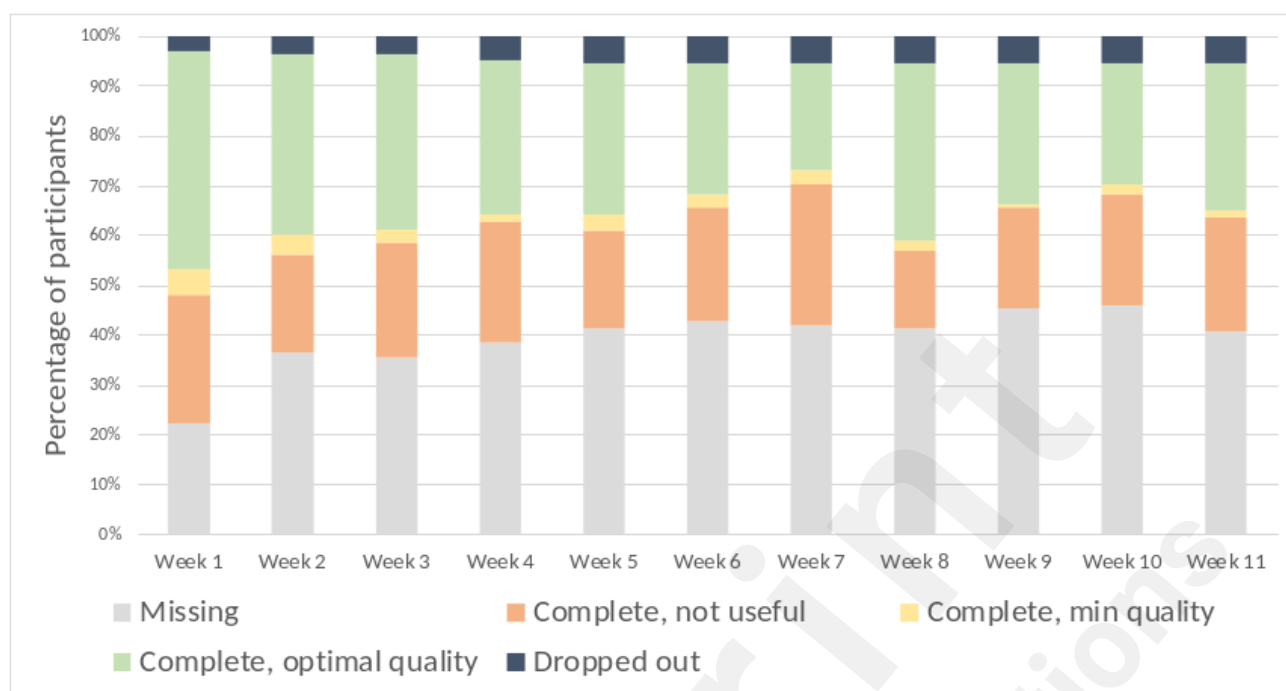
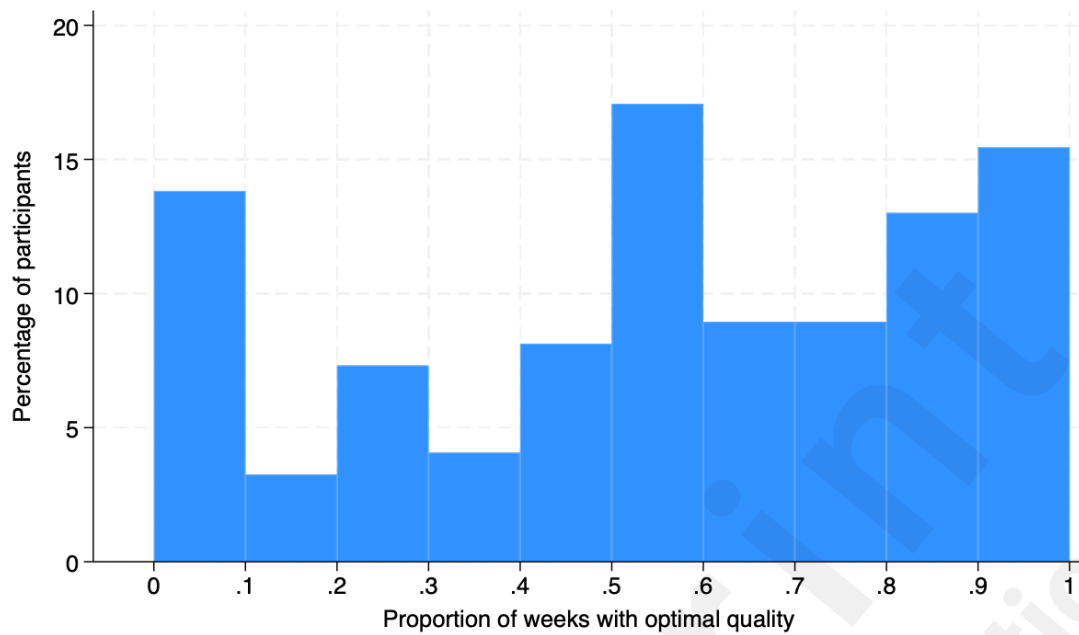
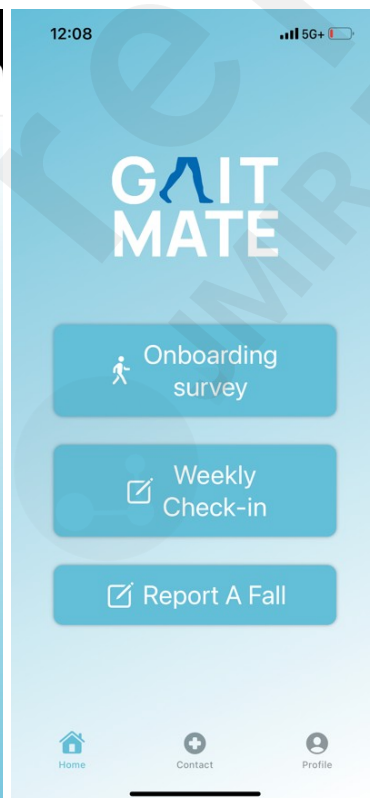
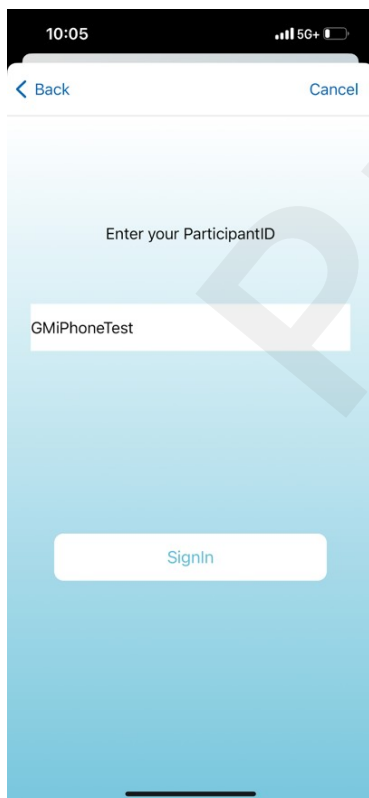
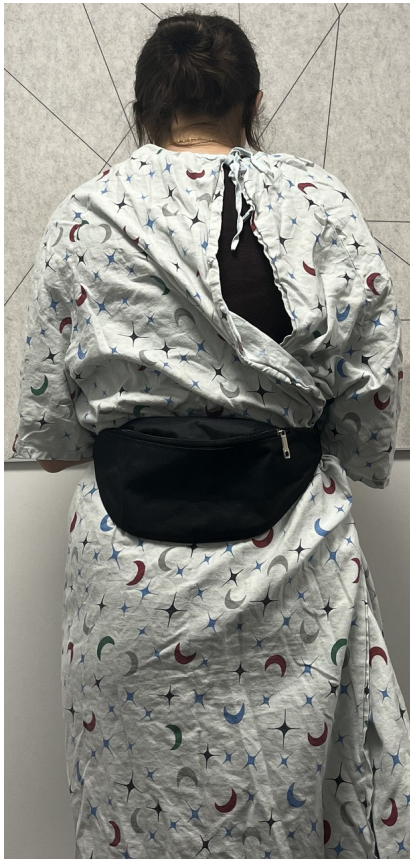


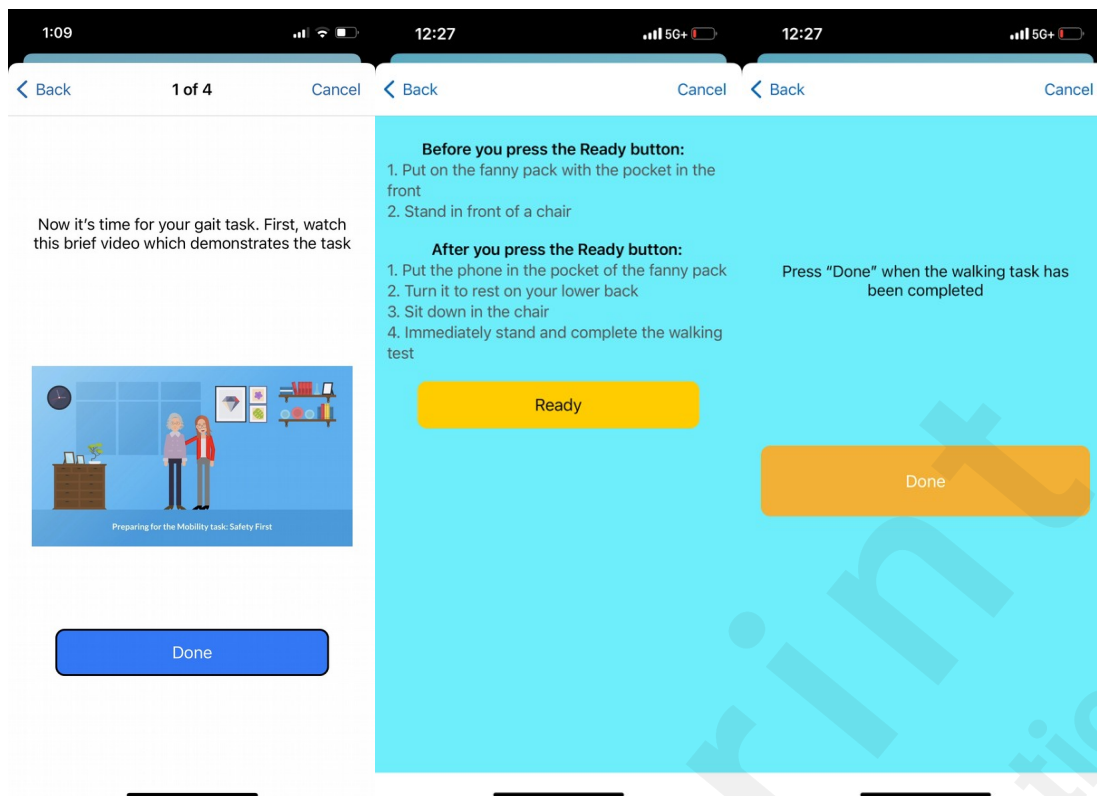
Figure 4: Distribution of participants by proportion of weeks with optimal data

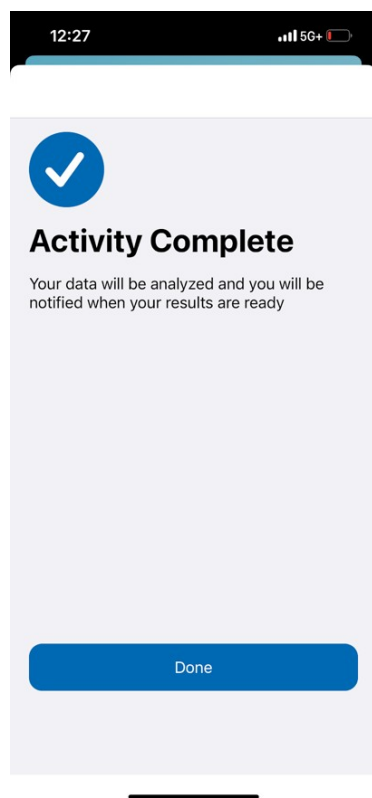


Supplemental Figures



1





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