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Smartphone assessment uncovers real-time relationships between depressed mood and daily functional behaviors after stroke

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Abstract

Introduction: The impact of depressed mood in daily life is difficult to investigate using traditional retrospective assessments, given daily or even within-day mood fluctuations in various contexts. This study aimed to use a smartphone-based ambulatory assessment to examine real-time relationships between depressed mood and functional behaviors among individuals with stroke.

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Author Contributions

Study concept: E JL, JML, DCM, and AWKW. Analysis and interpretation of data: QB, KJK, ECSM, MW MF, CLM, VP, and AWKW. Drafting the manuscript: QB, KJK, EGSM, and AWKW. Critical revision of the manuscript for intellectual content: E JL, JML, DCM, MW MF, CLM, VP, and SET. All authors read and approved the final draft of the manuscript.

Ethics approval

The study was approved by an Institutional Review Board (IRB) at the study sites. Informed consent was obtained from all individual participants included in the study.

Methods: A total of 202 participants with mild-to-moderate stroke (90% ischemic, 45% female, 44% Black) completed an ecological momentary assessment five times per day for 2 weeks by reporting their depressed mood and functional behaviors regarding where, with whom, and what activity was spent.

Results: Participants spent 28% of their wake-up time participating in passive leisure activities but spent the least time in physical (4%) and vocational (9%) activities. Depressed mood was concurrently lower when participants engaged in social activities ($\beta = -0.023 \pm 0.011$) and instrumental activities of daily living ($\beta = -0.061 \pm 0.013$); spent time with family members ($\beta = -0.061 \pm 0.014$), spouses ($\beta = -0.043, \pm 0.016$), friends ($\beta = -0.094, \pm 0.021$), and coworkers ($\beta = -0.050 \pm 0.021$); and were located in restaurants ($\beta = -0.068 \pm 0.029$), and homes of family members ($\beta = -0.039 \pm 0.020$) or friends ($\beta = -0.069 \pm 0.031$). Greater depressed mood was associated with worse ratings in satisfaction, performance, and engagement of activities in concurrent (β s = -0.036 ± 0.003 , -0.053 ± 0.003 , -0.044 ± 0.003) and time-lagged models ($\beta = -0.011 \pm 0.004$, -0.012 ± 0.004 , -0.013 ± 0.004).

Discussion: Smartphone-based ambulatory assessment can elucidate functional behaviors and associated mood after stroke. Findings support behavioral activation treatments to schedule social and instrumental activities for stroke survivors to reduce their depressed mood.

Keywords

Telemedicine; mobile health; smartphone; ambulatory assessment; ecological momentary assessment; stroke; depression; daily functional behaviors

Introduction

Stroke survivorship and its long-term consequences is a significant public health problem. Stroke is one of the leading causes of long-term disability worldwide.¹ Every year approximately 795,000 people in the United States experience a stroke. Of those 795,000 individuals, 610,000 are incident cases, and 185,000 are recurrent strokes. Depressed mood (DM) is the most common neuropsychiatric complication following a stroke, affecting a third of all stroke survivors. DM is associated with higher mortality rates, suicidal ideation, and cognitive deficits. It can persist multiple years after stroke, leading to a higher prevalence of long-term disabilities than those without depression² including reduced functional status, decreased social participation, and decreased quality of life.³ Despite its high prevalence, the etiology of poststroke DM is poorly understood and complex, involving multiple factors. Neurobiological factors such as a left-hemispheric lesion, cumulative lacunes in deep brain structures, and larger lateral and third ventricle to brain ratio have been associated with an increased likelihood of developing DM.⁴ As well as older age, female sex, genetics, history of depression, and diabetes mellitus. Psychosocial factors present during stroke recovery may also impact the occurrence of DM. Poor social support, divorce, and living alone increase the likelihood of developing DM.⁴ Stroke survivors report experiencing social marginalization, decreased self-esteem, and maladaptive psychological changes, such as passive coping and neuroticism.

Detailed insight into factors affecting the course of DM is needed to develop effective treatments. These efforts are impeded by the way depression is commonly measured via one-time psychiatric interviews or questionnaires, such as Hospital Anxiety and Depression Scale,⁵ and Patient Health Questionnaire, which query depressive symptoms retrospectively (e.g. past 2 weeks) making them more prone to recall bias.⁶ Bias may be amplified in instances where poststroke cognitive impairment may limit accurate recall. Moreover, traditional measures cannot capture dynamic variations in DM over time. Traditional measures produce a single score capturing the average level of DM in a given period. This single-score approach disregards that poststroke symptoms vary across or within days based on context (e.g. activities a person is engaged in and who they are with). Furthermore, traditional measures are administered in a controlled setting (e.g. lab or clinic), limiting the ability to compare dynamic variations in poststroke symptoms across real-world contexts.⁷ Thus, clinicians and researchers need more precise measurements to gain a deeper understanding of the variability and influence of DM in day-to-day contexts.

Technology-based ambulatory assessments, such as ecological momentary assessment (EMA) may overcome many limitations of traditional measures. EMA captures real-world data through repeated assessments administered via a smartphone. This method minimizes recall bias by measuring mood and daily behaviors in real time. It captures temporal and contextual variations by assessing study constructs multiple times a day within an individual's natural context. As such, a strength of EMA is the ability to capture intensive longitudinal data to study within- and between-subject variability of DM in daily life.⁸ EMA allows us to understand poststroke daily activity patterns which may provide novel insights into the etiology of DM and its impact on functional behaviors after stroke.

Despite these advantages, only a few studies have used EMA to study DM after stroke. One study supported the validity of EMA in measuring DM after stroke.⁹ Another EMA study showed that fewer social interactions but higher participation in sports and work activities after hospital discharge predicted higher DM at 3 months poststroke.¹⁰ Notably, both studies included a small sample ($n = 48^9$ and $n = 36^{10}$) and focused on the first 3 months poststroke, limiting generalizability to the broader poststroke population. In addition, Jean and colleagues¹⁰ measured a limited range of daily activities, highlighting a need for research to expand what is known about daily functional behaviors across multiple domains. The present study aimed to investigate DM related to daily functional behaviors during the chronic phase after stroke, i.e., at least 3-month poststroke. We used EMA to assess DM and time spent in and performance appraisals of daily activities. We investigated whether DM was predicted by activity type, locations, and social interactions. We hypothesized that higher DM was reported when participants stayed at home, alone, and did passive, sedentary activities. We also investigated if DM was momentarily related to perceived activity quality. We hypothesized that DM was higher during activities experienced as more difficult, less satisfied and engaged, and more help needed.

Methods

Participants

Participants were recruited between October 2018 and January 2021 using a hospital database. Inclusion criteria for the study included mild-to-moderate stroke defined by a National Institutes of Health Stroke Scale (NIHSS)¹¹ score ≥ 3 at the time of stroke onset, hemorrhagic or ischemic stroke, ≥ 3 months poststroke before enrollment, no or minor disability pre-stroke measured by a modified Rankin Scale (mRS)¹² score ≤ 2 , 18 + years old, and fluency in English. Exclusion criteria included previous co-morbid neurologic or psychiatric disorders, severe communication difficulty (Frenchay Aphasia Screening Test score <24), severe apraxia (Apraxia Screen of Test for Upper-Limb Apraxia score <6), visual inattention (Star Cancellation Test score <44), and visual acuity worse than 20/100 corrected version on the Lighthouse Near Visual Acuity test

We recruited 234 potential participants. Eight participants did not meet the eligibility criteria, and 14 participants dropped out before completing the study. Figure 1 shows a flow diagram of participant recruitment and completion. Two hundred and two participants completed the entire study protocol and had the EMA completion $\geq 30\%$. The institutional review board at the study sites approved this study. All participants provided written informed consent.

Procedures

At the initial lab-based visit, participants completed screening to ensure eligibility. Participants completed a 20-min individualized tutorial on how to use Status/Post, an EMA survey mobile application (app) using their iPhone or an iPod Touch device we provided. They completed a practice survey monitored by the examiner to assess any difficulties participants had with using the app or survey and were provided with hotlines to call should questions arise.

The EMA protocol started within 1 week of the initial visit. During the 14-day EMA period, participants completed five surveys per day between the hours of 8 a.m.–10 p.m., 11 a.m.–1 p.m., 2 p.m.–4 p.m., 5 p.m.–7 p.m., and 8 p.m.–10 p.m. On the first 2 days and periodically thereafter, the examiner called the participant to verify any questions completing the surveys at home. After the 14-day period, participants returned to the lab to return the device or delete the app from their iPhones. Participants were compensated up to \$125 for baseline and follow-up lab visits and EMA survey completed (maximum of 70 surveys).

EMA functioning survey

We developed an EMA functioning survey tailored to the unique needs of stroke survivors by modifying a validated EMA functioning survey.¹³ EMA surveys were predominantly checkbox or slider-format questions. Participants first reported their current location (home, workplace, restaurant, etc.). Next, participants reported with whom they were with (alone, family members, coworkers, etc.). The subsequent screens were customized to deliver home versus nonhome queries tapping a range of potential activities. Based on prior classification models,^{14,15} we categorized activities into domains (e.g. physical, cognitive,

social, activities of daily living (ADL), instrumental ADL (IADL), vocational, and passive leisure activities) and further grouped them into “total productive” versus “non-productive, sedentary” activities. After reporting what they were doing, participants rated the activity quality with the following questions: “How much help are you getting from someone else while doing this activity?”, “I am performing this activity well...”, “I am satisfied with doing this activity...”, and “I am engaged in doing this activity...”. Questions were answered on a 7-point scale from “no help/not well/not satisfied/not engaged” to “a lot of help/very well/very satisfied/very engaged”.

Participants also reported current levels of DM with the following question “Right now, I feel depressed...” with five possible response options (“not at all” to “very much”) during each EMA survey.

Statistical analysis

We used descriptive statistics to examine the study sample demographics. Multilevel models were used to investigate concurrent and lagged associations between DM and daily behaviors based on the time of day. EMA observations were thus modeled as level 1 (within-person level) units nested within individuals, who were modeled as level 2 (between-person level). The significance level was set at 0.05. We conducted two sets of models to examine (1) concurrent associations: either activity type, location, the person interacting with, or quality of activity as a predictor of same-survey DM and (2) lagged associations: either activity type, location, persons interacting with, or quality of activity as a predictor of next-survey DM. For time-lag associations, we included each predictor at the previous time point ($t - 1$) to predict the DM at the time point (t) and adjusted with the DM at a previous time point ($t - 1$). Activity quality variables were centered on creating person-centered deviation scores. The centered value indicated the momentary change (for moment-to-moment analyses) relative to each person’s own weekly average. Centering allows for examining within-person and between-person variance separately.¹⁶ We performed PROC MIXED in SAS version 9.4 to model between-person and within-person variances and account for the autocorrelation between adjacent observations. PROC MIXED is robust in handling missing data. This procedure does not omit the observations listwise, thus analyzing all data and assuming random missingness.¹⁷ We generated figures using R software, version 4.0.3.

Results

Sample characteristics

The EMA survey completion rate was 86.3% (14,140 surveys). We excluded 10 participants with low EMA adherence (<30%) from the analyses. The majority of included participants were men (55%), White (55%), married (52%), unemployed (57%), ischemic stroke (90%), and neurologic mild (75%) with the mean age of 59.7 years ($SD = 11.7$) (Table 1). Compared to the included participants, the excluded participants had lower total household income, higher DM, lower cognition, less mobile device usage experience, and required more mobility device use in the community.

Daily functional behaviors

Table 2 shows participants' daily functional behaviors. Participants mainly participated in passive leisure sedentary activities (28%), such as watching television and resting. Participants engaged in physical activities (3.6%) least, followed by vocational activities (8.6%). Participants spent most of their time at home (71%) and were primarily alone (27%) or with family members (17%). Regarding activity quality self-appraisal, participants rated their performance very well (mean = 5.19 [SD = 1.37]) and did not need much help (mean = 1.12 [SD = 2.04]). They were highly satisfied (mean = 5.16 [SD = 1.39]) and engaged (mean = 5.11 [SD = 1.45]) while doing activities.

Momentary DM

The average momentary DM was 0.24 (± 0.63). Participants experienced some DM in 16% of all observations (i.e. endorsed "A little bit" or higher in 1859 responded surveys).

DM and daily functional behaviors

Concurrent relationships between DM and daily functional behaviors.—Figure 2 shows relationships of DM with activity type, location, and the person interacting with. DM was reduced while doing productive activities ($\beta = -0.027 \pm 0.009$, p -value = 0.003), being in a location other than home ($\beta = -0.020 \pm 0.009$, p -value = 0.025), and being with others ($\beta = -0.050 \pm 0.010$, p -value < 0.0001).

Table 3 shows seven concurrent models, assessing granular relationships between each functional predictor and DM. For activity type (Figures 3a to 3c), DM was highest while doing passive leisure activity, mean DM = 0.266 ± 0.036 ($p < 0.0001$). Compared to the passive leisure, performing IADL and social activity had significantly lower mean DM scores ($\beta = -0.023 \pm 0.011$ ($p = 0.041$) and -0.061 ± 0.013 ($p < 0.0001$), respectively).

For the location, "at my home" was used as a reference with mean DM score = 0.252 ± 0.036 ($p < 0.0001$). When participants were "at the restaurant" ($\beta = -0.068 \pm 0.029$, $p = 0.022$), "at home of family member" ($\beta = -0.039 \pm 0.020$, $p = 0.047$), "at home of friends" ($\beta = -0.069 \pm 0.031$, $p = 0.026$), and "outside, other" ($\beta = -0.076$, ± 0.031 , $p = 0.013$), their mean DM score was significantly lower compared to "at my home". Mean DM score was significantly higher when patients were in the hospital ($\beta = 0.310 \pm 0.055$, $p < 0.0001$).

We used being alone as a reference with mean DM score = 0.279 ± 0.037 ($p < 0.0001$). Participants had significantly lower mean DM scores with "coworkers" ($\beta = -0.050 \pm 0.021$, $p = 0.019$), "family member" ($\beta = -0.061 \pm 0.014$, $p < 0.0001$), "friends" ($\beta = -0.094$, ± 0.021 , $p < 0.0001$), "other known people" ($\beta = -0.096$, ± 0.037 , $p = 0.010$), and "spouse or partner" ($\beta = -0.043$, ± 0.016 , $p = 0.007$). However, mean DM score was significantly higher while socializing electronically ($\beta = 0.174$, ± 0.052 , $p = 0.001$) compared to being alone.

Higher perceived performance ($\beta = -0.036 \pm 0.003$, $p < 0.0001$), satisfaction ($\beta = -0.053 \pm 0.003$, $p < 0.0001$) and engagement ($\beta = -0.044 \pm 0.003$, $p < 0.0001$) were associated with lower DM (Figure 4a to 4d).

Time-lagged relationships between DM and daily functional behaviors

After controlling for DM at the previous time point, current DM was significantly lower when participants were in a store ($\beta = -0.055 \pm 0.024$, $p = 0.019$) and socialized with people electronically ($\beta = -0.196 \pm 0.055$, $p = 0.0004$) at the previous time point (Table 4). Current DM was lower when participants had higher ratings in perceived performance, satisfaction, engagement, and help at the earlier time point.

Discussion

This study investigated the relationships of DM with daily functional behaviors among individuals with chronic stroke. Examining temporal associations between functional behaviors and DM using mobile health (mHealth) technology is a novel approach to understand mental health problems in stroke survivors. Our findings provide a framework for understanding DM in daily contexts after stroke, which enables progress toward developing effective treatments for DM and functional disability. We found that DM in daily life differed by activity type. Participants had the highest level of DM when engaged in passive leisure activities. Mentally passive sedentary behaviors, such as watching television, may contribute to an increased risk of depression.¹⁸ Yet, time-lagged analyses revealed that participation in passive leisure activities did not predict significantly higher DM at a later time when compared to other activities. This result suggests that passive leisure activities are a response to DM, rather than a contributor to DM.

We found that, compared with being alone, participants experienced lower concurrent DM while socializing with others, including their partner, family members, friends, and other known people. These significant associations between spending time with others and DM provide preliminary evidence to support clinical interventions promoting social activity for alleviating DM poststroke. An earlier EMA study supports this notion; increased perceived satisfaction, confidence, and success of social interactions were momentarily associated with reduced DM among mild stroke survivors.¹⁹ Similarly, we found that participants reporting lower DM were more likely to be in common locations for social gatherings, like restaurants, and homes of family members and friends. Interventions promoting in-person social interactions in poststroke life may warrant further investigation. Conversely, participants who socialized with people electronically experienced higher DM at the same survey but lower DM at the next survey. We cannot determine the reasons for this opposing relationship but a previous study of non-stroke populations found that participants engaging in online social media over more prolonged periods were at an increased risk of depression and other mental health problems.²⁰

Participating in IADLs was associated with lower DM, supporting previously published results that depression was associated with dependence on IADLs among stroke survivors.²¹ IADLs are complex activities needed to continue living in the community.²² It is possible that dependence on IADLs may induce feelings of helplessness, reduce self-esteem, and ultimately lead to depression.

Our results showed that reduced DM was associated with increased ratings in perceived satisfaction, performance, and engagement with daily functional behaviors among stroke

survivors in both concurrent and lagged models. These results align with the behavioral activation (BA) approach.²³ According to BA, behaviors and feelings influence each other. When people are depressed, they become less active. Likewise, when people do not engage in valued activities, they experience more depressed feelings. Our findings support that future intervention should focus on purposefully scheduling activities that stroke survivors perceive as satisfying, thereby reducing their DM. Interestingly, increased help needed for an activity predicted a decrease in DM at the next time point. The unexpected direction of this result warrants further investigation.

Prior literature has shown promising support for the application of telemedicine and mHealth technology in stroke care and management.^{24–26} This study adds to the literature by employing smartphone-based EMA to characterize the real-time relationships between DM and functional behaviors after stroke. To date, we use EMA as a research tool; however, it has potential clinical utility for monitoring poststroke functioning. Our findings support future interventions targeting structured social or community activities that evoke satisfaction from stroke survivors to reduce their DM.

Strengths and weakness

The use of real-time data in natural contexts, and the large and diverse sample are strengths of this study. However, there are several limitations. Most participants were community-dwelling and neurologically mild, which may not represent the general stroke population. Acute or severely impaired stroke survivors may show poorer functioning. Our sample was not diagnosed with poststroke depression but experienced a range of DM. Further research may include only patients diagnosed with mood disorders due to stroke as they may experience different daily life patterns. Sampling five surveys each day may miss activity episodes between consecutive surveys. Future research may consider using mobile sensing technologies to characterize daily functioning. Most current sensing research in stroke focuses on daily activity monitoring at home;²⁷ future studies integrating EMA with global positioning systems and accelerometers might expand this work to characterize activities in community settings. Assessing real-time associations between functional behaviors and mental health lays the groundwork for developing potential real-time interventions for improving DM and everyday functioning.

Conclusion

This study supports the use of smartphone assessment to examine DM and functional outcomes in real time among individuals after stroke. Our findings demonstrate the complexity of DM in daily life after stroke. Daily, repeated, and self-reported DM measured by smartphones may be clinically beneficial in detecting changes across day-to-day contexts. Future steps may integrate EMA with sensing technologies and behavioral strategies to deliver ecological momentary interventions to improve mental well-being and functional independence.²⁸

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Data availability

Data are available upon reasonable request. The data used as part of this study that support the reported findings are available from the corresponding author upon reasonable request.

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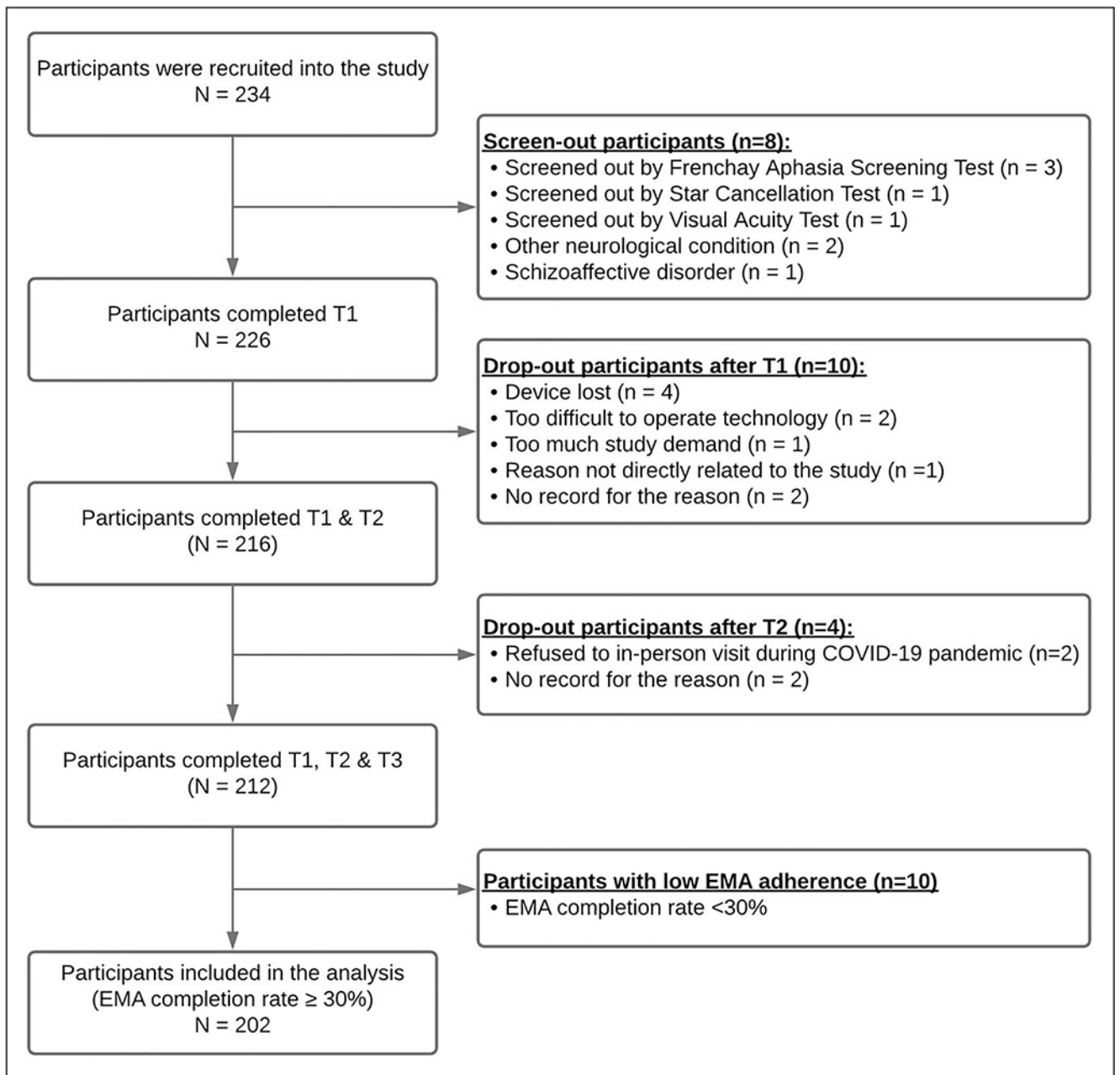


Figure 1.

Flow diagram showing participant recruitment and completion. T1: First lab visit, T2: 14 day EMA monitoring, T3: Second lab visit. EMA: ecological momentary assessment.

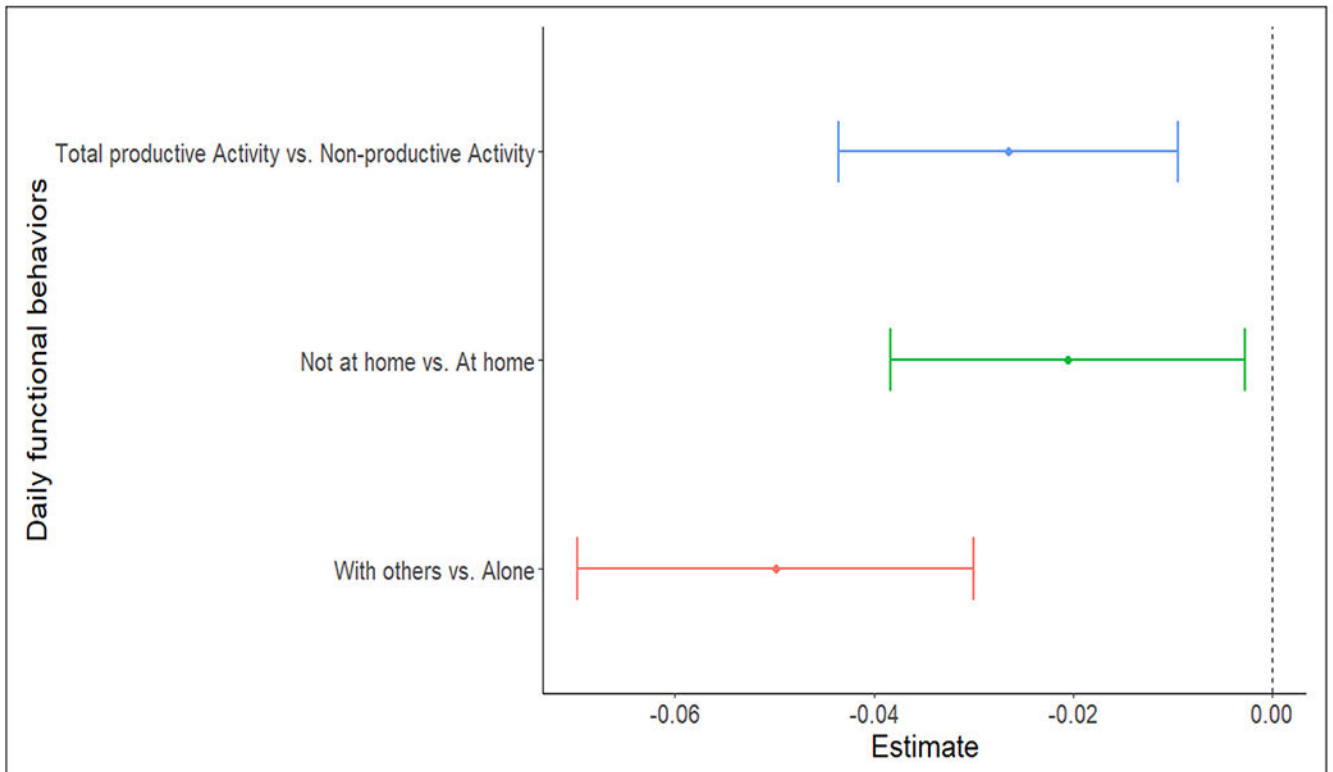


Figure 2.

Concurrent relationships of DM with dichotomized functional behaviors: type of activity (total productive activity vs. nonproductive activity), location (not at home vs. at home), and the person interacting with (with others vs. alone). The x -axis showed the estimated DM score. Dots represent the average DM scores, and whiskers represent 95% confidence intervals.

DM: depressed mood.

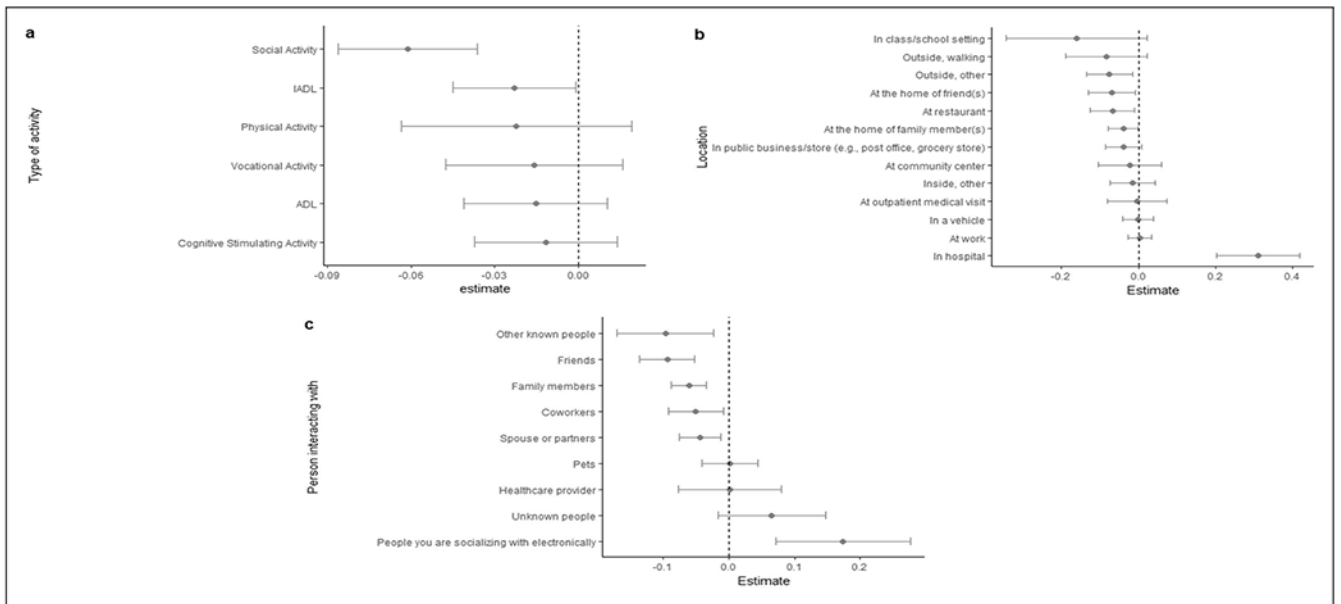


Figure 3. Concurrent relationships of DM with granular functional behaviors: (a) type of activity (reference: passive leisure, sedentary activity), (b) location (reference: home), and (c) the person interacting with (reference: alone). Dots represent the average DM scores, and whiskers represent 95% confidence intervals. DM: depressed mood.

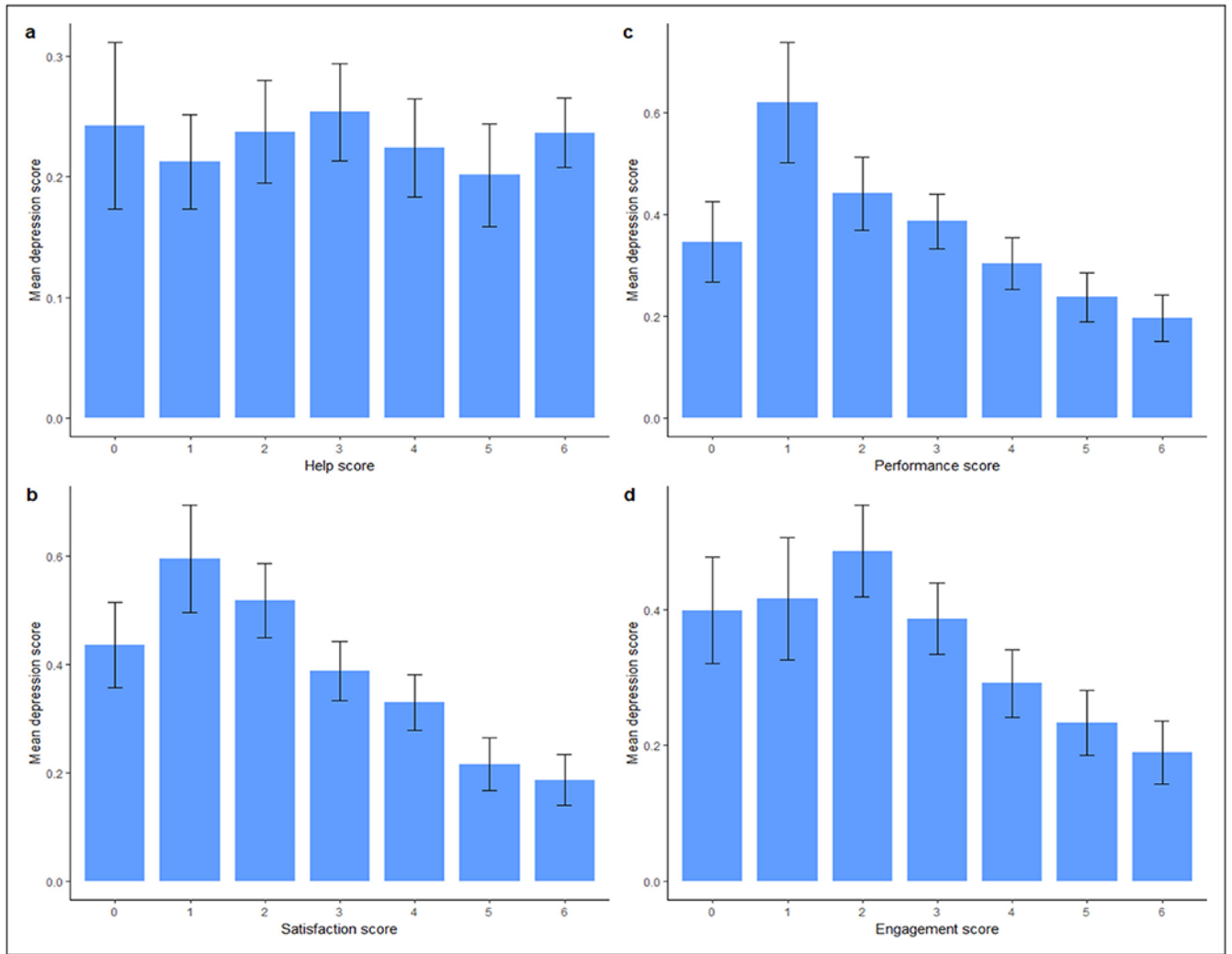


Figure 4. Average depression as a function of (a) help from someone while doing activities, (b) performance of activities, (c) satisfaction with doing activities, and (d) engagement of activities.

Table 1.

Clinico-demographic information of participants.

Variable	Overall, n = 212 ^a	Excluded participants, n = 10 ^a	Included participants, n = 202 ^a	P-value ^b
Stroke diagnosis				
Hemorrhagic	20 (9.4%)	0 (0%)	20 (9.9%)	0.603
Ischemic	192 (91%)	10 (100%)	182 (90%)	
Stroke side				
Bilateral	9 (4.3%)	0 (0%)	9 (4.5%)	0.278
Left	84 (40%)	3 (30%)	81 (40%)	
Right	83 (39%)	3 (30%)	80 (40%)	
Unknown	35 (17%)	4 (40%)	31 (15%)	
Handedness				
Both	1 (0.5%)	0 (0%)	1 (0.5%)	0.181
Left	28 (14%)	3 (30%)	25 (13%)	
Right	178 (86%)	7 (70%)	171 (87%)	
Premorbid disability (mRS)				
0	177 (84%)	8 (80%)	169 (84%)	0.188
1	20 (9.5%)	0 (0%)	20 (10.0%)	
2	14 (6.6%)	2 (20%)	12 (6.0%)	
Stroke severity (NIHSS)	3.23 (3.55)	3.50 (3.17)	3.21 (3.57)	0.549
Mean momentary depressed mood (DM)	0.24 (0.63)	0.46 (0.77)	0.24 (0.63)	<0.001
Time since stroke (days)	1449.33 (1025.35)	1639.00 (1378.75)	1439.95 (1008.29)	0.839
Age	59.85 (11.69)	62.10 (11.45)	59.74 (11.72)	0.459
Gender				
Female	93 (44%)	3 (30%)	90 (45%)	0.518
Male	119 (56%)	7 (70%)	112 (55%)	
Race				
Asian	1 (0.5%)	0 (0%)	1 (0.5%)	0.091
Black	97 (46%)	8 (80%)	89 (44%)	
White	114 (54%)	2 (20%)	112 (55%)	
Marital status				
				0.079

Variable	Overall, n = 212 ^a	Excluded participants, n = 10 ^a	Included participants, n = 202 ^a	P-value ^b
Married/Cohabiting	107 (50%)	2 (20%)	105 (52%)	
Separated/Divorced/Widowed	66 (31%)	6 (60%)	60 (30%)	
Single	39 (18%)	2 (20%)	37 (18%)	
Residential status				0.063
Alone	51 (24%)	5 (50%)	46 (23%)	
With others	161 (76%)	5 (50%)	156 (77%)	
Education years	14.13 (2.63)	12.80 (2.97)	14.20 (2.60)	0.069
Number of previous stroke	1.65 (1.76)	1.78 (1.09)	1.65 (1.78)	0.227
Employment status				0.195
Full time	57 (27%)	0 (0%)	57 (28%)	
Not employed	125 (59%)	9 (90%)	116 (57%)	
Part time	27 (13%)	1 (10%)	26 (13%)	
Volunteer	3 (1.4%)	0 (0%)	3 (1.5%)	
Total personal income				0.093
\$0-\$14,999	60 (28%)	6 (60%)	54 (27%)	
\$15,000-\$34,999	59 (28%)	1 (10%)	58 (29%)	
\$35,000-\$54,999	40 (19%)	3 (30%)	37 (18%)	
\$55,000-\$74,999	11 (5.2%)	0 (0%)	11 (5.4%)	
\$75,000 or more	42 (20%)	0 (0%)	42 (21%)	
Total household income				0.030
\$0-\$14,999	40 (19%)	5 (50%)	35 (17%)	
\$15,000-\$34,999	37 (17%)	2 (20%)	35 (17%)	
\$35,000-\$54,999	52 (25%)	3 (30%)	49 (24%)	
\$55,000-\$74,999	20 (9.4%)	0 (0%)	20 (9.9%)	
\$75,000 or more	63 (30%)	0 (0%)	63 (31%)	
Financial responsibilities				>0.999
Dependent, and living in residential treatment facility	7 (3.3%)	0 (0%)	7 (3.5%)	
Primary and partial responsibility	205 (97%)	10 (100%)	195 (97%)	
Social benefits				0.126
Commercial/Private Insurance	33 (16%)	0 (0%)	33 (16%)	
No benefit	30 (14%)	0 (0%)	30 (15%)	

Variable	Overall, n = 212 ^a	Excluded participants, n = 10 ^a	Included participants, n = 202 ^a	P-value ^b
SSB/Medicare/Medicaid	149 (70%)	10 (100%)	139 (69%)	
Mobility device use at home				0.149
Cane/Crutches/Walker	29 (14%)	4 (40%)	25 (12%)	
Manual wheelchair	4 (1.9%)	0 (0%)	4 (2.0%)	
None	176 (83%)	6 (60%)	170 (85%)	
Other	1 (0.5%)	0 (0%)	1 (0.5%)	
Power wheelchair	1 (0.5%)	0 (0%)	1 (0.5%)	
Mobility device use in community				0.011
Cane/Crutches/Walker	35 (17%)	3 (30%)	32 (16%)	
Manual wheelchair	4 (1.9%)	0 (0%)	4 (2.0%)	
None	168 (79%)	5 (50%)	163 (81%)	
Others	4 (1.9%)	2 (20%)	2 (1.0%)	
Power wheelchair	1 (0.5%)	0 (0%)	1 (0.5%)	
Global cognition function	25.86 (3.26)	22.10 (5.32)	26.04 (3.02)	0.017
Comorbidity	4.56 (2.87)	5.11 (2.26)	4.53 (2.90)	0.312
Mobile device use experience				0.023
No	6 (2.9%)	2 (22%)	4 (2.0%)	
Yes	203 (97%)	7 (78%)	196 (98%)	

^a n (%); Mean (SD).

^b Fisher's exact test; Wilcoxon rank-sum test.

mRS: modified Rankin Scale; NIHSS: National Institutes of Health Stroke Scale.

Table 2.

Distribution of EMA-measured functional behaviors and depressed mood (DM).

	EMA surveys ^a
Type of activity (14,140 EMA surveys)	
ADL	1419 (12%)
Cognitive stimulating activity	1467 (13%)
IADL	2505 (22%)
Passive leisure, sedentary activity	3258 (28%)
Physical activity	415 (3.6%)
Social activity	1523 (13%)
Vocational Activity	993 (8.6%)
Location (14,140 EMA surveys)	
At community center	93 (0.8%)
At my home	8298 (71%)
At outpatient medical visit	103 (0.9%)
At restaurant	186 (1.6%)
At the home of family member(s)	552 (4.7%)
At the home of friend(s)	181 (1.6%)
At work	1002 (8.6%)
In a vehicle	439 (3.8%)
In class/school setting	18 (0.2%)
In hospital	57 (0.5%)
In public business/store (e.g. post office, grocery store)	293 (2.5%)
Inside, other	187 (1.6%)
Outside, other	177 (1.5%)
Outside, walking	56 (0.5%)
Person interacting with (11,390 EMA surveys) ^b	
Alone	3118 (27.4%)
Coworkers	581 (5.1%)
Family members	1940 (17.0%)
Friends	441 (3.9%)

	EMA surveys ^d
Healthcare provider	118 (1.0%)
Pets	861 (7.6%)
Spouse or partners	1518 (13.3%)
Other known people	128 (1.1%)
Unknown people	91 (0.9%)
People you are socializing with electronically	64 (0.6%)
Help score^c, mean (SD)	1.12 (2.04)
Performance score^c, mean (SD)	5.19 (1.37)
Satisfaction score^c, mean (SD)	5.16 (1.39)
Engagement^c, mean (SD)	5.11 (1.45)
Depressed mood^d, mean (SD)	0.24 (0.63)

^aNumber of EMA surveys (%) unless otherwise specified.

^bIn the “Who is with you right now” question, participants could respond with more than one response. As participants might experience different depressive feelings across persons, we selected observations with a single response for data analysis.

^cIn the “How much help are you getting from someone else while doing this activity” question, participants rated on a 1 (no help) to 7 (a lot of help) scale. To match the semantic meaning, we converted it into the 0 (no help) to 6 (a lot of help) scale. In the “I am performing this activity well” question, participants rated on a 1 (not well) to 7 (very well) scale. To match the semantic meaning, we converted it into the 0 (not well) to 6 (very well) scale. In the “I am satisfied with doing this activity” question, participants rated on a 1 (not satisfied) to 7 (very satisfied) scale. To match the semantic meaning, we converted it into the 0 (not satisfied) to 6 (very satisfied) scale. In the “I am engaged in doing this activity” question, participants rated on a 1 (not engaged) to 7 (very engaged) scale. To match the semantic meaning, we converted it into the 0 (not engaged) to 6 (very engaged) scale.

^dIn the “Right now, I feel depressed” question, participants rated on a scale with 1: not at all; 2: a little bit; 3: somewhat; 4: quite a bit; and 5: very much. To match the semantic meaning, we converted it into a new scale with 0: not at all; 1: a little bit; 2: somewhat; 3: quite a bit; and 4: very much.

ADL: activities of daily living; EMA: ecological momentary assessment; IADL: instrumental activities of daily living.

Table 3. Multilevel models examining concurrent relationships between depressed mood (DM) and daily functional behaviors.

Models	Estimation	95% CI	p value
TYPE OF ACTIVITY			
<i>Fixed effect</i>			
Intercept	0.266	0.194 0.337	<0.0001
ADL	-0.015	-0.041 0.010	0.241
Cognitive stimulating activity	-0.012	-0.037 0.014	0.3693
IADL	-0.023 *	-0.045 -0.001	0.0403
Physical activity	-0.022	-0.063 0.019	0.2903
Social activity	-0.061 **	-0.086 -0.036	<0.0001
Vocational activity	-0.016	-0.048 0.016	0.3256
<i>Random effect</i>			
Intercept	0.254	0.203 0.306	<0.0001
Residual	0.151	0.147 0.155	<0.0001
LOCATION			
<i>Fixed effect</i>			
Intercept	0.252 **	0.182 0.323	<0.0001
At community center	-0.021	-0.103 0.061	0.6108
At outpatient medical visit	-0.004	-0.081 0.073	0.9204
At restaurant	-0.068 *	-0.125 -0.010	0.0216
At the home of family member(s)	-0.039 *	-0.078 -0.001	0.0467
At the home of friend(s)	-0.069 *	-0.129 -0.008	0.026
At work	0.003	-0.027 0.033	0.826
In a vehicle	-0.001	-0.040 0.038	0.9607
In class/school setting	-0.161	-0.343 0.022	0.084
In hospital	0.310 **	0.202 0.418	<0.0001
In public business/store (e.g. post office, grocery store)	-0.038	-0.085 0.008	0.1061
Inside, other	-0.015	-0.074 0.044	0.6251
Outside, other	-0.076 *	-0.136 -0.016	0.0127

Models	Estimation	95% CI	p value
Outside, walking	-0.084	-0.189	0.021
Random effect			
Intercept	0.255	0.196	0.314
Residual	0.151	0.147	0.155
PERSON INTERACTING WITH			
<i>Fixed effect</i>			
Intercept	0.279	0.207	0.352
Coworkers	-0.050 *	-0.092	-0.009
Family members	-0.061 **	-0.088	-0.033
Friends	-0.094 **	-0.135	-0.052
Healthcare provider	0.002	-0.076	0.080
Other known people	-0.096 *	-0.170	-0.023
People you are socializing with electronically	0.174 *	0.071	0.276
Pets	0.002	-0.041	0.045
Spouse or partners	-0.043 *	-0.075	-0.012
Unknown people	0.065	-0.017	0.147
Random effect			
Intercept	0.258	0.207	0.309
Residual	0.152	0.148	0.156
HELP FROM SOMEONE WHILE DOING ACTIVITIES			
<i>Fixed effect</i>			
Intercept	0.265	0.170	0.360
Help score (between-person)	-0.022	-0.077	0.032
Help score (within-person)	-0.002	-0.006	0.002
Random effect			
Intercept	0.244	0.195	0.292
Residual	0.131	0.127	0.135
PERFORMANCE OF ACTIVITIES			
<i>Fixed effect</i>			
Intercept	1.471	1.140	1.802

Models	Estimation	95% CI	p value
Performance score (between-person)	-0.240**	-0.303	-0.177 <0.0001
Performance score (within-person)	-0.036**	-0.042	-0.029 <0.0001
Random effect			
Intercept	0.190	0.153	0.228 <0.0001
Residual	0.130	0.126	0.133 <0.0001
SATISFACTION WITH DOING ACTIVITIES			
Fixed effect			
Intercept	1.571	1.261	1.880 <0.0001
Satisfaction score (between-person)	-0.262**	-0.321	-0.202 <0.0001
Satisfaction score (within-person)	-0.053**	-0.059	-0.046 <0.0001
Random effect			
Intercept	0.177	0.142	0.213 <0.0001
Residual	0.128	0.125	0.132 <0.0001
ENGAGEMENT of ACTIVITIES			
Fixed effect			
Intercept	1.363	1.066	1.661 <0.0001
Engagement score (between-person)	-0.223**	-0.281	-0.166 <0.0001
Engagement score (within-person)	-0.044**	-0.050	-0.037 <0.0001
Random effect			
Intercept	0.189	0.151	0.227 <0.0001
Residual	0.129	0.125	0.133 <0.0001

** p-value < 0.0001,

* p-value < 0.05.

ADL: activities of daily living; IADL: instrumental activities of daily living

Table 4. Multilevel models examining lagged relationships between depressed mood (DM) and daily functional behaviors.

Models	Estimation	95% CI	P-value
TYPE OF ACTIVITY			
Fixed effect			
Intercept	0.181	0.127	0.234 <0.0001
ADL	0.001	-0.026	0.027 0.952
Cognitive Stimulating Activity	0.003	-0.023	0.030 0.798
IADL	-0.011	-0.033	0.012 0.357
Physical Activity	-0.010	-0.052	0.032 0.635
Social Activity	0.002	-0.024	0.027 0.894
Vocational Activity	0.011	-0.021	0.043 0.506
Depression (previous time point)	0.275**	0.256	0.294 <0.0001
Random effect			
Intercept	0.135	0.107	0.163 <0.0001
Residual	0.137	0.133	0.141 <0.0001
LOCATION			
Fixed effect			
Intercept	0.182	0.130	0.234 <0.0001
At community center	-0.003	-0.085	0.080 0.952
At outpatient medical visit	-0.035	-0.114	0.044 0.386
At restaurant	0.027	-0.032	0.087 0.370
At the home of family member(s)	-0.006	-0.045	0.034 0.784
At the home of friend(s)	-0.003	-0.065	0.059 0.928
At work	-0.006	-0.037	0.025 0.712
In a vehicle	-0.020	-0.060	0.020 0.317
In class/school setting	-0.120	-0.323	0.083 0.248
In hospital	0.111	-0.008	0.230 0.068
In public business/store (e.g. post office, grocery store)	-0.055*	-0.102	-0.009 0.019
Inside, other	0.017	-0.044	0.078 0.580
Outside, other	0.007	-0.056	0.070 0.826

Models	Estimation	95% CI	P-value
Outside, walking	-0.024	-0.130	0.082
Depression (previous time point)	0.274**	-0.130	0.082
Random effect			
Intercept	0.135	0.107	0.163
Residual	0.136	0.132	0.140
PERSON INTERACTING WITH			
Fixed effect			
Intercept	0.177	0.122	0.233
Coworkers	0.017	-0.026	0.061
Family members	0.007	-0.021	0.035
Friends	0.000	-0.043	0.043
Healthcare provider	-0.035	-0.116	0.046
Other known people	0.003	-0.073	0.079
People you are socializing with electronically	-0.196*	-0.304	-0.088
Pets	0.041	-0.003	0.085
Spouse or partners	-0.004	-0.037	0.028
Unknown people	0.047	-0.039	0.133
Depression (previous time point)	0.282**	0.260	0.304
Random effect			
Intercept	0.139	0.109	0.169
Residual	0.136	0.132	0.140
HELP FROM SOMEONE WHILE DOING ACTIVITIES			
Fixed effect			
Intercept	0.182	0.118	0.245
Help score (between-person)	0.002	-0.028	0.031
Help score (within-person)	-0.005*	-0.009	0.000
Depression (previous time point)	0.245**	0.225	0.265
Random effect			
Intercept	0.141	0.112	0.170
Residual	0.126	0.122	0.129

Models	Estimation	95% CI	P-value
PERFORMANCE OF ACTIVITIES			
<i>Fixed effect</i>			
Intercept	0.565	0.364	0.765 <0.0001
Performance score (between-person)	-0.074*	-0.112	-0.037 0.0001
Performance score (within-person)	-0.011*	-0.019	-0.004 0.003
Depression (previous time point)	0.241**	0.221	0.261 <0.0001
<i>Random effect</i>			
Intercept	0.123	0.096	0.150 <0.0001
Residual	0.126	0.122	0.129 <0.0001
SATISFACTION WITH DOING ACTIVITIES			
<i>Fixed effect</i>			
Intercept	0.642	0.448	0.836 <0.0001
Satisfaction score (between-person)	-0.090**	-0.126	-0.053 <0.0001
Satisfaction score (within-person)	-0.012*	-0.019	-0.004 0.002
Depression (previous time point)	0.240**	0.219	0.260 <0.0001
<i>Random effect</i>			
Intercept	0.117	0.091	0.142 <0.0001
Residual	0.126	0.122	0.129 <0.0001
ENGAGEMENT OF ACTIVITIES			
<i>Fixed effect</i>			
Intercept	0.582	0.397	0.766 <0.0001
Engagement score (between-person)	-0.079**	-0.114	-0.044 <0.0001
Engagement score (within-person)	-0.013*	-0.020	-0.006 0.0004
Depression (previous time point)	0.239**	0.219	0.260 <0.0001
<i>Random effect</i>			
Intercept	0.121	0.094	0.147 <0.0001
Residual	0.126	0.122	0.129 <0.0001

** P-value < 0.0001,

* P-value < 0.05.

ADL: activities of daily living; IADL: instrumental activities of daily living.

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