

Activity patterns related to depression symptoms in stressed dementia caregivers

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ABSTRACT

Objectives: Self-reported activity restriction is an established correlate of depression in dementia caregivers (dCGs). It is plausible that the daily distribution of objectively measured activity is also altered in dCGs with depression symptoms; if so, such activity characteristics could provide a passively measurable marker of depression or specific times to target preventive interventions. We therefore investigated how levels of activity throughout the day differed in dCGs with and without depression symptoms, then tested whether any such differences predicted changes in symptoms 6 months later.

Design, setting, participants, and measurements: We examined 56 dCGs (mean age = 71, standard deviation (SD) = 6.7; 68% female) and used clustering to identify subgroups which had distinct depression symptom levels, leveraging baseline Center for Epidemiologic Studies of Depression Scale–Revised Edition and Patient Health Questionnaire-9 (PHQ-9) measures, as well as a PHQ-9 score from 6 months later. Using wrist activity (mean recording length = 12.9 days, minimum = 6 days), we calculated average hourly activity levels and then assessed when activity levels relate to depression symptoms and changes in symptoms 6 months later.

Results: Clustering identified subgroups characterized by: (1) no/minimal symptoms (36%) and (2) depression symptoms (64%). After multiple comparison correction, the group of dCGs with depression symptoms was less active from 8 to 10 AM (Cohen's $d \leq -0.9$). These morning activity levels predicted the degree of symptom change on the PHQ-9 6 months later (per SD unit $\beta = -0.8$, 95% confidence interval: $-1.6, -0.1, p = 0.03$) independent of self-reported activity restriction and other key factors.

Conclusions: These novel findings suggest that morning activity may protect dCGs from depression symptoms. Future studies should test whether helping dCGs get active in the morning influences the other features of depression in this population (i.e. insomnia, intrusive thoughts, and perceived activity restriction).

Key words: Dementia caregiving, depression symptoms, sleep, activity restriction, physical activity

There are approximately 6.7 million family caregivers for people with dementia in the U.S.A. alone (Wolff *et al.*, 2016). About 1.5 million of these dementia caregivers (dCGs) – over 20% – suffer from a depressive disorder (Cuijpers, 2005). Depression not only reduces the quality of daily lives, it is also associated with lower quality of caregiving (Smith *et al.*, 2011). Preventing depression in dCGs is therefore a high priority.

Major depressive episodes often do not “come out of the blue”; instead, the presence of some depressive symptoms predicts the development of more symptoms (Joling *et al.*, 2012) and major depressive disorder (Cuijpers and Smit, 2004). The modifiable characteristics associated with subsyndromal depressive symptoms in dCGs may therefore reflect the mechanisms of depression pathogenesis and represent logical targets for preventive interventions.

One important and potentially modifiable contributor to depression symptoms in dCGs is activity restriction (Mausbach *et al.*, 2011). Past research defines activity restriction as self-reports of reductions in the ability to engage in self-care and

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recreational activities due to the caregiving role. Given the strong relationship between the perception of activity restriction and depression in dCGs, it is plausible that differences in objectively measurable daily activity patterns also cotravel with depression in this group. Unlike self-reported activity restriction, objective activity measures related to depression cannot be due to confounding by negative affective or recall bias. Furthermore, activity patterns can be passively measured using wearable accelerometer devices. As such, evidence for depressogenic effects of particular activity patterns could lead to clinical applications, e.g. wearable actigraphy could be used to help stratify depression risk; and preventive interventions could be delivered to alter activity during key times. However, little evidence currently exists regarding which objectively measured activity patterns characterize depression symptoms in dCGs.

Our past study found that differences in dCGs' daily distribution of activity correlated with subsyndromal depression severity (Smagula *et al.*, 2017). This past work found that dCGs with more depression symptoms tended to have slower transitions between resting and active states, narrower periods of activity, and more activity during sleep. However, because these findings were based on cross-sectional data, it remains unknown whether differences in the daily distribution of activity also predict the changes in depression symptoms over time (i.e. rather than being an element or result of having depression symptoms). We therefore evaluated whether objectively measurable differences in daily activity patterns were related to subsyndromal depression symptoms and changes in symptoms over 6 months in dCGs.

Methods

Sample

Participants were initially contacted through outreach to local support groups and by accessing three registries (the University of Pittsburgh Alzheimer's Disease Research Center; the Pittsburgh Regional Caregiver Survey; and the University of Pittsburgh Clinical and Translational Science Institute "Pitt + Me" registry). To be included, caregivers were required to be at least 60 years of age, provide unpaid care to someone with dementia, have experienced stress or strain, either physical or emotional, related to caregiving, live with their care recipient or provide at least 15 hours of care per week, have Patient Health Questionnaire-9 (PHQ-9) scores less than 10 (consistent with the aim of studying subsyndromal symptoms and increases over time), and be able to undergo

magnetic resonance imaging (administered for an aim of the project that is not addressed in this report). Of the 239 potential participants screened, 86 were eligible and 66% of these eligible dCGs agreed to participate. This yielded a sample of 57 participants, of whom 56 provided adequate actigraphy data (defined below) and were included in the analysis.

Measures

OBJECTIVE ACTIVITY PATTERN

Participants wore a Philips Spectrum Plus (Philips Respironics, Bend, OR, USA) actigraphy device on their nondominant wrist for 14 days. Data were considered adequate if at least 3 days of valid wear were confirmed. Consistent with National Sleep Research Resource data processing standards (see Dean *et al.*, 2016 and associated website), valid days were defined as those that did not include periods of more than 4 hours of offwrist/invalid time or offwrist/invalid times during the main sleep period. Only one participant did not have adequate actigraphy data, and participants were highly compliant with the actigraphy protocol (mean recording length = 12.9 days, median = 14 days, range = 6–15 days).

Our previous publication using cosine-based analysis methods found that differences in the distribution of activity through 24-hour periods related to depression symptoms (Smagula *et al.*, 2017). To delineate which temporal patterns of activity relate to depression symptoms in dCGs, we defined the primary exposure variables in the current work as the average levels of actigraphy-defined activity level for each hour of the day. This set of objective measures can be passively assessed, thereby enabling potential applications tracking risk markers in interventions without active input from dCGs.

COVARIATES

We included several other sleep-wake measures that are plausibly related to depression in dCGs, including circadian preference measured with the Morningness-Eveningness Scale (Smith *et al.*, 1989); insomnia symptom severity measured using the Insomnia Severity Index (Bastien *et al.*, 2001); actigraphy-assessed sleep fragmentation (measured as the number of minutes awake after sleep onset, manually setting sleep intervals based on sleep diary data and then using an automated scoring system); and the Epworth Sleepiness Scale (Johns, 1991). We interviewed participants using a standardized questionnaire that asked: if they live with their care recipient; how many caregiving activities (from a list of 13 activities and instrumental activities of

daily living) they provided care for; how many years they had been in the caregiving role; and the Cumulative Illness Rating Scale for Geriatrics (Miller *et al.*, 1992) to measure medical comorbidity. Participants completed a questionnaire that included a question asking how many hours per day they provided care. Participants also completed questionnaires measuring constructs previously associated with depression: the Activity Restriction Scale (Williamson and Schulz, 1992), the Intrusive Thoughts Questionnaire adapted for caregivers (Schulz *et al.*, 2017), and the Five Facet Mindfulness Questionnaire (Baer *et al.*, 2006).

DEPRESSION SYMPTOMS

Depression symptoms were measured at study baseline with the Center for Epidemiologic Studies of Depression Scale–Revised Edition (Radloff, 1977) and the PHQ-9 (Kroenke *et al.*, 2001). We re-administered the PHQ-9 6 months after baseline via phone interview.

STATISTICAL ANALYSIS

First, the three available depression symptom measurements were entered into a person-centered clustering approach using finite normal mixture model implemented in the R Software package “MClust” (Scrucca *et al.*, 2016). As opposed to using traditional thresholds to create subgroups in our data, this clustering approach has key advantages, namely (1) it allows us to identify naturally occurring subgroups in the data without using an arbitrary cutpoint and (2) it allows us to leverage data from three depression symptom measurements, thereby providing a fuller picture than any single measure. We used the Bayesian Information Criterion to select the optimal model but specified that we would reject models that included small groups defined as <10% of the sample.

Next, we characterized differences between the identified subgroups (which had distinct depression symptom levels). We compared the above-listed characteristics using independent sample *t*-tests and Cohen’s *d* (Cohen, 1988) as a measure of effect size. We applied a Benjamini–Hochberg (1995) multiple comparison correction for the activity pattern analysis (because this analysis included 24 variables each reflecting the average activity levels for an hour of the day).

After identifying the correlates of prevalent depression symptoms subgroups, we conducted analyses aimed at identifying factors independently associated with changes in depression symptoms over time. We used multivariable regression analyses to determine if objectively measured activity levels (at specific times of the day) related to changes in depression symptoms independent of

self-reported covariates and actigraphy-assessed sleep fragmentation. We used a linear regression model with change in PHQ-9 scores (6 month minus baseline scores) as the outcome. As predictor variables, we included age, sex, baseline PHQ-9 score, and the statistically significant correlates of depression symptom subgroups identified in the initial analysis.

Results

Clustering identified groups with distinct depression symptom levels

The Bayesian Information Criterion indicated that a two-group solution was optimal (Supplemental Table 1). This empirical solution identified a majority subgroup of participants (64%) who had depression symptoms, and a minority of dCGs who had no/minimal symptoms of depression (Table 1). Note that the typical cutoff for predicting major depressive disorder using the Revised Center for Epidemiologic Studies – Depression Scale is ≥ 16 (Radloff, 1977); on the PHQ-9 ≥ 5 is considered mild and ≥ 10 is considered moderate depression symptom levels (Kroenke *et al.*, 2001). The means and SDs of symptom levels in the depression symptoms subgroup indicate mild symptoms approaching the range that is traditionally considered clinically meaningful.

Baseline differences associated with membership in depression symptom groups

There were large differences in two specific hourly measures of activity between the groups that survived multiple comparison correction (Figure 1 and Supplemental Table 2). dCGs with depression symptoms had significantly less actigraphy-assessed activity in the hours from 8 to 10 AM (Cohen’s *d* ≤ -0.9); given that associations of these adjacent hourly bins were in the same direction, we summed activity levels in these hours to simplify subsequent modeling. dCGs with depression symptoms also had more activity in the 3 AM bin, though this association was not statistically significant after correcting for multiple comparisons.

There were also large differences in self-reported measures of potential risk factors for depression between the groups. The group with depression symptoms had relatively more insomnia symptoms ($d = 1.0$), more actigraphy-assessed sleep fragmentation ($d = 0.5$), less preference for morningness ($d = -0.7$), more activity restriction ($d = 0.7$), more intrusive thoughts ($d = 0.6$), lower mindfulness scores ($d = -0.9$), and a greater medical comorbidity ($d = 0.4$).

Table 1. Sample characteristics by subgroups with distinct depression symptom levels identified using model-based cluster analysis ($n = 56$)

	CAREGIVERS WITH DEPRESSION SYMPTOMS, 64% ($n = 36$)	CAREGIVERS WITH MINIMAL/NO DEPRESSION SYMPTOMS, 36% ($n = 20$)	COHEN'S <i>D</i>	<i>p</i> -VALUE
Age	69.5 (5.6)	73.3 (8)	-0.5	0.06
White race, % (n)	97 (35)	85 (17)	-	0.09
Female sex, % (n)	72 (26)	60 (12)	-	0.35
Care recipient is spouse, % (n)	72 (26)	70 (14)	-	0.48
Caregiver lives with care recipient, % (n)	94 (34)	75 (15)	-	0.08
Depression symptom levels				
Baseline CES-D	15.2 (10.6)	2.6 (2.3)	1.2	<0.0001*
Baseline PHQ-9	6.6 (3.4)	1.3 (1.0)	1.6	<0.0001*
Six-month PHQ-9	5.4 (3.0)	1.1 (1.0)	1.4	<0.0001*
Caregiving characteristics				
Number of caregiving activities	8.7 (3.5)	9.2 (3.3)	-0.1	0.48
Hours providing care per day	7.5 (5.4)	6.1 (4.9)	0.2	0.39
Years in the caregiving role	6.3 (4.6)	4.9 (2.8)	0.3	0.19*
Sleep-wake factors				
Morningness-Eveningness Questionnaire (higher indicates more of a preference for morning)	39.6 (7.8)	45.4 (5.6)	-0.7	0.005
Insomnia Severity Index	10.2 (5.6)	3.8 (3.8)	1.0	<0.0001*
Sleep fragmentation (minutes awake after sleep onset)	40.9 (16.5)	31.0 (16.1)	0.5	0.03
Epworth Sleepiness Scale	8.0 (4.7)	5.7 (3.7)	0.4	0.10
Other caregiver characteristics				
Activity Restriction Scale	20.1 (6.7)	14.6 (6.8)	0.7	0.003
Intrusive Thoughts Questionnaire	9.3 (4.1)	6.0 (4.6)	0.6	0.006
Five Facet Mindfulness Questionnaire	99.3 (12.9)	112.5 (10.0)	-0.9	0.0002
Cumulative Illness Rating Scale for Geriatrics	5 (3.1)	3.6 (1.8)	0.4	0.04*

Means (standard deviations) shown unless otherwise noted.

Abbreviations: CES-D, Revised Center for Epidemiologic Studies – Depression Scale; PHQ-9, nine-item Patient Health Questionnaire. *p*-values are from *t*-tests or chi-squared tests except for: (1) the “live-in” status and race where Fischer’s exact test was required; and (2) asterisks indicate that a Satterthwaite method was used to account for the unequal variances.

Associations with changes in PHQ-9 scores over time

Only activity levels in the 8–10 AM period were associated with the degree of change in depression symptom levels over time (Table 2). For every SD unit higher activity level in the 8–10 AM period, the PHQ-9 score decreased by 0.8 more units over 6 months. Because morning activity plausibly can be affected by alcohol consumption, we further adjusted for the frequency of alcohol use, which did not alter these estimates.

We did not include objectively measured morning activity levels and the Morningness-Eveningness Scale in the same model, because these two measures tap similar constructs and their correlation were high (Spearman $r = 0.6$).

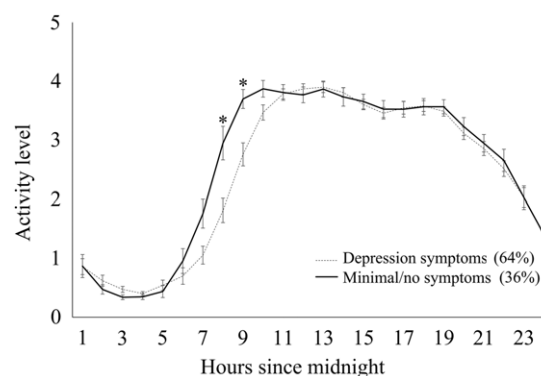


Figure 1. Activity levels in each hour since midnight in subgroups of dementia caregivers with distinct depression symptom levels. Asterisks indicate False Discovery Rates of less than 0.05 (see Supplemental Table 2).

Table 2. Associations of selected potential risk factors with 6-month changes in PHQ-9 scores ($n = 56$)

	β (95% CONFIDENCE INTERVAL)	p -VALUE
Activity level from 8 to 10 AM	-0.8 (-1.6, -0.1)	0.03
Insomnia Severity Index	0.4 (-0.5, 1.3)	0.36
Sleep fragmentation (minutes awake after sleep onset)	-0.1 (-0.8, 0.6)	0.79
Activity Restriction Scale	0.5 (-0.4, 1.3)	0.30
Intrusive Thoughts Questionnaire	0.4 (-0.5, 1.3)	0.42
Five Facets Mindfulness Questionnaire	0.2 (-0.7, 1.1)	0.64
Cumulative Illness Rating Scale for Geriatrics	0.01 (-0.3, 0.3)	0.98

Linear regression model shown predicting PHQ-9 changes scores (6 month minus baseline score). The predictor variables listed were standardized to a mean of 0 and a standard deviation of 1 to facilitate the effect size comparison. All factors are included in a single model that also includes baseline PHQ-9 scores, age, and sex.

Table 3. Baseline and 6-month PHQ-9 scores in groups above and below the median activity levels from 8 to 10 AM

ACTIVITY LEVEL FROM 8 TO 10 AM	BASELINE PHQ-9		6-MONTH PHQ-9	
	MEAN	(95% CONFIDENCE INTERVAL)	MEAN	(95% CONFIDENCE INTERVAL)
Above the median (more morning activity)	4.0	(2.6, 5.4)	2.6	(1.5, 3.8)
Below the median (less morning activity)	5.3	(3.9, 6.7)	5.0	(3.9, 6.1)

Similar results were obtained when altering the model shown in Table 2 using the Morningness–Eveningness Scale in place of objectively measured morning activity levels (per SD unit Morningness–Eveningness Scale, $\beta = -0.8$, 95% confidence interval: -1.6, 0.0, $p = 0.05$). Race and live-in status were not initially included as covariates due to concerns that there were very few non-Whites and non-live-in dCGs in our sample (Table 1), but further adjusting for these factors did not alter our findings.

Given these findings, we also examined whether diary-assessed morning wake and rise times (time the individual gets out of bed in the morning) were also related to changes in depression symptoms over time. Later wake and rise times were similarly associated with the degree of symptom changes (per SD unit wake time, $\beta = 1.0$, 95% confidence interval: 0.2, 1.8, $p = 0.01$; per SD unit time out of bed, $\beta = 1.1$, 95% confidence interval: 0.3, 1.9, $p = 0.006$).

To further illustrate these results, we also report PHQ-9 scores at baseline and the 6-month follow-up stratified by whether dCGs were above and below the median morning (8–10 AM) activity level (Table 3). Unlike dCGs above the median morning activity level, dCGs below the median morning activity level maintained symptom levels in the mild range at the 6-month follow-up.

Discussion

Using data-driven clustering, we identified two distinct subgroups of dCGs: a minority who had minimal/no depression symptoms, and a majority who had a considerable symptom burden. Because the subgroups were similar on several markers of caregiving intensity, the group with minimal/no symptoms might be considered resilient to the stressors of dementia caregiving. In contrast, the group that had depression symptoms was characterized by insomnia symptoms, sleep fragmentation, reports of activity restriction, intrusive thoughts, lower levels of self-reported mindfulness, and a greater burden of medical comorbidity. These factors are all potentially modifiable and had similarly large associations with prevalent depression symptoms, suggesting multiple important features of prevalent symptomology (i.e. the qualities of depression in dCGs) that should be considered in interventions. But among these correlates of depression symptoms, only markers of morning activity independently predicted the degree of change in depression symptoms over time; these associations were independent of baseline symptom levels, actigraphy-assessed sleep fragmentation, and the self-reported correlates of prevalent depression listed above.

Measures related to morning activity (i.e. self-reported preferences, morning wake times, and

times out of bed) were also associated with changes in depression symptoms over time. This suggests that, rather than being a characteristic of prevalent depression, staying in bed and not being active in the morning may shape the trajectory of depression symptoms in dCGs. Specifically, dCGs with higher morning activity levels had lower levels of depression symptoms 6 months later (Table 3); in contrast, dCGs with lower morning activity levels maintained their mild depression symptoms (indexed by the PHQ-9) 6 months later. Thus, low levels of morning activity may perpetuate depression symptoms, potentially impairing quality of life and increasing the risk of future conversion to major depression.

These observations are consistent with prior evidence that a preference for “morningness” is associated with lower rates of depression (Taylor and Hasler, 2018). While the mechanisms behind these associations are not clear, more “eveningness” is associated with several known neurobiological correlates of depression: default mode network connectivity (Facer-Childs *et al.*, 2019; Horne and Norbury, 2018), activation of frontal networks in the morning (Schmidt *et al.*, 2015), and grey matter density in these regions (Takeuchi *et al.*, 2015). It is plausible that morning inactivity reflects and/or leads to a lack of engagement in key circuits, e.g. those responsible for motivating action and deliberately regulating behavior. Our findings and this interpretation are consistent with evidence that negative mood is associated with a lower morning cortisol awakening response in caregivers (Leggett *et al.*, 2014). Future work is needed to understand how these hormonal and neurobiological factors interrelate in determining morning activity and its relationship with depression symptom.

Several limitations of this work should be noted. Because the follow-up period was relatively short and severe depression was not observed, we cannot be certain that these observations generalize to severe depression. Our finding that morning activity temporally precedes the degree of depression symptom change, independent of other factors, is consistent with a causal effect; but analyses were correlational in design and we cannot ascertain causality due to the potential of unmeasured (residual) confounding. Residual confounding, e.g. if morning activity marks an unmeasured aspect of depression, could account for the longitudinal association of morning activity with the degree of preexisting change. Additional limitations of the current work include a relatively small, older, and racially homogeneous sample, so

these findings do not necessarily generalize to caregivers in different groups.

In conclusion, we have extended the existing literature on activity restriction and depression in caregivers (Mausbach *et al.*, 2011) by examining objectively measured 24-hour activity patterns in relation to depression symptoms and their persistence 6 months later. The findings and limitations of the current work indicate that experiments are needed to test whether modification of morning activity patterns has a causal and clinically meaningful effect on dCG’s mood. The existing interventions targeting psychosocial factors in dCGs (e.g. Cheng *et al.*, 2016; Collins and Kishita 2018; Liu *et al.*, 2017) may yield additional benefits if they were to address the potential perpetuating effects of morning inactivity on depression. This proposal is consistent with the important role of behavioral activation in Cognitive Behavioral Therapy for depression (Beck, 1979). Behavioral activation aims to increase the frequency of active, positive, engagement by expanding exposures to potentially rewarding activities (Kanter *et al.*, 2010; Lewinsohn *et al.*, 1980). Our data specify that depression-vulnerable dCGs may be in particular need of help being active in the morning, e.g. by planning morning activities or respite services. Future studies are needed to test whether interventions that increase morning activity help mitigate subsyndromal symptoms and reduce the likelihood of their progression to major depression. In addition, studies with longer follow-up periods and experimental designs are needed to determine whether passive monitoring of morning activity may be developed to achieve precision medicine risk stratification approaches.

Conflict of interest

None.

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Description of authors' roles

SFS, BPH, RS, CFR, DJB, HJA, and MHH designed the study. All authors conceptualized the analysis. JLG, RTK, and SFS designed the analytic pipeline. SFS executed the analysis and drafted the paper. All authors critically revised the paper and approved the final submission.

Supplementary materials

To view supplementary material for this paper, please visit <https://doi.org/10.1017/S1041610219001601>

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