

Sleep Quality Partially Mediate the Relationship Between Depressive Symptoms and Cognitive Function in Older Chinese: A Longitudinal Study Across 10 Years

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Objective: This study aimed to examine the relationship between cognitive function and depressive symptoms and to explore the mediating role of sleep quality in the cognition-depression relationship in Chinese older adults (OAs).

Methods: Data came from a nationally representative sample of 16,209 Chinese OAs (aged 65+) from 2008, 2011, 2014, and 2018 waves of the Chinese Longitudinal Healthy Longevity Survey (CLHLS). A random intercept cross-lagged panel model (RI-CLPM) combined with mediation analysis was adopted to determine the relationship between cognitive function and depressive symptoms and the mediating effect of sleep quality on the ascertained cognition-depression relationship.

Results: Poorer cognitive function at prior assessment points were significantly associated with severe depressive symptoms at subsequent assessments, and vice versa. Sleep quality partially mediated the prospective relationship of cognition on depressive symptoms, which accounted for 3.92% of the total effect of cognition on depression.

Discussion: Cognitive decline may predict subsequent depressive symptoms, and vice versa. The impact of cognition on depression is partially explained by its influence on sleep quality. Multidisciplinary interventions aimed at reducing depression and cognitive decline per se as well as improving sleep quality would be beneficial for emotional well-being and cognitive health in OAs.

Keywords: cognitive function, depressive symptoms, sleep quality, random intercept cross-lagged panel model, mediation, Chinese older adults

Introduction

As the life expectancy of the world's population continues to increase, research on cognitive aging has been greatly promoted.¹ Cognitive deterioration and accompanying physical and mental health problems, such as dementia, depression, sleep disorders, and disability, are considered to be the most challenging problems faced by an aging society.² This will greatly damage the quality of life of the elderly and increase the burden of cognitive impairment.³ Mild Cognitive Impairment (MCI) is defined as a transitional stage between normal aging and dementia, which is characterized by cognitive deficits.⁴ The risk of dementia in MCI patients is significantly higher, with an annual progression rate between 8% and 15%.⁵⁻⁷ It is not only the main cause of disability and death among older adults (OAs) but has a profoundly detrimental effect on their families and caregivers. With a rapidly aging population, MCI was prevalent in recent years. In China, the prevalence of MCI among the aging population (aged 60 and older) was 14.71%.⁸ Therefore, it is necessary to reveal the influencing factors that can prevent and delay the harmful effects of aging.

In the past two decades, researchers have conducted to examine the relationship between cognitive function and depression.⁹⁻¹² Most of these studies have shown a link between depressive symptoms and dementia or cognitive decline,

but the nature of the relationship remains unclear. Many of them show that low cognitive function is a strong predictor of depression,^{13–15} or its manifestations, such as suicide completion,^{16,17} and suicidal thoughts.^{18,19} Research showed that depressive symptoms preceded cognitive decline.^{20–23} Depressed participants exhibited poorer performance at baseline and greater subsequent decline in all cognitive domains, which supports the theory that repeated depressive episodes may contribute to cognitive decline.^{22,23} More than just a predictor, though, many studies have shown that previous cognitive function is related to later depression,^{14,24–27} which suggests that cognitive function might play a causative role in depression. However, some studies did not report any significant association between cognitive function and depression.^{28,29} Although many studies have found a cross-sectional relation between depression and dementia or depressive symptomatology and cognitive functioning, the direction of this association is still inconsistent. For this reason, it is necessary to explore the relationship and the possible mechanism between depression and cognition in larger and longer follow-up samples.

Studies have found that poor sleep quality can affect many aspects of health, such as cognitive function,³⁰ health status,³¹ emotion,³² psychological well-being,³³ and others.³⁴ Thus, sleep quality might play an important role in considering the relationship between depression and cognition.³⁵

Firstly, a large number of investigations have shown that poor sleep quality is related to poor cognitive performance, such as processing speed, episodic memory recall, and executive function in the elderly.^{36–39} The link between sleep and neurodegenerative disease might be bi-directional; sleep disturbance serves not only as a marker but a promoter of cognitive impairment. Their mutual influence in many aspects is of great significance to the diagnosis and treatment of cognitive deterioration.⁴⁰ Recent studies have shown that in Alzheimer's disease (AD), the degeneration of the suprachiasmatic nucleus and prefrontal cortex in dementia disrupts sleep-wake rhythmicity and decline of cognitive ability. Markers for AD pathology are also related to sleep disorders in the elderly with normal cognition, suggesting that preclinical AD pathology may be the cause of sleep disorders. For example, in the elderly with normal cognition, amyloid deposits measured by positron emission tomography (PET) and A β 42 levels in cerebrospinal fluid increased with self-reported decreasing sleep duration,⁴¹ excessive daytime sleep,⁴² longer sleep latency,^{43,44} and poorer sleep quality.^{45,46} On the other hand, sleep deprivation increases the production of amyloid- β and reduces its clearance,⁴⁷ which leads to a higher risk of cognitive impairment.^{48–51}

In addition, there is a relationship between sleep quality and depression, and it may be mutual,^{52–55} which leads to a vicious circle between them.³⁵ Poor sleep quality may affect emotional regulation, in many cases leading to increased negative emotions, sleep interruption, and more severe depressive symptoms.^{31,56} Evidence shows that individuals with chronic insomnia are at a higher risk of developing depression, as chronic insomnia could disrupt the balance of the hypothalamic-pituitary-adrenal axis, leading to depression.^{57–61} Alternatively, people with depression are more likely to have sleep disturbances.⁶² A longitudinal study demonstrated that symptoms of depression at baseline would lead to more sleep complaints after 4 years in older adults.⁶³ The possible mechanism is that depression can inhibit the secretion of melatonin, interfere with the circadian rhythm, and disturb sleep.⁶⁴ In addition, the occurrences were not coinstantaneous, as longitudinal studies have shown the time lag of patients with sleep disorders to develop depression may exceed several months.^{60,61}

Those studies above showed sleep quality might be bi-directional correlated with depression and cognitive function, respectively, thus sleep quality might play a role in the relationship between cognitive function and symptoms of depression. However, the existing research on the relationship between depression, cognition, and sleep was mainly cross-sectional and bilateral, they failed to examine if one variable has a causal relation to the others or the trilateral relationship between them.^{65,66} Longitudinal studies were desperately needed to answer the question of whether the cognitive ability is causally related to depression (or vice versa), and the possible influence of sleep quality on this relationship.

This study aimed to examine the pathways through which depression and sleep quality affect cognitive function among Chinese OAs by using data from a longitudinal nationally representative survey of the Chinese older population. Firstly, we examine the reciprocal relationship between symptoms of depression and cognitive function using a national elderly sample. We hypothesized that previous depressive symptoms would negatively predict subsequent cognitive function, and previous cognitive function would also negatively predict subsequent depressive symptoms. Then, given

the potentially important role of sleep disturbance in the pathogenesis of cognitive impairment, we further explore the mediating role of sleep quality in this relationship. In addition, depression and cognitive function are influenced by many other factors, such as demographic,⁶⁷ socioeconomic status,⁶⁸ lifestyles,^{69,70} interpersonal relationships,⁷¹ and daily living activity.⁷² These factors may influence or cover up the relationship between depression and cognition. Therefore, in this study, we took these influencing factors as control variables and sleep quality as a mediator to investigate the mediating effect of sleep quality on the relationship between depression and cognition at a longitudinal level.

Method

Data and Sample

The data came from the Chinese Longitudinal Health and Longevity Study (CLHLS), a dynamic cohort study. The CLHLS recruited representative samples of China old adults from about half of 22 cities and counties in 31 provinces of China. The first wave of CLHLS data was collected in 1998 and conducted every three years with the latest wave in 2018. The survey covers 1.16 billion people, accounting for 85% of China's total population. Interviews were conducted to obtain data on demographic characteristics, socioeconomic status, lifestyle, and health. More detailed information on CLHLS could be found elsewhere.⁷³ The current study used data from the last four waves of CLHLS, each with a 3-year interval from 2008 (the "baseline" time) to 2018. Our study focused on the 2008 group, excluding respondents who were added to replace those who died in the subsequent waves. At baseline, there were 16,954 participants in the cohort. Those who reported their age at 106 or older were excluded because there was no reliable information to verify their age.⁷⁴ Based on previous studies of older respondents, respondents under 64 years of age and those who lacked MMSE in the first wave were excluded as well. Therefore, the final analysis sample consists of 16,209 respondents.

Measures

Cognitive Function

Cognitive function was assessed with a Chinese modified Mini-mental State Examination (CMMSE). The original MMSE has 30 items and tests 7 domains of cognitive function.⁷⁵ Considering that two-thirds of Chinese OAs are illiterate, the CLHLS deleted 2 items of language (write a complete sentence and follow a written instruction to close eyes) from the original version. To make questions for OAs easier to understand and answer, it further deleted 1 item of time orientation and 4 items of place orientation and culturally adapted the remaining 23 items.⁷⁶ Each item of the Chinese CMMSE is scored 1 if the answer is correct and 0 for an incorrect answer or "unable to answer", so higher scores indicate better cognitive function. Factor analyses revealed that a model with four factors consistently fitted the four waves' CMMSE data well. The four factors were orientation (orientation to time and place), memory (registration and delayed recall), attention and calculation (calculation and copying intersecting polygons), and language (repeating phrase, three-step command, and naming). Cronbach's α coefficients of the CMMSE ranged from 0.88 to 0.91 for the four waves of the CLHLS.

Depressive Symptoms

Depressive symptoms in CLHLS were assessed by a five-item scale. These five items have been often adopted to indicate depressive symptoms in various studies using the CLHLS data.^{77–80} Of the five questions on this scale, two questions measure positive feelings, and the other three measure negative effects. The first two questions include "Do you see the bright side of things?" and "Are you still as happy as you were when you were young?" The other three questions are "Do you often feel anxious or fearful?", "You often feel lonely and Is it isolated?" and "Do you think the older you are, the more useless you are?". The respondents were asked to choose from five frequency responses with 1 representing "Always", 2 representing "Often", 3 representing "Sometimes", 4 representing "Seldom" and 5 representing "Never". Then, the three negative items were reverse coded, and two positive items were added to calculate the total score. Therefore, the total score ranges from 5 to 25, with a higher value indicating more severe depressive symptoms. Previous research reported that the internal consistency reliability of the depressive emotion scale based on the 2008 sample is 0.66, above the acceptable values of 0.6.⁸¹ Principle component analysis based on the first wave generated one factor with eigenvalues ≥ 1 , explaining 43% of the total variance.

Sleep Quality

Sleep quality was assessed by a single item sleep quality scale. Participants were asked to answer the following question: “how do you rate your recent sleep quality?” Responses were rated in a 5-point format (“1 = very good”, “2 = good”, “3 = fair”, “4 = bad”, or “5 = very bad”). The single item sleep quality scale is a self-reported measurement to assess sleep quality that can be used when time is constrained or quick identification of sleep impairments.⁸² In addition, the single-item scale reduced the response burden on participants, especially in OAs with decreased cognitive function like in the present study. The self-reported measure of sleep quality in a single item has been widely used in previous studies and has been shown to be highly correlated with the multi-item versions of the Sleep Quality Scale.^{83–85}

Control Variables

To minimize the possibility of a third variable influence on the cognition-depression relationship and to keep the parsimony of our model, we controlled a limited number of covariates associated with depression and cognition. According to previous research, all covariates are based on baseline data.⁸⁶

Firstly, demographic control variables included age, gender, and education level. Then, lifestyle covariates included physical exercise, current smoking, and current drinking status. Respondents were asked whether they regularly participate in physical exercise (1 = yes; 2 = no), whether they currently smoke (1 = yes; 2 = no), and drinking (1 = yes; 2 = no). Interpersonal relationship covariates included marital status (1 = never married, separated, divorced, or widowed; 0 = married), living arrangement (1 = alone; 0 = with family members or in an institution) and social activity. Social activity was assessed by a single question: “Do you take part in some social activities at present?” (1 = almost every day; 2 = not daily, but once for a week; 3 = not weekly, but at least once for a month; 4 = not monthly, but sometimes; 5 = never). In addition, difficulties with six types of Activities of Daily Living (ADL) (dressing, walking, bathing, eating, getting in or out of bed, and using the toilet) were assessed by asking participants to indicate whether they had any difficulty performing these activities.

Analytic Strategies

The Cross-Lagged Panel Model (CLPM) is traditionally used as a typical method to study the causal impact between variables in longitudinal panel data. However, the CLPM only considers the temporal stabilities of variables, but fails to fully consider the time-invariant individual differences, which may lead to biased estimation.⁸⁷ To solve this problem, the random intercepts cross-lagged panel model (RI-CLPM) has been proposed. The RI-CLPM uses a multilevel perspective to distinguish the within-person internal process from between-person stable differences, which allows the separation of within-person effects from between-person effects.⁸⁸ In this study, the RI-CLPM analysis was adopted to examine the direction and strength of association between cognition and depressive symptoms. Four domain-representative item parcels were created as the indicators of cognition,⁸⁹ with each parcel representing each domain we defined in the measurement invariance. Figure 1 shows an illustration of the RI-CLPM examined in the current study. The random intercepts reflect an individual’s average, stable level of cognitive function, sleep quality, and depressive symptoms.

Longitudinal measurement invariance of cognition across measurement occasions is a prerequisite to modeling changes over time.⁹⁰ Using measurement invariance testing, Confirmatory Factor Analysis (CFA) found that our proposed construct of cognition was invariant over time on these parcels’ loadings and intercepts, and thus met the requirements for strong measurement invariance. The scalar longitudinal invariance model of cognition allowed correlations between residual error variances of the same parcel and imposed invariance constraints on the factor loadings and intercepts across the four-time points. Moreover, CFA of the five-item Depression Symptoms Scale also showed scalar longitudinal invariance, which indicates that the longitudinal cross-lagged analysis between depressive symptoms and cognitive function is permissible. To simplify the model structure, the symptom of depression was computed as the total score of all five items. Our RI-CLPM mainly involved the comparison of four-level nested models (Table 1). Model 1 evaluated the correlations of cognition and depressive symptoms within time points and autoregressive paths between time points. Model 2 assessed the stability in autoregressive paths of cognition and depressive symptoms. Added cross-lagged paths of cognition and depressive symptoms, Model 3 tested the reciprocal relationships between them. Model 4 tested “stationarity” in cross-lagged paths. Nested models were compared with the corrected

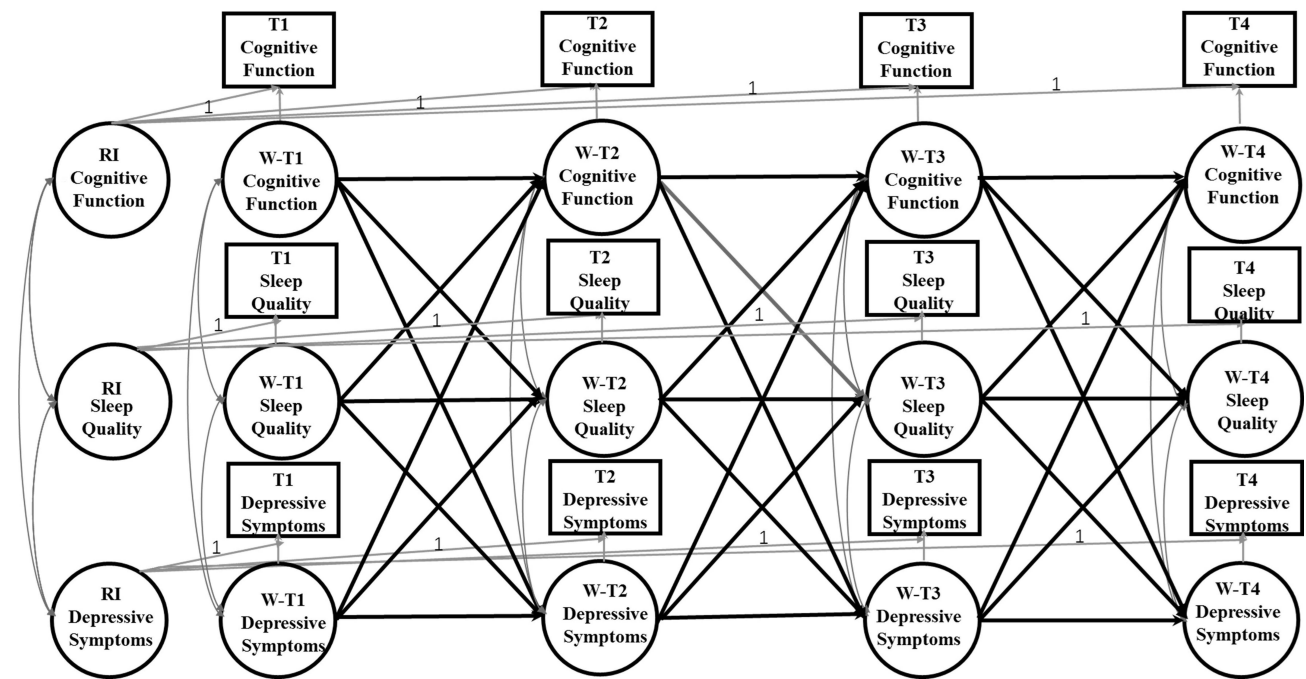


Figure 1 A depiction of the RI-CLPM including the relationships between cognitive function, depressive symptoms, and sleep quality across the three waves. W represents within-person variable; T1 represents 2008; T2 represents 2011; T3 represents 2014; T4 represents 2018. The observed variables are represented in the squares. Of these observed variables, latent variables at both the between and within level were constructed. These latent variables are displayed as ovals. The three random intercepts (RI Cognitive Function, RI Depressive Symptoms, and RI Sleep Quality) reflect the between-person variances. The nine latent within-person variables (WT1-WT4 Cognitive Function, WT1-WT4 Depressive Symptoms, and WT1-WT4 Sleep Quality) reflect the within-person variances across 4 waves. The within-person paths are illustrated by the modeling structure between the nine latent within-person variables: 3 autoregressive paths between the latent factors of Within Cognitive Function, 3 autoregressive paths between the latent factors of Within Depressive Symptoms, and 3 autoregressive paths between the latent factors of Within Sleep Quality; 3 cross-lagged paths from the latent factors of Within Cognitive Function to Within Depressive Symptoms and 3 cross-lagged paths from the latent factors of within Depressive Symptoms to Within Cognitive Function; 3 cross-lagged paths from the latent factors of Within Cognitive Function to Within Sleep Quality and 3 cross-lagged paths from the latent factors of Within Sleep Quality to Within Cognitive Function; 3 cross-lagged paths from the latent factors of Within Depressive Symptoms to Within Sleep Quality and 3 cross-lagged paths from the latent factors of Within Sleep Quality to Within Depressive Symptoms; correlations between Within Cognitive Function, Within Depressive Symptoms, and Within Sleep Quality at Wave 1, and between the residual correlations of Within Cognitive Function, Within Depressive Symptoms, and W Sleep Quality at Waves 2 and 3.

scaled Chi-square difference test.⁹¹ Sleep quality was hypothesized to mediate the reciprocal depression–cognition relationship ascertained in Model 4. The three indirect effects, previous depressive symptoms on posterior cognition via present sleep quality and previous cognition on posterior depressive symptoms via present sleep quality, were simultaneously tested in Model 5. The indirect effect of X predicting Y via M was the product of two direct effects on the path. Models 4i included all control variables as covariates predicting ~2008–2018 cognition and depressive symptoms based on Model 4, while Models 5i included all control variables as covariates predicting ~2008–2018 cognition, depressive symptoms, and sleep quality based on Model 5. Paths from covariates to cognition and depressive symptoms, correlations between variables at a given time, autoregressive paths of time-varying variables, and cross-lagged paths between time-varying variables were also added to Models 4i and 5i. Because RI-CLPM routinely assumes that prospective relationships between variables are stable over time,⁹² we applied equality constraints to the autoregressive and cross-lagged paths as well as paths from time-invariant covariates to cognition and depressive symptoms.

Available data of all the 16,209 respondents were used in this study, and the missing data were dealt with Full Information Maximum Likelihood (FIML) estimation.⁹³ The highest absolute values of skewness and kurtosis for our observed variables were 2.197 and 4.973 for the CMMSE score, therefore, the parameters were estimated with Robust Maximum Likelihood. We reported unstandardized regression coefficients (β s) and p values throughout.

The degree of model fit was assessed with Satorra–Bentler χ^2 statistic (S–B χ^2), Root Mean Square Error of Approximation (RMSEA), Comparative Fit Index (CFI), and Tucker–Lewis Index (TLI), however, we primarily depended on the results of RMSEA, CFI, and TLI, because χ^2 goodness-of-fit statistic is too sensitive for a very large sample such as used in our study.⁹⁴ RMSEA value under 0.05 represents a good fit to the data, and values between 0.05

Table 1 Fit Indices of Structured Equation Models for Cross-Lagged Panel and Mediation Analysis

Model	S-B χ^2	df	RMSEA	CFI	TLI	Δ CFI	Δ RMSEA	Δ S-B χ^2	Δ df	p	Scaling Correction Factor
1. Model 1, correlations within time points and autoregressive paths between time points	2225.798	179	0.027 (0.026 0.028)	0.967	0.964			—	—	—	1.2993
2. Model 2, equating autoregressive paths	2691.919	186	0.029 (0.028 0.030)	0.959	0.958	-0.008	-0.006	466.12	7	<0.001	1.3038
3. Model 3, plus cross-lagged paths	2426.934	177	0.028 (0.027 0.029)	0.963	0.96	0.004	0.002	-265	-9	<0.001	1.3011
4. Model 4, equating cross-lagged paths	2562.156	184	0.028 (0.027 0.029)	0.961	0.96	-0.002	0	135.22	7	<0.001	1.3066
4i. Model 4i, controlling for covariates based on Model 4	4548.292	289	0.030 (0.029 0.031)	0.948	0.934			—	—	—	1.1525
5. Model 5, plus two cross-lagged paths that shared sleep quality as a mediator	2780.717	251	0.025 (0.024 0.026)	0.963	0.96			—	—	—	1.2166
5i. Model 5i, controlling for covariates based on Model 5	5430.399	411	0.027 (0.027 0.028)	0.944	0.929			—	—	—	1.1222

Notes: "—" Represents these models are not nested, χ^2 comparisons to prior models are omitted, Δ represents the comparisons of the model fit indexes of S-B χ^2 , CFI, RMSEA and df.

and 0.08 indicate a reasonable fit. For the CFI/TLI, values greater than 0.90 are considered as an acceptable fit and 0.95 or more as a good fit.⁹⁵ Since the sensitivity of a large sample in the χ^2 difference test, the comparisons of the models were mainly determined by the difference in CFI (Δ CFI) and RMSEA (Δ RMSEA) between the base model and the constrained model.^{96–98} When the Δ CFI and Δ RMSEA are not greater than 0.01, the invariance model is acceptable.⁹⁸ All these analyses were conducted with Mplus, version 7.4.⁹³

Results

Descriptive Statistics

Sample characteristics are shown in Table 2. The mean scores of the existing samples from 2008 to 2018 were 17.76, 18.34, 19.24, and 19.00 for cognitive function, 17.70, 18.71, 18.64, and 18.16 for depressive symptoms, and 2.36, 2.35, 2.39, and 2.50 for sleep quality respectively. Among the 2365 survivors who were interviewed at each wave, the results of repeated-measures analysis of variance indicated that depressive symptoms increased over time between 2002 and 2011 ($F = 57.591$, $p < 0.001$), whereas mean CMMSE scores decreased ($F = 555.137$, $p < 0.001$).

Correlations between cognitive function, depressive symptoms, and sleep quality are presented in Table 3. On any given occasion, depressive symptoms and cognitive function were negatively and significantly correlated ($p < 0.001$). The score of depressive symptoms was negatively correlated with later cognitive function ($p < 0.001$) and cognitive function was negatively correlated with the later depressive symptoms ($p < 0.01$). Sleep quality was positively correlated with depressive symptoms and negatively correlated with cognition both on the same occasion ($p < 0.01$).

Table 2 Sample Characteristics of Depressive Symptoms, Cognition (MMSE Score), Sleep Quality and Covariates

Variables	2008	n = 16,209	2011	n = 8223	2014	n = 5111	2018	n = 2365
	Mean/%	SD	Mean/%	SD	Mean/%	SD	Mean/%	SD
Age (65–105 years)	87.05	11.09	85.73	10.65	85.13	9.76	85.38	8.03
Sex								
Male	42.81		45.30		46.35		46.09	
Female	57.19		54.70		53.65		53.91	
Education	2.01	3.39	2.35	3.56	2.53	3.56	2.95	3.89
Marriage statue								
Married	69.73		63.70		60.74		60.26	
Unmarried	30.27		36.30		39.26		39.74	
Living arrangement								
Living with others	84.45		83.33		81.88		78.85	
Living alone	15.55		16.67		18.12		18.77	
Smoking								
Current smoker	17.19		18.08		17.41		15.63	
Current no smoking	82.81		81.92		82.59		84.37	
Drinking								
Current drinker	16.95		17.07		15.74		14.46	
Current no drinking	83.05		82.93		84.26		85.54	
Physical exercise								
Regular physical exercise	27.73		36.45		32.19		32.35	
No physical exercise	72.26		63.55		67.81		67.65	
Frequency of social activity	4.75	0.80	4.69	0.88	4.67	0.90	4.74	0.83
Activities of daily living	6.97	2.38	7.21	2.64	7.14	2.60	7.00	2.45
Depressive symptom	11.75	2.55	11.29	3.37	11.36	3.44	11.84	2.41
Cognitive function	17.76	6.19	18.34	5.93	19.24	5.08	19.00	5.44
Sleep quality	2.36	0.95	2.35	0.97	2.39	0.96	2.50	1.01

Table 3 Correlations Between Latent Variables of Depressive Symptoms and Cognitive Function and a Manifest Variable of Sleep Quality

Variables	1	2	3	4	5	6	7	8	9	10	11	12
1. Cognitive function 2008	1											
2. Cognitive function 2011	1											
3. Cognitive function 2014	0.379***	1										
4. Cognitive function 2018	0.331***	0.507***	0.484***	1								
5. Depressive symptoms 2008	0.288***	0.395***	-0.117***	-0.098***	1							
6. Depressive symptoms 2011	-0.189***	-0.123***	-0.133***	-0.119***	0.172***	1						
7. Depressive symptoms 2014	-0.108***	-0.212**	-0.179***	-0.085***	0.160***	0.326***	1					
8. Depressive symptoms 2018	-0.072***	-0.121***	-0.062*	-0.087***	0.101***	0.129***	0.215***	1				
9. Sleep quality 2008	-0.071***	-0.038	-0.016	-0.032	0.184***	0.119***	0.123***	0.079**	1			
10. Sleep quality 2011	-0.073***	-0.023	-0.035*	-0.051*	0.095***	0.281***	0.157***	0.116***	0.234***	1		
11. Sleep quality 2014	-0.003	-0.073***	-0.041**	-0.024	0.087***	0.149***	0.277***	0.128***	0.207***	0.337***	1	
12. Sleep quality 2018	0.005	-0.040**	-0.044	-0.072**	0.062**	0.166***	0.156***	0.209***	0.258***	0.298***	0.343**	1

Notes: * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$.

The Reciprocal Relationship Between Depressive Symptoms and Cognitive Function

Figure 2 depicts Model 4i. After controlling for covariates, Model 4i fitted the data adequately (RMSEA = 0.030, CFI = 0.948, TLI = 0.933; Table 1), and the 3-year cross-lagged effects of prior depressive symptoms on cognition ($\beta = -0.016$, $p < 0.01$) and cognition on subsequent depressive symptoms ($\beta = -0.244$, $p < 0.001$) were significant. Lower prior cognitive function predicted higher subsequent depressive symptoms ($p < 0.01$), and higher prior depressive symptoms predicted lower subsequent cognitive function ($p < 0.001$).

Mediating Effect of Sleep Quality

As Figure 3 presents, after adding three indirect paths that shared sleep quality as a potential mediator and adjustment for control variables, Model 5i still had a good fit to the data (RMSEA = 0.027, CFI = 0.947, TLI = 0.932; Table 1). The 3-year cross-lagged effects of depressive symptoms on subsequent cognition ($\beta = -0.037$, $p < 0.01$) and prior cognition on subsequent depressive symptoms ($\beta = -0.435$, $p < 0.001$) were significantly reduced in size compared with those in Model 4i. The indirect effect of prior cognition on subsequent depressive symptoms via sleep quality was significant (effect size = -0.004 , $p < 0.05$), lower prior cognitive function predicted worse subsequent sleep quality (higher score) ($\beta = -0.026$, $p < 0.001$), and worse prior sleep quality predicted higher amounts of subsequent depressive symptoms ($\beta = 0.138$, $p < 0.01$). The effect size of the longitudinal mediation was computed with MacKinnon's formula for calculating the mediated percentage, which is the indirect effect divided by the total effect. However, this indirect effect size only accounted for 3.96% ($-0.004/-0.101$) of the total effects demonstrated in Model 4i.

Discussion

As far as we know, this is the first 10-year large-scale cohort study on the prospective relationship between cognition, depression, and sleep quality in China. Our findings not only strengthen the assertion that depressive symptoms have an adverse effect on cognitive function, but also show that cognitive dysfunction may exacerbate depression, forming a “vicious circle” that further damages over time Cognition. Finally, the indirect effect of cognition on depression through sleep quality was statistically significant, suggesting a mechanism that cognition can affect depression via its impact on sleep quality.

Most previous studies have found an association between cognition and depression.^{9–12} Some studies found that previous cognition was a predictor of follow-up depression,^{14,24–27} and some other studies showed depressive symptoms

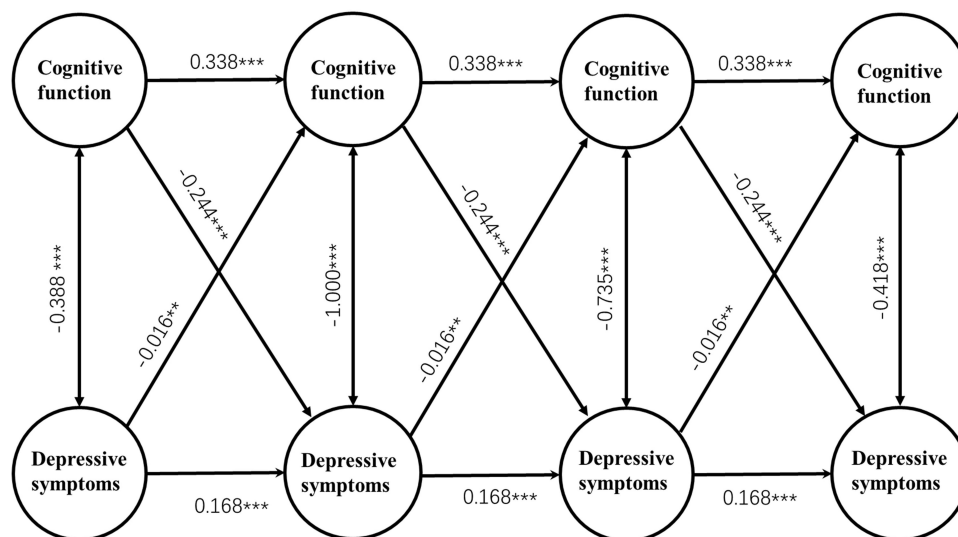


Figure 2 Latent variable cross-lagged panel model of the reciprocal relationship between depressive symptoms and cognitive function. Parcels of cognitive function, overtime correlations between parcel-specific residuals, fixed residual variances of depressive symptoms and control variables are not shown to enhance clarity. Unstandardized regression coefficients are presented. *** $p < 0.001$, ** $p < 0.01$.

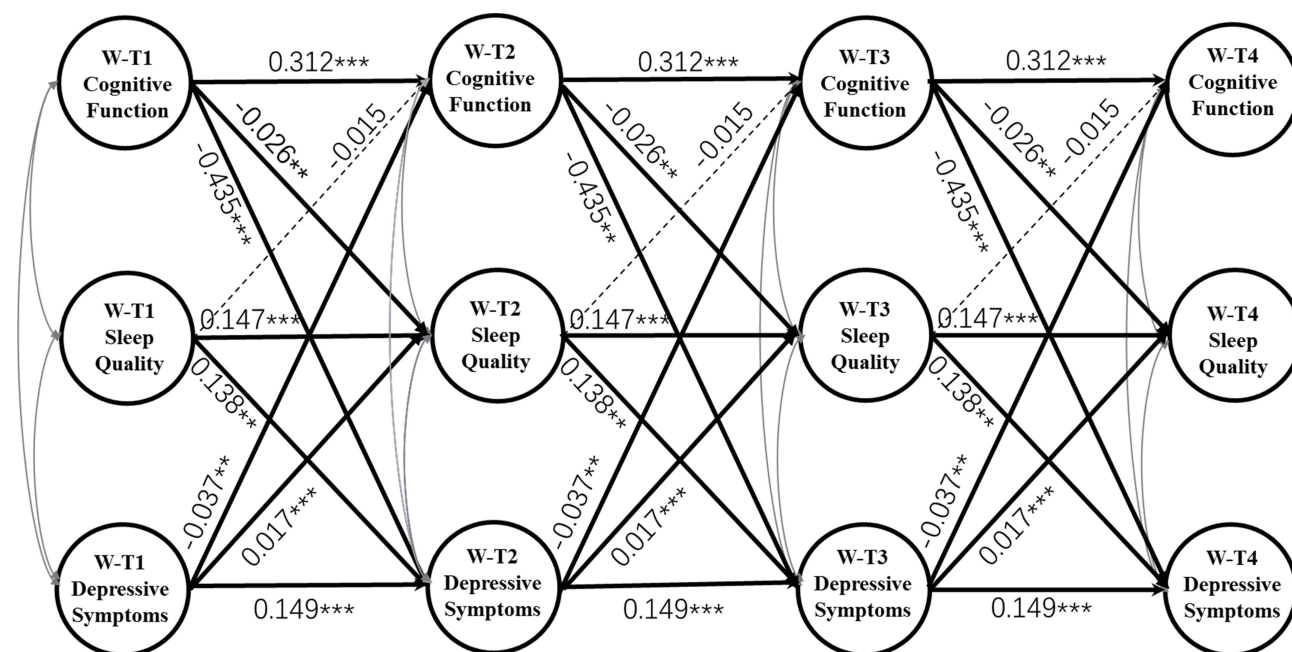


Figure 3 Mediating effects of Within Sleep Quality on the reciprocal relationship between Within Depressive Symptoms and Within Cognitive Function. W-T1 represents within-person variable in 2008; W-T2 represents within-person variable in 2011; W-T3 represents within-person variable in 2014; W-T4 represents within-person variable in 2018. Correlations among the three variables on each occasion and control variables are not shown to enhance clarity. Unstandardized regression coefficients are presented. *** $p < 0.001$, ** $p < 0.01$.

preceded cognitive decline.^{20–22} However, the directional relationship between them remains ambiguous.⁹⁹ Our study found that previous cognitive function was negatively associated with subsequent depressive symptoms; prior depressive symptoms were also negatively but marginally associated with subsequent cognitive function. Our study demonstrates there is a negative bidirectional causality between cognitive function and depression symptoms though exploring a nationally representative longitudinal dataset. This result contributes to some of the previous research which indicates that the interactions between depression and cognitive impairment can be bidirectional making it difficult to distinguish causality.¹⁰⁰

The bidirectional cognition–depression relationship over time remained significant after accounting for demographic (age, gender), socioeconomic status (education), lifestyles (physical exercise, smoking, and drinking), interpersonal relationships (social activity, marital status, and living arrangement), and daily living activity covariates. It indicates that the reciprocal cognition–depression relationship is not explained by these variables. This finding is especially important because previous studies have shown that cognition/depression has significant associations with these risk factors,^{101,102} in particular educational levels.¹⁰³ Our analysis rules out the third-variable effects caused by these covariates.

The bidirectional causal relationship between cognitive function and depressive symptoms could be explained by the cognitive theory of depression. It posits that people's thoughts, inferences, attitudes, and explanations, as well as the way they process and recall information, were associated with their risk for depression.¹⁰⁴ The relationship between biased cognitive processing and depressive disorder involves three mechanisms: the inhibitory process and defects of working memory, the rumination to negative emotions and events, and the inefficiency in regulating negative emotions.¹⁰⁴ Somatic symptoms of depression such as fatigue or attention deficits are also early indicators of cognitive decline.²⁰

It is worth noting that we only tested one possible mechanism of the reciprocal relationship between depression and cognition, and we found that the mediating effect of sleep quality, although statistically significant, can only explain a small portion of the direct effect. Therefore, we cannot determine whether this reciprocal effect over time is mainly caused by depression and cognition per se, or by their effects on potential psychosomatic processes, such as worsened

sleep quality, or other mechanisms. Previous research also notices that the process of reciprocal cognition-depression relationship might involve a “mediator chain” that includes many psychosomatic markers, such as cerebrovascular risk factors,¹⁰⁷ ruminations,¹⁰⁶ perceived stress and sleep quality.¹⁰⁵ The mediating effect of sleep quality found in this study may be just one of many steps between cognition and depression. Further research is needed to address this issue.

Several limitations of this study should be noted. First, a prerequisite for causal inferences in CLPM is that all possible false sources should be excluded. We have controlled a limited number of important covariates, but some unmeasured covariates, such as physical health status,¹⁰⁰ have not been adjusted. Second, Due to data collection limitations, sleep quality in this study might be underreported because this study only used a self-report question to assess sleep quality. Sleep quality is a complex construct to evaluate empirically, and yet self-reported subjective sleep quality is commonly used in studies as their only measurement. Further studies should include both subjective and objective measures when examining sleep quality in older adults.¹⁰⁸ Third, the five-item version of the depressive symptoms scale in this study was a crude measurement and contains only 5 items about symptoms of depression. However, these five items represent the most important aspects of depressive symptoms and have been used in many previous studies.^{77–80}

Conclusion

In summary, by adopting 10-year cross-lagged longitudinal data from the CLHLS demonstrated in Chinese OAs, we found that there was a negative bidirectional causality relationship between cognitive function and depressive symptoms, and the prospective cognition-depression relationship was partially mediated by sleep quality. Our results indicate lower cognitive function cause higher subsequent depressive symptoms partly through the worsening of sleep quality. Both depression and cognitive degeneration are major mental health challenges of contemporary Chinese OAs, and both are accompanied by sleep disturbance. Considering the reciprocal relationship between them, multidisciplinary interventions that try to relieve depression, maintain or delay the progress of cognitive degradation, or related risk factors, such as sleep quality, may contribute to the cognitive and mental health of OAs. More studies are needed to reveal the underlying pathways between depression and cognition to interrupt the possible vicious cycle between depression and cognitive degeneration.

Data Sharing Statement

All the datasets can be accessed at the Peking University Open Research Data after being authorized.

Disclosure

The authors report no conflicts of interest in this work.

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