The relationship between sleep, mental health, and performance on tests of pattern separation in young adults

Aina E Roenningen[1,2]ID, Devan Gill[1,3], and Brianne A Kent[1,2]ID

1. Department of Psychology, Simon Fraser University, Burnaby, Canada
2. Institute for Neuroscience and Neurotechnology, Simon Fraser University, Burnaby, Canada
3. Adler University, Clinical Psychology, 520 Seymour St, V6B 3J5, Vancouver, BC, Canada

Correspondence:

Brianne A Kent

bkent@sfu.ca

Simon Fraser University

Department of Psychology

8888 University Drive

Burnaby, BC

Canada

Abstract

Study objectives: Young adults experience the highest rates of mental health disorders of any age group. Depression and anxiety symptoms are often associated with both sleep disturbances and cognitive impairments. Here, we investigated whether the effects of sleep moderate the effects of depression and anxiety on cognitive performance. We were particularly interested in cognitive tasks designed to tax pattern separation, which we hypothesize are most sensitive to the cognitive impairments caused by mental health symptoms and sleep disturbances.

Methods: We recruited young adults (N=89; aged 18-30 years) and remotely monitored their sleep for 7 consecutive days using wrist actigraphy and daily sleep diaries. On day 7, participants completed in-person cognitive testing and mental health questionnaires. Cognitive tests included the Psychomotor Vigilance Task (PVT), Cambridge Neuropsychological Test Automated Battery (CANTAB), and the Mnemonic Similarity Task (MST). The CANTAB Delayed Matching to Sample (DMS) and the MST, are designed to tax pattern separation. Mental health questionnaires included Beck's Depression Inventory (BDI) and Beck's Anxiety Inventory (BAI).

Results: Eighty participants (mean age: 20.13 ± 2.00) were included in the final analyses. Depressive symptoms were significantly correlated with performance on the PVT and CANTAB DMS. Bedtime and wake-up times were significantly correlated with performance on DMS and MST.

Conclusion: Young adults experiencing depressive symptoms tend to have later wake-up times. Sleep timing and mental health independently affect performance on tests of pattern separation. Understanding the relationship between mental health, sleep, and cognitive performance is important for designing interventions to promote well-being in young adults.

Keywords: Sleep/wake, cognition, depression, anxiety, memory, actigraphy

1. Introduction

Mental health disorders can have significant impacts on quality of life, and increase the risk of illness, disability, and mortality [1,2]. Worldwide, one in eight people experience a mental health disorder, with depression and anxiety being the most prevalent conditions [3]. Depressive disorders include mood-related conditions characterized by persistent emotions of sadness, despair, and disinterest in once enjoyable activities [4]. Among the most prevalent of these disorders is Major Depressive Disorder (MDD), with a lifetime prevalence of 12% [5]. Anxiety disorders encompass mental health conditions characterized by persistent and intense emotions of fear and worry, often occurring alongside physical symptoms including restlessness and impaired concentration [4]. Common anxiety disorders such as Panic Disorder, Social Anxiety Disorder, and Generalized Anxiety Disorder (GAD) often co-occur with depression [6].

Although symptoms of anxiety and depression are complex and are expressed differently across individuals, sleep disturbances (e.g., insomnia or hypersomnia) are frequently reported in those who are diagnosed with anxiety and depression [7]. Those who are struggling with sleep disturbances or who experience disrupted sleep are also more likely to develop depressive and anxiety disorders [8,9], suggesting a bi-directional relationship between sleep and mental health. The relationship between sleep and mental health is particularly evident in college students, with the majority attaining insufficient amounts of sleep and reporting daytime sleepiness [10,11]. The sleep deficit experienced by the majority of college students has been associated with poorer mental health outcomes [7,12,13].

Mental health disorders (e.g., anxiety and depression) and sleep disturbances have been linked to poor cognitive performance [14,15]. For example, research has shown that anxiety, depression, and sleep disturbances are associated with episodic memory deficits, including

impaired performance on tests of pattern separation [16,17]. Pattern separation is a computational process thought to enable the formation of distinct neural representations from similar inputs [18,19], and is critically important for effective episodic memory. Anxiety, depression, and sleep disturbances have been found to impact pattern separation [17,20-22], possibly through suppression of neural plasticity in the hippocampus, such as hippocampal neurogenesis, which in rodent models has been shown to be critical for pattern separation [17,20].

As younger adults frequently experience depressive and anxiety symptoms, the present research aimed to explore whether depression and anxiety affect cognitive performance, including on tests designed to assess pattern separation, and whether sleep moderates the effect on cognition. If sleep is shown to moderate the effect of mental health disorders on cognition, then interventions to promote sleep should be prioritized in individuals experiencing cognitive impairments associated with depression or anxiety.

2. Methods

2.1 Participants

The study received approval from the Research Ethics Board (REB) at Simon Fraser University (SFU) under Protocol #30000539, adhering to the Tri-Council Policy Statement on Ethical Conduct for Research Involving Humans. Subjects were 89 undergraduate SFU students recruited via the SFU Psychology Research Participation System, and through advertisements and payment. Prior to any testing, participants completed an online consent form and demographics survey, which collected demographic information as well as information about mental, neurological, and physical health. We recruited individuals who were between 18 - 30

years of age, and they needed to understand written and oral English to participate. Participant demographic information is presented in Table 1.

2.2 Procedure

Participants in the study attended two in-person visits, scheduled 7 days apart between 10:00 am and 3:00 pm. During the first visit, participants were provided with wrist actigraphy to monitor their rest-activity over the week. On the seventh day, they returned to the lab for the second visit, during which they completed cognitive testing and filled out two mental health questionnaires (Figure 1).

2.3 Sleep Assessments

Participants wore wrist actigraphy (ActTrust2 model AT20101, Condor Instruments, Sao Paulo, Brazil) on their non-dominant hand and filled out an electronic sleep diary, modified from Lockley [23], over seven consecutive days. They were instructed to use the event button on the wrist actigraphy to record the time they went to bed and the time they woke up each morning.

2.4 Cognitive testing

2.4.1 Psychomotor Vigilance Task (PVT)

The PVT is a measure of reaction time and sustained attention [24,25]. We used the 5-min PVT version (Version 1.2.0., NASA, Washington, D.C., USA) [26,27]. Participants completed the task on an Apple iPad (OS 15.4.1, model A2602). In a small rectangular area, a millisecond counter was presented and appeared at random inter-stimulus intervals (ISI), in the range of 2 to 10 s [28]. The millisecond counter rapidly increased from zero in each trial. We instructed the participants to tap their dominant hand's thumb on the screen as fast as possible once they detected the millisecond counter. Participants completed practice trials before the experimental trials.

2.4.2 Mnemonic Similarity Task (*MST*)

The MST (version 0.96) is a cognitive task designed to assess pattern separation [29]. We used the computerized task pace, the 3-choice version, and Set 1 stimuli for the MST. The task included two phases. During phase 1 of the task, participants were presented with pictures of everyday objects. They were instructed to indicate via keyboard presses whether the object that appeared belonged indoors or outdoors [29,30]. We did not inform participants about the upcoming memory test in phase 2. Phase 2 had 192 trials and participants were shown pictures of everyday objects. This time, we instructed the participants to respond "old," "similar," or "new" to indicate whether the objects were a) the exact same as an object seen in phase 1 (i.e., old/target), b) similar to an object seen in phase 1 (i.e., similar/lure), or c) a new image not shown in phase 1 (i.e., new/foil). In the 192 trials, one-third (64) were targets, one-third (64) were lures, and one-third (64) were foils. The lures included 5 levels of similarity. The similarity levels ranged from L1 (highly similar) to L5 (least similar). In both phases of the task, all pictures were shown with a 2000 ms duration. There was a 500 ms inter-stimulus interval between the presentation of each image.

2.4.3 Cambridge Neuropsychological Test Automated Battery (CANTAB)

Participants completed 5 cognitive tasks from CANTAB, administered in the following order: Delayed Matching to Sample (DMS), Motor Screening Task (MOT), Paired Associates Learning (PAL), Reaction Time (RTI), and Spatial Working Memory (SWM). All tests were conducted using an iPad running OS 15.4.1.

DMS. The DMS assessed visual matching and short-term visual recognition memory for abstract and non-verbalizable patterns [31]. A complex sample pattern was shown for 4.5 s, followed by 4 choice patterns that had different structures and colors. Participants were asked to

identify the choice pattern that was identical to the sample pattern. The sample pattern and choice patterns either appeared simultaneously or there was a delay between 0, 4, or 12 s before the choices were shown. *PAL.* The PAL task measured new learning and visual episodic memory [32]. A number of boxes were shown on the screen. Randomly, two or more of the boxes opened one by one and presented a pattern for 2 s. Patterns were then shown, one at a time, in the screen center, and the participants were instructed to select the box in which the pattern had been displayed. *RTI.* The RTI test measured divided attention and psychomotor speed [33]. The participants were instructed to hold down a button at the bottom of the screen, followed by quickly tapping inside one of 5 circles at the top of the screen once a yellow stimulus flash appeared. They used the same finger to tap the flashing circle. *SWM.* The SWM task measured manipulation and retention of visuospatial information [34]. The screen showed a number of boxes, where a few contained a hidden token. The participants clicked the empty boxes to locate the tokens and were asked not to go back to a box where one had already been located. Trials increased in difficulty (i.e., number of boxes to search).

2.5 Mental health questionnaires

2.5.1 Beck Depression Inventory (BDI)

The BDI assessed symptoms associated with depression over the past two weeks including cognitive (e.g., concentration, pessimism, decisiveness), somatic (e.g., libido, appetite, sleep), and affective symptoms (e.g., sadness, crying, irritability). Each question can be rated in severity from zero to three, where three indicates the highest symptom severity [35]. Adding the scores together provides a total score between 0 and 63. Scores from 0 to 9 indicate minimal depression, 10 to 18 indicate mild depression, 19 to 29 indicate moderate depression, and 30 to 63 indicate severe depression [36].

2.5.2 Beck Anxiety Inventory (BAI)

The BAI assessed anxiety symptoms within the past month. The same assessment scale is used as in the BDI, with four statements in each item, and with a score of three indicating the highest anxiety symptom [37]. The items measure symptoms including physical (e.g., feeling hot, dizziness, difficulty breathing, increased heartrate, sweats), affective (e.g., feelings of fear, nervousness, being scared) and cognitive symptoms (e.g., catastrophizing thoughts). The scores from each item are added together to provide a total score between 0 and 63 [37]. Minimal anxiety is indicated by scores between 0 to 7, scores between 8 to 15 indicate mild anxiety, scores between 16 to 25 indicate moderate anxiety, and scores between 26 to 63 indicate severe anxiety.

3. Data Analysis

3.1 Actigraphy

Actigraphy data were analyzed using ActStudio (version 2.2.0, Condor Instruments, 2022). Only nocturnal sleep was assessed. The sleep parameters estimated were: 1) total sleep time, 2) sleep efficiency (i.e., percentage of time spent asleep of the total time in bed), 3) bedtime, 4) and wake-up time. A minimum of 5 nights of complete actigraphy data were required for each participant to be included in the analyses. The sleep diaries were used to confirm bed and wake-up times. Additionally, we used Clocklab (Version 6.1.11, Actimetrics, Wilmette, USA) to compute a cosinor analysis. Cosinor analysis was used to estimate amplitude, acrophase, intradaily variability (IV), and interdaily stability (IS) of the participants' daily activity rhythms.

3.2 PVT

The performance metric calculated for the 5-minute PVT was the slowest 10% reaction times, which is one of the most sensitive PVT measures to sleep disturbance [26,38].

3.3 MST

The Memory Recognition scores (REC) for old (i.e., repeat) images were calculated by subtracting the rate of "old" responses provided to foils from the rate of "old" responses provided to repeated images [p(correct old response to lures) - p(false old response to foils)] [39]. The Lure Discrimination Index (LDI) was calculated by subtracting the probability value of "similar" responses provided to the foils from the probability value of "similar" responses provided to the lures [p(correct similar response to lures) - p(false similar response to foils)] [39]. We also assessed performance on L1 and L2 specific lure bins. These lure images are most similar to target images, and thus tax pattern separation the most. For these specific lure bins, we assessed *Accuracy*, as the likelihood judging the lure as similar, and *False Memory Error Rate*, as the likelihood of incorrectly judging the lure as a target image.

3.4 CANTAB

The following performance metrics were used in the analyses:

DMS. We used DMS Pattern Errors (All Delays), which was calculated across all trials with a delay component [40]. The total score was derived from the number of times the participant initially selected the distractor stimulus with different patterns but that had identical color elements. *PAL.* We used the PAL First attempt and PAL Adjusted Errors. PAL First Attempt assessed the number of times a participant chose the correct box on their first attempt while recalling all pattern locations. PAL Adjusted Errors measured the number of times a participant chose the incorrect box, adding an adjustment to allow for a comparison of error rates across all participants [40]. *RTI.* We used the RTI Mean Five-Choice Movement Time, which measured the average time it took for a participant to release the response button and tap the target stimulus [40]. *SWM.* We used SWM Between Errors and SWM Strategy. The SWM Between Errors

assessed the number of times a participant re-opened a box where they had previously found a token [40]. The SWM Strategy measured the number of times the participant used the same or different search pattern in trials with 6-8 boxes.

3.5 Mental health

We used the total scores of BDI and BAI in the analyses. Both BDI and BAI had a total score of 63 points.

3.6 Statistical analyses

The relationships between sleep, mental health, and cognition were analyzed using multiple linear regressions in JMP 18.2.1 (JMP Statistical Discovery LLC, North Carolina, USA). Correlational analyses assessed relationships before performing the moderation analyses. We ran Spearman's Rank correlations (ρ) when data variables showed non-normality. In the regressions, the predictors included mental health, and the moderators included objective sleep measures (Figure 2). The relationships between the variables were linear shown by residual plots. The residuals were normally distributed, there was constant variance, and observations were independent shown by the Durbin Watson's test ($p > .05$). Thus, the assumptions for analysis were met. The cut-off used for statistical significance was $p < 0.05$ (two-tailed).

4. Results

4.1 Sample characteristics

Of the 89 participants, 9 were excluded from all analyses due to unreliable or missing sleep data ($n = 5$), external distractions during in-person testing session ($n = 1$), malfunction or incorrect task administration on both MST and CANTAB ($n = 1$), being outside of recruitment age range ($n = 1$), or missing mental health data ($n = 1$). The remaining participants included 80 participants (mean age: 20.13 ± 2.00) who were included in the analyses.

For the task-specific analyses, an additional participant was excluded from PVT analyses due to reporting major distractions during task performance. Eleven participants were excluded from MST analyses due to MST malfunction and incorrect test administration (n = 5), low effort or not understanding the task indicated by underuse of the "similar" response (n = 2) [41], not complying with task instructions (n = 2), and a REC score below the cut-off value of .50 (n = 2). Five younger adults were excluded from CANTAB analyses due to technical error (n = 2) and not complying with the protocol (n = 3). Twenty-two younger adults were excluded from Clocklab analyses as they did not have 5 or more days of continuous wrist actigraphy wearing (i.e., removed wrist actigraphy during the day).

4.2 Depression

Of the 80 participants, 43.75% of the students scored within 0 and 9, indicating minimal depression, 38.75% scored within 10 to 18 indicating mild depression, 11.25% scored within 19 to 29 indicating moderate depression, and 6.25% scored within 30 to 63 indicating severe depression [36].

4.3 Anxiety

Of the 80 participants, 33.75% scored between 0 - 7 on the BAI, indicating minimal anxiety, 26.25% scored between 8 - 15, indicating mild anxiety, 18.75% scored between 16 – 25, indicating moderate anxiety, and 21.25% scored 26 or higher, indicating severe anxiety [37]. The depression and anxiety scores were strongly correlated ($\rho(78) = .68$, $p < .0001$), indicating comorbidity.

4.4 Symptoms of depression were associated with later wake-up times

There was a significant correlation between BDI score and wake-up time ($\rho(78) = .35$, $p = .002$) (Figure 3). There were no other significant correlations between the objective sleep

measures or circadian parameters and mental health scores ($p > 0.05$). See Supplementary Table S1 for the correlations.

4.5 Symptoms of depression were associated with cognitive performance

There was a significant correlation between BDI total score and PVT 10% slowest reaction time ($\rho(77) = .26$, $p = .02$). See Supplementary Table S1 for the results. There was a significant correlation between BDI total score and CANTAB DMS ($\rho(71) = .24$, $p = .04$). No other correlations were significant ($p > 0.05$) (MST; Figure 4 and 5). See Supplementary Table S1 for the remaining correlations between mental health and cognitive performance.

4.6 Bedtime and wake-up time were significantly correlated with performance on tests of pattern separation

There was a correlation between DMS Pattern Errors (All Delays) and bedtime ($\rho(73) = .24$, $p = .04$) and wake-up time ($\rho(73) = .29$, $p = .01$). There were also significant correlations between LDI and bedtime ($r(67) = -.29$, $p = .01$) (Figure 6a) and wake-up time ($r(67) = -.29$, $p = .02$) (Figure 6b). There were significant correlations between bedtime and L2 Accuracy ($r(67) = -.26$, $p = .03$) (Figure 6m) and L2 False Memory Error Rate ($r(67) = .24$, $p = .050$) (Figure 6u). See Supplementary Table S2 for the remaining findings. In the circadian analyses, L1 Accuracy was significantly associated with later acrophase ($r(67) = -.29$, $p = .03$). See Supplementary Table S3 for the remaining results.

4.7 Sleep did not moderate the effect of mental health symptoms on cognitive performance

Sleep did not moderate the relationships between mental health (i.e., BDI and BAI) and cognitive performance ($p > .05$). When BDI was in the model, there were only direct effects of wake-up time (B = -.04, $p = .02$, 95% CI = -.07, .01) and bedtime (B = -.03, $p = .02$, 95% CI = -.06, .00) on LDI. There were also direct effects of wake-up time on LDI (B = -.03, $p = .02$, 95%

CI = -.06, .00) and DMS (B = .26, *p* = .02, 95% CI = .04, .47), and of bedtime on LDI (B = -.03, *p* = .02, 95% CI = -.06, .00), L2 Accuracy (B = -.03, *p* = .02, 95% CI = -.07, .00), and L2 False Memory Error Rate (B = .03, *p* = .02, 95% CI = -.00, .06), when BAI was in the model. See Supplementary Table S4 for the remaining results.

5. Discussion

It is well-documented that sleep disturbances, anxiety, and depression can independently and negatively affect cognitive functions [14-17]. In the present study, we aimed to investigate whether the sleep disturbances associated with symptoms of anxiety and depression in young adults moderates the relationship between mental health and cognitive performance. We were particularly interested in cognitive tasks designed to assess pattern separation. We predicted that the relationship between cognitive performance, anxiety, and depression, would be moderated by sleep quantity and quality.

We found that more than 50% of the students were in the range of mild to severe depression, and 25% met the threshold for moderate to severe anxiety. Research shows that depressive and anxiety symptoms are prevalent in college students and can have impairing effects on academic performance [42]. Symptoms of depression and anxiety also show a relationship with poor sleep, demonstrated in multiple studies [43-45]. In our study, we found a relationship with depression and late bed and wake-up time, which supports previous studies [46-50].

Bedtime and wake-up time showed correlations with cognitive performance, specifically, performance on MST and DMS, which are tasks designed to assess pattern separation. The sleep parameters did not moderate the relationship between depression and anxiety, and cognitive performance.

The relationships between depression and cognitive performance support previous research [17]. Becker et al. (2009) found that participants with a high score on the BDI showed impaired performance on DMS trials (i.e., DMS Correct Latency) with a longer delay compared to shorter delay trials, where their performance did not seem to be affected. We found that individuals scoring higher on the BDI showed more DMS pattern errors (i.e., remembered and selected the incorrect pattern during test trials) compared to those who had a lower BDI score. Individuals experiencing prolonged stress and/or depressive symptoms may have suppressed hippocampal plasticity, such as neurogenesis, which is important for optimal performance in tasks that tax pattern separation processes [17,22]. This is only a speculation as we cannot measure neurogenesis in humans directly. We also did not find effects of depression on MST performance, which contradicts some previous research [51-54].

Our study has limitations. We did not use a clinical assessment of depression or anxiety but relied on self-report using the BDI and BAI. We used actigraphy and sleep diaries to assess sleep, instead of polysomnography, which is the gold standard sleep assessment. Our sample included university students who may have different sleep patterns compared to other young adults who are not attending university [10]. University students may also experience stressors that are different from those experienced by other young adults, thus, the relationship between sleep, cognition, and mental health that we see in this sample may not generalize to the population. We also included limited control of extraneous variables in this study, thus, effects from medications, caffeine, and different demographic variables may also have affected the results. Therefore, the results need to be interpreted with caution.

6. Conclusion

To summarize, our findings suggest that there may be independent effects of mental health and sleep on cognitive performance in young adults. Future research is needed to measure whether cognitive tests designed to assess pattern separation are sensitive enough to identify improvements with sleep-targeting interventions and mental health therapeutics.

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Data availability

Data will be available upon request.

Conflict of Interest Statement

The authors listed in this manuscript confirm that they have no affiliations or involvement with any organization or entity that has a non-financial interest (e.g., personal or professional affiliations, relationships, beliefs, or knowledge) or financial interest (e.g., honoraria, stock ownership, expert testimony, or patent-licensing arrangements) that are related to the materials and content presented and discussed in this manuscript.

Disclosure Statement

a) Financial disclosure: There are no financial conflicts of interest to disclose.

b) Non-financial disclosure: The authors state that their submission has been posted as a preprint.

Ethics approval statement

The research study was reviewed and approved by the Research Ethics Board at Simon Fraser University (Protocol #30000539). All participants provided written informed consent to participate in the research. The privacy rights of the participants were observed.

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Tables

Table 1. Participant characteristics

	Sample
Age	
Age range	18-30
Mean	20.13 ± 2.00
Sex	
Female	49
Male	31
Education	
>12 years of education	100%
Ethnicity	
Hispanic	2
Middle Eastern	9
Black	2
Asian	42
White	19
Mixed ethnicity	6
Medical history	
Mental health disorder	
Anxiety	8
Depression	4
ADHD	3
PTSD	1
Comorbid	6
Not specified	1
Neurological condition	
Concussion	4
Dysautonomia	1
Sleep disorder	
Insomnia	1
Comorbid	1

Consumption of psychoactive substances	
Caffeine	
≥ Once a day	
≥ Once a week	17
Nicotine	15
≥ Once a day	
≥ Once a week	2
Marijuana	1
≥ Once a day	
≥ Once a week	2
≥ Once a month	4
Alcohol	4
≥ Once a week	
≥ Once a month	10
Not specified	6
	7
Time-zone crossing within the past month	
≥1 time-zones:	
Not specified	0
	4
Working nightshifts:	
Currently:	8
In the past:	11

Table 2. Summary of sleep and circadian rhythm parameters

	Bedtime	Wake-up time	TST (h)	Sleep efficiency (%)	SOL (h)	WASO (h)	Amplitude	Acrophase (h)	IV	IS
	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD	Mean ± SD
Participants	01.52 ± 1.49	8.96 ± 1.37	6.64 ± 0.83	89.51 ± 5.00	0.06 ± 0.02	0.68 ± 0.39	1.63 ± 0.59	17.18 ± 1.81	1.03 ± 0.29	0.39 ± 0.10

Abbreviations: TST, Total Sleep Time; SOL, Sleep Onset Latency; WASO, Wake After Sleep Onset; IV, Intradaily Variability; IS, Interdaily Stability. Bedtime, Wake-up time, TST, SOL, WASO, and acrophase are displayed in decimal time.

Figures

Figure 1. Illustration of the procedure, including two example images used in the MST.[29] Created with BioRender.com. Abbreviations: MST, Mnemonic Similarity Task; DMS, Delayed Matching to Sample; BDI, Beck's Depression Inventory; BAI, Beck's Anxiety Inventory.

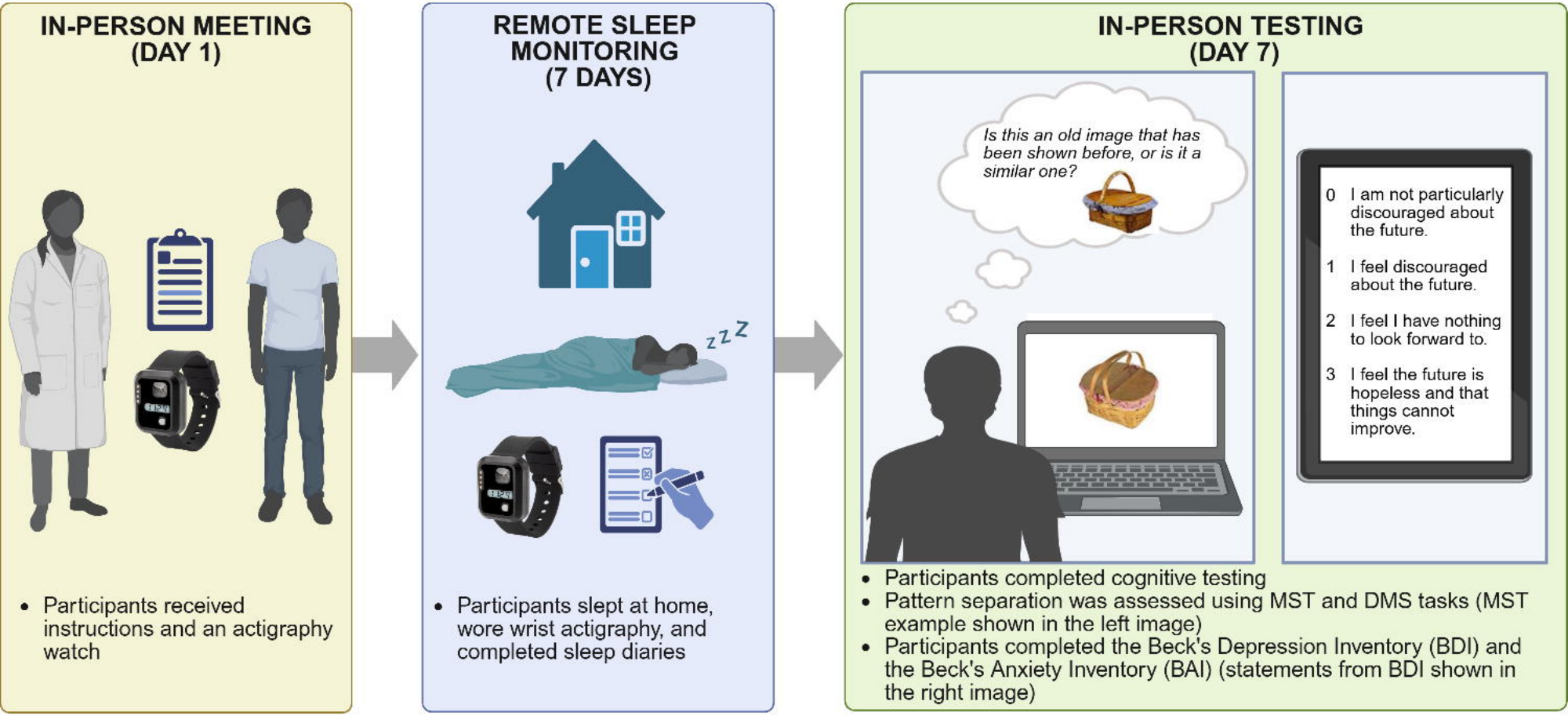
Figure 2. Illustration of the hypothesized moderation relationship between mental health, sleep, and cognitive performance. Abbreviations: BDI, Beck's Depression Inventory; BAI, Beck's Anxiety Inventory; TST, total sleep time; MST, Mnemonic Similarity Task; CANTAB, Cambridge Neuropsychological Test Automated Battery.

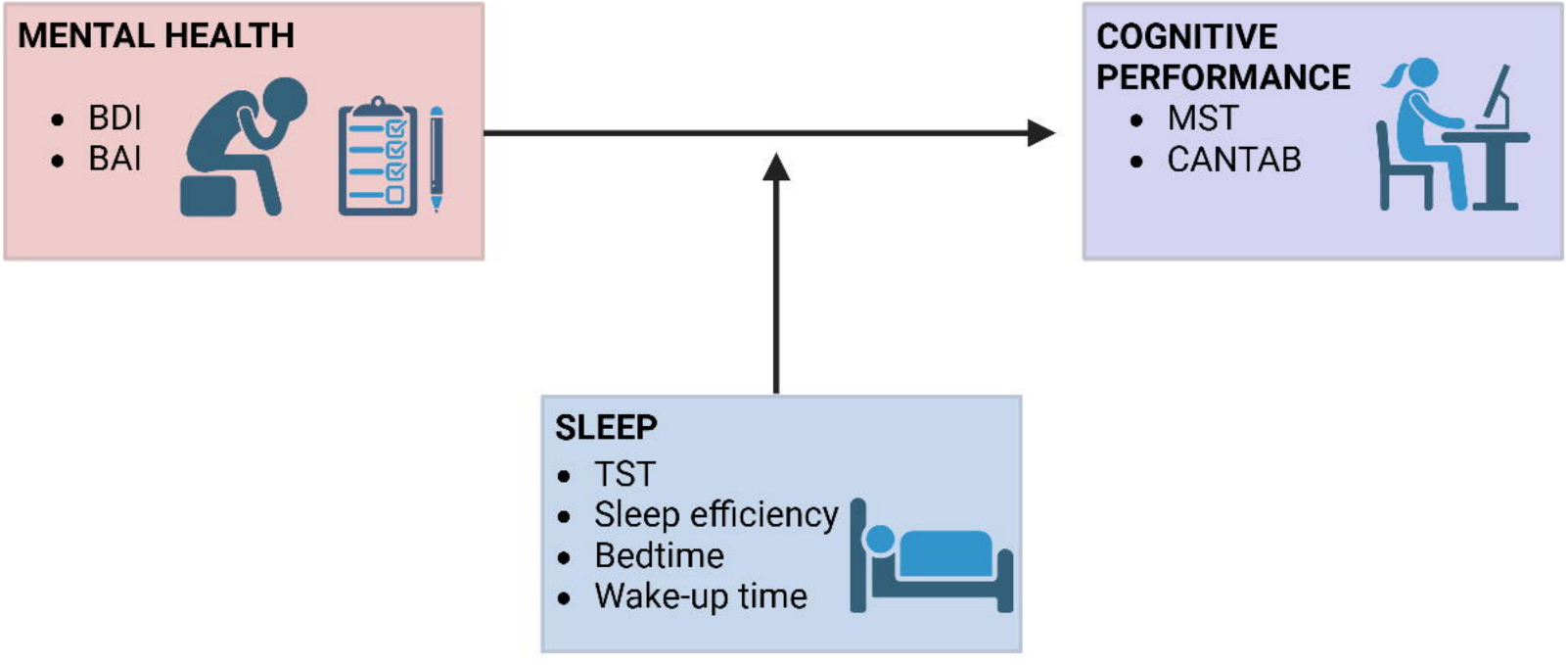
Figure 3. Scatterplots showing the relationship between mental health total scores and sleep: Bedtime (a, b), Wake-up time (c, d), TST (e, f), and sleep efficiency (g, h). Each black dot represents an individual participant. Spearman's (ρ) correlation, * = $p < 0.05$. Abbreviations: TST, total sleep time; BDI, Beck's Depression Inventory; BAI, Beck's Anxiety Inventory.

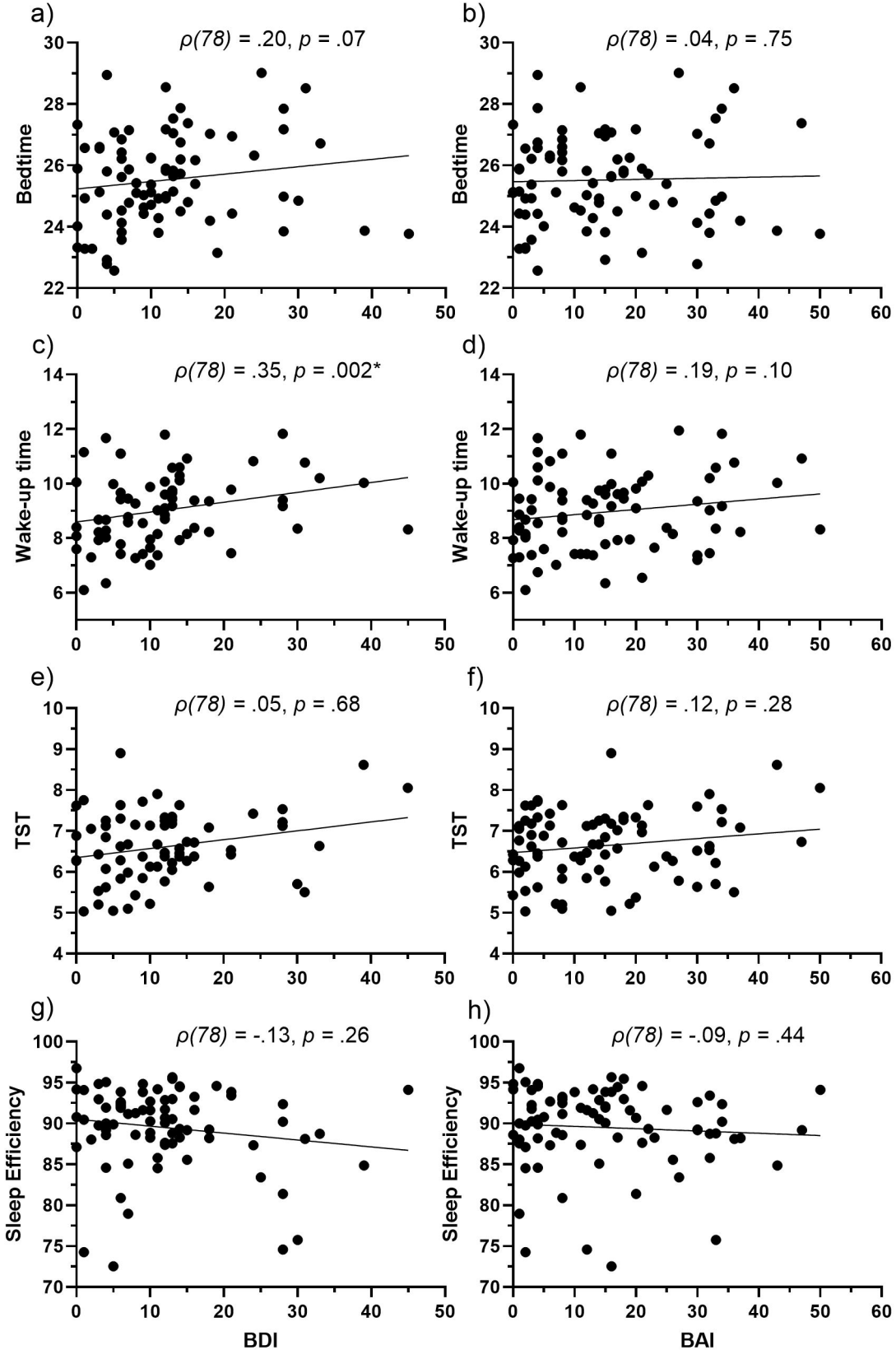
Figure 4. Scatterplots showing the relationship between BDI total score and MST performance: LDI (a), REC (b), L1 Accuracy (c), L2 Accuracy (d), L1 False Memory Error Rate (e), and L2 False Memory Error Rate (f). Each black dot represents an individual participant. Spearman's (ρ) correlation, * = $p < 0.05$. Abbreviations: BDI, Beck's Depression Inventory; LDI, Lure Discrimination Index; REC, Recognition Memory; L1, Lure Bin 1; L2, Lure Bin 2.

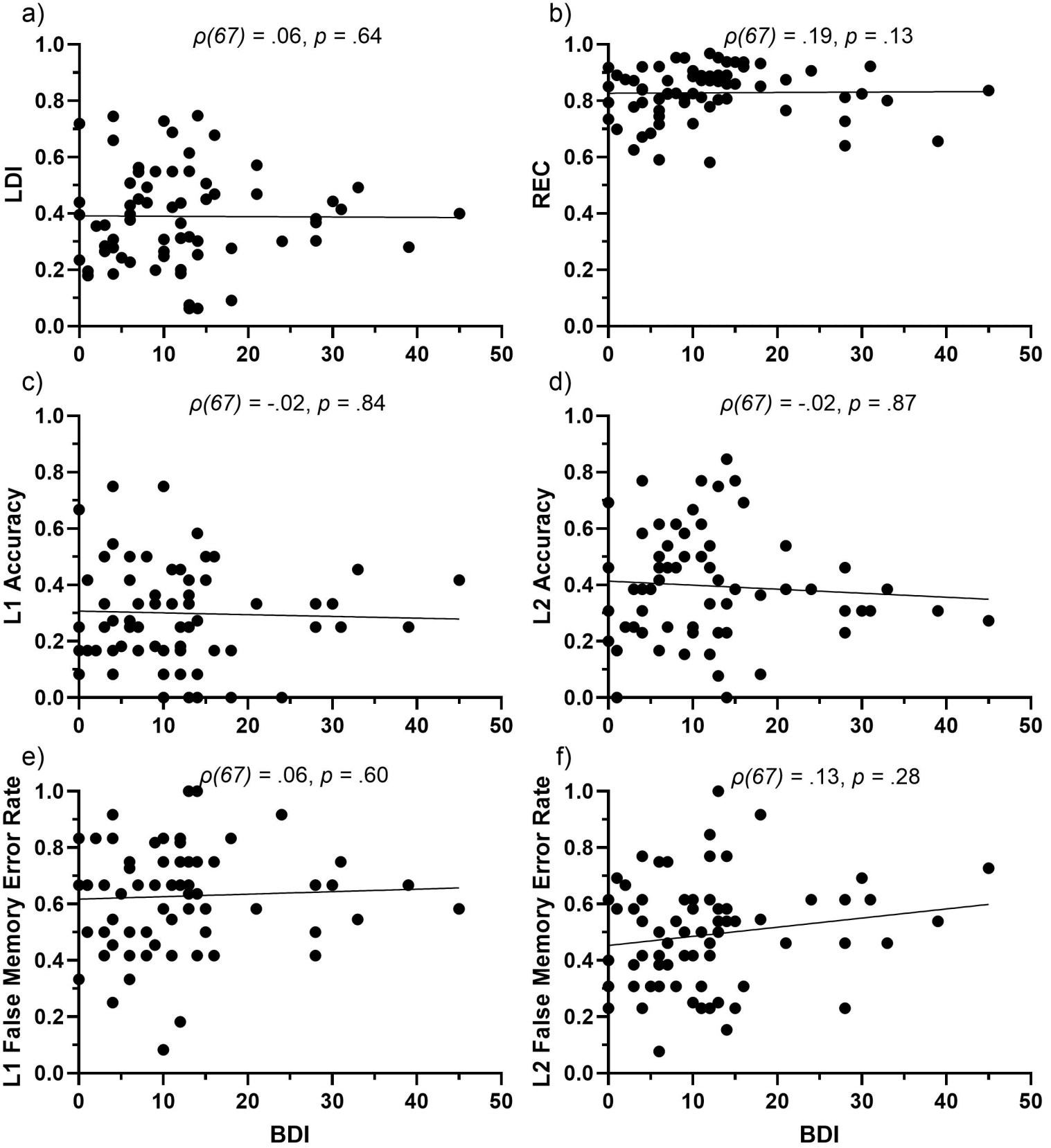
Figure 5. Scatterplots showing the relationship between BAI total score and MST performance: LDI (a), REC (b), L1 Accuracy (c), L2 Accuracy (d), L1 False Memory Error Rate (e), and L2 False Memory Error Rate (f). Each black dot represents an individual participant. Spearman's (ρ) correlation, * = $p < 0.05$. Abbreviations: BAI, Beck's Anxiety Inventory; TST, total sleep time; LDI, Lure Discrimination Index; REC, Recognition Memory; L1, Lure Bin 1; L2, Lure Bin 2.

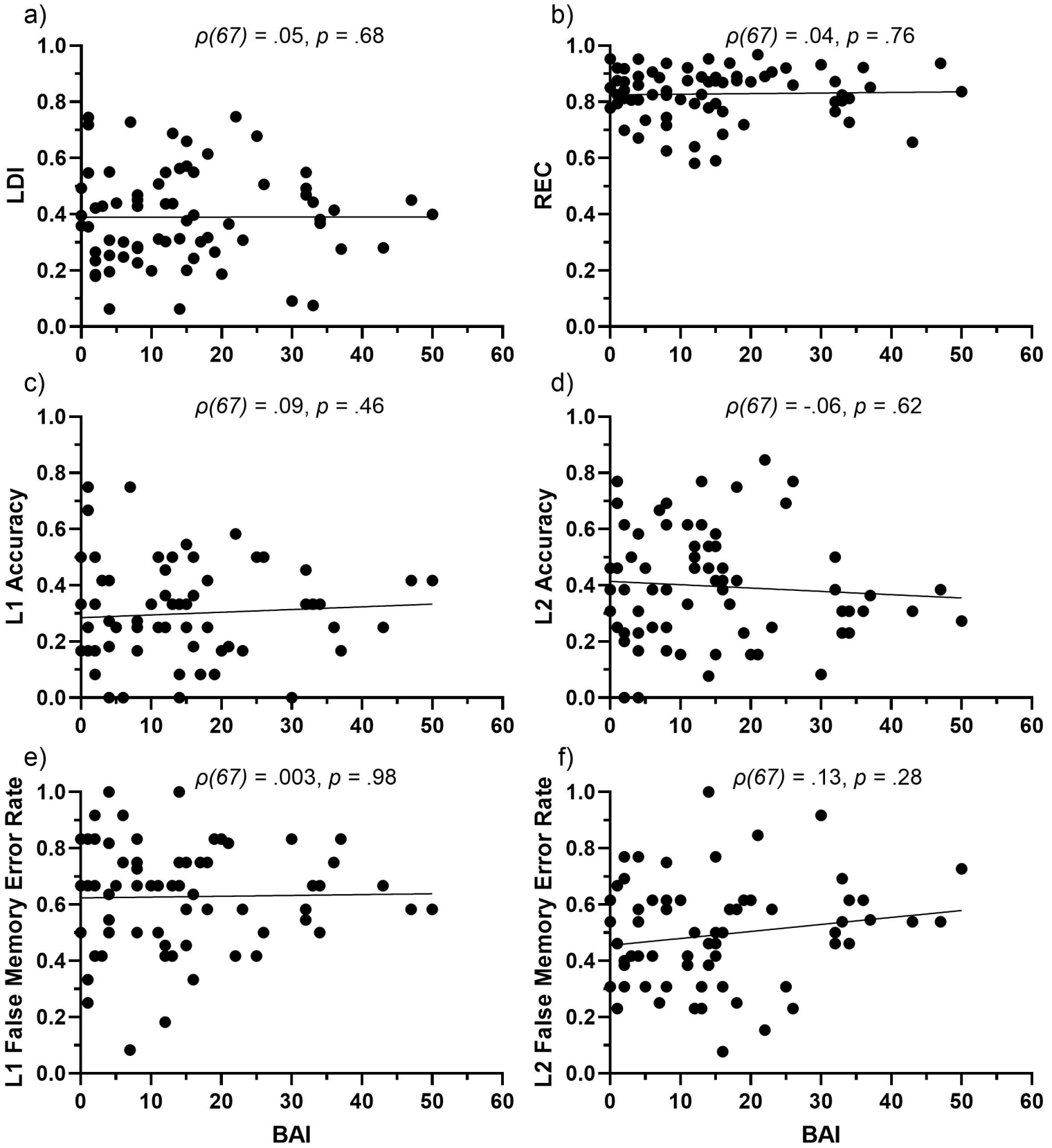
Figure 6. Scatterplots showing the relationship between sleep and MST performance: LDI (a, b, c, d), REC (e, f, g, h), L1 Accuracy (i, j, k, l), L2 Accuracy (m, n, o, p), L1 False Memory Error Rate (q, r, s, t), and L2 False Memory Error Rate (u, v, w, x). Each black dot represents an individual participant. Spearman's (ρ) correlation, * = $p < 0.05$. Abbreviations: TST, total sleep time; LDI, Lure Discrimination Index; REC, Recognition Memory; L1, Lure Bin 1; L2, Lure Bin 2.

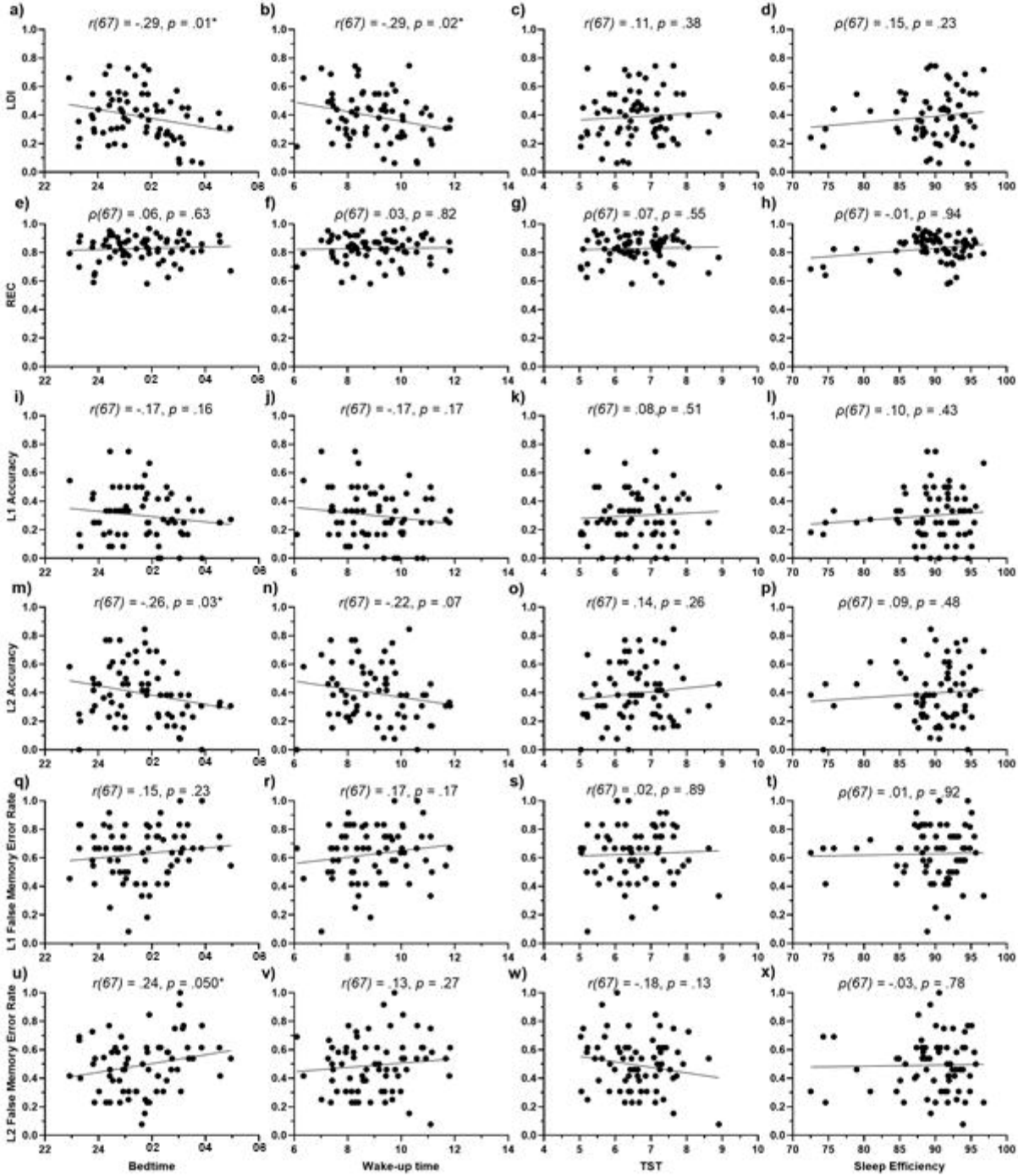

IN-PERSON MEETING
(DAY 1)
• Participants received instructions and an actigraphy watch
REMOTE SLEEP MONITORING
(7 DAYS)
• Participants slept at home, wore wrist actigraphy, and completed sleep diaries
IN-PERSON TESTING
(DAY 7)
Is this an old image that has been shown before, or is it a similar one?
0 I am not particularly discouraged about the future.
1 I feel discouraged about the future.
2 I feel I have nothing to look forward to.
3 I feel the future is hopeless and that things cannot improve.
• Participants completed cognitive testing
• Pattern separation was assessed using MST and DMS tasks (MST example shown in the left image)
• Participants completed the Beck's Depression Inventory (BDI) and the Beck's Anxiety Inventory (BAI) (statements from BDI shown in the right image)


MENTAL HEALTH
BDI
BAI
COGNITIVE PERFORMANCE
MST
CANTAB
SLEEP
TST
Sleep efficiency
Bedtime
Wake-up time


a)
ρ(78) = .20, p = .07
Bedtime
b)
ρ(78) = .04, p = .75
Bedtime
c)
ρ(78) = .35, p = .002*
Wake-up time
d)
ρ(78) = .19, p = .10
Wake-up time
e)
ρ(78) = .05, p = .68
TST
f)
ρ(78) = .12, p = .28
TST
g)
ρ(78) = -.13, p = .26
Sleep Efficiency
BDI
h)
ρ(78) = -.09, p = .44
Sleep Efficiency
BAI


a)
ρ(67) = .06, p = .64
LDI
b)
ρ(67) = .19, p = .13
REC
c)
ρ(67) = -.02, p = .84
L1 Accuracy
d)
ρ(67) = -.02, p = .87
L2 Accuracy
e)
ρ(67) = .06, p = .60
L1 False Memory Error Rate
f)
ρ(67) = .13, p = .28
L2 False Memory Error Rate
BDI
BDI


a)
ρ(67) = .05, p = .68
LDI
b)
ρ(67) = .04, p = .76
REC
c)
ρ(67) = .09, p = .46
L1 Accuracy
d)
ρ(67) = -.06, p = .62
L2 Accuracy
e)
ρ(67) = .003, p = .98
L1 False Memory Error Rate
f)
ρ(67) = .13, p = .28
L2 False Memory Error Rate
BAI
BAI


a) r(67) = -.29, p = .01*
b) r(67) = -.29, p = .02*
c) r(67) = .11, p = .38
d) ρ(67) = .15, p = .23
e) ρ(67) = .06, p = .63
f) ρ(67) = .03, p = .82
g) ρ(67) = .07, p = .55
h) ρ(67) = -.01, p = .94
i) r(67) = -.17, p = .16
j) r(67) = -.17, p = .17
k) r(67) = .08, p = .51
l) ρ(67) = .10, p = .43
m) r(67) = -.26, p = .03*
n) r(67) = -.22, p = .07
o) r(67) = .14, p = .26
p) ρ(67) = .09, p = .48
q) r(67) = .15, p = .23
r) r(67) = .17, p = .17
s) r(67) = .02, p = .89
t) ρ(67) = .01, p = .92
u) r(67) = .24, p = .050*
v) r(67) = .13, p = .27
w) r(67) = -.18, p = .13
x) ρ(67) = -.03, p = .78
LDI
REC
L1 Accuracy
L2 Accuracy
L1 False Memory Error Rate
L2 False Memory Error Rate
Bedtime
Wake-up time
TST
Sleep Efficiency