Polygenic evidence and overlapped brain functional connectivities for the association between chronic pain and sleep disturbance

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Abstract

Chronic pain and sleep disturbance are highly comorbid disorders, which leads to barriers to treatment and significant healthcare costs. Understanding the underlying genetic and neural mechanisms of the interplay between sleep disturbance and chronic pain is likely to lead to better treatment. In this study, we combined 1206 participants with phenotype data, resting-state functional magnetic resonance imaging (rfMRI) data and genotype data from the Human Connectome Project and two large sample size genome-wide association studies (GWASs) summary data from published studies to identify the genetic and neural bases for the association between pain and sleep disturbance. Pittsburgh sleep quality index (PSQI) score was used for sleep disturbance, pain intensity was measured by Pain Intensity Survey. The result showed chronic pain was significantly correlated with sleep disturbance (r = 0.171, p-value < 0.001). Their genetic correlation was r_g = 0.598 using linkage disequilibrium (LD) score regression analysis. Polygenic score (PGS) association analysis showed PGS of chronic pain was significantly associated with sleep and vice versa. Nine shared functional connectivity (FCs) were identified involving prefrontal cortex, temporal cortex, precentral/postcentral cortex, anterior cingulate cortex, fusiform gyrus and hippocampus. All these FCs mediated the effect of sleep disturbance on pain and seven FCs mediated the effect of pain on sleep disturbance. The chronic pain PGS was positively associated with the FC between middle temporal gyrus and hippocampus, which further mediated the effect of chronic pain PGS on PSQI score. Mendelian randomization analysis implied a possible causal relationship from chronic pain to sleep disturbance was stronger than that of sleep disturbance to chronic pain. The results provided genetic and neural evidence for the association between pain and sleep disturbance, which may inform future treatment approaches for comorbid chronic pain states and sleep disturbance.

Introduction

Sleep disturbance and chronic pain are two common disease states which are bi-directionally interrelated with reciprocal interactions. Each of them occurs with great frequency and results in significant public health and socioeconomic burden. There is considerable evidence linking pain to disturbed sleep. Sleep disturbance is present in 67–88% of chronic pain patients, and at least 50% of individuals with insomnia suffer from chronic pain. It’s crucial to identify overlapped genetic and neural pathways correlated to two symptoms to explore etiological contribution towards the reciprocal relationship.

Genetics has been shown to explain as much as 50% of the variance in pain syndromes and 25%–45% for...
Insomnia \cite{8}. High genetic correlation ($r_g = 0.69$) between pain and sleep disturbance has been reported using twins sample \cite{9}, which implies a possible overlap in the genes affecting both sleep disturbance and chronic pain \cite{10}. Genome-wide association studies (GWAS) have shown that chronic pain was associated with several genes involved in brain function and development and with a range of psychiatric traits \cite{11}. GWAS for sleep disturbance related traits \cite{12} and insomnia \cite{13, 14} have identified several significant loci and shared genetics with neuropsychiatric and metabolic traits. The large sample size GWAS data provided new ways to assess pleiotropy and genetic correlations between two related traits from a polygenic perspective. Multiple methods have been developed to accomplish this, including linkage disequilibrium (LD) score regression, which uses GWAS summary data to calculate genetic correlations between two traits \cite{15, 16}, polygenic score (PGS) association analysis \cite{17}, which uses GWAS summary data of a trait to test whether the effect of a set of single-nucleotide polymorphisms (SNPs) is associated with another phenotype in an independent sample \cite{18}. To our knowledge, the genetic correlation between sleep and pain has not been examined using GWAS data. The genetic correlation between pain and sleep disturbance requires further exploration especially from a polygenic perspective to reveal the mechanisms involved in this relationship.

Genetic association between chronic pain and sleep disturbance makes it especially interesting to search for possible neural mechanisms that may mediate the association. Neuroimaging exploration of the brain mechanisms underlying a bidirectional relationship between sleep disturbance and pain has been dominated by studies mapping the effects of sleep deprivation on pain responsivity \cite{19-23}. Hyperalgesia engendered by sleep loss involves increasing pain reactivity associated with heightened somatosensory cortex activity and a corresponding decrease in insula and striatum activity \cite{22}. Understanding the neural mechanisms underlying the interrelationship of sleep disturbance and chronic pain is a key factor in the development of new therapies, and network-based methods \cite{24-26} are useful for attaining the goal. Studies have shown that connectivity between the dorsal nexus and dorsolateral prefrontal cortex was susceptible to experimental sleep manipulation \cite{27}. The detection and inhibition of noxious inputs could be modulated by the functional dynamics of brain networks \cite{22}. Considering the complex relationship between genetic factors, brain networks and phenotypes, considerable efforts have been made to relate genetic variants to underlying neurobiological aspects using imaging-genetic methods \cite{28}.

One hypothesis is that genetic factors contribute to phenotypes through the mediation of neurobiological changes. Mediation analyses are employed to better understand a known relationship by exploring the underlying mechanism or process by which one variable influences another variable through a mediator variable \cite{29}. Two-sample Mendelian randomization (MR) is widely used to infer causal relationships between two traits based on GWAS summary data \cite{30} and has been used in many studies \cite{31, 32}. These analyses would facilitate to understand the causal relationship between chronic pain and sleep disturbance and the contribution of genetic factors and neurobiological features to this bi-directional relationship.

The goal of this study is to identify the genetic correlation and shared brain abnormalities of modulatory effects of the reciprocal relationships between sleep disturbance and pain, explore the interaction between genetics and aberrant neurobiology from polygenic aspects and inference their causal relationship, which will inform mechanisms involved in this relationship to facilitate new therapies and better treatment.

**Methods**

**Participants and data**

To explore the relationship between sleep disturbance and chronic pain, we analyzed phenotype data, imaging data and genotype data from 1206 participants (all obtained full informed consent) from the Human Connectome Project (HCP) database (March 2017 public data release) from the Washington University-University of Minnesota (WU-Minn HCP) Consortium. Research procedures and ethical guidelines were followed in accordance with Washington University institutional review board approval. To apply for the genotype data, the study was also approved by the Institutional Review Board of the Peking University Sixth Hospital. In addition, two large sample size GWAS summary data for sleep disturbance and chronic pain were downloaded from published studies for the combined analyses.

**Phenotypes**

The sleep phenotype of participants in the HCP was assessed by Pittsburgh Sleep Quality Index (PSQI), a validated self-reported assessment of sleep disturbance \cite{33}. Participants’ self-reported experience of pain was measured by the National Institutes of Health (NIH) Toolbox Pain Intensity Survey \cite{34}. A score was used to measure the average pain intensity in the past 7 days, which was used to measure their chronic pain (CP). More details about the phenotypes are described in the Supplementary Methods and Supplementary Fig. S1. Since the distribution of pain intensity score was not normal, we used a binary variable for pain (chronic pain binary, CPb). According to previous studies \cite{35}, a pain intensity score $\leq 6$ denotes severe pain, which was set as 1, and a pain intensity score $\leq 5$ was set as 0. Because an average pain
intensity above 5 leads to disruptions of function and mood, it’s conceptually distinct from mild pain.\textsuperscript{36,37}

**Genotype data**

The genotype data for the HCP participants was requested from dbGaP under access number phs001364. Among the total 1206 participants, 1142 subjects were genotyped using Infinium Multi-Ethnic Genotyping Array. Quality control for the genotype data was described in Supplementary Methods. Since HCP contained many twins and siblings, after quality control, 429 independent individuals with 1,169,182 SNPs remained for the PGS association analysis. Principal component analysis (PCA) was performed for the genotype of 429 individuals for population stratification. The top 10 principal components (PCs) with a Tracy-Widom test p-value < 0.05 (Supplementary Fig. S2) were used as covariates in the subsequent PGS association analyses.

**Imaging data and construction of whole-brain functional network**

The HCP participants were scanned on a 3-T connectome-Skyra scanner (Siemens) using standard multiband blood oxygen level dependent (BOLD) acquisition. We acquired the resting-state data preprocessed by the HCP with its uniform method.\textsuperscript{38} The participants with four resting-state data (two scans, and two directions for each scan) were used for the construction of whole-brain functional network. After preprocessing, the gray matter of the whole brain was parcellated into 250 regions employing the Shen atlas.\textsuperscript{39} Nodal signals were created by averaging the regional blood oxygen level-dependent signals of all voxels within each region. Pearson cross correlations between all pairwise combinations of region signals were calculated for each participant, followed by z transformation to improve normality. The whole-brain functional connectivity network (250 × 250 regions with 31,125 edges) was constructed by further averaging the correlation coefficients of two directions of two scans.

**GWAS summary statistics data of chronic pain and sleep disturbance**

Large sample size GWAS summary data on chronic pain and sleep disturbance were used to explore their genetic correlations. The GWAS summary data of chronic pain was from a multisite GWAS of chronic pain in the UK Biobank, which included ~380,000 participants.\textsuperscript{11} The PSQI is a widely used measure of sleep quality with high sensitivity (98.7%) and specificity (84.4%) for identifying primary insomnia.\textsuperscript{34,41,42} Because there were no GWAS summary data on sleep disturbance, GWAS summary data on insomnia (i.e., the most common sleep disturbance) from a study of 1,331,010 individuals\textsuperscript{13} was used as a conceptual approximation of sleep disturbance.

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**Statistical analysis**

The analyses flowchart is shown in Fig. 1 to explore the relationship between chronic pain and sleep disturbance. All the analyses below used age, gender, race (categorized as white or other), handedness, years of education, body mass index, blood pressure (systolic and diastolic), alcohol abuse diagnosis, smoke history, marijuana dependence, and lifetime as covariates.

**Phenotypic and genetic correlations between chronic pain and sleep disturbance**

The phenotypic correlation between PSQI score and chronic pain CPb was calculated using partial correlation after controlling the covariates as above using SPSS v26. To explore their genetic correlation, LD score regression analysis was conducted by using GWAS summary data on chronic pain and sleep disturbance. LD score regression has been used in many studies to test the genetic correlations between two traits. Then, we used PGS association analyses to investigate if the PGS of chronic pain was associated with sleep disturbance and vice versa. The PGS was calculated using GWAS summary data (discovery), and further regressed (linear regression for PSQI, logistic regression for CPb) with phenotype data in the HCP GWAS data (target) (see Supplementary Methods).
Identifying functional connectivity correlated with pain and sleep disturbance

To address the issue of multiple comparisons during identifying functional connectivity (FC) from the large brain network (31,125 edges), we used the network-based statistic (NBS) method to identify the FC correlated with PSQI total score. The NBS method is a well-validated tool for brain network association analyses that has been previously been used widely in neuroimaging studies. A linear regression model was constructed after controlling for covariates to yield a t-test statistic, then the NBS method identified topological clusters using the test statistic threshold of 3.0 and further computed a family-wise error rate (FWER) corrected p-value for each component. The component with FWER corrected p-value < 0.05 was considered significant. The PSQI score associated FCs were further analyzed using logistic regression model for CPb with the same covariates as above to test if the sleep disturbance related FCs were associated with pain. The FCs with FDR adjusted p-value < 0.05 was considered significant.

Association of polygenic score of chronic pain/sleep disturbance with shared FC

To further explore if the shared FCs were associated with the PGS of chronic pain/sleep disturbance, we used PRSice v2 to run the PGS association analysis. Multiple PGS were created for chronic pain/sleep disturbance using the SNPs at p-value thresholds from 0 to 0.5 increasing by 0.00005. The association of the PGS with the correlation between PSQI total score and CPb values; gender, race, handedness, education, BMI, alcohol abuse diagnosis and smoke history were associated with PSQI total score; gender, race and education were associated with CPb. After controlling for covariates, the correlation between PSQI total score and CPb values was significant (r = 0.171, p-value < 0.001).

Mediation analysis

To explore how the shared FCs contribute to the association between chronic pain and sleep disturbance, we used model 4 in PROCESS to construct mediation models to analyze if the shared FC (medication variable) mediated the association between chronic pain (independent variable/output variable) and sleep disturbance (output variable/independent variable). The same covariates as above were used in the model. N = 10,000 was used for bootstrap. Sobel test was further used to show the p-value which denotes if the indirect effect was different with zero. In addition, the mediation function of the shared FC for the association between the PGS and phenotype was assessed using the same models.

Two-sample Mendelian randomization for causal inference

To explore the causal relationship between chronic pain and sleep disturbance, we conducted two-sample MR analyses using the online package of database and analytical platform MR-Base. The SNP p-value threshold for instrument variables was defined as p-value < 5.0 × 10⁻⁸. To avoid bias in the MR estimates due to LD (r²), clumping was applied using the ‘clump_data’ function with an r² < 0.001. To combine estimates from individual genetic variants, we applied inverse-variance-weighted (IVW) regression to test the causal effect. Then, we used the MR Egger intercept test to test for directional horizontal pleiotropy, and IVW regression test for variant heterogeneity. Additionally, the MR Egger test, weighted median test, and MR-PRESSO methods were used for sensitivity tests.

Results

Phenotype correlation

After quality controls, total of 989 participants with all four resting-state imaging data, sleep data, pain data, and covariates were used for the phenotypic and neuroimaging analysis. Participants ranged from 22 to 37 years (mean = 28.73, standard deviation = 3.70) with 461 males and 523 females. The demographic data for the participants is shown in Table 1. In the regression models with all covariates, gender, race, handedness, education, BMI, alcohol abuse diagnosis and smoke history were associated with PSQI total score; gender, race and education were associated with CPb. After controlling for covariates, the correlation between PSQI total score and CPb values was significant (r = 0.171, p-value < 0.001).

Genetic correlation between pain and sleep disturbance

LD score regression analysis showed chronic pain and sleep disturbance was significantly correlated at the genetic risk (r_g = 0.5975, p-value = 3.20 × 10⁻⁴). PGS association analysis result (Supplementary Table 1) showed that PGS of chronic pain was significantly associated with PSQI total score (R² = 0.0159, p-value = 6.47 × 10⁻³, P_adjust = 0.0324, coefficient=1051.62, SE = 384.29); PGS of sleep disturbance was also significantly associated with CPb (R² = 0.1102, p-value = 9.50 × 10⁻⁴, P_adjust = 4.75 × 10⁻³, coefficient = 24399.30, SE = 7382.47).

Shared brain FC associated with chronic pain and sleep disturbance

The NBS method obtained one significant network component for PSQI (p-value = 0.0288), which included 239 edges and 144 nodes. Further regression for these 239 FCs with CPb showed nine FCs were significantly associated with pain. These nine shared FCs involved the prefrontal cortex (PFC), temporal gyrus, precentral/
postcentral gyrus, anterior cingulate gyrus, hippocampus and fusiform gyrus (Fig. 2a, b). All of them were significantly correlated with PSQI and CPb (Table 2).

Furthermore, we assessed whether the shared FCs mediated the association between PSQI and CPb. The mediation model for the FC “right middle temporal gyrus” - “right hippocampus” is shown in Fig. 2c. The mediation models for the other eight FCs correlated with PSQI and CPb are shown in Supplementary Table 2. Seven out of the nine shared FCs showed mediation function on both the effect of PSQI on CPb and the effect of CPb on PSQI. Connectivity “right precentral gyrus” - “left hippocampus” and “left superior frontal gyrus, dorsolateral” - “left anterior cingulate & paracingulate gyrus” mediated the effects of PSQI on CPb but not on the other direction.

**Association of polygenic score of chronic pain/sleep disturbance with brain FC**

Given the association of chronic pain and sleep disturbance at both genetic and neural levels, we investigated if the PGS of chronic pain or sleep disturbance was associated with the shared FC, and identified one significant result: the PGS of chronic pain was significantly associated with the FC “right middle temporal gyrus” - “right hippocampus” (permutation adjusted p-value = 0.0402) (Fig. 2d). The association of the PGS of sleep disturbance with this FC obtained p-value = 9.4 × 10^{-3}, but permutation adjusted p-value was not significant (adjusted p-value = 0.1517) (Supplementary Table 3).

We further analyzed if the PGS, which was associated with the FC (“right middle temporal gyrus” - “right hippocampus”), was associated with the phenotypes (CPb and PSQI) through the mediation function of this FC. The result showed the indirect effect of chronic pain PGS on PSQI mediated through this FC was significant (p-value = 0.0459, Fig. 2d), but the indirect effect of sleep disturbance PGS on CPb mediated through this FC was not significant (p-value = 0.1233) (Supplementary Table 4).

**Causal relationship between chronic pain and sleep disturbance**

The genetic correlation between chronic pain and sleep disturbance may arise from genes with pleiotropic effects. We used a bidirectional, two-sample Mendelian randomization approach to explore the causal relationship between chronic pain and sleep disturbance. As shown in Table 3, putative causal effects of both directions were detected using IVW regression analysis Mendelian randomization test (p-value = 1.65 × 10^{-14} for chronic pain as exposure, sleep disturbance as outcome, p-value = 9.69 × 10^{-7} for opposite). IVW regression test for variant heterogeneity showed that the heterogeneity was significant for both directions (p-value = 2.81 × 10^{-4} and 1.23 × 10^{-6} separately) but the MR Egger tests for
Directional horizontal pleiotropy were not significant (p-value = 0.95 and 0.051 separately). Additional sensitivity tests showed, for chronic pain as exposure, sleep disturbance as outcome, all other four methods had effects in the same direction as the IVW test although p-value for MR Egger test was not significant, but for sleep disturbance as exposure, chronic pain as outcome, weighted median, MR-PRESSO and Contamination mixture had the same direction with IVW test with significance but MR Egger had different direction (Table 3).

**Discussion**

To our knowledge, this is the first study to examine the underlying genetic and neurologic components of the association between chronic pain and sleep disturbance using a large study sample. Based on the phenotypic correlation between chronic pain and sleep disturbance, which is consistent with earlier findings, we identified their high genetic correlation. At the neural level, we identified shared functional connectivities involving the prefrontal cortex, temporal gyrus, precentral/postcentral gyrus, anterior cingulate gyri, hippocampus and fusiform gyrus, which further mediate the association of chronic pain with sleep disturbance. Furthermore, the chronic pain PGS was associated with the FC “right middle temporal gyrus – right hippocampus”, which mediated the association of chronic pain PGS with sleep disturbance. The Mendelian randomization analyses implied a possible

Fig. 2 Chronic pain and sleep disturbance shared functional connectivities (FC) and the mediation function of the FC on the association between two phenotypes and the association between polygenic score and phenotypes. a Area with the nine FCs shared by PSQI and CPb, color is proportional to the links connected with the node. b Nine significant FCs with nodes and edges indicated, node size is proportional to the links connected with the node. c Mediation model for the mediation of the FC (“right middle temporal gyrus - right hippocampus”) for the correlation between PSQI and CPb. The mediation model for eight other significant FCs are shown in Supplementary Table 2. d Association results of PGS of chronic pain with FC (“right middle temporal gyrus” - “right hippocampus”) and mediation model for the indirect effect of PGS of chronic pain (CP) on PSQI through FC (“right middle temporal gyrus” - “right hippocampus”). SFGdor: superior frontal gyrus, dorsolateral; MFG: middle frontal gyrus; MTG: middle temporal gyrus; ITG: inferior temporal gyrus; ACC&PaCG: anterior cingulate & paracingulate gyri; HIP: hippocampus; FFG: fusiform gyrus; IFGtriang: interior frontal gyrus, triangular part; PreCG: precentral gyrus; PoCG: postcentral gyrus. For mediation model, X is independent variable, Y is outcome variable, M is mediation variable. Path a is the effect of X on M, path b is the effect of M on Y, path c is the effect of X on Y (total effect), c’ is the indirect effect of X on Y. PSQI: Pittsburgh sleep quality index; CP: chronic pain; CPb: chronic pain binary; PGS: polygenic score; #SNP: number of SNPs used to calculate PGS; P-th: p-value threshold to define SNPs with P < P-th to calculate the PGS.
causal relationship from chronic pain to sleep disturbance with stronger evidence than the other direction. These findings provide genetic and neural evidence for the co-occurrence of chronic pain and sleep disturbance.

The pleiotropic nature of genes implies that the PGS of chronic pain may also be associated with disturbed sleep and vice versa. The PGS association between chronic pain and sleep disturbance are consistent with phenotypic cross-sectional links and findings from previous twin studies which showed the relationship between sleep and pain to be confounded by shared genetic and environmental factors.

Genetic components for the development of sleep disturbance and chronic pain likely reflect inherited vulnerability to the development of abnormal functional connectivity. Co-occurrence of sleep disturbance and chronic pain suggests that both might share some common underlying neurophysiological mechanism. Consistent with this possibility, we found that disturbed sleep and chronic pain were significantly correlated with nine FCs involving the prefrontal cortex, temporal gyrus, precentral/postcentral gyrus, anterior cingulate gyri, hippocampus and fusiform gyrus.

Increasing evidence shows that the prefrontal cortex (PFC) is a key brain region correlated with executive function, pain processing and vigilance, awareness, attention and development of chronic pain. People with higher PSQI have worse performance in tests of working memory and attentional set shifting. Chronic pain patients exhibit significantly less deactivation in medial PFC and abnormal default mode network (DMN) activity while performing a visual attention task. Sleep deprivation research also suggests increased pain sensitivity is positively correlated with FC between executive control network (ECN) and DMN, implies that sleep deprivation impaired cognitive networks could partially contribute to the sleep-pain dyad. The abnormal FC between PFC and anterior cingulate cortex (ACC) we found may demonstrate the deficiency in executive control, thus impairing the ability to transfer from DMN to ECN. Connections of PFC and postcentral gyrus fits the latest model of perception of pain which requires the involvement of PFC. Hyperalgesia might happen through impaired PFC function caused by sleep deficits.

As part of the limbic system, hippocampus could potentiate the sleep disturbance and make it vulnerable to neuropsychiatric disorders such as depression and chronic pain. Decreased hippocampal neurogenesis is closely associated with memory deficits and aversive affective states in patients with chronic pain. The memory network deficit revealed in chronic pain and sleep disturbed patient might be caused by the aberrant FC between temporal cortex and hippocampus which is demonstrated in our research. We also found the FC
between middle temporal gyrus and hippocampus could mediate the association between chronic pain and sleep disturbance. Disrupted hippocampus-PFC connectivity predicts the transition from subacute to chronic pain and impairment in memory encoding and retrieval, decision-making and extinction learning. Furthermore, the shared brain FCs might in charge of properties of chronic pain and sleep disturbance through shared genetic risk, which may lead to sleep disturbances and chronic pain. The association of chronic pain PGS with the FC between right middle temporal gyrus and right hippocampus confirmed this hypothesis. The FC further mediated the indirect effect of the chronic pain PGS on sleep disturbance. Patterns of phenotypic overlap and comorbidity might share a direct etiological links which risk genes confer pleiotropic risk for multiple distinct brain phenotypes.

Pain can be both a cause and a consequence of sleep disturbance. Meta-analysis of longitudinal studies found that a decline in sleep quality and quantity was associated with a two- to three-fold increase in the risk of developing a pain condition. Our research supports both directions of causation, but the causation of sleep disturbance by chronic pain is more strongly supported (all methods had the same effect direction). Longitudinal studies report that pain and sleep often interact and negatively affect each other and the relationship is not unidirectional but more complex. Future longitudinal studies are needed to investigate direction of causation with more ingenious designs.

Finding that chronic pain and sleep disturbance shared neural and genetic underpinnings is significant. Strengths of the present investigation are the large number of participants leading to robust findings, and evidence from both genetic and neural mechanisms obtained by integrating the imaging-genetics results. The mediation and MR causal analyses further helped to uncover the relationship between these two clinically important comorbidities. Currently, pharmacotherapy for chronic pain and insomnia has been woefully inadequate. The mediation effect of the aberrant FC that we identified in the present study may indicate potential increased efficacy and reproducibility target of neurostimulation to propose therapies to the patients who suffer from sleep disturbance and complicated by chronic pain.

There were several limitations in our study. The sample in the HCP database is from the young general population. The phenotypic correlation between sleep and pain is weaker than clinical samples, which resulted in relatively few reported pain intensity scores greater than 6. We used a pain intensity score threshold ≥6 to define the binary variable CPb. We also tried to define a threshold ≥3 for the regression analysis of PSQI and FCs, but there was no significant result (data not shown). The logistic regression model with a threshold ≥6 had significant intercept, whereas the threshold ≥3 did not, which may indicate that only severe pain affects brain FC. The PSQI score and pain variable that we used could only represent a portion of sleep disturbance and chronic pain. Additionally, gender is a factor influencing pain and sleep.

Further research to elucidate the mechanism underlying sex difference in pain and sleep is needed to reduce the disparities. Although the number of participants (n = 989) for imaging analysis is relatively large, the sample size for the PGS association analysis was small. Also, causal and mediation relationships between chronic pain and chronic pain.

### Table 3 Two-sample MR analysis results for chronic pain and sleep disturbance.

| Outcome          | #SNP | Mendelian randomization | IVW regression test for variant heterogeneity | MR Egger test for directional horizontal pleiotropy |
|------------------|------|-------------------------|---------------------------------------------|---------------------------------------------------|
|                  |      | B   | SE   | p-value | Q | df | p-value | Egger intercept | SE   | p-value |
| Sleep disturbance| 35   | 0.670 | 0.087 | 1.65 × 10⁻¹⁴ | 69.879 | 34 | 2.81 × 10⁻⁴ | −0.00042 | 0.0066 | 0.95 |
|                  |      | 0.696 | 0.046 | 0.096   | 6.58 × 10⁻⁹ | 1.31 × 10⁻⁶ |
| Chronic pain     | 13   | 0.205 | 0.042 | 9.69 × 10⁻⁷ | 0.00944 | 0.0043 | 0.051 |
|                  |      | −0.030 | 0.114 | 0.794   | 3.33 × 10⁻⁴ | 3.67 × 10⁻⁴ |
|                  |      | 0.135 | 0.038 | 3.67 × 10⁻⁴ | 0.370 | 0.038 | 5.89 × 10⁻⁵ |

Bold values denote p < 0.05.
pain results from MR analysis and mediation analyses were analyzed based on the sectional data. The results need further validation and exploration using larger independent samples in studies with longitudinal designs.

Conclusions
This study explored the association between chronic pain and sleep disturbance from different perspectives. The results revealed their high genetic correlation, identified shared brain functional connectivities mediated their association, built the link from the PGS of chronic pain to brain functional connectivity and further sleep disturbance, and possible causal relationship from chronic pain to sleep disturbance. Our results would have significant implications for understanding the mechanism of sleep and pain comorbidity and imply new treatment and interventions strategy for the comorbidity.

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Conflict of interest
The authors declare that they have no conflict of interest.

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