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Common Genetic Variants in *ARNTL* and *NPAS2* and at Chromosome 12p13 are Associated with Objectively Measured Sleep Traits in the Elderly

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Study Objectives: To determine the association between common genetic variation in the clock gene pathway and objectively measured actigraphic sleep and activity rhythm traits.

Design: Genetic association study in two population-based cohorts of elderly participants: the Study of Osteoporotic Fractures (SOF) and the Osteoporotic Fractures in Men (MrOS) study.

Setting: Population-based.

Participants: SOF participants (n = 1,407, 100% female, mean age 84 years) and MrOS participants (n = 2,527, 100% male, mean age 77 years) with actigraphy and genotype data.

Interventions: N/A.

Measurements and Results: Common genetic variation in 30 candidate genes was captured using 529 single nucleotide polymorphisms (SNPs). Sleep and activity rhythm traits were objectively measured using wrist actigraphy. In a region of high linkage disequilibrium on chromosome 12p13 containing the candidate gene *GNB3*, the rs1047776 A allele and the rs2238114 C allele were significantly associated with higher wake after sleep onset (meta-analysis: rs1047776 P_{ADD} = 2 × 10^{-5}, rs2238114 P_{ADD} = 5 × 10^{-5}) and lower *LRRC23* gene expression (rs1047776: ρ = -0.22, P = 0.02; rs2238114: ρ = -0.50, P = 5 × 10^{-8}). In MrOS participants, SNPs in *ARNTL* and *NPAS2*, genes coding for binding partners, were associated with later sleep and wake onset time (sleep onset time: *ARNTL* rs3816358 P_{2DF} = 1 × 10^{-4}, *NPAS2* rs3768984 P_{2DF} = 5 × 10^{-5}; wake onset time: rs3816358 P_{2DF} = 3 × 10^{-3}, rs3768984 P_{2DF} = 2 × 10^{-4}) and the SNP interaction was significant (sleep onset time P_{INT} = 0.003, wake onset time P_{INT} = 0.001). A SNP association in the *CLOCK* gene replicated in the MrOS cohort, and rs3768984 was associated with sleep duration in a previously reported study. Cluster analysis identified four clusters of genetic associations.

Conclusions: These findings support a role for common genetic variation in clock genes in the regulation of inter-related sleep traits in the elderly.

Keywords: Genetic, aging, circadian, actigraphy, SNP

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INTRODUCTION

Circadian clocks influence many aspects of physiology and behavior such as body temperature, release of hormones, and sleep-wake cycles. Sleep consolidation and the timing of sleep-wake cycles are regulated by the complex interplay between a circadian process and a homeostatic process.[1] The central circadian pacemaker located in the suprachiasmatic nucleus (SCN) of the hypothalamus entrains peripheral clocks in extra-SCN brain regions to regulate sleep-wake cycles.[2]

Aging is accompanied by multiple changes in sleep quality and sleep timing.[3] In more than 9,000 participants 65 yr and older, 43% reported difficulty initiating or maintaining sleep.[4] Cross-sectional analysis has demonstrated that older age is as-

sociated with lower sleep efficiency (SE) and lower percentage of slow wave sleep.[5-9] Longitudinal analysis also points to an age-related decline of SE and slow wave sleep. Over a 3-yr follow-up, SE and percentage of slow wave sleep declined more significantly in subjects age 75 to 87 yr than in subjects age 61 to 74 yr, and this decline was not correlated with changes in chronic medical burden.[10] Comparisons of sleep timing across age groups indicate that older age is also associated with sleep occurring at earlier clock times and increased morning preference; however, this relationship has not been examined longitudinally.[11-14]

Studies in humans measuring the output of the circadian pacemaker, e.g., core body temperature or melatonin, have provided strong evidence for an age-related advancement of the phase of circadian rhythms; however, the evidence for an age-related reduction of circadian rhythm amplitude is mixed.[12,13,15-18] Circadian period, on the other hand, does not appear to differ by age.[19] Multiple lines of evidence indicate that age-related changes in sleep are related to an interaction between age-related changes in the circadian system and the sleep homeostatic process.[12,13,20-22] Contributing to an early habitual wake time in older people is a shorter interval between the core body temperature minimum (CBT_{min}) and wake time compared with young adults.[13] In addition, age-related changes

in the circadian regulation of sleep-wake propensity results in decreased sleep consolidation at CBT$_{min}$ among older people, which could result in early awakening and exposure to light at an earlier circadian phase, which in turn could shift circadian phase to an earlier hour.[13,21-24] Accompanying these age-related changes in sleep are alterations in the neural organization of the SCN.[25] There is also evidence from a transgenic rat study that age-related deterioration of rhythmic clock gene expression is more prominent in some peripheral tissues than the SCN, suggesting that the SCN loses the ability to entrain some peripheral clocks with age.[26] Results from a human sleep laboratory study using a 90-min sleep-wake cycle are consistent with the notion of an age-related decline in the SCN's ability to entrain peripheral clocks.[24] In summary, age-related changes in the circadian and sleep homeostatic processes and in central and peripheral clocks accompany age-related changes in sleep.

A highly conserved set of clock genes regulates the timing of the central pacemaker in the SCN through a transcription-translation feedback loop.[27] CLOCK and ARNTL heterodimerize and activate transcription of the period (*PER1*, *PER2*, and *PER3*) and cryptochrome (*CRY1* and *CRY2*) genes.[27] NPAS2 plays a similar role to that of CLOCK in the forebrain and possibly in the SCN.[28,29] PER and CRY protein levels accumulate over the course of the day, leading to PER:CRY heterodimers that translocate back into the nucleus where they repress CLOCK and ARNTL activity, thus repressing their own transcription. Degradation of the PER:CRY heterodimers during the night allows CLOCK:ARNTL to activate a new cycle of *PER* and *CRY* gene transcription. Posttranslational modification of the components of the molecular clock regulates clock function. Subcellular localization and ubiquitin-dependent degradation of PER and CRY is regulated by CSNK1D/CSNK1E/GSK3β-mediated phosphorylation. FBXL3 is involved in the ubiquitination of phosphorylated CRY proteins, and βTrCP1 and βTrCP2 are involved in the ubiquitination of phosphorylated PER proteins. An auxiliary clock feedback loop regulates ARNTL levels and makes the cyclic pattern of clock gene expression more robust. This auxiliary loop is composed of REV-ERBα/NR1D1 and REV-ERBβ/NR1D2, which repress *ARNTL* expression, and RORα, RORβ, and RORγ, which activate *ARNTL* expression. PGC1α enhances RORα's activation of *ARNTL*. A second auxiliary feedback loop consists of DBP, TEF, HLF, and E4BP4/NFIL3. Several additional molecular factors are involved in the regulation of the mammalian molecular clock, as reviewed by Lowrey and Takahashi.[27] In addition to exhibiting a cyclic pattern of gene expression within an SCN neuron, clock gene expression is also synchronized among the SCN neurons. Specifically, VIP and VIPR2 are required to couple SCN neurons to produce synchronized rhythms of clock gene expression.[27]

Clock genes have been shown to influence sleep traits in model organisms, and there is growing evidence that genetic variants in clock genes are associated with sleep traits in humans. A missense mutation in *PER2* disrupting the phosphorylation site of CSNK1E has been linked to familial advanced sleep phase syndrome, a rare mendelian sleep disorder.[30] A variant in *CSNK1D*, a paralog of *CSNK1E*, has also been found to be associated with familial advanced sleep phase syndrome.[30] Genetic variants in *PER2*, *PER3*, *CLOCK*, *CSNK1E*, *DBP*, and *DEC2* have been found to be associated with various sleep traits in humans, including diurnal preference and delayed sleep phase.[30-36] Variants in *NPAS2* and *ARNTL* have been associated with seasonality and seasonal affective disorder, phenotypes that could reflect circadian rhythm disruption.[33,37,38] However, some clock gene SNP associations with sleep traits have failed to replicate.[39] Furthermore, many of the previously reported genetic association studies of sleep traits have examined a small number of candidate polymorphisms in clock genes rather than systematically assaying common genetic variation within the entire gene region; thus, allelic heterogeneity has not been explored in many studies. To date, three studies have attempted to systematically capture common genetic variation in the known clock genes to examine genetic associations with sleep traits, but subjective measures of sleep-related traits were analyzed in all three studies.[32,40,41] SNPs in *TIMELESS* were significantly associated with case-control status where cases were defined using trait combinations of depression, early morning awakening, and fatigue.[40] However, the combination of traits used in the case definition presents a challenge in selecting appropriate replication cohorts. The other studies that systematically captured common genetic variation in clock genes examined self-reported sleep duration using the Munich ChronoType Questionnaire (MCTQ).[32,41] Objectively measured sleep traits have not been examined in genetic association studies with SNPs designed to systematically capture common genetic variation in clock genes. Moreover, study populations composed entirely of elderly individuals have not been used in genetic association studies examining variation in clock genes and sleep traits measured objectively or subjectively.

In this study, we aim to examine the association between common genetic variation within clock gene regions and objectively measured sleep and activity rhythm traits in the elderly, a population group in which sleep continuity is frequently disrupted and the circadian and homeostatic sleep processes have undergone multiple age-related changes. We investigated the association between these sleep and activity rhythm traits that were objectively measured using actigraphy and 529 tagSNPs capturing common genetic variation in 30 clock genes in two large population-based cohorts of elderly participants 65 yr and older. The 30 clock genes that were examined included the canonical circadian pathway genes and genes previously shown to be associated with sleep traits in humans. Hierarchical cluster analysis was performed to further elucidate the relationship between genetic associations of correlated sleep traits.

METHODS

Study Participants

Data from two studies, the Osteoporotic Fractures in Men (MrOS) study and the Study of Osteoporotic Fractures (SOF), contributed to this analysis. The design of the MrOS study was based on the SOF study to allow for comparable analyses in both men and women. Both are large prospective studies recruited from communities in the United States, with the primary aim of studying osteoporotic fractures. Both had the limited exclusion criteria at enrollment of age younger than 65 yr, history of a bilateral hip replacement, or inability to walk without assistance. In the MrOS study, 5,994 men 65 yr and older were recruited from 2000 to 2002.[42,43] The MrOS Sleep

Study, an ancillary study of the parent MrOS cohort, was conducted between December 2003 and March 2005 and recruited 3,135 MrOS participants for a comprehensive sleep assessment. Among the participants of the MrOS Sleep Study, 2,527 self-identified white MrOS participants had DNA extracted, SNPs genotyped, and technically adequate sleep actigraphy data collected. In the SOF study, 9,704 self-identified white women 65 yr and older were recruited and baseline examinations were conducted from 1986 to 1988.[44] From January 2002 to February 2004, 4,727 women attended the eighth SOF clinic visit. Among these 4,727 women, 3,219 were provided with an actigraph, as described in Tranah et al.,[45] and technically adequate sleep actigraphy data were collected from 2,799 women. Among these 2,799 participants, DNA was extracted and SNPs were genotyped for 1,407 participants. All data were collected with written informed consent as approved by the review boards of the participating institutions.

Actigraphic Sleep Phenotypes

Sleep-wake patterns were assessed using actigraphy in both cohorts (Sleep-Watch-O, Ambulatory Monitoring, Inc., Ardsley, NY), as was described previously.[46,47] Briefly, the actigraph was similar in size to a wristwatch and was worn on the non-dominant wrist. Activity was detected by a piezoelectric bimorph-ceramic cantilever beam that generated a voltage each time the actigraph was moved. Data were collected in three modes but are reported here in proportional integration mode given our work showing that this mode correlates best with the gold standard of polysomnography for measurement of sleep traits.[47,48] While the actigraph was worn, participants completed sleep diaries that included time into and out of bed and times the actigraph was removed. This information was used when editing the actigraphy data files to set intervals for when the participant was in bed trying to sleep (after "lights off"), and to delete time when the actigraph was removed. A sleep scoring algorithm was applied to the data to determine sleep-wake boundaries. Actigraphy data were available in MrOS and SOF participants for 5.2 ± 0.8 nights (mean ± standard deviation [SD]) and 4.1 ± 0.7 nights, respectively.

The variables estimated by actigraphy were time in bed (the time spent in bed trying to sleep at night, from "lights off" to the time of getting out of bed), total sleep time (TST, the hours per night spent sleeping during the time spent in bed), sleep onset time (the start of the first 20 minute continuous block scored as sleeping after "lights off"), wake onset time (the last minute during the in-bed interval scored as sleeping), number of long wake episodes (NWAK, number of awakenings 5 minutes or longer in duration while in bed), sleep latency (SL, minutes from "lights off" to sleep onset), wake after sleep onset (WASO, minutes spent awake during the in-bed interval between sleep and wake onset times), sleep efficiency (SE, the percentage of time in bed spent sleeping between sleep and wake onset times), and nap minutes (minutes scored as sleep for continuous blocks of at least 5 minutes while not in bed at night trying to sleep). These actigraphy variables reflect data averaged over all nights or days, as appropriate, that participants wore the actigraph in order to obtain a more representative characterization of usual sleep patterns.

The activity data gathered by actigraphy was used to compute measures of activity rhythms using an extension to the traditional cosine curve to allow the activity data to fit a more squared wave rather than a cosine curve.[49] Nonlinear least squares regression was used to estimate the activity parameters. The following activity rhythm traits were computed: acrophase (the time of day of peak activity, measured in portions of hours), amplitude (the difference between the minimum and maximum of the function, measured in arbitrary units of activity as counts/minute), mesor (the minimum of the function plus half of the amplitude, which represents the middle of the fitted activity peak with units of counts/minute), and the pseudo-F statistic (a measure of overall fit of the activity data to the extended cosine curve, with higher values indicating stronger activity rhythms).

Measurement of Other Nongenetic Characteristics

All participants completed questionnaire data, which included questions about alcohol use, caffeine intake, and place of residence. Prescription and nonprescription medications used within the preceding 30 days were identified, recorded by the clinics, and stored in an electronic medications inventory database (San Francisco Coordinating Center, San Francisco, CA); each medication was matched to its ingredient(s) based on the Iowa Drug Information Service Drug Vocabulary (College of Pharmacy, University of Iowa, Iowa City, IA).[50] A comprehensive examination included measurements of body weight and height; body mass index was calculated as weight in kilograms divided by the square of height in meters.

Clock Gene Selection, Genotyping and Quality Control of TagSNPs, and Analysis of Population Stratification

The California Pacific Medical Center investigators and the University of California, San Diego group collaborated to develop a custom Illumina Golden Gate assay (Illumina, San Diego, CA) to genotype polymorphisms in circadian and sleep related genes. Candidate genes (Table S1) were selected after review of experimental studies involving circadian rhythms in common model organisms (e.g., *Drosophila melanogaster* and *Mus musculus*) and association studies performed in human participants. The role of the canonical circadian pathway genes listed in Table S1 has been previously described.[27] In addition, *FMR1* and *FXR2* regulate the circadian rhythm in mice, but their role in the regulation of sleep in humans has not been examined.[51] Evidence for the association between the C825T *GNB3* allele and seasonality is mixed.[52,53] It has been reported that the C825T *GNB3* allele interacts with *CLOCK* variants in an association with diurnal preference.[54] *PGC1B* has been found to be associated with bipolar disorder,[55] and its paralog, *PGC1A*, regulates *ARNTL*.[56] Evidence for the role of *TIMELESS* in the mammalian molecular clock is mixed; nevertheless, *TIMELESS* was included as a candidate gene.[57]

TagSNPs were selected using Tagger[58] ($r^2 \geq 0.8$, minor allele frequency (MAF) ≥ 0.01) with HapMap CEU Phase II (release 22) genotype data in the candidate gene regions including 10 kb upstream and downstream of transcript boundaries. Using these Tagger settings, 314 tagSNPs would have been selected from the 798 SNPs found within the 761 kb of genomic DNA that includes the *RORA* gene and 20 kb of flanking DNA. To reduce our *RORA* genotyping burden, the linkage disequilibrium (LD) threshold in Tagger was lowered to 0.6 and a maximum of 100 tagSNPs were selected for genotyping, resulting in

81% of the 798 *RORA* SNPs being captured at r² ≥ 0.6. In total, 658 SNPs within the candidate gene regions were selected for genotyping. Genotypes were called using Beadstudio software (Illumina, San Diego, CA). Genotype concordance rate was > 0.99 (8% of MrOS samples and three SOF samples were plated in duplicate). Samples with < 90% SNP call rate were excluded. SNPs with missing call rate frequency > 0.05 and MAF < 0.01 were excluded. Autosomal SNPs with a Hardy-Weinberg equilibrium (HWE) exact P value < 8×10^{-5} (Bonferroni-corrected P value for 658 SNPs) were excluded.[59] Among the all-male MrOS samples, X-linked SNPs were excluded if they were found to have heterozygous genotypes. Among the 658 genotyped SNPs, 529 in MrOS and 508 in SOF passed quality control (QC) filters (Table S1).

To correct for residual population stratification in these self-identified European American cohorts, 195 independent, autosomal SNPs were used in multidimensional scaling analysis (MDS) as implemented in PLINK for the MrOS and SOF participants, and the resulting first two MDS components were included as covariates in all regression models.[60]

Statistical Analysis

All actigraphic sleep traits, except for NWAK, were continuous variables and linear regression analysis was performed. Effect size was reported as the β parameter estimate. Interaction analysis was performed by inclusion of a multiplicative term in linear regression models. Poisson regression models were explored for the count variable NWAK, but there was evidence of overdispersion (P < 0.0001 for the likelihood ratio test of the dispersion parameter in negative binomial regression models); thus, negative binomial regression was used to model NWAK. To obtain rate ratios for NWAK, time in bed (h) was included as an offset variable in models. Results presented here were based on regression models adjusted for age, body mass index, clinic site as indicator variables, and the first two MDS components. Secondary analysis was performed to assess the sensitivity of our results to various factors. In the secondary analysis, each of the following steps was performed individually: exclusion of participants who reported taking nonbenzodiazepine nonbarbiturate sedative hypnotic prescription sleep medication (Chloral Hydrate, Dexmedetomidine, Eszopiclone, Zopiclone, Zolpidem, Zaleplon, or Ramelteon) (MrOS n = 51, SOF n = 18), adjustment for current alcohol consumption, adjustment for current caffeine consumption, adjustment for residence type, and adjustment for season of the year when actigraphy was performed. SNP associations whose significance exceeded the multiple testing threshold in primary analysis did not change in secondary analysis (data not shown).

To correct for multiple hypothesis testing in the presence of LD, the effective number of independent SNPs using our QC-filtered genotype data was estimated and a multiple testing significance threshold of 1.7×10^{-4} was adopted.[61,62] Results for all three modes of inheritance (additive, dominant, and recessive) were shown for SNPs whose association significance exceeded the multiple testing threshold under the additive inheritance mode or the model-free two degrees of freedom (2DF) test. Regression analysis was performed in each cohort separately, and then combined by fixed-effect meta-analysis using inverse variance weighting of effect estimates. P values from

the 2DF test were combined by a Z-statistic based approach weighted by the square root of the sample size. Heterogeneity between studies was assessed using the I² statistic and the P value from the Q-test. Association analysis was performed using R (www.r-project.org), meta-analysis was performed using METAL,[63] and power calculations were performed using QUANTO.[64]

Hierarchical agglomerative cluster analysis (pvclust R package) was performed using the complete linkage clustering method on the distance matrix of the absolute value of the sample correlation of the additive SNP test statistics from the meta-analysis for all examined actigraphic traits. Cluster significance was assessed using multiscale bootstrap resampling with 10,000 replications. Principal component analysis was performed on the correlation matrices of the actigraphic traits and the additive SNP test statistics from the meta-analysis.

BioGPS was used to determine the distribution of *LRRC23* gene expression across 79 tissues in a human microarray panel.[65,66] Genevar was used to access HapMap CEU expression quantitative trait loci (eQTL) data, and the association between SNP genotype and gene expression was estimated using the Spearman rank correlation coefficient.[67,68]

RESULTS

In this genetic association study of common genetic variation in clock genes and objectively measured actigraphic sleep traits, two population-based cohort studies (MrOS and SOF) of elderly participants were examined. Age was significantly lower and body mass index was significantly higher in the all-male MrOS cohort compared with the all-female SOF cohort (Table 1). The age range of the MrOS and SOF participants was 67 to 96 yr and 79 to 98 yr, respectively. Except for sleep onset time and NWAK, all other sleep and activity rhythm traits were significantly different between the two cohorts (Table 1). Actigraphy-based TST and SE were reported in a population cohort (Rotterdam Study) with a similar participant age distribution as the MrOS and SOF cohorts, and the TST and SE estimates from the three cohorts were remarkably similar (Rotterdam mean TST in h: 6.4 in men, 6.7 in women; mean SE %: 77.8 in men, 79.0 in women).[69]

This candidate gene study examined 529 SNPs in 30 clock genes (Table S1). SNP associations that passed multiple testing under the additive inheritance mode or the model-free 2DF test in either cohort alone or in the meta-analysis are described in detail. Briefly, two SNPs (rs1047776 and rs2238114) at chromosome 12p13 were significantly associated with sleep continuity in the meta-analysis of results from MrOS and SOF. In MrOS, rs3816358 in *ARNTL* and rs3768984 in *NPAS2* were significantly associated with later sleep timing, and the SNP interaction was significant. In SOF, the *NPAS2* SNP rs895520 was significantly associated with the pseudo-F statistic, a measure of activity rhythm robustness, and rs1047776 at chromosome 12p13 was significantly associated with sleep latency.

SNP Associations With Sleep Continuity at Chromosome 12p13

In an attempt to tag common genetic variation in and around the *GNB3* gene at chromosome 12p13, tagSNPs were genotyped within *GNB3* and two surrounding genes, *LEPREL2* and *USP5*. The *LEPREL2* SNP rs1047776 and the *USP5* SNP

rs2238114 (Figure 1) were both significantly associated with multiple sleep continuity traits after multiple test correction (Table 2, Table S2). Under an additive mode of inheritance, rs1047776 in the 3' untranslated region (UTR) of *LEPREL2* was associated with higher WASO in MrOS ($\beta_{ADD} \pm SE$, P_{ADD}: 4.22 ± 1.27, 9x10^{-4}) and SOF ($\beta_{ADD} \pm SE$, P_{ADD}: 5.04 ± 1.82, 0.006), and the meta-analysis P value passed multiple test correction ($\beta_{META} \pm SE$, P_{META}: 4.49 ± 1.04, 2×10^{-5}) (Table 2). Consistent with an increase in WASO, rs1047776 was also associated with a decrease in SE under an additive mode of inheritance in MrOS ($\beta_{ADD} \pm SE$, P_{ADD}: -0.99 ± 0.29, 8×10^{-4}) and SOF ($\beta_{ADD} \pm SE$, P_{ADD}: -0.97 ± 0.38, 0.01), and the meta-analysis P value passed multiple test correction ($\beta_{META} \pm SE$, P_{META}: -0.98 ± 0.23, 2×10^{-5}) (Table 2). Consistent with an increase in WASO and

Table 1—Characteristics of the study populations

|	MrOS		SOF		
Trait (unit)	n[a]	mean ± SD	n[a]	mean ± SD	
Age (yr)	2,527	76.57 ± 5.61	1,407	83.88 ± 3.37	**
Body mass index (kg/m^2)	2,527	27.22 ± 3.79	1,407	26.73 ± 4.81	*
Sleep traits					
Sleep onset time (h:min)	2,527	23:16 ± 01:16	1,407	23:19 ± 01:17	
Wake onset time (h:min)	2,527	06:52 ± 01:07	1,407	07:13 ± 01:08	**
Total sleep time (h)	2,527	6.42 ± 1.23	1,407	6.80 ± 1.24	**
Number of wake episodes	2,527	6.89 ± 3.28	1,407	6.67 ± 3.08	
Wake after sleep onset (min)	2,527	78.36 ± 44.73	1,407	74.32 ± 46.41	**
Sleep efficiency (%)	2,527	82.57 ± 10.44	1,407	84.44 ± 9.68	**
Sleep latency (min)	2,527	30.72 ± 32.78	1,407	39.48 ± 37.84	**
Nap minutes (min)	2,524	55.31 ± 51.83	1,390	64.54 ± 57.40	**
Activity rhythm traits					
Acrophase (h:min)	2,527	14:16 ± 01:13	1,397	14:41 ± 01:11	**
Pseudo-F statistic	2,527	1,052.29 ± 513.79	1,397	862.48 ± 452.56	**
Amplitude (counts/min)	2,527	3,598.71 ± 1,057.46	1,397	3,435.37 ± 992.48	**
Mesor (counts/min)	2,527	2,159.15 ± 481.49	1,397	2,104.46 ± 463.61	**

[a]Number of genotyped samples with nonmissing data for sleep traits and covariates used in regression models. *P < 0.05, unpaired two-sided *t*-test for sleep onset time, wake onset time, acrophase, pseudo-F statistic, amplitude, mesor, and total sleep time; Kruskal-Wallis test for number of wake episodes, wake after sleep onset, sleep efficiency, sleep latency, and nap min. **P < 0.001 for statistical tests listed above. MrOS, The Osteoporotic Fractures in Men study; SD, standard deviation; SOF, The Study of Osteoporotic Fractures.

a decrease in SE, meta-analysis indicated that rs1047776 was also significantly associated with a higher rate of long (> 5 min) wake episodes while in bed (Table S2). Similarly, the intronic SNP rs2238114 within the *USP5* gene was also significantly associated with WASO and NWAK, but the effect of the minor allele was in the opposite direction to that of rs1047776 (Table 2, Table S2). The association significance between rs2238114 and SE nearly exceeded the multiple testing threshold (Table 2). There was no evidence for interaction between rs1047776 and rs2238114 for WASO or SE (P_{int} > 0.05).

Although rs2238114 was in high LD (HapMap CEU $r^2 = 0.95$, Figure 1) with the *GNB3* SNP rs5443 that was previously reported to interact with a *CLOCK* SNP in the association with diurnal preference,[54] rs1047776 in *LEPREL2* was selected to capture common genetic variation upstream of *GNB3* and was not in high LD with any of the HapMap SNPs within *GNB3* (maximum HapMap CEU $r^2 = 0.16$), including rs5443 (HapMap CEU $r^2 = 0.12$) (Figure 1). The SNPs rs1047776 and rs2238114 were not in high LD with each other in MrOS participants ($r^2 = 0.12$) or in HapMap CEU individuals ($r^2 = 0.15$, Figure 1). Despite the low LD between rs1047776 and rs2238114, the effect sizes and P values for the WASO and SE SNP associations were shifted toward the null in conditional analysis that included both SNPs in the same regression model (Table S3).

In addition to being in high LD with SNPs in *GNB3*, the *USP5* SNP rs2238114 was also in high LD (HapMap CEU $r^2 \geq 0.8$) with SNPs in many neighboring genes. In fact, rs2238114 and the previously reported rs5443 reside within a 101kb haplotype block (HapMap CEU) that spans the following genes: *GNB3*, *CDCA3*, *USP5*, *TPI1*, *SPSB2*, *LRRC23*, *ENO2*, *ATN1*, and *PTPN6* (Figure S1).

Publicly available eQTL data from HapMap CEU lymphoblastoid cell lines were used to determine whether the sleep continuity-associated SNPs that we identified (rs2238114 and rs1047776) were associated with expression of any of the genes in the 101 kb haplotype block. The rs2238114 and rs1047776 alleles associated with higher WASO were significantly associated with lower *LRRC23* gene expression in HapMap CEU lymphoblastoid cell lines (Table 3). Significant association was observed between rs2238114 and both *LRRC23* gene expression probes ($P = 5 \times 10^{-8}$ and 2×10^{-4}) (Table 3, Figure S2A). Although the significance of the correlation between *LRRC23* gene expression and rs1047776 was not as significant as that for rs2238114, rs1047776 was more correlated with *LRRC23* gene expression than with the expression of any other gene in the LD block (Table 3). *LRRC23* gene expression is found in a broad range of human tissues, including multiple brain regions (Figure S2B).

SNP Associations with Sleep Timing in the ARNTL and NPAS2 Genes

The SNPs rs3816358 in *ARNTL* and rs3768984 in *NPAS2* were associated with later sleep timing in the MrOS cohort but not the SOF cohort. The intronic SNP rs3816358 within *ARNTL* was significantly associated with later sleep onset time in MrOS participants ($\beta_{ADD} \pm SE$, P_{ADD}: 0.19 ± 0.06, 7×10^{-4}) and the P value for the 2DF test ($P_{2DF} = 1.2 \times 10^{-4}$) exceeded the multiple testing threshold (Table 4). The SNP rs3816358 was also nominally associated with a later wake onset time in MrOS participants ($\beta_{ADD} \pm SE$, P_{ADD}: 0.16 ± 0.05, 0.001; $P_{2DF} = 0.003$) (Table 4).

Consistent with a nearly equal association effect size with sleep onset time and wake onset time, rs3816358 was not associated (P > 0.05) with TST in the MrOS cohort (data not shown). In addition, rs3816358 was not associated with WASO,
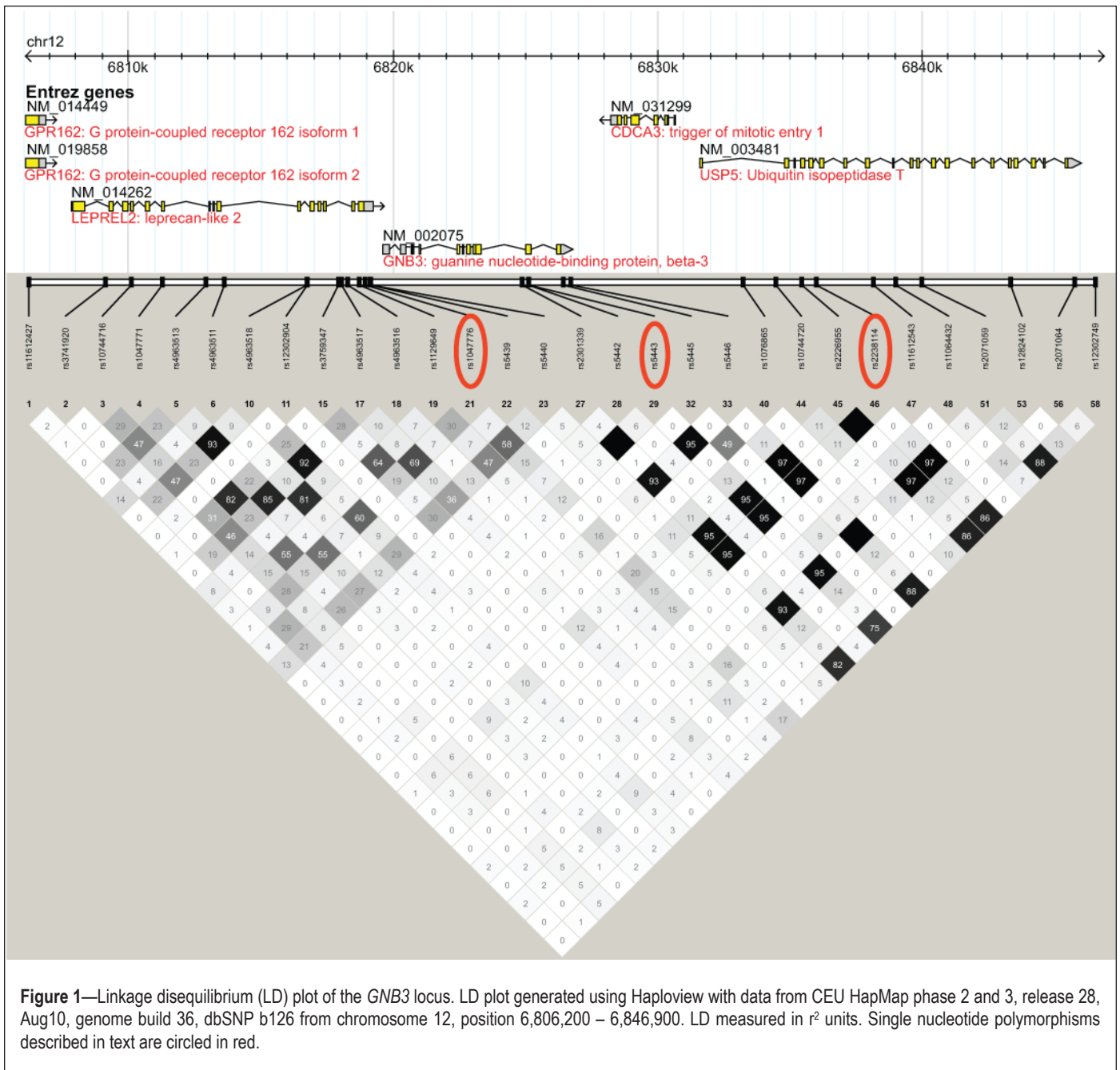
Figure 1—Linkage disequilibrium (LD) plot of the *GNB3* locus. LD plot generated using Haploview with data from CEU HapMap phase 2 and 3, release 28, Aug10, genome build 36, dbSNP b126 from chromosome 12, position 6,806,200 – 6,846,900. LD measured in r² units. Single nucleotide polymorphisms described in text are circled in red.

SE, or NWAK in the MrOS cohort (P > 0.05, additive mode of inheritance, data not shown). The SNP rs3816358 was nominally associated with sleep latency in the MrOS cohort ($\beta_{ADD} \pm$ SE in units of h, P_{ADD}: 0.05 ± 0.03, 0.04), raising the possibility that the association between rs3816358 and a later sleep onset time could be mediated by an increase in sleep latency. Although the association in the MrOS cohort between rs3816358 and sleep onset time was slightly attenuated after adjustment for sleep latency ($\beta_{ADD} \pm$ SE, P_{ADD}: 0.15 ± 0.05, 0.004; $P_{2DF} = 0.001$), the SNP association with sleep onset time remained nominally significant. The association between rs3816358 and wake onset time was essentially unchanged after adjustment for sleep latency ($\beta_{ADD} \pm$ SE, P_{ADD}: 0.16 ± 0.05, 0.002; $P_{2DF} = 0.006$).

An intronic SNP rs3768984 within *NPAS2* was also associated with later sleep onset time and wake onset time in the MrOS cohort, and the P value for the 2DF test exceeded the multiple testing

threshold (Figure 2, Table 4). Plots of adjusted mean sleep onset time and mean wake onset time by genotype indicated that a recessive genetic model is most appropriate for this SNP (Figure S3). Under a recessive mode of inheritance, rs3768984 was significantly associated with a later sleep onset time in MrOS participants ($\beta_{REC} \pm$ SE, P_{REC}: 0.47 ± 0.11, 9×10^{-6}) (Table 4). The direction of effect was the same in SOF participants, but the association failed to reach a nominal significance level (Table 4). This SNP also passed multiple test correction after meta-analysis under the recessive mode of inheritance ($\beta_{META} \pm$ SE, P_{META}: 0.34 ± 0.08, 6×10^{-5}) (Table 4). Similarly, rs3768984 was significantly associated with a later wake onset time in MrOS participants ($\beta_{REC} \pm$ SE, P_{REC}: 0.40 ± 0.09, 3×10^{-5}; $P_{2DF} = 1.7 \times 10^{-4}$) but not in SOF participants, and the meta-analysis results were also significant ($\beta_{META} \pm$ SE, P_{META}: 0.31 ± 0.08, 4×10^{-5}) (Table 4). Power was 0.56 to detect the meta-analysis effect size for sleep onset time (0.34 h) among 3,930 par-

Table 2—Genetic association with sleep continuity traits at chromosome 12p13

SNP/ location/ gene	Alleles MAF	Mode	Wake after sleep onset (min) $\beta \pm SE$ (P value)			Sleep efficiency (%) $\beta \pm SE$ (P value)		
			MrOS n = 2,488	SOF n = 1,401	Meta n = 3,889	MrOS n = 2,488	SOF n = 1,401	Meta n = 3,889
rs1047776/ 3' UTR/	G/**A** 0.37	Add	4.22 ± 1.27 (9×10^{-4})	5.04 ± 1.82 (0.006)	4.49 ± 1.04 (**2×10^{-5}**)	-0.99 ± 0.29 (8×10^{-4})	-0.97 ± 0.38 (0.01)	-0.98 ± 0.23 (**2×10^{-5}**)
LEPREL2		Dom	4.96 ± 1.77 (0.005)	5.86 ± 2.51 (0.02)	5.26 ± 1.45 (3×10^{-4})	-1.10 ± 0.41 (0.007)	-1.04 ± 0.52 (0.05)	-1.08 ± 0.32 (8×10^{-4})
		Rec	6.55 ± 2.51 (0.009)	7.93 ± 3.65 (0.03)	7.00 ± 2.07 (7×10^{-4})	-1.66 ± 0.58 (0.004)	-1.70 ± 0.76 (0.03)	-1.68 ± 0.46 (3×10^{-4})
		2DF	(0.004)	(0.02)	(2×10^{-4})	(0.003)	(0.03)	(3×10^{-4})
			MrOS n = 2,527	SOF n = 1,406	Meta n = 3,933	MrOS n = 2,527	SOF n = 1,406	Meta n = 3,933
rs2238114/ intron	C/**A** 0.30	Add	-4.39 ± 1.32 (9×10^{-4})	-4.29 ± 1.90 (0.02)	-4.36 ± 1.08 (**5×10^{-5}**)	0.94 ± 0.30 (0.002)	0.76 ± 0.40 (0.05)	0.87 ± 0.24 (3×10^{-4})
USP5		Dom	-4.38 ± 1.71 (0.01)	-5.23 ± 2.48 (0.04)	-4.66 ± 1.41 (9×10^{-4})	0.97 ± 0.40 (0.01)	0.83 ± 0.52 (NS)	0.92 ± 0.31 (0.004)
		Rec	-9.18 ± 2.97 (0.002)	-6.12 ± 4.23 (NS)	-8.17 ± 2.43 (8×10^{-4})	1.86 ± 0.69 (0.007)	1.37 ± 0.88 (NS)	1.68 ± 0.54 (0.002)
		2DF	(0.002)	(NS)	(4×10^{-4})	(0.006)	(NS)	(0.002)

NS: P > 0.05. MAF is weighted average of MAF in MrOS and SOF. The effect allele and P values passing multiple test correction are bolded. Add, additive, 2DF, two degrees of freedom; Dom, dominant; MAF, minor allele frequency; MrOS, The Osteoporotic Fractures in Men study; Rec, recessive; SE, standard error; SNP, single nucleotide polymorphism; SOF, The Study of Osteoporotic Fractures.

ticipants (two-sided $\alpha = 1.7 \times 10^{-4}$, recessive mode of inheritance). Both rs3816358 in *ARNTL* and rs3768984 in *NPAS2* were nominally associated with a later acrophase (peak activity time later in the day) in the MrOS cohort and in the meta-analysis, but the association P value did not pass multiple test correction (Table S4). The *NPAS2* SNP rs3768984 was not significantly associated with TST, WASO, SE, sleep latency, or NWAK in the MrOS cohort (P > 0.05, recessive mode of inheritance, data not shown).

As rs3816358 and rs3768984 were significantly associated with later sleep timing and are located in genes coding for protein binding partners, we investigated the statistical interaction between these two SNPs. With an additive mode of inheritance for rs3816358 and a recessive mode of inheritance for rs3768984, there was a significant interaction between these SNPs for sleep onset time and wake onset time ($\beta_{INT} \pm SE$, $P_{INT} = 0.72 \pm 0.24$, 0.003 and 0.71 ± 0.22, 0.001, respectively) in MrOS participants (Figure 3). The effect size of the association between the *ARNTL* SNP rs3816358 and sleep onset time was much larger among MrOS participants with an *NPAS2* SNP rs3768984 C/C genotype ($\beta_{ADD} \pm SE$, $P_{ADD} = 0.78 \pm 0.34$, 0.02) than those with A/A or A/C rs3768984 genotypes ($\beta_{ADD} \pm SE$, $P_{ADD} = 0.15 \pm 0.06$, 0.007). Consistently, the effect size of the association between rs3768984 and sleep onset time increased with the number of minor alleles of rs3816358 (rs3816358 G/G: $\beta_{REC} \pm SE$, $P_{REC} = 0.32 \pm 0.11$, 0.006; rs3816358 G/T: $\beta_{REC} \pm SE$, $P_{REC} = 1.03 \pm 0.27$, 0.00014; rs3816358 T/T: $\beta_{REC} \pm SE$, $P_{REC} = 2.36 \pm 1.86$, NS). These two SNPs also interacted in their association with acrophase in the MrOS cohort ($\beta_{INT} \pm SE$, $P_{INT} = 0.96 \pm 0.24$, 5×10^{-5}).

Table 3—Single nucleotide polymorphism associations with gene expression in HapMap CEU lymphoblastoid cell lines at chromosome 12p13

Gene[a]	rs2238114 C allele Correlation (P value)[b]	rs1047776 A allele Correlation (P value)[b]
GNB3	-0.10 (0.31)	-0.10 (0.33)
CDCA3	-0.05 (0.64)	0.10 (0.29)
USP5	0.02 (0.86)	-0.04 (0.68)
TPI1	-0.01 (0.92)	0.03 (0.75)
SPSB2	-0.23 (0.02)	-0.14 (0.14)
LRRC23 (exon 7)	-0.35 (2×10^{-4})	-0.17 (0.08)
LRRC23 (exon 8)	-0.50 (5×10^{-8})	-0.22 (0.02)
ENO2	0.00 (0.99)	-0.05 (0.63)
ATN1	0.07 (0.50)	0.03 (0.76)
PTPN6	0.10 (0.32)	0.10 (0.29)

[a]Gene in which expression probe is located. Exonic location of a probe is indicated for genes with multiple probes. [b]Spearman rank correlation coefficient.

SNP Associations With Sleep Continuity and Activity Rhythm Traits in the NPAS2 Gene and at Chromosome 12p13

In the SOF cohort, P values from the recessive inheritance mode and the 2DF test exceeded multiple testing thresholds for the association between the *NPAS2* SNP rs895520 and the pseudo-F statistic, a measure of activity rhythm robustness (Table 5,

Table 4—Genetic association with sleep timing-related traits

SNP/ location/ gene	Alleles MAF	Mode	Sleep onset time (h) β ± SE (P value)			Wake onset time (h) β ± SE (P value)		
			MrOS n = 2,524	SOF n = 1,406	Meta n = 3,930	MrOS n = 2,524	SOF n = 1,406	Meta n = 3,930
rs3816358/ intron/	G/**T** 0.10	Add	0.19 ± 0.06 (7 × 10⁻⁴)	0.01 ± 0.08 (NS)	0.13 ± 0.05 (0.005)	0.16 ± 0.05 (0.001)	0.07 ± 0.07 (NS)	0.13 ± 0.04 (0.002)
ARNTL		Dom	0.24 ± 0.06 **(9 × 10⁻⁵)**	-0.02 ± 0.09 (NS)	0.16 ± 0.05 (0.002)ᵃ	0.19 ± 0.06 **(8 × 10⁻⁴)**	0.07 ± 0.08 (NS)	0.15 ± 0.05 (0.001)
		Rec	-0.22 ± 0.24 (NS)	0.32 ± 0.31 (NS)	-0.01 ± 0.19 (NS)	0.08 ± 0.22 (NS)	0.24 ± 0.28 (NS)	0.14 ± 0.17 (NS)
		2DF	**(1 × 10⁻⁴)**	(NS)	(0.006)ᵃ	(0.003)	(NS)	(0.007)
			MrOS n = 2,527	SOF n = 1,407	Meta n = 3,934	MrOS n = 2,527	SOF n = 1,407	Meta n = 3,934
rs3768984/ intron/	A/**C** 0.24	Add	0.10 ± 0.04 (0.01)	-0.01 ± 0.05 (NS)	0.06 ± 0.03 (NS)	0.08 ± 0.04 (0.02)	0.07 ± 0.05 (NS)	0.08 ± 0.03 (0.008)
NPAS2		Dom	0.04 ± 0.05 (NS)	-0.05 ± 0.07 (NS)	0.01 ± 0.04 (NS)	0.04 ± 0.05 (NS)	0.06 ± 0.06 (NS)	0.05 ± 0.04 (NS)
		Rec	0.47 ± 0.11 **(9 × 10⁻⁶)**	0.11 ± 0.14 (NS)	0.34 ± 0.08 **(6 × 10⁻⁵)ᵃ**	0.40 ± 0.09 **(3 × 10⁻⁵)**	0.16 ± 0.13 (NS)	0.31 ± 0.08 **(4 × 10⁻⁵)**
		2DF	**(5 × 10⁻⁵)**	(NS)	(2 × 10⁻⁴)	**(1.7 × 10⁻⁴)**	(NS)	(0.01)

NS: nominal P > 0.05. MAF is weighted average of MAF in MrOS and SOF. The effect allele and P values passing multiple test correction are bolded. ᵃHeterogeneity P < 0.05. Add, additive; 2DF, two degrees of freedom; Dom, dominant; MAF, minor allele frequency; MrOS, The Osteoporotic Fractures in Men study; Rec, recessive; SE, standard error; SNP, single nucleotide polymorphism; SOF, The Study of Osteoporotic Fractures.

Figure 2). Results from the recessive model were based on 252 TT SOF participants (18% homozygous TT genotype frequency). The two *NPAS2* SNPs that exceeded multiple testing thresholds in this study (rs895520 and rs3768984) are not in LD ($r^2 = 0.02$ based on SNP genotypes from MrOS and SOF participants).

Also in the SOF cohort, P values from the recessive inheritance mode and the 2DF test exceeded multiple testing thresholds for the association between rs1047776 at chromosome 12p13 and sleep latency (Table 5). The homozygous AA genotype frequency was 13% (183 AA genotypes) in the SOF cohort. Given the rs1047776 SNP association with WASO and SE (Table 2), conditional analysis was performed. Adjustment for WASO or SE did not greatly change the association between rs1047776 and sleep latency in the SOF cohort (WASO adjusted $P_{2DF} = 3 \times 10^{-4}$, SE adjusted $P_{2DF} = 4 \times 10^{-4}$).

Rare RORC SNP Associated With Sleep Continuity

The SNP rs4284267 within the *RORC* 5' flanking region was significantly associated with WASO and SE in the MrOS cohort after multiple test correction, but the MAF was 5%, resulting in low power and potentially unstable, biased effect estimates (Table S5). Given the low MAF of rs4284267 combined with the recessive genetic model, this result should be considered preliminary.
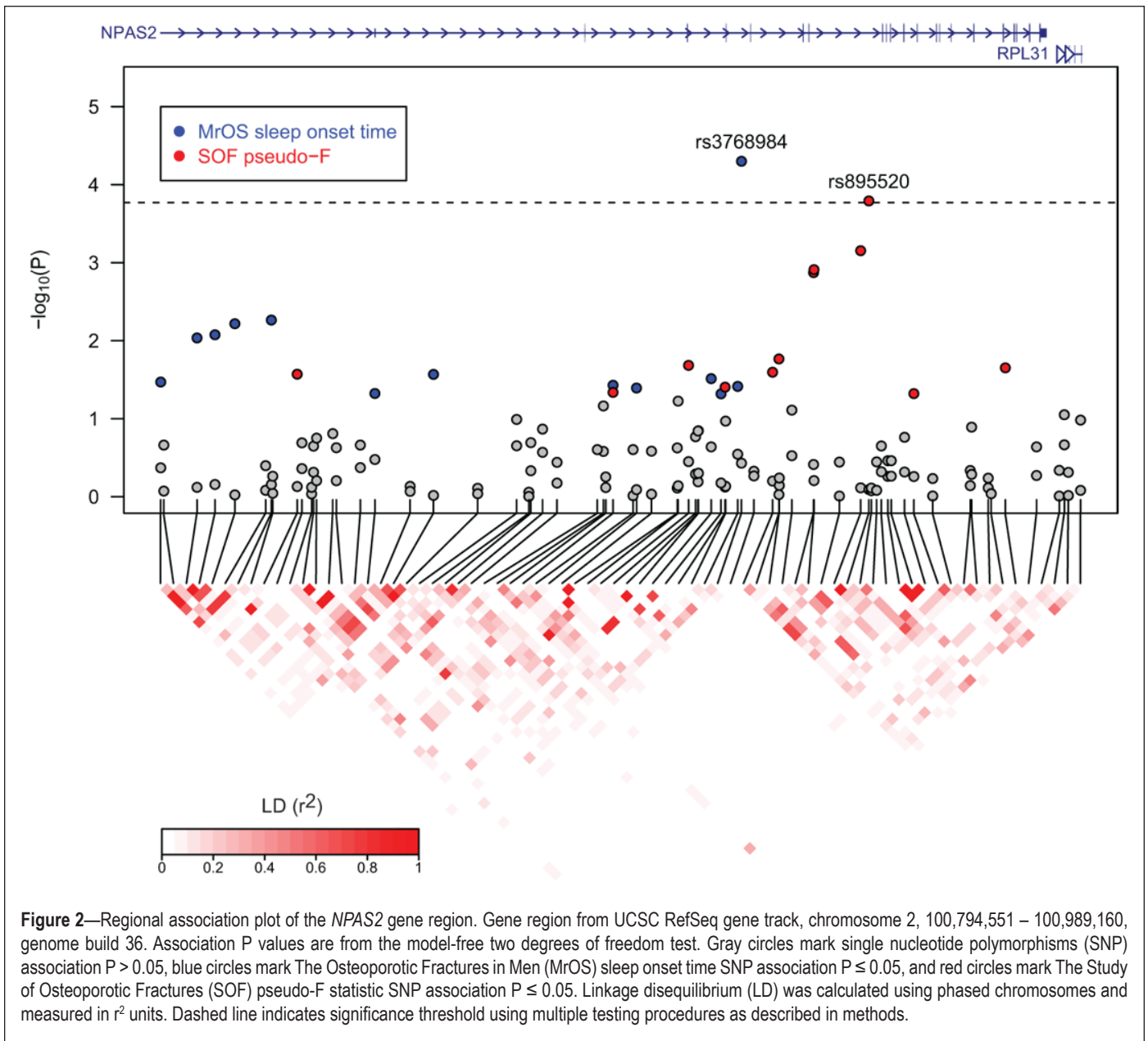
Phenotypic and Genotypic Correlations

It could be expected that rs3768984 would be associated with sleep onset time, wake onset time, and acrophase, as these three traits are highly correlated. To explore the relationship be-

tween genetic associations of multiple correlated traits, the correlation matrices of phenotypes and genetic associations were compared (Figure 4). Genetic associations are represented by SNP association test statistics so that significance level and effect direction are considered, and the genetic correlation is the correlation of these test statistics for different actigraphic traits. Traits with strong phenotypic correlation also showed strong genetic correlation, as depicted by nearly complete overlap between black and blue ellipses (Figure 4, upper triangle) and nearly identical correlation values (Figure 4, lower triangle).

Hierarchical cluster analysis was performed to identify clusters such that SNP associations within clusters were more similar to each other than to those in other clusters. Genetic associations for the actigraphic traits can be clustered into four groups: wake onset time and acrophase; sleep onset time and TST; pseudo-F statistic, amplitude, mesor, and nap min; and NWAK, WASO, SE, and sleep latency (Figure 4, gray triangles, and Figure S4). The existence of four clusters identified by hierarchical cluster analysis was supported by principal component analysis of the correlation matrices. The first four principal components explained 97% and 96% of the phenotypic and genotypic variance, respectively.

Replication Analysis

A previously reported candidate gene study of 194 SNPs (Bonferroni corrected P value = 2.6×10^{-4}) in clock genes and self-reported sleep duration using the MCTQ was conducted in a discovery sample of approximately 360 participants, and the top

Figure 2—Regional association plot of the *NPAS2* gene region. Gene region from UCSC RefSeq gene track, chromosome 2, 100,794,551 – 100,989,160, genome build 36. Association P values are from the model-free two degrees of freedom test. Gray circles mark single nucleotide polymorphisms (SNP) association P > 0.05, blue circles mark The Osteoporotic Fractures in Men (MrOS) sleep onset time SNP association P ≤ 0.05, and red circles mark The Study of Osteoporotic Fractures (SOF) pseudo-F statistic SNP association P ≤ 0.05. Linkage disequilibrium (LD) was calculated using phased chromosomes and measured in r[2] units. Dashed line indicates significance threshold using multiple testing procedures as described in methods.

two SNP associations from the discovery stage (rs12649507 and rs11932595, both located in the *CLOCK* gene) were tested for replication in 1,011 participants.[32] In a subsequent meta-analysis of genome-wide association studies (GWAS) of MCTQ-based self-reported sleep duration, these two *CLOCK* SNPs were not imputed in many of the cohorts, and as such, association statistics for the replication of these two *CLOCK* SNPs were not reported.[41] In the original candidate gene study, the association between sleep duration and rs12649507 was significant in the discovery stage (P = 0.0051), the replication stage (P = 0.045), and the meta-analysis (P = 0.0087).[32] In the same report, the association between sleep duration and rs11932595 was significant in the discovery stage (P = 0.0080) and the replication stage (P = 0.047), but apparently the direction of effect was not the same, as the association in the meta-analysis was not significant (P = 0.24).[32] Unfortunately, the effect size and effect direction were not reported, thus preventing the determination of the power to replicate or whether the effect direction replicates.

Of the top two SNPs, rs11932595 was genotyped in our study but rs12649507 was not. The SNP rs9312661 was in high LD with rs12649507 based on HapMap CEU release 22 genotypes (r[2] = 0.80) and was selected as a proxy SNP. There was no evidence that rs11932595 was associated with TST in the MrOS or SOF cohorts, but the A allele of the proxy SNP for rs12649507 was associated with less TST at a nominal level of significance (P ≤ 0.05) in the MrOS cohort and the meta-analysis but not in the SOF cohort (Table S6). Associations between the two *CLOCK* SNPs and all sleep traits and activity rhythm traits listed in Table 1 were also examined. In addition to TST, the A allele of the proxy SNP for rs12649507 was associated (P ≤ 0.05) with more napping minutes and lower mesor of the activity rhythm in the MrOS cohort and in the meta-analysis (Table S6). The decrease in TST (-4.06 min) associated with rs12649507 was similar to the increase in nap minutes (3.59 min) (Table S6). The rs11932595 G allele was associated (P ≤ 0.05) with higher activity rhythm amplitude and higher
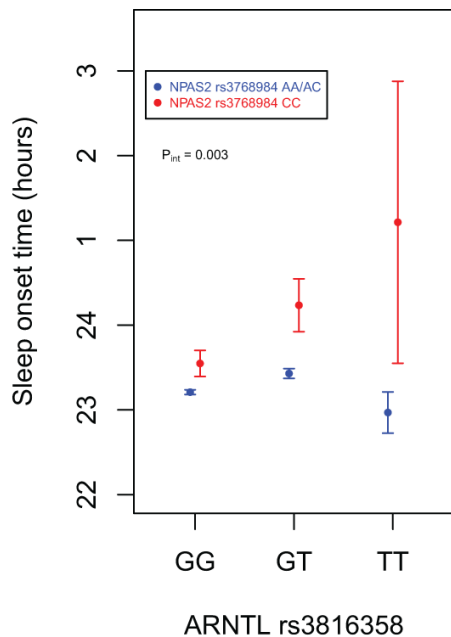
Figure 3—Interaction between rs3768984 (*NPAS2*) and rs3816358 (*ARNTL*) on sleep onset time. Adjusted means (filled circles) with standard errors of the predicted means are shown using results from The Osteoporotic Fractures in Men study participants. Sleep onset time adjusted for the same covariates as regression models, as described in the Methods section.

pseudo-F statistic (indicative of a more robust activity rhythm) in the SOF cohort (Table S6).

The previously published GWAS of MCTQ-based self-reported sleep duration reported SNPs near and within clock genes associated at a nominal significance level ($P \le 0.05$). Although we did not identify significant SNP associations with TST, we investigated whether the SNPs associated with traits other than TST in our current study might have been previously reported to be associated with MCTQ-based sleep duration.[41] The C allele of the *NPAS2* SNP rs3768984, which was associated with later sleep timing in the MrOS cohort, was significantly associated with lower MCTQ-based sleep duration ($\beta \pm$ SE, P = -0.07 ± 0.03, 0.008). None of the other SNPs we identified were reported to be associated at a nominal significance level with MCTQ-based sleep duration.

DISCUSSION

We present evidence from the analysis of two large population-based cohorts of elderly participants that common genetic variation in *ARNTL* and *NPAS2* and at chromosome 12p13 is significantly associated with objectively measured actigraphic sleep and activity rhythm traits. Two SNPs (rs1047776 and rs2238114) at chromosome 12p13 were associated with sleep continuity traits in both cohorts, and the meta-analysis P value exceeded the multiple testing significance threshold. SNP rs1047776 was also significantly associated with sleep latency in the SOF cohort. SNP rs2238114 resided in a region of high

Table 5—Genetic association with sleep latency and activity rhythm

SNP/ location/ gene	Alleles MAF	Mode	Sleep latency (min) β ± SE (P value)			Pseudo-F statistic β ± SE (P value)		
			MrOS n = 2,526	SOF n = 1,406	Meta n = 3,932	MrOS n = 2,526	SOF n = 1,396	Meta n = 3,922
rs895520/ intron/	C/**T** 0.43	Add	-0.39 ± 1.00 (NS)	1.21 ± 1.55 (NS)	0.08 ± 0.84 (NS)	3.10 ± 15.21 (NS)	-59.60 ± 18.05 (0.001)	-22.93 ± 11.63 (0.05)[a]
NPAS2		Dom	-0.36 ± 1.47 (NS)	1.99 ± 2.27 (NS)	0.34 ± 1.23 (NS)	-0.31 ± 22.40 (NS)	-35.61 ± 26.57 (NS)	-14.97 ± 17.12 (NS)
		Rec	-0.68 ± 1.72 (NS)	0.87 ± 2.70 (NS)	-0.23 ± 1.45 (NS)	9.59 ± 26.16 (NS)	-132.10 ± 31.55 (**3 × 10⁻⁵**)	-48.14 ± 20.14 (0.02)[a]
		2DF	(NS)	(NS)	(NS)	(NS)	(**1.6 × 10⁻⁴**)	(0.02)[a]
			MrOS n = 2,488	SOF n = 1,401	Meta n = 3,889	MrOS n = 2,488	SOF n = 1,391	Meta n = 3,879
rs1047776/ 3'UTR/	G/**A** 0.37	Add	1.68 ± 0.96 (NS)	3.76 ± 1.50 (0.01)	2.28 ± 0.81 (0.005)	-1.42 ± 14.48 (NS)	-31.79 ± 17.43 (NS)	-13.82 ± 11.14 (NS)
LEPREL2		Dom	1.99 ± 1.33 (NS)	1.12 ± 2.07 (NS)	1.74 ± 1.12 (NS)	-7.26 ± 20.20 (NS)	-37.19 ± 24.08 (NS)	-19.63 ± 15.48 (NS)
		Rec	2.60 ± 1.89 (NS)	12.80 ± 3.00 (**2 × 10⁻⁵**)	5.51 ± 1.60 (**6 × 10⁻⁴**)[a]	9.10 ± 28.70 (NS)	-49.51 ± 35.01 (NS)	-14.46 ± 22.20 (NS)
		2DF	(NS)	(**8 × 10⁻⁵**)	(**8 × 10⁻⁴**)	(NS)	(NS)	(NS)

NS: P > 0.05. MAF is weighted average of MAF in MrOS and SOF. The effect allele and P values passing multiple test correction are bolded. [a]Heterogeneity P < 0.05. Add, additive; 2DF, two degrees of freedom; Dom, dominant; MAF, minor allele frequency; MrOS, The Osteoporotic Fractures in Men study; Rec, recessive; SE, standard error; SNP, single nucleotide polymorphism; SOF, The Study of Osteoporotic Fractures.

Figure 4—Phenotypic and genetic correlation of actigraphic sleep and activity rhythm traits. The upper triangle depicts correlation graphically, and the lower triangle shows correlation values, with phenotypic correlations in black and genetic correlations in blue. Phenotypic correlations (Pearson's *r*) are calculated using The Osteoporotic Fractures in Men study data; similar results are seen using The Study of Osteoporotic Fractures phenotypic data. Genetic correlations (Pearson's *r*) are based on the meta-analysis single nucleotide polymorphisms test-statistics using the additive genetic model. A correlation of 0 is depicted with a circle, and correlations approaching 1 or -1 are depicted with more narrow ellipses to the right or left, respectively. Gray triangles mark the clusters formed by hierarchical cluster analysis of genetic associations. NWAK, number of awakenings; TST, total sleep time; WASO, wake after sleep onset.

LD that included nine genes, and the rs1047776 and rs2238114 alleles that were associated with higher WASO were also significantly associated with lower *LRRC23* gene expression in HapMap CEU lymphoblastoid cell lines. SNP associations that were significant in a single cohort were also identified. The *ARNTL* SNP rs3816358 and the *NPAS2* SNP rs3768984 were significantly associated with later sleep and wake onset time in the MrOS cohort. ARNTL and NPAS2 biochemically interact to form a heterodimeric transcription factor that acts at the core of the circadian rhythm pathway, and the *ARNTL* SNP rs3816358 and the *NPAS2* SNP rs3768984 statistically interact as well (sleep onset time $P_{INT} = 0.003$, wake onset time $P_{INT} = 0.001$). In addition, a second *NPAS2* SNP (rs895520) was significantly associated with the pseudo-F statistic, a measure of activity rhythm robustness, in the SOF cohort. Our study

achieved adequate power (power > 0.8) to detect an effect size of 0.15 SD units at MAF values even lower than 0.20 and approached adequate power to detect an effect size of 0.10 SD units at an MAF of 0.40 (Table S7); thus, undetected SNP associations are likely to have small effect sizes. Cluster analysis revealed that the genetic associations for all examined actigraphic traits could be grouped into four clusters. A previously identified SNP association in the *CLOCK* gene with self-reported sleep duration replicated in the MrOS cohort, and the *NPAS2* SNP rs3768984 we discovered to be associated with later sleep timing was also associated with sleep duration from a previous GWAS meta-analysis.[32,41]

The *GNB3* locus was included in our investigation based on the previous report that the 825C/T allele (rs5443) of *GNB3* interacted with the 3111C/T allele of *CLOCK* in an associa-

tion with diurnal preference among healthy Korean college students.[54] We attempted to directly genotype rs5443, but the Illumina designability score (range 0 - 1.1, designability cut-off = 0.6) was only 0.65 for rs5443, and not surprisingly, this SNP failed our genotyping assay. The HapMap project also had difficulty genotyping this SNP (50% missing rate in CEU samples). Although the *GNB3* SNP rs5443 was not successfully genotyped in our samples, a proxy SNP, rs2238114, was successfully genotyped. The rs2238114 SNP was in high LD with rs5443 (HapMap CEU r^2 = 0.95), but rather than being within the *GNB3* gene, it was located downstream of *GNB3* in a near-by gene, *USP5* (Figure 1). The SNP rs2238114 was not associated with sleep timing in the MrOS or SOF cohorts, but it was significantly associated with sleep continuity traits in the meta-analysis of MrOS and SOF results. The SNP rs1047776 located in *LEPREL2*, a gene upstream of *GNB3*, was also observed to be associated with sleep continuity traits in both cohorts. The low LD between the *LEPREL2* SNP rs1047776 and the *USP5* SNP rs2238114 raised the possibility that genetic variation in *LEPREL2* might be independently associated with sleep traits, but conditional analysis proved otherwise. Thus, the *LEPREL2* and *USP5* SNP associations appear to represent the same genetic signal.

To further examine the *GNB3* genetic association region, we characterized the LD structure of the *GNB3* region from the HapMap CEU population of European descent, which revealed that the *GNB3* SNP rs5443 and the *USP5* SNP rs2238114 reside on a 101 kb haplotype at chromosome 12p13 that spans multiple genes (*GNB3*, *CDCA3*, *USP5*, *TPI1*, *SPSB2*, *LRRC23*, *ENO2*, *ATN1*, and *PTPN6*) (Figure S1). The fact that SNPs in the *GNB3* region reside on a large haplotype aids in the interpretation of our results and previously published associations between rs5443 and multiple diverse traits. The initial discovery that the coding synonymous *GNB3* SNP rs5443 alters *GNB3* transcript splicing, resulting in an in-frame deletion of 41 amino acids, has led to the hypothesis that this SNP represents the causal allele for this locus.[70] Subsequent genetic analysis of the *GNB3* locus has focused on this single SNP, and a wide variety of phenotypes, including hypertension, circadian-related traits, depression, response to antidepressant medication, obesity, and insulin resistance have been reported to be associated with rs5443.[54,70-73] Certainly, these traits could have a common molecular etiology that involves *GNB3*. Considering the large LD block that includes rs5443, genetic associations with rs5443 could also reflect the influence of SNPs in the other genes located in the large haplotype block in this genomic region. In fact, rs5443 is in high LD (HapMap CEU r^2 = 0.96) with rs2226955, a coding synonymous SNP within the fourth exon of *USP5*, a protein involved in ubiquitin-mediated proteolysis.[74] The *GNB3* SNP rs5443 is also in high LD (HapMap CEU r^2 = 0.73) with rs710415, an E276G missense SNP (glutamine to glycine) in the *LRRC23* gene. In an attempt to narrow down the list of genes that might be regulated by the chromosome 12p13 SNPs associated with sleep continuity in the MrOS and SOF cohorts, publicly available eQTL data were examined. The rs1047776 and rs2238114 alleles associated with higher WASO were significantly associated with lower *LRRC23* gene expression in HapMap CEU lymphoblastoid cell lines, suggesting a link between lower levels of *LRRC23* gene expression and higher

levels of sleep fragmentation. Little is known about *LRRC23* (leucine-rich repeat containing 23) gene function. *LRRC23* is expressed in a wide range of tissues (Figure S2B). The leucine-rich repeat (LRR) protein domain is involved in protein-protein interactions, and a recent bioinformatic categorization of 375 human LRR-containing proteins grouped LRRC23 into a nuclear and centrosome cluster.[75] Publicly available eQTL data along with our SNP association results point toward *LRRC23* as an interesting candidate for further study, however, human lymphoblastoid cell lines might not faithfully reproduce gene expression regulation in human tissue relevant to the regulation of sleep-wake patterns. Follow-up eQTL studies in neural tissues relevant to sleep-wake regulation will be necessary to strengthen the evidence for *LRRC23*'s role in sleep regulation. Furthermore, future studies conducted in populations of African descent where the LD in this region is greatly reduced can help to disentangle the associations between multiple traits and SNPs in this region. Our examination of the chromosome 12p13 region surrounding the SNPs we genotyped near our candidate gene *GNB3* contributes to a more complete understanding of genetic associations in this region.

To our knowledge, our study is the first to identify significant associations between *NPAS2* SNPs and sleep traits and activity rhythm traits in humans. SNPs in *NPAS2* and *ARNTL* were significantly associated with a shift toward later sleep timing in the MrOS cohort, a shift that was particularly striking given the well-established age-related shift toward earlier sleep timing.[3,21] Replication analysis in well-powered studies of younger participants will inform whether these SNP associations are specific to particular age groups. Although *NPAS2* has been shown to regulate sleep in the mouse,[29,76,77] its role in the regulation of sleep in humans has been less clear. For instance, seasonal affective disorder, a condition that is related to circadian rhythm disruption but is not a direct measure of sleep, has been associated with *NPAS2* genetic variants.[33,37] It is thought that ARNTL:CLOCK heterodimers control clock gene expression in the SCN, but in the forebrain, ARNTL partners with NPAS2.[28,30] NPAS2 might also partner with ARNTL in the SCN, but NPAS2 likely plays a less prominent role than CLOCK in the SCN.[29] Our observation of a statistical interaction between rs3768984 in *NPAS2* and rs3816358 in *ARNTL* is consistent with the biochemical interaction between the two proteins and strengthens the case that these SNP associations reflect a functional alteration in the circadian clock. Germline polymorphisms such as SNPs cannot provide insight into the tissue-specific role of NPAS2; thus, further studies will be needed to determine whether NPAS2:ARNTL heterodimers in central clocks (SCN) and/or peripheral clocks (forebrain) regulate sleep timing.

In addition to examining TST in the replication analysis of the *CLOCK* SNPs rs12649507 and rs11932595,[32] we also examined whether these two SNPs were associated with any of the actigraphic sleep and activity rhythm traits that were analyzed in our genetic association study. In the MrOS cohort but not the SOF cohort, the *CLOCK* SNP rs12649507 was associated with less TST, more nap minutes, and lower mesor (mean activity level from the fitted activity rhythm curve). Although the association between TST and the other *CLOCK* SNP, rs11932595, did not replicate, the association between this SNP and two ac-

tivity rhythm traits (higher amplitude and higher pseudo-F statistic) in the SOF cohort was indicative of a more robust activity rhythm. Examination of these SNP associations in cohorts of younger participants may help to determine whether the associations are age-dependent.

We also assessed whether the SNPs we discovered in the MrOS and SOF cohorts were reported in a GWAS of MCTQ-based average weekly sleep duration.[41] The weighted mean age of the female and male participants in the discovery cohorts of the GWAS meta-analysis was 45.8 yr and 46.9 yr, respectively, and the participants in the replication cohorts were even younger.[41] Sleep duration was the only trait examined in the previous GWAS, and although we did not discover SNPs that were significantly associated with total sleep time in our study, we did identify SNPs associated with related traits. The *NPAS2* SNP rs3768984, which was significantly associated with later sleep timing in the MrOS cohort, was reported to be associated with less MCTQ-based self-reported sleep duration. Later sleep timing, i.e., later chronotype, is related to sleep duration when work and free days are examined separately, although the relationship is nearly abolished when sleep duration is averaged across all days.[14] A later chronotype is associated with less sleep on work days, which is normally compensated for by more sleep on free days.[14] In the cohorts contributing to the previously reported GWAS meta-analysis of sleep duration, the *NPAS2* SNP rs3768984 might be associated with a later chronotype and shorter sleep duration on work days that is not fully compensated for on free days, resulting in shorter average weekly sleep duration. Under this model, rs3768984 would not be associated with sleep duration in the elderly who do not typically have work days, and indeed, rs3768984 was not associated with TST in the MrOS or SOF cohorts. Alternatively, the *NPAS2* SNP might regulate different sleep traits (sleep timing and sleep quantity) in different age groups, or the association might differ based on subjective and objective sleep measurements. Follow-up of the *NPAS2* SNP with chronotype measures in young and old individuals should help to resolve these questions.

Replication will be critical to establish the validity and generalizability of the SNP associations discovered in the MrOS and SOF participants. However, the SNP associations discovered in the MrOS and SOF cohorts of elderly participants might not replicate in cohorts of younger participants if the SNP associations with sleep traits differ by age, i.e., interaction between SNPs and age. Age-dependent relationships between sleep traits and circadian rhythm markers and age-related changes in the circadian and sleep homeostatic processes make SNP interactions with age plausible. For instance, the strong relationship found in young adults between circadian period and morningness-eveningness and the related trait, self-selected wake times, was found to be much weaker in older adults.[20] Thus, genetic variants affecting circadian period in the same manner across all age ranges could still be differentially associated with morningness-eveningness across age ranges, simply due to the nature of the age-dependent relationship between circadian period and morningness-eveningness. Other sleep traits that are related to circadian parameters in an age-dependent manner could also exhibit age-dependent associations with genetic variants. In addition, age-related changes in the circadian and sleep homeostatic processes are well documented. For in-

stance, the amplitude of circadian gene expression and other circadian outputs decreases with age.[25] Age-related decline of circadian rhythms potentially places the circadian clock on a threshold where subtle genetic variants that had little to no effect during the early years of life could have an effect during the later years of life. If age-specific genetic associations with sleep traits are identified, and the corresponding causal variants are discovered, the characterization of the effect of causal variants on the function of the molecular clock and the resultant circadian parameters could help to elucidate the fundamental clock properties that regulate sleep in different age groups. Our study represents a first step toward this goal, and further studies in cohorts composed of elderly and nonelderly participants will be needed to establish the existence of age-specific genetic associations with sleep traits.

The examination of SNP associations with multiple correlated actigraphic traits provided a more complete view of SNP associations compared with the conventional analysis of a single sleep trait. For instance, the consistency of SNP associations among correlated traits could be examined. Not surprisingly, SNP associations for correlated traits were similar, and this applied to all significance levels (nearly complete overlap of black and blue ellipses in Figure 4). In addition, we explored whether SNP associations were mediated by related sleep traits, e.g., whether *ARNTL* and *NPAS2* SNP associations with later sleep timing were mediated by sleep latency. We also performed cluster analysis and found that genetic associations fell into four clusters that largely mirrored phenotypic correlations. The existence of four clusters of genetic associations supports the notion that rather than examining 12 independent traits, our study examined four groups of traits. Even if multiple test adjustment was based on the number of independent SNPs (310) and clusters of traits (four), the significance threshold would be 4×10^{-5}, and many of the SNP associations reported here would remain significant.

One strength of this study is the use of objectively measured sleep and activity rhythm traits. A second strength is the examination of these traits in two cohorts composed of elderly individuals, an age group that experiences high rates of sleep disturbance. Sleep and activity rhythm traits have been associated with multiple adverse outcomes in the elderly, and thus are relevant to health in the elderly.[45,78,79] The disadvantages of the candidate gene approach have been discussed previously.[80] Briefly, candidate gene studies that were commonly performed prior to GWAS were plagued by the following problems: risk factors (candidate genes) chosen based on diverse considerations, studies were underpowered (especially considering what is now known about the effect size of common genetic variants), nominal significance thresholds of 0.05 were adopted, and confounding by population stratification was unaccounted for.[80] Our study avoids many of the pitfalls that plagued previously performed candidate gene studies. The clock gene pathway is well established from work in multiple model organisms, our choice of clock genes was clearly documented, and we systematically surveyed common genetic variation in these genes rather than choosing individual variants to test within these genes. Our sample size was sufficient to achieve adequate power to observe small effect sizes (Table S7). Achieving adequate power also means that our reported effect sizes are less

likely to be inflated.[81] Multiple test correction was applied using modern techniques that take LD into account. We accounted for population stratification by restricting our analysis to self-identified Caucasian participants and by adjusting for genetic ancestry using components from multidimensional scaling analyses in our regression models. Candidate gene studies can also miss genetic associations because common genetic variation in the entire genome is not surveyed. Indeed, genetic associations outside of the candidate gene regions may have been missed in this study. Despite this limitation, we made the serendipitous discovery that SNPs near the candidate gene *GNB3* reside on a large LD block on chromosome 12 that includes genes not previously known to be associated with sleep traits, which could eventually lead to the potential identification of novel sleep regulatory genes.

The identification of significant associations between objectively measured sleep and activity rhythm traits and SNPs in the highly conserved circadian rhythm gene pathway supports the notion that common genetic variation in these genes plays a role in sleep regulation in humans. SNPs at chromosome 12p13 were significantly associated with sleep continuity traits in both cohorts. Although associations between SNPs in the *ARNTL/NPAS2* genes and delayed sleep timing were limited to a single cohort, the statistical interaction between SNPs mirrored the known biochemical interaction between these proteins, suggesting that these SNPs or genetic variants in LD might be functionally relevant. The significance of *NPAS2* and *ARNTL* SNP associations in the MrOS cohort but not the SOF cohort could indicate that these SNP associations are sex-specific. Alternatively, the lack of association significance in the SOF cohort could indicate a lack of replication in an independent cohort or it could be due to the smaller sample size in SOF. Replication studies using sex-stratified analysis can help to resolve this question. Our results highlight the relevance of common genetic variation in clock genes in the regulation of sleep traits in the elderly and could motivate the further exploration of genetic associations with objectively measured sleep traits in multiple human populations.

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Table S1—Number of genotyped SNPs passing quality control filters per gene

Gene symbol	NCBI Gene ID	kb[a]	QC-passing SNPs in MrOS	QC-passing SNPs in SOF	kb/SNP[b]
BHLHE40/DEC1	8553	21.41	7	7	3.06
BHLHE41/DEC2	79365	16.5	7	7	2.36
ARNTL	406	113.81	27	26	4.22
CLOCK	9575	131.49	22	22	5.98
CRY1	1407	93.48	9	9	10.39
CRY2	1408	49.33	10	10	4.93
CSNK1D	1453	27.54	4	4	6.88
CSNK1E	1454	42.35	14	14	3.02
DBP	1628	8.28	4	4	2.07
FMR1	2332	47.81	16	0	2.99
FXR2	9513	14.57	4	3	3.64
GNB3	2784	18.01	6	6	3
GSK3A	2931	NA	1	1	NA
GSK3B	2932	278.38	44	44	6.33
HLF	3131	87.22	25	25	3.49
E4BP4/NFIL3	4783	31.86	13	13	2.45
NPAS2	4862	194.61	73	72	2.67
NR1D1/REVA	9572	1.48	6	6	0.25
NR1D2/RVRB	9975	38.89	7	7	5.56
OPN4	94233	25.57	10	11	2.56
PER1	5187	21.57	5	5	4.31
PER2	8864	52.95	20	19	2.65
PER3	8863	73.22	26	25	2.82
PGC1B	133522	127.43	40	40	3.19
RARA	5914	35.95	2	2	17.98
RORA	6095	723.48	74	74	9.78
RORC/RORγ	6097	38.73	13	13	2.98
RXRA	6256	130.63	27	26	4.84
TEF	7008	30.24	8	8	3.78
TIMELESS	8914	40.64	5	5	8.13
Total			529	508	

[a]Kilobases (kb) between genotyped SNPs passing QC on the most extreme 5' and 3' ends of the targeted locus. [b]Locus kb divided by the number of genotyped SNPs passing QC in MrOS. MrOS, Osteoporotic Fractures in Men; NCBI, National Center for Biotechnology Information; QC, quality control; SNP, single nucleotide polymorphism; SOF, Study of Osteoporotic Fractures.

Table S2—Genetic association with number of wake episodes at chromosome 12p13

SNP/location/gene	Alleles MAF	Mode	Number of wake episodes RR [95% CI] (P value)[a]		
			MrOS n = 2,488	SOF n = 1,401	Meta n = 3,889
rs1047776/ 3' UTR/	G/**A** 0.37	Add	1.04 [1.01-1.07] (0.002)	1.04 [1.01-1.07] (0.02)	1.04 [1.03-1.05] (**1 × 10⁻⁴**)
LEPREL2		Dom	1.05 [1.01-1.09] (0.006)	1.04 [1.00-1.09] (NS)	1.05 [1.03-1.06] (9 × 10⁻⁴)
		Rec	1.06 [1.01-1.11] (0.02)	1.06 [1.00-1.13] (0.05)	1.06 [1.04-1.08] (0.003)
		2DF	(0.008)	(NS)	(0.001)
			MrOS n = 2,527	SOF n = 1,406	Meta n = 3,933
rs2238114/ intron	C/**A** 0.30	Add	0.95 [0.93-0.98] (3 × 10⁻⁴)	0.96 [0.93-1.00] (0.02)	0.96 [0.95-0.97] (**2 × 10⁻⁵**)
USP5		Dom	0.94 [0.91-0.98] (7 × 10⁻⁴)	0.96 [0.92-1.00] (NS)	0.95 [0.94-0.96] (**1 × 10⁻⁴**)
		Rec	0.93 [0.87-0.99] (0.02)	0.94 [0.87-1.01] (NS)	0.93 [0.91-0.95] (0.003)
		2DF	(0.001)	(NS)	(3 × 10⁻⁴)

NS: P > 0.05. MAF is weighted average of MAF in MrOS and SOF. The effect allele and P values passing multiple test correction are bolded. [a]RR, rate ratio, ratio of number of wake episodes per h of time in bed. Add, additive; CI, confidence interval; 2DF, two degrees of freedom; Dom, dominant; MAF, minor allele frequency; MrOS, Osteoporotic Fractures in Men; Rec, recessive; SNP, single nucleotide polymorphism; SOF, Study of Osteoporotic Fractures.

Table S3—Conditional analysis of rs1047776 and rs2238114

SNP	Mode	Wake after sleep onset (min) β ± SE (P value)		Sleep efficiency (%) β ± SE (P value)	
		MrOS n = 2,488	SOF n = 1,400	MrOS n = 2,488	SOF n = 1,400
rs1047776[a]	Add	3.06 ± 1.35 (0.02)	4.13 ± 1.96 (0.04)	-0.75 ± 0.31 (0.02)	-0.83 ± 0.41 (0.04)
rs2238114[a]	Add	-3.51 ± 1.41 (0.01)	-2.73 ± 2.04 (NS)	0.73 ± 0.33 (0.03)	0.44 ± 0.43 (NS)

NS: P > 0.05. [a]Both SNPs included in the same regression model. Add, additive; MrOS, Osteoporotic Fractures in Men; SE, standard error; SNP, single nucleotide polymorphism; SOF, Study of Osteoporotic Fractures.

Table S4—Genetic association with acrophase

SNP/ location/ gene	Alleles MAF	Mode	Acrophase (h) $\beta \pm$ SE (P value)		
			MrOS n = 2,524	**SOF** n = 1,396	**Meta** n = 3,920
rs3816358/ intron/	G/**T** 0.10	Add	0.13 ± 0.06 (0.02)	0.00 ± 0.08 (NS)	0.08 ± 0.04 (NS)
ARNTL		Dom	0.15 ± 0.06 (0.01)	0.02 ± 0.08 (NS)	0.11 ± 0.05 (0.03)
		Rec	0.00 ± 0.24 (NS)	-0.18 ± 0.29 (NS)	-0.07 ± 0.18 (NS)
		2DF	(0.04)	(NS)	(NS)
			MrOS n = 2,527	**SOF** n = 1,397	**Meta** n = 3,924
rs3768984/ intron/	A/**C** 0.24	Add	0.08 ± 0.04 (0.05)	0.12 ± 0.05 (0.03)	0.09 ± 0.03 (0.003)
NPAS2		Dom	0.05 ± 0.05 (NS)	0.12 ± 0.06 (0.05)	0.08 ± 0.04 (0.04)
		Rec	0.29 ± 0.10 (0.004)	0.22 ± 0.13 (NS)	0.27 ± 0.08 (0.001)
		2DF	(0.02)	(NS)	(0.003)

NS: nominal P > 0.05. MAF is weighted average of MAF in MrOS and SOF. The effect allele and P values passing multiple test correction are bolded. Add, additive; 2DF, two degrees of freedom; Dom, dominant; MAF, minor allele frequency; MrOS, Osteoporotic Fractures in Men; Rec, recessive; SE, standard error; SNP, single nucleotide polymorphism; SOF, Study of Osteoporotic Fractures.

Table S5—*RORC* SNP with MAF ≤ 0.05 associated with sleep continuity traits

SNP/ location/ gene	Alleles MAF	Mode	Wake after sleep onset (min) $\beta \pm$ SE (P value)			Sleep efficiency (%) $\beta \pm$ SE (P value)		
			MrOS n = 2,527	**SOF** n = 1,406	**Meta** n = 3,933	**MrOS** n = 2,527	**SOF** n = 1,406	**Meta** n = 3,933
rs4284267/ 5' flank/	T/**A** 0.05	Add	1.57 ± 2.78 (NS)	-2.26 ± 4.00 (NS)	0.32 ± 2.29 (NS)	-0.47 ± 0.64 (NS)	0.58 ± 0.83 (NS)	-0.08 ± 0.51 (NS)
RORC		Dom	-0.81 ± 2.98 (NS)	-2.69 ± 4.13 (NS)	-1.46 ± 2.42 (NS)	0.12 ± 0.69 (NS)	0.66 ± 0.86 (NS)	0.33 ± 0.54 (NS)
		Rec	49.85 ± 12.99 **(1 × 10⁻⁴)**	11.83 ± 26.65 (NS)	42.53 ± 11.68 (3 × 10⁻⁴)	-12.53 ± 3.01 **(3 × 10⁻⁵)**	-1.67 ± 5.55 (NS)	-10.06 ± 2.65 **(1 × 10⁻⁴)**
		2DF	(4 × 10⁻⁴)	(NS)	(0.002)	**(1 × 10⁻⁴)**	(NS)	(8 × 10⁻⁴)

NS: P > 0.05. MAF is weighted average of MAF in MrOS and SOF. The effect allele and P values passing multiple test correction are bolded. Add, additive; 2DF, two degrees of freedom; Dom, dominant; MAF, minor allele frequency; MrOS, Osteoporotic Fractures in Men; Rec, recessive; SE, standard error; SNP, single nucleotide polymorphism; SOF, Study of Osteoporotic Fractures.

**Table S6**—Replication analysis of *CLOCK* SNPs

			TST (min)		
			MrOS n = 2,525-2,526 β ± SE (P value)	SOF n = 1,407 β ± SE (P value)	Meta-analysis n = 3,932-3,933 β ± SE (P value)
SNP	Alleles MAF[a]	Mode			
rs9312661[b]	G/**A**	ADD	-4.06 ± 2.08 (0.05)	-1.74 ± 2.89 (0.55)	-3.27 ± 1.69 (0.05)
	0.37	2DF	(0.06)	(0.61)	(0.07)
rs11932595	A/**G**	ADD	0.58 ± 2.09 (0.78)	-2.11 ± 2.81 (0.45)	-0.38 ± 1.68 (0.82)
	0.41	2DF	(0.95)	(0.57)	(0.77)
			Nap min		
			MrOS n = 2,522-2,523 β ± SE (P value)	SOF n = 1,390 β ± SE (P value)	Meta-analysis n = 3,912-3,913 β ± SE (P value)
SNP	Alleles MAF[a]	Mode			
rs9312661[b]	G/**A**	ADD	3.59 ± 1.48 (0.02)	2.26 ± 2.27 (0.32)	3.19 ± 1.24 (0.01)
	0.37	2DF	(0.03)	(0.61)	(0.04)
rs11932595	A/**G**	ADD	0.91 ± 1.48 (0.54)	-2.71 ± 2.21 (0.22)	-0.22 ± 1.23 (0.86)
	0.41	2DF	(0.79)	(0.25)	(0.37)
			Mesor		
			MrOS n = 2,525-2,526 β ± SE (P value)	SOF n = 1,397 β ± SE (P value)	Meta-analysis n = 3,922-3,923 β ± SE (P value)
SNP	Alleles MAF[a]	Mode			
rs9312661[b]	G/**A**	ADD	-21.58 ± 13.75 (0.12)	-1.17 ± 17.90 (0.95)	-14.01 ± 10.90 (0.20)
	0.37	2DF	(0.03)	(0.48)	(0.03)
rs11932595	A/**G**	ADD	1.70 ± 13.81 (0.90)	30.78 ± 17.39 (0.08)	12.95 ± 10.82 (0.23)
	0.41	2DF	(0.47)	(0.19)	(0.85)
			Amplitude		
			MrOS n = 2,525-2,526 β ± SE (P value)	SOF n = 1,397 β ± SE (P value)	Meta-analysis n = 3,922-3,923 β ± SE (P value)
SNP	Alleles MAF[a]	Mode			
rs9312661[b]	G/**A**	ADD	-53.68 ± 29.72 (0.07)	-13.50 ± 37.88 (0.72)	-38.36 ± 23.38 (0.10)
	0.37	2DF	(0.10)	(0.80)	(0.24)
rs11932595	A/**G**	ADD	12.60 ± 29.85 (0.67)	82.71 ± 36.76 (0.02)	40.46 ± 23.17 (0.08)
	0.41	2DF	(0.69)	(0.05)	(0.40)
			Pseudo-F statistic		
			MrOS n = 2,525-2,526 β ± SE (P value)	SOF n = 1,397 β ± SE (P value)	Meta-analysis n = 3,922-3,923 β ± SE (P value)
SNP	Alleles MAF[a]	Mode			
rs9312661[b]	G/**A**	ADD	-24.95 ± 14.32 (0.08)	1.14 ± 17.50 (0.95)	-14.49 ± 11.08 (0.19)
	0.37	2DF	(0.15)	(0.93)	(0.22)
rs11932595	A/**G**	ADD	-4.48 ± 14.38 (0.76)	39.54 ± 16.98 (0.02)	13.91 ± 10.97 (0.21)
	0.41	2DF	(0.74)	(0.03)	(0.31)

[a]MAF is weighted average of MAF in MrOS and SOF. The effect allele is bolded. [b]rs9312661 serves as a proxy SNP for rs12649507. ADD, additive; 2DF, two degrees of freedom; MAF, minor allele frequency; MrOS, Osteoporotic Fractures in Men; SE, standard error; SNP, single nucleotide polymorphism; SOF, Study of Osteoporotic Fractures; TST, total sleep time.

Table S7—Power to detect SNP associations in the present study

| Effect size (SD)	Power[a]				
	0.01 MAF	0.10 MAF	0.20 MAF	0.30 MAF	0.40 MAF
0.05	0.0005	0.0089	0.0254	0.0438	0.0573
0.10	0.002	0.1629	0.4429	0.6359	0.7286
0.15	0.0074	0.6585	0.9526	0.992	0.9975
0.20	0.0231	0.9643	0.9998	0.9999	0.9999
0.25	0.0603	0.9993	0.9999	0.9999	0.9999
0.30	0.1334	0.9999	0.9999	0.9999	0.9999
0.35	0.2519	0.9999	0.9999	0.9999	0.9999
0.40	0.4105	0.9999	0.9999	0.9999	0.9999

[a]Additive mode of inheritance, linear regression, two-sided $\alpha = 1.7 \times 10^{-4}$, 3,934 participants. MAF, minor allele frequency; SD, standard deviation.
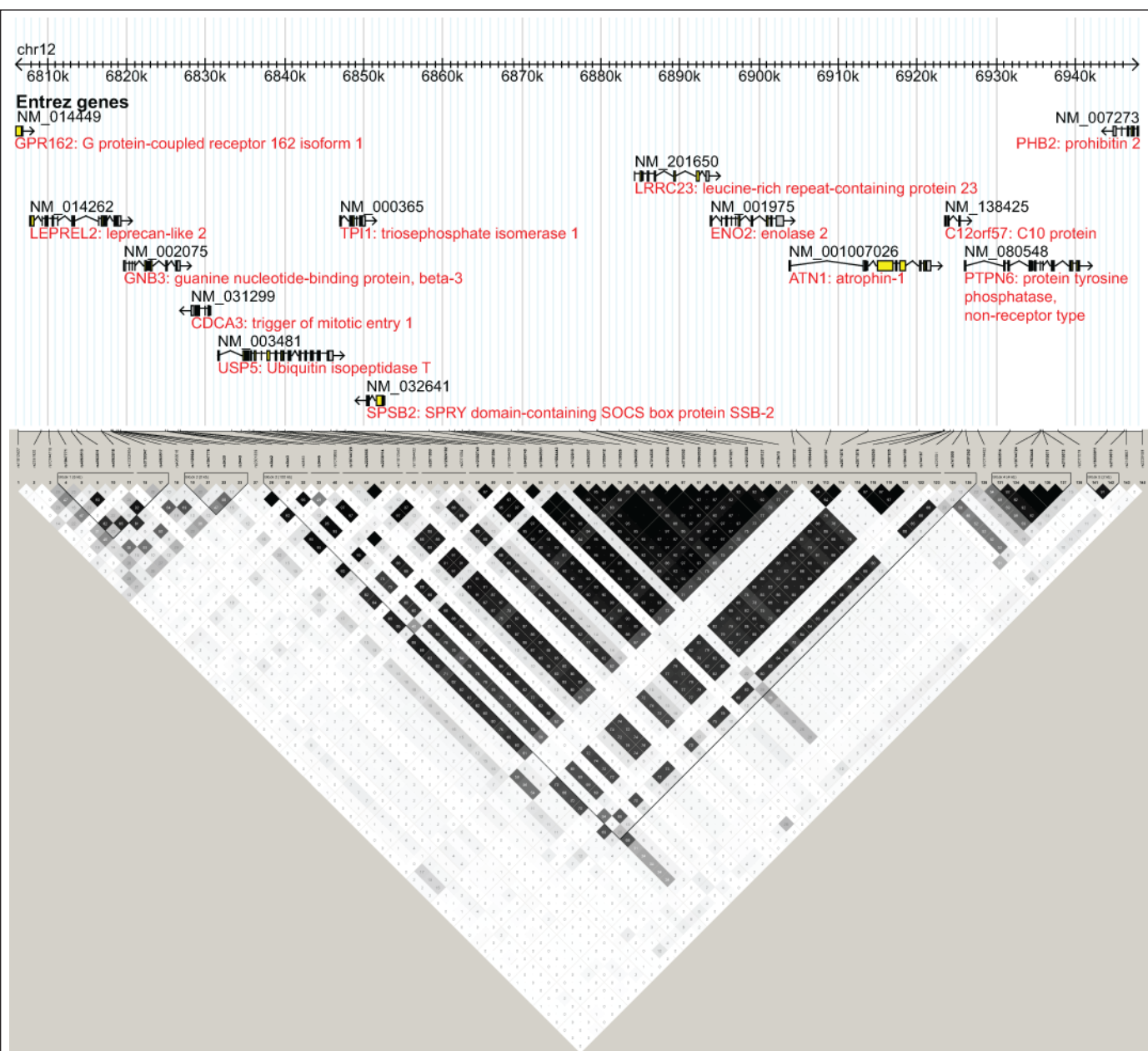

Figure S1—Linkage disequilibrium (LD) plot of single nucleotide polymorphisms (SNPs) at chromosome 12p13. LD plot generated using Haploview with data from CEU HapMap phase 2 and 3, release 28, Aug10, genome build 36, dbSNP b126 from chromosome 12, position 6,806,200 – 6,950,000. LD measured in r^2 units. Haplotype blocks estimated using the confidence interval method.[1]
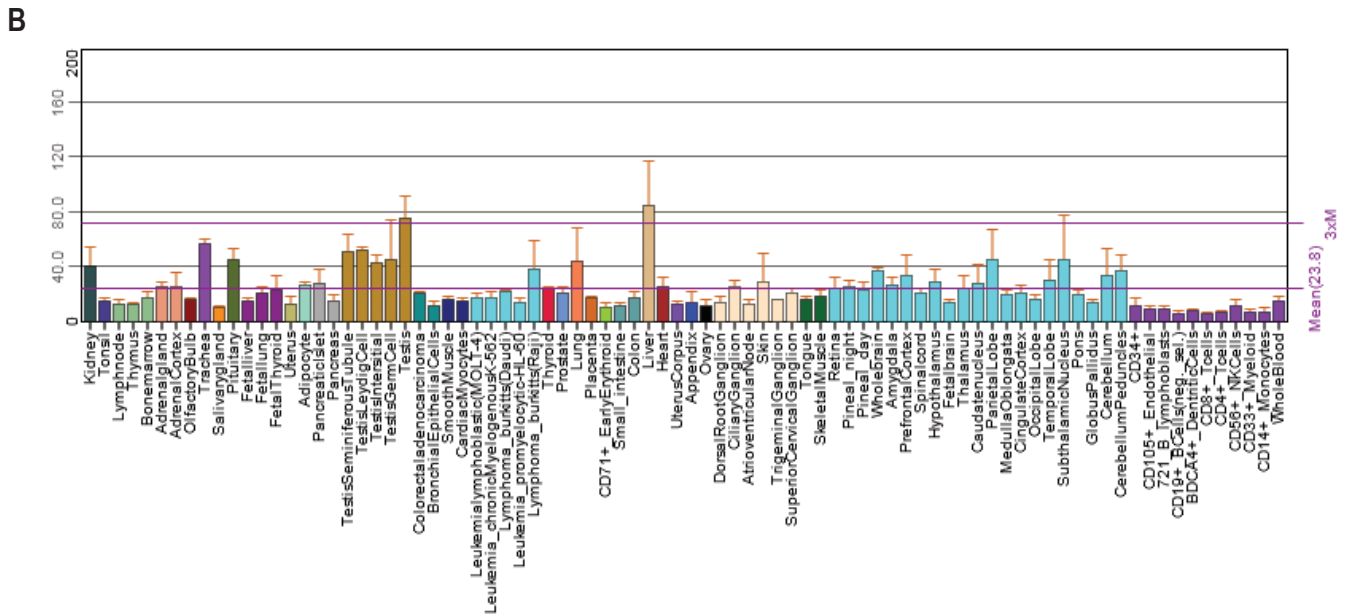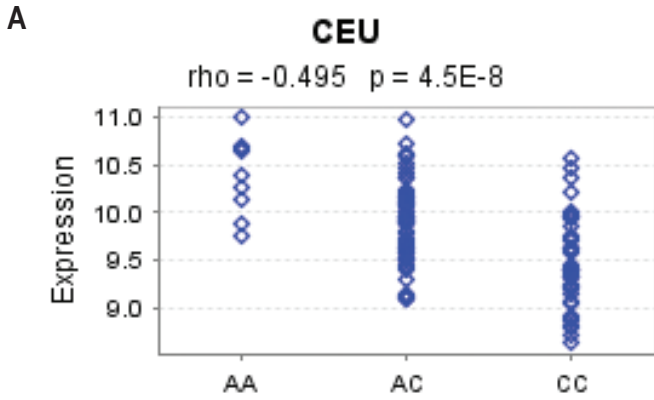
Figure S2—*LRRC23* gene expression association and tissue distribution. **(A)** SNP rs2238114 association with *LRRC23* gene expression in HapMap CEU lymphoblastoid cells. **(B)** Tissue distribution of *LRRC23* gene expression.
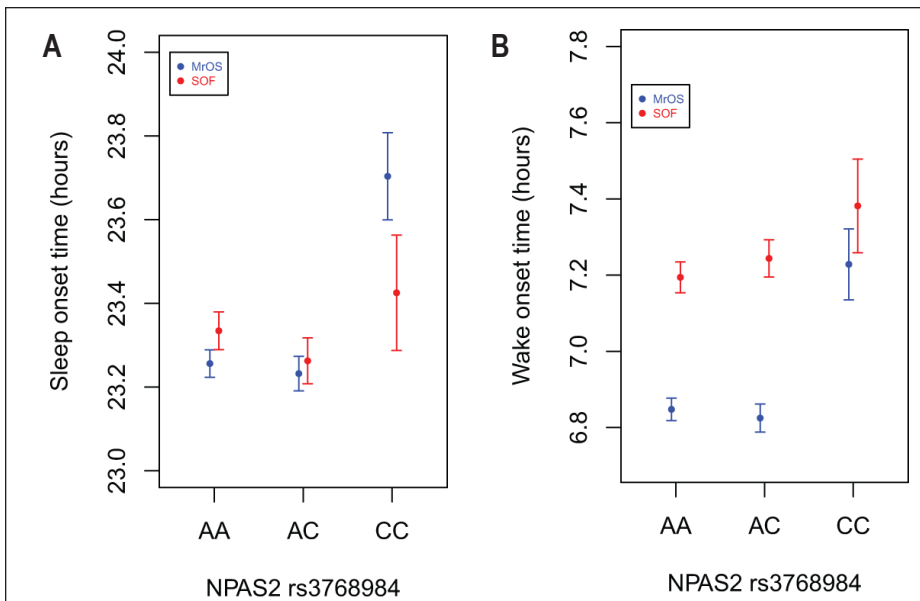
Figure S3—Adjusted mean sleep timing related traits by *NPAS2* rs3768984 genotype. Adjusted means (filled circles) with standard errors of the predicted means are shown. Sleep traits adjusted for the same covariates as regression models, as described in methods. **(A)** Sleep onset time. **(B)** Wake onset time. MrOS, Osteoporotic Fractures in Men; SOF, Study of Osteoporotic Fractures.
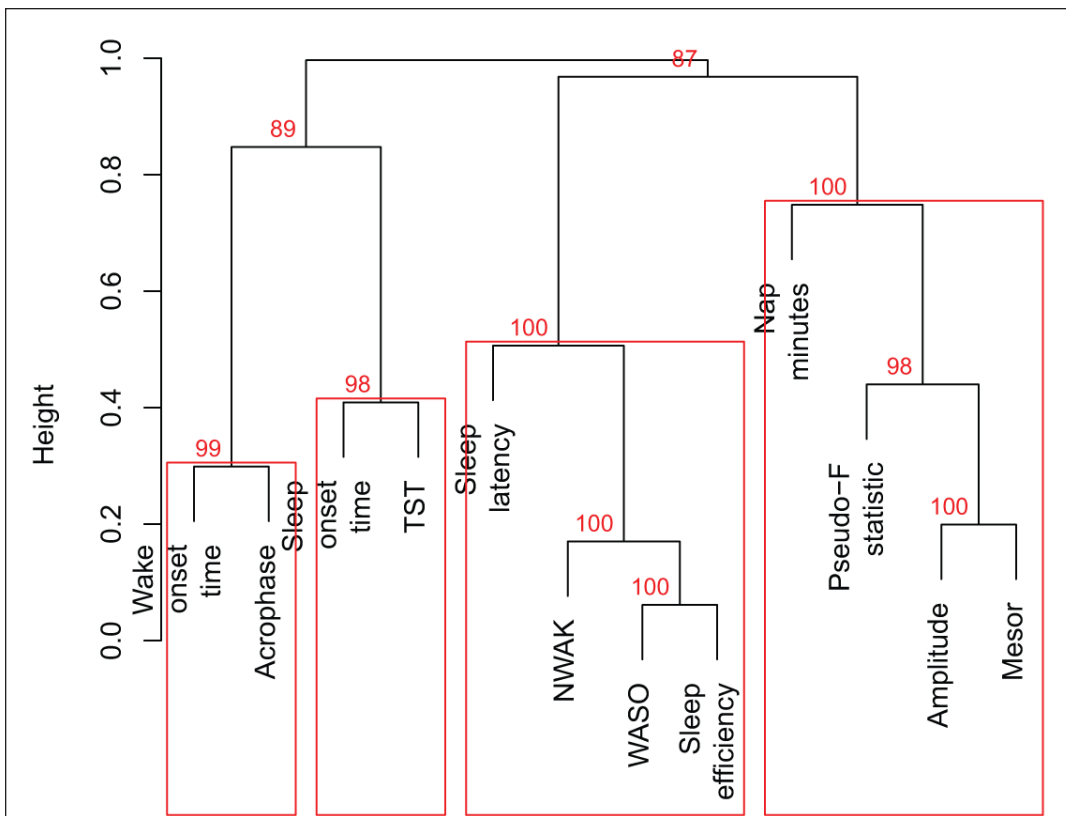
Figure S4—Hierarchical cluster dendrogram of genetic association results. Multiscale bootstrap resampling-based P values shown in red above edges. For clusters with P > 95, the hypothesis that the cluster does not exist is rejected with significance level 0.05. Red boxes mark clusters with multiscale bootstrap resampling- based P > 95. Height on the y-axis indicates the complete clustering method values. NWAK, number of awakenings; TST, total sleep time; WASO, wake after sleep onset.

SUPPLEMENTAL REFERENCE

1. Gabriel SB, Schaffner SF, Nguyen H, et al. The structure of haplotype blocks in the human genome. Science 2002;296:2225-9.