Identification of novel risk loci for restless legs syndrome in genome-wide association studies in individuals of European ancestry

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Identification of novel risk loci for restless legs syndrome in genome-wide association studies in individuals of European ancestry: a meta-analysis


Summary

Background Restless legs syndrome is a prevalent chronic neurological disorder with potentially severe mental and physical health consequences. Clearer understanding of the underlying pathophysiology is needed to improve treatment options. We did a meta-analysis of genome-wide association studies (GWASs) to identify potential molecular targets.

Methods In the discovery stage, we combined three GWAS datasets (EU-RLS GENE, INTERVAL, and 23andMe) with diagnostic data collected from 2003 to 2017, in face-to-face interviews or via questionnaires, and involving 15 126 cases and 95 725 controls of European ancestry. We identified common variants by fixed-effect inverse-variance meta-analysis. Significant genome-wide signals (p≤5x10⁻⁸) were tested for replication in an independent GWAS of 30 770 cases and 286 913 controls, followed by a joint analysis of the discovery and replication stages. We did gene annotation, pathway, and gene-set-enrichment analyses and studied the genetic correlations between restless legs syndrome and traits of interest.

Findings We identified and replicated 13 new risk loci for restless legs syndrome and confirmed the previously identified six risk loci. MEIS1 was confirmed as the strongest genetic risk factor for restless legs syndrome (odds ratio 1·92, 95% CI 1·85–1·99). Gene prioritisation, enrichment, and genetic correlation analyses showed that identified pathways were related to neurodevelopment and highlighted genes linked to axon guidance (associated with SEMAGD), synapse formation (NTNG1), and neuronal specification (HOXB cluster family and MYTI).

Interpretation Identification of new candidate genes and associated pathways will inform future functional research. Advances in understanding of the molecular mechanisms that underlie restless legs syndrome could lead to new treatment options. We focused on common variants; thus, additional studies are needed to dissect the roles of rare and structural variations.


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Introduction

Despite the prevalence of restless legs syndrome being up to 10% in populations of European ancestry, its genetic basis and underlying pathophysiology remain unclear. The restless legs syndrome phenotype is an unusual composite of sensory and motor symptoms that present with distinct circadian rhythmicity. The symptoms worsen or are only present in the evening or at night and markedly lessen in the early morning. People feel an overwhelming urge to move, often in conjunction with unpleasant sensations, usually in the legs. Rest and inactivity provoke the symptoms, whereas movement and other external stimuli lead to temporary relief. Due to the chronic progressive nature of the disorder, it has long-lasting effects on patients’ mental and physical health. People with restless legs syndrome have substantially impaired sleep, reduced overall quality of life, and increased risk of depression, anxiety disorders, hypertension, and, possibly, cardiovascular disease. Around 2–3% of the general population have severe restless legs syndrome, and most need chronic treatment with dopaminergics, α₁ ligands, or even opioids. However, long-term use of dopaminergics can lead to severe side-effects, including the worsening of symptoms

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Research in context

Evidence before this study

We searched PubMed for articles published up to July, 2017, with combinations of the search term “restless legs AND (genomewide OR genome-wide OR GWAS)”, without restrictions on language of publication. This search yielded 42 original articles, including reports on genetic linkage analyses and genome-wide association studies (GWASs). Although linkage results on restless legs syndrome have not been reproducible, GWASs so far have revealed six risk loci, with the strongest signal being in MEIS1, a member of the three aminocacid loop extension homeobox gene class. Subsequent studies have suggested altered embryonic development of the striatum is also important in the aetiology of restless legs syndrome. Nevertheless, the pathogenesis needs further elucidation. Changes in dopaminergic signalling and brain iron deficiency seem to be involved and are targeted by approved drugs. Other suggested causes include peripheral hypoxia and unknown metabolic factors related to uraemia and pregnancy. The current recommended treatment for restless legs syndrome, although effective, can lead to serious adverse effects, including worsening of symptoms, thus alternatives are needed.

Added value of this study

We did a meta-analysis of three GWAS datasets, which yielded a total sample size more than one order of magnitude larger than any previously published restless legs syndrome GWAS. We discovered 13 new risk loci for restless legs syndrome, taking the total from six to 19. Assessment of these candidate genes will enable more granular dissection of the pathogenesis of restless legs syndrome, which could improve determination of shared genetic architecture with other neurological phenotypes and the prospects of developing novel and more effective treatment options. Our results strongly support the link to neurogenesis, changes in neuronal circuit formation, synaptogenesis, and axonal guidance, thereby strengthening the concept of restless legs syndrome as a neurodevelopmental disorder. The novel risk loci include the genes CRBN, which encodes cerebulin, and MEIS2, which encodes its physiological substrate homeobox protein Meis2. As thalidomide targets this interaction, this drug could be a candidate for the treatment of restless legs syndrome in patients beyond reproductive age.

Implications of all the available evidence

The genes and associated pathways identified provide a much-needed basis for future investigation of restless legs syndrome, informing which pathophysiological and pharmacological concepts to examine in laboratory and clinical trials. Our findings further suggest investigating the role of neurodevelopmental processes in restless legs syndrome and the mechanism of the interaction between cerebulin and homeobox protein Meis2 in the context of assessing the repurposing of thalidomide.

Methods

Study populations and phenotype definitions

For the discovery meta-analysis, we combined three GWAS datasets. The EU-RLS-GENE consortium GWAS includes cases and population-matched controls recruited in eight European countries, Canada, and the USA. People with restless legs syndrome were recruited in specialist outpatient clinics for movement disorders and in sleep units. Restless legs syndrome was diagnosed in face-to-face interviews by an expert neurologist, based on the International Restless Legs Syndrome Study Group diagnostic criteria.1,2 A subset of the samples had been used in previous GWASs for restless legs syndrome (appendix p 10).4,5

The INTERVAL study11,12 included whole-blood donors recruited in England, enrolled between 2012 and 2014. The validated Cambridge-Hopkins Restless Legs Questionnaire6 was used to identify patients with restless legs syndrome to exclude mimicking disorders. Probable and definite cases were combined in one group.

The 23andMe GWAS dataset comprised samples drawn from participants of the customer base of 23andMe (Mountain View, CA, USA), which is a genetic testing company. The restless legs syndrome phenotype was determined with one research question, “Have you ever been diagnosed with restless legs syndrome?”, which had three response options: yes, no, and not sure. We include respondents who answered “yes” as cases and those who answered “no” as controls, and we excluded those who answered “not sure”.

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For the replication study, we used the 23andMe customer base to identify a distinct population. We used the same selection methods as for the discovery population.

Participants of all the original studies had provided informed consent, and the study protocols had been approved by the respective ethics committees. Demographics of the study samples are provided in the appendix (p 10).

Genotyping, quality control, imputation, and statistical analysis

Genotyping, quality-control procedures, imputation, and statistical analysis make independent inheritance signals in the appendage study. By adapting an additive model with logistic regression on genotype dosage, adjusted for age, sex, and components from either multivariable scaling or principal components analysis to correct for population stratification.

Meta-analysis procedures

For the discovery stage, summary statistics from each individual GWAS dataset were subjected to further quality control with EasyQC, version 8.5, and we did a fixed-effect inverse-variance meta-analysis with METAL, release 2011-03-25 (appendix p 5). To address heterogeneity between studies in the identified association signals, we did a random-effects meta-analysis with METASOFT, version 2.0.1 (appendix p 5). Genomic control was done in each study separately before meta-analysis by calculating the inflation factor λ and adjusting for it.16

To define independent genome-wide significant signals (p<5×10⁻⁸) in the discovery meta-analysis results, we used a two-step procedure (appendix p 5). Briefly, we first assigned variants to clusters with the “clump” command in PLINK software (version 1.90b3.36) based on the association p value and short-range (500 kb) linkage disequilibrium (LD), with each clump defined by one lead single-nucleotide polymorphism (SNP). Second, we tested for statistical independence of the lead SNPs with the stepwise model selection procedure implemented in GCTA, version 1.25.3, taking into account long-range LD (10 Mb) between lead SNPs (appendix p 5). Finally, we did standard conditional analysis in the remaining independent SNP clusters, adjusted for the top-associated lead SNP, to identify secondary independent signals within the clump.

For the joint analysis of the discovery and replication stages, summary statistics for independent association signals were subjected to fixed-effect inverse-variance and random-effects meta-analysis, as in the discovery stage (appendix p 5).

Heritability, partitioned heritability, and genetic correlation analysis

Heritability and partitioned heritability were estimated by LD-score regression (with LDSC, version 1.0.0) and using the summary statistics of the discovery meta-analysis without genomic control correction. Partitioned heritability analysis used publicly available partitioned LD scores for 52 functional categories based on the phase 3 dataset of the 1000 Genomes Project, as precomputed by the developers of LDSC. For the genetic correlation analysis, we used the LD Hub database (version 1.2.2), which provides access to summary-level GWAS statistics of more than 200 traits. The analytical procedures and settings are detailed in the appendix (pp 5–6). To estimate the variance explained by genome-wide significant association signals, we calculated Nagelkerke’s pseudo-R² in the EU-RLS-GENE dataset with tenfold cross validation.

Genetic risk score analysis

Genetic risk profiles were generated by estimating weighted polygenic risk scores in the EU-RLS-GENE dataset. The weights were based on the effect size estimates of the discovery meta-analysis (fixed-effect model). Polygenic risk scores were calculated with PRSice software, version 1.25 (appendix p 6).

Biological interpretation of association signals

As a first step, we did a literature-based annotation of protein-coding genes located in the genome-wide significant risk loci, defined based on LD, and in their vicinity (appendix p 6). To identify the biological and cellular pathways underlying the association signals, we applied bioinformatic methods for gene prioritisation and did enrichment analyses for pathways and tissues. In restless legs syndrome, the overall number of risk loci is small, restricting the use of standard tools, such as DEPICT.17 Moreover, the enrichment analysis could be hampered by the clinical heterogeneity of restless legs syndrome, its complex phenotype with somatosensory and motor symptoms, and pleiotropic effects of the candidate genes. Therefore, to enable efficient exploration of our data, we developed a new algorithm, called BI-ENRICH, which builds on the concept of biclustering used in gene-expression analysis (appendix pp 6–8). Additionally, we used the DEPICT software (version rel19413) to do gene prioritisation for each locus and to search for gene set and tissue enrichment among these genes (appendix p 8).

Data sharing

BI-ENRICH code is available on GitHub. GWAS summary statistics will be made available for researchers (appendix p 3).

Role of the funding source

The funders of the study had no role in study design, data collection, data analysis, data interpretation, or writing of the report. The corresponding author had full access to all the data in the study and had final responsibility for the decision to submit for publication.
Results

The discovery stage association study was done in 15126 cases and 95725 controls: 6228 and 10992 from EU-RLS-GENE; 3065 and 24923 from INTERVAL; and 5833 and 59810 from 23andMe. After quality control, 6864281 SNPs and indels with minor allele frequency 1% or greater were available for statistical analysis. We identified 20 independent association signals meeting genome-wide significance (figure 1, table 1, appendix p 13). No secondary independent association signals with genome-wide significance were found in these 20 regions by standard conditional analysis. We assigned the 20 independent association signals to 19 independent genomic risk loci (two mapped to the same gene, PTPRD, and are <500 kb apart; appendix p 65). Lead SNPs rs1836229 and rs62535767 of these two signals were not correlated \((r^2=0.0002, \text{EU-RLS-GENE dataset})\). Genomic control showed negligible inflation of the median test statistic \((\lambda_{1000}=1.004, \text{rescaled to adjust for the large sample})\). Lead SNPs rs113851554, located in a putative regulatory element in intron eight of MEIS1, is a low-frequency variant with odds ratio (OR) estimates of 1.82–2.16, which clearly distinguish it from the other risk loci for restless legs syndrome. The amount of heritability of restless legs syndrome attributable to all SNPs available in our dataset in the LD-score regression was 19–6%. Focusing on the 20 independent association signals, these explained 11–7% of the observed variance (60% of the SNP heritability). In the assessment of distribution of genetic risk, calculated from polygenic risk scores based on the 20 association signals (appendix p 6), individuals in the highest risk group (polygenic risk score >4.4, 95% CI 3.8–5–42.3, \(p=6.9\times10^{-26}\)) as this group represents only a very small subset of the population, we also compared the genetic

[Figure 1: Manhattan plot showing results of the discovery meta-analysis]

Due to their close proximity, two independent risk loci on chromosome 2 (both previously known) and on chromosome 6 (one previously known, one new) appear as single peaks. Thus, 19 loci (six previously known and 13 new) are represented. SNP=single-nucleotide polymorphism.
All p values were obtained by fixed-effect inverse-variance meta-analyses. The threshold for genome-wide significance was $p \leq 5 \times 10^{-8}$.

### Table 1: Association results for lead single-nucleotide polymorphisms reaching genome-wide significance in the discovery meta-analysis

<table>
<thead>
<tr>
<th>Chromosome</th>
<th>Position (bp)</th>
<th>Effect allele</th>
<th>Other allele</th>
<th>Effect-allele frequency</th>
<th>Protein-coding gene context</th>
<th>Discovery stage meta-analysis</th>
<th>Replication stage</th>
<th>Joint stage meta-analysis</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>T</td>
<td>C</td>
<td>0.59</td>
<td>PRMT6*†,**, NTNG1†‡</td>
<td>3.22 x 10^{-7}</td>
<td>2.03 x 10^{-7}</td>
<td>3.25 x 10^{-7}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>G</td>
<td>A</td>
<td>0.36</td>
<td>DCDC2C†‡</td>
<td>3.78 x 10^{-5}</td>
<td>7.74 x 10^{-5}</td>
<td>1.41 x 10^{-5}</td>
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<tr>
<td></td>
<td></td>
<td>T</td>
<td>G</td>
<td>0.07</td>
<td>MEIS1*†,**</td>
<td>1.1 x 10^{-6}</td>
<td>4.8 x 10^{-7}</td>
<td>2.00 x 10^{-6}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>A</td>
<td>C</td>
<td>0.53</td>
<td>MEIS1*, C1D,*</td>
<td>1.23 x 10^{-5}</td>
<td>1.98 x 10^{-5}</td>
<td>1.39 x 10^{-5}</td>
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<tr>
<td></td>
<td></td>
<td>G</td>
<td>A</td>
<td>0.33</td>
<td>ATP2C1*‡, AST21†‡</td>
<td>3.18 x 10^{-4}</td>
<td>1.4 x 10^{-5}</td>
<td>2.55 x 10^{-5}</td>
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<tr>
<td></td>
<td></td>
<td>T</td>
<td>A</td>
<td>0.26</td>
<td>CNTN4*, CRBN*, LRRN1‡</td>
<td>5.38 x 10^{-4}</td>
<td>1.93 x 10^{-7}</td>
<td>2.01 x 10^{-7}</td>
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<td></td>
<td>A</td>
<td>T</td>
<td>0.09</td>
<td>AKT2B1‡, MR111†‡</td>
<td>4.37 x 10^{-3}</td>
<td>5.9 x 10^{-9}</td>
<td>3.96 x 10^{-9}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>A</td>
<td>C</td>
<td>0.23</td>
<td>RNF111*, CDC61C*, MDGA1†‡</td>
<td>6.43 x 10^{-3}</td>
<td>7.61 x 10^{-6}</td>
<td>2.55 x 10^{-5}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>C</td>
<td>G</td>
<td>0.36</td>
<td>BRD9*, GLO2†‡</td>
<td>1.36 x 10^{-4}</td>
<td>1.22 x 10^{-5}</td>
<td>3.58 x 10^{-5}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>G</td>
<td>A</td>
<td>0.13</td>
<td>ADAM23, STEAP4*, ZNF804B†‡</td>
<td>1.86 x 10^{-5}</td>
<td>5.01 x 10^{-7}</td>
<td>1.73 x 10^{-7}</td>
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<td></td>
<td></td>
<td>G</td>
<td>A</td>
<td>0.48</td>
<td>PTPRD†‡</td>
<td>1.94 x 10^{-5}</td>
<td>1.12 x 10^{-7}</td>
<td>1.13 x 10^{-7}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>T</td>
<td>C</td>
<td>0.32</td>
<td>PTPRD†‡</td>
<td>3.13 x 10^{-5}</td>
<td>8.77 x 10^{-7}</td>
<td>3.23 x 10^{-5}</td>
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<td>T</td>
<td>G</td>
<td>0.20</td>
<td>DACH1†‡, DCL1†‡</td>
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<td>3.9 x 10^{-6}</td>
<td>3.23 x 10^{-5}</td>
</tr>
<tr>
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<td></td>
<td>T</td>
<td>A</td>
<td>0.06</td>
<td>DPH6*, MEIS2†‡</td>
<td>2.96 x 10^{-5}</td>
<td>5.45 x 10^{-5}</td>
<td>3.39 x 10^{-5}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>G</td>
<td>T</td>
<td>0.10</td>
<td>SEMA6D††</td>
<td>1.05 x 10^{-5}</td>
<td>3.81 x 10^{-7}</td>
<td>2.69 x 10^{-6}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>A</td>
<td>C</td>
<td>0.32</td>
<td>Sema3A†, MAP2K5*, SKOR1*, C1N6†</td>
<td>1.09 x 10^{-6}</td>
<td>9.23 x 10^{-6}</td>
<td>5.48 x 10^{-6}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>G</td>
<td>C</td>
<td>0.42</td>
<td>TOX3C‡†‡</td>
<td>4.72 x 10^{-6}</td>
<td>4.36 x 10^{-6}</td>
<td>7.27 x 10^{-6}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>A</td>
<td>G</td>
<td>0.21</td>
<td>Hoxb8 cluster†‡, PRAC1†‡</td>
<td>4.87 x 10^{-6}</td>
<td>2.01 x 10^{-6}</td>
<td>4.27 x 10^{-6}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>T</td>
<td>C</td>
<td>0.25</td>
<td>SETBP1†‡</td>
<td>1.37 x 10^{-6}</td>
<td>6.59 x 10^{-6}</td>
<td>1.11 x 10^{-6}</td>
</tr>
<tr>
<td></td>
<td></td>
<td>G</td>
<td>A</td>
<td>0.27</td>
<td>MYST1††</td>
<td>3.36 x 10^{-6}</td>
<td>7.83 x 10^{-6}</td>
<td>1.73 x 10^{-6}</td>
</tr>
</tbody>
</table>

All p values were obtained by fixed-effect inverse-variance meta-analyses. The threshold for genome-wide significance was $p \leq 5 \times 10^{-8}$. For each locus, only the selected protein-coding genes are listed by genomic position (direction 5' to 3' on chromosome). GWAS = genome-wide association study. Position = GRCh37/hg19 coordinates. bp = base pair. *Selected by nearest gene. †Selected by Bi-ENRICH prioritisation (nominal p < 0.05). ‡Selected by manual annotation. §Loci already discovered in previous GWASs. ¶Gene reported in previous GWASs.
Articles

linked to the risk loci and regional association plots for all loci are provided in the appendix (pp 16–19, pp 44–86).

These observations were substantiated by bioinformatic enrichment and gene prioritisation analyses. The BI-ENRICH algorithm showed that most of the top pathways identified for restless legs syndrome were related to neurodevelopment, including neurogenesis (genes MDGA1, MYT1, NTNG1, and SEMA6D), cell-junction organisation (PKP4 and SMAD3), and axon guidance (NTNG1 and SEMA6D). Moreover, locomotor behaviour (BTBD9, CLN6, HOXB8, and MEI5) was highlighted (figure 2, appendix pp 20–22). Pathways related to DNA repair and maintenance (APLF, ASTE1, DIS3, PRMT6, and RNF8) were also detected by the enrichment analysis. Consistent with these results, BI-ENRICH prioritised genes related to neurodevelopment and DNA-damage repair (appendix pp 23–30). Use of the standard version of DEPICT returned no significant results (appendix pp 31–32). However, including the UniProtKB biological process annotations in the gene-set definitions, genes prioritised by DEPICT were enriched for neurogenesis (false discovery rate <0.05, appendix p 33). Finally, we did stratified LD-score regression to partition the heritability carried by all SNPs to specific functional categories. 14–8% of the SNPs that reside in regions associated with histone marks in the CNS explain 44.7% of the variance.

This CNS enrichment is significant (enrichment factor 44.7/14.8 = 3.0, p = 0.0043). Consistent with this observation, tissue enrichment analyses with BI-ENRICH and DEPICT ranked brain and spinal cord tissues at the top, but false discovery rates were greater than 0.2, and only spinal-cord tissue (BI-ENRICH) and mesencephalon (DEPICT) were significant (appendix pp 34–37).

Finally, we used LD-score regression to assess the genetic correlation between restless legs syndrome and diseases and traits of interest (figure 3, appendix p 38).

Table 2: Candidate genes linked to functions in neurodevelopment

<table>
<thead>
<tr>
<th>Genes</th>
<th>Functions related to neurodevelopment</th>
</tr>
</thead>
<tbody>
<tr>
<td>rs12046503</td>
<td>NTNG1</td>
</tr>
<tr>
<td>rs10208712</td>
<td>DCDC2C</td>
</tr>
<tr>
<td>rs11385154</td>
<td>MEI5</td>
</tr>
<tr>
<td>rs8031944</td>
<td>PKP4</td>
</tr>
<tr>
<td>rs1848460</td>
<td>CRBN, CNTN4</td>
</tr>
<tr>
<td>rs17636328</td>
<td>MDGA1</td>
</tr>
<tr>
<td>rs10952927</td>
<td>ZNF804B, ADAM22</td>
</tr>
<tr>
<td>rs1836229</td>
<td>PTPRD</td>
</tr>
<tr>
<td>rs6255767</td>
<td>DACH1</td>
</tr>
<tr>
<td>rs340651</td>
<td>DACH1</td>
</tr>
<tr>
<td>rs996064</td>
<td>MEI52</td>
</tr>
<tr>
<td>rs111652004</td>
<td>SEMA6D</td>
</tr>
<tr>
<td>rs45544231</td>
<td>TOX3</td>
</tr>
<tr>
<td>rs12450895</td>
<td>HOXB cluster family</td>
</tr>
<tr>
<td>rs365031</td>
<td>MYT1</td>
</tr>
</tbody>
</table>

For the 1000 Genomes Project see http://www.internationalgenome.org
For GitHub see https://github.com/Chen-Zhao/BI-ENRICH

Figure 2: Representation of significantly enriched functional gene sets found by the BI-ENRICH analysis

Similarities between gene sets are measured with the Jaccard index, with low being J ≤ 0.3, medium being J > 0.3 to < 0.5, and high being J > 0.5. Empirical p values are shown.
Restless legs syndrome showed some positive correlations with neuropsychiatric diseases, whereas correlations were mainly negative with neurodegenerative disorders (figure 3), although none reached nominal significance. Few supporting data are available from corresponding epidemiological studies. Two small studies showed increased prevalence of restless legs syndrome among people with amyotrophic lateral sclerosis. One study showed prevalence of 4% among people with Alzheimer’s disease, which is lower than the average 6–12% in elderly people of European descent, but making a diagnosis of restless legs syndrome in cognitively impaired individuals is challenging and the findings are difficult to interpret. Data on restless legs syndrome and Parkinson’s disease were inconclusive, showing wide-ranging prevalence (0–50%), with some studies proposing restless legs syndrome as a risk factor for Parkinson’s disease and others showing protective effects. Moreover, dopaminergic treatment of Parkinson’s disease has been suggested to precipitate restless legs syndrome in some patients. The risk locus on chromosome 15q23 (encompassing MAP2K5 and SKOR1) overlaps with GWAS signals of four different traits, including the posterior cortical atrophy variant of Alzheimer’s disease (appendix pp 39–41). The ORs of the shared SNP rs11637445 were 1.48 for posterior cortical atrophy and 0.89 for restless legs syndrome, which is consistent with the negative correlation between these two disorders. We found a significant positive correlation between restless legs syndrome and depressive symptom phenotypes and with neuroticism (figure 3, appendix p 38). These genetic correlations are in line with epidemiological associations, which show increased prevalence of these phenotypes in patients with restless legs syndrome. Overall, the correlations between restless legs syndrome and neurological or psychiatric phenotypes were low to moderate. We found a positive genetic correlation with number of children ($r=0.22$, $p=0.0079$). Pregnancy is a validated risk factor for restless legs syndrome, and parity increases the risk of developing restless legs syndrome in later life. Sleep-related phenotypes and iron-related traits showed correlations in line with published data, but none reached nominal significance. Finally, negative genetic correlations were generally seen between restless legs syndrome and measures of educational attainment (figure 3), although this association does not seem to be widely reported.

**Discussion**

This association study identified 20 independent association signals that reached genome-wide significance in 19 risk loci, of which 13 had not been previously reported. A subset of these signals showed between-study heterogeneity in our meta-analyses; in a random-effects meta-analysis that accounted for heterogeneity, all reached genome-wide significance. Effect-size estimates and strength of the association signal in these loci varied between studies, probably because of differences in phenotyping methods (clinical face-to-face interview, self-reporting by participants via a validated questionnaire, and self-reporting by use of one question). These differences can lead to variation in the proportions of misclassified
cases and controls and might have modified some of the association signals. Moreover, the specific features of the restless legs syndrome phenotype, such as disease severity or preponderance of motor or sensory symptoms, might have varied between the studies. Detailed clinical data were scarce, and the restless legs syndrome phenotype in the INTERVAL and 23andMe datasets were not clinically validated and, therefore, we cannot further address the underlying causes of heterogeneity. Studies including detailed assessments of restless legs syndrome symptoms are needed to dissect the specific roles of and possible interactions between genetic variants in the risk loci and the resulting phenotypes.

Together, the 20 association signals accounted for 60% of the SNP-based heritability of 19.6% estimated in this study. Our dataset was limited to SNPs and small indels with minor allele frequency of 1% or greater. Therefore, we could not assess the contribution of rarer or structural variants, which are likely to have larger effect sizes and might explain another part of the heritability of restless legs syndrome. Future large-scale whole-genome sequencing efforts might have the power to detect such variants.

The main aim of our study was to provide new clues to understanding the biology of restless legs syndrome. Candidate genes and pathway analyses across the 19 risk loci converge on functions important in the development of the CNS, such as neurogenesis and neural-circuit formation, including axon guidance and synaptogenesis. Additionally, the BI-ENRICH analysis highlighted DNA-damage repair, which is important for development and maintenance of the nervous system, as having a relevant role. The involvement of perturbations in neurodevelopmental processes is in line with previous functional studies of MEIS1 that identified the embryonic ganglionic eminences as relevant structures of restless legs syndrome biology. Our pathway analysis was designed to have high sensitivity to avoid downward bias, such as false-negative results due to incomplete information provided by the annotation databases. To avoid upward bias (ie, false-positive enrichment results), we corrected by sampling and null-GWAS permutation (phenotype label permutation).

Significance of results in the tissue enrichment analyses might have been hindered by a lack of appropriate samples from relevant anatomical regions or developmental stages in the input datasets. Nonetheless, neuronal tissues were consistently prioritised, and nominally significant regions, such as midbrain structures and spinal cord, have previously been implicated in restless legs syndrome.

At present, major concepts in the pathophysiology of restless legs syndrome address alterations in the dopaminergic neurotransmitter system and in brain iron metabolism, and are supported by evidence from animal, imaging, and human post-mortem studies. Even though our enrichment analyses did not specifically highlight corresponding biological processes, the new risk loci offer potential bases for functional studies that might shed light on mechanisms underlying the suspected changes of the dopaminergic neurotransmitter system or brain iron metabolism. Moderate, albeit non-significant, positive genetic correlations are being reported with neuropsychiatric disorders, such as schizophrenia or bipolar disorder, for which evidence of contributing pathological events that affect early neurodevelopment and the correct setup of neuronal circuitry is growing.

Finally, annotation of the risk loci identified a gene with a product targeted by an existing drug that is readily available for repurposing. The ubiquitin ligase substrate receptor, cereblon, which is encoded by CRBN (locus on chromosome 3p26), is a target of the drug thalidomide, and homeobox protein Meis2, which is encoded by MEIS2 (locus on chromosome 15q14), has been identified as an endogenous substrate of cereblon. Thalidomide was initially licensed as a hypnotic, but was withdrawn from the market because of teratogenicity. Thalidomide and its analogues are now used as immunomodulatory drugs in cancer. They block binding of homeobox protein Meis2 to the ubiquitin ligase, thereby modulating its activity. Thalidomide and thalidomide-like substances, therefore, could be potential candidates for therapeutic use in restless legs syndrome, including in women beyond reproductive age. Moreover, this finding might also provide clues to the sleep-promoting mechanisms of this drug. Carefully designed clinical trials are needed to further investigate these drugs.

With the 13 new risk loci identified for restless legs syndrome, the total becomes 19. Our results suggest that altered embryonic neurodevelopment, impaired neurogenesis at later age, or both, could underlie restless legs syndrome pathogenesis. The exact molecular mechanisms, the relevant times in life, and the connection to brain iron metabolism and the dopaminergic system are unknown, but the implicated loci, candidate genes, and candidate pathways will provide impetus for further functional research in restless legs syndrome. Finally, the possibility of repurposing thalidomide could lead to translation of these research findings into the care of patients.

Contributors BS and JW designed and oversaw the study. BS, CZ, AVS, KO, BM-M, and JW did primary interpretation of the data. BS, CZ, KO, and JW wrote the paper, and SB and DAH contributed substantial edits. BS did the EU-RLS-GENE GWAS. CZ did the meta-analyses. ET and BP participated in quality control and statistical analysis of the EU-RLS-GENE GWAS. BM-M supervised statistical analyses for EU-RLS-GENE GWAS and meta-analyses. YD, AS, BH, WPo, DK, KS, CGB, WPa, CT, WHOe, MH, MT-L, AM, GMH, OP, IF, OAR, ZW, WGO, KB, and JW recruited patients for the EU-RLS-GENE GWAS. RPA, CJE, LX, JM, and GAR recruited patients and controls for the EU-RLS-GENE GWAS. MP, PV, CD, AF, LT, AFRS, SHS, CG, and AP provided control samples or genotype data for the EU-RLS-GENE GWAS. JW acquired funding for the EU-RLS-GENE GWAS and all meta-analyses. SB did statistical analysis of the INTERVAL GWAS for restless legs syndrome. ASB and NS were responsible for data curation of the INTERVAL study. ASB, NS, WHOn, DJR, JD, and EDA acquired funding for and supervised the INTERVAL study. EDA conceived and oversaw the INTERVAL GWAS for restless legs syndrome. DAH performed data analysis of the 23andMe restless legs syndrome GWAS. All authors contributed and critically reviewed the final version of the manuscript.
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References
Articles


