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A meta-analysis of genome-wide association studies identifies 17 new Parkinson's disease risk loci

Diana Chang1, **Mike A Nalls**2,3, **Ingileif B Hallgrímsdóttir**4,6, **Julie Hunkapiller**1, **Marcel van der Brug**1,6, **Fang Cai**1, **International Parkinson's Disease Genomics Consortium**5, **23andMe ResearchTeam**5, **Geoffrey A Kerchner**1, **Gai Ayalon**1, **Baris Bingol**1, **Morgan Sheng**1, **David Hinds**4, **Timothy W Behrens**1, **Andrew B Singleton**2, **Tushar R Bhangale**1,7, and **Robert R Graham**1,7,iD

¹Genentech, Inc., South San Francisco, California, USA

²Laboratory of Neurogenetics, National Institute on Aging, US National Institutes of Health, Bethesda, Maryland, USA

³Data Tecnica International, Glen Echo, Maryland, USA

⁴23andMe Inc., Mountain View, California, USA

Abstract

Common variant genome-wide association studies (GWASs) have, to date, identified >24 risk loci for Parkinson's disease (PD). To discover additional loci, we carried out a GWAS comparing 6,476 PD cases with 302,042 controls, followed by a meta-analysis with a recent study of over 13,000 PD cases and 95,000 controls at 9,830 overlapping variants. We then tested 35 loci ($P < 1 \times$ 10−6) in a replication cohort of 5,851 cases and 5,866 controls. We identified 17 novel risk loci (^P $< 5 \times 10^{-8}$) in a joint analysis of 26,035 cases and 403,190 controls. We used a neurocentric strategy to assign candidate risk genes to the loci. We identified protein-altering or *cis*–expression quantitative trait locus (*cis*-eQTL) variants in linkage disequilibrium with the index variant in 29 of the 41 PD loci. These results indicate a key role for autophagy and lysosomal biology in PD risk, and suggest potential new drug targets for PD.

AUTHOR CONTRIBUTIONS

COMPETING FINANCIAL INTERESTS

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Correspondence should be addressed to R.R.G. (graham.robert@gene.com).

A list of members appears in Supplementary Note 1

⁶Present addresses: Amgen, South San Francisco, California, USA (I.B.H.); E-Scape Bio, South San Francisco, California, USA (M.v.d.B.). 7These authors contributed equally to this work

Robert R Graham D <http://orcid.org/0000-0001-7151-4277>

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PD is the second most common neurodegenerative disorder^{1,2}, with a prevalence of $3-4\%$ in individuals over 80 years of age³. PD is characterized by the loss of dopaminergic neurons in the substantia nigra and the presence of Lewy bodies^{1,2}. These neuropathologies manifest in affected individuals primarily as motor-related symptoms, but the involvement of other brain regions can lead to nonmotor symptoms⁴.

Early-onset, familial PD (onset at ≤ 60 years of age) accounts for a small fraction of cases⁵, but the identified associated genes, including LRRK2, GBA, and SNCA, provide insight into disease pathogenesis^{6,7}. For the later-onset, common form of PD, at least 24 loci have been associated at a genome-wide significant level with disease risk in individuals of European ancestry⁸. The narrow-sense heritability (h^2) explained by the confirmed PD risk loci is low $(0.033)^9$; however, the heritability explained by common variants is estimated at 0.227 (s.d.: 0.08 ⁹, which suggests that additional loci with smaller effect sizes remain to be discovered.

We carried out a GWAS of 6,476 subjects from a 23andMe PD cohort (PDWBS (Web-Based Study of Parkinson's Disease)) and 302,042 controls genotyped on custom Illumina arrays (Fig. 1). The 6,476 PD cases of European ancestry were independent from those previously reported⁸ but met the same inclusion criteria, except that carriers of the $LRRK2$ G2019S mutation were not removed^{8,10}. The 302,042 controls did not report having PD and were of similar ancestry as the cases. The data were imputed with Minimac2 using 1000 Genomes phase 1 haplotypes^{11,12}. Single-nucleotide polymorphisms (SNPs) with low imputation quality or that failed general quality control metrics were removed (Online Methods). After correcting for age, sex, and the top principal components (Online Methods), we observed minimal inflation for P values genome-wide ($\lambda_{gc} = 1.057$; $\lambda_{1000} = 1.004$; Supplementary Fig. 1).

A total of 12 loci had $P < 5 \times 10^{-8}$ in the PDWBS analysis, including 11 of the loci that were reported in a previous GWAS in individuals of European ancestry⁸ (Table 1). For the remaining 13 previously reported loci, we observed $P < 0.05$ for 11 loci, with no significant evidence for association observed in the PDWBS sample for *CHMP2B* (rs115185635) or TMEM229B (rs155399). The remaining novel locus in the PDWBS analysis, rs9468199 (P $= 1.77 \times 10^{-9}$), is more than 4 Mb from the nearest PD association in the HLA class II region and is independent of rs9275326 ($P_{\text{conditional}} = 2.64 \times 10^{-9}$).

Using genome-wide summary statistics from the PDWBS analysis, we estimated the h^2 value for PD explained by common variants as 0.209 (95% confidence interval (CI): 0.148– 0.271, assuming a prevalence of 0.01), which is similar to the h^2 value reported previously^{9,10}. Regions contributing to PD heritability were significantly enriched for acetylation of histone H3 at lysine 27 ($P = 0.001$; Supplementary Table 1), a mark of active regulatory regions. PD heritability was also enriched for histone marks in central nervous system, adrenal, and pancreatic cell types (Supplementary Table 2), in agreement with a previous study¹³.

We next carried out a meta-analysis between the PDWBS GWAS and results for the top 10,000 variants available from a large-scale meta-analysis for PD with over 13,000 cases and 95,000 controls⁸ (PDGene) (Fig. 1). For the 9,830 overlapping SNPs between the PDWBS

and PDGene studies, we used an inverse-variance weighted method to combine association statistics for meta-analysis¹⁴. The odds ratios and P values for the 9,830 overlapping SNPs in the PDWBS and PDGene studies were correlated ($\rho_{-\log 10(P \text{ value})} = 0.85$, $\rho_{\text{OR}} = 0.58$). Furthermore, quantile–quantile (Q-Q) plots indicated an increase in the number of variants with low P values (Supplementary Fig. 1), even after the exclusion of variants in regions previously reported as associated with PD risk at a genome-wide significant level (Supplementary Table 3).

The meta-analysis identified 35 loci associated at $P < 1 \times 10^{-6}$, including 15 loci with $P < 5$ $\times 10^{-8}$ (Fig. 2, Supplementary Figs. 2 and 3, Supplementary Table 4). Only two of the previously reported loci (*BCKDK*:rs14235 and *MAPT*:rs17649553) and 2 of the 20 suggestive loci (FCGR2A.rs4657041, ITGA2B.rs5910) were in linkage disequilibrium (LD) $(r^2 > 0.8)$ with variants associated (at $P < 5 \times 10^{-8}$) with any phenotype in the NHGRI GWAS catalog15. Significant pleiotropy of PD risk loci with other complex diseases has not been identified¹⁶, but this pleiotropy landscape may change as more modest effects are uncovered.

We next sought validation of these 35 candidate loci in an independent cohort of 5,851 cases and 5,866 controls of European ancestry genotyped with the semi-customized NeuroX Illumina array 8.17 (Fig. 1). Twenty-nine of the 35 loci either were directly genotyped on the NeuroX array or had suitable proxies ($t^2 > 0.9$ with the original SNP; Supplementary Table 5). Weaker proxies at four additional SNPs ($t^2 > 0.5$) were available but were not used for validation in this study (Supplementary Table 5). In a replication-phase joint analysis of these 29 loci (meta-analysis of PDGene, PDWBS, and NeuroX), 16 had $P < 5 \times 10^{-8}$ (Table 2). Of these 16, all but 3 (rs4073221, rs10906923, and rs9468199) were also nominally associated in the NeuroX study (one-sided $P < 0.05$). A genetic risk score^{8,18,19} defined by these 16 loci, in addition to the previously reported loci, had a non-negligible ability to predict PD case status (area under the curve, 0.6518; 95% CI, 0.6419–0.6616). This represents a significant improvement over the predictive power of risk scores defined by previously reported loci alone ($P = 6 \times 10^{-8}$) (Supplementary Note 1). In sum, we identified 16 independent PD risk loci with a joint $P < 5 \times 10^{-8}$ and 1 locus (rs601999) with $P < 5 \times$ 10−8 in the discovery cohort with no suitable proxy for replication in the NeuroX cohort (Table 2).

Overall, 11 of 17 novel loci were in high LD ($r^2 > 0.8$) with at least one variant predicted to affect transcription factor binding (Supplementary Table 6). Of the 17 novel loci and 24 previously reported loci, 10 contained residual associations with $P < 1 \times 10^{-3}$ after conditioning on each region's most significant SNP in the PDWBS data (Supplementary Table 7). These regions included three of the four independent secondary signals reported by Nalls *et al.*⁸, as well as one variant previously reported at a non-genome-wide significant level $(P = 5.15 \times 10^{-7})^{20}$.

We note that the HLA region association with PD is particularly complex. Two candidate genes from the HLA region were nominated on the basis of support from either a proteincoding variant or an eQTL (Fig. 3). This is in line with a previous study that suggested that the PD association in the HLA region may point to multiple HLA factors, including

independent regulatory factors²¹. The association pattern observed at this locus may be reminiscent of the HLA association observed in schizophrenia and linked to C4 copy $number²²$.

The identification of the causal variants and genes underlying regions associated with common, complex disease is a major challenge²³. Several statistical methods have been proposed for the fine-mapping of causal variants^{23–25}. Alternatively, some studies have narrowed down lists of candidate genes by combining multiple levels of evidence with scoring-based strategies^{26,27}. Here we implemented a neurocentric strategy to nominate candidate genes for PD-associated loci.

We incorporated seven sources of data to annotate the index variant and linked variants from PD-associated loci (including eQTLs and expression data from $GTEx^{28}$, as well as expression data from brain cell types in mice²⁹; a full list is provided in the Online Methods). We used a two-stage approach to assign candidate genes to each locus (see the Online Methods for further details, and Supplementary Fig. 4 for a graphical visualization). In the first stage, we assigned a gene to a locus if (i) the index SNP or linked variants (t^2 > 0.6) altered the protein sequence or (ii) the index variant was a cis-eQTL for the gene. When no candidate genes were identified by the first stage, we ranked neighboring gene(s) on the basis of neurologically related phenotypes and expression and assigned the gene with the highest score to the locus (Online Methods).

With this strategy we identified a single candidate gene for 28 loci, and multiple candidate genes with similar levels of supporting evidence for 13 loci (Fig. 3, Supplementary Figs. 5 and 6). The candidate-gene nomination strategy confirmed several known PD risk genes, including *GBA*, *LRRK2*, *SNCA*, and *MAPT*. Among the 41 PD risk loci, a total of 29 loci (71%) had either a protein-altering or a cis-eQTL variant linked to the index SNP (Supplementary Tables 8 and 9). In addition, we carried out a colocalization analysis to determine whether the GWAS signal and the eQTL signal pointed to the same causal variant 30 (Supplementary Note 1). Seven candidate genes also had evidence for protein– protein interaction (Online Methods, Supplementary Table 10). Further studies are needed to experimentally determine the causal genes in the PD risk loci; however, the identification of candidate genes provides testable hypotheses for functional studies.

To gain insight into the biology, we tested the identified candidate genes in the 41 PD risk loci for association with any pathways or gene sets compared with a background gene list (Online Methods). We investigated whether candidate genes were enriched for pathways previously implicated in PD: autophagy, lysosomal, and mitochondrial biology¹. PDassociated signals were enriched (at a threshold of $P < 0.05/3 = 0.017$) for lysosomal and autophagy genes ($P = 3.35 \times 10^{-6}$ and $P = 5.71 \times 10^{-3}$, respectively). The addition of candidate genes more than doubled the number of lysosomal genes observed in PD loci and improved the enrichment significance ($P_{all_loci} = 3.35 \times 10^{-6}$, $P_{novel_loci} = 3.64 \times 10^{-5}$). We also observed that one previously identified gene (MCCC1) and two novel candidate genes (COQ7 and ALASI) mapped to the mitochondrial gene set (Supplementary Table 11).

Lysosomal biology and its role in the degradation of protein aggregates emerged as a highly significant pathway in PD risk. Among the five candidate genes linked to lysosomal biology, two were previously identified candidate genes (GBA (glucocerebrosidase) and TMEM175 (transmembrane protein 175)), and three were newly identified candidate genes (CTSB (cathepsin B), $ATP6V0A1$ (ATPase H⁺ transporting V0 subunit a1), and GALC (galactosylceramidase)). Glucocerebrosidase is required for normal lysosomal activity and α-synuclein degradation. In addition, GBA loss-of-function alleles are a common PD risk factor³¹. *TMEM175* was recently shown to encode a potassium channel that can regulate lysosomal function32, and the missense variant TMEM175 M393T is strongly linked to the index variant in the region (Supplementary Table 8). CTSB is a lysosomal cysteine protease. A PD risk allele is linked to a cis-eQTL for CTSB in multiple tissues (Supplementary Table 9), where the risk allele is associated with reduced levels of CTSB mRNA. Double-knockout mice for Ctsb and Ctsl (cathepsin L) show a tremor phenotype with cerebral and cerebellar atrophy³³. CTSB is also capable of degrading membrane-bound and soluble α -synuclein in mice³⁴.

Autophagy is the catabolic process that targets long-lived proteins and dysfunctional organelles for lysosomal degradation. Autophagy and lysosomal degradation have been implicated in PD by rare familial and common GWAS-associated GBA variants. We note that a strong cis-eQTL for lysine acetyltransferase 8 (KAT8) is associated with PD risk, with lower levels of KAT8 mRNA linked to increased PD risk. Inhibition of KAT8 was recently shown to decrease autophagic flux 35 .

Next, we used INRICH³⁶ to investigate whether PD-associated regions were enriched for gene sets in an unbiased fashion. Once again, we found significant enrichment of the lysosomal pathway ($P_{\text{adjusted}} = 0.02$) (Supplementary Table 12). We further examined the expression of the PD candidate genes in a brain-specific cell-type expression data set in mice²⁹; however, we observed broad expression across the major brain cell types, and no clear cell-type-specific pattern was evident (Supplementary Fig. 7).

Among the candidate genes newly identified in this study is *SH3GL2* (SH3 domaincontaining GRB2-like 2, endophilin A1), a gene recently demonstrated to be phosphorylated by LRRK2 and which may have a role in clathrin-mediated endocytosis of synaptic vesicles³⁷. Dysregulation of *Elovl7* (elongation of very long chain fatty acids protein 7) in mice results in several neurological phenotypes, including inflammatory astrocytosis and microgliosis in the brain, and neuronal degeneration³⁸. Upregulation of the candidate gene SCN3A (sodium voltage-gated channel α-subunit 3) enhances neuronal excitability and is associated with epilepsy in both humans and animal models 39 .

The new loci also encode three transcription factors: SATB1, ZNF184, and TOX3. TOX3 has been implicated in neuronal survival⁴⁰, and SATB1 has been associated with T cell function, particularly the development of regulatory T cells⁴¹.

Several of the PD candidate genes are within the 'druggable' genome⁴², including the previously identified serine/threonine kinase 39 (STK39) and the novel candidate gene inositol 1,4,5-trisphosphate kinase B (ITPKB). An in-frame deletion of ITPKB

(rs147889095) is linked to a PD-associated variant, and complete loss of ITPKB was reported in a patient with common-variable immunodeficiency⁴³. STK39 is a kinase linked to hypertension⁴⁴, regulation of K^+ levels, and the cellular stress response.

In summary, this study presents what to our knowledge is the largest meta-analysis of PD so far, involving a total of 26,035 cases and 403,190 controls. We identified 17 novel PD loci and, using a neurocentric candidate-gene nomination pipeline, found that several of the newly identified PD risk genes have a role in lysosomal biology and autophagy. The identification of these candidate genes allows for the prioritization of functional studies to determine causal genes for PD and possible therapeutic targets.

ONLINE METHODS

PDWBS GWAS

The PDWBS is a genome-wide analysis of 6,476 PD cases and 302,042 control subjects, all of whom were customers of 23andMe Inc. and consented to participate in research. The study protocol was approved by the external AAHRPP-accredited institutional review board, Ethical and Independent Review Services (E&I Review). Cases and controls were designated on the basis of surveys¹⁰. Controls were selected from 23andMe Inc. research participants who did not self-report as having been diagnosed with PD. Although the use of self-reported controls can result in a reduction of power, the effect of this on the current study was probably minimal (Supplementary Note 1). Any samples present in the PDGene study⁸ were removed from the PDWBS analysis. The average age of cases and controls was 67.6 and 50.8 years, respectively. The study also included 147 cases (2.3%) and 554 controls $(0.18%)$ that were LRRK2 G2019S carriers. Removing LRRK2 G2019S carriers from the analysis removed genome-wide significant associations at the LRRK2 locus.

DNA extraction and genotyping were performed on saliva samples by CLIA-certified CAPaccredited clinical laboratories of the Laboratory Corporation of America. Samples were genotyped on one of the following four platforms: V1 and V2, two variants of the Illumina HumanHap550+ BeadChip, with ~25,000 custom SNPs and ~950,000 total SNPs; V3, Illumina OmniExpress+BeadChip with custom SNPs to increase overlap with the V2 chip, with a total of ~950,000 SNPs; and V4, a custom chip that included SNPs overlapping V2 and V3 chips, low-frequency coding variants and ~570,000 SNPs. Samples with a call rate lower than 98.5% were reanalyzed, and research participants with samples that failed repeatedly were re-contacted and asked to provide additional samples.

Research participants were restricted to those of mainly $(>97%)$ European ancestry^{10,45}. All research participants in the study were also required to share <700 cM identity by descent (IBD) (estimated by a segmental IBD estimation algorithm⁴⁶), corresponding approximately to the sharing expected between first cousins. We additionally excluded individuals who shared >700 cM IBD with any 23andMe research participant whose data was used in the PDGene GWAS. Data were imputed on 1000 Genomes phase 1 haplotypes (September 2013 release) with Minimac2 on default settings^{11,47}. Imputation was run separately on data from each genotyping platform.

For genotyped SNPs, SNPs were removed if they were genotyped on only the V1 and/or the V2 chip, if they failed a parent–offspring transmission test on trio data, if they were not in Hardy–Weinberg equilibrium ($P < 10^{-20}$), or if they had a call rate < 0.90. For imputed SNPs, SNPs were removed if they had an average $r^2 < 0.5$ or minimum $r^2 < 0.3$ in any imputation batch, or failed a test for imputation batch effect (testing imputation dosage with imputation batch; $P < 10^{-50}$).

We applied logistic regression assuming an additive model to test for association between case/control status and either genotypes or imputed dosages (for imputed SNPs). Only SNPs with minor allele frequency $(MAF) > 0.1\%$ were analyzed. Covariates were added to adjust for age, sex, the first five principal components, and genotyping platform version. A total of 12,896,220 variants (11,933,700 SNPs) were analyzed. The genomic inflation factor was calculated from the median P value of analyzed variants. Scaling of the genomic inflation factor by sample size was carried out as described previously for 1,000 cases and 1,000 controls⁴⁸.

Meta-analysis of PD GWASs

Summary odds ratios, 95% CIs, and P values of the 10,000 most significant GWAS metaanalysis results were obtained from PDGene ("URLs"). Cohort descriptions, quality control, and meta-analysis for this study have been described previously⁸. SNP s.e. was derived from the reported P values and odds ratios. More specifically, the z-statistic was calculated as the square root of the inverse γ -square transformation of the P value, and the s.e. was calculated as follows: s.e.m. $=$ ln(odds ratio)/absolute(*z*-score).

There were 9,830 SNPs in common between the PDGene and the PDWBS data sets. A fixed-effects model based on inverse-variance weighting, as implemented in METAL, was used to combine summary statistics from the two studies¹⁴. Heterogeneity values (\hat{P} and \hat{Q}) were obtained with PLINK⁴⁹. Novel signals of association were defined as genome-wide significant associations in the meta-analysis that did not overlap loci associated with PD at genome-wide significant thresholds in the PDGene data (35 loci with $P < 1 \times 10^{-6}$).

Joint analysis with NeuroX

The NeuroX cohort was previously described $8,17$. Briefly, 5,851 cases and 5,866 controls of European ancestry were genotyped on a semi-custom NeuroX array. A logistic regression was carried out to test for association, with covariates to adjust for age, sex, and population ancestry (the first five principal components). Twenty-five of the 35 novel loci were directly genotyped on the chip, and four additional SNPs had suitable proxies ($r^2 > 0.9$). At these 29 SNPs, we carried out a fixed-effects inverse-variance weighted meta-analysis14 for all three studies (PDGene, PDWBS, and NeuroX) as described above.

Conditional analysis

Conditional analysis was run on all 17 loci that were significantly associated with PD in the joint meta-analysis ($P < 5 \times 10^{-8}$) and the 24 previously reported PD loci using the PDWBS study. For each locus, SNPs within 500 kb of the index SNP (the SNP with the most

significant P value) were tested for association by the same methods as described above for the PDWBS GWAS with the index SNP added as an additional covariate.

Heritability estimates

We used LD score regression (LDSC)^{50,51} to compute the narrow-sense heritability (h^2) estimates of PD in the PDWBS GWAS data (described above). Several methods exist for estimating h^2 with GWAS data^{50–52}. We used LDSC to estimate h^2 in this study because it requires only summary-level data and is more computationally efficient for larger data sets. Reference LD scores were computed with the European ancestry subset of the 1000 Genomes data for SNPs within 500 kb of the SNP to be scored. Strict filtering was applied to ensure the robustness of heritability estimates as recommended^{50,51}. After filtering, Z scores for 7,629,099 SNPs from the 23andMe study were used as input to LDSC. We further used the stratified LD-score regression approach to partition heritability into 24 different cell-type-agnostic annotation categories including conserved regions, histone marks, DNase I hypersensitivity sites, ENCODE chromatin states, and enhancers⁵¹, as well as 10 different cell-type-specific histone annotations. Significant enrichment was assessed at a strict Bonferroni threshold of 0.0021 (0.05/24) for the 24 general categories, and 0.005 for the cell-type-specific enrichment.

Pleiotropy analysis: overlap with EBI-NHGRI GWAS catalog

Data were downloaded from the EBI-NHGRI catalog15 (version available on 17 April 2016). If a variant in the meta-analysis was within 500 kb and in LD ($r^2 > 0.8$) with an association ($P < 5 \times 10^{-8}$) in the catalog, the meta-analysis signal was considered to be overlapping the reported signal.

A neurocentric strategy to identify candidate causal variants and genes

Associated index SNPs were paired to candidate genes on the basis of two broad levels of evidence: variant-level support and gene-level support (see Supplementary Fig. 4 for a graphic representation). In the former category, index SNPs were paired with candidate genes if there was evidence that the index SNP or an SNP in LD ($t^2 > 0.6$) with the index SNP was annotated with a putative high-impact variant (chromosome number variation, exon loss variant, frame-shift variant, rare amino acid variant, splice donor or acceptor variant, start-lost, stop-gained or stop-lost, and transcript ablation) or moderate-impact variant $3'$ or $5'$ UTR truncation and exon loss, coding sequence variant, disruptive inframe deletion or insertion, in-frame deletion or insertion, missense variant, regulatory region ablation, splice region variant, and transcription factor binding-site ablation). We obtained variant annotations by running SnpEff⁵³ on dbSNP build 142. A second source of variant-level support consisted of cis-eQTL evidence. Cis-eQTLs as pre-computed by GTEx $(v6)^{28}$ were downloaded directly from the GTEx portal ("URLs"). Although eQTL results were available for 46 tissues, including ten regions from the brain, our search for eQTLs was limited by the sampled tissues and cell types, and therefore we might have missed any eQTLs that are cell-type or tissue specific, in addition to eQTLs that are present only under certain stimuli (for example, 'response' eQTLs). The index SNP was tested for significant association with any gene where the TSS was within 250 kb of the index SNP. As roughly 90% of eQTLs are within 250 kb of a gene²⁸, it is likely that we captured the majority of

eQTLs while missing rarer, more distal events. For brain eQTLs, a strict Bonferroni correction was applied to the raw eQTL P values to adjust for multiple testing of genes within 250 kb of the index SNP. For other tissues, only eQTLs with a false discovery rate of ≤ 0.05 as determined by GTEx²⁸ were considered. We weighted brain and non-brain eQTLs equally.

Gene-level support was used when an index SNP had no candidate genes supported by variant-level data as described above. A list of genes within 250 kb of the index SNP was obtained (gene models used by GTEx were downloaded from the GTEx portal), and each gene was scored for neurological relevant features or annotations. Genes were first weighted for neurological relevant phenotypic annotations. Genes were (i) scored for being differentially expressed between PD patients and healthy controls (see Supplementary Note 1 for further details) (311 genes total genome-wide); (ii) annotated with 'neuro'-associated phenotypes in FlyBase54 (1,521 genes); (iii) scored for behavioral, neurological, and olfactory phenotypes annotated in MGI55 (3,890 genes); and (iv) annotated with any phenotypes related to neurological disorders or the brain in $OMIM⁵⁶$ (521 genes). Lastly, genes were scored for being expressed (median expression across samples > 2 reads per kilobase per million mapped reads) in any cohort of GTEx brain region samples (15,197 genes) or in at least one brain cell type in the mouse expression data set²⁹ (astrocyte, microglia, neuron, or oligodendrocyte) (12,092 genes). For the gene-level support, we used a tiered scoring scheme to weight phenotypic annotations more heavily than expression in the brain (scores demarked in Supplementary Fig. 4) to enrich for genes with demonstrated neurological related roles. At each locus, the gene (or tied genes) with the highest score was nominated as the candidate gene for the region.

Protein–protein and coexpression analysis

All protein-coding genes within 250 kb of PD-associated loci (Supplementary Table 13) were used as input to STRING⁵⁷. Gene pairs that were either coexpressed or involved in experimentally validated protein–protein interactions with a medium score or higher (score 0.4) are reported in Supplementary Table 10.

Pathway enrichment analysis

Previously reported PD loci and novel PD associations were tested for enrichment in particular pathways or gene sets. First, the nominated candidate genes for these PDassociated loci were tested for enrichment in several targeted gene sets by a hypergeometric test. The background list of genes for comparison was matched to the neurological-centric candidate-gene nomination pipeline. The background list thus consisted of genes that had mouse knockout phenotypes, had fly mutant phenotypes, had OMIM-related phenotype annotations, had nominally significant cis-eQTLs in GTEx, were differentially expressed in PD patients versus controls, and were expressed in GTEx brain tissue or mouse brain cell types in the Barres data set.

We obtained mitochondrial genes from MitoMiner⁵⁸ using the MitoCarta^{59,60} reference set after excluding genes that mapped to the mitochondria (genes that map to the mitochondria were not included in this metaanalysis). Lysosomal genes were obtained from the hlGDB⁶¹

using only the proteomics and literature resources. Finally, we obtained autophagy genes from the Human Autophagy Database⁶², as well as ten additional genes reported in a recent siRNA screen of autophagic flux modulators³⁵. A list of all genes in each pathway is provided in Supplementary Note 1. The minimum P value per gene (for genes that an SNP within the 9,830 variants assayed in this meta-analysis mapped to) is provided in Supplementary Tables 14–16.

Second, we applied a non-targeted gene-set enrichment approach using $INRICH³⁶$ to assess whether regions associated with PD were enriched for genes in KEGG⁶³ and Gene Ontology $(GO)⁶⁴$ gene sets. The 24 previously reported PD index variants and the 17 novel PDassociated variants reported in this study were used as input into PLINK's⁶⁵ "show-tags" function. The European 1000 Genomes¹² samples were used for reference LD patterns. An interval for each PD-associated variant was defined as the region from the leftmost tag variant to the rightmost tag variant in the 1000 Genomes data.

We ran INRICH on these 41 intervals with the default settings, with the exception of increasing the number of replicates and bootstraps to 5,000 (-r 5000 --q 5000) and setting the pre-compute feature to false for software stability (-c). Enrichment for KEGG and GO gene sets was assessed separately.

Data availability

A **Life Sciences Reporting Summary** for this paper is available. Summary statistics for the 9,830 variants presented in the discovery phase meta-analysis are available at [http://research](http://research-pub.gene.com/chang_et_al_2017)[pub.gene.com/chang_et_al_2017.](http://research-pub.gene.com/chang_et_al_2017) The full GWAS summary statistics for PDWBS will be made available through 23andMe and Genentech to qualified researchers under an agreement with 23andMe that protects the privacy of the 23andMe participants and an agreement with Genentech for data sharing. Please contact D.H. [\(dhinds23andme.com](http://dhinds23andme.com)) for more information and to apply to access the data.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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Figure 1.

A flow chart of the two-stage meta-analysis design. In stage 1, we carried out a metaanalysis of 9,830 SNPs between the PDWBS and PDGene studies. Thirty-five loci with P < 1×10^{-6} were carried forward into the replication-phase meta-analysis. In stage 2, we carried out a meta-analysis between the two discovery-phase studies and the NeuroX study for these 35 loci. Of these loci, 16 of the 29 available in NeuroX and 1 locus without replication data were carried forward for downstream analyses (see the main text for further details).

Figure 2.

Results of the Parkinson's disease discovery-phase meta-analysis. The top SNPs in associated regions are indicated by pink symbols. Candidate genes for previously associated loci are labeled in black ($P < 5 \times 10^{-8}$ in the discovery phase) or gray text ($P > 5 \times 10^{-8}$ in the discovery phase); candidate genes for newly identified loci are labeled in red. The y-axis shows the two-sided unadjusted $-\log_{10}(P)$ values for association with PD. SNPs with $P < 1$ \times 10⁻²⁵ are indicated by triangles.

Figure 3.

The candidate genes for regions associated with Parkinson's disease. The most likely candidate gene is annotated for each region that was significantly associated with PD in the final joint analysis. Black or gray text indicates previously reported loci that had P values less than or greater than 5×10^{-8} in the discovery phase, respectively. Red text indicates newly identified loci that were significantly associated with PD in the final joint analysis. Gray lines at the outer edge spanning multiple genes indicate candidate genes within a single locus. Chromosome numbers are shown in the gray shaded ring, and support for candidate genes is indicated by color-coding in the inner rings. The innermost ring indicates expression of the gene in brain cell types (in a mouse expression data set) or in human brain regions (in GTEx), or differential expression between PD brains and healthy control brains.

Table 1

Parkinson's disease risk loci previously reported at genome-wide significance levels Parkinson's disease risk loci previously reported at genome-wide significance levels

Effect allele frequency (EAF) measured in PDWBS controls or cases.

P value for SNP in the publicly available PDGene data (13,708 cases, 95,282 controls). Publicly available data for the following SNPs include an additional 5,450 cases and 5,798 controls genotyped on NeuroX: rs115185635, rs35749011, rs117896735, rs62120679, rs9275326, rs3793947, rs1555399, rs1474055, and rs8118008.

P value for SNP in PDWBS (6,476 cases, 302,042 controls).

The alternate SNP is genome-wide significant (rs12651582; $P = 3.51 \times 10^{-8}$). g The alternate SNP is genome-wide significant (rs76904798; $P = 4.45 \times 10^{-75}$).

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Table 2

Seventeen novel regions associated with Parkinson's disease at genome-wide significance levels Seventeen novel regions associated with Parkinson's disease at genome-wide significance levels

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Summary statistics are shown for the discovery cohort (PDWBS and PDGene), NeuroX (5,851 cases, 5,866 controls), and the joint meta-analysis of the discovery and NeuroX data.

Additional summary statistics for NeuroX and the joint meta-analysis are available in Supplementary Table 5. EAF, effect allele frequency.

Additional summary statistics for NeuroX and the joint meta-analysis are available in Supplementary Table 5. EAF, effect allele frequency.

 $^4\rm Chromosome$ and physical position according to Hg19. Chromosome and physical position according to Hg19.

 b betails regarding the assignment of candidate genes are provided in the Online Methods. Details regarding the assignment of candidate genes are provided in the Online Methods.

^CNeuroX and joint statistics are shown for proxy SNP rs1293298. NeuroX and joint statistics are shown for proxy SNP rs1293298.