

1 **Excessive burden of lysosomal storage disorder gene variants in Parkinson's disease**

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40 Running Head: Lysosomal storage disorder genes and Parkinson's disease

1 **Abstract**

2 Mutations in the *glucocerebrosidase* gene (*GBA*), which cause Gaucher disease, are also potent  
3 risk factors for Parkinson’s disease. We examined whether a genetic burden of variants in other  
4 lysosomal storage disorder genes is more broadly associated with Parkinson’s disease  
5 susceptibility. The sequence kernel association test was used to interrogate variant burden among  
6 54 lysosomal storage disorder genes, leveraging whole exome sequencing data from 1,156  
7 Parkinson’s disease cases and 1,679 control subjects. We discovered a significant burden of rare,  
8 likely damaging lysosomal storage disorder gene variants in association with Parkinson’s disease  
9 risk. The association signal was robust to the exclusion of *GBA*, and consistent results were  
10 obtained in 2 independent replication cohorts, including 436 cases and 169 controls with whole  
11 exome sequencing and an additional 6,713 cases and 5,964 controls with exome-wide  
12 genotyping. In secondary analyses designed to highlight the specific genes driving the aggregate  
13 signal, we confirmed associations at the *GBA* and *SMPD1* loci and newly implicate *CTSD*,  
14 *SLC17A5*, and *ASAHI* as candidate Parkinson’s disease susceptibility genes. In our discovery  
15 cohort, the majority of Parkinson’s disease cases (56%) have at least one putative damaging  
16 variant in a lysosomal storage disorder gene, and 21% carry multiple alleles. Our results  
17 highlight several promising new susceptibility loci and reinforce the importance of lysosomal  
18 mechanisms in Parkinson’s disease pathogenesis. We suggest that multiple genetic hits may act  
19 in combination to degrade lysosomal function, enhancing Parkinson’s disease susceptibility.

20

21 **Key Words**

22 Parkinson’s disease

23 Lysosomal storage disorders

1 Genetics

2 Whole exome sequencing

3

#### 4 **Abbreviations**

5 CADD Combined Annotation Dependent Depletion

6 GBA Glucocerebrosidase

7 IPDGC International Parkinson's Disease Genomics Consortium

8 LSD Lysosomal Storage Disorder

9 MAF Minor allele frequency

10 PPMI Parkinson's Progression Markers Initiative

11 QC Quality control

12 RSX-1 Rotterdam Study exome dataset version 1

13 SKAT-O Sequence kernel association test – optimized

14 WES Whole exome sequencing

15

#### 16 **Introduction**

17 Parkinson's disease is a common neurodegenerative disorder with evidence for a substantial

18 genetic etiology (Kalia and Lang, 2015). Studies in families as well as large population-based

19 cohorts have implicated more than 30 genes (Bras *et al.*, 2015; Singleton *et al.*, 2013;

20 Verstraeten *et al.*, 2015); however, the risk alleles identified to date explain only a fraction of

21 Parkinson's disease heritability estimates (Do *et al.*, 2011; Hamza and Paymi, 2010; Keller *et al.*,

22 2012), suggesting the involvement of additional loci. Beyond discovering the responsible genes,

23 a major challenge remains to understand the mechanisms by which these factors alter disease

1 onset and/or progression, including whether they act independently or interact within coherent  
2 biologic pathways.

3       Substantial evidence highlights the importance of lysosomal mechanisms in Parkinson’s  
4 disease susceptibility and pathogenesis (Kalia and Lang, 2015; Moors *et al.*, 2016; Vekrellis *et*  
5 *al.*, 2011; Wong and Krainc, 2016). Prior to its discovery as a Parkinson’s disease risk locus, the  
6 *glucocerebrosidase* gene, *GBA*, was known to cause Gaucher disease, an autosomal recessive  
7 lysosomal storage disorder (LSD). Increased risk for Parkinson’s disease in heterozygous carriers  
8 of *GBA* loss-of-function alleles was first recognized in families of individuals with Gaucher  
9 disease (Goker-Alpan *et al.*, 2004; Tayebi *et al.*, 2003). Follow-up studies in large, case-control  
10 samples confirmed that heterozygous *GBA* variants confer at least a five-fold increased risk of  
11 Parkinson’s disease (Aharon-Peretz *et al.*, 2004; Sidransky *et al.*, 2009). *GBA* variants may also  
12 modify Parkinson’s disease clinical manifestations (Brockmann *et al.*, 2015; Clark *et al.*, 2007;  
13 Davis *et al.*, 2016; Winder-Rhodes *et al.*, 2012), causing earlier age-of-onset, higher risk of  
14 cognitive impairment, and accelerated progression. LSDs—of which there are more than 50—are  
15 strictly Mendelian-inherited, metabolic disorders collectively caused by dysfunction in lysosomal  
16 biogenesis or function, and similarly characterized by the abnormal accumulation of non-  
17 degraded metabolites in the lysosome (Boustany, 2013; Filocamo and Morrone, 2011). The  
18 strong genetic evidence linking Gaucher disease and Parkinson’s disease risk leads to the  
19 intriguing, generalized hypothesis that LSDs and Parkinson’s disease may share a common  
20 genetic mechanism. Other LSD genes have therefore become attractive candidate risk factors for  
21 Parkinson’s disease (Deng *et al.*, 2015; Shachar *et al.*, 2011). Several studies have consistently  
22 supported a role for *SMPD1* (Clark *et al.*, 2015; Foo *et al.*, 2013; Gan-Or *et al.*, 2013; 2015; Wu  
23 *et al.*, 2014), which causes Niemann-Pick disease, Type A/B. Initial reports evaluating other

1 LSD genes, including *NPC1*, *NPC2*, *MCOLN1*, *NAGLU* and *ARSB*, have either shown  
2 conflicting results or await further replication (Clark *et al.*, 2015; Jansen *et al.*, 2017;  
3 Klunenmann *et al.*, 2013; Winder-Rhodes *et al.*, 2012; Zech *et al.*, 2013). LSDs are individually  
4 quite rare in populations of European ancestry, as are the known genetic variants established to  
5 cause these disorders (Boustany, 2013; Filocamo and Morrone, 2011). However, with the  
6 exception of *GBA*, most studies of LSD gene candidates have been small and therefore likely  
7 underpowered to detect the effects of rare alleles or those with more modest effect sizes.  
8 Genome-wide association studies in large Parkinson's disease case-control cohorts have  
9 independently implicated more common risk alleles at another LSD gene, *SCARB2* (Do *et al.*,  
10 2011; Nalls *et al.*, 2014), which encodes a membrane protein required for correct targeting of  
11 glucocerebrosidase to the lysosome. Besides this growing genetic evidence, studies in cellular  
12 and animal models also implicate the lysosome in the clearance of alpha-synuclein (Cuervo *et*  
13 *al.*, 2004; H. J. Lee, 2004; Vogiatzi *et al.*, 2008), which aggregates to form Lewy body pathology  
14 in Parkinson's disease. Reciprocally, alpha-synuclein disrupts neuronal vesicle trafficking and  
15 lysosomal function (Cooper *et al.*, 2006; Mazzulli *et al.*, 2011; Moors *et al.*, 2016; Wong and  
16 Krainc, 2016).

17 In this study, we leverage the largest Parkinson's disease whole exome sequencing  
18 (WES) dataset currently available to systematically examine the overlap between genes  
19 responsible for LSDs and Parkinson's disease. Our results reveal an aggregate burden for genetic  
20 variants among 54 genes established to cause LSDs and suggest that many genes besides *GBA*  
21 likely contribute to susceptibility for Parkinson's disease.

22

## 23 **Materials and Methods**

## 1 **Subjects**

2 Clinical and demographic features for our study cohorts, which have also been described in other  
3 recent reports (Giri *et al.*, 2017; Jansen *et al.*, 2017), are shown in Supplemental Table 1. The  
4 International Parkinson's Disease Genomics Consortium (IPDGC) WES discovery dataset used  
5 for this study consists of 2,835 samples of Northern and Western European ancestry, including  
6 1,156 Parkinson's disease cases and 1,679 controls not known to have Parkinson's disease.  
7 Subjects were recruited from academic medical centers across the United States and Europe.  
8 Cases were recruited at a mean age of 51.5 years (SD=11.5) and diagnosed with Parkinson's  
9 disease at a mean age of 41.2 years (SD=10.8); 40.4% reported a positive family history. Control  
10 subjects were on average 63.7 years of age (SD=17.1). 1,201 control exomes originated from the  
11 Rotterdam Study exome dataset version 1 (RSX-1) (van Rooj *et al.*, 2017; Giri *et al.*, 2017). The  
12 Rotterdam Study is a prospective population-based cohort study based in Rotterdam, the  
13 Netherlands. WES was performed on DNA from participants from the RSX-I subcohort, enrolled  
14 in 1990, with an average age at baseline of 68.6 (SD=8.6, 54.4% female) (Hofman *et al.*, 2015).  
15 All IPDGC and RSX-1 subjects gave written informed consent for participation in genetic  
16 research, which was approved by relevant oversight committees and institutional review boards.  
17 Subjects with pathogenic variants in established Mendelian Parkinson's disease genes (*SNCA*,  
18 *LRKK2*, *VPS35*, *PARK2/parkin*, *PARK7/DJ-1*, or *PINK1*) were excluded from analysis (Jansen *et*  
19 *al.*, 2017). Following quality control filters, the Parkinson's Progression Markers Initiative  
20 (PPMI) replication dataset (Parkinson Progression Marker Initiative, 2011) includes 436 cases  
21 and 169 controls of Northwest European descent. Cases were recruited at a mean age of 61.7  
22 years (SD 9.7) and diagnosed with Parkinson's disease at an average age of 59.8 years  
23 (SD=10.0); 27.1% reported a positive family history. PPMI controls were an average of 61.8

1 years of age (SD=10.1) at the time of evaluation. Data used in the preparation of this article were  
2 obtained from the PPMI database ([www.ppmi-info.org/data](http://www.ppmi-info.org/data)); for up-to-date information on the  
3 study, visit [www.ppmi-info.org](http://www.ppmi-info.org). Samples analyzed for both the IPDGC and PPMI cohorts were  
4 derived from whole blood. The NeuroX cohort has also been previously described in detail  
5 (Jansen *et al.*, 2017; Nalls *et al.* 2015). A minority of subjects overlapping with the IPDGC WES  
6 discovery sample were removed, such that the NeuroX replication cohort was a completely  
7 independent sample, including 6,713 individuals with Parkinson's disease and 5,964 controls.  
8 NeuroX cases were diagnosed at an average age of 61.6 (SD=12.4) and controls were evaluated  
9 at an average age of 64.1 (SD=14.3).

10

## 11 **Sequencing/Genotyping and Quality Control**

12 Data generation and detailed quality control procedures for the IPDGC and RSX-1 samples has  
13 recently been reported (Giri *et al.*, 2017; Jansen *et al.*, 2017; van Rooj *et al.*, 2017). WES was  
14 performed using the Roche Nimblegen SeqCap v2 or Illumina exome capture kits to prepare  
15 sample libraries, followed by paired-end sequencing with Illumina HiSeq2000. The generation of  
16 the PPMI WES dataset are described elsewhere ([www.ppmi-info.org](http://www.ppmi-info.org)). Although the datasets  
17 originate from different consortia, the same algorithms were used for read processing. The  
18 Burrows-Wheeler Aligner-MEM algorithm (Li and Durbin, 2010) was used for alignment of  
19 sequencing reads to the human reference genome (hg19). Using Picard tools  
20 (<http://broadinstitute.github.io/picard>), Binary Alignment/Map files were generated in a sorted  
21 and indexed manner. Alignments were Base-Quality score recalibrated and indels realigned  
22 using the Genome Analysis Toolkit (McKenna *et al.*, 2010) v3.3-0, after which single nucleotide  
23 variants and small insertions/deletions were called with the HaplotypeCaller to one genomic

1 Variant Call Format file per individual. The IPDGC and RSX-1 WES datasets (hereafter referred  
2 to as simply the IPDGC discovery dataset) were merged by joint variant calling from the  
3 individual genomic Variant Call Format files. Variants that were not assigned with the standard  
4 Genome Analysis Toolkit quality annotation 'PASS' were excluded for subsequent analyses.  
5 94.4% and 98.0% of the IPDGC and PPMI exomes, respectively, achieved a minimum of 10x  
6 coverage.

7 As previously described (Giri *et al.*, 2017; Jansen *et al.*, 2017), for individual quality  
8 control, samples were excluded for ambiguous gender, deviating heterozygosity/genotype calls,  
9 low genotype call rates, or cryptic relatedness following identity-by-descent analyses. Population  
10 structure was further evaluated using multi-dimensional scaling component analysis based on  
11 linkage disequilibrium-pruned, genome-wide common variant markers. Prior to these  
12 calculations, our datasets were merged with available genotypes from 1000 Genomes Project  
13 (1000GP) ancestry-based population samples, including African (AFR), East Asian (EAS),  
14 European (EUR) and the Americas (AMR) (1000 Genomes Project Consortium 2012). Using the  
15 European samples as a reference, population outliers were excluded, resulting in the removal of  
16 39 or 9 individuals from the IPDGC and PPMI datasets, respectively. All remaining samples  
17 cluster tightly with European ancestry subjects on multi-dimensional scaling plots (Supplemental  
18 Figure 1). Genotype and variant quality control was accomplished by removal of low-quality  
19 genotypes (Phred-scaled genotype quality score < 20, depth < 8) and variants with low call rates  
20 or departure from Hardy-Weinberg equilibrium. Furthermore, for the IPDGC discovery dataset,  
21 variants were only considered when located within the overlapping targeted regions of the  
22 applied library preparation capture kits. Post-quality control procedures, a total of 462,946 and  
23 192,421 variants were called for the IPDGC and PPMI datasets, respectively.

1 Data generation and quality control for the NeuroX cohort has also previously described  
2 in detail (Jansen *et al.* 2017; Nalls *et al.*, 2015). NeuroX consists of 242,901 exonic variants from  
3 the Illumina Infinium HumanExome BeadChip and 24,706 custom variants related to neurologic  
4 disease. For individual quality control, as above, samples were excluded for gender ambiguity,  
5 dubious heterozygosity/genotype calls, evidence of relatedness, or poor clustering on multi-  
6 dimensional scaling plots (Supplemental Figure 1). We similarly excluded variants for low call  
7 rates, departure from Hardy-Weinberg equilibrium, or for significant differences in missingness  
8 rate between cases and controls. Post-quality control, we called 177,028 exonic variants from the  
9 NeuroX dataset.

10 Where allowable based on individual consents and institutional review board approval,  
11 the datasets used in this study, including WES and NeuroX data from the IPDGC, are publicly  
12 available. Data availability is detailed in Jansen *et al.* 2017 and at <http://pdgenetics.org/resources>.  
13 Data from PPMI is also available for download at [http://www.ppmi-info.org/access-data-  
14 specimens/download-data/](http://www.ppmi-info.org/access-data-specimens/download-data/).

15

## 16 **Variant Selection**

17 Our analyses initially considered 54 LSDs (Table 1), defined based on widely accepted clinical,  
18 pathologic, and metabolic criteria (Amberger *et al.*, 2015; Boustany, 2013; Filocamo and  
19 Morrone, 2011). All variants within the LSD gene set were extracted from the three datasets. For  
20 the IPDGC WES dataset, no variants in the genes *CLN5* and *NEUI* passed the pre-specified  
21 maximum missingness criteria of 15%, yielding 1,136 total exonic variants for consideration in  
22 these analyses. In addition, there were no non-synonymous variants identified in *SUMF1*.  
23 Variants were categorized in nested groups (Figure 1) including (1) nonsynonymous (n=760

1 variants in 51 genes), (2) likely damaging (n=596 variants in 51 genes), or (3) loss-of-function  
2 (n=69 variants in 27 genes) (see Table 1 and Supplemental Table 2). Loss-of-function variants  
3 included stop gain/loss, frameshift, and splicing mutations falling within two base pairs of exon-  
4 intron junctions. Predictions of variant pathogenicity were obtained from ANNOVAR (Wang *et*  
5 *al.*, 2010), based on the Combined Annotation Dependent Depletion (CADD) algorithm (v1.3,  
6 <http://cadd.gs.washington.edu>) (Kircher *et al.*, 2014). CADD integrates predictions from  
7 numerous bioinformatic algorithms into a single “C-score” and ranks all possible nucleotide  
8 changes in the genome based on potential to disrupt gene/protein function. In accordance with  
9 prior work (Amendola *et al.*, 2015), we selected a stringent CADD C-score $\geq$ 12.37, representing  
10 the top ~2% most damaging of all possible nucleotide changes in the genome—this subset is  
11 enriched for known pathogenic alleles. For descriptive purposes, all putative damaging variants  
12 within the IPDGC discovery cohort were further cross-referenced with ClinVar (Landrum *et al.*,  
13 2016) to identify those previously established with pathogenicity for LSDs (Supplemental Table  
14 3). For the PPMI cohort, no variants were called in *DNAJC5*, resulting in a dataset of 515 total  
15 exonic variants, of which 256 variants from 49 genes were nonsynonymous and 187 variants in  
16 47 genes met the CADD criteria for putative damaging changes (Supplemental Table 2). For the  
17 NeuroX cohort, all genes in the 54-gene set were represented, resulting in 467 nonsynonymous  
18 variants, of which 348 were classified as likely damaging (Supplemental Table 2). Within these  
19 categories, variants were filtered based on two minor allele frequency (MAF) thresholds: (a)  
20 <1% and (b) <3% (Figure 1). The latter, more relaxed frequency threshold is based on the  
21 population prevalence (de Lau and Breteler, 2006; Pringsheim *et al.*, 2014) and known  
22 incomplete penetrance of Parkinson’s disease risk alleles (Anheim *et al.* 2012; Marder *et al.*,  
23 2015; Rana *et al.* 2013; Trinh *et al.*, 2014). For a subset of individuals in the IPDGC (n=572) and

1 PPMI (n=566) WES cohorts, array-based genotyping data was also available, allowing us to  
2 compute concordance rates for genotyping calls present in both datasets using 2 independent  
3 assays (Supplemental Table 4). We observe complete concordance for *GBA* variants as well as  
4 nearly perfect concordance (>>99%) for variant genotype calls in the full LSD gene set.

## 6 **Statistical Analysis**

7 The sequence kernel association test – optimal (SKAT-O) (S. Lee *et al.*, 2012; 2016) was  
8 implemented in R using SKAT v1.0.9 to determine the difference in the aggregate burden of  
9 rare LSD gene variants between Parkinson’s disease cases and controls. SKAT-O aggregates  
10 genetic information across defined genomic regions to test for associations. Covariates were  
11 included to adjust analyses for gender and WES coverage (pre-quality control missingness).  
12 Twenty multi-dimensional scaling components were also included to account for other possible  
13 confounding factors (4 components for analyses of the NeuroX genotyping cohort). An  
14 empirical  $p$ -value ( $p$ ) was derived from the distribution of null results based on 10,000  
15 permutation trials in which case/control assignment was randomized. As diagrammed in Figure  
16 1, SKAT-O analysis was initially performed for the complete LSD gene set, considering each  
17 class of variants defined based on frequency and functional characteristics. In order to adjust for  
18 multiple comparisons, we applied the Bonferroni-Holm stepwise procedure to control for the  
19 familywise error rate and establish a corrected statistical significance threshold and adjusted  $p$ -  
20 value ( $p_{adj}$ ) based on a significance level,  $alpha$ , of 0.05 (Holm 1979). For those categories with  
21 a significant SKAT-O association in the full gene set, a secondary analysis was performed  
22 excluding all *GBA* variants in order to confirm the involvement of additional genes. For  
23 example, in the IPDGC discovery cohort, we adjusted for  $k=5$  or 2 comparisons for the number

1 of variant categories evaluated in the primary and secondary analyses, respectively. Due to the  
2 nested variant categories (Figure 1) and the highly interdependent nature of the respective  
3 burden tests, we separately considered those results with an empirical SKAT-O  $p$ -value  $< 0.05$ ,  
4 but not surviving the Bonferonni-Holm correction, as “suggestive”. Unadjusted, empiric SKAT-  
5 O  $p$ -values for all gene set analyses are included in Supplemental Table 5. Lastly, in order to  
6 highlight those loci driving associations detected in the gene set, secondary analyses were also  
7 performed using SKAT-O to evaluate variants in each LSD gene independently. For these per  
8 gene analyses, which we considered exploratory due to limited statistical power (below), we  
9 report all findings with an empirical unadjusted  $p$ -value  $< 0.05$ .

10 To estimate statistical power, we performed 1,000 SKAT simulations of causal  
11 subregions within the discovery or replication datasets. We assumed a Parkinson’s disease  
12 prevalence of 0.0041 and 0.0017 for the IPDGC and PPMI datasets, respectively, based on their  
13 distinct ages of onset (Supplemental Table 1) (Pringsheim *et al.*, 2014). For gene set simulations,  
14 subregion length was defined as the sum of individual LSD gene coding region lengths (169.5 kb  
15 or 170.4 in IPDGC and PPMI, respectively). For single gene simulations, the average gene  
16 length was used (3.5kb or 3.2 kb, respectively). The MAF cutoff for causal variants was set to  
17 0.00035 (based on the frequency of rare *GBA* loss-of-function alleles in the IPDGC data set) or  
18 0.03 for the rare or more common variant models, respectively, and penetrance was assumed to  
19 be either 100% or 10%. Because we predict that LSD gene variants associated with Parkinson’s  
20 disease will have a damaging effect, all causal variants were assumed to have a positive  
21 coefficient (risk rather than protective alleles).

22

## 23 **Results**

1 Variants were extracted from 54 genes responsible for LSDs, defined based on widely accepted  
2 criteria (Table 1), and filtered into nested categories based on 2 frequency thresholds and 3 tiers  
3 of functional criteria (Figure 1A). Our overall analytic approach is diagrammed in Figure 1B. To  
4 test our hypothesis that an aggregate burden of variants in the LSD gene set contributes to  
5 Parkinson's disease, we first implemented SKAT-O within the IPDGC WES discovery cohort  
6 (Table 2). Following adjustment for multiple comparisons (see Methods), significant associations  
7 were detected for the LSD gene set considering either all non-synonymous variants (category 1b,  
8  $p_{adj}=0.014$ ) or likely damaging variants (category 2b,  $p_{adj}=0.0055$ ), when using the more relaxed  
9 frequency threshold of  $MAF < 3\%$ . When considering only the subset of rare ( $MAF < 1\%$ )  
10 nonsynonymous or likely damaging variants, the SKAT-O result was attenuated and no longer  
11 significant (category 1a,  $p_{adj}=0.056$  and category 2a, 0.066, respectively). No association was  
12 observed when considering only loss-of-function alleles (category 3,  $p_{adj}=0.464$ ), possibly due to  
13 the relative paucity of such variants limiting statistical power (Supplemental Table 1). We next  
14 repeated analyses with significant results, but excluding all *GBA* variants. As expected, the  
15 strength of the associations was attenuated; however, both SKAT-O results including either all  
16 nonsynonymous variants ( $MAF < 3\%$ ) or the subset of likely damaging variants was robust to the  
17 exclusion of *GBA* and remained significant (category 1b,  $p_{adj}=0.026$  and category 2b,  
18  $p_{adj}=0.0198$ ). Our results indicate that the association between variant burden and Parkinson's  
19 disease risk in the IPDGC discovery cohort is mediated, at least in part, by the effects of LSD  
20 genes other than *GBA*, an established Parkinson's disease susceptibility locus.

21 To replicate our findings, we leveraged two independent cohorts, including an additional  
22 WES dataset from PPMI (436 Parkinson's disease cases and 169 controls) (Parkinson  
23 Progression Marker Initiative, 2011) and the NeuroX exome-wide genotyping dataset from

1 IPDGC (6,713 Parkinson’s disease cases and 5,964 controls) (Nalls *et al.*, 2015). We again  
2 implemented SKAT-O to detect a potential variant burden in Parkinson’s disease cases versus  
3 controls. In the smaller PPMI replication cohort, we discovered suggestive evidence for an  
4 excessive LSD variant burden in Parkinson’s disease (Table 2); however, this finding was not  
5 significant following adjustment for multiple comparisons (category 1a,  $p_{adj}=0.096$ ). The  
6 association signal—which appeared independent of *GBA* (Supplemental Table 5)—was detected  
7 exclusively among rare alleles ( $MAF < 1\%$ ) and only when considering all non-synonymous  
8 variants. It is possible that SKAT-O is sensitive to cohort differences between PPMI and the  
9 IPDGC, including both sample size and pertinent demographic features (e.g. age of onset and  
10 family history; Supplemental Table 1). However, in the substantially larger NeuroX dataset,  
11 significant burden associations were detected for the same 2 variant categories implicated by  
12 SKAT-O in the IPDGC discovery cohort (Table 2), despite the less comprehensive genotyping  
13 coverage compared to WES. A major driver for the robust LSD gene set association in NeuroX  
14 (category 1b,  $p_{adj}=0.0004$  and category 2b,  $p_{adj}=0.0003$ ) appears to be the more common  
15 *GBA*<sup>E326K</sup> variant ( $Freq_{Cases}=0.021$ ,  $Freq_{Controls}=0.011$ ), which has been reported to be associated  
16 with Parkinson’s disease risk in several large studies (Duran *et al.*, 2012; Pankratz *et al.*, 2012).  
17 Importantly, consistent with our findings in the IPDGC discovery cohort, the LSD gene set  
18 burden association for both of these variant categories remained significant in NeuroX following  
19 exclusion of *GBA* (category 1b,  $p_{adj}=0.002$  and category 2b,  $p_{adj}=0.020$ ). When considering only  
20 the subset of rare ( $MAF < 1\%$ ) variants in the NeuroX dataset, the SKAT-O result for the LSD  
21 gene set was attenuated and no longer significant; although, the association in the  
22 nonsynonymous variant group remained suggestive, and this association was independent of  
23 *GBA* (Supplemental Table 5). In sum, based on analyses in three independent Parkinson’s

1 disease case-control datasets, we demonstrate a burden of variants in LSD genes associated with  
2 Parkinson's disease risk, and this signal is at least partially independent of *GBA*.

3 To determine which additional LSD genes/variants may be responsible for the observed  
4 association with Parkinson's disease risk, we performed exploratory analyses using SKAT-O to  
5 assess for potential contribution of variants within each gene considered independently. For these  
6 analyses, we returned to the IPDGC discovery dataset, and again focused on likely damaging  
7 variants, which showed the strongest association signal in our primary analysis (category 2b). In  
8 these gene-based analyses, besides the expected result for *GBA* ( $p = 0.0001$ ) and confirmation of  
9 *SMPD1* ( $p = 0.029$ ), we discover evidence of novel aggregate associations for variants in *CTSD*  
10 ( $p = 0.002$ ), *SLC17A5* ( $p = 0.005$ ), and *ASAHI* ( $p = 0.031$ ). The specific variants implicated for  
11 each of these genes are included in Supplemental Table 3, along with all other putative damaging  
12 variants considered in our full LSD gene set analysis. While our datasets are underpowered to  
13 definitively assess the contributions of a particular rare variant in any single gene (see  
14 Discussion), these results identify the most likely specific loci driving the aggregate LSD gene  
15 set association signal detected in the IPDGC discovery sample.

16 Lastly, we examined the distribution of putative damaging LSD gene variants ( $MAF <$   
17  $3\%$ , category 2b) within the IPDGC WES cohort (Figure 2). Consistent with our finding of an  
18 excessive variant burden in Parkinson's disease, the distribution of variants appeared modestly  
19 right-skewed in cases. The average variant burden among IPDGC cases was 0.9 alleles per  
20 individual, which was slightly higher than that seen in controls (0.8 alleles per individual). Given  
21 their commonality, the majority of IPDGC cases (56%) have at least one putative damaging  
22 variant in an LSD gene, and 21% carry multiple alleles. Notably, only 22 out of 1156 total  
23 Parkinson's disease cases are homo- or hemizygous for putative damaging LSD variants

1 (Supplemental Table 6), suggesting that Mendelian recessive or X-linked inheritance may  
2 contribute minimally to the overall burden association. As discussed further below, our findings  
3 are consistent with a hypothetical model in which multiple LSD gene variants may interact to  
4 influence Parkinson's disease risk.

## 6 **Discussion**

7 This study reveals an important connection between the genetic factors broadly responsible for  
8 LSDs, which are predominantly pediatric Mendelian disorders, and Parkinson's disease, an  
9 adult-onset neurodegenerative disorder with complex genetic etiology. Specifically, among 54  
10 genes that cause LSDs, we find evidence for a burden of damaging alleles in association with  
11 Parkinson's disease risk. This association persisted after excluding *GBA*, consistent with a  
12 contribution from additional LSD genes. More than half of Parkinson's disease cases in our  
13 cohort harbor one or more putative damaging variants among the LSD genes. Thus, our results  
14 implicate several promising new Parkinson's disease susceptibility loci and reinforce the  
15 importance of lysosomal mechanisms in Parkinson's disease pathogenesis.

16 The strengths of this study include a large Parkinson's disease case/control discovery  
17 cohort as well as two independent datasets for replication of our findings. The IPDGC WES  
18 discovery sample is characterized by younger-onset Parkinson's disease cases (mean age~41  
19 years) and those with a positive family history, thereby enriching for individuals with a potential  
20 genetic contribution. Recruitment of a substantially older IPDGC control group (mean age~64),  
21 reduces the possibility of latent, unrecognized Parkinson's disease (i.e. with minimal or absent  
22 symptoms), likely further increasing power for genetic discovery. By contrast, our PPMI and  
23 NeuroX replication cohorts include older cases (mean age~62 years) and age-matched controls,

1 making them more broadly representative of the older adult population commonly affected by  
2 Parkinson's disease. Consistent findings of an excessive LSD variant burden across these 3  
3 datasets, especially the large NeuroX sample (n~12,677), strongly enhances the generalizability  
4 of our conclusions. To minimize the possibility of population stratification, stringent quality  
5 control filters were implemented to ensure a homogeneous European ancestry sample in all study  
6 cohorts (Supplemental Figure 1). Nevertheless, it will also be important to examine other ethnic  
7 populations in the future, especially those potentially enriched for LSD-causing variants due to  
8 genetic bottlenecks.

9         Since our understanding of the characteristics of causal alleles—including in both  
10 Parkinson's disease and LSDs—is incomplete, our initial analyses systematically considered  
11 multiple variant classes binned into categories based on frequency and putative functional  
12 impact. In the IPDGC and PPMI cohorts, WES offers comprehensive characterization of LSD  
13 gene variants. By contrast, since the NeuroX data is restricted to those variants included on the  
14 genotyping array, it is possible that many potential pathogenic variants would be missed.  
15 Nevertheless, a total of 348 putative damaging variants were detected, including alleles for all  
16 LSD genes (Supplemental Table 2). Importantly, the selected analytic tool, SKAT-O, is robust to  
17 a wide frequency spectrum, including rare and more common alleles, and to variants with  
18 different magnitudes and directions of effect (S. Lee *et al.*, 2012; 2016). Our results suggest that  
19 consideration of likely damaging alleles based on bioinformatic predictions, including more  
20 common LSD variants (MAF < 3%), appeared to offer optimal sensitivity for detection of a  
21 significant aggregate variant association. Many of these variants are known to be pathogenic for  
22 LSDs (Supplemental Table 3). For example, of the *GBA* variants considered in our analyses,  
23 27% of those with annotations available in ClinVar (Landrum *et al.*, 2016) are rated as likely or

1 definitively pathogenic. Critically, the implementation of burden association tests for joint  
2 consideration of LSD genes significantly improves statistical power over single gene and variant  
3 tests (Zuk *et al.*, 2014). In populations of European ancestry similar to our study cohorts, loss-of-  
4 function alleles, including those established to cause LSDs, are individually rare (Supplemental  
5 Table 2), and based on post-hoc simulations (see Methods), we estimate poor power for  
6 discovery of rare Parkinson's disease risk alleles at single loci. For example, assuming a rare  
7 variant model (MAF = 0.035%, as for *GBA* loss-of-function alleles in our sample) and even  
8 assuming full penetrance, the IPDGC discovery cohort has only 30% power to discover an  
9 association for a single gene. However, a similar simulation considering the full set of 54 LSD  
10 genes was fully powered (100%). Our consideration of higher frequency variants further  
11 enhances power for both discovery and replication, especially when coupled with filtering based  
12 on potential pathogenicity. For example, allowing for more common variants (MAF < 3%) and  
13 assuming 10% of such alleles are causal, we estimate that the smaller PPMI cohort achieves 95%  
14 power for replication of a gene set association, whereas negligible power (1%) is available for  
15 interrogation of a single gene candidate. We anticipate that larger WES datasets will significantly  
16 improve power, including for per gene analyses.

17 We also performed analyses in the IPDGC cohort to pinpoint the specific drivers from the  
18 LSD gene set responsible for increasing Parkinson's disease risk. Our results (i) recapitulate the  
19 established association with *GBA*, (ii) strengthen the emerging evidence in support of *SMPDI*,  
20 and (iii) newly implicate *SLC17A5*, *ASAH1*, and *CTSD* as candidate Parkinson's disease  
21 susceptibility genes. Recessive mutations in *SMPDI* cause Niemann-Pick type A/B disease and  
22 this locus has been independently implicated in Parkinson's disease risk based on several  
23 published studies (Clark *et al.*, 2015; Foo *et al.*, 2013; Gan-Or *et al.*, 2013; 2015; S. Lee *et al.*,

1 2012). While our analysis identified 21 candidate, putative damaging *SMPD1* risk alleles  
2 (Supplemental Table 3), most appear distinct from those reported in other studies of Parkinson's  
3 disease. One notable exception, *SMPD1p.L304P* (also referred to as p.L302P), was previously  
4 implicated in a study of Ashkenazi Jewish subjects (Gan-Or *et al.*, 2013). Another non-  
5 synonymous variant, p.P332L implicated in the IPDGC sample is at the same amino acid  
6 position as a different substitution, p.P332R, that was previously implicated in a Chinese  
7 Parkinson's disease cohort (Foo *et al.*, 2013). Among the novel candidate genes, *SLC17A5*,  
8 *ASAHI*, and *CTSD*, most of the implicated variants are rare (MAF<1%). Only 2 of these variants  
9 (*rs16883930* and *rs141068211* in *SLC17A5* and *ASAHI*, respectively) are present in the 1000  
10 Genomes reference (The 1000 Genomes Project Consortium, 2012), having been previously  
11 examined in genome-wide scans, and both were non-associated with Parkinson's disease risk  
12 ( $p>0.05$ ) based on available data (Lill *et al.*, 2012). Mutations in *SLC17A5*, *ASAHI*, and *CTSD*  
13 cause the rare LSDs, Salla disease, Farber Lipogranulomatosis, and Neuronal Ceroid  
14 Lipofuscinosis (CLN10), respectively. Whereas Sialin (the protein product of *SLC17A5*) is a  
15 lysosomal membrane transporter for sialic acid, Acid Ceramidase (*ASAHI*) participates in  
16 ceramide metabolism, similar to Glucocerebrosidase and Sphingomyelinase (*SMPD1*). In  
17 addition to promoting lysosomal stress, glucosylceramide, which accumulates in Gaucher  
18 disease, has been suggested to directly promote the aggregation of alpha-synuclein (Mazzulli *et*  
19 *al.*, 2011; Moors *et al.*, 2016). Interestingly, *CTSD* encodes a lysosomal aspartyl proteinase  
20 which has been independently implicated in alpha-synuclein degradation (Cullen *et al.*, 2009;  
21 McGlinchey and J. C. Lee, 2015). In sum, the LSD genes and variants implicated by our studies  
22 are excellent candidates for further replication, including resequencing and/or genotyping in the  
23 largest available Parkinson's disease case/control samples. Although we employed standard

1 quality control procedures for calling variants from WES and genotyping data, definitive  
2 confirmation of specific variants will require additional studies.

3         There is a growing recognition of the importance of lysosomal biology in Parkinson's  
4 disease pathogenesis (Moors *et al.*, 2016; Wong and Krainc, 2016). First, the lysosome is an  
5 important route for alpha-synuclein degradation (Cuervo *et al.*, 2004; H. J. Lee, 2004; Vogiatzi  
6 *et al.*, 2008). Genomic variants that elevate alpha-synuclein protein levels—such as rare locus  
7 multiplication (Singleton *et al.*, 2003) or a common polymorphism that enhances promoter  
8 activity (Soldner *et al.*, 2016)—also increase Parkinson's disease risk. Knockdown of selected  
9 LSD genes, including *GBA* or *SCARB2*, in neuronal cells or in mouse models impairs alpha-  
10 synuclein clearance (Cooper *et al.*, 2006; Rothaug *et al.*, 2014; Sardi *et al.*, 2011), whereas  
11 increasing glucocerebrosidase activity has the opposite effect (Mazzulli, Zunke, Tsunemi, *et al.*,  
12 2016; Migdalska-Richards *et al.*, 2016; Sardi *et al.*, 2011). Second, lysosomal autophagy plays a  
13 critical role in mitochondrial quality control, and substantial evidence, including from genetics,  
14 highlight mitochondrial dysfunction in Parkinson's disease (Haelterman *et al.*, 2014). Third,  
15 there is accumulating evidence from numerous experimental models that alpha-synuclein  
16 interferes with endoplasmic reticulum-to-Golgi vesicle trafficking, inducing reciprocal  
17 disruptions in lysosomal biogenesis (Cooper *et al.*, 2006). Expression of alpha-synuclein  
18 impeded trafficking of multiple hydrolases linked to LSDs, including *GBA*, within human  
19 dopaminergic neurons (Mazzulli, Zunke, Isacson, *et al.*, 2016). In one recent study, subjects with  
20 idiopathic Parkinson's disease, in which *GBA* carriers were excluded, were found to have modest  
21 but significantly reduced glucocerebrosidase enzymatic activity based on peripheral blood testing  
22 (Alcalay *et al.*, 2015). Fourth, besides *GBA* and the other genes implicated in our study,  
23 mutations in *ATP13A2*, a rare cause of recessive juvenile-onset parkinsonism and dementia has

1 been independently implicated to cause the LSD Neuronal Ceroid Lipofuscinosis (Bras *et al.*,  
2 2012). Lastly, many other common and rare Parkinson's disease risk alleles, including at  
3 *RAB7L1*, *GAK*, *LRRK2*, and *VPS35* have strong functional links to vesicle trafficking, including  
4 for lysosomal biogenesis and function. Together, these findings support a model in which partial  
5 loss-of-function in genes regulating lysosomal activity, such as those that cause LSDs, may  
6 increase vulnerability to alpha-synuclein-mediated mechanisms in Parkinson's disease.

7         While our analyses reveal a robust and replicable LSD variant burden in Parkinson's  
8 disease cases, the overall magnitude of the difference between variant frequencies in cases and  
9 controls appears modest (Figure 2). We speculate that this is probably an underestimate of the  
10 true difference due to several assumptions. Specifically, only a subset of the 54 LSD genes and  
11 760 nonsynonymous variants considered in our burden analyses are likely to be truly involved in  
12 Parkinson's disease risk. Further, as noted above, while the CADD framework allowed us to  
13 prioritize 596 variants as putative damaging alleles, larger Parkinson's disease exome datasets  
14 with improved statistical power will be required to resolve the specific LSD genes and variants  
15 that contribute to Parkinson's disease risk. Lastly, similar to *GBA* (Anheim *et al.* 2012; Rana *et*  
16 *al.* 2013), we expect that many of the other LSD gene variants contributing to Parkinson's  
17 disease risk may have individually modest and therefore incompletely penetrant effects, perhaps  
18 modified by alleles at other loci (Cooper *et al.* 2013). In sum, the likely (i) incomplete  
19 penetrance of many pathogenic variants along with (ii) contamination of our analyses by many  
20 benign variants would be expected to inflate estimates for the LSD variant burden among  
21 controls and attenuate the overall SKAT-O association.

22         Parkinson's disease heritability remains incompletely explained by the genes and variants  
23 identified to date (Do *et al.*, 2011; Hamza and Paymi, 2010; Keller *et al.*, 2012; Verstraeten *et*

1 *al.*, 2015). Besides the likelihood of yet undiscovered loci, alternative explanations for familial  
2 aggregation of disease include epigenetic changes due to shared environmental exposures or  
3 even false positive diagnoses due to phenocopies (Mullin and Schapira 2015; Pihlstrom 2011). In  
4 complex genetic disorders such as Parkinson’s disease, the cumulative impact of common and  
5 rare variants at multiple genomic loci, as well as non-additive interactions among alleles, likely  
6 also play an important role (Cooper *et al.* 2013; Lupski *et al.* 2011). Polygenic modeling  
7 approaches have previously demonstrated how common risk alleles can cumulatively impact  
8 Parkinson’s disease risk and age-of-onset (Escott-Price *et al.*, 2015; Nalls *et al.*, 2014). In  
9 addition, a recently published analysis in the IPDGC WES and NeuroX cohorts identified  
10 evidence for oligogenic interactions underlying Parkinson’s disease risk, including alleles for  
11 *GBA* and those for established Mendelian Parkinson’s disease genes (Lubbe *et al.*, 2016). In the  
12 IPDGC, WES reveals a substantial proportion of Parkinson’s disease cases (21%) carrying two  
13 or more likely damaging variants in LSD genes. Consistent with other reports (Clark *et al.*,  
14 2015), our observation suggests the possibility that multiple LSD gene variants may interact in a  
15 multi-hit, combinatorial manner to degrade lysosomal function, causing the accumulation of  
16 alpha-synuclein and potentially other toxic substrates, and increasing susceptibility for  
17 Parkinson’s disease. Oligogenic interactions such as those proposed here may be an important  
18 source for “missing heritability” in Parkinson’s disease (Mullin and Schapira 2015; Pihlstrom  
19 2011). Recent work has also implicated oligogenic inheritance in other neurologic disorders,  
20 including amyotrophic lateral sclerosis (Cady *et al.*, 2015; Kenna *et al.*, 2013; van Blitterswijk *et*  
21 *al.*, 2012) and idiopathic peripheral neuropathy (Gonzaga-Jauregui *et al.*, 2015), and further  
22 reveals how pleiotropic genes causing early-onset, monogenic disorders may act in combination  
23 to additionally trigger late-onset, complex genetic disorders (Cooper *et al.* 2013; Lupski *et al.*

1 2011). Future studies, including even-larger, case-control cohorts with WES and complementary  
2 experiments in Parkinson's disease cellular or animal models, are needed to further investigate  
3 whether a variant burden in LSD genes, perhaps in combination with other susceptibility loci,  
4 underlies oligogenic risk and contributes substantially to Parkinson's disease heritability.

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## 6 **Supplementary Material**

7 Supplementary material, including 2 Figures and 6 Tables, is available at Brain online.

## 9 **Appendix**

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16

## 1 **Figure Legend**

2

3 **Figure 1 Overall analytic strategy.** (Left) Variant categories. Because the number, frequency,  
4 and effect sizes of Parkinson's disease risk variants remains incompletely defined, our analyses  
5 considered three nested categories based on increasing variant pathogenicity: (1) all non-  
6 synonymous variants (Nonsyn), (2) likely damaging variants based on combined annotation  
7 dependent depletion (CADD) score, and (3) loss-of-function (LoF) variants. Based on the known  
8 prevalence of Parkinson's disease and incomplete penetrance documented for many risk alleles,  
9 we also considered 2 frequency thresholds, including rare ( $MAF < 1\%$ ) and somewhat more  
10 common ( $MAF < 3\%$ ) variants. (Right) Analysis Flowchart. The Sequence Kernel Association  
11 Test-Optimal (SKAT-O) was initially performed for the complete LSD gene set in the IPDGC  
12 discovery cohort, considering each variant category separately. For those categories with a  
13 significant SKAT-O association in the full gene set, a secondary analysis was performed  
14 excluding all *GBA* variants in order to confirm the involvement of additional genes. This was  
15 repeated in each of the replication cohorts (PPMI and NeuroX). Lastly, in order to highlight  
16 those loci driving associations detected in the gene set, secondary analyses were performed using  
17 SKAT-O to evaluate variants in each LSD gene independently.

18

19

20 **Figure 2 Distribution of LSD variants in the IPDGC cohort.** The number of likely damaging  
21 LSD variants ( $MAF < 3\%$ ,  $CADD\ C\text{-score} \geq 12.37$ ) per individual is shown versus the proportional  
22 representation in the IPDGC discovery cohort. Cases (Red) and Controls (Blue) are plotted  
23 separately. Many individuals harbor multiple LSD alleles, and the distribution is right-skewed  
24 among Parkinson's disease cases. The analysis considers variants in all 54 LSD genes.  
25 Supplemental Figure 2 shows a similar plot restricted to the 5 top driver genes.

26

27

# 1 Tables

## 2 Table 1 LSD Genes and Variants in the IPDGC cohort

Disease	Gene	Variants <sup>a</sup>
Aspartylglucosaminuria	<i>AGA</i>	13 (10)
Metachromatic Leukodystrophy	<i>ARSA</i>	5 (5)
Maroteaux-Lamy disease	<i>ARSB</i>	11 (10)
Farber Lipogranulomatosis	<i>ASAHI</i>	20 (17)
Kufor-Rakeb syndrome	<i>ATP13A2</i>	24 (18)
Neuronal Ceroid Lipofuscinosis (CLN3)	<i>CLN3</i>	18 (17)
Neuronal Ceroid Lipofuscinosis (CLN5)	<i>CLN5</i>	-
Neuronal Ceroid Lipofuscinosis (CLN6)	<i>CLN6</i>	10 (7)
Neuronal Ceroid Lipofuscinosis (CLN8)	<i>CLN8</i>	9 (4)
Cystinosis	<i>CTNS</i>	13 (12)
Galactosialidosis	<i>CTSA</i>	14 (11)
Neuronal Ceroid Lipofuscinosis (CLN10)	<i>CTSD</i>	7 (4)
Neuronal Ceroid Lipofuscinosis (CLN13)	<i>CTSF</i>	11 (9)
Pycnodysostosis	<i>CTSK</i>	6 (5)
Neuronal Ceroid Lipofuscinosis (CLN4B)	<i>DNAJC5</i>	5 (5)
Fucosidosis	<i>FUCA1</i>	15 (12)
Pompe disease	<i>GAA</i>	15 (10)
Krabbe disease	<i>GALC</i>	36 (30)
Morquio A disease	<i>GALNS</i>	22 (14)
Gaucher disease	<i>GBA</i>	39 (32)
Fabry disease	<i>GLA</i>	9 (7)
GM1-Gangliosidosis/Morquio B	<i>GLB1</i>	8 (4)
GM2-Gangliosidosis	<i>GM2A</i>	1 (1)
I-Cell disease	<i>GNPTAB</i>	39 (31)
Sanfilippo D syndrome	<i>GNS</i>	20 (11)
Neuronal Ceroid Lipofuscinosis (CLN11)	<i>GRN</i>	19 (12)
Sly disease	<i>GUSB</i>	17 (10)
Tay-Sachs disease	<i>HEXA</i>	20 (18)
Sandhoff disease	<i>HEXB</i>	8 (6)
Sanfilippo C syndrome	<i>HGSNAT</i>	18 (15)
Mucopolysaccharidosis Type IX	<i>HYAL1</i>	13 (9)
Hunter syndrome	<i>IDS</i>	9 (8)
Hurler syndrome	<i>IDUA</i>	8 (4)
Neuronal Ceroid Lipofuscinosis (CLN14)	<i>KCTD7</i>	4 (3)
Danon disease	<i>LAMP2</i>	9 (7)
Wolman disease	<i>LIPA</i>	14 (10)
Alpha-Mannosidosis	<i>MAN2B1</i>	12 (11)
Beta-Mannosidosis	<i>MANBA</i>	18 (15)
Mucopolysaccharidosis Type IV	<i>MCOLN1</i>	19 (14)
Neuronal Ceroid Lipofuscinosis (CLN7)	<i>MFSD8</i>	18 (14)
Schindler Disease/Kanzaki disease	<i>NAGA</i>	9 (8)
Sanfilippo B syndrome	<i>NAGLU</i>	10 (9)
Sialidosis	<i>NEU1</i>	-
Niemann-Pick Disease Type C1	<i>NPC1</i>	43 (35)
Niemann-Pick Disease Type C2	<i>NPC2</i>	2 (2)
Neuronal Ceroid Lipofuscinosis (CLN1)	<i>PPT1</i>	9 (7)
Sphingolipid-activator deficiency	<i>PSAP</i>	22 (16)
Action mycolonus-renal failure syndrome	<i>SCARB2</i>	10 (7)
Sanfilippo A syndrome	<i>SGSH</i>	10 (8)
Salla disease	<i>SLC17A5</i>	18 (17)
Niemann-Pick Disease Type A/B	<i>SMPD1</i>	25 (21)
GM3-Gangliosidosis	<i>ST3GAL5</i>	11 (11)
Multiple Sulfatase Deficiency	<i>SUMF1</i>	-
Neuronal Ceroid Lipofuscinosis (CLN2)	<i>TPP1</i>	15 (13)

3 <sup>a</sup>The number of variants (MAF < 3%) in each LSD gene is shown for the IPDGC  
4 discovery cohort, including total number of nonsynonymous variants and likely  
5 damaging variants based on CADD (in parentheses). Of the 54 LSD genes  
6 considered, no exonic variants in *CLN5* or *NEU1* passed quality control filters (see  
7 Methods), and no nonsynonymous variants were identified in *SUMF1*.  
8 LSD=Lysosomal storage disorder; CADD=Combined Annotation Dependent  
9 Depletion.

**Table 2 Analyses of LSD Variant Burden in Parkinson’s disease**

Cohort	Cases (n)	Controls (n)	Variants <sup>a</sup>	(a) MAF < 1%		(b) MAF < 3%	
				n <sup>b</sup>	<i>p</i> <sub>LSD</sub> <sup>c</sup>	n	<i>p</i> <sub>LSD</sub> ( <i>p</i> - <i>GBA</i> ) <sup>c</sup>
<i>Discovery</i>							
<b>IPDGC</b>	1,167	1,685	(1) nonsyn	746 (709)	0.056	760 (721)	0.014 (0.026)
			(2) CADD	585 (555)	0.066	596 (564)	0.0055 (0.0198)
			(3) LoF	69 (65)	0.464	- <sup>d</sup>	-
<i>Replication</i>							
<b>PPMI</b>	436	169	(1) nonsyn	243 (237)	0.096	256 (248)	0.320
			(2) CADD	179 (174)	0.294	187 (180)	0.281
<b>NeuroX</b>	6,713	5,964	(1) nonsyn	452 (443)	0.068	467 (456)	0.0004 (0.002)
			(2) CADD	338 (331)	0.057	348 (339)	0.0003 (0.020)

<sup>a</sup>Variants were classified into nested categories (Figure 1A) based on two frequency thresholds, MAF < 1% (a) or 3% (b), and three functional filters, all nonsynonymous (1), CADD likely damaging (2), and LoF (3).

<sup>b</sup>n=total number of LSD variant (number of variants excluding *GBA*). In parentheses, the number of variants excluding those in *GBA* are shown.

<sup>c</sup>Empirical SKAT-O *p*-values are based on 10,000 permutations following randomization of case/control status, and adjusted for multiple comparisons using the Bonferroni-Holm method (see Methods). As shown in Figure 1, primary analyses consider the variant burden among 54 LSD genes (*p*<sub>LSD</sub>). For significant SKAT-O results, secondary analyses were performed excluding all variants in *GBA* (*p*-*GBA*). Unadjusted *p*-values are reported in Supplemental Table 5.

<sup>d</sup>No additional LoF variants met the relaxed frequency threshold (MAF < 3%).

LSD=lysosomal storage disorder; MAF=minor allele frequency; IPDGC=International Parkinson’s Disease Genomics Consortium Discovery Cohort; PPMI= Parkinson's Progression Markers Initiative Replication Cohort; NeuroX = NeuroX exome array cohort; nonsyn=nonsynonymous variants; CADD=Combined Annotation Dependent Depletion; LoF= loss of function variants

## Figures

Figure 1 Overall Analytic Strategy.

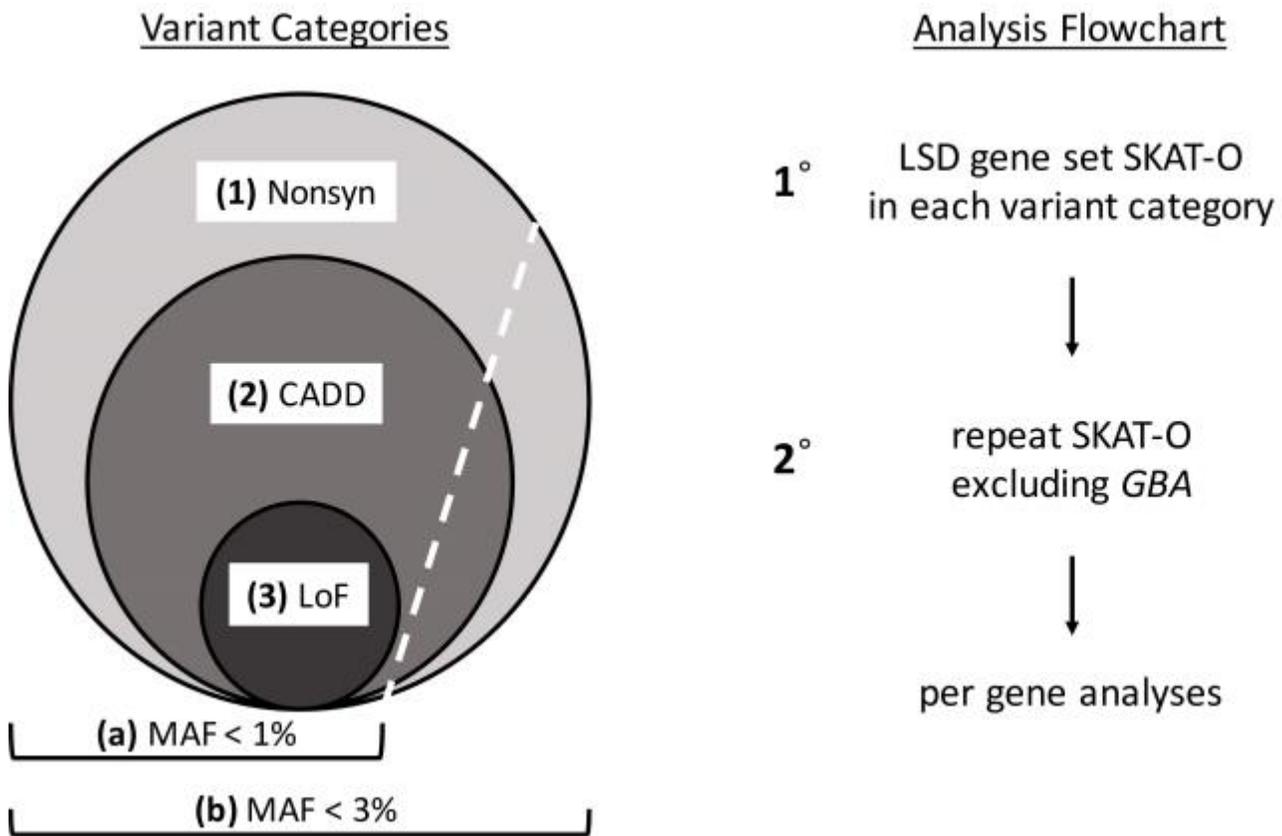


Figure 2 Distribution of LSD variants in the IPDGC cohort.

