

Genomic Landscape and Mutational Signatures of Deafness-Associated Genes

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The classification of genetic variants represents a major challenge in the post-genome era by virtue of their extraordinary number and the complexities associated with ascribing a clinical impact, especially for disorders exhibiting exceptional phenotypic, genetic, and allelic heterogeneity. To address this challenge for hearing loss, we have developed the Deafness Variation Database (DVD), a comprehensive, open-access resource that integrates all available genetic, genomic, and clinical data together with expert curation to generate a single classification for each variant in 152 genes implicated in syndromic and non-syndromic deafness. We evaluate 876,139 variants and classify them as pathogenic or likely pathogenic (more than 8,100 variants), benign or likely benign (more than 172,000 variants), or of uncertain significance (more than 695,000 variants); 1,270 variants are re-categorized based on expert curation and in 300 instances, the change is of medical significance and impacts clinical care. We show that more than 96% of coding variants are rare and novel and that pathogenicity is driven by minor allele frequency thresholds, variant effect, and protein domain. The mutational landscape we define shows complex gene-specific variability, making an understanding of these nuances foundational for improved accuracy in variant interpretation in order to enhance clinical decision making and improve our understanding of deafness biology.

Introduction

Genomic technologies have revolutionized medicine in the post-genome era by offering the promise of personalized, precision healthcare based on DNA sequencing.¹ Prior to and immediately after the completion of the human genome project, the primary bottleneck in advancing precision medicine was generating DNA sequencing and genetic variant data. With the advent of massively parallel sequencing technologies, the bottleneck shifted to clinically meaningful variant interpretation that is comprehensive, easily understandable, free from contradictory categorization, curated by experts, and freely available to the public. Guidelines developed by the American College of Medical Genetics and Genomics (ACMG) aid classification using a structured framework that defines 28 evidence codes by which to score a variant. There are 20 rules for combining codes to reach one of five conclusions that predict variant effect: pathogenic (P), likely pathogenic (LP), variant of uncertain significance (VUS), likely benign (LB), or benign (B).^{2,3}

The challenging and dynamic process of variant interpretation has spurred the creation of two major variant databases—ClinVar⁴ and the Human Gene Mutation Database (HGMD)⁵—to catalog the rapidly increasing volume of reported genetic variants. ClinVar is a freely accessible, public database that archives reports of the relationships

between variations and phenotypes with varying degrees of supporting evidence. HGMD, a pay-for-access service, is a comprehensive reference database of published germline mutations that are associated with human inherited diseases based on curation of published literature.⁶ These databases are invaluable resources but because of their broad all-encompassing design, are not disease specific.

Hearing loss is the most common sensory deficit in humans. It affects an estimated 5% of the world's population (360 million individuals) and in developed countries is most frequently genetic, segregating in a Mendelian fashion in the case of non-syndromic hearing loss (NSHL) or as a complex genetic disease in the case of age-related hearing loss.⁷ Its clinical evaluation has been facilitated by the use of comprehensive genetic testing with massively parallel sequencing, which has evolved to become the most informative test in the diagnostic evaluation of the hearing-impaired person. A positive diagnosis is made in more than 40% of persons who undergo this type of testing, and to date more than 6,000 mutations in more than 150 genes have been causally implicated in deafness.⁸ As the number of genes implicated in NSHL has continued to increase, we sought to provide a freely and continually updated comprehensive database to inform variant classification for deafness.

Called the Deafness Variation Database (DVD, see [Web Resources](#)), this resource is collated from major public

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databases. It provides a single classification for each variant based on collected evidence and is curated by experts in genetic hearing loss to provide a single-source guide to variant interpretation. By capitalizing on the wealth of data the DVD provides to assess the genomic and mutational landscape of deafness, we provide a deeper understanding of hereditary hearing loss and the molecular mechanisms at play.

Material and Methods

Gene Selection

The DVD v.8.1 includes 152 genes and microRNAs known to cause hearing loss-related phenotypes including NSHL, NSHL mimics such as Usher, Perrault, and Pendred syndromes (PDS [MIM: 274600]), and common forms of syndromic hearing loss like Alström (ALMS [MIM: 203800]), branchio-oto-renal (BOR1 [MIM: 113650], BOR2 [MIM: 610896]), Jervell and Lange-Nielsen (JLNS1 [MIM: 220400], JLNS2 [MIM: 612347]), and Wolfram (WFS1 [MIM: 222300], WFS2 [MIM: 604928]) syndromes (Table S1). The genes are curated from the Hereditary Hearing Loss Homepage and published literature after careful review of the supporting evidence including the strength of the genetic data (linkage data, allele frequency and deleteriousness of the candidate variant, segregation analysis) and functional data (gene expression in inner ear, *in vivo* experiments, animal models). The gene list is regularly updated by adding or removing genes based on newly published data.

Annotation Collection

Data for the DVD are collected, combined, filtered, and analyzed using a custom-built internal computational pipeline we have developed called Kafeen. The pipeline was built using the Ruby programming language and integrates open-source and freely available bioinformatics utilities including BCFTools and BEDTools.^{9,10} Variants are collected and annotated from multiple data sources including the 1000 Genomes Project (phase 3.5a),¹¹ Exome Sequencing Project (ESP) (ESP6500SI-V2-SSA137 release) (see Web Resources), Exome Aggregation Consortium (ExAC) (v.0.3),¹² HGMD (2015.2 release),⁵ ClinVar (2016-03-02 release),⁴ dbSNP 146,¹³ and our manual curations (Figure 1A). Additional annotations for pathogenicity prediction algorithms are collected from dbNSFP (v.3.0a).^{14,15} Transcripts on which the variant has the most deleterious impact are selected. When all transcripts are equally impacted (e.g., the variant is missense in all of them), then the canonical transcript is selected. All tab-delimited files are converted to VCF. All VCF sources are further left-aligned, normalized, quality filtered, and de-duplicated before input to the pipeline. The pipeline is extensible to incorporate additional variant and annotation sources as they become available. Copy number variants (CNVs) are not included in the DVD.

Scoring System and Interpretation

Available annotations are utilized to make an informed interpretation about the pathogenicity of each variant. MAF data from 1000 Genomes, ESP, ExAC, and our in-house database are used to determine whether a variant is too common to be considered pathogenic (Figure 1B). When considering multiple MAF annotations for a variant across databases and populations, we select the highest population-specific MAF to use in our computational evaluation.

As a general rule, we use a MAF threshold of 0.5%,¹⁶ with the exception of select variants in specific genes (i.e., *GJB2* [MIM: 121011]) (Table S2). A minimum of 400 alleles in the population with the highest MAF is required to use this classification threshold.

We consider a $MAF \geq 0.5$ to be inconsistent with P/LP for NSHL and implement this MAF cutoff in our pipeline. Any variant with a $MAF \geq 0.5\%$ is automatically classified as B (Figure 1C), although it remains important to know whether other databases classify common variants as P/LP (Figure 1D). To capture this information, we append an asterisk (*) to a B classification for common variants that are classified as P/LP by other databases. For example, if a variant has been classified as P in ClinVar but has a $MAF \geq 0.5\%$ in any population, the DVD classification is B*.

If no population-based MAF meets or exceeds the defined cutoff, the pipeline uses variant classification data from ClinVar and HGMD (Figures 1C and 1D). ClinVar provides user-submitted data, and a single variant can have multiple and conflicting classifications with varying degree of supporting evidence. In such cases, we select the most deleterious ClinVar classification as it carries the highest clinical significance for individuals with hearing loss. When ClinVar and HGMD classifications are concordant, the DVD uses that classification. For ClinVar-HGMD discrepancies, the DVD default classification is based on the level of discordance (Figure 1D). If the variant is reported in only one database, that classification is used for the DVD.

For a variant with $MAF < 0.5\%$ that is absent from both ClinVar and HGMD, the DVD relies on functional prediction annotations to classify the variant as either LB or VUS (Figure 1). Our pipeline currently supports two evolutionary conservation algorithms (phyloP¹⁷ and GERP++¹⁸) and four deleteriousness prediction algorithms (SIFT,¹⁹ PolyPhen-2,²⁰ MutationTaster,²¹ and LRT²²). The DVD calculates a composite pathogenicity score (PS), assigning 1 point for each conserved and damaging prediction to make a final prediction of either VUS or LB. When multiple scores for the same prediction or conservation algorithm are provided, DVD selects the most damaging prediction from the set to consider in its algorithms. The LB classification is warranted if a variant has $\leq 40\%$ pathogenic predictions and at least 5 algorithms make a prediction. In all other instances, the variant is computationally classified as a VUS (Figure 1C).

Classification Metrics

To validate this scoring metric and threshold, we tested all known deafness pathogenic variants with $MAF < 0.5\%$ and at least 5 algorithms calls (Figure S1). In calculation of performance of Kafeen variant classification prediction, we make the distinction between two classes of variants: positive as pathogenic classification and negative as benign classification. Within these classes, we consider the subclasses: true positive as predicted ($PS \geq 60\%$) and classified pathogenic variants; true negative as predicted ($PS \leq 40\%$) and classified benign variants; false positive as predicted pathogenic but classified benign variants; and false negative as predicted benign but classified pathogenic variants. Then, in calculating the traditional binary classification metrics of sensitivity, specificity, positive predictive value (PPV), and negative predictive value (NPV): sensitivity = true positive / all positive; specificity = true negative / all negative; PPV = true positive / (true positive + false positive); NPV = true negative / (true negative + false negative).

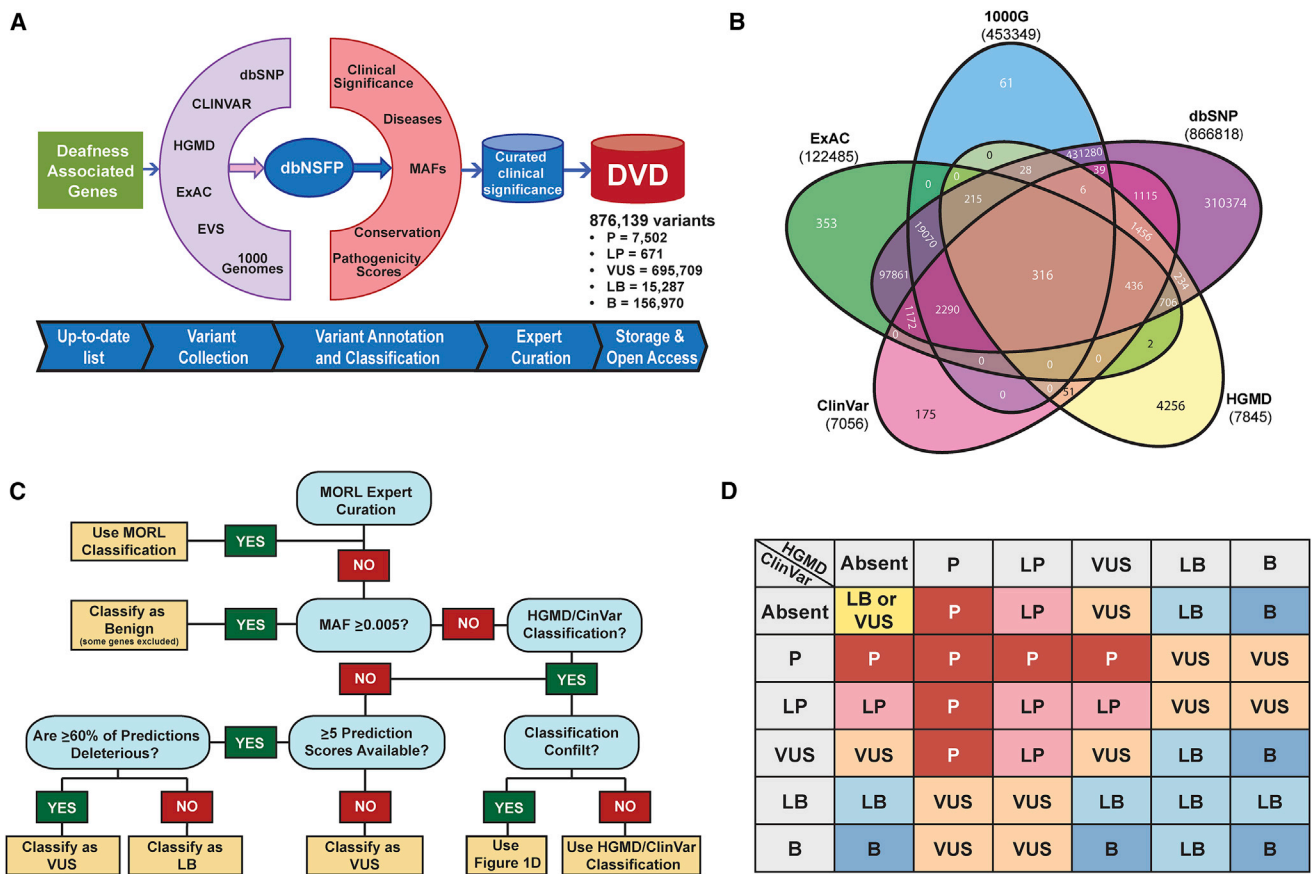


Figure 1. The Deafness Variation Database

(A) Kafeen, a custom internal pipeline, gathers data for the DVD by collecting variants and annotations from multiple data sources. Deleteriousness predictions collected from dbNSFP and MAF data are extracted from our local database, EVS, 1000 Genomes, and ExAC to inform the DVD classification. A comparison between DVD versus ClinVar and HGMD classifications captures all changes that result in medically significant differences (defined as up-grading a variant to P/LP or down-grading a variant from P/LP), each of which is manually curated to ensure that the DVD reclassification is appropriate.

(B) Venn diagram showing the number of variants collated from major population-scale MAF databases and the count of variants that are shared among them and those that are database specific.

(C) Decision tree for Kafeen classification.

(D) Decision matrix detailing Kafeen logic regarding variants classified in ClinVar and HGMD.

Implementation, Manual Curation, and Override

The DVD was implemented in our internal NGS pipeline, which we use to generate a clinical report for each subject evaluated with our targeted gene panel OtoSCOPE.⁸ A multidisciplinary expert panel, including clinicians, geneticists, scientists, bioinformaticians, and genetic counselors, reviews all genetic results in the context of available phenotypic data. When the expert panel does not agree with the variant classification in the DVD, the variant is added to an internal list of manually curated variants with the revised classification. This list is continually updated and integrated back into the DVD to prevent the propagation of an incorrect variant classification. The manually curated list includes pathogenic variants that have been identified exclusively in our screen of more than 5,000 individuals with hearing loss, as these variants were not found in other public databases.^{8,23,24} It also includes known pathogenic variants with MAFs $\geq 0.5\%$ (founder mutations and variants in specific genes, see Table S2), which have been deemed exempt from the MAF restriction.

Versioning the DVD

To keep the DVD up to date, we regularly update it by adding newly discovered deafness-associated genes and the most recent

versions of all input data sources. Testing and validation of each new DVD version is performed via a comparison between the newest dataset and the previous dataset to capture all variants that have undergone a major reclassification in pathogenicity (any changes from or to P and LP) resulting in medically significant differences (P/LP versus VUS/LB/B). We evaluate these variants to ensure that the reclassification is appropriate. If upon further review we do not agree with the reclassification, we preserve the previous classification.

Results

The Deafness Variation Database

The DVD classifies and interprets variants in 152 genes and microRNAs implicated in genetic hearing loss. The included genes are associated with a variety of hearing loss-related phenotypes including NSHL, NSHL mimics, and common forms of syndromic hearing loss (Table S1). 876,139 genetic variants in these genes were extracted

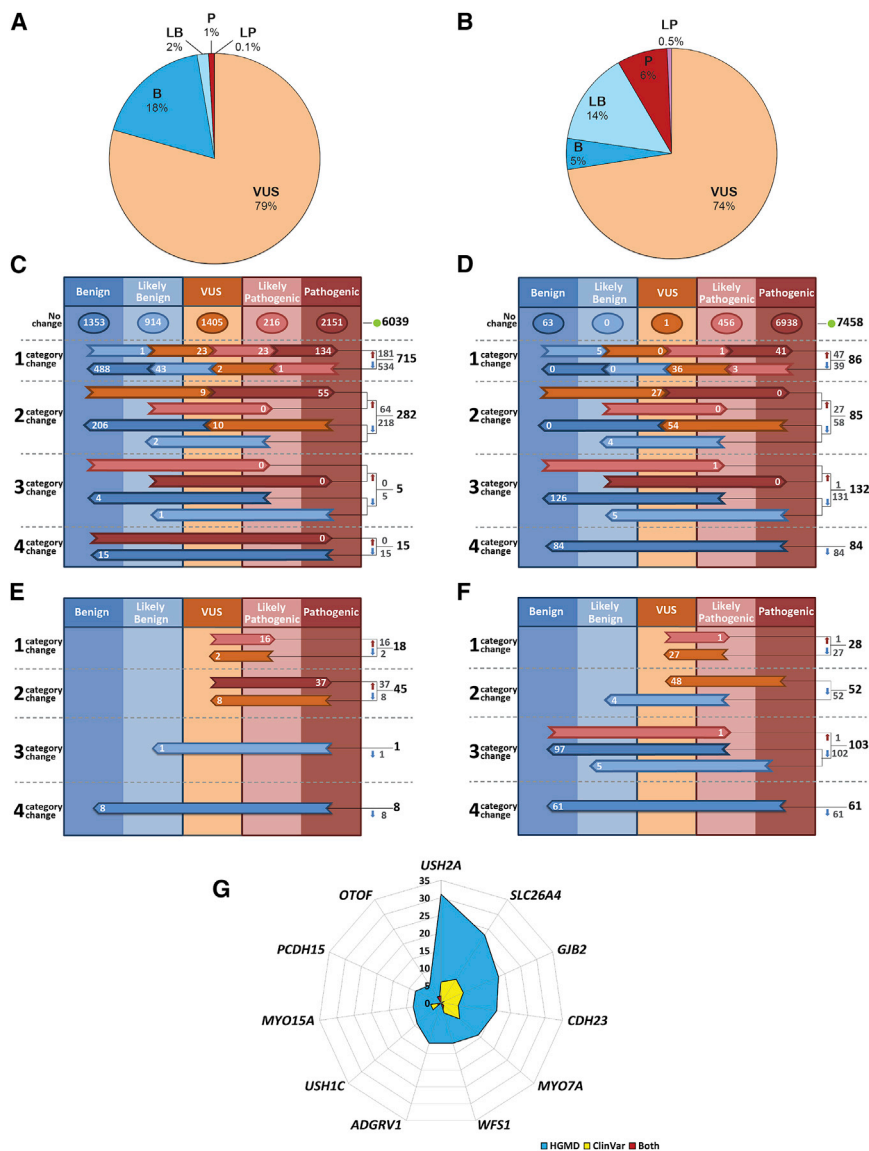


Figure 2. Variant Classification by the DVD

(A) Fractions of different classification categories for variants in the whole DVD.

(B) A slightly different picture emerges when only clinically relevant regions and deafness-associated variants (variants that were associated with other non-related deafness phenotypes are excluded) are considered.

(C) Comparative overview of DVD versus ClinVar. 7,056 classifications from ClinVar were identified within our specified gene regions (each variant in ClinVar with multiple submissions for pathogenicity has been represented by its most pathogenic submission). Of this number, 6,039 ClinVar classifications agreed with the corresponding DVD classification whereas there was disagreement for 1,017 variants.

(D) Comparative overview of DVD versus HGMD. 7,845 classifications from HGMD were identified within our specified gene regions. Of this number, 7,458 classifications agreed with the corresponding DVD classification and discrepancies were found for 387 variants.

(E) There were 72 major categorical changes between ClinVar and DVD that resulted in medically significant differences (53 up-classifications and 19 down-classifications).

(F) 244 medically significant reclassifications were found when DVD was compared to HGMD (2 up-classifications and 242 down-classifications).

(G) Of the 20% of genes carrying the greatest numbers of medically significant changes, 6 are implicated in Usher syndrome.

For (C) through (F), the horizontal arrows show discordant calls, with the number of discordant classifications shown within each arrow; totals are listed to the right of the colored columns.

from dbSNP, ExAC, 1000 Genomes, ESP, ClinVar, HGMD, and our internal manual curation database.^{8,23,24} All variants were annotated for MAF (from large-scale population databases), variant effect (intronic, UTR, splice-site, missense, nonsense, synonymous, inframe indels, frame-shift indels, start loss, stop loss), deleteriousness predictions (dbNSFP), and classification (ClinVar, HGMD, our internal manual curation) (Figure 1).

All available data were used to classify variants computationally, with supplemental expert manual curation as detailed in the Material and Methods (Figure 1). We integrated predictions from six algorithms—two assessing conservation (PhyloP and GERP++) and four evaluating deleteriousness (SIFT, PolyPhen-2, MutationTaster, and LRT)—to calculate a composite pathogenicity score (PS) and annotate variants with MAF < 0.5%. Variants with MAFs above this threshold were automatically classified as benign with the exception of known common founder

mutations (Figure 1C, Table S2).¹⁶ To validate the PS, we plotted all variants classified as pathogenic by MAF and PS (Figure S1) and found that of 3,591 pathogenic variants with predictions from at least five pathogenicity prediction tools, 95.4% have a composite PS \geq 60%. The calculated sensitivity, specificity, PPV, and NPV were 0.95, 0.51, 0.74, and 0.88, respectively. We used this threshold for variant classification, labeling variants with a MAF < 0.5 and a PS \leq 40%, based on at least five pathogenicity predictions, as LB.

In aggregate, DVD v.8.1 reports 876,139 variants from 152 genes and microRNAs. Of these variants, 7,502 (0.85%) are classified as P, 671 (0.077%) are LP, 15,287 (1.74%) are LB, 156,970 (17.9%) are B, and 695,709 (79.4%) are VUSs (Figure 2A). To assess only medically relevant variants for deafness, we considered *only* the 97,007 variants within coding and splice-site regions (exons as defined by RefSeq and Ensembl coding transcripts,

±20 bp from exon boundaries, 3' and 5' UTRs, and any deep intronic variant classified as P or LP) as these regions are routinely screened in clinical diagnostics settings. We also considered any variant that is P or LP for a phenotype *other than* deafness as a VUS for the purpose of this analysis. For example, 20 P/LP variants have been reported in *MET* (MIM: 164860), but only one has been linked to hearing loss. Of 97,007 variants we considered, 6.2% were P (6,045), 0.5% were LP (445), 14.2% were LB (13,823), 4.8% were B (4,628), and 74.3% were VUSs (72,066) (Figure 2B).

Computational and Expert Manual Curation Led to Medically Significant Changes in Pathogenicity

To assess differences in variant interpretation between the DVD and ClinVar and HGMD, we compared the number of downgraded (from more severe to more benign) and upgraded (from more benign to more severe) classifications (Figures 2C–2F). Of the variants listed in the DVD, 7,056 are found in ClinVar (filtered to represent each variant by only its most pathogenic classification). Of these variants, 175 are unique to ClinVar (Figure 1B). There was classification agreement for 6,039 (85.6%) variants. Of the 1,017 (14.4%) discordant calls, classification discrepancies of one degree were most common (715 of 1,017 changes), with the DVD being more likely to downgrade a ClinVar classification (772 downgrades versus 245 upgrades) (Figure 2C). Major classification changes for deafness-related variants that resulted in *medically significant differences* (variants that were upgraded to or downgraded from P/LP) were identified for 72 variants. Of these variants, there were 53 up-classifications of a variant by the DVD to P/LP and 19 down-classifications of a variant from P/LP (Figure 2E).

A total of 7,845 DVD variants are found in HGMD. DVD and HGMD classifications were concordant in 7,458 (95%) cases. Of the 387 (5%) discordant calls, classification discrepancies of three degrees were most common (132 of the 387 changes), with DVD downgrades of HGMD calls more common than upgrades (312 downgrades versus 75 upgrades) (Figure 2D). There were 244 major classification changes that resulted in medically significant differences, with all except two representing downgrades by the DVD from an HGMD call of P/LP (Figure 2F).

Following computational and manual curation, variants in 101 genes were reclassified in the DVD. These reclassifications included major categorical changes representing medically significant changes (P/LP versus VUS/LB/B) for 300 variants in 52 genes (Table S3). Of the 20% of genes carrying the greatest number of medically significant differences, six are associated with the diagnosis of Usher syndrome (Figure 2G, Table S4). For both ClinVar and HGMD, the same five genes carry the greatest number of major categorical changes (*USH2A* [MIM: 608400], *SLC26A4* [MIM: 605646], *GJB2*, *MYO7A* [MIM: 276903], *CDH23* [MIM: 605516]) (Figure 2G, Table S4). The remaining frequently impacted genes are *WFS1* (MIM: 606201) (DFNA6/14/38

[MIM: 600965] and Wolfram syndrome), *USH1C* (MIM: 605242) (DFNB18A [MIM: 602092] and *USH1C* [MIM: 276904]), *ADGRV1* (MIM: 602851) (*USH2C* [MIM: 605472]), and *MYO15A* (MIM: 602666) (DFNB3 [MIM: 600316]).

Most Genetic Variants in Deafness-Associated Genes Are Missense and Rare

Having built a comprehensive resource that collates and annotates all variants in hearing loss genes and provides a clinical interpretation, we sought to explore the genomic and mutational landscape of deafness-associated genes. Our first objective was to evaluate the distribution of variants with respect to their MAF and type. Of all variants in the DVD, novel, ultra-rare ($0% < \text{MAF} \leq 0.05%$), and rare ($0.05% < \text{MAF} < 0.5%$) variants represented 36%, 11%, and 35%, respectively (Figure 3A). When only clinically relevant variants within coding and splice-site regions were considered, the general tendency did not change. Variants with $\text{MAF} < 0.5%$ remained the most prevalent (96%) although the distribution within this set changed, with ultra-rare variants ($0% < \text{MAF} \leq 0.05%$) now representing the major category (59%) (Figure 3B). The finding that variants with a $\text{MAF} < 0.5%$ (the threshold above which a variant is too common to be deafness causing¹⁶) account for 96% of all the variants falling within coding and splice-site regions implies that only 4% of variants can be excluded as deafness causing on the basis of MAF filtering.

Of all variants within deafness-associated genes, ~12% were located in the coding regions and canonical splice sites (Figure 3C). Missense variants represent the major set of all coding variants at 62%. The second most common type are synonymous variants (28%) followed by indels (4% frameshift and 2% inframe), nonsense (2%), canonical splice-site (2%), and start/stop loss (<1%) (Figure 3D).

Disparity in Gene Variation Rates

As expected, the number of variants per gene correlated with gene size, with larger genes carrying higher numbers of variants (Figures 4A and S2A, Table S5). The greatest variant load was found in *PCDH15* (MIM: 605514), *USH2A*, *ADGRV1*, and *CDH23*, but when the number of variants was normalized for gene size, different trends emerged (Figures S2B and S2C, Table S5). *ACTG1* (MIM: 102560) had the highest variant rate at 41% (4 of 10 bases carry reported variants), with most genes (85%) having a variation rate below 10%. If we restricted the analysis to coding and splice-site regions, again there was a correlation between the number of variants and the size of the coding regions, with *USH2A*, *ADGRV1*, and *CDH23* carrying the highest number of variants (Table S5). Normalizing to the size of the coding region, however, gave strikingly different results: *GJB2* carried the greatest variation at ~69% (nearly 7 of 10 bases carry reported variants) and six other genes had variation rates higher than

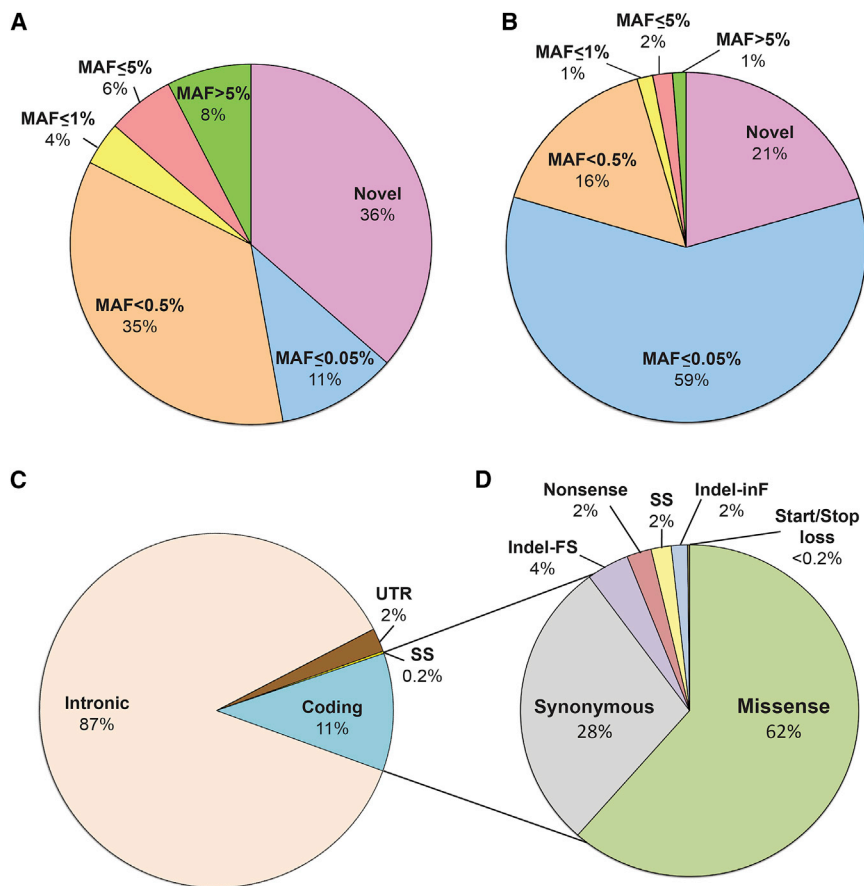


Figure 3. Distribution of Variants by Location, MAF, and Type

(A and B) MAF (all variants in DVD including intronic (A) and only variants in gene coding regions (B)). Most coding variants (96%) in deafness-associated genes are novel or rare (MAF < 0.5%).

(C) Distribution of variant by their gene location.

(D) Coding variant breakdown by type showing that missense variants constitute the major set of all coding variants.

Abbreviations: FS, frameshift; SS, splice-site; inF, in-frame.

30%: *WFS1* (53%), *KCNQ1* (MIM: 607542) (44%), *ACTG1* (39%), *SLC26A4* (37%), and *KCNE1* (MIM: 176261) (36%). The average variation rate was ~22% (Figures 4B and S3A).

To determine whether gene-specific variation rates correlated with tolerance or intolerance to variation, we focused on the 6,490 variants classified as P and LP for deafness and normalized to the total number of coding variants. We found that ~69% of coding variants in *GJB2* are disease causing (P and LP variants), meaning that for any new variant identified in the coding sequence of *GJB2*, there is a 70% chance that it is pathogenic. Both *COL4A5* (MIM: 303630) (55.3%) and *SLC26A4* (47.2%) also had high (P+LP)/(Coding Variant) ratios (Figures 4C and S3B, Table S5).

Variants Are Differentially Distributed across Classifications

To characterize the molecular profile of variants within different classification categories, we focused on variants in coding and splice regions and grouped them by type (nonsense, splice-site, frameshift indels, start loss, stop loss, in-frame indels, missense, UTRs, intronic, synonymous) across variant classifications (P, LP, VUS, LB, B). Overall, missense variants were most prevalent in all categories (Figure 5A). For P variants, loss-of-function (LoF) variants and non-LoF were equally represented (~50%). Of

LoF variants, frameshift indels were most common (47.8%), followed by nonsense and splice-site at 27.65% and 22.27%, respectively. LP variants showed a slightly different profile with mostly missense variants at ~70%. As expected, P variants are enriched in LoF and B variants are depleted. VUSs are enriched for missense (53.5%) and synonymous (34.1%) variants, with LoF variants representing only ~7%. We found an enrichment of LoF variants in the P (50%) and LP (16.9%) categories, whereas they represent only 7% of VUSs and a negligible proportion of

LB (0.03%) and B (0.47%) variants (Figure 5A). Missense variants are most common in the LB classification at ~95% and represent ~70% of the LP variants and ~50% of P variants.

Diverse Mutational Spectrum across Deafness-Associated Genes

We next examined the distribution of LoFs and missense and synonymous variants by gene and observed disparity among genes as some are depleted of LoF (such as *SIX1* [MIM: 601205]) whereas others are enriched in synonymous (such as *ACTG1*) or missense (such as *ADGRV1*) variants (Figures 5B and S4A). Of all LoF variants, the fraction contributing to the P/LP pool differs across genes, showing that for some genes a LoF variant is most likely to be pathogenic (*SOX10* [MIM: 602229], *TCOF1* [MIM: 606847], *COL2A1* [MIM: 120140], *COL4A5* [MIM: 303630], *EYA1* [MIM: 601653], *GATA3* [MIM: 131320], *POU3F4* [MIM: 300039]), whereas for others it is not (e.g., *ACTG1*, *AIFM1* [MIM: 300169]) (Figures 5C and S4B). A similar disparity is also observed for missense variants, where for some genes more than half of all missense variants are P/LP (*GJB2*, *KCNQ1*, *PRPS1* [MIM: 311850]) and for others this contribution is marginal (e.g., *TRIOBP* [MIM: 609761], *ADGRV1*) (Figures 5D and S4C). Interestingly, for some genes such as *BSND* (MIM: 606412), *TCOF1*, and *TRIOBP*, approximately half of all missense

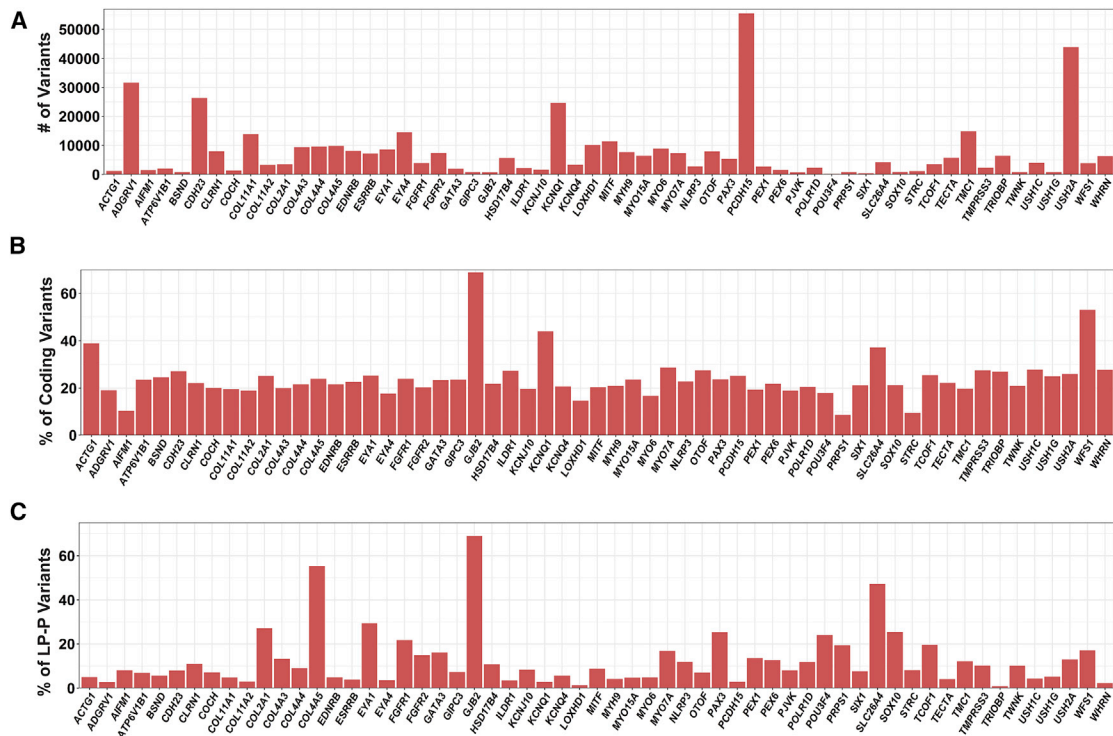


Figure 4. Variation Rate for Deafness-Associated Genes

(A) Total number of variants per gene.

(B) Normalized number of coding variants based on the size of the coding and splice regions.

(C) Normalized number of deafness-associated variants (P+LP) based on the total number of coding variants.

Only genes with ≥ 14 reported deafness-associated variants are included in this figure; the remaining genes are shown in Figures S2 and S3.

variants are classified as B/LB, implying that a missense variant in those genes is more likely to be non-disease causing.

We also noted wide variation across genes in the fractional contribution of missense versus LoF variants to the P/LP category (Figures 5E and S4D). Some genes have exclusively missense mutations (*ACTG1*, *PRPS1*, *COCH* [MIM: 603196], and *AIFM1*) while other genes were enriched in LoF mutations (*TCOF1*, *LOXHD1* [MIM: 613072], *ADGRV1*, *EYA1*, and *PCDH15*). A more detailed analysis of the different types of mutations within the LoF group revealed greater variability in the fractions of nonsense, splice-site, and frameshift indels across genes (Figures 5E and S4D). For example, the majority of LoF mutations in *LOXHD1* are nonsense, whereas for *COL11A1* (MIM: 120280) they are splice sites.

MAF Thresholds for Disease-Causing Variants Are Gene Specific

Gene-specific MAF thresholds for P+LP variants ranged from 0% to 7.34%. *GJB2*, *MYO15A*, *OTOF* (MIM: 603681), *PEX6* (MIM: 601498), and *CLRN1* (MIM: 601498) had the highest MAFs at 7.34%, 2.45%, 0.79%, 0.71%, and 0.69%, respectively (Table S6). However, these maximum MAFs are misleading and do not provide an accurate MAF for the majority of disease-causing variants associated with these genes. For example, while the

maximum MAF for any pathogenic variant reported in *GJB2* (GenBank: NM_004004.5) is 7.3% for c.109G>A (p.Val37Ile), the median MAF for all mutations in *GJB2* is surprisingly 0, reflecting the huge number of ultra-rare P+LP variants in this gene (Figures 6A, 6B, and S5, Table S6). Similar results were found for *SLC26A4*, *USH2A*, and *WFS1*. These discrepancies also reflect founder effects as some mutations occur solely in a single population or ethnicity and account for a large portion of that population's hearing loss (Table S2). These critical exceptions emphasize the importance of expert curation and review of variants that exceed the 0.5% MAF cut-off.

MAF Thresholds for Disease-Causing Variants Are Type Specific

To determine whether MAFs for P+LP variants were mutation-type dependent, we subdivided all variants by effect and plotted against their MAF. Although the median MAF is 0 for all variant types, synonymous and UTR variants had the highest mean MAF (0.023% and 0.027%, respectively), followed by missense (0.017%), nonsense (0.009%), splice-site (0.0047%), in-frame indels (0.0036%), and frameshift indels (0.0028%) (Figure 6C, Table S7). These results compare closely to the gene-level results and demonstrate that regardless of type and gene, disease-causing mutations are ultra-rare and are heavily comprised of novel/private variants.

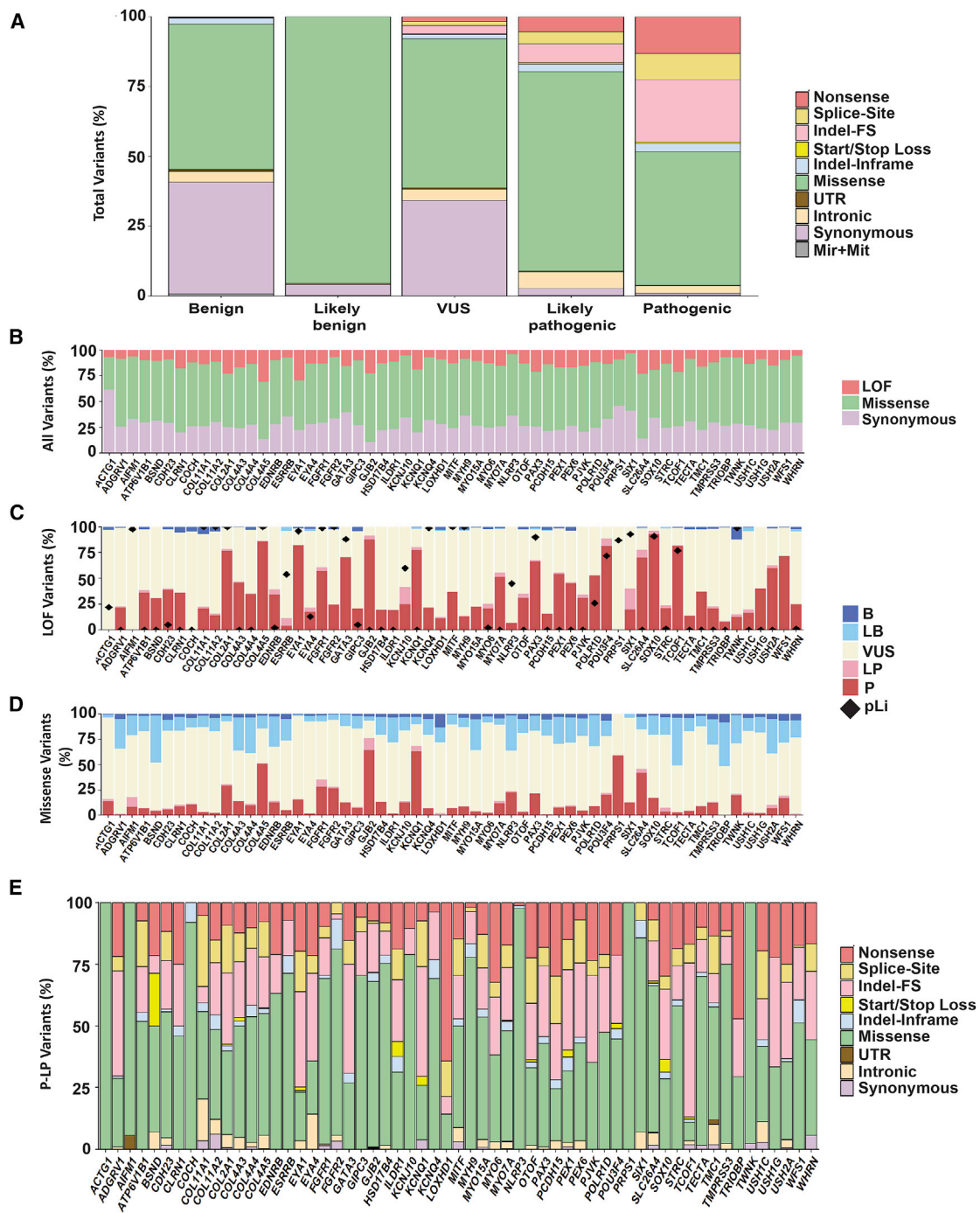


Figure 5. Genomic Landscape of Deafness-Associated Genes

(A) Variant architecture by each classification category shows a strikingly distinct distribution of variant types across the five classifications.

(B) Distribution of LoF, missense, and synonymous variants is different across genes.

(C) Most LoF variants are P/LP and some genes are highly enriched in this type of variant.

(D) The contribution of missense variants to the mutational pool of hearing loss is variable across genes. However, in most genes, the majority of missense variants are VUSs.

(E) The mutational spectrum is gene specific. Splice-site indicates variants in canonical splice sites.

Only genes with ≥ 14 reported deafness-associated variants are included in this figure; the remaining genes are shown in [Figures S4](#) and [S5](#).

Kafeen and the DVD Are Configurable, Customizable, and Open-Access Resources

The DVD is freely available. It is widely used by the scientific and clinical communities worldwide with $\sim 3,000$

users and 13,000 sessions over the past 12 months ([Figure S6](#)). The Kafeen bioinformatic pipeline, upon which the DVD was built, is configurable, adaptable, and extensible, allowing incorporation of additional variant

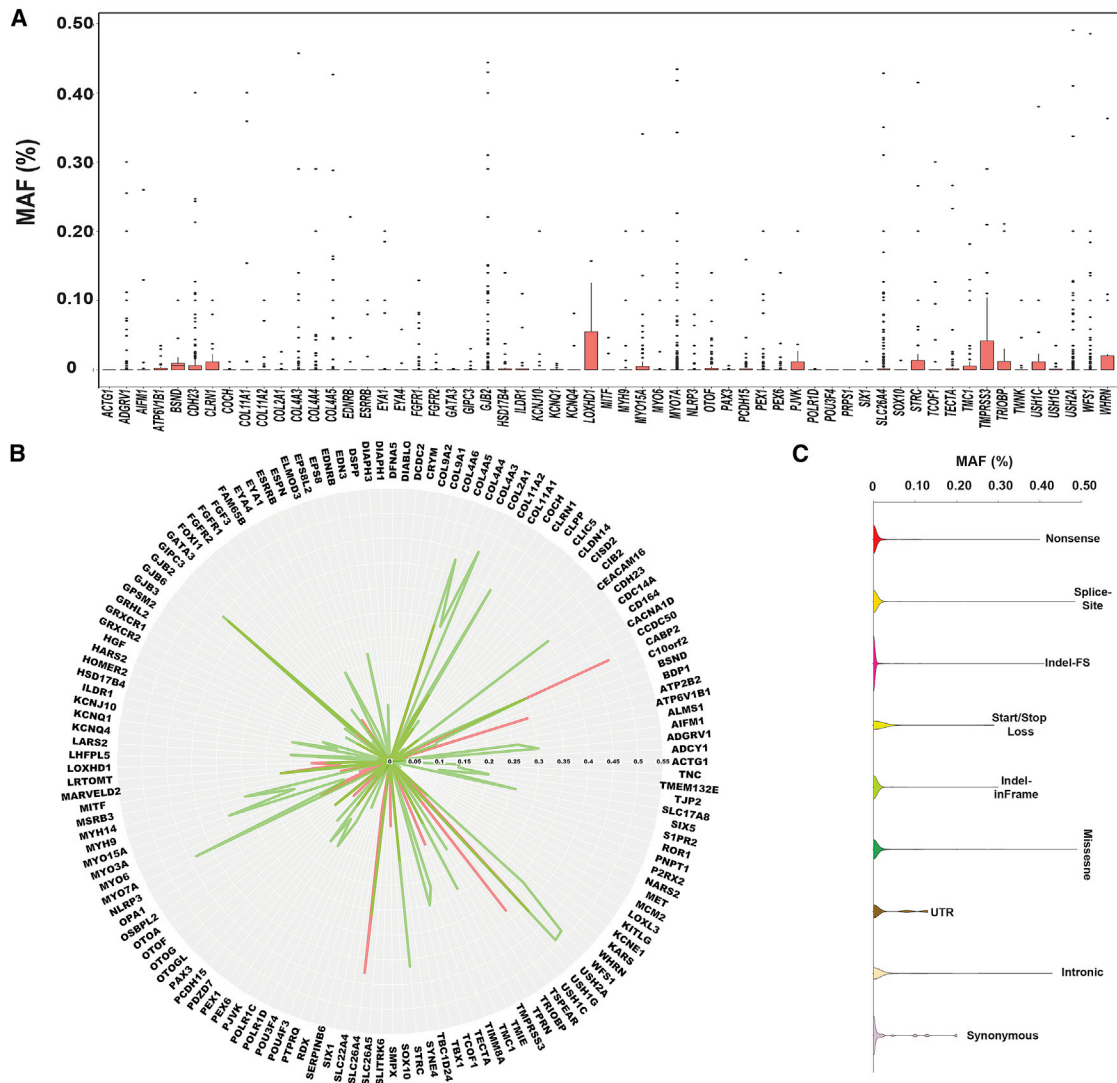


Figure 6. MAFs Thresholds for Deafness-Associated Variants Are Gene and Type Specific

(A) Plot of MAFs of all P/LP variants in each deafness-associated gene.

(B) Maximum MAF is gene specific and there is a clear distinction between LoF versus missense variants.

(C) Overall, missense variants exhibit the highest MAFs when compared to all other variants.

Only genes with ≥ 14 reported deafness-associated variants are included in this figure; the remaining genes are shown in [Figure S5](#).

and annotation sources as well as deleteriousness prediction tools. It also allows for customizable thresholds of MAF to classify variants. Consequently, its use is not limited to deafness and could be implemented for a variety of other genetic disorders.

Discussion

Genetic variant classification is crucial to accurate genetic diagnoses and represents a major challenge in the post-genome era, particularly for a disorder with genetic and phenotypic heterogeneity like deafness. The DVD was designed as a deafness-specific, comprehensive, open-access database that collates and summarizes all available data in addition to providing expert curation of genetic variants implicated in deafness ([Figure 1](#)). We integrate its use into a

weekly multidisciplinary conference where a person's genotypic data are reviewed in the context of available phenotypic data to provide expert contextual interpretation of the genetic results. As a first step, DVD annotations are used for prioritization of a person's variant list, automatically flagging variants known to be reported as pathogenic in the DVD, as well as retaining DVD-classified LP/P variants that may have been filtered out of our NGS processing pipeline due to poor quality or ambiguous mapping. This type of curation reduces false negative rates and highlights the importance of disease-specific knowledge and disease-specific databases. CNVs are not integrated in the current release of the DVD. We have shown previously that they are major contributor to hearing loss and are implicated in $\sim 18\%$ of all positive diagnoses.²⁵ The challenge to their incorporation in the DVD resides in the lack of data regarding their exact breakpoint

junctions. As more data become available, integration of CNVs should be an integral part of any variant database.

While it is difficult to provide “universal” MAF thresholds, the ACMG does recommend using MAF data as a key filter in their guidelines for variant interpretation. We deem a MAF $\geq 0.5\%$ to be incompatible with a classification of P/LP for hereditary hearing loss aside from specific cases such as variants in *GJB2* and *SLC26A4*, and we use this threshold to automatically classify any variant as B (Table S2).¹⁶ It is important to note that with the availability of new datasets from large-population sequencing projects such as gnomAD, MAF for some variants will change, which in turn may affect their clinical significance. While using a universal MAF cutoff is beneficial, for a common disease such as deafness this filter aided in classifying only 4% of coding variants as benign, illustrating that MAF cutoffs and rarity alone are not sufficient to determine deleteriousness.

As an additional aid, the DVD integrates predictions from six algorithms—two assessing conservation (PhyloP and GERP++) and four evaluating deleteriousness (SIFT, PolyPhen-2, MutationTaster, and LRT)—from which to calculate a composite PS. As more than 95% of known P variants have a pathogenicity score $> 40\%$, the PPV of this approach reaches 0.995 (Figure S1). Using this threshold, we classify variants as either VUS (PS $\geq 60\%$) or LB (PS $\leq 40\%$). ACMG guidelines also endorse predictions from *in silico* algorithms as one of the eight evidence criteria recommended for variant clinical interpretation, and although outcomes and results from several studies vary depending on the algorithms used, these studies all agree on the utility of such tools for improving accuracy and reducing VUS burden in clinical diagnosis.^{3,26}

Questions remain regarding the strength and amount of evidence needed to sway a classification from a VUS to P/LP or B/LB. Since in clinical settings substantial evidence is needed to reach a P/LP classification, we have opted to use *in silico* algorithms exclusively to shift a VUS classification to LB.²⁷ We require additional evidence (genotype, phenotype, family history, segregation, and functional studies) to upgrade a VUS to a P or LP variant.

Discrepancies in variant classification between the DVD versus ClinVar and HGMD were observed at 14.5% and 5%, respectively (Figure 2). Differences were due in part to the misclassification of B and LB variants as P or LP and have been reported in other studies highlighting the limitations of ClinVar and HGMD.^{28–30} ClinVar is based on submissions from researchers and clinical diagnostic laboratories. It is an invaluable resource that creates an open platform for sharing genetic data and variant interpretation, but it has some disadvantages. Most obvious are the differences in the methods used to detect, validate, curate, and derive variant interpretation, which understandably vary between groups and thus can lead to conflicting classifications.^{27,31–33} Unlike ClinVar, HGMD relies on published literature and is primarily a disease-causing

focused variant database. Although the variants reported in HGMD have been published and therefore have undergone peer review, the HGMD curation process is error prone due to the potential for subjective misinterpretation of the literature and a lack of disease-specific experts reviewing the material.

We implemented major categorical reclassifications that led to medically significant changes in 52 genes. In 33 genes, the change affected three or fewer variants (20 genes, 1 variant change; 11 genes, 2 variant changes; 2 genes, 3 variant changes); however, of the top 20% of genes carrying the greatest number of reclassifications, six cause Usher syndrome (*ADGRV1* [USH2C], *CDH23* [USH1D and DFNB12], *MYO7A* [USH1B, DFNB2, and DFNA11], *PCDH15* [USH1F and DFNB23], *USH2A* [USH2A], and *USH1C* [USH1C and DFNB18A]) (Figure 2G, Table S4). Differentiating USH1 from NSHL is possible if a directed developmental history is obtained, because sitting and walking milestones are significantly delayed in USH1 due to the associated vestibular dysfunction, emphasizing the need to correlate clinical history with the interpretation of genetic data. We also consider audioprofiles, noting any progression of hearing loss, age at diagnosis and symmetry; imaging studies if available; and family history, as it is often possible to refine a diagnosis when more clinical information is provided. For example, a genetic diagnosis consistent with either USH1C or DFNB18A would be changed to USH1C if the child had delayed developmental milestones.

Recognizing the importance of more stringent filtering strategies to improve variant classification prompted us to use the DVD to define the molecular landscape of deafness-associated genes. When normalized to genomic size, some genes show remarkably high variation rates, such as *ACTG1*, although for the majority of genes the variation rate is below 10% (Figures 4B and S2B). This trend changes dramatically when only *clinically relevant* regions (coding and splice regions) are considered, implying that most variation is intronic. The coding/splice-site variation rate is highest for *GJB2* (~69%) and ranges from 8.5% to 53% for all other genes (Figures 4B and S2B, Table S5). Other studies, notably by Petrovski et al.,³⁴ Lek et al.,¹² and Samocho et al.,³⁵ have used population-scale databases of variant numbers and allele frequencies to infer gene constraint or tolerance to genetic variation. Their assumption is that genes carrying more variants than expected have low constraint, while those with lesser variants have higher constraint and are intolerant to genetic variation. Our data showed that *GJB2* does not fit into this model. Although it has the highest variation rate, it also carries the highest fraction of pathogenic variants (Figure 4C). This observation contrasts with its z-score of -1.07 (ExAC), which implies tolerance to variation and decreased constraint (Table S5). Similar findings are seen for *SLC26A4*, where every other variant is disease causing although its Z score is -3.23 . These findings highlight the need to integrate real variant clinical

interpretation data for each gene-phenotype association as large-scale population data can be misleading.

Several studies have also emphasized the importance of moving from gene-wide constraint calculations to protein domain-specific constraints as a method of identifying regions of functional importance.^{35–38} This refinement is particularly important for proteins involved in hearing loss, as most have various structurally different domains with distinct functions. Furthermore, some show an extraordinary pleiotropy and cause both autosomal-dominant NSHL (ADNSHL) and autosomal-recessive NSHL (ARNSHL) (*TECTA* [MIM: 602574] and *TMC1* [MIM: 606706]) or both syndromic hearing loss and NSHL (Usher type 1-associated genes, *WFS1*, *TBC1D24* [MIM: 613577], and *COL11A1* [MIM: 120290]).^{39–44} Classifying variants by domain or regional constraint can minimize both false-positive and false-negative pathogenicity predictions and facilitate proper diagnosis, especially for genes associated with NSHL mimics.

Our assessment of variant distribution by mutation type, classification, and gene-specific MAFs across all 152 genes and microRNAs uncovered gene-specific variant architecture (Figures 5, 6, S4, and S5). For example, some genes (*GJB2*, *SLC26A4*, and *COL4A5*) are relatively depleted of synonymous changes when compared to other genes (Figure 5B). Interestingly, these same genes possess the highest intolerance to variation with 70%, 55%, and 47% of all coding variants being P/LP for *GJB2*, *COL4A5*, and *SLC26A4*, respectively (Figure 4C, Table S5). The involvement of synonymous variants in disease is secondary to splice alteration by changing exonic splice enhancers or silencers, or through codon usage bias that impacts gene expression by affecting mRNA folding and stability, messenger ribonucleoprotein (mRNP) complex formation, translation rate, and protein folding and function.^{45,46} Synonymous variants may be under selective pressure in *GJB2*, *COL4A5*, and *SLC26A4*, implying a potential unrecognized disease mechanism that would affect their proper expression in inner ear. This highlights the need to carefully review these variants when interpreting sequence data from persons with hearing loss. Conversely, for other genes like *ACTG1*, synonymous variants predominate, while the only reported deafness-associated pathogenic variants are missense, suggesting intolerance to changes at the protein level, which is in line with the reported gain-of-function mechanism for these variants.

Overall, there was a great diversity in the contribution of LoF and missense variants to the mutational load across genes (Figures 5C–5E and S4). Of all the LoF variants, the fraction contributing to the P/LP group was highest for genes for which haploinsufficiency is the mechanism of action (autosomal-dominant and X-linked genes) such as *EYA1*, *TCOF1*, *SOX10*, and *GATA3* (Figure 5C). This trend was further accentuated when assessing the contribution of LoF variants to the mutational spectrum of these genes (Figure 5E). For example, LoF mutations in *TCOF1* and

EYA1 represent ~90% and 80% of all reported pathogenic variants, respectively.

The variability in the fractional contribution of nonsense, splice-site, and frameshift indels to the mutational load across genes is intriguing. Although for autosomal-recessive genes this variability may not affect the outcome at the protein level (as most of these variants are expected to result in null alleles), the story is quite different for autosomal-dominant genes. The latter exert their effect via haploinsufficiency or a gain-of-function/dominant-negative mode of action, and the specific type of mutation might be crucial. LoF variants in genes known to have a dominant-negative/gain-of-function mechanism of action are not traditionally predicted to be pathogenic for a dominant disease. However, this caveat ignores the position-dependent effect of these variants on nonsense-mediated mRNA decay (NMD).⁴⁷ For example, truncating pathogenic variants in *DIAPH1* are linked to two different phenotypes: (1) autosomal-recessive seizures, cortical blindness, and microcephaly syndrome (MIM: 616632) due to null alleles through NMD and (2) autosomal-dominant DFNA1 hearing loss with thrombocytopenia due to gain-of-function truncating variants (that escape NMD) in the C-terminal DAD domain, which disrupt the autoinhibitory activity of the DAD and renders the protein constitutively active.^{48,49} *SOX10* and *PTPRQ* are other examples where the impact of a LoF variant is position dependent.^{50,51}

COL11A1 has the largest proportion of splice-site pathogenic variants when compared to all other deafness-associated genes. The majority of these variants are located in the triple-helical domain and cause inframe exon skipping rather than frameshifts. The mutant proteins exert their effect through a dominant-negative mechanism to cause Marshall syndrome (MRSHS) and Stickler syndrome type II (STL2).^{52,53} However, biallelic null alleles cause fibrochondrogenesis (FBCG1), a severe recessive often neonatally lethal disease.⁵⁴ This genotype-phenotype correlation explains the enrichment of splice-altering disease variants in *COL11A1*.

For genes where most missense variants are classified as B/LB, we estimate that a majority of the variants that are currently classified as VUS will be subsequently downgraded to B/LB. Similar to the diversity of variant distribution across classifications, we exposed clear distinctions in the maximum MAFs of P/LP variants depending on the gene and variant type (LoF versus missense) (Figures 6 and S5).

The emerging global picture from our findings is an intricate and complex portrait of the genomic landscape and mutational signature of deafness-associated genes. Although this work lays the foundation for improved variant interpretation, which greatly enhances clinical decision making, significant challenges remain. For example, of coding variants with a MAF < 0.5%, missense variants predominate. They constitute 70% of all VUSs and their accurate reclassification will require

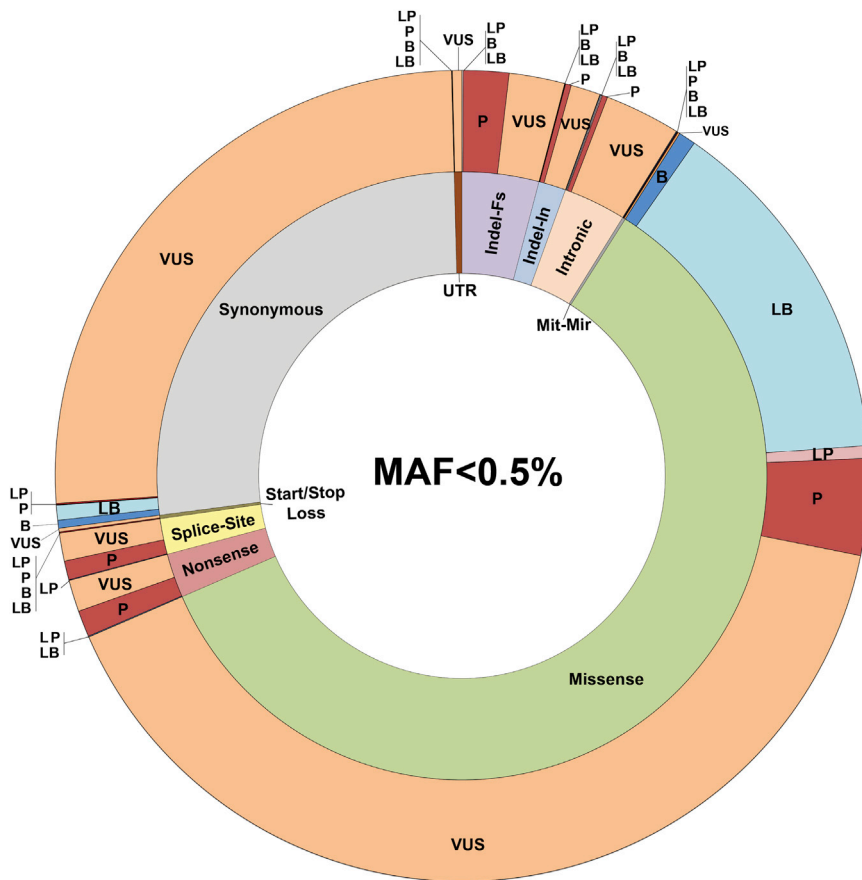


Figure 7. The Challenge of VUSs
 Variant architecture correlating variant type (inner ring) and clinical significance (outer ring) for variants with MAF less than 0.5% and located within the clinically relevant regions. Of all coding variants with MAF < 0.5%, missense variants represent the majority at 61.5%; of these missense variants, 70% are classified as VUSs. Abbreviations: Indel-In, in-frame indel; Indel-Fs, frameshift indel; Mit-Mir, mitochondrial and microRNA.

better computational tools (Figure 7). The non-coding pathogenic landscape also must be defined, warranting coordinated studies to integrate expression and genomic data.

In summary, using decision support tools and human expert curation, we have developed an integrated approach to facilitate the application of comprehensive genetic testing to the clinical care of persons with hearing loss. We believe that detailed disease-specific knowledge of the genomic landscape is requisite to establish a framework for variant interpretation and show that there are gene-specific mutational signatures, the knowledge of which will refine guidelines for variant interpretation for deafness and advance our understanding of disease biology. This resource is freely available to the public and configurable to allow its implementation for any Mendelian genetic disorder.

Supplemental Data

Supplemental Data include six figures and seven tables and can be found with this article online at <https://doi.org/10.1016/j.ajhg.2018.08.006>.

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Declaration of Interests

R.J.H.S. directs the Molecular Otolaryngology and Renal Research Laboratories (MORL) which developed and offers comprehensive genetic testing for persons with hearing loss.

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Web Resources

1000 Genomes, <http://www.internationalgenome.org/>
 ClinVar, <https://www.ncbi.nlm.nih.gov/clinvar/>
 DBNSFP, <https://sites.google.com/site/jpopgen/dbNSFP>
 dbSNP, <https://www.ncbi.nlm.nih.gov/projects/SNP/>
 Deafness Variation Database, <http://deafnessvariationdatabase.com/>
 ExAC Browser, <http://exac.broadinstitute.org/>
 GenBank, <https://www.ncbi.nlm.nih.gov/genbank/>
 gnomAD Browser, <http://gnomad.broadinstitute.org/>
 Hereditary Hearing Loss Homepage, <http://hereditaryhearingloss.org>
 Human Gene Mutation Database (HGMD), <https://www.qiagenbioinformatics.com/products/human-gene-mutation-database/>
 Kafeen, <https://github.com/clcg/Kafeen>

MutationTaster, <http://www.mutationtaster.org/>
NHLBI Exome Sequencing Project (ESP) Exome Variant Server,
<http://evs.gs.washington.edu/EVS/>
OMIM, <http://www.omim.org/>
PolyPhen-2, <http://genetics.bwh.harvard.edu/pph2/>
RefSeq, <https://www.ncbi.nlm.nih.gov/RefSeq>
Ruby programming language, <https://www.ruby-lang.org/>
SIFT, <http://sift.bii.a-star.edu.sg/>

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