

Genome-wide association meta-analyses of drug-resistant epilepsy

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Summary

Background: Epilepsy is one of the most common neurological disorders, affecting over 50 million people worldwide. One-third of people with epilepsy do not respond to currently available anti-seizure medications, constituting one of the most important problems in epilepsy. Little is known about the molecular pathology of drug resistance in epilepsy, in particular, possible underlying genetic factors are largely unknown.

Methods: We performed a genome-wide association study (GWAS) in two epilepsy cohorts of European ancestry, comparing drug-resistant (N=4,208) to drug-responsive individuals (N=2,618) followed by meta-analyses across the studies. Next, we performed subanalyses split into two broad subtypes: acquired or non-acquired focal and genetic generalized epilepsy.

Findings: Our drug-resistant versus drug-responsive epilepsy GWAS meta-analysis showed no significant loci when combining all epilepsy types. Sub-analyses on individuals with focal epilepsy (FE) identified a significant locus on chromosome 1q42.11-q42.12 (lead SNP: rs35915186, $P=1.51 \times 10^{-8}$, OR[C]=0.74). This locus was not associated with any epilepsy subtype in the latest epilepsy GWAS (lowest uncorrected $P=0.009$ for FE vs healthy controls), and drug resistance in FE was not genetically correlated with susceptibility to FE itself. Seven genome-wide significant SNPs within this locus, encompassing the genes *CNIH4*, *WDR26*, and *CNIH3*, were identified to protect against drug-resistant FE. Further transcriptome-wide association studies (TWAS) imply significantly higher expression levels of *CNIH3* and *WDR26* in drug-resistant FE than in drug-responsive FE. *CNIH3* is implicated in AMPA receptor assembly and function, while *WDR26* haploinsufficiency is linked to intellectual disability and seizures. These findings suggest that *CNIH3* and *WDR26* may play a role in mediating drug response in focal epilepsy.

Interpretation: We identified a contribution of common genetic variation to drug-resistant focal epilepsy. These findings provide insights into possible mechanisms underlying drug response variability in epilepsy, offering potential targets for personalised treatment approaches.

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Research in context

Evidence before this study

The causes of drug resistance in epilepsy are not well understood, leading to a stagnation of drug therapy in epilepsy in the last 40 years. We searched PubMed with the terms 1.) “epilepsy” and “common variants”, 2.) “seizure outcomes” and “genetics”, 3.) “drug-resistant epilepsy” and “genetics” OR “association” OR “GWAS” for reports published before 1st July 2024, with no language restrictions. While rare genetic variants have been established as causal factors in epilepsy, and evidence suggests their potential overlap with drug resistance, this primarily applies to rare/monogenic epilepsy syndromes. These syndromes represent only a small fraction of all epilepsy cases. The majority of epilepsy cases exhibit a complex/polygenic genetic architecture, well-characterized by numerous successful genome-wide association and polygenic risk-scoring studies. However, to date, no study has successfully identified genome-wide significant common genetic factors influencing drug response in all forms of epilepsy.

Added value of this study

Prior studies, although inconclusive, suggested the involvement of common genetic variants in drug response and a potential heritable component to drug resistance in epilepsy. This study provides evidence for common genetic variants associated with drug response in focal epilepsy, confirming these earlier suggestions. To investigate the genetic basis of drug resistance, we leveraged data from two large-scale initiatives: EpiPGX, an international multicenter research project on epilepsy pharmacogenetics, and Epi25, the largest sequencing study in epilepsy. In the combined cohort of 6,826 individuals with drug-resistant and -responsive epilepsy, we identified a locus on chromosome 1q42.11-q42.12, encompassing the genes *CNIH4*, *WDR26*, and *CNIH3*, associated with protection against drug resistance in focal epilepsy. Additionally, we observed

significantly higher predicted expression levels of *CNIH3* and *WDR26* in individuals with drug-resistant focal epilepsy compared to those with drug-responsive focal epilepsy.

Implications of all the available evidence

The present study provides two key insights into understanding drug resistance in epilepsy. First, we demonstrate that drug resistance in focal epilepsy has a common genetic component, which may enable quantification of each individual's polygenic risk for drug resistance in (focal) epilepsy and, thus, inform treatment strategies. The common genetic basis of drug resistance also suggests a future need to target multiple pathways rather than single molecules/genes. Second, fine-mapping of the association signal for drug response in focal epilepsy implicates three candidate genes: *CNIH4*, *WDR26*, and *CNIH3*. Pathogenic variants in *WDR26* have been shown to cause a drug-responsive seizure phenotype consistent with the protective effect observed in our meta-analysis and the higher expression levels in drug-resistant cases suggested by our transcriptome-wide association study. *CNIH3* acts as an auxiliary subunit that regulates AMPA receptor gating and trafficking, and abnormal AMPA receptor trafficking could contribute to seizure activity. The findings of this study provide a foundation for future research exploring the common genetic origins of drug resistance in epilepsy.

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Introduction

Epilepsy is a burdensome neurological disorder affecting over 50 million people worldwide.¹ One-third of people with epilepsy experience ongoing seizures despite treatment with appropriate antiseizure medications (ASMs). The standard operational definition of drug resistance in epilepsy, formulated by the International League Against Epilepsy (ILAE), is “failure of adequate trials of two tolerated, appropriately chosen and used ASM schedules (whether as monotherapies or in combination) to achieve sustained seizure freedom”.² Drug-resistant epilepsy (DRE) is associated with reduced quality of life, treatment side effects, comorbidities, lowered socioeconomic status, stigmatisation, and premature mortality.^{3–9} Despite the availability of more than 25 registered ASMs, the proportion of people with DRE has remained steady over time.¹⁰ A single-centre 30-year longitudinal cohort study found a similar proportion of people with DRE over the study period despite a marked increase in the use of newer ASMs.¹¹

The causes of DRE are unknown. Evidence suggests the existence of general mechanisms of drug resistance that act regardless of epilepsy syndrome or specific drug.¹² Several hypotheses have arisen as putative explanations for DRE, including the target,¹³ multidrug transporter,¹⁴ intrinsic severity,¹⁵ epigenetic,¹⁶ network,¹⁷ and others.¹⁸ However, evidence for these hypotheses remains limited.¹² While genetic factors have been suspected to play a role in drug resistance, definitive evidence has been limited. Only one epilepsy GWAS on drug response has been published (N=889), which did not find any genome-wide significant loci.¹⁹ More recently, a familial aggregation of a history of uncontrolled seizures (≥ 4 tonic-clonic seizures per year) was demonstrated, suggesting a genetic component of seizure outcomes.²⁰

We hypothesised a common-variant genetic component to DRE. We performed genome-wide SNP-based association studies (GWAS) in two independent international epilepsy cohorts (EpiPGX and Epi25) with drug response phenotypes, followed by meta-analyses. Given existing evidence that focal and generalised epilepsies have distinctive biologies and that DRE is more common in focal than generalised epilepsies, we hypothesised that any genetic basis for DRE would differ between these two categories²¹ and performed subanalyses in focal and generalised epilepsies.

Methods

Ethics

All individuals from the EpiPGX Consortium and Epi25 Collaborative gave written informed consent. Each centre's ethics committees/institutional review boards approved data collection and use. For the EpiPGX consortium, all participants provided written informed consent for appropriately coded use of their clinical data. Ethical approval for this study was obtained from the Camden and Kings Cross Research Ethics Committee (reference number: 11/LO/2016). Consent from parents or legal guardians was obtained from those unable to consent. For the Epi25 cohort, patients or their legal guardians provided signed informed consent/assent according to local IRB requirements;²² as samples had been collected over 20 years in some centres, forms reflected standards at the time of collection. For Epi25 Collaborative samples collected after 25th January 2015, forms required specific language according to the NIH Genomic Data Sharing Policy.²³

Study cohorts

Individuals were recruited from EpiPGX, an international multicenter research project on epilepsy pharmacogenetics, and Epi25, the largest sequencing study in epilepsy.²²

The EpiPGX database contains coded demographic and clinical details of about 10,000 individuals with a diagnosis of epilepsy confirmed by an epilepsy specialist. The database includes detailed data on >39,000 treatment regimens collected retrospectively from contemporary records. Participants were recruited mainly from tertiary referral centres in the UK, Ireland, Belgium, the Netherlands, Germany, and Italy. Data collection spanned from 2012-2016. All individuals were classified for treatment response, following a modification of the International League Against Epilepsy (ILAE) definition² of DRE. According to the ILAE definition of DRE, individuals with very rare seizures (for example, one seizure in 12 months) may be classified as drug-resistant²⁴ and preclude the identification of clinically meaningful DRE phenotypes. Therefore, this study adopted a threshold of four or more seizures per year, consistent with established practice in pharmacogenetic and pharmacogenomic investigations. This modified DRE definition was: "seizures occurring at a frequency of ≥ 4 /year during the year preceding the latest data entry, despite adequate trials of ≥ 2 tolerated and appropriately chosen (and used) ASM schedules, whether as monotherapies or in combination." It is important to note that the ILAE study advises adaptation of the definitions for particular circumstances and studies. Given that the phenotypic data for this study were collected retrospectively and that pre-intervention inter-seizure intervals were not consistently documented, drug-responsive epilepsy was defined as freedom from seizures for ≥ 12

months up to the latest recorded visit.¹¹ Consequently, individuals with 1-3 seizures in the 12 months preceding the latest data entry were excluded from the study. This usage is within the ILAE definition,² which categorises a treatment outcome as "seizure-free" (Category 1 response) if "the treatment results in seizure freedom for 12 months, or for a minimum of three times the longest pre-intervention inter-seizure interval, whichever is longer".² Our usage aligns with the seizure-free interval that often actually leads to changes in daily life (e.g., permitting reinstatement of driving privileges) and ensures that those who are considered drug-responsive have experienced a seizure-free interval of at least 12 months. Of note, none of the individuals classified as drug-responsive were seizure-free without medication (Table 1; an average of 1.9 adequate ASM trials). An AED trial was considered adequate if administered at an appropriate dose for a sufficient duration. Appropriateness was determined by prior evidence of efficacy, ideally from randomised controlled trials. Minimum therapeutic doses for adults were established by a panel of EpiPGX principal investigators (SMS, JC, ND, CD, HL, AGM, JWS, GJS), informed by World Health Organization (WHO) defined daily doses (DDD) (atcddd.fhi.no/atc_ddd_index/). It is important to note that the agreed appropriate AED daily doses only apply to monotherapy trials and that the list was used as guidance rather than a set of strict rules. Clinical judgment was required to evaluate the adequacy of AED trials in the context of polytherapy, extremely low or high body weight, and for AED trials taking place in an individual's childhood. Laboratory reports of AED levels were taken into account if available. If the AED levels were below the local reference range while the individual was taking a stable dose of the AED and there were no signs indicating CNS toxicity, the AED trial was considered inadequate. Individuals with non-epileptic seizures or known non-adherence were excluded from the study. Individuals who underwent epilepsy surgery were classified as drug-resistant if they met the DRE criteria before surgery and excluded from analysis if they achieved remission following epilepsy surgery. This classification approach required substantial efforts and resources. Of the ~10,000 individuals in the EpiPGX database, only those who could be robustly classified in one of the two response groups were included. We note that this level of phenotyping depth requires significant time and effort and is not generally feasible. The EpiPGX cohort thus represents a deeply-phenotyped group nested within the broader framework of the Epi25 cohort. The deep phenotyping used for EpiPGX, designed as a pharmacogenomics study, was not undertaken for the second cohort from the Epi25 Collaborative, the primary purpose of which was gene discovery. Overall, the joint cohort achieves robust and

aligned classification of seizure freedom (and thus drug responsiveness, as all patients achieving seizure freedom were on ASMs) and real-world usages for response to individual ASMs. This approach will facilitate both ease of independent replication and enlargement of the cohort in our own future work.

For the Epi25 Collaborative, the unmodified ILAE definitions of DRE (failure of adequate trials of two tolerated, appropriately chosen, and used ASM schedules)² and drug responsiveness (“seizure-free for a minimum of three times the longest pretreatment inter-seizure interval, or 12 months, whichever is longer”) were used. This, too, ensures that individuals were seizure-free for at least 12 months (or longer). Across both cohorts, therefore, those deemed drug-responsive had been seizure-free for at least 12 months, a meaningful and consequential period of seizure freedom aligned across the two cohorts. Because detailed drug response data was only provided in a minority of Epi25 participants, we could only include a fraction of the whole Epi25 study. The study cohorts are detailed in Table 1. Both cohorts displayed similar demographics, apart from the mean age at epilepsy onset of drug-responsive individuals, which was higher in the EpiPGX compared to the Epi25 cohort. Age is, however, not considered a factor in the development of drug resistance.¹² The EpiPGX and Epi25 GWAS cohorts included individuals with possible genetic causes (EpiPGX: 3.7% of the drug-resistant and 5.9% of the drug-responsive individuals; Epi25: 17% of the drug-resistant and 13% of the drug-responsive individuals; Supplementary Tables 6 and 7), without a significant enrichment of individuals with a possible genetic cause in either of the drug response groups across both cohorts ($P=0.083$ [Cochran-Mantel Haenszel test stratified for the two cohorts]). These individuals were included in the analyses following evidence that common genetic risk variants are enriched in individuals with a family history of the phenotype or unique causal variants.^{25–29} Epilepsy type and epilepsy sub-syndromes were diagnosed in all cohorts based on the primary mode of seizure onset (generalised vs focal), taking into account clinical interview data, neurological examination, EEG, and imaging data, following ILAE schemata.³⁰

Single nucleotide polymorphism genotyping

All EpiPGX samples were genotyped at deCODE Genetics (Reykjavik, Iceland) using Illumina single nucleotide polymorphism (SNP) arrays (OmniExpress-12 v1.1, OmniExpress-24 v1.1, Human610-Quad, HumanHap550v3). SNP genotypes were called with the Genotyping Module of

the GenomeStudio Software (Illumina, CA, USA). Epi25 samples were genotyped at the Broad Institute of Harvard and MIT (Cambridge, MA, USA) using the Illumina Global Screening Array with Multi-disease drop-in (GSA-MD v1.0). SNP genotypes were called using Illumina's genotyping analysis software Autocall. Rare SNPs (minor allele frequency, MAF<0.1) were called with the zCall software³¹ into the Autocall output.

Data quality control and imputation - EpiPGX cohort

For the EpiPGX samples, data quality control (QC) and imputation were performed separately for each chip type and genotyping batch. Before imputation, we excluded genotyped individuals based on the following criteria: (1) genotyping call rate (CR) <0.98; (2) heterozygosity rate outliers with >5 standard deviations (SD) from the median of the whole sample, using a subset of uncorrelated SNPs (pairwise r^2 <0.1 in 100 Kbp sliding windows with a step size of 25 SNPs); (3) missing, ambiguous, or sex mismatch between X-chromosome genotype and reported sex; (4) one individual from each pair of closely related individuals with >0.9 identity by state; (5) individuals with <90% European ancestry, as identified using STRUCTURE-v2.2,³² with HapMap European samples as the reference population and 2,766 ethnicity-sensitive SNPs. We then excluded SNPs based on the following criteria: (1) SNP-CR<0.95; (2) MAF<0.01; (3) deviation from the Hardy-Weinberg equilibrium (HWE) with P <10⁻⁶. We applied pre-imputation checks according to scripts available on the website of Will Rayner of the Wellcome Trust Centre for Human Genetics (Supplementary material, URLs) to align the QC-filtered dataset to the imputation reference (variant name, variant position, and strand orientation), remove all A/T and C/G SNPs to avoid strand issues, and to remove SNPs with allele frequencies deviating >20% from the frequency in the 1000 Genomes phase 3 reference.³³ We then split genotypes up according to chromosome arms (and in the case of chromosome X, we split additionally into pseudo-autosomal regions, PAR, and non-PAR) and created phased haplotypes using SHAPEIT-v215³⁴ with recommended effective size setting (HapMap2 European, N=11,418), and using the 1000 Genomes phase 1 integrated (v3) map files as reference. Following haplotype phasing, we imputed genotypes into our dataset using IMPUTE-v2.3.0³⁵ with recommended effective population size settings (20,000) and 1000 Genomes phase 1 integrated (v3) genotypes as reference.³⁶ The haplotype phasing and imputation were performed in separate batches for each genotyping dataset.

Post-imputation QC filters were applied first separately for every imputation batch to remove genotyped variants with low concordance between the observed genotype and masked, imputed genotype (IMPUTE2 r^2_{type0} score <0.90 , $\text{concordance_type0} < 0.90$). We then performed further QCs on the merged datasets for GWAS cases and controls separately, removing variants based on the following criteria: (1) SNPtest $v2^{37}$ imputation quality info score <0.97 ; (2) SNPtest $\text{average_maximum_posterior_call} < 0.90$; (3) $\text{MAF} < 0.01$; (4) deviation from HWE with $P < 10^{-6}$ in controls only. QC-filtered imputed genotypes were converted for subsequent analyses to hard calls using GTOOL (Supplementary material, URLs). At the individual level, we removed duplicate samples across imputation batches (using the same parameter as in the pre-imputation step).

Data quality control and imputation - Epi25 cohort

Before imputation, genotyped Epi25 individuals were excluded based on the following sample-level QC filters: (1) heterozygous/homozygous SNP ratio outliers with >4 SD from the mean of the whole sample; (2) individuals with missing, ambiguous, or mismatch between genetically inferred and reported sex; (3) one individual from each pair of closely related individuals with >0.2 proportion of identity by descent; (4) population outliers not clustering with the 1000 Genomes Project³³ European samples in a principal component analysis (PCA). SNPs were filtered out with the following criteria: (1) $\text{SNP-CR} < 0.98$; (2) monomorphic SNPs; (3) SNPs with batch association ($P < 10^{-4}$); (4) deviation from HWE with $P < 10^{-10}$. The resulting QC-filtered SNPs were used for imputation to the Haplotype Reference Consortium reference r1.1³⁸ using Minimac4³⁹ and reference-based phasing with Eagle-v2.4,⁴⁰ as implemented on the Michigan Imputation Server.³⁹ All Epi25 samples were imputed as one single batch.

Post-imputation, we randomly removed one individual from each pair of individuals with 3rd-degree relationships and higher (kinship coefficient >0.0442) using KING.⁴¹ Imputed genotypes were converted to hard calls using PLINK-v1.9⁴² and filtered for high quality based on the following criteria: (1) Minimac4 imputation quality score, $R^2 \geq 0.3$; (2) Minimac4 squared correlation value between masked genotypes of genotyped SNPs and the imputed dosages, $\text{Emp-}R^2 \geq 0.3$.

Detection of overlapping individuals across the EpiPGX and Epi25 cohorts

To identify individuals that were ascertained in the EpiPGX and the Epi25 study without sharing individual-level data between sites, we used a protocol inspired by the one-way cryptographic hash function.⁴³ One-way cryptographic hashes are a security algorithm form that alters input data so that the resulting output data cannot be reverted feasibly to the original form. We first generated ten batches of SNPs, which did not have missing genotypes in any of the studies. We then computed hash values (checksums) for each of the ten batches for each individual, using the Linux “cksum” command. The “cksum” command will always generate the same unique hash value when using the same SNPs, with the same information (same non-missing genotype), and in the same order (sorted by physical position). We then marked every pair of individuals with one or more identical hash values (out of the ten) as duplicate and excluded the corresponding individual from the Epi25 cohort. The procedure is implemented in Perl and is freely available (Supplementary material, URLs). We removed 22 samples from the Epi25 cohort duplicated between the EpiPGX and Epi25 cohorts before generating the GWAS statistics.

Genetic correlation analyses

We used LDSC to calculate the genetic correlation (R_g) of the drug response phenotype in focal epilepsy with epilepsy and the two main subtypes (focal and generalised epilepsy) (Supplementary Table 5). The summary statistics for epilepsy vs. (healthy) controls were obtained from the most recent GWAS in epilepsy.⁴⁴ We used pre-computed LD scores suitable for GWASs based on European individuals, generated as described in Bulik-Sullivan et al. (2015).⁴⁵

Genome-wide association and meta-analysis

We used logistic regression adjusted for sex and the first ten principal components of ancestry in PLINK-v1.9⁴² to perform separate GWASs in the EpiPGX and Epi25 cohorts. We did not adjust our analysis for potential non-genetic predictors of drug resistance. We performed three GWASs for each cohort in drug-resistant vs. drug-responsive individuals with (1) any type of epilepsy, ‘all-EPI’; (2) non-acquired or lesional focal epilepsy, ‘FE’; or (3) generalised epilepsy, ‘GE’. SNPs for GWASs were selected based on the following criteria: (1) $CR \geq 0.98$ in the combined case/control dataset; (2) $MAF \geq 0.01$; (3) deviation from HWE with $P > 10^{-5}$. Sample and SNP QC procedures were performed using PLINK-v1.9.⁴² To minimise confounding due to population stratification, we performed a stringent, post-imputation selection of individuals clustering exclusively with

Western European and British individuals from the 1000 Genomes Project³³ in a PCA using GCTA.⁴⁶ Of note, as well as excluding individuals with Finnish ancestry, as is standard (best) practice for GWASs in the European population, we also excluded European individuals that clustered with Tuscan⁴⁷ and Iberian⁴⁸ individuals to avoid population stratification within the largely Western and Central European GWAS cohort.

Next, we performed *P*-value-based fixed-effects meta-analyses with GWAMA⁴⁹ for each of the three epilepsy phenotypes (all-EPI, FE, and GE). The threshold for genome-wide significance in the meta-analyses was set to the commonly used $\alpha=5\times 10^{-8}$. Fine-mapping of the meta-analysis association signals was performed using FUMA,⁵⁰ LocusZoom,⁵¹ and Haploview.⁵² Gene-based association analyses were performed using MAGMA⁵³ as implemented in FUMA. The Bonferroni-corrected threshold for a significant association in the MAGMA analysis was set to $\alpha=2.63\times 10^{-6}$ (19,005 tested protein-coding genes).

Transcriptome-wide association analysis (TWAS) was performed using the S-MultiXcan framework⁵⁴ on all available brain-specific GTEx v8 transcriptome datasets (N=13). S-MultiXcan⁵⁴ leverages the substantial sharing of quantitative trait loci (QTL) across tissues to increase the power of identifying associated gene expression or alternative splicing variation.⁵⁵ Expression and splicing predictions were generated using multivariate adaptive shrinkage (mash) models⁵⁶ for GTEx v8 expression QTL (eQTL) and splicing QTL (sQTL) data.⁵⁷ We then applied the S-MultiXcan framework on all brain-specific GTEx v8 transcriptome datasets (N=13). The Bonferroni-corrected thresholds for a significant association were set to $\alpha=2.69\times 10^{-6}$ (18,562 tested genes) in the eQTL-based TWAS and $\alpha=3.78\times 10^{-7}$ (132,272 tested splicing events) in the sQTL-based TWAS. Power calculations were performed *post hoc* using the PGA Power Calculator,⁵⁸ assuming a disease prevalence of 0.1%, an additive risk model, and linkage disequilibrium (LD) $r^2=0.9$ between a causal variant and a genotyped marker.

Role of funders

The funding institutions had no role in the design and conduct of the study, including data collection, analysis, and interpretation of results, or the preparation, review, and decision to submit the manuscript for publication.

Results

Genome-wide association meta-analysis reveals one locus associated with drug resistance in focal epilepsy

To test for a possible genetic basis of DRE, we performed European ancestry-focused genome-wide association (GWA) meta-analyses in 4,208 individuals with DRE vs. 2,618 individuals with drug-responsive epilepsy. We did not identify any genome-wide significant loci in the all-EPI analysis (Figure 1) despite 80% power to detect a genetic predictor of relative risk ≥ 1.33 (Supplementary Figures 1 and 2). Subanalyses were performed in drug-resistant vs. drug-responsive individuals with FE or GE (see cohorts in Table 1). The sample size for drug-resistant GE was underpowered to detect common risk factors and SNPs showing association trends not overlapping with ‘all-EPI’ or FE association signals (Figure 1, Supplementary Figure 1). Fixed-effects GWA meta-analysis for drug resistance in FE identified seven genome-wide significant SNPs in a region of strong linkage disequilibrium on chromosome 1q42.11-q42.12 encompassing *CNIH4*, *WDR26*, and *CNIH3* (lead SNP: rs35915186, $P=1.51 \times 10^{-8}$ [logistic regression], odds ratio OR[C]=0.74, 95% confidence interval [95%-CI]:0.66-0.82) (Figure 1). Interestingly, all associated SNPs at the identified locus had OR<1, indicating that the minor allele (MAF=0.14) protects against drug resistance. This locus was not significantly associated with any epilepsy subtype in the most recent epilepsy GWAS⁴⁴ (lowest uncorrected $P=0.009$ [linear mixed model] for FE vs healthy controls, Supplementary Tables 2 and 3). The GWAS Catalog listed 86 associations with $P < 5 \times 10^{-8}$ within ± 500 Kb of the lead SNP, of which 33 were in strong LD with rs35915186 ($r^2 > 0.8$), but none were related to neurological or psychiatric traits (Supplementary Table 4). Notably, we did not find any genetic correlation between drug resistance in FE and susceptibility to FE itself, based on genetic correlation analyses with the ILAE 2023 GWAS for FE⁴⁴ (linkage disequilibrium score regression genetic heritability=-0.22, standard error=0.38, $P=0.28$ [regression]; Supplementary Table 5).

WDR26, *CNIH3*, and *CNIH4* are candidate drivers of drug response in focal epilepsy

Fine-mapping of the region associated with drug response in FE narrowed down the critical region to a 161Kb LD block of 106 SNPs in high LD with at least one of the seven genome-wide significant SNPs ($r^2 \geq 0.8$ using 1000g Phase 3 EUR data, Figure 2). The identified LD block featured three genes: *CNIH4*, *WDR26*, and the first two exons of a *CNIH3* transcript variant (ENST00000471578.5). All three genes emerged as genome-wide significant after Bonferroni

correction for multiple testing ($P < 2.63 \times 10^{-6}$ [multiple regression with F-test]) in a MAGMA⁵³ gene-based association analysis of drug-resistant FE (Supplementary Table 1, Supplementary Figure 3).

We then performed two multi-tissue TWASs for eQTL and sQTL GTEx v8 data using S-MultiXcan⁵⁴ to identify expression or splicing events associated with drug response in FE. eQTL-based TWAS across 13 GTEx v8 brain tissues implied significantly higher expression levels of *CNIH3* and *WDR26* in drug-resistant compared to drug-responsive FE ($P_{CNIH3} = 1.10 \times 10^{-6}$, $Z_{MEAN} = 3.55$; $P_{WDR26} = 1.60 \times 10^{-6}$, $Z_{MEAN} = 3.44$; multivariate regression with F-test; Table 2) at a Bonferroni-corrected significance threshold $\alpha = 2.69 \times 10^{-6}$. sQTL-based TWAS across the same brain tissues revealed 18 unique splicing events associated with drug response in FE, mapping exclusively to the three candidate genes at a Bonferroni-corrected significance threshold $\alpha = 3.78 \times 10^{-7}$ (*CNIH3*, *WDR26*, and *CNIH4*, Supplementary Table 9).

CNIH3 is one of two members of the cornichon family of transmembrane proteins coassembled with AMPA receptors (along with *CNIH2*)⁵⁹ and a brain-specific expressed gene that shows the highest expression in the frontal cortex (BA9).⁶⁰ Upon successful assembly, *CNIH3* increases the surface expression of AMPA receptors and slows deactivation and desensitisation kinetics.^{59,61} *Cnih3* knock-out in mice depresses AMPA receptor synaptic transmission only when combined with *Cnih2* knock-out, suggesting that *CNIH2* can compensate for the lack of *CNIH3*.⁶² All four genes encoding AMPA receptors have been reported to cause monogenic autosomal dominant neurodevelopmental disorders with seizures (*GRIA1*,⁶³ *GRIA2*,⁶⁴ *GRIA3*,⁶⁵ *GRIA4*⁶⁶). *CNIH4* is a brain-expressed but not brain-specific gene that shows the highest expression in cultured fibroblasts⁶⁰ and is a distantly related member of the cornichon family,⁶⁷ which lacks key residues responsible for binding to AMPA receptors.⁶⁸ *Cnih4* knock-out mice were reported as viable without any “overt” developmental abnormalities.⁶⁸ *WDR26* is a brain-expressed but not brain-specific gene with the highest expression levels in the skin.⁶⁰ *WDR26* haploinsufficiency is known to cause an (ultra-rare) distinct clinical phenotype characterised by intellectual disability and seizures (*WDR26*-related intellectual disability / Skraban-Deardorff syndrome).⁶⁹ The exact biological function of *WDR26* is not established; studies suggest roles in MAPK signalling,⁷⁰ PI3K/AKT signalling,⁷¹ and the negative regulation of β -catenin degradation within the Wnt signalling pathway⁷² (among other possible functions^{73–76}). Notably, the seizure types described in affected individuals were self-limited or responded well to standard treatments.⁷⁷ Upon screening

samples that also had whole-exome sequencing, we identified 10 individuals with FE, eight individuals with GE and one individual with DEE who carried rare variants in the candidate genes ($N_{WDR26}=10$, $N_{CNIH3}=7$, and $N_{CNIH4}=2$). Only one of these variants was classified as likely pathogenic according to ACMG criteria (without considering gene-disease relationships), while all others were classified as variants of uncertain significance. There was no clear over-representation of rare variant carriers in either group (drug-resistant or drug-responsive) (Supplementary Table 8).

Discussion

We performed case-case GWAS meta-analyses for drug response in the EpiPGX Consortium and the Epi25 Collaborative cohorts. Following evidence from previous studies that showed significant differences between the genetic architectures of epilepsy sub-syndromes,^{78,79} we performed additional GWAS meta-analyses for drug resistance in focal (FE) and generalised epilepsy (GE). We found a genome-wide significant locus at 1q42.11-q42.12 associated with protection against drug resistance in FE. This common risk locus driving drug response in FE was not previously reported as a risk factor for FE itself or any other epilepsy type.⁴⁴ We had insufficient power to identify genetic factors associated with drug-resistant GE. In line with our hypothesis that different mechanisms drive drug response in FE compared to GE, we found no significant risk factors when combining FE and GE in an ‘all epilepsies’ (all-EPI) analysis. This study and one of our previous GWAS studies in mesial temporal lobe epilepsy with febrile seizures⁸⁰ demonstrate the value of focusing on more narrowly defined subtypes to identify common risk factors for traits of interest in FE.

Fine-mapping the association signal for drug response in FE revealed three candidate genes: *CNIH4*, *WDR26*, and *CNIH3*. Among these, pathogenic variation in *WDR26* has been shown to cause a drug-responsive seizure phenotype⁷⁷ consistent with the protective effect from drug-resistant epilepsy we observed from the meta-analysis, and the higher expression levels for *CNIH3* and *WDR26* in drug-resistant cases suggested by the TWAS. Although *CNIH3* has not been identified as a monogenic epilepsy gene, common *CNIH3* variants could plausibly act as a modifier of drug response. *CNIH3* acts as an auxiliary subunit that regulates AMPA receptor gating and trafficking,^{59,61,81} and abnormal AMPA receptor trafficking could contribute to seizure activity.⁸² Our result should spark further research to uncover novel therapies, as no drug-gene

interactions are currently reported for the three candidate genes.⁸³ Our eQTL- and sQTL-based TWAS framework could not conclusively prioritise between the three candidate genes. However, as the underlying gene expression and splicing variation predictions are based on GTEx post-mortem bulk transcriptomics data, our analyses may suffer from sensitivity limitations and not fully capture cell-type-specific expression and transcriptional patterns of living tissues or under disease-specific conditions.⁸⁴

While we identified common variants predicting drug response in FE, additional genetic (and environmental) factors are likely to play a role in DRE. There is accumulating evidence that rare genetic variation is important in epilepsy causation, and such variation can overlap with poor response to ASMs.⁸⁵ Rare variants known to cause monogenic forms of epilepsy can also influence drug response. For example, sodium channel blockers aggravate seizures in most people with Dravet syndrome due to loss-of-function *SCN1A*⁸⁶ mutations or epilepsy due to loss-of-function variants in *SCN2A*⁸⁷ or *SCN8A*.⁸⁸ Conversely, sodium channel blockers are an effective treatment for people with epilepsy due to gain-of-function variants in *SCN1A*,⁸⁹ *SCN2A*,⁸⁷ or *SCN8A*.⁸⁸

Further research in larger cohorts is needed to detect the causal genes and mechanisms for drug resistance in epilepsy. Our GWA meta-analyses were underpowered to capture significant single-SNP associations with drug-resistant GE. We focused on overall drug resistance in large epilepsy subgroups. Testing in larger cohorts that allow drug-specific sub-analyses, drug-matched control usage, and stratification for comorbid disorders may help uncover biomarkers for drug-specific resistance in epilepsy. For example, a recent study suggested rare variants underlie resistance to two common ASMs:⁸⁵ rare variants in ADME (absorption, distribution, metabolism, and excretion) genes were associated with resistance to valproic acid and rare variants in drug target genes were associated with resistance to levetiracetam. We opted for a very stringent selection of individuals with Western and Central European-like ancestry to reduce potential confounding of association statistics by population sub-structure.⁹⁰ Therefore, the generalizability of these results to individuals beyond European-like ancestry remains to be determined. Operational definitions, typically applied at a single point in time to define drug-resistant and drug-responsive cases, cause additional challenges in drug-resistance research in epilepsy. Such definitions do not consider the dynamic relationship between drug resistance and seizure remission and recurrence. Most people with epilepsy attain remission early, later in their disease history, or never, with only a minority fluctuating between periods of seizure freedom and relapse.⁹¹ Because

a dynamic course is more common in individuals with infrequent seizures,⁹² the EpiPGX definition of DRE (which requires a minimum of four seizures in the past 12 months) partially addresses this issue. Continued efforts are needed in the field to reach a consensus on addressing the temporal course of drug resistance in epilepsy for research purposes. Finally, phenotyping and clinical information collection for the EpiPGX cohort was completed over a decade ago, utilising terminology and classifications predating the current definition of Developmental and Epileptic Encephalopathies (DEEs).⁹³ Consequently, the presence of individuals with DEE within the FE GWAS meta-analyses cannot be entirely ruled out. However, we note that even if there were an over-representation of individuals with DEE in the drug-resistant cohorts, and even if these individuals have a monogenic basis for their epilepsy and the drug-resistant nature of that epilepsy, this would serve only to reduce the power of our current analysis. The same applies to the possibility that any focal epilepsies might have been monogenic.

In conclusion, we show that drug resistance in focal epilepsy has a common genetic component. More large-scale projects are needed to identify biomarkers for drug resistance in epilepsy. Potentially, such work could provide new clues to the aetiology and pathophysiology of drug-resistant epilepsy, especially focal epilepsy. The common polygenic nature of the genetic contribution to drug resistance could inform treatment strategies and may point to the need for alternative approaches focused broadly on pathways rather than single molecular targets.

Contributors

CL and SMS designed the study. AA, AGM, ALJ, BMN, BPCCK, CAB, CD, FZ, GJS, GLC, HL, HS, JJC, JWS, KS, LJ, MRJ, ND, PM, PS, SFB, SMS, TJOB, UU, the EpiPGX Consortium, and the Epi25 Collaborative collected data. AA, CL, EH, HMC, LJ, MEO, PM, RS, SC, and UU analysed the data. AA, CL, DS, RS, and SMS accessed and verified the underlying data, interpreted the analyses, and were responsible for the decision to submit the manuscript. BMN, DL, HS, PM, RK, and SMS provided the computational infrastructure. SMS supervised the study. AA, CL, RS, and SMS wrote the manuscript. All authors interpreted the data, read and revised the manuscript, and approved the final version of the manuscript.

Declaration of Interests

AA is an employee of Regenon and owns Regenon stocks.. AGM received institutional consulting fees from Jazz Pharmaceuticals and UCB Pharma; institutional honoraria for lectures from Sanofi and GSK, institutional support for meeting attendance/travel from Angelini Pharma, and has unpaid leadership roles at the European Academy of Neurology and Epilepsy Research Institute.. BMN is a member of the Neumora Scientific and Deep Genomics Scientific Advisory boards.. CD received institutional honoraria from UCB Pharma, support for meeting attendance/travel from Angelini Pharma, and is a member of Angelini Pharma and Neuraxpharm Advisory Boards.. DL received institutional research funding from the National Institutes of Health (NIH), National Institute of Neurological Disorders and Stroke (NINDS) under R01 NS117544.. GC received research funding from Janssen Pharmaceuticals and consultancy fees from Ono Pharmaceuticals.. GJS received personal consulting fees from Angelini Pharma, and personal honoraria for lectures from Angelini Pharma, Bial Pharma UK, and UCB Pharma.. HL received personal consulting fees from Praxis Precision Medicine and institutional fees from Lario Therapeutics, personal lecture honoraria from Eisai and UCB Pharma, personal payment for expert testimony for Fondazione Telethon, and is an advisory board member of IntraBio.. HS, KS, and UU are employees of deCODE genetics/Amgen Inc.. JJC received honoraria from UCB-Pharma, Glaxo Smith Kline, Janssen-Cilag, Sanofi-Synthelabo, Pfizer, and Eisai to attend advisory boards, present lectures / tutorials.. JWS received institutional research funding from the UK National Institute for Health Research, Angelini Pharma, UCB Pharma, Epilepsy NL, the Academy of Medical Sciences, and the Wellcome Trust Alliance, institutional consulting fees from the UK Competition & Markets

Authority and Eisai, personal honoraria from Angelini Pharma, Eisai, and UCB, and personal fees for participation on an advisory board of Angelini Pharma, and holds unpaid roles as Medical Director of the UK Epilepsy Society and Editorial Board member for Lancet Neurology.. LJ received institutional research funding from the Swedish state under the ALF agreement (ALFGBG-966370), Genomic Medicine Sweden (GMS K131050263), and the Söderström König Foundation.. MRJ received research funding from the UK Research and Innovation (UKRI) Medical Research Council (MRC) (Award Nos. MR/S02638X/1 and MR/W029790/1).. ND received consulting fees and honoraria from Angelini Pharma, Actiobio, Eisai, UNEEG Medical, and Jazz Pharma, and owns stocks from Actiobio.. PS received consulting fees from Jazz Healthcare and UCB Pharma for participation on advisory boards, and honoraria from Proveca and UCB Pharma for presentation at congresses/workshops.. SFB received institutional research funding from the National Health and Medical Research Council (NHMRC) (Grant IDs 1091593, 1196637, 2010562), UCB Pharma, Eisai, SEER, Chiesi, and LivaNova, personal consulting fees from Praxis Precision Medicines, personal honoraria from Eisai and DeltaMed, co-owns a patent held by Bionomics Inc. licensed to Athena Diagnostics, Genetics Technologies Ltd, and is the Chief Medical Officer of the Epilepsy Foundation (Victoria).. SMS received institutional research funding from CURE Epilepsy, Epilepsy Society, and MRC, honoraria from Angelini Pharma, Eisai, Zogenix, UCB, Eisai, Jazz Pharmaceuticals, and UCB, travel support from UCB Pharma, and is a member of Advisory Boards of Biocodex, Stoke Therapeutics, and Takeda.. TJOB received institutional research funding from NHMRC, MRFF, DOD, and NIH, institutional consulting fees from Kinaxis Therapeutics, Jazz Pharmaceuticals, and Livanova, institutional honoraria from UCB, institutional travel support from Longboard Pharmaceuticals, is a member of the Kinaxis Therapeutics Advisory Board, and hold an unpaid position as Cabrini Health Board Member. All other authors have no conflict of interests to declare.

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Data Sharing Statement

GWAS/meta-analysis summary statistics that support the findings of this study are available on application to the Data Access Committees of the EpiPGX Consortium and the Epi25 Collaborative through the corresponding authors. Due to ethical restrictions, we cannot publicly share individual-level raw data of the EpiPGX cohort. However, subject to data use agreements and collaborator approval, we can provide regulated access to bona fide researchers on our secure servers. Interested researchers can contact the corresponding author for further details. Individual-level raw data of the Epi25 cohort are deposited in dbGaP (accession number: phs001489.v3.p2). Summary statistics of GWAS in epilepsy used in this study for correlation analyses are available from the open-access Epilepsy Genetic Association database (epiGAD, <https://www.epigad.org/>).

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Table 1: GWA meta-analysis cohorts after quality control.

Study cohorts of individuals with drug-resistant or drug-responsive epilepsy. Epilepsy and epilepsy sub-syndromes were diagnosed in all cohorts according to clinical criteria (clinical interview, neurological examination, EEG, imaging data), following ILAE classifications.³⁰ Abbreviations: All-EPI: all epilepsies; FE: focal epilepsy; GE: generalised epilepsy; DEE: developmental and epileptic encephalopathy; Epilepsy-NOS: epilepsy, not otherwise specified. *Due to ethical restrictions at the time of data collection, the average age at epilepsy onset was based on only 32% of the EpiPGX GWAS sample size.

Table 2: Gene-based TWAS in drug-resistant vs. drug-responsive FE.

TWAS *P*-values were calculated using S-MultiXcan⁵⁴ with MASHR models for GTEx v8 eQTLs across 13 brain-specific tissues. Shown are all genes with $P < 10^{-3}$ [multivariate regression with F-test] in a TWAS in drug-resistant vs. drug-responsive FE. The threshold for significant associations after Bonferroni correction was set to $\alpha = 2.69 \times 10^{-6}$ (18,562 tested genes). Significant associations are highlighted in bold. Legend: N: number of "tissues" available for this gene, *P*_i_{best}: best p-value of single-tissue S-PrediXcan association (plotted in Supplementary Figure 4), *T*_i_{best}: name of best GTEx v8 single-tissue S-PrediXcan association, *Z*_{MEAN}: mean z-score among single-tissue S-PrediXcan associations.

Fig. 1: Manhattan plot of the GWAS meta-analyses in drug-resistant vs drug-responsive individuals with epilepsy.

The red line shows the threshold for genome-wide significance ($P < 5 \times 10^{-8}$). Chromosome and position are displayed on the x-axis and $-\log_{10}(P\text{-values})$ [logistic regression] on the y-axis. **a:** GWAS meta-analysis in 3,231 drug-resistant vs. 1,578 drug-responsive individuals with focal epilepsy (FE). Annotated genes were tagged by SNPs in high linkage disequilibrium with the lead SNP rs35915186 ($r^2 \geq 0.8$). **b:** GWAS meta-analysis in 4,208 drug-resistant vs. 2,618 drug-responsive individuals with epilepsy (all-EPI). **c:** GWAS meta-analysis in 506 drug-resistant vs. 751 drug-responsive individuals with generalised epilepsy (GE).

Fig. 2: Chromosome 1q42.11-q42.12 locus associated with drug response in FE.

The SNPs in the upper plot are coloured according to their linkage disequilibrium (LD) r^2 value with the lead SNP rs35915186. The linkage disequilibrium pattern with corresponding LD blocks (black triangles) is shown in the lower plot. The pairwise LD values are displayed in shades of grey, with black representing SNP pairs in full LD ($r^2 = 1$).

Cohort name		All-EPI	FE	GE	DEE	Epilepsy-NOS	Males (%)	Mean age at epilepsy onset (years) (SD)	Mean age at last follow-up (years) (SD)	Mean number of adequate ASM trials
EpiPGX	Drug-resistant	2,105	1,802	179	0	124	47.5%	15.5 (SD 13.6)*	45.6 (SD 13.4)	4.3 (SD 2.3)
	Drug-responsive	1,394	999	233	0	162	49.3%	23.1 (SD 17.2)*	44.4 (SD 17.7)	1.9 (SD 1.3)
Epi25	Drug-resistant	2,103	1,429	327	337	10	49.2%	15.6 (SD 14.8)	33.6 (SD 17.0)	Not available
	Drug-responsive	1,224	579	518	107	20	45.1%	10.7 (SD 16.6)	31.0 (SD 19.6)	Not available
Meta-analysis		6,826	4,809	1,257	444	316				

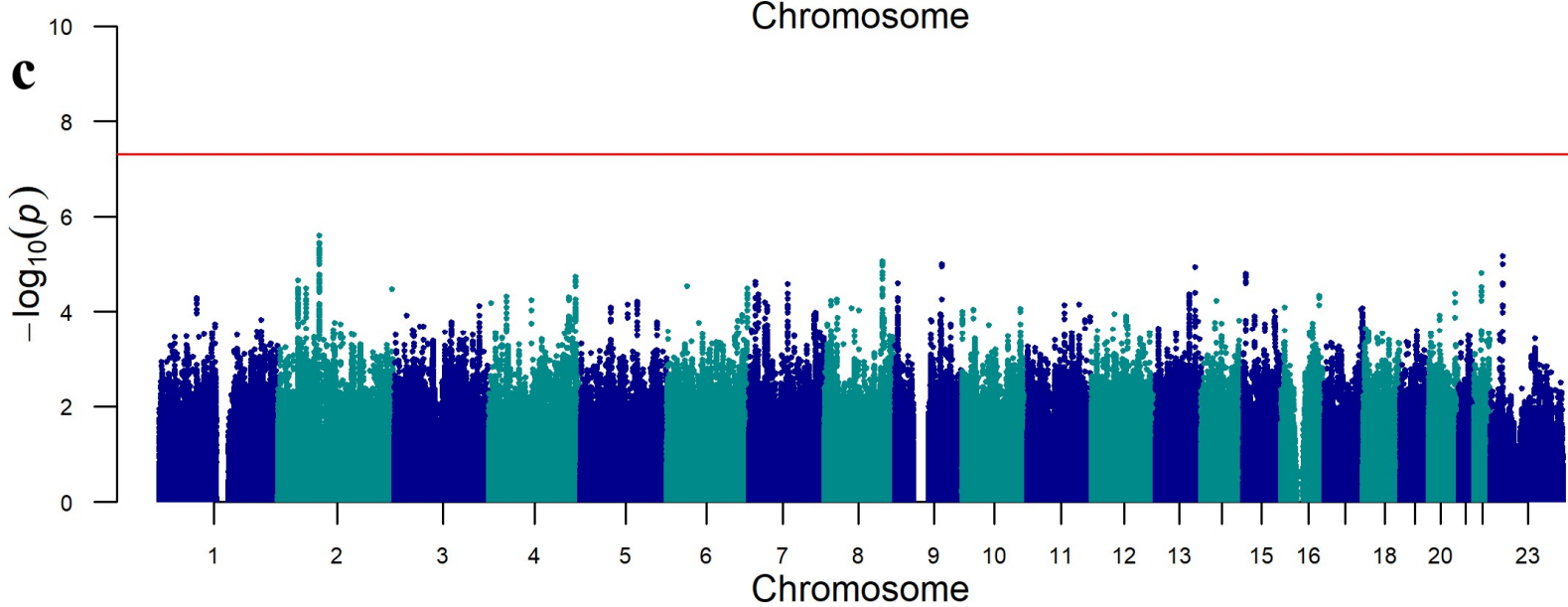
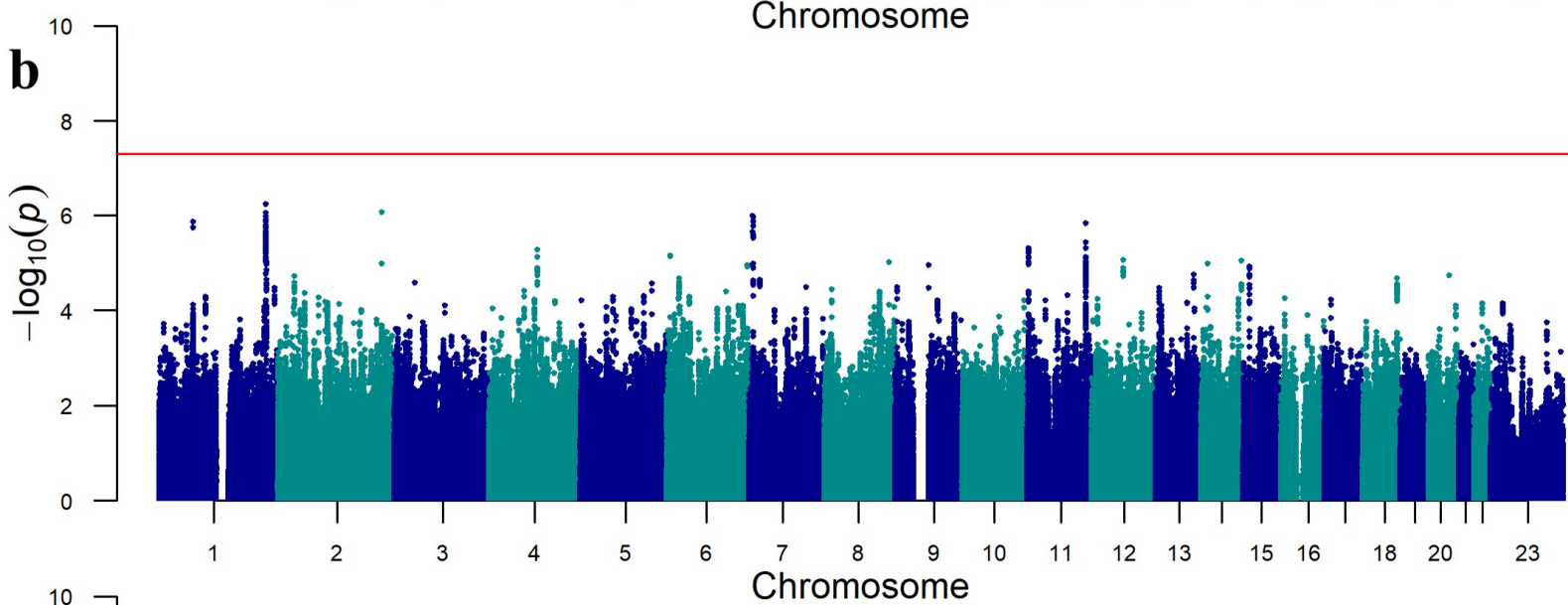
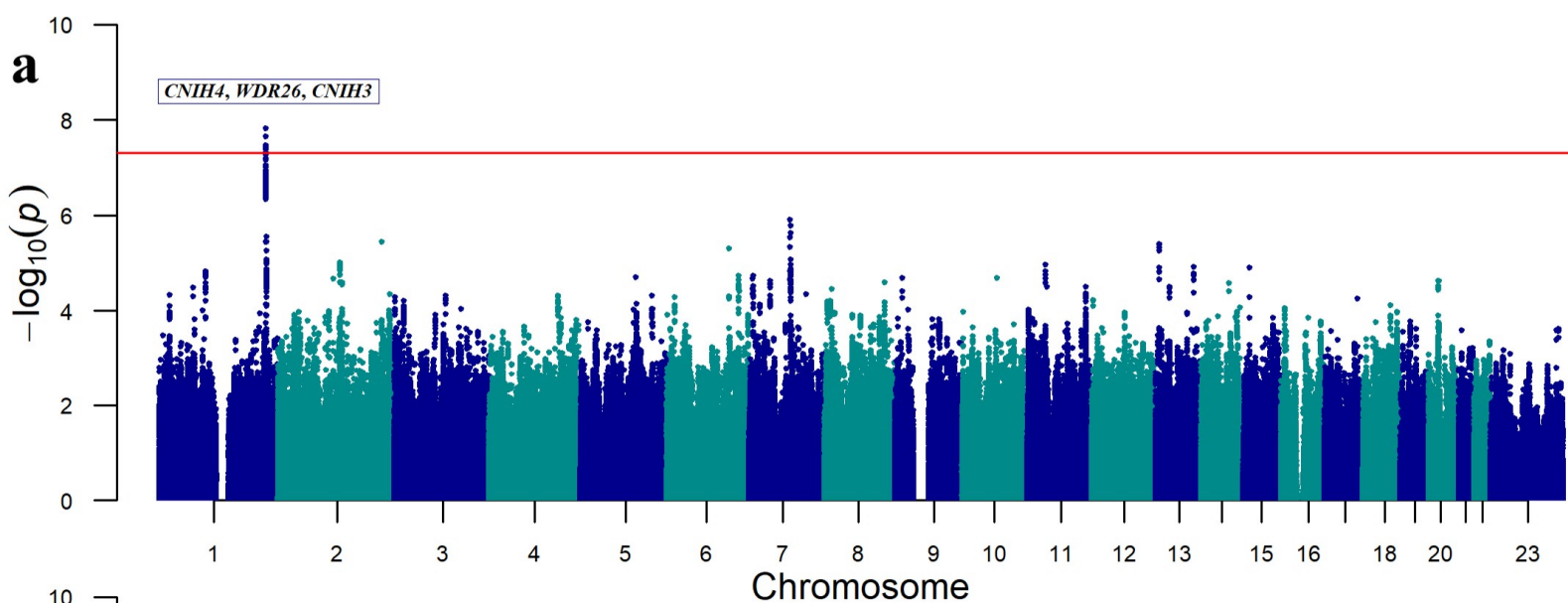
Study cohorts of individuals with drug-resistant or drug-responsive epilepsy. Epilepsy and epilepsy sub-syndromes were diagnosed in all cohorts according to clinical criteria (clinical interview, neurological examination, EEG, imaging data), following ILAE classifications.²⁹ Abbreviations: All-EPI: all epilepsies; FE: focal epilepsy; GE: generalised epilepsy; DEE: developmental and epileptic encephalopathy; Epilepsy-NOS: epilepsy, not otherwise specified. *Due to ethical restrictions at the time of data collection, the average age at epilepsy onset was based on only 32% of the EpiPGX GWAS sample size.

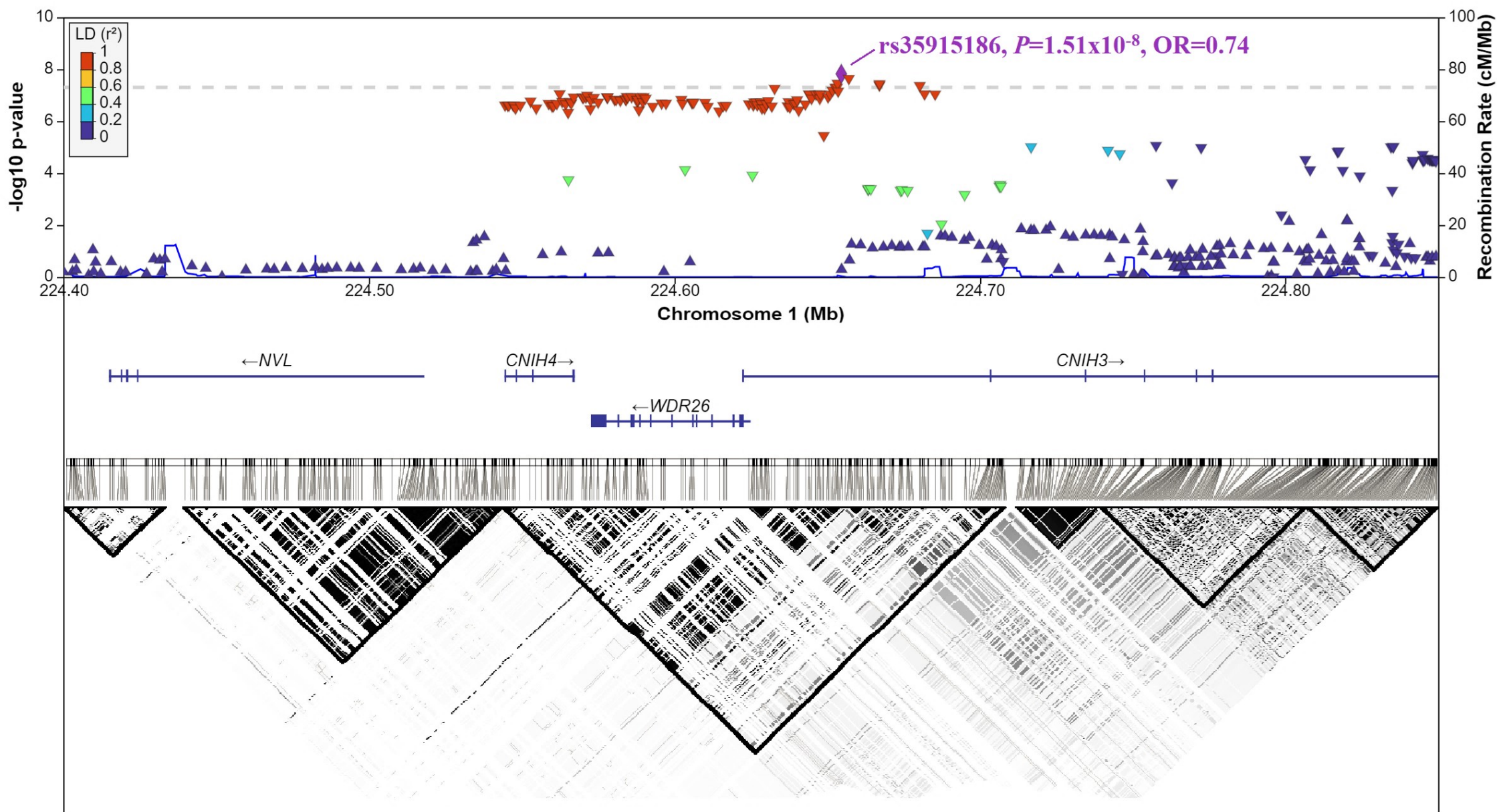
Table 1: GWA meta-analysis cohorts after quality control

Ensemble ID	Gene name	P-value	N	P_i_best	T_i_best	Z _{MEAN}
ENSG00000143786.7	<i>CNIH3</i>	1·10E-06	2	2·13E-07	Brain_Amygdala	3·55
ENSG00000162923.14	<i>WDR26</i>	1·60E-06	8	2·13E-07	Brain_Frontal_Cortex_BA9	3·44
ENSG00000143771.11	<i>CNIH4</i>	7·68E-06	13	2·37E-07	Brain_Spinal_cord_cervical_c-1	2·85
ENSG00000085563.14	<i>ABCB1</i>	5·09E-05	13	4·50E-04	Brain_Cerebellum	0·26
ENSG00000225924.2	<i>RP11-111D6.4</i>	2·28E-04	2	5·94E-05	Brain_Hippocampus	3·24
ENSG00000254480.1	<i>RP11-23F23.2</i>	2·63E-04	10	2·03E-04	Brain_Putamen_basal_ganglia	-1·74
ENSG00000114446.4	<i>IFT57</i>	2·80E-04	13	4·58E-03	Brain_Hippocampus	-1·01
ENSG00000166268.10	<i>MYRFL</i>	3·46E-04	13	2·07E-04	Brain_Amygdala	-3·52
ENSG00000247970.2	<i>RP11-543C4.1</i>	5·96E-04	9	0·18	Brain_Cerebellar_Hemisphere	0·35
ENSG00000173465.7	<i>SSSCA1</i>	6·19E-04	5	3·07E-03	Brain_Anterior_cingulate_cortex_BA24	-1·07
ENSG00000088930.7	<i>XRN2</i>	7·34E-04	10	8·76E-03	Brain_Cerebellum	-0·31
ENSG00000251562.7	<i>MALAT1</i>	7·51E-04	3	5·39E-03	Brain_Amygdala	1·51
ENSG00000139168.7	<i>ZCRB1</i>	7·83E-04	13	1·33E-03	Brain_Hippocampus	-1·78
ENSG00000006634.7	<i>DBF4</i>	8·00E-04	12	2·54E-03	Brain_Anterior_cingulate_cortex_BA24	-0·89

TWAS P-values were calculated using S-MultiXcan⁵⁶ with MASHR models for GTEx v8 eQTLs across 13 brain-specific tissues. Shown are all genes with $P < 10^{-3}$ [multivariate regression with F-test] in a TWAS in drug-resistant vs. drug-responsive FE. The threshold for significant associations after Bonferroni correction was set to $\alpha = 2 \cdot 69 \times 10^{-6}$ (18,562 tested genes). Significant associations are highlighted in bold. Legend: N: number of "tissues" available for this gene, P_i_best: best p-value of single-tissue S-PrediXcan association (plotted in Supplementary Figure 4), T_i_best: name of best GTEx v8 single-tissue S-PrediXcan association, Z_{MEAN}: mean z-score among single-tissue S-PrediXcan

Table 2: Gene-based TWAS in drug-resistant vs. drug-responsive FE





Supplementary material

Genome-wide association meta-analyses of drug-resistant epilepsy

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7. **References**

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3. URLs

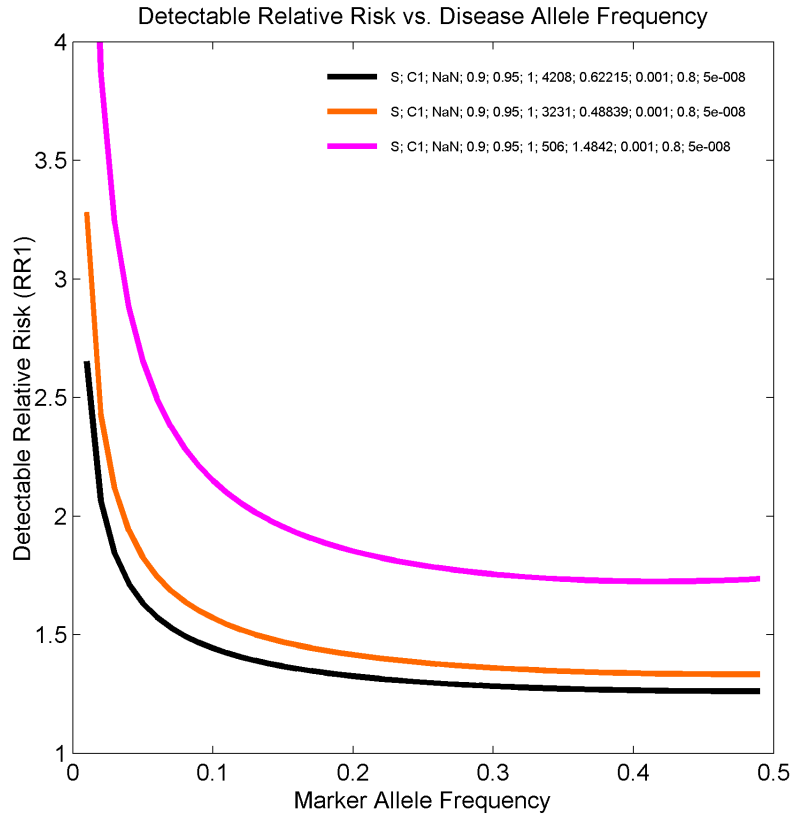
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Epi25 Collaborative: <https://epi-25.org>
STRUCTURE: <https://web.stanford.edu/group/pritchardlab/structure.html>
SHAPEIT: https://mathgen.stats.ox.ac.uk/genetics_software/shapeit/shapeit.html
IMPUTE v2: https://mathgen.stats.ox.ac.uk/impute/impute_v2.1.0.html
GTOOL, <https://hpc.nih.gov/apps/IMPUTE.html>
Will Rayner imputation preparation and checking scripts, <https://www.well.ox.ac.uk/~wrayner/tools>
Cksum script, https://personal.broadinstitute.org/sripke/share_links/checksums_download/
Epilepsy Genetic Association Database (epiGAD) database, <https://www.epigad.org>
Eagle v2.4: <https://alkesgroup.broadinstitute.org/Eagle>
Minimac4, <https://genome.sph.umich.edu/wiki/Minimac4>
Michigan Imputation Server, <https://imputationserver.sph.umich.edu/index.html>
SNPTEST v2, https://mathgen.stats.ox.ac.uk/genetics_software/snpctest/snpctest.html
KING, <https://www.kingrelatedness.com>
PLINK-v1.9: <https://www.cog-genomics.org/plink>
GCTA: <https://yanglab.westlake.edu.cn/software/gcta>
GWAMA: <https://genomics.ut.ee/en/tools>
FUMA, <https://fuma.ctglab.nl/>
LocusZoom, <http://locuszoom.org/>
Haploview: <https://www.broadinstitute.org/haploview/haploview>
LDSC: <https://github.com/bulik/ldsc>
PrediXcan/S-MultiXcan, <https://github.com/hakyimlab/MetaXcan>
PGA Power Calculator: <https://dceg.cancer.gov/tools/design/pga>

4. Supplementary Results

4.1. Power analysis

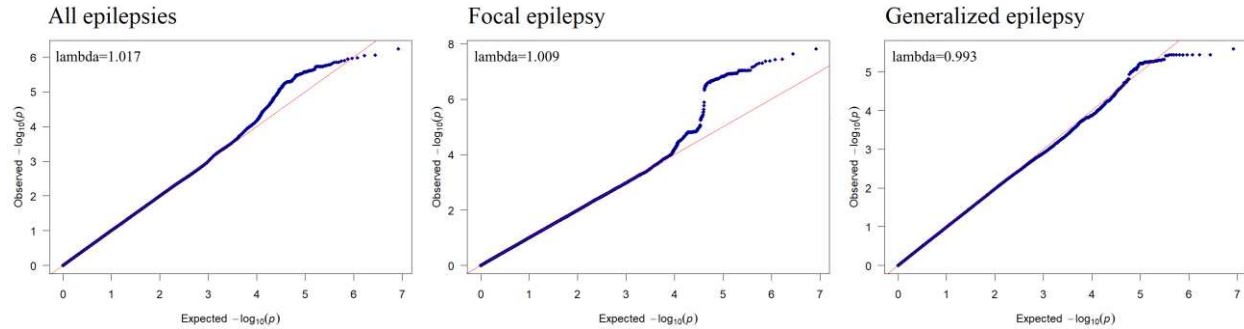
The GWA meta-analysis in DRE had 80% power to detect a genetic predictor of relative risk ≥ 1.33 (which can be approximated to odds ratio for rare disorders²) with $MAF \geq 20\%$ at $\alpha = 5 \times 10^{-8}$, assuming a disease prevalence of 0.1%, an additive genetic model, and a linkage disequilibrium $r^2 = 0.9$ between the causal variant and the genotyped marker (Supplementary Figure 1). However, no genome-wide signal was found in the ‘all epilepsies’ group (Supplementary Figure 2). The GWA meta-analysis in individuals with FE had 80% power at $\alpha = 5 \times 10^{-8}$ to detect genome-wide significant SNPs with $MAF \geq 20\%$ and relative risk ≥ 1.42 .

5. Supplementary Figures



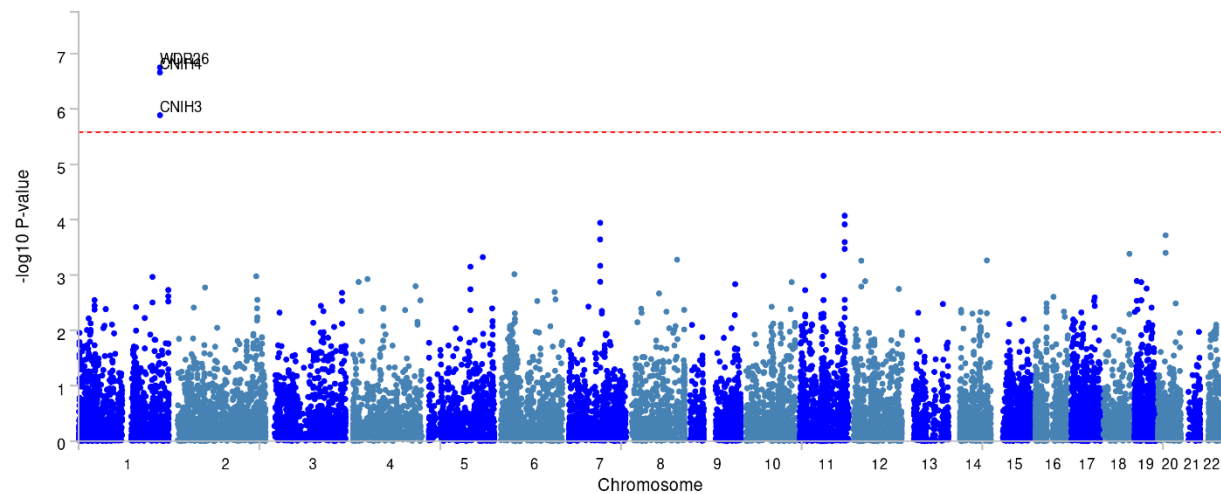
Supplementary Figure 1: Power analysis for the genome-wide association meta-analysis in DRE and subphenotypes

Shown are the detectable relative risks with 80% power at $\alpha = 5 \times 10^{-8}$, assuming a disease prevalence of 0.1%, an additive risk model, and a linkage disequilibrium $r^2 = 0.9$ between a causal variant and a genotyped marker for all individuals with DRE (black curve), drug-resistant focal epilepsy (orange curve), and drug-resistant generalised epilepsy (purple curve). Power calculations were performed using the PGA Power Calculator³.



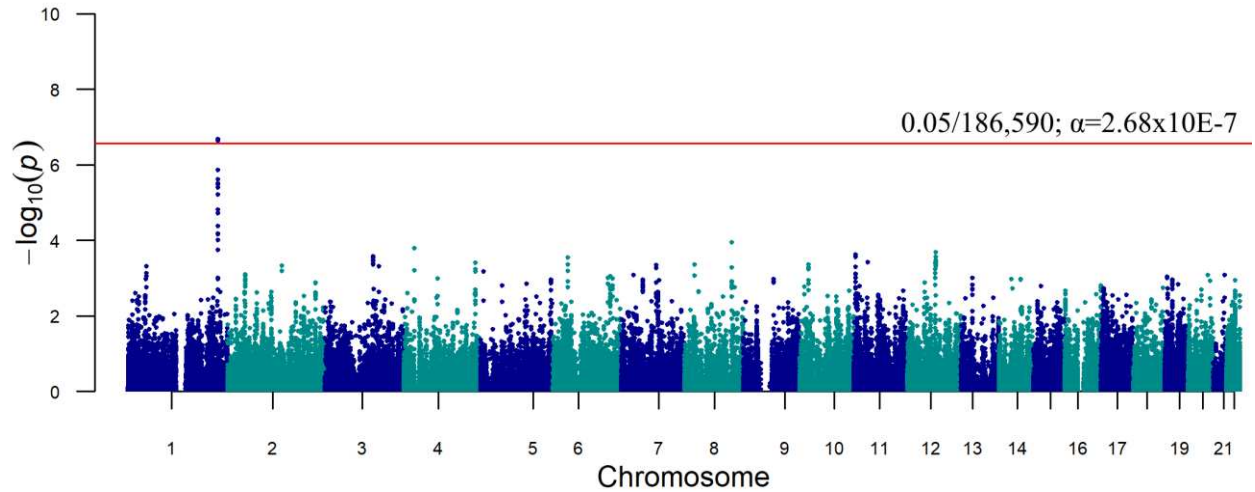
Supplementary Figure 2. Quantile-quantile plots of all three genome-wide meta-analyses of drug resistance in epilepsy, focal epilepsy, and generalised epilepsy

The observed $-\log_{10}(P\text{-values})$ [logistic regression] are plotted as a function of the expected $-\log_{10}(P\text{-values})$ for the genome-wide meta-analyses of drug resistance in epilepsy (DRE), focal epilepsy (DR-FE), and generalised epilepsy (DR-GE). The lambda inflation values were: $\lambda_{\text{DRE}}=1.017$, $\lambda_{\text{DR-FE}}=1.009$, and $\lambda_{\text{DR-GE}}=0.993$.



Supplementary Figure 3. MAGMA gene-based association study in drug-resistant vs. drug-responsive FE

Manhattan plot of the MAGMA gene-based association analysis. The chromosomes and positions of the analysed genes are displayed on the x-axis, and $-\log_{10}(P\text{-values})$ [multiple regression with F-test] are displayed on the y-axis. The red line shows the Bonferroni-corrected threshold for genome-wide significance (19,005 tested genes, $\alpha = 2.63 \times 10^{-6}$).



Supplementary Figure 4. Single-tissue transcriptome-wide association study (TWAS) in drug-resistant vs. drug-responsive FE

Manhattan plot of the single-tissue TWAS results in drug-resistant vs. drug-responsive FE. Shown are all predicted S-PrediXcan single-tissue expression level associations ($-\log_{10}(P\text{-values})$ [linear regression] on the y-axis) in each of the 13 brain-specific GTEx tissues before integration across tissues with S-MulTiXcan⁴. The red line shows the Bonferroni-corrected threshold for genome-wide significance (186,590 tested genes, $\alpha=2.68 \times 10^{-6}$).

6. Supplementary Tables

Supplementary Table 1: Gene-based association analysis in drug-resistant FE

Gene	Chr	Start (hg19)	Stop (hg19)	NSNPS	NPARAM	N	Z-score	P-value
WDR26	1	224562845	224634735	68	2	4800	5.0907	1.78E-07
CNIH4	1	224534552	224577161	48	4	4791	5.051	2.20E-07
CNIH3	1	224612362	224938251	527	24	4799	4.7001	1.30E-06
OR6T1	11	123803492	123824580	63	9	4791	3.7592	8.52E-05
RUNDC3B	7	87246864	87471611	222	9	4805	3.685	1.14E-04
OR4D5	11	123800250	123821344	64	8	4794	3.668	1.22E-04
XRN2	20	21273942	21380463	133	9	4798	3.5509	1.92E-04
ABCB1	7	87123175	87352611	315	26	4802	3.5051	2.28E-04
OR10S1	11	123837368	123858488	60	10	4795	3.4757	2.55E-04
TMEM225	11	123743633	123766349	64	9	4803	3.3988	3.38E-04
NKX2-4	20	21366005	21388666	26	5	4801	3.3542	3.98E-04
FAM69C	18	72092963	72135179	38	9	4794	3.3428	4.15E-04
PDE6A	5	149227519	149334356	97	14	4792	3.3033	4.78E-04
TMEM65	8	125314231	125394933	129	9	4796	3.274	5.30E-04
SERPINA12	14	94943611	94994181	112	20	4792	3.2653	5.47E-04
IAPP	12	21497893	21542912	104	14	4792	3.2617	5.54E-04
SLC25A40	7	87452883	87515672	77	5	4803	3.2014	6.84E-04
AP3S1	5	115167178	115259778	49	8	4781	3.1902	7.11E-04
ZFAND3	6	37777275	38132400	680	17	4798	3.0991	9.70E-04

Gene-based *P*-values were calculated using MAGMA⁵ as implemented in FUMA⁶. Shown are all genes with $P < 10^{-3}$ [multiple regression with F-test] in the gene-based association analysis for drug-resistant FE. The thresholds for a significant association after Bonferroni correction was set to $\alpha = 2.63 \times 10^{-6}$ (19,005 tested protein-coding genes). Significant associations are highlighted in bold. Legend: NSNPS: number of SNPs annotated to a gene after internal SNP quality control, NPARAM: number of relevant parameters used in the model, N: sample size used when analysing a gene.

Supplementary Table 2: ILAE epilepsy GWAS¹ *P*-values of the drug resistance meta-analysis top hits in FE

GWAS meta-analysis in 3,231 drug-resistant vs. 1,578 drug-responsive individuals with FE							ILAE GWAS 2023		
SNP	CHR	BP (hg19)	A1	A2	OR (95%-CI)	P-value	'all EPI' P-value	FE P-value	GE P-value
rs35031726	1	224562410	C	T	0.75 (0.68-0.83)	8.87E-08	0.063	0.015	0.26
rs1544196	1	224632782	A	G	0.75 (0.67-0.83)	5.48E-08	0.078	0.022	0.30
rs12743200	1	224644060	A	G	0.75 (0.68-0.83)	9.13E-08	0.039	9.22E-03	0.27
rs12758579	1	224644167	T	C	0.75 (0.68-0.83)	9.13E-08	0.039	9.22E-03	0.27
rs6663522	1	224646328	T	C	0.75 (0.68-0.83)	9.13E-08	0.064	0.017	0.29
rs35067427	1	224647230	G	T	0.75 (0.68-0.83)	9.13E-08	0.063	0.017	0.28
rs2185684	1	224647741	G	T	0.75 (0.68-0.83)	9.13E-08	0.063	0.017	0.28
rs12731630	1	224650947	C	T	0.75 (0.68-0.83)	9.13E-08	0.067	0.017	0.29
rs7365151	1	224651782	C	T	0.75 (0.67-0.83)	6.54E-08	0.090	0.026	0.28
rs11577170	1	224652687	G	A	0.75 (0.67-0.83)	5.03E-08	0.063	0.016	0.29
rs11800613	1	224653213	G	A	0.75 (0.67-0.83)	4.96E-08	0.054	0.015	0.26
rs10753468	1	224653388	C	T	0.74 (0.67-0.83)	3.50E-08	0.064	0.018	0.25
rs12062871	1	224653597	G	T	0.75 (0.67-0.83)	7.03E-08	0.055	0.017	0.24
rs35915186	1	224654623	C	T	0.74 (0.66-0.82)	1.51E-08	0.061	0.022	0.28
rs7536608	1	224657092	A	C	0.74 (0.67-0.82)	2.28E-08	0.044	9.63E-03	0.34
rs35804313	1	224667102	T	C	0.75 (0.67-0.83)	4.11E-08	0.071	0.024	0.32
rs11578709	1	224667131	G	C	0.75 (0.67-0.83)	3.65E-08	0.070	0.022	0.33
rs6698343	1	224680332	C	T	0.75 (0.67-0.83)	4.27E-08	0.043	0.011	0.26
rs17570292	1	224681894	G	T	0.75 (0.68-0.83)	8.96E-08	0.045	0.012	0.26
rs11584057	1	224685350	A	T	0.75 (0.68-0.83)	9.13E-08	0.076	0.032	0.24

Listed are the top association signals with $P < 10^{-7}$ [logistic regression] in the GWA meta-analysis in 3,231 drug-resistant vs. 1,578 drug-responsive individuals with focal epilepsy. The threshold for genome-wide significance was set to $P < 5 \times 10^{-8}$. Significant associations are highlighted in bold. The association *P*-values [linear mixed model] in the ILAE epilepsy GWAS¹ are stated for each top hit. Legend: BP: physical position, A1: effect allele, A2: non-effect allele, 'all EPI': any type of epilepsy, FE: focal epilepsy, GE: generalised epilepsy.

Supplementary Table 3: Effect allele frequencies of the FE drug resistance meta-analysis top hits in external datasets

Drug resistance meta-analysis top hits in FE (as in Supplementary Table 2)					Frequency of the effect allele (A1)			
SNP	CHR	BP (hg19)	A1	A2	Meta-analysis FE cases (drug- resistant)	Meta-analysis FE controls (drug- responsive)	ILAE GWAS 2023 cohort (cases and controls) ¹	gnomAD v4.0 (European, non- Finnish)
rs35031726	1	224562410	C	T	0.2064	0.2494	0.2259	0.2259
rs1544196	1	224632782	A	G	0.203	0.2457	0.2252	0.2278
rs12743200	1	224644060	A	G	0.2057	0.2481	0.2251	0.2253
rs12758579	1	224644167	T	C	0.2057	0.2481	0.2251	0.2251
rs6663522	1	224646328	T	C	0.2057	0.2481	0.2253	0.2251
rs35067427	1	224647230	G	T	0.2057	0.2481	0.2253	0.2268
rs2185684	1	224647741	G	T	0.2057	0.2481	0.2253	0.2252
rs12731630	1	224650947	C	T	0.2057	0.2481	0.2253	0.2253
rs7365151	1	224651782	C	T	0.2053	0.2484	0.23	0.2249
rs11577170	1	224652687	G	A	0.2049	0.2484	0.2253	0.2255
rs11800613	1	224653213	G	A	0.2053	0.2487	0.2253	0.2254
rs10753468	1	224653388	C	T	0.2062	0.2503	0.2315	0.2271
rs12062871	1	224653597	G	T	0.2057	0.2487	0.225	0.2254
rs35915186	1	224654623	C	T	0.2053	0.2505	0.2231	0.2241
rs7536608	1	224657092	A	C	0.2061	0.2508	0.2251	0.2255
rs35804313	1	224667102	T	C	0.2082	0.2521	0.2258	0.2262
rs11578709	1	224667131	G	C	0.2076	0.2517	0.2251	0.2251
rs6698343	1	224680332	C	T	0.2079	0.2517	0.2256	0.2261
rs17570292	1	224681894	G	T	0.2089	0.2519	0.226	0.2264
rs11584057	1	224685350	A	T	0.2075	0.2503	0.2237	0.2245

Listed are the effect allele frequencies of the top association signals (Supplementary Table 2) in the GWA meta-analysis in 3,231 drug-resistant vs. 1,578 drug-responsive individuals with focal epilepsy. Significant associations are highlighted in bold. Legend: BP: physical position, A1: effect allele, A2: non-effect allele, FE: focal epilepsy.

Supplementary Table 4: Associations with $P < 5 \times 10^{-8}$ within ± 500 Kb of the lead SNP rs35915186 listed in the NHGRI-EBI GWAS Catalog

SNP	P-value	Mapped genes	GWAS trait	PMID	r ² with rs35915186
rs10916619	3.00E-169	CNIH3	Height	36224396	0.966
rs7542507	6.00E-34	CNIH3	Height	36224396	0.049
rs10916606	2.00E-29	WDR26,CNIH4	Height	30595370	0.944
rs35571080	1.00E-28	CNIH3	White blood cell count	32888493	0.966
rs12062871	5.00E-27	CNIH3	White blood cell count	32888493	0.987
rs35571080	1.00E-25	CNIH3	Neutrophil count	32888493	0.966
rs7536608	7.00E-25	CNIH3	Neutrophil count	32888493	0.996
rs10916617	8.00E-25	CNIH3	Neutrophil count	32888494	0.966
rs35571080	4.00E-24	CNIH3	White blood cell count	32888494	0.966
rs28434172	2.00E-21	WDR26	White blood cell count	30595370	0.953
rs145934561	2.00E-19	WDR26	Neutrophil count	34594039	
rs2051105	4.00E-19	DEGS1,FBXO28	Mean platelet volume	32888493	0.021
rs145934561	7.00E-19	WDR26	White blood cell count	34594039	
rs2051105	2.00E-18	DEGS1,FBXO28	Mean platelet volume	32888493	0.021
rs7519734	8.00E-18	CNIH4	Total cholesterol levels	34887591	0.941
rs11586729	4.00E-17	CNIH4	Height	36224396	0.956
rs12756984	4.00E-16	NVL,DEGS1	Total Dihydroceramide levels	35668104	0.014
rs12756984	6.00E-16	NVL,DEGS1	Dihydroceramide (d18:0/24:0) levels	35668104	0.014
rs12076788	1.00E-15	FBXO28,DEGS1	Mean spheric corpuscular volume	32888494	0.010
rs56105022	6.00E-15	CNIH4	Smoking initiation	36477530	0.013
rs7517754	6.00E-15	NVL,CNIH4	Height (standard GWA)	37106081	0.940
rs12125241	1.00E-14	CNIH3	Smoking initiation	36477530	0.014
rs35913393	1.00E-14	CNIH4	Reticulocyte fraction of red cells	32888494	
rs6682551	1.00E-14	DEGS1,FBXO28	Sphingomyelin (d18:0/20:0, d16:0/22:0) levels	36635386	0.016
rs12756984	3.00E-14	NVL,DEGS1	Dihydroceramide (d18:0/22:0) levels	35668104	0.014
rs752521494	3.00E-14	DEGS1	Palmitoyl dihydrosphingomyelin (d18:0/16:0) levels	35347128	
rs12062871	2.00E-13	CNIH3	Red cell distribution width	32888493	0.987
rs35767322	2.00E-13	CNIH3	Neutrophil count	34469753	0.880
rs61732863	2.00E-13	DEGS1	Behenoyl dihydrosphingomyelin (d18:0/22:0) levels	36635386	0.008
rs11452219	3.00E-13	WDR26	Monocyte percentage of white cells	32888494	
rs7519734	3.00E-13	CNIH4	Total cholesterol levels	34887591	0.941
rs6673347	4.00E-13	DEGS1	Dihydroceramide (d18:0/24:1) levels	35668104	0.020
rs6675858	7.00E-13	CNIH4	Mean corpuscular hemoglobin	32888493	0.955
rs12062871	1.00E-12	CNIH3	Red cell distribution width	32888493	0.987
rs6673347	1.00E-12	DEGS1	Red blood cell count	32888494	0.020
rs35913393	5.00E-12	CNIH4	High light scatter reticulocyte percentage of red cells	32888494	
rs6675858	5.00E-12	CNIH4	Total cholesterol levels	33462484	0.955
rs7365151	5.00E-12	CNIH3	Mean corpuscular hemoglobin	32888494	0.987
rs56105022	9.00E-12	CNIH4	High density lipoprotein cholesterol levels	34887591	0.013
rs12751807	1.00E-11	NVL,CNIH4	Red cell distribution width	32888494	0.707

rs56105022	1.00E-11	CNIH4	Apolipoprotein A1 levels	32203549	0.013
rs6675858	2.00E-11	CNIH4	Mean corpuscular hemoglobin	32888493	0.955
rs6675858	2.00E-11	CNIH4	Mean corpuscular hemoglobin	30595370	0.955
rs12756984	6.00E-11	NVL,DEGS1	Sphingomyelin (d18:0/22:0) levels	35668104	0.014
rs10707541	7.00E-11	WDR26	Hemoglobin A1c levels	34594039	0.586
rs10707541	1.00E-10	WDR26	Glycated hemoglobin levels	33462484	0.586
rs11803981	1.00E-10	CNIH4	Low density lipoprotein cholesterol levels	34887591	0.941
rs35913393	1.00E-10	CNIH4	Reticulocyte count	32888494	
rs4653568	1.00E-10	DEGS1,NVL	Phosphatidylcholine (36:0) levels	35668104	0.013
rs11388086	2.00E-10	CNIH4	Neutrophil side scatter	37596262	
rs12089565	2.00E-10	CNIH3	Drinks per week	36477530	0.062
rs34791963	2.00E-10	CNIH3	Neutrophil percentage of white cells	32888494	0.001
rs4653568	2.00E-10	DEGS1,NVL	Ceramide [N(24)S(19)] product-precursor ratio	33437986	0.013
rs6675858	2.00E-10	CNIH4	Low density lipoprotein cholesterol levels	33462484	0.955
rs10916619	3.00E-10	CNIH3	Height	36224396	0.966
rs12117480	3.00E-10	NVL	Mean corpuscular hemoglobin	32888494	0.013
rs34119581	3.00E-10	CNIH3	Red cell distribution width	30595370	0.516
rs28434172	4.00E-10	WDR26	Basophil percentage of white cells	32888494	0.953
rs56105022	4.00E-10	CNIH4	HDL cholesterol levels	32203549	0.013
rs6663522	5.00E-10	CNIH3	Non-HDL cholesterol levels	34887591	0.965
rs7527044	5.00E-10	WDR26	Lung function (forced vital capacity)	36914875	0.954
rs12130576	6.00E-10	DNAH14	Smoking initiation	36477530	0.002
rs16851979	6.00E-10	CNIH3	Protein quantitative trait loci (liver)	32778093	0.008
rs12129540	8.00E-10	CNIH3	Apolipoprotein A1 levels	33462484	0.016
rs12751807	1.00E-09	NVL,CNIH4	Low density lipoprotein cholesterol levels	32154731	0.707
rs6673347	1.00E-09	DEGS1	Red blood cell count	30595370	0.020
rs35913393	2.00E-09	CNIH4	High light scatter reticulocyte count	32888494	
rs4653568	2.00E-09	DEGS1,NVL	Sphingomyelin (41:0) levels	35668104	0.013
rs73115953	3.00E-09	CNIH3	Drinks per week	36477530	0.059
rs7517754	3.00E-09	NVL,CNIH4	Height (weighted GWA)	37106081	0.940
rs10799590	4.00E-09	CNIH3	Opioid dependence	26239289	0.0004
rs34687215	4.00E-09	WDR26	White blood cell count	27863252	0.930
rs16844823	5.00E-09	DEGS1,FBXO28	Mini-mental state examination / Folstein test (baseline)	35086473	0.004
rs56105022	5.00E-09	CNIH4	High density lipoprotein cholesterol levels	34887591	0.013
rs116524503	6.00E-09	DNAH14	Heel bone mineral density x serum urate levels interaction	34046847	0.000
rs7519734	8.00E-09	CNIH4	Low density lipoprotein cholesterol levels	34887591	0.941
rs16844823	2.00E-08	DEGS1,FBXO28	Mini-mental state examination / Folstein test (baseline)(adjusted for APOE e4 dosage)	35086473	0.004
rs2897048	2.00E-08	WDR26	Lung function (FVC)	30595370	0.959
rs56105022	2.00E-08	CNIH4	HDL cholesterol	34594039	0.013
rs7517754	2.00E-08	NVL,CNIH4	Total cholesterol levels	34594039	0.940
rs7519734	2.00E-08	CNIH4	Non-HDL cholesterol levels	34887591	0.941
rs1533589	3.00E-08	NVL; NVL; NVL; NVL	Core binding factor acute myeloid leukemia	27903959	0.00004

rs145916104	4.00E-08	DNAH14	Metabolonic lactone sulfate levels	34563731	0.001
rs6426153	4.00E-08	CNIH3,CNIH3-AS1	Sphingomyelin (d42:0) levels	35393526	0.012
rs1533589	5.00E-08	NVL; NVL; NVL; NVL	Core binding factor acute myeloid leukemia	27903959	0.00004
rs6662242	5.00E-08	DNAH14,LINC02813	Principal component-derived dietary pattern 42	32193382	0.005

Shown are all association signals with $P < 5 \times 10^{-8}$ listed in the NHGRI-EBI GWAS Catalog for chr1:224154623-225154623 (hg19). Linkage disequilibrium (LD) r^2 values are indicated where available, showing 33 of all association signals in strong LD with rs35915186, the lead SNP of GWA meta-analysis in FE.

Supplementary Table 5: Genetic correlations of the genome-wide meta-analyses of drug resistance in epilepsy with the ILAE epilepsy GWAS¹

Genetic correlations		ILAE 2023 GWAS (PMID: 37653029)		
		All epilepsies [rg, SE]	Focal epilepsy [rg, SE]	Generalised epilepsy [rg, SE]
Drug-resistance GWAS	All epilepsies	-0.73 (0.89)	-0.40 (0.56)	-0.53 (0.61)
	Focal epilepsy	-0.06 (0.27)	-0.22 (0.38)	0.18 (0.31)
	Generalised epilepsy	0.02 (0.36)	0.15 (0.67)	0.09 (0.35)

Shown are the genetic correlations between drug-resistance epilepsy GWAS and the most recent case-control epilepsy GWAS for all, focal, and generalised epilepsy¹. Genetic correlation coefficients (rg), as calculated with LDSC, are displayed with standard errors (SE) in parentheses. None of the genetic correlations were significant (all $P > 0.05$ [regression]).

Supplementary Table 6: EpiPGX cohort samples with a possible genetic cause

Cohort	Phenotype	N individuals with CNV / WES data	N individuals with data that have a possible genetic cause		Percentage of individuals with data that have a possible genetic cause [% of individuals with data]		N unique individuals with CNV or WES data**	N unique individuals with data that have a possible genetic cause (CNV or SNV)	Percentage of individuals with data that have a possible genetic cause [%]
			CNV	SNV*	CNV	SNV			
EpiPGX Drug-resistant	FE	953 / 124	25	0	0.026	0	977	25	0.026
	GE	140 / 54	4	14	0.029	0.26	141	18	0.13
	Epi-NOS	99 / 7	2	0	0.020	0	102	2	0.020
	TOTAL	1192 / 185	31	14	0.026	0.08	1220	45	0.037
EpiPGX Drug-responsive	FE	408 / 125	10	0	0.025	0	418	10	0.024
	GE	159 / 87	4	21	0.025	0.24	160	25	0.16
	Epi-NOS	95 / 28	4	1	0.042	0.036	96	5	0.052
	TOTAL	662 / 240	18	22	0.027	0.092	674	40	0.059

The numbers of possible genetic cases in the EpiPGX cohort are shown by drug response and epilepsy subphenotype classification. Of the 3499 EpiPGX cohort samples, 1854 were screened for epilepsy/seizures-associated CNVs, and 425 had whole-exome sequencing (WES) to screen for rare variants. Rare variant calling from WES was detailed in our previous work⁷. CNVs were called with the Illumina Genome studio plugin cnvPartition⁸ with standard settings. After CNV calling, samples with >3 standard deviations in CNV count, log R ratio SD, B allele frequency mean, and waviness factor from the mean of the calling set were excluded. High-quality CNVs were retained using the following parameters: 1.) covered by >20 markers, 2.) CNV size >20000 bp, 3.) SNP density >0.0001 if size <1Mbp, and 4.) cnvPartition CNV confidence score ≥ 35 . To identify possible genetic causes, we considered all CNVs identified as genome-wide significantly associated with seizures or epilepsy⁹. Rare variants were filtered to 1.) affect established epilepsy genes with an autosomal or X-linked dominant inheritance mode (N=102)¹⁰; 2.) lead to protein or canonical splice site change; 3.) have a maximal population allele frequency in RGC-ME¹¹, RGC-MCPS¹², GnomAD¹³ v4, UK Biobank¹⁴, TogoVar¹⁵, hrcr1¹⁶, GME¹⁷, ABraOM¹⁸, $\leq 10^{-5}$; 4.) a non-reference allele frequency in the sequenced EpiPGX cohort $\leq 10\%$; and 5.) have an in silico pathogenicity prediction score that satisfies at least a moderate ACMG PP3¹⁹ pathogenicity criterion (i.e., REVEL ≥ 0.773 or BayesDel ≥ 0.27)²⁰. Deleteriousness *in silico* prediction scores were annotated from the dbNSFP v4.7 database²¹ using ANNOVAR²². Legend: N: number; FE: focal epilepsy; GE: generalised epilepsy; Epi-NOS: epilepsy, not otherwise specified; VUS: variant of uncertain significance. *Individuals who were carrying an epilepsy-associated CNV and a possible pathogenic SNV were counted only once (in the CNV carrier group). **The union of all individuals with CNV or WES data was counted.

Supplementary Table 7: Epi25 cohort samples with a possible genetic cause

Cohort	Phenotype	N individuals	N individuals that have a possible genetic cause		Percentage of individuals with data that have a possible genetic cause [%]		N individuals with data that have a possible genetic cause (CNV or SNV)	Percentage of individuals that have a possible genetic cause [%]
			CNV	SNV*	CNV	SNV		
Epi25 Drug-resistant	DEE	337	29	38	0.086	0.11	67	0.20
	FE	1429	186	51	0.13	0.036	237	0.17
	GE	327	33	14	0.10	0.043	47	0.14
	Epi-nos	10	1	1	0.10	0.10	2	0.20
	TOTAL	2103	249	104	0.12	0.049	353	0.17
Epi25 Drug-responsive	DEE	107	7	8	0.065	0.075	15	0.14
	FE	579	48	29	0.083	0.050	77	0.13
	GE	518	51	16	0.10	0.031	67	0.13
	Epi-nos	20	0	1	0	0.050	1	0.05
	TOTAL	1224	106	54	0.087	0.044	160	0.13

The numbers of possible genetic cases in the Epi25 cohort are shown by drug response and epilepsy subphenotype classification. All 3327 Epi25 cohort samples had whole-exome sequencing and were screened for epilepsy/seizures-associated CNVs and rare pathogenic variants. CNV and rare variants calling for the Epi25 were detailed in our previous work^{9,23}. To identify possible genetic causes, we considered all CNVs identified as genome-wide significantly associated with seizures or epilepsy⁹. For rare variants, we performed additional filtering, as the scope of our previous study was not to identify monogenic causes but to identify rare variant-burdened genes²³. Rare variants were filtered to 1.) affect established epilepsy genes with an autosomal or X-linked dominant inheritance mode (N=102)¹⁰; 2.) lead to protein or canonical splice site change; 3.) have a maximal population allele frequency in RGC-ME¹¹, RGC-MCPS¹², GnomAD¹³ v4, UK Biobank¹⁴, TogoVar¹⁵, hrcr1¹⁶, GME¹⁷, ABraOM¹⁸, $\leq 10^{-5}$; 4.) a non-reference allele frequency in the Epi25 cohort $\leq 10\%$; and 5.) have an in silico pathogenicity prediction score that satisfies at least a moderate ACMG PP3¹⁹ pathogenicity criterion (i.e., REVEL ≥ 0.773 or BayesDel ≥ 0.27)²⁰. Deleteriousness *in silico* prediction scores were annotated from the dbNSFP v4.7 database²¹ using ANNOVAR²². Legend: N: number; DEE: developmental and epileptic encephalopathy; FE: focal epilepsy; GE: generalised epilepsy; Epi-NOS: epilepsy, not otherwise specified. *Individuals who were carrying an epilepsy-associated CNV and a possible pathogenic SNV were counted only once (in the CNV carrier group).

Supplementary Table 8: Rare variants in *WDR26*, *CNIH3*, and *CNIH4* observed in the GWAS samples which also had whole-exome sequencing

De-identified variant carrier ID	Phenotype	Drug-resistant	AAChange.refGene	GnomAD v4.1 AF exome / genome	REVEL score ²⁴	BayesDel (addAF) score ²⁵	ACMG classification
Epi25 1	DEE	Yes	WDR26:NM_001379403.1:c.C1267T:p.L423F	7.53E-06 / 6.57E-06	0.233	-0.08528	VUS
Epi25 2	FE	Yes	CNIH3:NM_001322303.2:c.G122C:p.R41T	4.58E-05 / 3.94E-05	0.174	-0.15321	VUS
Epi25 3	FE	Yes	CNIH3:NM_001322303.2:c.G190C:p.E64Q	2.60E-05 / 3.29E-05	0.249	-0.00306	VUS
Epi25 4	FE	Yes	CNIH3:NM_001322303.2:c.G190C:p.E64Q	2.60E-05 / 3.29E-05	0.249	-0.00306	VUS
Epi25 5	FE	Yes	CNIH3:NM_001322303.2:c.G428A:p.C143Y	6.84E-07 / NA	0.51	0.24683	VUS
Epi25 6	FE	Yes	CNIH4:NM_001277199.2:c.C61T:p.R21X	1.07E-05 / 6.58E-06	0.378	-0.05004	VUS
Epi25 7	FE	Yes	WDR26:NM_001379403.1:c.C1192G:p.P398A	1.37E-06 / NA	0.483	0.14187	VUS
Epi25 8	FE	Yes	WDR26:NM_001379403.1:c.G1097A:p.R366H	2.05E-06 / 6.57E-06	0.126	0.01533	VUS
Epi25 9	FE	Yes	WDR26:NM_001379403.1:c.G580T:p.A194S	NA	0.062	-0.12765	VUS
Epi25 10	GE	Yes	CNIH4:NM_001277199.2:c.G214A:p.A72T	5.82E-05 / 4.60E-05	0.22	0.00773	VUS
Epi25 11	GE	Yes	WDR26:NM_001379403.1:c.C1403G:p.T468S	6.84E-07 / NA	0.271	0.05827	VUS
EpiPGX 1	GE	Yes	WDR26:NM_001379403.1:c.174_176del:p.S67del	7.79E-06 / NA	NA	NA	VUS
Epi25 12	FE	No	CNIH3:NM_001322303.2:c.C163T:p.H55Y	4.11E-06 / NA	0.201	0.07147	VUS
Epi25 13	FE	No	WDR26:NM_001379403.1:c.C1085T:p.A362V	3.01E-05 / NA	0.207	-0.04492	VUS
Epi25 14	GE	No	CNIH3:NM_001322303.2:c.C287T:p.T96M	7.81E-06 / 1.97E-05	0.382	0.04988	VUS
Epi25 15	GE	No	CNIH3:NM_001322303.2:c.C287T:p.T96M	7.81E-06 / 1.97E-05	0.382	0.04988	VUS
Epi25 16	GE	No	WDR26:NM_001379403.1:c.A1505G:p.Y502C	6.84E-07 / NA	0.61	0.27853	VUS
Epi25 17	GE	No	WDR26:NM_001379403.1:c.G1196A:p.R399Q	NA	0.248	0.05563	Likely Pathogenic
Epi25 18	GE	No	WDR26:NM_001379403.1:c.G556A:p.A186T	2.17E-06 / NA	0.033	-0.17619	VUS

Shown are filtered variants in the genes *WDR26*, *CNIH3*, and *CNIH4* screened in 425 out of 3499 EpiPGX and 3327 out of 3327 Epi25 cohort samples who also had whole-exome sequencing. Variants were annotated using ANNOVAR²². Variants were filtered for: 1.) a maximal population allele frequency in RGC-ME¹¹, RGC-MCPS¹², GnomAD¹³ v4, UK Biobank¹⁴, TogoVar¹⁵, hrcr1¹⁶, GME¹⁷, ABraOM¹⁸, $\leq 10^{-4}$; 2.) protein changing or canonical splice site change; 3.) non-reference allele frequency in the sequenced cohort $\leq 10\%$. Deleteriousness *in silico* prediction scores were annotated from the dbNSFP v4.7 database²¹. As no variant displayed strong predictions for pathogenicity, we also generated American College of Medical Genetics (ACMG)¹⁹-level pathogenicity classification labels, with the caveat that none of the genes has established gene-disease relationships corresponding to the clinical phenotypes of the carriers. Legend: FE: focal epilepsy; GE: generalised epilepsy; DEE: developmental and epileptic encephalopathy; VUS: variant of uncertain significance.

Supplementary Table 9: sQTL-based TWAS in drug-resistant vs. drug-responsive FE

Gene name	GTEX Intron ID (hg38 positions) (Transcript)	P-value	N	P _i _best	T _i _best	Z _{MEAN}
<i>CNIH3</i>	intron_1_224456921_224459124 (transcript with undefined CDS)	8.94E-08	1	8.94E-08	Brain_Cerebellum	5.35
<i>WDR26</i>	intron_1_224401069_224402004 (alternative transcript ENST00000486652.5, intron 10)	1.13E-07	1	1.13E-07	Brain_Cerebellum	5.30
<i>WDR26</i>	intron_1_224401069_224404430 (canonical transcript ENST00000414423.9, intron 8)	1.13E-07	1	1.13E-07	Brain_Cerebellum	-5.30
<i>WDR26</i>	intron_1_224389860_224393828 (canonical transcript ENST00000414423.9, intron 13)	1.14E-07	3	1.14E-07	Brain_Cortex	-5.30
<i>CNIH3</i>	intron_1_224454343_224454449 (transcript with undefined CDS)	1.18E-07	2	9.26E-08	Brain_Cerebellar_Hemisphere	5.29
<i>CNIH3</i>	intron_1_224454343_224456859 (transcript with undefined CDS)	1.48E-07	5	8.94E-08	Brain_Nucleus_accumbens_basal_ganglia	5.25
<i>WDR26</i>	intron_1_224418416_224419518 (canonical transcript ENST00000414423.9, intron 5)	1.86E-07	3	1.86E-07	Brain_Cerebellum	-5.21
<i>CNIH3</i>	intron_1_224456921_224459213 (transcript with undefined CDS)	1.95E-07	12	8.94E-08	Brain_Nucleus_accumbens_basal_ganglia	4.99
<i>CNIH4</i>	intron_1_224371423_224375795 (alternative transcript ENST00000465271.6, intron 4)	2.05E-07	9	2.05E-07	Brain_Amygdala	-5.19
<i>CNIH4</i>	intron_1_224371423_224379069 (canonical transcript ENST00000366858.7, intron 3)	2.05E-07	9	2.05E-07	Brain_Anterior_cingulate_cortex_BA24	-5.19
<i>CNIH3</i>	intron_1_224451208_224451871 (transcript with undefined CDS)	2.10E-07	2	2.10E-07	Brain_Cerebellar_Hemisphere	-5.19
<i>CNIH3</i>	intron_1_224566264_224583164 (transcript with undefined CDS)	2.10E-07	1	2.10E-07	Brain_Hippocampus	5.19
<i>CNIH3</i>	intron_1_224434862_224454278 (transcript with undefined CDS)	2.13E-07	4	2.13E-07	Brain_Caudate_basal_ganglia	5.19
<i>CNIH3</i>	intron_1_224434862_224439634 (transcript with undefined CDS)	2.13E-07	2	2.13E-07	Brain_Frontal_Cortex_BA9	5.19
<i>CNIH3</i>	intron_1_224456921_224459143 (transcript with undefined CDS)	2.33E-07	12	8.94E-08	Brain_Hippocampus	4.77
<i>CNIH3</i>	intron_1_224439774_224451871 (transcript with undefined CDS)	2.34E-07	1	2.34E-07	Brain_Cerebellum	5.17
<i>CNIH4</i>	intron_1_224356993_224360495 (canonical transcript ENST00000366858.7, intron 1)	2.37E-07	7	2.37E-07	Brain_Putamen_basal_ganglia	-5.17
<i>CNIH4</i>	intron_1_224356993_224364109 (alternative transcript ENST00000366860.9, intron 1)	2.37E-07	7	2.37E-07	Brain_Anterior_cingulate_cortex_BA24	5.17

TWAS *P*-values were calculated using S-MultiXcan⁴ with MASHR models for GTEx v8 sQTLs across 13 brain-specific tissues. Shown are all splicing events found to be significantly associated with drug response in FE after Bonferroni correction for 132,272 tested splicing events ($\alpha=3.78 \times 10^{-7}$). Legend: P-value: significance p-value of S-MultiXcan association [multivariate regression with F-test], N: number of "tissues" available for this gene, P_i_best: best p-value of single-tissue S-PrediXcan association, T_i_best: name of best GTEx v8 single-tissue S-PrediXcan association, Z_{MEAN}: mean z-score among single-tissue S-PrediXcan associations.

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