Enhancing Discovery of Genetic Variants for Posttraumatic Stress Disorder Through Integration of Quantitative Phenotypes and Trauma Exposure Information

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ABSTRACT

BACKGROUND: Posttraumatic stress disorder (PTSD) is heritable and a potential consequence of exposure to traumatic stress. Evidence suggests that a quantitative approach to PTSD phenotype measurement and incorporation of lifetime trauma exposure (LTE) information could enhance the discovery power of PTSD genome-wide association studies (GWASs).

METHODS: A GWAS on PTSD symptoms was performed in 51 cohorts followed by a fixed-effects meta-analysis ($N = 182,199$ European ancestry participants). A GWAS of LTE burden was performed in the UK Biobank cohort ($N = 132,988$). Genetic correlations were evaluated with linkage disequilibrium score regression. Multivariate analysis was performed using Multi-Trait Analysis of GWAS. Functional mapping and annotation of leading loci was performed with FUMA. Replication was evaluated using the Million Veteran Program GWAS of PTSD total symptoms.

RESULTS: GWASs of PTSD symptoms and LTE burden identified 5 and 6 independent genome-wide significant loci, respectively. There was a 72% genetic correlation between PTSD and LTE. PTSD and LTE showed largely similar patterns of genetic correlation with other traits, albeit with some distinctions. Adjusting PTSD for LTE reduced PTSD heritability by 31%. Multivariate analysis of PTSD and LTE increased the effective sample size of the PTSD GWAS by 20% and identified 4 additional loci. Four of these 9 PTSD loci were independently replicated in the Million Veteran Program.

CONCLUSIONS: Through using a quantitative trait measure of PTSD, we identified novel risk loci not previously identified using prior case-control analyses. PTSD and LTE have a high genetic overlap that can be leveraged to increase discovery power through multivariate methods.

POSTTRAUMATIC STRESS DISORDER (PTSD) MAY DEVELOP AFTER EXPOSURE TO TRAUMATIC LIFE EVENTS. PTSD CAN SEVERELY IMPACT THE MENTAL AND PHYSICAL HEALTH OF AFFECTED INDIVIDUALS AND IMPAIR THEIR INTERPERSONAL RELATIONSHIPS (1). WHILE THE ESTIMATED COMMUNITY PREVALENCE OF PTSD IN THE UNITED STATES IS 5% TO 10% (2), THE RATE OF PTSD DIFFERS BASED ON THE NATURE OF TRAUMA (3) AND OTHER ENVIRONMENTAL (4) AND GENETIC (5) FACTORS. IDENTIFYING THE BIOLOGICAL MECHANISMS ASSOCIATED WITH THE ETIOLOGY OF PTSD WILL FACILITATE THE DISCOVERY OF BIOMARKERS FOR SCREENING AND DIAGNOSTIC PURPOSES (7) AND THE DEVELOPMENT OF NEW TREATMENTS.

GENOME-WIDE ASSOCIATION STUDIES (GWAS) FACILITATE BIOLOGICAL UNDERSTANDING OF PTSD (8,9), BUT ARE WELL KNOWN TO BE LIMITED BY STATISTICAL POWER TO IDENTIFY RISK VARIATION (10). QUANTITATIVE MEASURES OF PTSD ENHANCE DISCOVERY POWER OVER BINARY TRAIT DEFINITIONS (9,11). APPROPRIATELY ACCOUNTING FOR TRAUMA EXPOSURE HYPOTHETICALLY ENHANCES POWER, AS INDIVIDUALS WILL NOT DEVELOP THE DISORDER UNLESS THEY ARE EXPOSED TO TRAUMA, REGARDLESS OF HIGH GENETIC VULNERABILITY FOR PTSD (12,13). MOREOVER, THE NOTION THAT GENETIC VARIANTS CAN PREDISPOSE TO TRAUMA EXPOSURE IS ONLY STARTING TO BE EXPLORED (14).

As trauma exposure is a prerequisite for the development and manifestation of PTSD, investigating the genetics of trauma exposure will hypothetically lead to a clearer picture of PTSD genetics.

The Psychiatric Genomics Consortium (PGC)–PTSD is a global collaborative effort to study the genetic basis of PTSD through meta-analysis of diverse cohorts (13). Subsequent to a case-control GWAS (8), our collaborators have provided quantitative measures of PTSD and lifetime trauma exposure (LTE). To obtain genomic insights from the quantitative PTSD phenotype, we performed a GWAS of PTSD symptoms in 182,199 participants from the PGC-PTSD Freeze 2 dataset. To determine if accounting for LTE would provide the hypothesized increase in discovery power, we performed a GWAS of PTSD with covariate adjustment for LTE, showing that it lowers PTSD signal. We investigated the possibility that multicollinearity arising from high genetic correlation ($r_g$) of PTSD and LTE was responsible for this result. To perform this investigation, we performed a GWAS of LTE in the most powered and unbiased (15) subsample of the data, 132,988 participants from the UK Biobank (UKBB) (16), then evaluated the $r_g$ of PTSD and LTE. To explore the $r_g$ further, we contrasted the $r_g$s that PTSD and LTE have with other traits. We showed that the high $r_g$s of PTSD and LTE can be leveraged to enhance the power of PTSD GWASs using multivariate methods. We replicated PTSD GWAS findings in the Million Veteran Program (MVP) GWAS of total PTSD symptoms (MVP TOT). We contextualized genomic findings through functional annotation, tissue expression analyses, and phenome-wide association study (PheWAS).

METHODS AND MATERIALS

Study Population and Phenotyping

Participants were drawn from a collection of 51 cohorts within the PGC-PTSD Freeze 2 dataset, as previously described in Nievergelt et al. (8). All participants included in the present study were of genetically estimated European ancestry. PTSD symptoms and LTE were measured within each cohort using structured clinical interviews, self-reported inventories, or clinical evaluation. A summary of the assessment and scoring methods for the various studies is presented in Table S1 in Supplement 2, and a complete description is available in Nievergelt et al. (8). All participants provided written informed consent, and studies were approved by the relevant.
institutional review boards and the University of California San Diego Human Research Protection Program.

**GWAS Quality Control**

Genotyping, quality control (QC), and imputation methods for the included studies have been described in detail (8). In brief, participating cohorts provided phenotype and genotype data or GWAS summary statistics to the PGC-PTSD for quality control and analysis. For studies in which the PGC-PTSD analyst had direct access to genotype data, RICOPILI (17) was used to perform QC and imputation. QC included standard filters for single nucleotide polymorphism (SNP) call rates (exclusion of SNPs with call rate <98% or a missing difference >0.02 between cases and controls), call rate for participant genotypes (samples with <98% call rate excluded), Hardy-Weinberg equilibrium (p < 1 × 10⁻⁶ in controls), and heterozygosity (within ± 0.2). Datasets were phased using SHAPEIT (18) and imputed using IMPUTE2 (19) with the 1000 Genomes Phase 3 reference panel data (20). For the UKBB, QC and imputation were carried out centrally by UKBB investigators as previous described (16) and GWAS was carried out by the PGC-PTSD analyst. For cohorts with data-sharing restrictions, analyses were performed using similar protocols by the study team that had individual-level data access, and GWAS summary statistics were provided to the PGC-PTSD.

**Genome-wide Association Study**

Only unrelated (r < 0.2) participants were retained for analysis. Principal components (PCs) were calculated within each cohort using EIGENSOFT v6.3.4 (21). The PTSD GWAS was performed within cohorts using PLINK 2.0 alpha with the –glm option, with the exception of UKBB and VETS (Vietnam Era Twin Study of Aging) data, which were analyzed using BOLT-LMM v2.3.4 (22). Where available, PTSD symptom scores were analyzed using linear regression (n = 36 cohorts); PTSD case-control status was used if symptom scores were not available, using logistic regression (n = 15 cohorts). In both cases, 5 PCs were included as covariates to account for population stratification and genotyping artifacts. The UKBB PTSD GWAS included an additional PC as well as batch and assessment center covariates. Studies providing summary data used similar analytic strategies, as previously described (8). For each GWAS, SNPs with minor allele frequency <1% or imputation information score <0.6 were excluded. To perform a GWAS of PTSD conditioned on LTE, the GWAS was performed with LTE included as an additional covariate as either a count of LTEs or a binary variable, depending on data availability. The GWAS of the LTE count phenotype in the UKBB sample was performed in BOLT-LMM using 6 PCs, batch, and assessment center as covariates.

**PTSD Meta-analysis**

Sample size-weighted fixed-effects meta-analysis was performed using METAL (23). To account for different analytic methods and measure scales, effect estimates were converted into z scores by dividing effect sizes by standard errors (24). Case-control and quantitative GWAS subsets were evaluated for r₂ to determine if they could be meta-analyzed. To account for differences in ascertainment, heritability, and power between case-control and quantitative subsets, modified sample size weights were derived as previously described (25), assuming 10% population prevalence of PTSD, the estimates of SNP-based heritability (h² SNP), r₂, and sample PTSD prevalence. Meta-analysis was conducted on the reweighted z scores. Only SNPs available in >90% of all samples (N ≥ 163,979) were included in analyses. Regional annotation plots of genome-wide significant loci were produced using Locus-Zoom (26).

**Heritability and Genetic Correlation Estimation With Linkage Disequilibrium Score Regression**

Trait h² SNP and r₂ were estimated from GWAS summary statistics using linkage disequilibrium score regression (27). The linkage disequilibrium score intercept was used to test for inflation of test statistics owing to residual population stratification or other artifacts, and the attenuation factor [(intercept – 1)/mean (y²) – 1]) was used to determine the proportion of inflation of test statistics owing to residual population stratification (Table S2 in Supplement 2). Heritabilities were contrasted using a z test where standard errors were estimated using the block-jackknife approach. To estimate r₂ with other disorders, the LD Hub web interface was used (28). To identify genetic differences between PTSD and LTE, the r₂ s observed for PTSD and LTE were contrasted using z tests, where significance level was determined using Bonferroni correction for the 772 traits tested (p < 6.47 × 10⁻⁵).

**FUMA**

FUMA v1.3.6a (29) was used with the default settings (Supplement 1) to visualize and annotate GWAS results. The FUMA pipeline integrates the MAGMA (30) tool to perform gene-based, gene-pathway, and tissue-enrichment analyses, with significance based on Bonferroni correction. 1000 Genomes Europeans were used as reference genotypes. Tissue-enrichment analysis included Genotype-Tissue Expression (GTEx) v8 expression data (31).

**Cis-Quantitative Trait Locus Mapping**

The effects of GWAS loci on transcriptomic regulation of surrounding genes (locus within ± 1 Mb of the gene transcription starting site) were tested for 49 tissues in GTEx v8 with genome-wide false discovery rate correction applied. Using the same criteria, GTEx v8 data were also used to investigate the effects of GWAS loci on the regulation of alternative splicing isoforms. A detailed description regarding GTEx v8 quantitative trait locus (QTL) mapping data by the GTEx Consortium is available (32). Briefly, cis-expression QTL (eQTL) and cis-splicing QTL mapping was performed using FastQTL (33) including the top 5 genotyping PCs, probabilistic estimation of expression residuals factors (34), sequencing platform, sequencing protocol, and sex as covariates.

**Replication Analysis**

Summary data from MVP TOT (dbGaP study accession phs001672.v4.p1) was used to replicate GWAS results. MVP TOT included 186,689 European ancestry participants who completed the PTSD Checklist–Civilian Version and passed QC. Details of MVP TOT have been published (35). SNPs were
deemed replicated in MVP\textsubscript{TOT} if they had matching effect direction and were nominally significant after Bonferroni correction for the 9 SNPs tested ($p < .006$).

### Multi-Trait Analysis of GWAS

Multi-Trait Analysis of GWAS (MTAG) (36) performs multivariate analysis of genetically correlated traits to increase discovery power for each input trait, providing trait-specific effect estimates and $p$ values. MTAG was used to perform multivariate analysis with PTSD and LTE GWASs. The maxFDR statistic was used to test for MTAG model assumptions (Supplement 1).

### Phenome-wide Association Study

To understand further how functional changes of significant loci are associated with human traits and diseases, we conducted a Phenome-WAS of leading SNPs from PTSD and LTE loci using data from the GWAS Atlas (available at https://atlas.ctglab.nl/) (37). Bonferroni correction was applied to account for the 4756 phenotypes available that were tested ($p < 1.05 \times 10^{-5}$).

### RESULTS

The PTSD GWAS meta-analysis included 182,199 participants of European ancestry from 51 cohorts (Table S1 in Supplement 2). The largest cohort was the UKBB ($N = 134,586$ participants). Across the cohorts, PTSD was assessed using a variety of different methods ($n = 19$ methods); the most common methods were versions of the Clinician-Administered PTSD Scale ($n = 18$ studies) and PTSD Checklist ($n = 14$ studies). The majority of participants ($91\%$, $n = 165,825$) were analyzed based on PTSD symptom scores; the remaining participants ($9\%$, $n = 16,374$, 15 studies) did not have symptom scores available and were analyzed based on PTSD case-control status.

### PGC-PTSD GWAS Meta-analysis

The $h^2_{\text{SNP}}$ of meta-analysis of cohorts analyzed by symptom scores was $0.0547$ (SE = $0.0042$, $p = .00004$) in the PGC-PTSD GWAS. The $h^2_{\text{SNP}}$ in the larger meta-analysis of case-control cohorts was $0.0580$ (SE = $0.0259$, $p = .17$). The $r_g$ between the symptom score and case-control analyses was very high ($r_g = 0.9646$, SE = $0.036$, $p = .0074$). Thus, symptom score and case-control GWASs were meta-analyzed. We identified 5 genome-wide significant loci (Table 1, Figure 1A). Leading variants in significant loci mapped to an intergenic locus on chromosome 1, the intronic region of the GABBR1 gene on chromosome 6, the intronic regions of MPPE6 and DFNAS on chromosome 7, an intron of FOXP2 on chromosome 7, and the intronic region of FAM120A on chromosome 9. Gene-based analysis identified 6 significant genes (DCAF5, EXD2, FAM120A, FOXP2, GALNT16, and PHF2) (Table S3 in Supplement 2).

### PGC-PTSD GWAS Covariate Adjustment for LTE

We repeated the GWAS of PTSD with covariate adjustment for LTE. $h^2_{\text{SNP}}$ was $0.0389$ (SE = $0.00340$, $p = 2.6 \times 10^{-30}$), $31\%$ lower than the PTSD GWAS without LTE covariate adjustment ($p = 8.6 \times 10^{-26}$). There was a genome-wide significant locus in an uncharacterized region, CTC-340A15.2, on chromosome 5 that was not identified in the PTSD GWAS (Table S4 in Supplement 2). Effects changed slightly for the loci previously identified in the unadjusted PTSD GWAS (Table S4 in Supplement 2). Gene-based analysis identified no significant genes.

### UKBB LTE GWAS

We performed GWAS of LTE count in the UKBB subset of the PGC-PTSD GWAS data ($132,988$ UKBB participants). Of participants, $30.9\%$ reported 1 LTE, $14.8\%$ reported 2 LTES, $6.3\%$ reported 3 LTEs, and $3.3\%$ reported 4 or more LTES (Table S5 in Supplement 2). The $h^2_{\text{SNP}}$ of LTE count was $0.0734$ ($SE = 0.005$, $p = 8.7 \times 10^{-48}$). Six loci showed genome-wide significance (Figure 1B, Table 2). Leading variants in significant loci mapped to an intron of PRUNE on chromosome 1, the intron of noncoding RNA AC068490.2 on chromosome 2, the intronic regions of SGCD on chromosome 5, an intron of FOXP2 on chromosome 7 (also identified in the PGC-PTSD GWAS), an intergenic region in chromosome 14 near MDGA, and the smaller meta-analysis of case-control cohorts (observed scale $h^2_{\text{SNP}} = 0.0580$, SE = $0.0259$, $p = .17$).

### Table 1. Genome-wide Significant Loci From PTSD GWASs and MTAGs With Replication in MVP\textsubscript{TOT} GWAS

| Analysis | rsID | Chr | Position$^a$ | A1 | A2 | A1 Freq | A2 Freq | z Score | $p$ Value | A1 freq | z Score | $p$ Value$^b$ |
|----------|------|-----|--------------|----|----|---------|---------|---------|----------|----------|---------|---------|----------|
| Identified in GWAS | rs72657988 | 1 | 35688541 | T | G | 0.08 | 0.92 | 6.44 | $1.2 \times 10^{-10}$ | 0.34 | 9.4 | $10^{-8}$ | 0.07 | 2.18 | .029 |
| | rs146918648 | 6 | 28548674 | A | G | 0.04 | 0.96 | 6.04 | $1.5 \times 10^{-9}$ | 6.50 | 8.0 | $10^{-11}$ | 0.04 | 2.00 | .045 |
| | rs27218161$^c$ | 2 | 24699329 | A | G | 0.83 | 0.17 | -5.27 | $1.4 \times 10^{-7}$ | -5.80 | 6.5 | $10^{-9}$ | 0.82 | -1.45 | .15 |
| | rs100626927 | 7 | 114143407 | T | C | 0.59 | 0.41 | 5.38 | $7.4 \times 10^{-8}$ | 6.72 | 1.8 | $10^{-11}$ | 0.59 | 4.97 | 6.7 | $10^{-7}$ |
| | rs10821140 | 9 | 96253169 | A | C | 0.35 | 0.65 | -5.71 | $1.2 \times 10^{-6}$ | -6.02 | 1.8 | $10^{-9}$ | 0.34 | -3.89 | 1.0 | $10^{-4}$ |

| Identified in MTAG | rs4557006 | 2 | 22443840 | A | G | 0.45 | 0.55 | 4.26 | $2.0 \times 10^{-5}$ | 5.83 | 5.7 | $10^{-9}$ | 0.45 | 5.53 | 3.2 | $10^{-8}$ |
| | rs1504930 | 5 | 155852066 | A | G | 0.62 | 0.38 | -4.26 | $2.0 \times 10^{-5}$ | -5.58 | 2.5 | $10^{-8}$ | 0.62 | -4.20 | 2.7 | $10^{-5}$ |
| | rs8059002 | 16 | 25417390 | T | G | 0.86 | 0.14 | -4.43 | $9.3 \times 10^{-6}$ | -5.46 | 4.8 | $10^{-8}$ | 0.85 | -1.50 | .13 |
| | rs7264419 | 20 | 47070309 | A | G | 0.75 | 0.25 | -5.06 | $4.1 \times 10^{-7}$ | -5.85 | 5.0 | $10^{-9}$ | 0.76 | 0.55 | .58 |

A1, allele 1 (coded); A2, allele 2; Chr, chromosome; GWAS, genome-wide association study; MTAG, Multi-Trait Analysis of GWAS; MVP\textsubscript{TOT}, MVP total PTSD symptoms; PGC-PTSD, Psychiatric Genomics Consortium–posttraumatic stress disorder; rsID, reference SNP ID number.

$^a$Base pair position on chromosome (hg19/GR37 Human Genome Build).

$^b$Significant in MVP if $p < .006$ (Bonferroni-corrected for 9 loci).

$^c$Linkage disequilibrium proxy for rs2721817, the leading single nucleotide polymorphism in this locus.
upstream of CCDC8 on chromosome 19. Gene-based analysis identified SGCD (chromosome 5: 155,297,354–156,194,799 base pairs, 2965 SNPs, 99 parameters, $z = 5.53$, $p = 1.5 \times 10^{-28}$) and C20orf112 (chromosome 20:31,030,862–31,172,876 base pairs, 296 SNPs, 21 parameters, $z = 4.73$, $p = 1.13 \times 10^{-26}$). GWAS of LTE count weighted by trauma-specific PTSD

**Figure 1.** Manhattan plots of genome-wide association study (GWAS) associations. The x-axis is the position on the genome, ordered by chromosome and base-pair position. The y-axis is the $-\log_{10} p$ value of association. Each dot represents the association between a given single nucleotide polymorphism and the trait. Colors alternate between chromosomes, with odd chromosomes colored blue and even chromosomes colored teal. (A) Results of posttraumatic stress disorder GWASs. (B) Results of lifetime trauma exposure GWASs. (C) Posttraumatic stress disorder–specific results of MTAG (Multi-Trait Analysis of GWAS) analysis of posttraumatic stress disorder and lifetime trauma exposure.

**Table 2. Genome-wide Significant Loci From GWASs of LTE**

| rsID      | Chr | Position* | A1 | A2 | A1 Frequency | $z$ Score | $p$ Value |
|-----------|-----|-----------|----|----|--------------|-----------|-----------|
| rs6661135 | 1   | 150999414 | C  | T  | 0.93         | −5.52     | $3.3 \times 10^{-8}$ |
| rs4665501 | 2   | 22546151  | G  | T  | 0.44         | −5.77     | $7.7 \times 10^{-9}$ |
| rs4704792 | 5   | 155757946 | A  | T  | 0.26         | 5.75      | $9.2 \times 10^{-9}$ |
| rs1476535 | 7   | 114071035 | C  | T  | 0.44         | −5.77     | $8.0 \times 10^{-9}$ |
| rs2933196 | 14  | 47285917  | G  | A  | 0.59         | −5.51     | $3.6 \times 10^{-8}$ |
| rs770444611| 19  | 46917381  | INS| T  | 0.59         | 5.66      | $1.5 \times 10^{-8}$ |

A1, allele 1 (coded); A2, allele 2; Chr, chromosome; GWAS, genome-wide association study; LTE, lifetime trauma exposure; rsID, reference SNP ID number.

*Base pair position on chromosome (hg19/GR37 Human Genome Build).

*Insertion of TGAGGCCAGGAGTTC.
Extended Phenotyping Identifies PTSD Genetic Variants

Genetic Overlap Between LTE and PTSD

The $r_g$ between PTSD and LTE was high ($r_g = 0.7239, p < 1 \times 10^{-16}$). To explore this genetic overlap, we contrasted patterns of $r_g$ of PTSD and LTE to other traits. Testing 772 human traits and diseases, we observed 269 and 217 $r_g$s that survived Bonferroni adjustment. Noncolored (gray) dots indicate traits where genetic correlation with PTSD than with LTE. Red dots indicate traits with significantly higher genetic correlation with PTSD than with LTE. The top 5 traits with a significantly higher correlation to PTSD than LTE and top trait with significantly higher correlation to LTE than PTSD have been labeled.

Multivariate Analysis of PTSD and Trauma Exposure

MTAG analysis that combined PTSD GWAS meta-analysis and UKBB LTE GWAS reported an effective sample size increase of PTSD GWAS from 182,199 to 217,491. There were 8 genome-wide significant loci for the MTAG PTSD analysis, including 4 loci not identified in the PTSD GWAS meta-analysis (Table 1, Figure 1C). Leading variants from additional loci mapped to an intergenic region in chromosome 2, the intron of SGCD on chromosome 5, an intergenic region on chromosome 16 near ZKSCAN2 and AQP8, and the intron of STAUF1 on chromosome 20. In gene-based analysis, there were 8 significant genes, including 5 genes not identified from the original GWAS gene-based analysis (CE11, DIP5A5, FOXP1, SGCD, TRIM26) (Table S3 in Supplement 2).

Cross-Replication in MVPTot

Of the 9 loci identified across the PTSD GWASs (5 from the PGC GWAS and 4 from the MTAG), 4 replicated significantly in MVPTot ($p < .006$) (Table 1, Figures S2–S10 in Supplement 1). Of the 11 genes identified in gene-based analyses (6 in the GWAS and 5 in the MTAG), 7 replicated at least at a nominally significant level in MVPTot (Table S3 in Supplement 2). Additionally, of 15 loci identified in MVPTot GWASs, 9 nominally replicated in PGC-PTSD (Table S7 in Supplement 2). Overall, $r_g$ between PGC-PTSD and MVPTot was high ($r_g = 0.8359, SE = 0.0376, p = 2.5 \times 10^{-10}$).

Functional Consequences of Risk Loci

We examined the functional impact of the 9 GWAS variants associated with PTSD (5 from the GWAS and 4 from the MTAG) (Table 1). We observed that 7 loci were related to multiple tissue-specific eQTLs (Table S8 in Supplement 2), where 11% of false discovery rate–significant eQTLs were in brain regions. A similar trend was present for splicing QTLs (Table S9 in Supplement 2), where only 7% of gene-tissue combinations were related to brain regions. Further details of the eQTL analysis are provided in Supplement 1.

We found enrichment of genes involved in brain transcriptomic regulation in PTSD (Table S10 in Supplement 2). All brain regions tested were at least nominally significant, with several remaining significant after Bonferroni correction (MTAG: cortex $p = 2.9 \times 10^{-9}$, frontal cortex Brodmann area (BA) 9 $p = 3.53 \times 10^{-5}$, cerebellum $p = 1.09 \times 10^{-4}$, anterior cingulate cortex BA 24 $p = 1.29 \times 10^{-4}$, cerebellar hemisphere $p = 1.43 \times 10^{-3}$, nucleus accumbens/basal ganglia $p = 3.6 \times 10^{-4}$). There was no significant enrichment detected in any sets from the list of curated gene sets and Gene Ontology terms (Table S11 in Supplement 2).

Phenome-wide Association Study

We identified 200 phenome-wide significant associations (Table S12 in Supplement 2), with more than half of the significant associations related to two domains: psychiatry (34%) and metabolism (18%). The strongest PheWAS associations with PTSD and LTE loci included height and body mass phenotypes, educational attainment, social interaction, sexual activity, risk tolerance, and sleep phenotypes (Supplement 1). Several PTSD loci showed widespread pleiotropy across multiple psychiatric traits: rs10266297 (35 significant
associations, 40% psychiatric domain, top psychiatric result: risk taking $p = 1.27 \times 10^{-11}$, rs10821140 (37 significant associations, 38% psychiatric domain, top psychiatric result: loneliness $p = 1.11 \times 10^{-11}$, rs146918648 (44 significant associations, 48% psychiatric domain, top psychiatric result: tenseness/restlessness $p = 2.13 \times 10^{-5}$).

**DISCUSSION**

Our GWASs aimed to advance understanding of PTSD genetics by integrating quantitative PTSD phenotypes and LTE exposure information in 182,199 participants of European ancestry from 51 cohorts. Overall, quantitative PTSD phenotyping captured similar genetic signal to our prior case-control analysis ($r_g = 0.92–1.14$) (8), but with substantially higher power. However, by using LTE as a covariate, which hypothetically accounts for unexpressed genetic vulnerability among unexposed participants (12), we found a significant reduction in heritability and gene discovery. As high $r_g$ between PTSD and LTE would be one hypothetical explanation for this result (i.e., multicollinearity), we performed a GWAS of LTE and contrasted it to GWAS results for PTSD. We found that LTE has $h^2_{SNP}$ comparable to PTSD and high $r_g$ compared with PTSD. We leveraged the $r_g$ to significantly enhance PTSD discovery power using a multivariate approach (36).

One explanation for $h^2_{SNP}$ of PTSD adjusted for LTE being lower than the unadjusted estimate is that it may have removed genetic effects on PTSD mediated by trauma exposure (12,13). Given that trauma is a prerequisite for PTSD, genetic effects on trauma exposure can have mediated (i.e., indirect) effects on PTSD. Indeed, this seems plausible, as our LTE GWAS suggested a substantial amount of $h^2_{SNP}$ related to trauma exposure. Therefore, the estimated $h^2_{SNP}$ of PTSD conditional on LTE would theoretically reflect only nonmediated (i.e., direct) effects and thus would be smaller.

We used $r_g$ to quantify the genetic overlap between LTE and PTSD, finding similar magnitude to findings from twin studies (5,6). At the same time, incomplete $r_g$ between these two phenotypes also suggested meaningful genetic differences. To investigate this, we contrasted the magnitudes of $r_g$ that PTSD and LTE shared with other traits. For most traits, $r_g$ with PTSD was quite similar in magnitude to $r_g$ with LTE. However, we also found that negative affect traits, such as neuroticism and irritability, were more strongly correlated with PTSD than LTE, whereas risk-taking behavior showed higher correlation with LTE than PTSD. This suggests that some variants influence PTSD and LTE through somewhat distinct psychological and behavioral mechanisms (5).

The high $r_g$ between PTSD and LTE facilitates the application of multivariate approaches to PTSD GWASs. Whereas the $r_g$ between PTSD and LTE induces loss of power in the PTSD analysis when conditioned on LTE, a multivariate approach can benefit from it. Our multivariate (36) analysis resulted in a 19% increase in the effective sample size by adding LTE count data from the UKBB and identified replicable loci and patterns of tissue expression not identified in a standard PTSD GWAS.

The biological mechanisms associated with several of the protein products of identified genes have been linked to PTSD pathophysiology in animal and cell models: amygdala-mediated fear extinction [FAM120A (38)], neuronal transcriptional regulation [FOXP2 (39)], brain excitatory/inhibitory balance [ARFGEF2, GABBR1, STAUI1 (40)], intracellular vesicular trafficking and other synaptic activities [ARFGEF2 (41), MPP6 (42), SEMA6C (43), SGCD (44)], and inflammation [HIAT1, TRIM26 (45), TRIM27 (46), ZMYM4, ZNF165 (47)]. Blood and brain transcription-wide association and differential gene expression studies of PTSD have also implicated some of these genes, including a blood-based prediction of downregulation of ARFGEF2 in the dorsolateral prefrontal cortex (48) and a postmortem study of human PTSD cortex indicating downregulation of CTSS expression in the dorsal anterior cingulate cortex and downregulation of OSBPRL3 expression in the dorsolateral prefrontal cortex (49).

Interestingly, PTSD loci show widespread pleiotropic associations in PheWAS (50–52). Some loci point to factors associated with existing clinical presentations of PTSD (e.g., sleep), while others point to potential risk/protective factors for PTSD, such as educational attainment and cognitive functioning. Loci may affect PTSD through their direct influence on these risk/protective factors. Alternatively, the high degree of pleiotropy shown by these loci suggests that they could influence PTSD risk through a more general alteration of biological function (37), such as general predisposition to psychiatric illness (53). In particular, metabolic phenotypes such as height and body mass also appeared to be enriched in our PheWAS. This could be the influence of these loci on previously implicated inflammatory mechanisms for PTSD (8) or simply an artifact of their overrepresentation in the GWAS Atlas. Nevertheless, the broad variety of behavioral and clinical domains associated with these loci suggest complex etiologic heterogeneity of PTSD that could relate to subtypes (54).

Further characterization of significant loci via eQTL analyses identified expression across a variety of tissue types. Given the high degree of shared eQTL architecture between tissues, the presence of some of these tissues might not be directly related to PTSD pathogenesis. Indeed, on the genome-wide level, our tissue enrichment analysis suggests that only brain tissues are relevant. The brain regions implicated are consistent with functional magnetic resonance imaging and structural magnetic resonance imaging findings of PTSD. BA 24 (as part of the ventral anterior cingulate cortex) is implicated in PTSD response to trauma-, fear-, and threat-related stimuli (55,56). BA 9 (as part of the dorsomedial prefrontal cortex) reflects response to self-referential thought, theory of mind, empathy, and moral judgments and shows greater engagement in people with PTSD and trauma-exposed individuals (55,57,58). Nucleus accumbens expression is consistent with the neuroimaging evidence of its role in the reward system, which is prominently affected with emotional numbing symptoms of PTSD (59–62).

**Limitations**

Stress-related disorders are phenotypically complex and heterogeneous (63), which limits discovery power and complicates translation to clinical application. The strategies proposed for understanding and addressing heterogeneity in major depressive disorder, such as harmonization of measures, additional phenotypic measures, and investigations of subtypes, could be applied to PTSD as additional avenues to enhance discovery power (64). Sex differences may also contribute a significant
source of heterogeneity (8,65–68). Our analyses were restricted to participants of European ancestry given power limitations for other ancestry groups. However, urgent scientific and ethical reasons call for extending analyses to individuals of non-European ancestry (69). The PGC-PTSD group has actively been gathering data to increase representation from diverse ancestry and developing methods to optimize analyses in admixed populations (70). As sample sizes increase, future investigations will be powered to investigate ancestry and sex-specific genetic influences on PTSD and trauma exposure. In performing a GWAS of cumulative LTE, we identified several significant loci, including loci previously identified in GWASs of childhood trauma exposure (14). A full investigation of the genetic basis of LTE is clearly warranted. Future work could also examine the relationship between PTSD and specific types or numbers of trauma exposure, as they plausibly have different relationships with PTSD (6) and may therefore be more informative than our cumulative measure for LTE. Finally, trauma was assessed via participant self-report, which may vary with mood and PTSD symptoms at the time of reporting (71) and could inflate genetic associations with PTSD.

Conclusions

Novel replicable risk loci for PTSD identified by incorporating quantitative symptom data and trauma exposure information into GWASs offer us new insights into the genetic architecture of PTSD. Beyond the nature of LTE as an environmental exposure, there is a heritable component to LTE that overlaps highly with PTSD to impart an enhanced understanding of PTSD genetics. In future investigations, the genetic architectures of PTSD and LTE could be further delineated using causal mediation analysis (72), which can provide estimates of LTE-related mediation and gene-by-environment interaction. Our results reinforce the notion that in addition to larger LTE-related mediation and gene-by-environment interaction. Our results reinforce the notion that in addition to larger

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