Title: Analysis of Shared Heritability in Common Disorders of the Brain

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One Sentence Summary: Comprehensive heritability analysis of brain phenotypes demonstrates a clear role for common genetic variation across neurological and psychiatric disorders and behavioral-cognitive traits, with substantial overlaps in genetic risk.

Abstract: Disorders of the brain exhibit considerable epidemiological comorbidity and frequently share symptoms, provoking debate about the extent of their etiologic overlap. We quantified the genetic sharing of 24 brain disorders and 16 phenotypes based on summary statistics from genome-wide association studies. Psychiatric disorders show substantial sharing of common variant risk, while neurological disorders appear more distinct from one another. We observe limited evidence of sharing between neurological and psychiatric disorders, but do identify sharing between several quantitative measures and brain disorders. We also performed extensive simulations to explore power and the impact of diagnostic misclassification and phenotypic heterogeneity. These results highlight the importance of common genetic variation as a source of risk for brain phenotypes and the value of heritability-based methods in understanding their etiology.

Main Text:

The classification of brain disorders has evolved over the past century, reflecting the medical and scientific communities’ best assessments of the presumed root causes of clinical phenomena such as behavioral change, loss of motor function, spontaneous movements or alterations of consciousness. A division between neurology and psychiatry developed, with the more directly observable phenomena (such as the presence of emboli, protein tangles, or unusual electrical activity patterns) generally defining the neurological disorders(1). Applying modern methods to understand the genetic underpinnings and categorical distinctions between brain disorders may be helpful in informing next steps in the search for the biological pathways that underlie their pathophysiology(2, 3).

In general, brain disorders (here excepting those caused by trauma, infection or cancer) show substantial heritability from twin and family studies (4). Epidemiological and twin studies have explored patterns of phenotypic overlaps(5-7), and substantial comorbidity has been reported for many pairs of disorders, including bipolar disorder-migraine(8), stroke-major depressive disorder(MDD)(9), epilepsy-autism spectrum disorders (ASD) and epilepsy-attention deficit hyperactivity disorder (ADHD)(10). Furthermore, neurological and psychiatric research has shown that mutations in the same ion channel genes confer pleiotropic risk for multiple distinct brain phenotypes(11-13). Recently, genome-wide association studies (GWAS) have demonstrated that individual common risk variants show overlap across traditional diagnostic boundaries (14,
and that disorders like schizophrenia, MDD and bipolar disorder can have strong genetic correlations. GWAS have also demonstrated that common genetic variation substantially contributes to the heritability of brain disorders. In most cases, this occurs via many common variants with small risk effects, as has been the case with Alzheimer’s disease, bipolar disorder, migraine, Parkinson’s disease, and schizophrenia. In addition to locus discovery, larger sample sizes enable analyses of shared genetic influences, to improve our understanding of the degree of distinctiveness of brain disorders. Using recently developed heritability-based methods, we can now extend our evaluation of the nature of these diagnostic boundaries and explore the extent of shared common variant genetic influences across a wide set of neurological and psychiatric phenotypes.

**Study design**

We formed the Brainstorm consortium, a collaboration among GWAS meta-analysis consortia of 24 disorders (see Data sources), to perform the first comprehensive heritability and correlation analysis of brain disorders. The study sample consists of 212,367 brain disorder cases and 647,979 controls (Table 1), including every common brain disorder for which we could identify a meta-analysis consortium and reflecting the respective current diagnostic gold standards. This includes at least the most common representative in each ICD-10 block in F20-59 (mental and behavioral disorders) and G20-G47 (diseases of the central nervous system). Also included are 1,113,280 samples for 12 “behavioral-cognitive” phenotypes (n=666,178) chosen for being traditionally viewed as brain-related, and four “additional” phenotypes (n=447,102), selected to represent known, well-delineated etiological processes (e.g. immune disorders [Crohn’s disease] and vascular disease [coronary artery disease]; Table 2).

GWAS summary statistics for the 40 disorders and phenotypes were centralized and underwent uniform quality control and processing. We generated European-only meta-analyses for each disorder to avoid potential biases arising from ancestry differences, as many of the brain disorder datasets included sample sets from diverse ancestries. We have recently developed a novel heritability estimation method, linkage disequilibrium score regression (LDSC), which was used to calculate heritability estimates and correlations, as well as to estimate their statistical significance from block jack-knife-based standard errors. Heritability for binary disorders and phenotypes was transformed to liability-scale. We further performed a weighted-least squares regression analysis to evaluate whether differences relating to study makeup (such as sample size) were correlated with the magnitude of the correlation estimates. We also performed a heritability partitioning analysis using stratified LD score regression to examine whether the observed heritability was enriched in any tissue-specific regulatory partitions of the genome, using the ten top-level tissue-type and 53 functional partitions from Finucane et al.
Finally, simulated phenotype data was generated under several different scenarios by permuting the 120,267 genotyped individuals from the UK Biobank (24) to both evaluate power and aid in interpreting the results (see Supplementary Text).

Heritability and correlations among brain disorders

We observed a similar range of heritability estimates among the disorders and the behavioral-cognitive phenotypes (Fig. S1A-B and Table S1, S2), roughly in line with previously reported estimates in smaller datasets (see Table S3 and Supplementary Text). Three ischemic stroke subtypes (cardioembolic, large-vessel disease and small-vessel disease) as well as the “agreeableness” personality measure from NEO Five-Factor Inventory (26) had insufficient evidence for additive heritability for robust analysis and were excluded from further analysis (24). We did not observe a correlation between heritability estimates and factors relating to study makeup (Table S4; Fig S1C-F).

In expanding on the number of pairwise comparisons in brain disorders, we observed widespread sharing across psychiatric disorders (Fig. 1 and S3) beyond those previously reported (16), but not among neurological disorders. Among the psychiatric disorders, schizophrenia in particular showed significant genetic correlation with most of the psychiatric disorders. Further, schizophrenia, bipolar disorder, anxiety disorders, MDD and ADHD each showed a high degree of correlation to the four others (average $r_g=0.40$; Table S5). Anorexia nervosa, OCD and schizophrenia also demonstrated significant sharing amongst themselves. Tourette’s syndrome (TS) was only significantly correlated with OCD and migraine and not with any of the other disorders. From this analysis, the common variant risk of both ASD and TS appear to be somewhat distinct from other psychiatric disorders, although with some evidence of correlation between TS and OCD and between ASD and schizophrenia. The modest power of the ASD and TS meta-analyses, however, limits the strength of this conclusion.

Neurological disorders revealed greater specificity, and a more limited extent of genetic correlation than the psychiatric disorders (Fig. 2 and S4, Table S5). Parkinson’s disease, Alzheimer’s disease, generalized epilepsy and multiple sclerosis showed little to no correlation with any other brain disorders. Focal epilepsy showed the highest degree of genetic correlation among the neurological disorders (average $r_g=0.46$, excluding other epilepsy datasets), though none were significant, reflecting the relatively modest power of the current focal epilepsy meta-analysis. However, the modest heritability and the broad pattern of sharing observed for focal epilepsy may be consistent with considerable heterogeneity and potentially even diagnostic misclassification across a range of neurological conditions.

In the cross-category correlation analysis, the overall pattern is consistent with limited sharing across the included neurological and psychiatric disorders (Fig. 3; average $r_g=0.03$). The only significant cross-category correlations were with migraine, suggesting it may share some of
its genetic architecture with psychiatric disorders; migraine-ADHD ($r_g=0.26$, $p=8.81 \times 10^{-8}$), migraine-TS ($r_g=0.19$, $p=1.80 \times 10^{-5}$), and migraine-MDD ($r_g=0.32$, $p=1.42 \times 10^{-22}$ for all migraine), $r_g=0.23$, $p=5.23 \times 10^{-5}$ for migraine without aura, $r_g=0.28$, $p=1.00 \times 10^{-4}$ for migraine with aura).

We observed several significant genetic correlations between the behavioral-cognitive or additional phenotypes and brain disorders (Fig. 4 and S5, Table S6). Years of education showed a significant correlation to each psychiatric disorder (negative for ADHD, anxiety disorders, MDD and TS; positive for the others), while five neurological phenotypes (Alzheimer’s disease and the four “vascular” phenotypes, intracerebral hemorrhage, all and early-onset stroke and migraine) showed a significant negative correlation (Fig. S6; correlations with bipolar disorder(23), Alzheimer’s disease and schizophrenia have been previously reported(27)). Among the personality measures, significant positive correlations were observed for neuroticism (anxiety disorders, migraine, migraine without aura and OCD; Fig. S7), depressive symptoms (ADHD, anxiety disorder, bipolar disorder, MDD, and schizophrenia) and subjective well-being (anxiety disorder, bipolar disorder, MDD, as well as replicating the previously reported correlation between neuroticism with both MDD and schizophrenia(28)). For smoking-related measures, the only significant genetic correlations were to never/ever smoked (MDD: $r_g=0.33$, $p=3.10 \times 10^{-11}$ and ADHD: $r_g=0.37$, $p=3.15 \times 10^{-6}$).

Among the additional phenotypes, the two diseases chosen as examples of disorders with well-defined etiologies had different results: Crohn’s disease, representing immunological pathophysiology, showed no correlation with any of the study phenotypes, while the phenotype representing vascular pathophysiology (coronary artery disease) showed significant correlation to MDD ($r_g=0.19$, $p=8.71 \times 10^{-5}$) as well as the two stroke-related phenotypes ($r_g=0.69$, $p=2.47 \times 10^{-6}$ to ischemic stroke and $r_g=0.86$, $p=2.26 \times 10^{-5}$ for early-onset stroke), suggesting shared genetic effects across these phenotype. Significant correlations were also observed for BMI, which was positively correlated with ADHD and MDD, and negative correlated with anorexia nervosa (as previously reported with a different dataset(23)) and schizophrenia.

Functional enrichment analysis (Table S7 and S8) replicated the previously reported central nervous system (CNS) enrichment for schizophrenia and bipolar disorder (here in a larger dataset compared to the original report (25), but with considerable sample overlap). We also demonstrated novel significant heritability enrichments to CNS (for generalized epilepsy, MDD, TS, college attainment, neuroticism, never/ever smoked) and to immune system cells and tissues (multiple sclerosis).

Discussion

By integrating and analyzing the current genome-wide association summary statistic data from consortia of 24 brain disorders, we find that psychiatric disorders broadly share a
considerable portion of their common variant genetic risk, especially across schizophrenia, MDD, bipolar disorder, anxiety disorder and ADHD, while neurological disorders are more genetically distinct. Across categories, psychiatric and neurologic disorders share relatively little of their common genetic risk, suggesting that multiple different and largely independently regulated etiological pathways may give rise to similar clinical manifestations (e.g., psychosis, which manifests in both schizophrenia(29) and Alzheimer’s disease(30)). Except for migraine, which appears to share genetic architecture with psychiatric disorders, the existing clinical delineation between neurology and psychiatry is recapitulated at the level of common variant risk for the studied disorders.

Given that the broad and continuous nature of psychiatric disorder spectra in particular has been clinically recognized for a long time(31-33), we evaluated whether diagnostic misclassification (given that genetic correlations are always a function of the underlying ascertained phenotypes) could cause the observed correlations. For example, substantial numbers of cases progressing through multiple diagnoses over lifetime or some the diagnostic boundaries between some phenotype pairs being particularly porous to misclassification, congruent with the clinical controversies in classification, could give rise to the appearance of correlation. Previous work(34) suggests that even substantial misclassification is likely insufficient to introduce high levels of genetic correlation. We expanded on this work by performing a series of simulations across a variety of scenarios. These simulations demonstrate that reasonable levels of misclassification or changes in the exact level of heritability appear unlikely to induce substantial changes in the estimated genetic correlation (Fig. S8, S9, Table S9 and Supplementary Text), though the lower resulting heritability (Fig. S8A) would decrease the power to estimate the genetic overlap (Fig S10). Further, such evidence of genetic overlap is unlikely to appear in the absence of underlying genetic correlation (Table S10), as it is apparent that very high (up to 79%) degree of misclassification would be required to produce the observed correlations in the absence of any true genetic correlation. Therefore, the observed correlations suggest true sharing of a substantial fraction of the common variant genetic architecture among psychiatric disorders as well as between behavioral-cognitive measures and brain disorders.

The high degree of genetic correlation among the psychiatric disorders adds further evidence that current clinical diagnostics do not reflect the underlying genetic etiology of these disorders, and that genetic risk factors for psychiatric disorders do not respect clinical diagnostic boundaries. This suggests an interconnected nature for their genetic etiology, in contrast to neurological disorders, and underscores the need to refine psychiatric diagnostics. This study may provide important ‘scaffolding’ to support a new research framework for investigating mental disorders, incorporating many levels of information to understand basic dimensions of brain function, such as through the National Institute of Mental Health’s RDoC initiative.

The observed positive genetic correlations are consistent with a few different scenarios. For example, $r_g$ may reflect the existence of some portion of common genetic risk factors conferring equal risks to multiple disorders where other distinct additional factors contribute to the
eventual clinical presentation. The presence of significant genetic correlation may also reflect the phenotypic overlap between any two disorders; for example, the sharing between schizophrenia and ADHD might reflect underlying difficulties in executive functioning, which are well-established in both disorders(35). Similarly, the sharing between anorexia nervosa, OCD and schizophrenia may reflect a shared mechanism underlying cognitive biases that extend from overvalued ideas to delusions. Another scenario is that a heritable intermediate trait confers risk to multiple outcomes, thereby giving rise to the genetic correlation, as the genetic influences on this trait will be shared for both outcomes (e.g., obesity as a risk factor for both type 2 diabetes and coronary artery disease), or even that the majority common genetic effects are shared between a pair of traits, but each individual effect may confer different degrees of risk and lead to different aggregate genetic risk profiles. While a combination of these is likely, it will become increasingly feasible to evaluate these overlaps at the locus level in the future as more genome-wide significant loci are identified.

The low correlations observed across neurological disorders suggest that the current classification reflects relatively specific genetic etiologies, although the limited sample size for some of these disorders and lack of inclusion of disorders conceived as “circuit-based” in the literature, such as restless legs syndrome, sleep disorders and possibly essential tremor, constrains the generalizability of this conclusion. Generally, this analysis recapitulates the current understanding of the relatively distinct primary etiology underlying these disorders; degenerative disorders (such as Alzheimer’s and Parkinson’s diseases) would not be expected a priori to share their polygenic risk profiles with a neuro-immunological disorder (like multiple sclerosis) or neurovascular disorder (like ischemic stroke). Similarly, we see limited evidence for the reported co-morbidity between migraine with aura and ischemic stroke(36) ($r_g=0.29$, $p=0.099$); however, the standard errors of this comparison are too high to draw strong conclusions. At the disorder subtype level, migraine with and without aura ($r_g=0.48$, $p=1.79 \times 10^{-5}$) show substantial genetic correlation, while focal and generalized epilepsy ($r_g=0.16$, $p=0.388$) show much less.

The few significant correlations across neurology and psychiatry, namely between migraine and ADHD, MDD and TS, suggest modest shared etiological overlap across the neurology/psychiatry distinction. The co-morbidity of migraine with MDD and ADHD has been previously reported in epidemiological studies (37-39), while in contrast, the previously reported co-morbidity between migraine and bipolar disorder seen in epidemiological studies (40) was not reflected in our estimate of genetic correlation ($r_g=-0.03$, $p=0.406$).

Several phenotypes show only very low-level correlations with any of the other studied disorders and phenotypes, despite large sample size and robust evidence for heritability, suggesting their common variant genetic risk may largely be unique. Alzheimer’s disease, Parkinson’s disease, and multiple sclerosis show extremely limited sharing with the other phenotypes and with each other. Neuroinflammation has been implicated in the pathophysiology of each of these conditions(41-43), as it has for migraine(44) and schizophrenia(45), but no considerable shared heritability was observed with either of those conditions nor with Crohn’s disease. While this
observation does not preclude shared neuroinflammatory mechanisms in these disorders, it does
suggest that on a large scale, common variant genetic influences on these inflammatory
mechanisms are not shared between the disorders. Further, we only observed significant
enrichment of heritability to immunological cells and tissues in multiple sclerosis, showing that
inflammation-specific regulatory marks in the genome do not show overall enrichment for
common variant risk for either Alzheimer’s or Parkinson’s diseases (though this does not preclude
the effects of specific, non-polygenic neuroinflammatory mechanisms(46)). Among psychiatric
disorders, ASD and TS showed a similar absence of correlation with other disorders, although this
could reflect small sample sizes.

Analysis of the Big Five personality measures suggests that the current sample sizes for
personality data are now starting to be sufficiently large for correlation testing; neuroticism, which
has by far the largest sample size, shows a number of significant correlations. Most significant of
these was to MDD ($r_g=0.737$, $p=5.04 \times 10^{-96}$), providing further evidence for the link between
these phenotypes, reported previously with polygenic risk scores(47) and twin studies(48, 49);
others included schizophrenia, anxiety disorders, migraine, migraine without aura, and OCD
(Table S6). Further, the observation of strong correlation between MDD and anxiety disorders
together with their remarkably strong and similar patterns of correlation between each of these
disorders and the dimensional measures of depressive symptoms, subjective well-being, and
neuroticism suggests that they all tag a fundamentally similar underlying etiology. The novel
significant correlation between coronary artery disease and MDD supports the long-standing
epidemiological observation of a link between MDD and CAD(50), while the observed correlation
between ADHD and smoking initiation ($r_g=0.374$, $p=3.15 \times 10^{-6}$) is consistent with the
epidemiological evidence of overlap(51) and findings from twin studies(52), supporting the
existing hypothesis that impulsivity inherent in ADHD may drive smoking initiation and
potentially dependence (though other explanations, such as reward system dysfunction would fit
as well).

For the neurological disorders, five (Alzheimer’s disease, intracerebral hemorrhage,
ischemic and early-onset stroke, and migraine) showed significant negative genetic correlation to
the cognitive measures, while a further two (epilepsy and focal epilepsy) showed moderate
negative genetic correlation (Fig. S6). For Alzheimer’s disease, poor cognitive performance in
early life has been linked with increased risk for developing the disorder in later life(53), but to
our knowledge no such connection has been reported for the other phenotypes. ADHD, anxiety
disorders and MDD show a significant negative correlation to cognitive and education attainment
measures, while the remaining five of the eight psychiatric disorders (anorexia nervosa, ASD,
bipolar disorder, OCD, and schizophrenia) showed significant positive genetic correlation with
one or more cognitive measures. These results strongly suggest the existence of a link between
cognitive performance already in early life and the genetic risk for both psychiatric and
neurological brain disorders. The basis of the genetic correlations between education, cognition
and brain disorders may have a variety of root causes including, indexing performance differences
based on behavioral dysregulation (e.g., ADHD relating to attentional problems during cognitive
tests) or may reflect ascertainment biases in certain disorders conditional on impaired cognition
(e.g., individuals with lower cognitive reserve being more rapidly identified for Alzheimer’s
disease).

BMI shows significant positive genetic correlation to ADHD, consistent with a meta-
analysis linking ADHD to obesity(54), and negative genetic correlation with anorexia nervosa,
OCD and schizophrenia. These results connect well with the evidence for enrichment of BMI
heritability in CNS tissues(25) and that many reported signals suggest neuronal involvement(55);
this may also provide a partial genetic explanation for lower BMI in anorexia nervosa patients
even after recovery(56). Given that no strong correlations were observed between BMI and any of
the neurological phenotypes, it is possible to hypothesize that BMI’s brain-specific genetic
architecture is more closely related to behavioral phenotypes. Ischemic stroke and BMI show
surprisingly little genetic correlation in this analysis ($r_g=0.07$, $p=0.26$), suggesting that although
BMI is a strong risk factor for stroke(57), there is little evidence for shared common genetic effects.
These analyses also suggest that the reported reduced rates of cardiovascular disease in individuals
with histories of anorexia nervosa (58, 59) are due to BMI-related effects; with the limited evidence
of genetic correlation of anorexia nervosa with intracerebral hemorrhage, ischemic stroke, early-
onset stroke and coronary artery disease, these results suggest that any lower cardiovascular
mortality is more likely due to direct BMI-related effects rather than shared common genetic risk
variants.

Broadly it is apparent that the current clinical boundaries for the studied psychiatric
phenotypes do not appear to reflect distinct underlying pathogenic processes based on the genetic
evidence, while in contrast, the studied neurological disorders show much greater genetic
specificity. Although it is important to emphasize that while some disorders are under-represented
here (e.g. personality disorders in psychiatry and circuit-based disorders such as restless leg
syndrome in neurology), these results clearly demonstrate the limited evidence for widespread
common genetic risk sharing between psychiatric and neurological disorders, while on the other
hand providing strong evidence for links between them and behavioral-cognitive measures. We
highlight the need for some degree of restructuring of psychiatric nosology and that genetically
informed analyses may provide a good basis for such activities, consistent with the historical
knowledge from twin- and family-based results. Further elucidation of individual disorders and
their genetic overlap, especially as distinct loci map onto a subset of disorders and etiological
processes, may form the basis for either defining new clinical phenotypes or support a move to a
more continuous view of psychiatric phenotypes. Ultimately, such developments give hope for
reducing diagnostic heterogeneity and eventually improving the diagnosis and treatment of
psychiatric disorders.
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Materials and methods are available as supplementary materials on Science Online.


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Data sources

Disorder or phenotype – Consortium or dataset identifier – web address:

Psychiatric disorders

OCD – IOCDFGC - https://iocdf.org/

Neurological disorders

Epilepsy and subtypes, focal and generalized(60) – ILAE – http://www.epigad.org/page/show/gwas_index
Intracerebral hemorrhage(61) – ISGC - http://www.strokegenetics.com/
Migraine and subtypes, migraine with and without aura – IHGC – www.headachegenetics.org
Multiple sclerosis(63) – IMSGC - http://eaglep.case.edu/imsgc_web
Parkinson's disease(20) – IPDGC – www.pdgene.org

Behavioral-cognitive phenotypes

College attainment, years of education(64) – SSGAC – http://www.thessgac.org/data
Childhood cognitive performance(65) – SSGAC – http://www.thessgac.org/data
Extraversion, agreeableness, conscientiousness and openness (26) – GPC – http://www.tweelingenregister.org/GPC/
Never/ever smoked, cigarettes per day(67) - TAG - http://www.med.unc.edu/pgc/results-and-downloads

Additional phenotypes

BMI(55) – GIANT – https://www.broadinstitute.org/collaboration/giant
Height(68) – GIANT – https://www.broadinstitute.org/collaboration/giant
Coronary artery disease(70) – Cardiogram – http://www.cardiogramplusc4d.org/downloads/

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Figure 1. Genetic correlation matrix across psychiatric phenotypes.

Color of each box indicates the magnitude of the correlation, while size of the boxes indicates its significance, with significant correlations filling each box completely. Asterisks indicate genetic correlations which are significant after Bonferroni correction. ADHD – attention deficit hyperactivity disorder; ASD – autism spectrum disorder; MDD – major depressive disorder; OCD – obsessive-compulsive disorder.
Figure 2. Genetic correlation matrix across neurological phenotypes.

Color of each box indicates the magnitude of the correlation, while size of the boxes indicates its significance, with significant correlations filling each box completely. Asterisks indicate genetic correlations which are significant after Bonferroni correction. Some phenotypes have substantial overlaps (see Table 1), e.g. all cases of generalized epilepsy are also cases of epilepsy. Asterisks indicate significant genetic correlation after multiple testing correction. ICH – intracerebral hemorrhage.
Figure 3. Genetic correlation matrix across neurological and psychiatric phenotypes.

Color of each box indicates the magnitude of the correlation, while size of the boxes indicates its significance, with significant correlations filling each box completely. Asterisks indicate genetic correlations which are significant after Bonferroni correction. ADHD – attention deficit hyperactivity disorder; ASD – autism spectrum disorder; ICH – intracerebral hemorrhage; MDD – major depressive disorder; OCD – obsessive-compulsive disorder.
**Figure 4.** Genetic correlation matrix across brain disorders and behavioral-cognitive phenotypes.

Color of each box indicates the magnitude of the correlation, while size of the boxes indicates its significance, with significant correlations filling each box completely. Asterisks indicate genetic correlations which are significant after Bonferroni correction. ADHD – attention deficit hyperactivity disorder; ASD – autism spectrum disorder; ICH – intracerebral hemorrhage; MDD – major depressive disorder; OCD – obsessive-compulsive disorder; BMI – body-mass index.
Table 1. Brain disorder phenotypes used in the Brainstorm project. Indented phenotypes are part of a larger whole, e.g. the epilepsy study consists of the joint analysis of focal epilepsy and generalized epilepsy. Numbers in gray denote a sample set which is non-unique, e.g. cardioembolic stroke samples are a subset of ischemic stroke samples. ADHD – attention deficit hyperactivity disorder; OCD – obsessive-compulsive disorder. ‘Anxiety disorders’ refers to a meta-analysis of five subtypes (generalized anxiety disorder, panic disorder, social phobia, agoraphobia and specific phobias). Source details are listed under Data Sources and the references in Table S1.

<table>
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<th>Psychiatric disorders</th>
<th>Neurological disorders</th>
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<td>Autism spectrum disorder</td>
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<td>Major depressive disorder</td>
<td>PGC-MDD2</td>
</tr>
<tr>
<td>OCD</td>
<td>PGC-OCDTS</td>
</tr>
<tr>
<td>Schizophrenia</td>
<td>PGC-SCZ2</td>
</tr>
<tr>
<td>Tourette Syndrome</td>
<td>PGC-OCDTS</td>
</tr>
<tr>
<td>Migraine</td>
<td></td>
</tr>
<tr>
<td>Migraine with aura</td>
<td></td>
</tr>
<tr>
<td>Migraine without aura</td>
<td></td>
</tr>
<tr>
<td>Multiple sclerosis</td>
<td></td>
</tr>
<tr>
<td>Parkinson's disease</td>
<td></td>
</tr>
</tbody>
</table>

Total psychiatric                          | 105,177         | 229,329      | Total neurologic                           | 107,190         | 418,650      |
Table 2. Behavioral-cognitive and additional phenotypes used in the study. Numbers in gray denote overlapping study sets, e.g. samples in the college attainment analysis are a subset of those in the analysis for years of education. (d) – dichotomous phenotype, (q) – quantitative phenotype. BMI – body-mass index. Source details are listed under Data Sources, while references are listed in Table S2.

<table>
<thead>
<tr>
<th>Phenotype</th>
<th>Source</th>
<th>Samples</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Cognitive</strong></td>
<td></td>
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</tr>
<tr>
<td>Years of education (q)</td>
<td>SSGAC</td>
<td>293,723</td>
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<tr>
<td>College attainment (d)</td>
<td>&quot;</td>
<td>120,917</td>
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<tr>
<td>Cognitive performance (q)</td>
<td>&quot;</td>
<td>17,989</td>
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<tr>
<td><strong>Personality measures</strong></td>
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<td></td>
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<tr>
<td>Subjective well-being</td>
<td>SSGAC</td>
<td>298,420</td>
</tr>
<tr>
<td>Depressive symptoms</td>
<td>&quot;</td>
<td>161,460</td>
</tr>
<tr>
<td>Neuroticism (q)</td>
<td>&quot;</td>
<td>170,911</td>
</tr>
<tr>
<td>Extraversion (q)</td>
<td>GPC</td>
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<td>Agreeableness (q)</td>
<td>&quot;</td>
<td>17,375</td>
</tr>
<tr>
<td>Conscientiousness (q)</td>
<td>&quot;</td>
<td>17,375</td>
</tr>
<tr>
<td>Openness (q)</td>
<td>&quot;</td>
<td>17,375</td>
</tr>
<tr>
<td><strong>Smoking-related</strong></td>
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<td></td>
</tr>
<tr>
<td>Never/ever smoked (d)</td>
<td>TAG</td>
<td>74,035</td>
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<tr>
<td>Cigarettes per day (q)</td>
<td>TAG</td>
<td>38,617</td>
</tr>
<tr>
<td><strong>Additional phenotypes</strong></td>
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<td></td>
</tr>
<tr>
<td>BMI (q)</td>
<td>GIANT</td>
<td>339,224</td>
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<tr>
<td>Height (q)</td>
<td>&quot;</td>
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<td>Coronary artery disease (d)</td>
<td>Cardiogram</td>
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<tr>
<td>Crohn’s disease (d)</td>
<td>IIBDGC</td>
<td>20,883</td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td></td>
<td>1,124,048</td>
</tr>
</tbody>
</table>

595  Table 2. Behavioral-cognitive and additional phenotypes used in the study. Numbers in gray denote overlapping study sets, e.g. samples in the college attainment analysis are a subset of those in the analysis for years of education. (d) – dichotomous phenotype, (q) – quantitative phenotype. BMI – body-mass index. Source details are listed under Data Sources, while references are listed in Table S2.

599  Total 1,124,048
Supplementary Materials

Materials and methods

Supplementary Text

Comparison with previous heritability estimates

Effect of misclassification

Correlation by misclassification alone

Study-specific acknowledgements

Consortium memberships

Figures S1-10

Tables S1-9