Supplementary Information for:

The shared genetic risk architecture of neurological and psychiatric disorders: a genomewide analysis

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1 Supplementary Methods

1.1 Descriptions of GWAS summary data

GWAS summary statistics for 28 phenotypes were included in this cross-trait analysis. A summary of the samples contributing to each GWAS, including the phenotype definition, is provided below. Detailed information about cohort design, genotyping, quality control and imputation can be found in the original publications (Table 1).

1.1.1 Neurological diseases

Alzheimer's disease (ALZ). We retrieved GWAS summary statistics based on 12 cohorts contributing to the Wightman *et al.* (2021) GWAS on ALZ¹, yielding a total sample of 86,531 cases and 676,386 controls, all of European ancestry. Depending on the cohort, cases had been clinically diagnosed with ALZ according to ICD-9 or ICD-10 criteria, were found to have definite, probable or possible ALZ at a memory clinic, or were identified in medical registries as having been prescribed Donepezil (Aricept). Additionally, a proxy phenotype for ALZ casecontrol status was generated from a self-report questionnaire from the UK Biobank², in which participants reported whether their biological mother or father ever suffered from ALZ/dementia, and they reported each parent's current age (or age at death, if applicable).

Amyotrophic lateral sclerosis (ALS). 117 cohorts were meta-analyzed, yielding a total sample of 27,205 cases and 110,881 controls. All cases were diagnosed and ascertained through specialized motor neuron disease clinics where they were diagnosed with ALS according to the (revised) El Escorial Criteria3 by neurologists specialized in motor neuron diseases. All individuals were of European ancestry.

Essential tremor (ET). Two clinical cohorts, comprising of samples from multiple centers, were meta-analyzed together with data from two population-based cohorts from the UK Biobank and 23andMe, yielding a total sample of 7,177 cases and 475,877 controls⁴. In the clinical cohorts, cases were diagnosed according to consensus criteria from the Movement Disorder Society (MDS), the Tremor Research Investigational Group (TRIG), other clinical or research criteria, or review of longitudinal movement disorder clinical notes. In the populationbased cohorts, cases self-reported as having ET. All individuals were of European ancestry.

Focal epilepsy (FE) and genetic generalized epilepsy (GGE). The multi-ancestry GWAS⁵ on the broad epilepsy subtypes FE and GGE by the International league against epilepsy (ILAE) included 14,939 cases with FE, 6,952 cases with GGE and 42,436 controls of European ancestry that were included in the present analysis. All cases were classified by neurologists.

Lewy body dementia (LBD). The GWAS on LBD⁶ comprised of multiple sites/consortia from Europe and North America, of which GWAS summary data on 2,591 cases and 4,027 controls of European ancestry were included in the present analysis. Cases were diagnosed with pathologically definite or clinically probable disease according to consensus criteria.

Migraine (MIG). We retrieved GWAS summary data based on four study collections contributing to the GWAS on MIG by the International Headache Genetics Consortium (IHGC) 7 , totaling 48,975 cases and 540,381 controls. Most migraine diagnoses were self-reported, while a subset of cases was phenotyped in headache specialist centers. All individuals were of European ancestry.

Multiple sclerosis (MS). We retrieved GWAS summary data on 14,802 cases and 26,703 controls contributing to the GWAS on MS by the International Multiple Sclerosis Genetics Consortium (IMSGC)⁸. Cases were determined according to standard diagnostic criteria. All individuals were of European ancestry.

Parkinson's disease (PD). The GWAS summary data was based on a meta-analysis from three different GWAS⁹⁻¹¹, yielding a total sample of 53,858 cases and 846,380 controls. A subset of cases was ascertained by clinicians¹⁰, a subset in the 23andMe cohort self-reported as having PD in a web-based survey¹¹, while another subset was determined as proxy-cases using UK Biobank data (individuals who do not have Parkinson's disease but have a first degree relative that does)⁹. All individuals were of European ancestry.

Stroke. We included GWAS summary statistics on any stroke (comprising ischemic stroke, intracerebral hemorrhage, and stroke of unknown or undetermined type) contributing to a recent GWAS on stroke¹². Cases were determined according to various clinical criteria depending on the cohort. All individuals were of of European ancestry.

1.1.2 Psychiatric disorders

Anorexia nervosa (AN). Thirty-three datasets were meta-analyzed, yielding 16,992 cases and $55,525$ controls¹³. Cases met criteria for lifetime diagnosis of AN via hospital or register records, structured clinical interviews, or online questionnaires based on standardized criteria (DSM-III-R, DSM-IV, ICD-8, ICD-9 or ICD-10), while in one contributing cohort, the UK Biobank, cases self-reported diagnosis of AN. All individuals were of European ancestry.

Anxiety disorders (ANX). Three cohorts were meta-analyzed¹⁴, yielding a total sample of 31,977 cases and 82,114 controls. In the first cohort from the UK Biobank, cases self-reported a lifetime professional diagnosis for at least one of the core five ANX (generalized anxiety disorder, social phobia, panic disorder, agoraphobia or specific phobia). Additionally, some cases met criteria for a likely lifetime diagnosis of DSM-IV generalized anxiety disorder based on anxiety questions from the Composite International Diagnostic Interview Short-form questionnaire. In the second cohort from the ANGST study¹⁵, cases met DSM criteria for at least one of the five core ANX. In the third cohort from iPSYCH¹⁶, diagnosis of one of the five ANX was made by a psychiatrist. All individuals were of European ancestry.

Attention deficit hyperactive disorder (ADHD). Twelve cohorts, including data from iPSYCH, deCODE genetics and ten cohorts collected by the PGC, were meta-analyzed, yielding a sample of $38,691$ cases and $186,843$ controls¹⁷. In the iPSYCH cohort, cases were diagnosed by psychiatrists according to ICD-10 criteria identified using the Danish Psychiatric Central Research Register and the Danish National Patient register. In the deCODE cohort, cases had been diagnosed according to ICD-10 criteria, or had been prescribed medication specific for ADHD symptoms. Depending on the PGC cohorts, cases were recruited from clinics, hospitals or through medical registries and diagnosed using research-based assessments administered by clinicians or trained staff. All individuals were of European ancestry.

Autism spectrum disorder (ASD). Five family-based cohorts and a population-based casecontrol sample from iPSYCH were meta-analyzed¹⁸, yielding a sample of 18,381 cases and 27,969 controls. In the family-based cohorts, diagnosis was confirmed for all affected individuals using standard research tools and expert clinical consensus diagnosis. In the

population-based sample, cases were identified using the Danish Psychiatric Central Research Register and were diagnosed with ASD before 2014 by a psychiatrist according to ICD-10. All individuals were of European ancestry.

Bipolar disorder (BD). Fifty-seven cohorts collected in Europe, North America and Australia were meta-analyzed, yielding a total sample of $41,917$ cases and $371,549$ controls¹⁹. For 52 cohorts, cases met international consensus criteria (DSM-IV, ICD-9 or ICD-10) for lifetime diagnosis of BD, established using structured diagnostic interviews, clinician-administered checklists or medical record review. For five external cohorts (iPSYCH, deCODE genetics, Estonian Biobank, Trøndelag Health Study (HUNT) and UK Biobank), cases were ascertained using ICD codes or self-report during a nurse interview. All individuals were of European ancestry.

Major depressive disorder (MDD). We meta-analyzed separate GWAS data on MDD, from the PGC²⁰, 23andMe²¹, UK Biobank²², Million Veteran Program²³, and FinnGen²⁴, yielding a total sample of $412,305$ cases and $1,588,397$ controls. In the PGC cohort²⁰, lifetime diagnosis of MDD was ascertained using structured clinical interviews (DSM-V, ICD-9, ICD-10), clinician-administered checklists, or review of medical records, while some met broader criteria. In the 23andMe cohort²¹, cases were defined according to a self-reported clinical diagnosis of depression. In the UK Biobank cohort²², case and control status of broad depression was defined by the participants' response to the questions 'Have you ever seen a general practitioner for nerves, anxiety, tension or depression?' or 'Have you ever seen a psychiatrist for nerves, anxiety, tension or depression?', while participants who either identified with bipolar disorder, schizophrenia, or personality disorder, or were prescribed antipsychotic medications, were excluded. In both the Million Veteran Program and FinnGen cohorts, case

definitions were derived from electronic health record data based on ICD-9 and -ICD-10 codes for MDD. All individuals were of European ancestry.

Obsessive compulsive disorder (OCD). Two cohorts were meta-analyzed yielding 2,688 cases and 7,037 controls²⁵. All cases met DSM-IV criteria for OCD. All individuals were of European ancestry.

Post-traumatic stress disorder (PTSD). We included only individuals of European ancestry of an ancestrally diverse meta-analysis of 60 different PTSD-cohorts²⁶, yielding 20,329 cases and 124,440 controls. Cases were determined either by lifetime or current PTSD. Diagnosis was established using various instruments and different versions of the DSM (DSM-III-R, DSM-IV, DSM-5).

Schizophrenia (SCZ). We excluding all individuals of non-European ancestry from the recent GWAS on SCZ from the PGC²⁷, yielding a sample of 53,386 cases and 77,258 controls across 75 cohorts. Depending on the cohort, cases were defined as individuals diagnosed with schizophrenia, schizoaffective disorder, or schizophrenia spectrum disorders, determined by research diagnostic interview, expert clinical consensus diagnosis, or review of medical records or hospital registers.

Tourette syndrome (TS). The GWAS consisted of three case-control cohorts and one familybased cohort including 4,819 cases and 9,488 controls²⁸. All cases met DSM-IV-TR or DSM-5 criteria for TS, while 12 cases met DSM-5 criteria for chronic motor or vocal tic disorder. Cases were recruited through TS specialty clinics or were identified by email or online recruitment combined with validated, web-based phenotypic assessments. All individuals were of European ancestry.

1.1.3 Comparators

Average cortical thickness (CRT-TH) and total cortical surface area (CRT-SA). To generate the neuroimaging data, we used the Freesurfer $v5.3$ automatical cortical pipeline²⁹ to process T1-weighted MRI data from all white Europeans participating in the UK Biobank, under accession number 27412. The final neuroimaging sample included 32,877 individuals, after excluding individuals with poor scan quality, and all individuals with a neurological, psychiatric or brain-related disorder, specifically the ICD10 diagnoses 'Malignant neoplasms of eye, brain and other parts of central nervous system' (C69-C72), 'Mental, Behavioral and Neurodevelopmental disorders' (F00-F99), 'Diseases of the nervous system' (G00-G99), 'Cerebrovascular diseases' (I60-I69), and 'Congenital malformations of the nervous system' (Q00-07), to avoid bias when estimating genetic overlap with neurological and psychiatric disorders. Age, sex, scanner location, scan quality and the first twenty genetic principal components were included as covariates and a rank-based inverse normal transformation was applied to the phenotype when conducting association analysis. We further made use of the UK Biobank v3 imputed data. The GWAS on brain morphology measures were carried out using the standard additive model of linear association in regenie $v3.2.5.^{30}$

General cognitive ability (COG). 14 epidemiological cohorts contributed to the GWAS on COG31, including a total of 269,867 participants of European ancestry. Various neurocognitive tests, primarily assessing fluid domains of cognitive functioning, were conducted across the cohorts.

Chronic kidney disease (CKD). We retrieved GWAS summary data on 41,395 cases and 439,303 controls of European ancestry contributing to a GWAS on CKD³². CKD was defined as an eGFR below 60 ml min⁻¹ per 1.73 m^2 .

Coronary artery disease (CAD). We included GWAS summary statistics on 71,602 cases and 260,875 controls of European ancestry contributing to GWAS on CAD33-35. Cases were defined by an inclusive CAD phenotype, for example fatal or nonfatal myocardial infarction, acute coronary syndrome, percutaneous transluminal coronary angioplasty, coronary artery bypass grafting, chronic ischemic heart disease, angina and coronary stenosis of >50%.

Inflammatory bowel disease (IBD). A total of 25,042 cases and 34,915 controls contributed to the GWAS on IBD, all of European ancestry³⁶. Diagnosis of IBD was based on accepted radiological, endoscopic and histopathological criteria.

Type 2 Diabetes (T2D). The GWAS summary statistics on T2D was based on a meta-analysis of 32 GWAS, including 74,124 cases and 824,006 controls, all of European ancestry37. Cases were defined according to standard diagnostic criteria.

Height. We included GWAS summary statistics on 4,080,687 individuals of European ancestry contributing to the recent GWAS on height 38 .

2 Supplementary Results

2.1 Global genetic correlations

Using linkage disequilibrium score regression $(LDSC)^{39}$ (Fig. 3; Supplementary Table 4), we estimated significant genetic correlations between ET and five psychiatric disorders at FDR < 0.05. Specifically, ET was positively correlated with PTSD $(r_g=0.40, p=4.00 \times 10^{-4})$, ANX $(r_g=0.37, p=4.63x10^{-6})$, MDD $(r_g=0.22, p=6.08x10^{-7})$, AN $(r_g=0.17, p=2.80x10^{-3})$ and ADHD $(r_g=0.15, p=7.40x10^{-3})$, indicating shared genetic effects. MIG, which was positively correlated with ET $(r_g=0.17, p=3.90 \text{ x}10^{-3})$, showed a similar pattern of genetic correlations with psychiatric disorders as ET. Specifically, MIG was significantly correlated with ANX (r_g =0.35, p=8.59x10-16), MDD (*r*g=0.29, p=1.71x10-45), PTSD (*r*g=0.22, p=1.00x10-4), ADHD (*r*g=0.20, $p=6.83x10^{-10}$), ASD ($r_g=0.12$, $p=3.30x10^{-3}$) and TS ($r_g=0.12$, $p=1.39x10^{-2}$) at FDR < 0.05. Stroke was also significantly correlated with multiple psychiatric disorders at FDR < 0.05. Stroke was positively correlated with ADHD $(r_g=0.29, p=6.70 \times 10^{-13})$, PTSD $(r_g=0.28,$ $p=1.00x10^{-4}$), ANX ($r_g=0.20$, $p=5.09x10^{-5}$) MDD ($r_g=0.16$, $p=1.22x10^{-7}$) and BD ($r_g=0.09$, $p=9.30x10^{-3}$), while it was negatively correlated with OCD ($r_g=-0.19$, $p=5.90x10^{-3}$).

Except for negative correlations between CRT-TH and the psychiatric disorders ADHD $(r_g=-0.19, p=1.67x10^{-7})$, ANX $(r_g=-0.14, p=1.60x10^{-3})$ and MDD $(r_g=-0.08, p=1.40x10^{-3})$, neither of the two neuroimaging measures were significantly correlated with neurological or psychiatric disorders at FDR < 0.05. CRT-TH and CRT-SA were negatively correlated with each other (r_g =-0.36, p=3.43x10⁻¹⁷), while CRT-TH was positively correlated with COG (r_g =0.23, $p=4.75x10^{-17}$) and height ($r_g=0.20$, $p=6.86x10^{-16}$), and negatively correlated with CAD ($r_g=$ 0.12, $p=5.00x10^{-4}$). Cognitive impairment is a clinical hallmark shared between many psychiatric and neurological conditions⁴⁰⁻⁴³. As expected³¹, COG was negatively correlated with several neurological and psychiatric disorders, but positively correlated with ASD $(r_g=0.23, p=5.60x10^{-14})$ and AN $(r_g=0.09, p=3.60x10^{-3})$. Notably, COG was also significantly

correlated with somatic disorders CAD (r_g =-0.19, p=2.14x10⁻¹⁸), T2D (r_g =-0.05, p=1.22x10⁻²), IBD (r_g =0.10, p=2.00x10⁻⁴) as well as height (r_g =0.10, p=2.65x10⁻¹⁶), indicating that genetic variants linked to COG influence a wide spectrum of health-related traits. CAD, which was genetically correlated with stroke (r_g =0.52, p=1.65x10⁻³³), showed a similar pattern of genetic correlations as stroke. Specifically, CAD was significantly correlated with four psychiatric disorders (MDD $(r_g=0.22, p=4.15x10^{-24})$, ADHD $(r_g=0.31, p=5.06x10^{-26})$, ANX $(r_g=0.25,$ $p=1.85x10^{-10}$) and PTSD ($r_g=0.27$, $p=9.66x10^{-6}$) as well as FE ($r_g=0.20$, $p=5.60x10^{-3}$). T2D, a cardiovascular disease risk factor, was also significantly correlated with MDD (r_g =0.11, $p=2.86x10^{-10}$), ADHD ($r_g=0.18$, $p=4.82x10^{-11}$) and ANX ($r_g=0.11$, $p=2.20x10^{-3}$), but negatively correlated with AN $(r_g=0.10, p=1.40x10^{-3})$ and OCD $(r_g=-0.14, p=5.20x10^{-3})$. Apart from psychiatric disorders, T2D was significantly correlated with stroke $(r_g=0.27, p=2.92x10^{-13})$, MS $(r_g=0.08, p=1.26x10^{-2})$, and CAD $(r_g=0.33, p=4.18x10^{-27})$. IBD was significantly positively correlated with five psychiatric disorders SCZ ($r_g=0.14$, p=9.83x10⁻⁹), BD ($r_g=0.09$, p=8.00x10⁻ ⁴), MDD (r_g =0.11, p=6.11x10⁻⁷), ANX (r_g =0.11, p=3.50x10⁻³) and OCD (r_g =0.15, p=1.03x10⁻⁴) ²), but only with MS (r_g =0.25, p=4.45x10⁻⁹) among the neurological disorders. CKD was negatively correlated with four psychiatric disorders SCZ (r_g =-0.14, p=4.67x10⁻⁵), BD (r_g =-0.11, p=2.80x10⁻³), PTSD (r_g =-0.36, p=1.00x10⁻⁴), and AN (r_g =-0.17, p=3.60x10⁻³), but not with any neurological disorder. Height showed multiple significant genetic correlations at FDR $<$ 0.05. Specifically, height was negatively correlated with MDD (r_g =-0.04, p=2.00x10⁻⁴), ANX (*r*g=-0.07, p=9.00x10-4), PTSD (*r*g=-0.11, p=6.10x10-3), ALZ (*r*g=-0.09, p=1.80x10-2), and CAD $(r_g=-0.13, p=2.55x10^{-12})$, and positively correlated with CRT-TH ($r_g=0.20, p=6.86x10^{-16}$), COG $(r_g=-0.10, p=2.65x10^{-16})$ and CKD $(r_g=0.10, p=3.00x10^{-4})$, demonstrating multiple healthrelated correlations with this anthropometric measure. As observed for other measures of genetic overlap in this study, the presence of statistically significant genetic correlations appeared to be driven by statistical power. The GWAS with low power (e.g., TS and OCD)

showed fewer significant correlations than other psychiatric disorders, while this is likely to change as their respective GWAS increase in sample size.

2.2 Differential expression of MAGMA genes in specific tissues

Using MAGMA, we evaluated whether the implicated genes for each GWAS were differentially expressed in 54 specific human tissue types based on Genotype-Tissue Expression v.8 data⁴⁴, as implemented in FUMA⁴⁵ (Supplementary Fig. 3) Risk genes for ALZ were significantly upregulated in the spleen, the terminal ileum of the small intestine and in whole blood, all linked to the immune system. Risk genes for ADHD were significantly upregulated in the cerebral cortex only. Risk genes for AN were significantly downregulated in the putamen and the hippocampus, and differentially expressed in the putamen, the hippocampus and the amygdala (considering both sides). Genes implicated for CRT-TH were significantly upregulated in the cerebellar hemisphere only. Genes implicated for COG were significantly upregulated in several brain tissues, specifically the anterior cingulate cortex, frontal cortex, cerebral cortex, nucleus accumbens, the cerebellar hemisphere and the cerebellum, and downregulated in the renal cortex. Moreover, COG genes were differentially expressed in the following brain tissues, the putamen, anterior cingulate cortex, hypothalamus, amygdala, hippocampus, frontal cortex, cerebral cortex, nucleus accumbens and the substantia nigra. Height genes were significantly upregulated, downregulated or differentially expressed in 37 out of all 54 tissue types assessed, indicating body-wide expression of height genes. Risk genes for IBD were significantly upregulated in the spleen and whole blood, and downregulated in several brain tissues, specifically the putamen, amygdala, hippocampus, C-1 of the spinal cord, the substantia nigra, hypothalamus and frontal cortex. Risk genes for MDD were significantly upregulated in several brain tissues, specifically the hypothalamus, nucleus accumbens, putamen, hippocampus, the frontal cortex, nucleus caudate, substantia nigra, amygdala, anterior cingulate cortex, cerebral cortex, cerebellum, and the cerebellar hemisphere. Moreover, risk genes for MDD were differentially expressed in the hypothalamus, nucleus accumbens, putamen, hippocampus, and the frontal cortex. Risk genes for MS were significantly upregulated in EBV-transformed lymphocytes, the spleen, whole blood and in the terminal ileum of the small intestine. PD risk genes were significantly upregulated in the substantia nigra only. Risk genes for SCZ were significantly upregulated in several brain tissues, specifically the hippocampus, frontal cortex, cerebellum, the anterior cingulate cortex, the cerebral cortex, the cerebellar hemisphere, the amygdala, and the nucleus accumbens. Furthermore, SCZ risk genes were differentially expressed in the hippocampus, frontal cortex, cerebellum. and the anterior cingulate cortex. Finally, T2D risk genes were significantly downregulated in the substantia nigra, amygdala, hippocampus, C-1 of the spinal cord, putamen, caudate nucleus, nucleus accumbens and the hypothalamus.

3 Supplementary Figures

3.1 Supplementary Fig. 1 | GWAS power plots

Supplementary Fig. 1 | GWAS power plots. Power plots of current and future GWAS estimated using MiXeR⁴⁶, displaying the proportion of SNP- heritability explained by genomewide significant SNPs as a function of sample size, N. Asterisks indicate the current effective sample sizes of the GWAS, and circles indicate the estimated sample sizes needed to capture 90% of the genetic variance for that phenotype.

3.2 Supplementary Fig. 2 | Genetic correlations

Supplementary Fig. 2 | Genetic correlations. Global pairwise genetic correlations across all phenotypes estimated using linkage disequilibrium score regression³⁹. One asterisk denotes statistical significance at FDR < 0.05, two asterisks denote statistical significance after Bonferroni correction. The color denotes the magnitude and direction of correlation.

3.3 Supplementary Fig. 3 | Bivariate MiXeR estimates of genetic overlap

Supplementary Fig. 3 | Bivariate MiXeR estimates of genetic overlap. For each pair of phenotypes, the unique and overlapping genetic architectures estimated using MiXeR47, excluding phenotypes with poor model fit due to insufficient GWAS power (ANX, PTSD, TS, OCD, FE, ET and LBD). MiXeR-modeled Venn diagrams displaying the number (in thousands) of shared and phenotype-specific "causal" variants for each pair of phenotypes, conditional QQ plots showing real-life vs modelled cross trait enrichment and log-likelihood plots, plotting the adjusted negative log-likelihood function against the number of shared 'causal' variants, constrained to minimum and maximum possible overlap.

3.4 Supplementary Fig. 4 | Tissue specific expression of mapped genes

Attention-deficit/hyperactivity disorder

Anorexia nervosa

Cortical thickness

General cognitive ability

Height

Inflammatory bowel disease

Major depressive disorder

Multiple sclerosis

Parkinson's disease

Schizophrenia

Type 2 Diabetes

Supplementary Fig. 4 | Tissue specific enrichment. Results from MAGMA tissue specificity analysis across 54 different tissue types based on GTEx v.8 data⁴⁴, as implemented in FUMA⁴⁵. Red bars indicate significant results after Bonferroni correction..

3.5 Supplementary Fig. 5 | Cell type specific expression of mapped genes

Alzheimer's disease

Genetic generalized epilepsy

Multiple sclerosis

Attention-deficit/hyperactivity disorder

GSE168408_Human_Prefrontal_Cortex_level2_Neona Linnarsson_GSE76381_Human_Midbrain

Bipolar disorder

Allen Human LGN level2 GSE104276 Human Prefrontal cortex per ages GSE168408 Human Prefrontal Cortex level2 Adolescer GSE168408 Human Prefrontal Cortex level2 Adult GSE168408 Human Prefrontal Cortex level2 Fetal GSE168408 Human Prefrontal Cortex level2 Neonatal Linnarsson_GSE76381_Human_Midbrain

Schizophrenia

Coronary artery disease

GSE168408_Human_Prefrontal_Cortex_level1_Neonatal GSE168408_Human_Prefrontal_Cortex_level2_Adolescer GSE168408_Human_Prefrontal_Cortex_level2_Adult GSE168408_Human_Prefrontal_Cortex_level2_Childhood GSE168408_Human_Prefrontal_Cortex_level2_Fetal GSE168408_Human_Prefrontal_Cortex_level2_Infancy GSE168408_Human_Prefrontal_Cortex_level2_Neonatal

Cortical surface area

Allen_Human_MTG_level1 Allen Human MTG level2 GSE104276_Human_Prefrontal_cortex_per_ages GSE168408 Human Prefrontal Cortex level1 Adolescer GSE168408 Human Prefrontal Cortex level1 Adult GSE168408_Human_Prefrontal_Cortex_level1_Childhood GSE168408_Human_Prefrontal_Cortex_level1_Fetal
GSE168408_Human_Prefrontal_Cortex_level1_Fetal GSE168408 Human Prefrontal Cortex level2 Adolesce GSE168408 Human Prefrontal Cortex level2 Adult GSE168408 Human Prefrontal Cortex level2 Childhood GSE168408 Human Prefrontal Cortex level2 Fetal GSE168408_Human_Prefrontal_Cortex_level2_Infancy GSE168408_Human_Prefrontal_Cortex_level2_Neonatal PsychENCODE_Adult

General cognitive ability

Allen Human LGN level1 DroNc Human Hippocampus GSE104276 Human Prefrontal cortex all a GSE104276 Human Prefrontal cortex per GSE67835 Human Cortex Linnarsson GSE76381 Human Midbrain PsychENCODE Developmental

Inflammatory bowel disease

Type 2 diabetes

Supplementary Fig. 5 | Cell type specificity analysis. Results from MAGMA cell type specificity analysis using 24 single-cell RNA sequencing datasets from the developing and adult human brain, as implemented in FUMA⁴⁸. Significant cell type associations were determined after Bonferroni correction, followed by conditional analysis within each dataset to account for correlated cell type expression profiles.

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