GWAS of 19,629 individuals identifies novel genetic variants for regional brain volumes and refines their genetic co-architecture with cognitive and mental health traits

Running title: GWAS of 101 ROI volumes

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Abstract

Volumetric variations of human brain are heritable and are associated with many brain-related complex traits. Here we performed genome-wide association studies (GWAS) and post-GWAS analyses of 101 brain volumetric phenotypes using the UK Biobank (UKB) sample including 19,629 participants. GWAS identified 287 independent SNPs exceeding genome-wide significance threshold of $4.9*10^{-10}$, adjusted for testing multiple phenotypes. Gene-based association study found 142 associated genes (113 new) and functional gene mapping analysis linked 122 more genes. Many of the discovered genetic variants have previously been implicated with cognitive and mental health traits (such as cognitive performance, education, mental disease/disorders), and significant genetic correlations were detected for 29 pairs of traits. The significant SNPs discovered in the UKB sample were supported by a joint analysis with other four independent studies (total sample size 2,192), and we performed a meta-analysis of five samples to provide GWAS summary statistics with sample size larger than 20,000. Using genome-wide polygenic risk scores prediction, up to 4.36% of phenotypic variance (p-value=$2.97*10^{-22}$) in the four independent studies can be explained by the UKB GWAS results. In conclusion, our study identifies many new genetic variants at SNP, locus and gene levels and advances our understanding of the pleiotropy and genetic co-architecture between brain volumes and other traits.

Keywords: Genetic co-architecture; Genetic correlation; Pleiotropy; UK Biobank; Brain structure; Regional brain volumes.

Regional brain volumes are heritable measures of brain functional and structural changes. Volumetric variations of human brain are known to be phenotypically and genetically associated with heritable cognitive and mental health traits (1-5), and it is an active research area to understand the shared genetic influences in these traits (6). Individual variations of human brain volume are usually quantified by magnetic resonance imaging (MRI). In region of interest (ROI)-based analysis, whole brain MRIs are processed and annotated onto many per-defined ROIs, and then regional volumetric phenotypes are generated to measure the structure of brain ROIs. Family and population-based studies have both shown that these volumetric phenotypes are highly heritable (7-9), and common single-nucleotide polymorphism (SNP) markers collected across the genome can account for a large proportion of phenotypic variation (10). A highly polygenic or omnigenic (11, 12) genetic architecture has been observed, which indicates that a large number of genetic variants influence regional brain volumes and their genetic contributions are widespread across the whole genome.

Several genome-wide association studies (GWAS) (3, 7, 8, 13-18) have been conducted to identify genetic risk variants for brain volumetric phenotypes. However, except for the whole brain volume and volumes of few specific ROIs (e.g., hippocampus in subcortical area (3, 8, 19)), GWAS of most brain volumetric phenotypes were insufficiently powered, for which the largest sample size of discovery GWAS was less than 10,000 in (7). Such GWAS sample size is much smaller than those of recent GWAS of other heritable brain-related traits, such as cognitive function (20), neuroticism (21), and intelligence (22), where sample sizes ranged from 269,867 to 449,484. Given the polygenic nature of brain volumes, most of the genetic risk variants may remain undetected, and GWAS with larger sample size can uncover more associated variants and enrich the pleiotropy and genetic co-architecture with other traits. Recently, the UK Biobank (UKB, (23)) study team has collected and released MRI data for more than 20,000 participants. In addition, publicly available imaging genetic datasets also emerge from several other independent studies, including Philadelphia Neurodevelopmental Cohort (PNC, (24)), Alzheimer's Disease Neuroimaging Initiative (ADNI, (25)), Pediatric Imaging, Neurocognition, and Genetics (PING, (26)), and the Human Connectome

Project (HCP,(27)), among others. These datasets provide a new opportunity to perform better-powered GWAS of all ROI brain volumes.

Here we downloaded the raw MRI data from these data resources and processed the data using consistent standard procedures via advanced normalization tools (ANTs, (28)) to generate 101 regional (and total) brain volume phenotypes (referred as ROI volumes), including the total brain volume (TBV), gray matter (GM), white matter (WM), and cerebrospinal fluid (CSF). 19,629 UKB individuals of British ancestry were used in the main discovery GWAS. Other four datasets with relatively small sample sizes (total sample size 2,192 after quality controls) were used to validate the UKB findings and finally a meta-analysis was performed to combine all the data. We started our analysis of UKB data with estimating the SNP heritability, which is the proportion of phenotypic variation that can be explained by the additive effects of all common autosomal SNPs (29). Particularly, the UKB MRI data were released at different time points. We organized them in two parts: the first part was released in 2017 (referred as phase 1 in this paper, n=9,198), most of which has been analyzed in (7), and the second part was released in 2018 (referred as phase 2, n=10,431). To detect any potential heterogeneity of the two phases, we compared the SNP heritability estimated in phase 2 data to those in phase 1 data, which were reported in (10). We then carried out GWAS to identify the associated genetic variants for each ROI volume. We performed gene-based association analysis via MAGMA (30) to uncover gene-level associations, and performed post-GWAS functional mapping and annotation (FUMA, (31)) to explore the functional consequences of the significant SNPs. We calculated the pairwise genetic correlation between ROI volumes and 50 brain-related complex traits by the LD score regression (LDSC, (32)). To confirm the robustness of UKB GWAS findings, we jointly analyzed the UKB GWAS results with those from PNC, ADNI, PING and HCP. We developed genome-wide polygenic risk scores (PRS) to assess the predictive ability of the UKB GWAS results on the other four datasets. GWAS summary statistics of the UKB sample and meta-analysis for the five studies have been made available to public at https://med.sites.unc.edu/bigs2/data/gwas-summary-statistics/.

RESULTS

SNP heritability estimates of the two UKB phases data

Supplementary Fig. 1 compares the SNP heritability (h^2) estimated separately from UKB phase 1 and 2 data. The correlation of these estimates was 0.79, indicating moderate to high level of agreement in terms of the degree of genetic contributions to ROI between the two phases. Six ROIs had >0.6 h^2 estimates in both phases, including TBV, cerebellar vermal lobules VIII-X, cerebellar vermal lobules I-V, brain stem, and left/right cerebellum exterior. The h^2 estimates from the combined data were highly correlated with those from phase 1 (correlation=0.91) and phase 2 (correlation=0.92) **(Supplementary Figs. 2-3)**. The SNP heritability estimates, standard errors, raw and Bonferroni-corrected p-values from the one-sided likelihood ratio tests are provided in **Supplementary Table 1**. Significant genetic controls widely spread across most ROIs of the whole brain (mean h^2=0.35, h^2 range=[0.15,0.71], standard error=0.15). Heritability of left/right basal forebrain (h^2=0.09/0.11) and optic chiasm (h^2=0.01) were insignificant.

Significant GWAS associations of 101 ROI volumes

We carried out GWAS of the 101 ROI volumes with using 8,944,375 SNPs after genotyping quality controls. Manhattan and QQ plots of all the 101 phenotypes are displayed in **Supplementary Fig. 4**. There were 22,353 significant associations at the conventional $5*10^{-8}$ GWAS significance level and 12,060 significant ones at the $4.9*10^{-10}$ significance level (that is, $5*10^{-8}/101$, additionally adjusted for all 101 GWAS performed) **(Supplementary Fig. 5, Supplementary Table 2)**. TBV had the largest number of significant associations, which was 3,408 at $4.9*10^{-10}$ significance level. In addition to TBV, left/right hippocampus, left/right putamen, and cerebellar vermal lobules VIII-X had more than 500 significant associations. In the rest of this paper, we refer $4.9*10^{-10}$ as the significance threshold for SNP-level associations unless otherwise stated.

287 independent significant SNPs had 392 significant associations with 54 ROIs (**Supplementary Table 3**). Independent significant SNPs were defined as significant SNPs that were independent of other significant SNPs by FUMA (Online Methods, (31)). The number of associations for each ROI is displayed in **Figure 1 and Supplementary Table 4.** Left/right hippocampus, cerebellar vermal lobules VIII-X, left/right putamen, and cerebellar vermal lobules I-V had at least 19 independent significant SNPs. Other ROIs

that had at least 10 independent significant SNPs included left/right precentral, brain stem, X4th ventricle, left/right lateral ventricle, left/right cerebellum white matter, and TBV. The number of independent significant associations on each chromosome is shown in **Supplementary Table 5**, and clearly chromosome 12 had the largest number of SNP-level associations with ROI volumes **(Supplementary Fig. 6)**.

The 392 independent significant SNP-level associations can be further characterized (Online Methods) as 134 significant associations between genetic risk loci and ROI volumes **(Table 1, Supplementary Table 6)**. Brain stem, cerebellar vermal lobules VIII-X, left/right lateral ventricle, TBV and WM had at least five genetic risk loci **(Supplementary Table 7)**. Each chromosome had at least one genetic risk locus except for chromosomes 13 and 21 **(Supplementary Tables 8)**. Results at significance thresholds $5*10^{-8}$ and $5*10^{-9}$ are also provided in above tables and are summarized in **Supplementary Table 9**.

Concordance with previous GWAS results

We performed association lookups for the 287 independent significant SNPs and their correlated SNPs in genetic risk loci (Online Methods) on the NHGRI-EBI GWAS catalog (33). We found that 117 independent significant SNPs (associated with 36 ROI volumes) have previously reported GWAS associations with any traits **(Supplementary Table 10)**. Our results tagged many SNPs that were previously reported in GWAS of ROI volumes, including 14 SNPs in van der Meer, Rokicki (3) for hippocampal subfield volumes, 11 in Hibar, Stein (8) for subcortical brain region volumes, 5 in Chen, Wang (34) for putamen volume, 4 in Bis, DeCarli (18) for hippocampal volume, 2 in Hibar, Adams (14) for hippocampal volume, 2 in Stein, Medland (35) for brain structure, 2 in Ikram, Fornage (17) for intracranial volume, 1 in Furney, Simmons (36) for whole brain volume, and 1 in Baranzini, Wang (37) for normalized brain volume **(Supplementary Table 11)**. For the other traits, we highlighted previous associations of 29 different SNPs with mental health disease/disorders (such as schizophrenia, autism spectrum disorder [ASD], and depression), 78 with cognitive functions, 17 with educational attainment, 20 with neuroticism, 14 with Parkinson's disease, 3 with reaction time, and 1 with Alzheimer's disease. More previous GWAS results were found when the significance threshold was

relaxed to $5*10^{-8}$ **(Supplementary Table 12).** We also compared our results with those reported in (7). Elliott, Sharp (7) performed GWAS of 3,144 imaging phenotypes (including brain volume phenotypes processed by FreeSurfer (38)) using the UKB phase 1 data (n=8,428). When both being corrected for the number of performed GWAS, 26 of the 78 unique SNPs (covered 66 of the 368 significant associations) reported in (7) were within LD of our independent significant SNPs **(Supplementary Table 13)**. When both being relaxed to the $5*10^{-8}$ significance threshold, 119 of their 616 unique SNPs (covered 493 of the 1,262 significant associations) were within LD of our independent significant SNPs.

Gene-based association analysis and functional mapping

We performed gene-based association analysis with GWAS summary statistics for 18,796 candidate genes (Online Methods). We found 237 significant gene-level associations (p-value<$2*10^{-8}$, adjusted for multiple traits) between 142 genes and 47 ROIs **(Table 2, Supplementary Table 14).** Our results replicated 29 genes discovered in previous studies, including *FOXO3* in Baranzini, Wang (37), *HMGA2* and *HRK* in Stein, Medland (35), *KANSL1, MAPT, STH* and *CENPW* in Ikram, Fornage (17), *SLC44A5* in Furney, Simmons (36), *MSRB3, BCL2L1, DCC, CRHR1* in Hibar, Stein (8), *LEMD3, WIF1* and *ASTN2* in Bis, DeCarli (18), *FAM53B, METTL10* and *FAF1* in van der Meer, Rokicki (3), *DSCAML1* in Chen, Wang (34), *SLC39A1, GATAD2B, DENND4B* in Hibar, Stein (39), and *ZIC4, VCAN, PAPPA, DRAM1, GNPTAB, DAAM1*, and *ALDH1A2* in Elliott, Sharp (7). 13 genes were novel and were not linked to ROI volumes. 57 genes have previously been implicated with cognitive functions, intelligence, education, neuroticism, neuropsychiatric and neurodegenerative diseases/disorders, such as *IGF2BP1* (22, 40, 41), *WNT3* (20, 21, 42, 43), *PLEKHM1* (43-45), and *AGBL2* (21, 43, 46, 47). Particularly, 40 of the 57 pleiotropic genes were novel genes of ROI volumes, and thus these findings substantially uncovered the gene-level pleiotropy between ROI volumes and these traits **(Figure 2)**.

The independent significant SNPs were also annotated by functional consequences on gene functions **(Supplementary Table 15, Supplementary Fig. 7)**, and were subsequently mapped to genes according to physical position, expression quantitative

trait loci (eQTL) association (for brain tissues), and 3D chromatin (Hi-C) interaction (Online Methods). Functional gene mapping yielded 389 significant associations on 214 genes and 49 ROIs **(Supplementary Table 16)**. 122 genes were not discovered in the above gene-based association analysis, which replicated more previous findings on ROI volumes, such as *TBPL2* and *KTN1* in Chen, Wang (34), *FAT3* in Hibar, Stein (8), *SLC4A10, RNFT2, TESC, DMRTA2, CDKN2C* and *DPP4* in van der Meer, Rokicki (3), and *EPHA3, SLC39A8, BANK1, WNT16, CHPT1, ACADM, FAM3C, FBXW8, L3HYPDH, JKAMP,* and *AQP9* in Elliott, Sharp (7). 31 (23 new) of the 122 genes were associated with cognitive functions, intelligence, education, neuroticism, neuropsychiatric and neurodegenerative disorders, such as *NT5C2* (21, 44, 48, 49), *ADAM10* (49, 50), and *GOSR1* (20, 44) **(Supplementary Fig. 8)**.

Gene-priority analysis was performed for 14 brain tissues to examine whether the tissue-specific gene expression levels were related to the associations between genes and ROI volumes (Online Methods). After adjusting for multiple testing (that is, 14*101=1,414 tests) by the Benjamini-Hochberg (B-H) procedure (51) at 0.05 level, we detected nine significant associations, including gene expression in brain hippocampus tissue and gene's association significance with left hippocampus volume, and gene expression in brain cerebellar hemisphere and cerebellum tissues and gene's association significance with pallidum and putamen volumes (p-value<$2.46*10^{-4}$) **(Supplementary Table 17)**. These results showed that genes with higher transcription levels on these brain tissues also had stronger associations with the corresponding brain ROI volumes**.**

Joint analysis with four independent datasets

To validate the UKB GWAS results, we repeated GWAS of 101 ROI volumes separately on data obtained from four other independent studies: PNC (n=537), HCP (n=334), PING (n=461), and ADNI (n=860). Due to the small sample size of these four datasets, the probability of replicating significant findings in the UKB was low. Instead, we checked whether the SNP effect signs were concordant in the five studies and whether the p-value of top UKB SNPs decreased after meta-analysis (Online Methods). Smaller p-values after meta-analysis indicates similar SNP effects in independent samples (52, 53).

The joint analysis was carried out on 3,841,911 SNPs which were present in all five sets of GWAS results. For the 5,940 significant associations (at $4.9*10^{-10}$ significance level), 64.6% (3,839) associations had the same effect signs across the five studies, and 97.5% (5,791) associations had the same effect signs in at least four studies (including UKB). 94.0% (1,880) of the top 2,000 significant associations had smaller p-value after meta-analysis, and 92.3% (5,484) of all the 5,940 associations were enhanced. We then performed meta-analysis on all the 8,944,375 UKB GWAS SNPs (SNPs were allowed to be missing in the four independent datasets). There were more significant associations after meta-analysis: 25,083 significant associations at $5*10^{-8}$ significance level and 14,004 at $4.9*10^{-10}$ significance level **(Supplementary Table 18, Supplementary Fig. 9).**

Genetic correction with other traits

The meta-analysis GWAS results were used to estimate the genetic correlation (gc) with other traits via LDSC (32). As positive controls, we first estimated the genetic correlation between several UKB ROIs volumes (TBV, left/right thalamus proper, left/right caudate, left/right putamen, left/right pallidum, left/right hippocampus, left/right accumbens area) and their corresponding traits studied in the ENIGMA consortium (54). The gc estimates were all significant (p-value<$1.20*10^{-5}$) and average correlation was 0.93 **(Supplementary Table 19).** We then collected 50 sets of publicly available GWAS summary statistics **(Supplementary Table 20)** and calculated their pairwise genetic correlation with ROI volumes **(Supplementary Tables 21)**. We mainly focused on traits that showed evidence of pleiotropy in association lookups. There were 29 significant associations after adjusting for multiple testing (4,900 tests) by the B-H procedure at 0.05 level **(Supplementary Tables 22, Supplementary Fig. 10).**

Significant genetic correlations linked 16 ROI volumes with general cognitive functions, education (education years, college completion), intelligence, numerical reasoning, reaction time, depressive symptoms, neuroticism, worry, ASD, and bipolar disorder (BD) **(Figure 3)**, which matched our findings in SNP and gene level lookups. Particularly, TBV had positive correlations with cognitive functions, education, intelligence, and numerical reasoning (gc range=[0.20,0.24], mean=0.22, p-value range=[$1.54*10^{-11}$,$2.73*10^{-5}$]). Left

posterior cingulate showed positive correlations with cognitive functions, intelligence, and numerical reasoning (gc range=[0.16,0.17], p-value range=[$6.09*10^{-5}$,$1.85*10^{-4}$]). We note that TBV has been adjusted in GWAS of ROIs other than TBV. Right rostral anterior cingulate showed positive correlation with ASD (gc=0.32, p-value=$2.00*10^{-4}$), left rostral middle frontal had positive correlation with BD (gc=0.20, p-value=$1.00*10^{-4}$), and right precuneus had positive correlation with neuroticism (gc=0.17, p-value=$1.20*10^{-4}$). Reaction time had positive genetic correlations with left/right lateral ventricle and X3rd ventricle (gc range=[0.16,0.18], p-value range=[$3.13*10^{-5}$,$1.58*10^{-4}$]), and had negative correlations with left/right pallidum, left/right ventral DC, and white matter (gc range=[-0.20, -0.15], p-value range=[$3.64*10^{-7}$,$1.44*10^{-5}$]). Negative genetic correlations were also found on depressive symptoms (gc=-0.25, p-value=$3.33*10^{-5}$), neuroticism (gc=-0.14, p-value=$2.20*10^{-4}$), and worry (gc=-0.14, p-value=$2.94*10^{-4}$).

Predictive ability of the UKB GWAS results

We examined the out-of-sample prediction power of the UKB GWAS summary statistics using polygenic risk scores prediction (55). We focused the analysis on total brain volume. We first used a ten-fold cross-validation design to examine the prediction power within the UKB sample (Online Methods). Five polygenic profiles were created with p-value thresholds 1, 0.5, 0.05, $5*10^{-4}$ and $5*10^{-8}$, respectively, and we examined the incremental R-squared (Online Methods). The PRS can explain 1.51% of the variance in total brain volume (p-value=$4.42*10^{-110}$) **(Supplementary Table 23)**. We then used the GWAS summary statistics of 19,629 UKB individuals to construct polygenic profiles on subjects in PNC, HCP, PING, and ADNI. The UKB-derived PRS were all significantly associated with the phenotype in all the four independent datasets, and can account for 1.38%-4.36% phenotypic variation (p-value range=[$2.97*10^{-22}$, $1.44*10^{-6}$]). The largest R-squared 4.36% was in PNC dataset with threshold 1 and 224,657 SNP predictors.

DISCUSSION

In this study, we presented GWAS of 101 ROI volumes using data of 19,629 UKB individuals. Our novel contributions include 1) identification of many newly associated genetic variants at SNP, locus, and gene levels; 2) revealing the genetic co-architecture

of brain volume phenotypes and other brain-related complex traits; 3) validation of the UKB results in independent studies; and 4) assessment of the predictive power of UKB GWAS results. Significant (p-value<$4.9*10^{-10}$) associations were found for 54 of the 101 ROIs. With larger sample size, the present study replicated many known genetic variants but also prioritized new ones. Compared to (7), our GWAS not only discovered more genetic variants, but also enriched the degree of (statistical) pleiotropy (56) of the associated genes and characterized the shared genetic influences with cognitive and mental health traits.

However, the current GWAS sample size of ROI volumes (and many other brain imaging phenotypes) is still far from being sufficient. The highly polygenic genetic architecture of ROI volumes requires a larger number of subjects to identify many weak causal SNPs. In the era of sharing GWAS summary statistics, well powered GWAS is essential for ROI volumes to be linked to the genetic co-architecture atlas with other complex traits. For example, a recent study of Watanabe, Stringer (56) to discover the global overview of genetic co-architecture of 2,965 traits only focused on GWAS with sample size larger than 50,000, with the average sample size of selected traits being 256,276. In our genetic correlation analysis, we only obtained limited number of significant correlations, even though many pleiotropic genes were found in association lookups. Therefore, we expect that GWAS of ROI volumes with larger sample size will be available and can further improve our understating of genetic overlaps underlying other traits. Besides increasing the sample size, combining SNP data with external omic information, such as gene expression data (57), may also help elucidate the causal mechanism, improve the prediction performance of SNP data and reveal the genetic connections among traits.

URLs.

ANTs, http://stnava.github.io/ANTs/;

PLINK, https://www.cog-genomics.org/plink2/ ;

GCTA, http://cnsgenomics.com/software/gcta/ ;

METAL, https://genome.sph.umich.edu/wiki/METAL;

FUMA, http://fuma.ctglab.nl/;

MGAMA, https://ctg.cncr.nl/software/magma;

LD Score Regression, https://github.com/bulik/ldsc/ ;

LD Hub, http://ldsc.broadinstitute.org/ldhub/;

MaCH-Admix, http://www.unc.edu/~yunmli/MaCH-Admix ;

NHGRI-EBI GWAS Catalog, https://www.ebi.ac.uk/gwas/home;

The atlas of GWAS Summary Statistics, http://atlas.ctglab.nl/;

UK Biobank, http://www.ukbiobank.ac.uk/resources/ ;

PING, http://pingstudy.ucsd.edu/resources/genomics-core.html ;

PNC,

https://www.ncbi.nlm.nih.gov/projects/gap/cgi-bin/study.cgi?study_id=phs000607.v1.p1;

ADNI, http://adni.loni.usc.edu/data-samples/;

HCP, https://www.humanconnectome.org/.

METHODS

Methods are available in the ***Online Methods*** section.

Note: One supplementary information pdf file and one supplementary zip file are available.

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AUTHOR CONTRIBUTIONS

B.Z., H.Z., and Y.L. designed the study. B.Z. and T.L. performed the experiments and analyzed the data. T.L., J.Z. Y.S., X.W., L.Y., F.Z., and Z.Z. downloaded the datasets,

preprocessed MRI and SNP data, and undertook the quantity controls. B.Z., H.Z., and Y.L. wrote the manuscript with feedback from all authors.

COMPETETING FINANCIAL INTERESTS

The authors declare no competing financial interests.

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ONLINE METHODS

GWAS participants and phenotypes

We performed GWAS separately on five publicly available datasets: the UK Biobank (UKB) study, the Human Connectome Project (HCP) study, the Pediatric Imaging, Neurocognition, and Genetics (PING) study, the Philadelphia Neurodevelopmental Cohort (PNC) study, and the Alzheimer's Disease Neuroimaging Initiative (ADNI) study. The main GWAS made use of data of 19,629 individuals of British ancestry from the UKB study, and the other four GWAS were performed on individuals of European ancestry, see **Supplementary Table 24** for a summary of sample size of each GWAS.

The raw MRI, covariates and SNP data were downloaded from each data resource. We processed the MRI data locally using consistent procedures via advanced normalization tools (ANTs) to generate ROI volume phenotypes for each dataset. The processing steps were detailed in Zhao, Ibrahim (10) and we removed three ROIs (X5th ventricle and left/right lesion) with missing rates > 99%. For each phenotype and continuous covariate variable, we further removed values greater than five times the median absolute deviation from the median value. All individuals were aged between 3 and 92 years. More information about study cohorts can be found in **Supplementary Table 25** and **Supplementary Note.**

Heritability estimation and Genome-wide association analysis

We estimated the proportion of variation explained by all autosomal SNPs in UKB with using univariate GCTA-GREML analysis (40). The adjusted covariates included age (at imaging), age-squared, gender, age-gender interaction, age-squared-gender interaction,

TBV (for ROIs other than TBV itself), as well as the top 40 genetic principle components (PCs) provided by UKB ((58), Data-Field 22009). We performed GWAS for each ROI volume with PLINK (59). The same set of covariates as in GCTA-GREML analysis were adjusted. GWAS were also separately performed on PING, PNC, ADNI, and HCP data. In these four datasets, we adjusted for age, age-squared, gender, age-gender interaction, age-squared-gender interaction, TBV (for ROIs other than TBV itself), and top ten genetic PCs estimated from the SNP data. We also adjusted for the Alzheimer's disease status in ADNI GWAS.

Genomic risk loci characterization and comparison with previous findings

Genomic risk loci were defined using FUMA (31) online platform (version: 1.3.4). We input the UKB GWAS summary statistics obtained from PLINK (59). FUMA first identified independent significant SNPs, which were defined as SNPs with a p-value smaller than the predefined threshold and independent of other significant SNPs at R-squared < 0.6. Using these independent significant SNPs, FUMA then constructed LD block for independent significant SNPs by tagged all SNPs that had a MAF $\geq$ 0.0005 and were in LD (R-squared$\geq$0.6) with at least one of the independent significant SNPs. These SNPs included those from the 1000 Genomes reference panel and may not have been included in the present study. Based on these independent significant SNPs, (independent) lead SNPs were also identified as those that were independent from each other (R-squared<0.1). If LD blocks of independent significant SNPs were closed (<250 kb based on the closest boundary SNPs of LD blocks), they were merged to a single genomic locus. Thus, each genomic locus could contain more than one independent significant SNPs and lead SNPs. More details can be found in Watanabe, Taskesen (31). Independent significant SNPs and all the tagged SNPs were subsequently searched by FUMA on NHGRI-EBI GWAS catalog ((33), version: 2019-01-31) to look for their reported SNP associations (p-value<$9*10^{-6}$) with any traits.

Gene-based association analysis, functional annotation and gene-property analysis

Gene-based association analysis was carried out for 18,796 protein-coding genes using MAGMA (30), which was also implemented in FUMA (31). SNPs were mapped according

to their psychical positions, then the gene-based p-values were calculated by the GWAS summary statistics of mapped SNPs.

In functional annotation and mapping analysis, SNPs signals were annotated with their biological functionality and then were linked to genes by a combination of positional, eQTL, and 3D chromatin interaction mappings (31). Specifically, independent significant SNPs and all the tagged SNPs were first annotated for functional consequences on gene functions (e.g., intergenic, intronic, exonic) using ANNOVAR (60). Functionally-annotated SNPs were then mapped to 35,808 candidate genes based on physical position on the genome (tissue/cell types for 15-core chromatin state: brain), eQTL associations (tissue types: GTEx v7 brain (61), BRAINEAC (62), and CommonMind Consortium (63)) and chromatin interaction mapping (built-in chromatin interaction data: dorsolateral prefrontal cortex, hippocampus (64); annotate enhancer/promoter regions: E053-E082 brain (65)). We used default values for all other parameters.

For the detected genes, we performed lookups on NHGRI-EBI GWAS catalog ((33), version: 2019-01-31) again to explore the previously reported associations with the same or other traits. We focused on traits including cognitive functions (such as general cognitive ability, cognitive performance, and empathy quotient), intelligence, educational attainment, math ability (such as highest math class taken and self-reported math ability), reaction time, neuroticism, neurodegenerative diseases (such as Alzheimer's disease and Parkinson's disease), and neuropsychiatric disorders (such as major depressive disorder, Schizophrenia, and bipolar disorder). For the 14 brain tissues (GTEx v7, (61)), we also performed gene-priority analysis via MAGMA (30). That is, for each candidate gene, whether its tissue-specific expression levels can be linked to the strength of its association with ROI volumes.

Meta-analysis of GWAS results

We meta-analyzed the UKB, PING, PNC, ADNI, and HCP GWAS summary results by METAL (https://genome.sph.umich.edu/wiki/METAL) with the sample-size weighted approach. Since the sample sizes of other four datasets were small, we removed the SNPs that were not presented in the UKB data.

Genetic correlation estimation with LDSC

The LD Hub (v1.9.1, http://ldsc.broadinstitute.org/ldhub/) was used to estimate the genetic correlation between several UKB ROIs volumes and their corresponding traits studied in the ENIGMA consortium (54). Then the LDSC software (v1.0.0, https://github.com/bulik/ldsc) was used to estimate the pairwise genetic correlation with 50 sets of collected GWAS summary statistics. We used the pre-calculated LD scores provided by LDSC (https://data.broadinstitute.org/alkesgroup/LDSCORE/), which were computed using 1000 Genomes European data. We used HapMap3 SNPs and removed all SNPs on chromosome 6 in the MHC region.

Polygenic scoring

Polygenic profiles were created to examine the out-of-sample prediction power of the GWAS results. Specifically, we used PLINK (59) to generate risk scores in testing data by summarizing across pruned (window size 50, step 5, R-squared = 0.2) SNP alleles, weighed by their effect sizes estimated from training data. We randomly divided the 19,629 UKB individuals into ten folds, then used nine of these folds as training data to rerun GWAS, and created polygenic profiles on the individuals in the remaining fold, which served as testing data. We repeated this procedure ten times such that each fold alternated to serve as the testing data for exactly one time. Then we used the UKB GWAS results to perform prediction on ADNI, PING, PNC and HCP data. The prediction accuracy was evaluated on all samples in the four testing sets (with phenotype and SNP data available), not limited to individuals of European ancestry used in GWAS. We tried five p-value thresholds for SNP predictor selection: 1, 0.5, 0.05, $5*10^{-4}$ and $5*10^{-8}$. The association between polygenic profile and brain volume was estimated and tested in linear regression model, adjusting for the effects of age and gender. The additional variance of brain volume that can be explained by polygenic profile was used to measure the prediction power.

Data availability

All UKB and meta-analysis GWAS summary statistics of 101 ROI volumes can be found at: https://med.sites.unc.edu/bigs2/data/gwas-summary-statistics/.

Figure and Table Legends

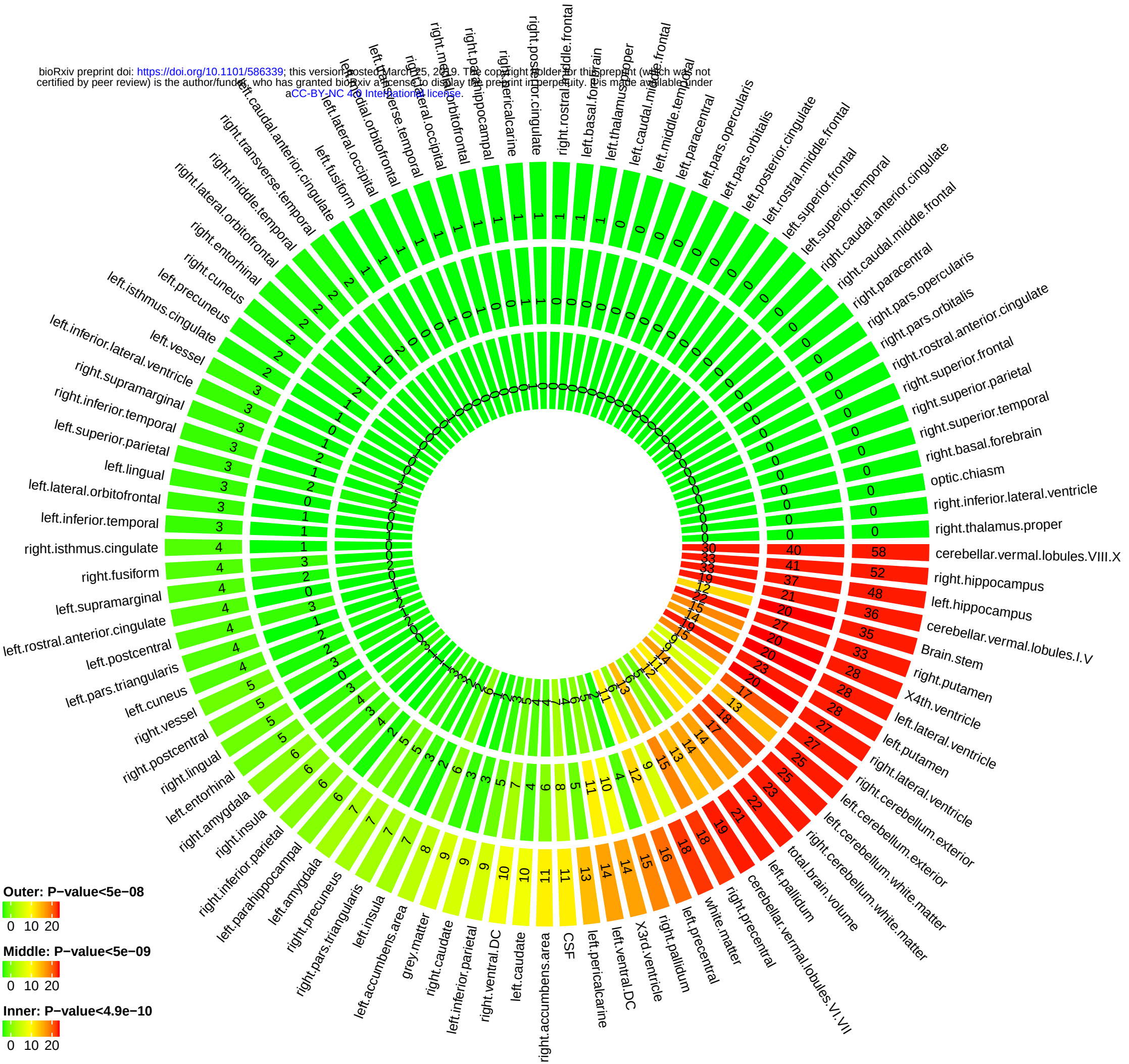
Figure 1. Number of independent significant SNP associations discovered in UKB GWAS at different significance levels. Outer layer: p-value $<5*10^{-8}$; middle layer: p-value $<5*10^{-9}$; and inner layer: p-value $<4.9*10^{-10}$.

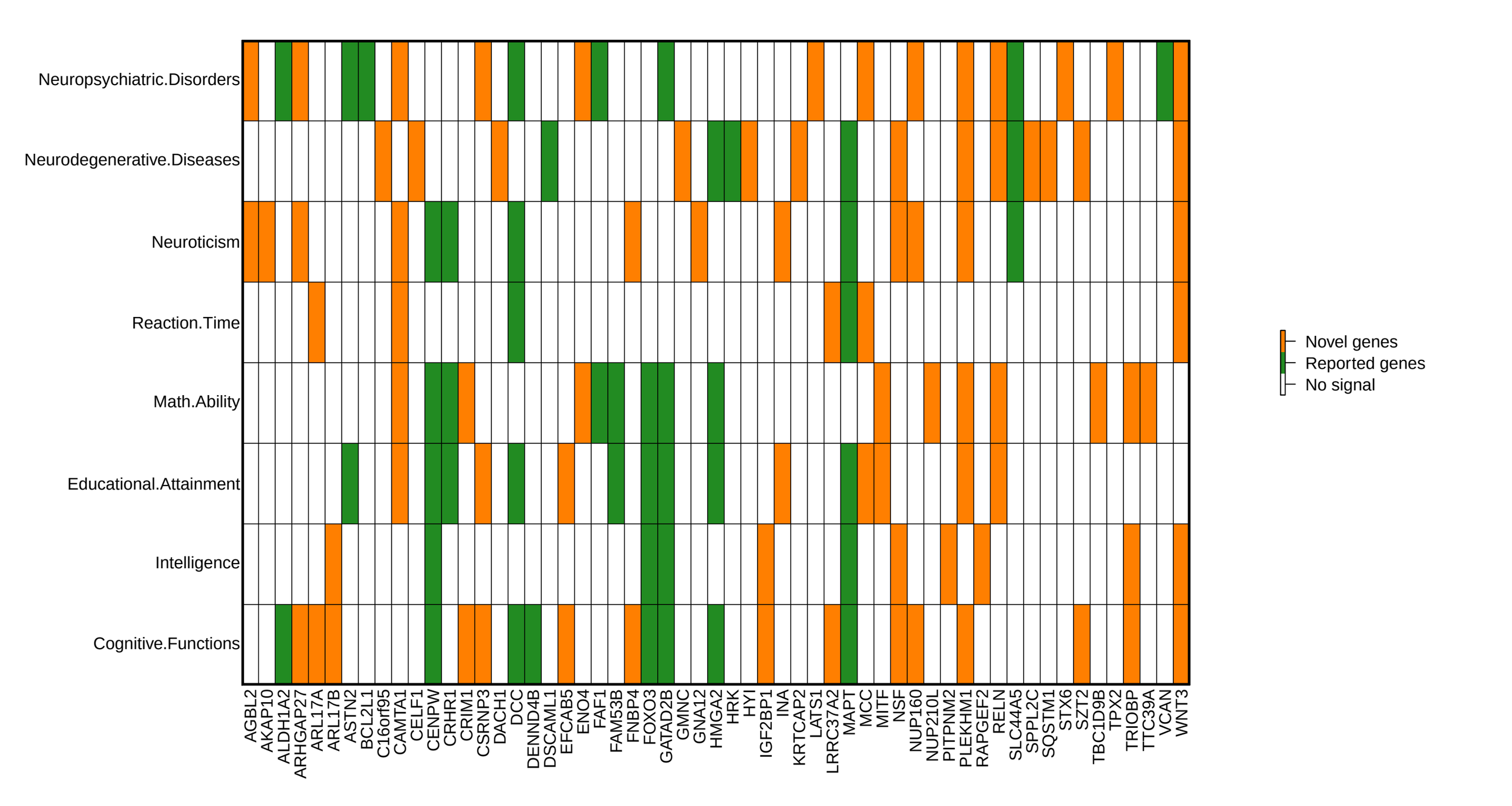
Figure 2. Genes identified in gene-based association analysis of ROI volumes that have been linked to cognitive traits and mental health disease/disorders in previous GWAS.

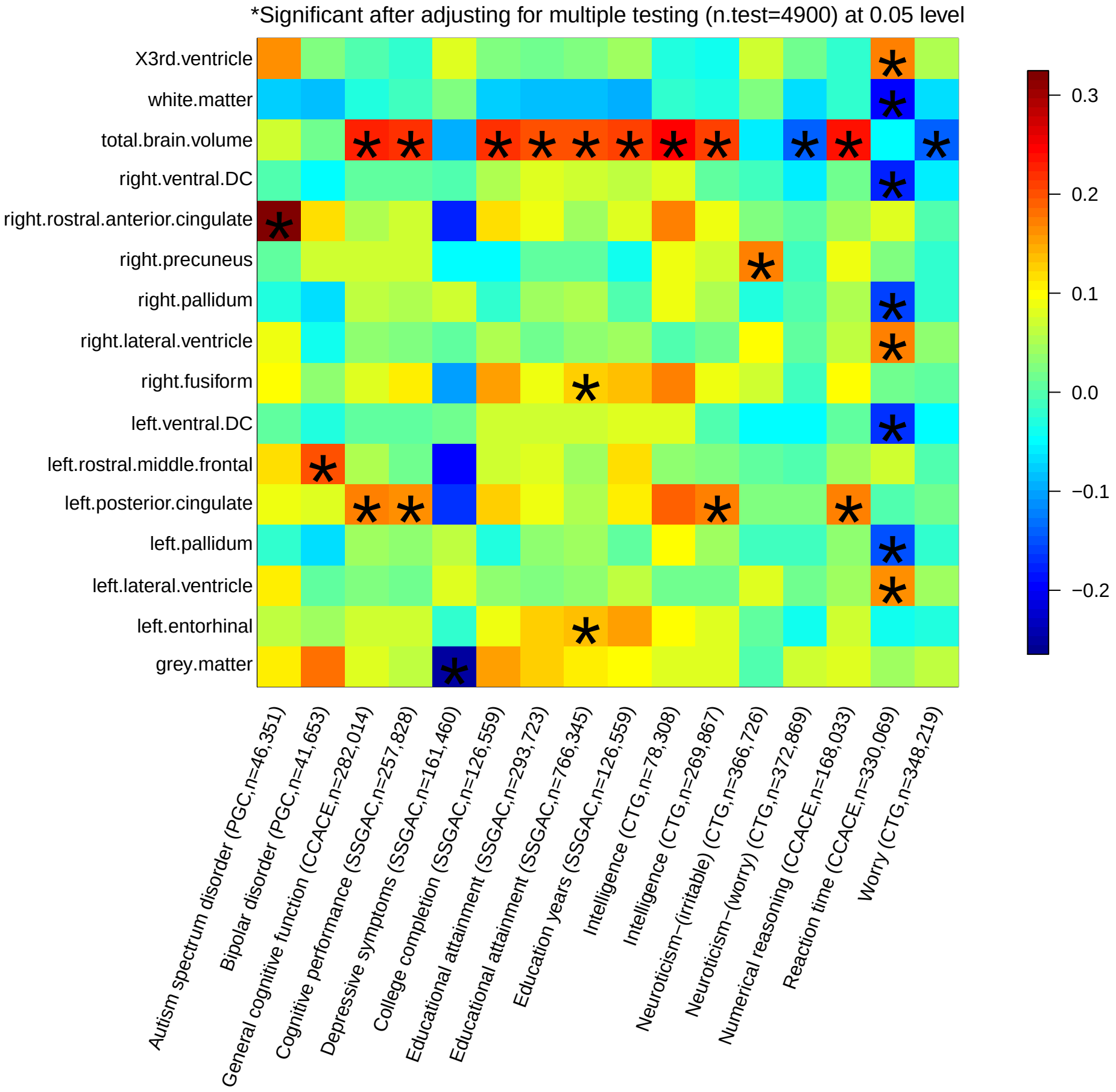
Figure 3. Selected pairwise genetic correlations between ROI volumes and other traits

Table 1. List of significant genetic risk loci identified by UKB GWAS at $4.9*10^{-10}$ significant level. RSID: rsID of the top lead SNP; Position: position of top lead SNP.

Table 2. List of significant genes identified by gene-based association analysis in UKB data at $2*10^{-8}$ significant level.




Neuropsychiatric.Disorders
Neurodegenerative.Diseases
Neuroticism
Reaction.Time
Math.Ability
Educational.Attainment
Intelligence
Cognitive.Functions
AGBL2
AKAP10
ALDH1A2
ARHGAP27
ARL17A
ARL17B
ASTN2
BCL2L1
C16orf95
CAMTA1
CELF1
CENPW
CRHR1
CRIM1
CSRNP3
DACH1
DCC
DENND4B
DSCAML1
EFCAB5
ENO4
FAF1
FAM53B
FNBP4
FOXO3
GATAD2B
GMNC
GNA12
HMGA2
HRK
HYI
IGF2BP1
INA
KRTCAP2
LATS1
LRRC37A2
MAPT
MCC
MITF
NSF
NUP160
NUP210L
PITPNM2
PLEKHM1
RAPGEF2
RELN
SLC44A5
SPPL2C
SQSTM1
STX6
SZT2
TBC1D9B
TPX2
TRIOBP
TTC39A
VCAN
WNT3
Novel genes
Reported genes
No signal


*Significant after adjusting for multiple testing (n.test=4900) at 0.05 level
X3rd.ventricle
white.matter
total.brain.volume
right.ventral.DC
right.rostral.anterior.cingulate
right.precuneus
right.pallidum
right.lateral.ventricle
right.fusiform
left.ventral.DC
left.rostral.middle.frontal
left.posterior.cingulate
left.pallidum
left.lateral.ventricle
left.entorhinal
grey.matter
0.3
0.2
0.1
0.0
−0.1
−0.2
Autism spectrum disorder (PGC,n=46,351)
Bipolar disorder (PGC,n=41,653)
General cognitive function (CCACE,n=282,014)
Cognitive performance (SSGAC,n=257,828)
Depressive symptoms (SSGAC,n=161,460)
College completion (SSGAC,n=126,559)
Educational attainment (SSGAC,n=293,723)
Educational attainment (SSGAC,n=766,345)
Education years (SSGAC,n=126,559)
Intelligence (CTG,n=78,308)
Intelligence (CTG,n=269,867)
Neuroticism−(irritable) (CTG,n=366,726)
Neuroticism−(worry) (CTG,n=372,869)
Numerical reasoning (CCACE,n=168,033)
Reaction time (CCACE,n=330,069)
Worry (CTG,n=348,219)

RSID	ROI	CHR	Position	p-value	p-value (second ROI, if any)	p-value (third ROI, if any)	Start position of the locus	End position of the locus
rs2817145	left.caudate	1	3133422	9.29E-16			3118674	3149789
rs2817145	right.caudate	1	3133422	7.69E-13			3121877	3149789
rs3120124	left.pallidum	1	43764165	7.95E-11			43760070	43949718
rs6658111	X4th.ventricle	1	47980916	2.60E-18			47712057	47980916
rs6658111	left.parahippocampal	1	47980916	2.62E-16			47915175	47993332
rs6658111	CSF	1	47980916	3.58E-10			47974123	47980916
rs3176459	left.amygdala	1	51437247	1.48E-10			50828316	51526458
rs12072311	left.cerebellum.exterior	1	51572465	1.51E-12			50882506	52215089
rs12072311	right.cerebellum.exterior	1	51572465	5.49E-14			50882506	52368501
rs74091739	cerebellar.vermal.lobules.VIII.X	1	76028726	3.21E-45			75818830	76095959
rs2748444	cerebellar.vermal.lobules.I.V	1	76010959	1.23E-19			75946751	76091842
rs76934732	cerebellar.vermal.lobules.VI.VII	1	76013268	2.72E-11			75946751	76073887
rs11392431, rs2991716	right.inferior.parietal, left.inferior.parietal	1	88423056, 88423966	3.10E-10	3.22E-10		88196463	88435308
rs1044595	white.matter	1	180943529	4.64E-13			180940588	181017348
rs1452628	grey.matter	1	215139887	7.34E-12			215134041	215327772
rs1030088	left.inferior.temporal	2	65937745	4.22E-10			65937041	65999623
rs34111434	right.transverse.temporal	2	150048607	3.60E-10			149945701	150084200
rs5835889	right.hippocampus	2	162845572	3.78E-10			162314229	162891848
rs5835889	left.hippocampus	2	162845572	9.80E-12			162796517	162891848
rs759663	cerebellar.vermal.lobules.VIII.X	2	208080975	2.76E-12			208017033	208102505
rs2234675	left.cerebellum.exterior	2	223085955	1.19E-10			223046616	223085955
rs12636275	right.inferior.temporal	3	89523038	9.00E-11			89451721	89751353
rs12633109	right.vessel	3	146392141	2.89E-11			146337667	146396844
rs2279829	right.pars.triangularis	3	147106319	4.52E-14			147090583	147224629
rs2279829	left.pars.triangularis	3	147106319	3.56E-10			147095294	147224629
rs2279829	left.postcentral	3	147106319	1.87E-10			147095294	147140680
rs2279829	right.postcentral	3	147106319	7.02E-12			147095294	147268599
rs34859790, rs61500084, rs61500084	left.inferior.lateral.ventricle, left.lateral.ventricle, right.lateral.ventricle	3	190652830, 190636749,190636749	4.39E-11	2.80E-13	2.59E-17	190591418	190678743
rs10706986	left.accumbens.area, right.accumbens.area	3	190646282	2.04E-10	1.40E-13		190592836	190678743
rs13073466	cerebellar.vermal.lobules.VIII.X	3	192617930	1.38E-10			192613518	192716379
rs10034561	cerebellar.vermal.lobules.VIII.X	4	66701964	2.28E-16			66629087	66986038
rs17755998	cerebellar.vermal.lobules.VI.VII	4	66714289	1.08E-21			66653473	67034895
rs13107325	left.accumbens.area	4	103188709	5.05E-19			102702364	103387161
rs13107325	right.accumbens.area	4	103188709	9.64E-16			102702364	103388441
rs13107325	white.matter	4	103188709	2.52E-13			103001649	103387161
rs10478102	cerebellar.vermal.lobules.VIII.X	5	112398770	4.99E-15			112379434	112442272
rs152234	white.matter	5	179335060	2.73E-12			179287576	179370127
rs2819861	left.putamen	6	45424940	8.20E-12			45407654	45510107
rs35405209	right.pallidum	6	45428508	2.42E-10			45407654	45461253
rs12215796	Brain.stem	6	55715715	4.20E-10			55679959	55717728
rs9296804	left.cerebellum.white.matter	6	55717728	4.14E-10			55712608	55717728
rs9345125	right.precentral	6	92002653	1.14E-10			92002569	92010010
rs2153960	Brain.stem, left.ventral.DC, right.ventral.DC	6	108988184	7.69E-12	1.05E-11	6.32E-12	108861264	109020032
rs2764264	total.brain.volume	6	108934461	5.16E-13			108861264	109019323
rs11759026	total.brain.volume	6	126792095	1.25E-16			126659043	127167072
rs62435771	CSF	6	169470668	3.26E-10			169469707	169555989
rs368699386	left.lateral.ventricle	7	2803883	1.09E-14			2745766	2912928
rs368699386, rs529794364	X3rd.ventricle, right.lateral.ventricle	7	2803883, 2791033	3.28E-12	3.17E-12		2752152	2912928
rs151057105	left.ventral.DC, right.ventral.DC	7	54944920	7.38E-14	5.51E-14		54909928	55015505
rs4006770	cerebellar.vermal.lobules.VIII.X	7	103361695	8.81E-26			103200019	103396186
rs12705150	cerebellar.vermal.lobules.VI.VII	7	103356275	2.53E-11			103356062	103373126
rs142005327	white.matter	7	120969969	2.21E-11			120954908	121040782
rs7778054	cerebellar.vermal.lobules.I.V	7	156095814	1.40E-11			156066471	156170722
rs196814	left.pallidum	8	24716594	2.18E-10			24679371	24764896
rs13268108	right.cerebellum.exterior	8	142039586	1.30E-10			142033034	142042870
rs11789773	right.postcentral	9	76157130	1.10E-10			75849823	76193770
rs12344396	left.cerebellum.white.matter	9	118921327	3.75E-16			118917007	119098518
rs1405	Brain.stem	9	118954624	7.48E-16			118917007	119002278
rs72754248	left.cerebellum.exterior, right.cerebellum.exterior	9	119061396	7.72E-39	1.25E-37		118917007	119134875
rs72754266	right.cerebellum.white.matter	9	119093757	2.51E-15			118917007	119098518
rs147269950	X4th.ventricle	9	119098518	1.00E-14			119061396	119098518
rs7030607	left.hippocampus, right.hippocampus	9	119245183	6.81E-19	1.37E-14		119241165	119479868
rs77641763	cerebellar.vermal.lobules.VIII.X	9	140265782	1.20E-10			140241209	140278206
rs4748994	left.caudate	10	25300609	3.10E-11			25233290	25377328
rs6584542	total.brain.volume	10	104995495	3.69E-14			104965551	105176914
rs4752582	left.cerebellum.exterior	10	123443605	1.82E-10			123443605	123443605
rs4046908	white.matter	11	1483117	7.69E-11			1472475	1632590
rs10770131	left.lateral.ventricle, right.lateral.ventricle	11	10682248	5.99E-12	1.89E-11		10655021	10699750
rs1187162	left.caudate	11	92011126	4.02E-11			92001738	92036512
rs1187162	right.caudate	11	92011126	1.38E-10			92001738	92315907
rs688968	left.putamen, right.putamen	11	117406739	7.36E-13	5.10E-11		117383215	117429845

rs7936928	CSF	11	130279168	1.28E-16			130260673	130297957
rs11062908	left.amygdala, right.amygdala	12	4005222	1.42E-16	1.34E-16		4004752	4013260
rs17178006	right.hippocampus	12	65718299	2.84E-32			65386792	66065136
rs17178006	left.hippocampus	12	65718299	9.50E-33			65400292	66065136
rs17178006	right.amygdala	12	65718299	3.93E-10			65718299	65874956
rs8756	total.brain.volume	12	66359752	4.42E-14			66326943	66389968
rs11111094	left.cerebellum.white.matter, right.cerebellum.white.matter	12	102328806	3.03E-31	9.74E-28		102096012	102946220
rs11111094	Brain.stem	12	102328806	3.54E-22			102319002	102866662
rs35872193	right.cerebellum.exterior	12	102442428	9.61E-12			102394872	102962583
rs12146713	left.lateral.ventricle, right.lateral.ventricle	12	106476805	1.86E-11	1.88E-13		106476805	106510413
rs10778498	Brain.stem	12	107063340	5.77E-14			106960182	107522203
rs146607495	left.hippocampus	12	117319202	1.24E-23			117229539	117506632
rs146607495	right.hippocampus	12	117319202	5.04E-25			117234676	117528064
rs17126556	CSF	14	53942788	6.98E-12			53917808	53996110
rs8014725	right.putamen	14	56186953	2.81E-32			55921561	56206214
rs8017172	left.putamen	14	56199048	2.57E-28			55998163	56206214
rs8014725	left.pallidum	14	56186953	7.17E-11			56177508	56201197
rs76341705	left.pericalcarine	14	59628679	9.97E-21			59449138	59916960
rs76341705	left.cuneus	14	59628679	2.90E-12			59585670	59669948
rs147148763, rs5809016	left.precuneus, right.precuneus	14	59631075, 59627434	2.87E-10	4.22E-12		59588323	59860859
rs5809016, rs73313052	left.inferior.parietal, right.pericalcarine	14	59627434, 59625997	5.71E-11	2.56E-10		59588323	59669948
rs2033939	right.precentral	15	39633904	4.46E-63			39355860	39670175
rs2033939	left.precentral	15	39633904	5.11E-73			39534404	39670175
rs28520337, rs4924345	left.supramarginal, right.supramarginal	15	39647894, 39639898	1.86E-10	2.18E-12		39617295	39664000
rs34680120	left.superior.parietal	15	39664000	5.81E-11			39631771	39664000
rs4775006	X4th.ventricle	15	58215727	1.01E-20			58188859	58385264
rs12921170	right.lateral.ventricle	16	87227397	4.01E-24			87211783	87357985
rs4843552	left.lateral.ventricle	16	87233516	2.93E-22			87211963	87266245
rs4843550, rs4843560	X3rd.ventricle, X4th.ventricle	16	87236383, 87225143	1.86E-14	4.78E-17		87220694	87265438
rs3833159	right.precuneus	17	27953390	3.89E-11			27925812	28553639
rs3110494	left.insula	17	27976597	1.01E-13			27935546	28553639
rs3110494	right.insula	17	27976597	5.44E-13			27935546	28538715
rs78777685	cerebellar.vermal.lobules.I.V	17	35257232	5.46E-23			35215422	35266724
rs118087478	total.brain.volume	17	44051589	1.03E-13			43463493	44865603
rs11665242, rs6508230	left.putamen, right.putamen	18	50907127, 50884924	2.25E-15	2.52E-15		50555225	51061399
rs60575064	X3rd.ventricle	19	33565344	9.33E-12			33525467	33631912
rs35255138	cerebellar.vermal.lobules.I.V	19	41197268	3.51E-11			41148797	41197268
rs6121038	right.pallidum	20	30254773	2.20E-12			29420066	30437522
rs6121038	left.pallidum	20	30254773	1.17E-12			29509439	30439298
rs6121038	right.putamen	20	30254773	3.01E-10			30235302	30434084
rs6062237, rs6062264	right.ventral.DC, left.ventral.DC	20	61154437, 61154871	2.55E-12	2.16E-14		61141981	61158050
rs34134374	left.lateral.ventricle	22	38138284	4.52E-10			38102776	38304858

Gene Symbol	Gene ID	ROI	CHR	Start	Stop	P-value					
ABHD12	ENSG00000100997	left.pallidum	20	25275379	25371619	4.41E-09					
ADAMTS8	ENSG00000134917	left.basal.forebrain, CSF	11	130274820	130298888	3.62E-09	1.76E-15				
AGBL2	ENSG00000165923	left.pallidum, right.pallidum	11	47681143	47736941	4.33E-10	1.72E-10				
AKAP10	ENSG00000108599	left.pallidum	17	19807615	19881656	9.90E-09					
AL133373.1	ENSG00000268657	right.accumbens.area	14	92038788	92041383	7.51E-10					
AL162431.1	ENSG00000234237	white.matter	1	180941695	180949868	1.60E-10					
ALDH1A2	ENSG00000128918	X4th.ventricle	15	58245622	58790065	6.50E-13					
AMZ1	ENSG00000174945	X3rd.ventricle, X4th.ventricle, left.lateral.ventricle, right.lateral.ventricle	7	2719156	2815134	2.98E-11	1.34E-08	1.39E-15	9.19E-13		
ANGPT1	ENSG00000154188	left.paracentral	8	108261721	108510283	1.81E-08					
ARHGAP27	ENSG00000159314	total.brain.volume	17	43471275	43511787	3.65E-09					
ARL17A	ENSG00000185829	total.brain.volume	17	44594068	44657088	4.91E-11					
ARL17B	ENSG00000228696	total.brain.volume	17	44352150	44439130	1.09E-09					
ASTN2	ENSG00000148219	left.hippocampus, right.hippocampus	9	119187504	120177348	1.89E-14	1.00E-13				
BCL2L1	ENSG00000171552	left.putamen, left.pallidum, right.putamen, right.pallidum	20	30252255	30311792	1.60E-08	2.54E-11	2.58E-09	4.86E-11		
BLVRA	ENSG00000106605	cerebellar.vermal.lobules.I.V	7	43798279	43846939	1.41E-08					
BMP4	ENSG00000125378	right.lateral.orbitofrontal	14	54416454	54425479	3.07E-09					
BMP5	ENSG00000112175	Brain.stem	6	55618443	55740362	7.34E-09					
C16orf95	ENSG00000260456	X3rd.ventricle, X4th.ventricle, left.lateral.ventricle, right.lateral.ventricle	16	87117168	87351022	4.44E-15	3.51E-12	1.80E-19	1.47E-22		
C1orf185	ENSG00000204006	cerebellar.vermal.lobules.I.V, left.cerebellum.exterior, right.cerebellum.exterior	1	51567906	51613752	8.68E-09	1.09E-12	2.60E-13			
C20orf166	ENSG00000174407	left.ventral.DC, right.ventral.DC	20	61147660	61167971	1.28E-10	3.46E-09				
CAMTA1	ENSG00000171735	left.cerebellum.white.matter, right.cerebellum.white.matter	1	6845384	7829766	1.71E-08	3.58E-10				
CCDC53	ENSG00000120860	Brain.stem, left.cerebellum.white.matter, right.cerebellum.exterior, right.cerebellum.white.matter	12	102406705	102455927	6.83E-09	2.73E-13	9.11E-11	1.24E-13		
CDC20	ENSG00000117399	left.pallidum, right.pallidum	1	43824626	43828874	4.85E-10	1.54E-08				
CELF1	ENSG00000149187	left.pallidum, right.pallidum	11	47487496	47587121	5.90E-09	8.56E-10				
CENPW	ENSG00000203760	left.lateral.orbitofrontal, X4th.ventricle, total.brain.volume	6	126661320	126670021	1.17E-11	1.84E-08	5.15E-14			
CMPK1	ENSG00000162368	X4th.ventricle	1	47799469	47844511	1.18E-08					
COA1	ENSG00000106603	cerebellar.vermal.lobules.I.V	7	43648055	43769316	5.92E-09					
CORO6	ENSG00000167549	left.insula, right.precuneus, right.insula	17	27941774	27949925	2.47E-09	8.82E-09	1.47E-08			
CRHR1	ENSG00000120088	total.brain.volume	17	43699267	43913194	4.39E-11					
CRIM1	ENSG00000150938	left.inferior.lateral.ventricle	2	36583069	36778278	9.77E-10					
CSRNP3	ENSG00000178662	total.brain.volume	2	166326157	166545917	4.31E-09					
DAAM1	ENSG00000100592	left.pericalcarine, right.precuneus	14	59655364	59838123	6.26E-11	3.76E-09				
DACH1	ENSG00000165659	right.vessel	13	72012098	72441330	2.26E-09					
DCC	ENSG00000187323	left.insula, left.putamen, right.putamen	18	49866542	51057784	1.60E-08	3.49E-22	1.95E-22			
DEFB119	ENSG00000180483	left.pallidum	20	29964967	29978406	8.39E-09					
DEFB121	ENSG00000204548	left.pallidum	20	29992648	30000641	7.51E-09					
DEFB124	ENSG00000180383	left.pallidum	20	30053309	30064560	5.62E-10					
DENND4B	ENSG00000198837	left.inferior.lateral.ventricle	1	153901977	153919172	1.60E-08					
DRAM1	ENSG00000136048	Brain.stem, left.cerebellum.white.matter, right.cerebellum.white.matter	12	102271129	102405908	4.35E-15	1.22E-19	9.82E-18			
DSCAML1	ENSG00000177103	left.putamen	11	117298489	117688240	8.59E-09					
DUSP15	ENSG00000149599	left.pallidum	20	30435440	30458550	8.23E-09					
EFCAB5	ENSG00000176927	left.insula, right.precuneus, right.insula	17	28256218	28435470	2.26E-11	4.18E-10	5.36E-11			
EFNA1	ENSG00000169242	white.matter	1	155099936	155107333	1.28E-08					
EGFR	ENSG00000146648	left.ventral.DC	7	55086714	55324313	4.76E-09					
ENKUR	ENSG00000151023	left.caudate, right.caudate	10	25270908	25305089	5.35E-12	4.47E-10				
ENO4	ENSG00000188316	right.insula, left.putamen, right.putamen	10	118609023	118671299	1.17E-08	1.67E-08	3.16E-09			
FAF1	ENSG00000185104	right.cerebellum.exterior	1	50905150	51425935	1.42E-08					
FAM53B	ENSG00000189319	left.isthmus.cingulate	10	126307861	126432838	1.56E-08					
FNBP4	ENSG00000109920	left.pallidum, right.pallidum	11	47738072	47788995	2.45E-10	9.00E-11				
FOXO3	ENSG00000118689	Brain.stem, left.cerebellum.white.matter, left.ventral.DC, right.ventral.DC, total.brain.volume	6	108881038	109005977	4.47E-11	1.81E-08	9.51E-11	3.34E-11	1.21E-11	
GATAD2B	ENSG00000143614	left.inferior.lateral.ventricle	1	153777201	153895451	6.42E-09					
GCAT	ENSG00000100116	left.lateral.ventricle	22	38203912	38213183	3.53E-09					
GINM1	ENSG00000055211	left.hippocampus, right.hippocampus	6	149887430	149912884	2.22E-09	3.94E-10				
GINS1	ENSG00000101003	left.pallidum	20	25388363	25433264	1.86E-09					
GMNC	ENSG00000205835	left.lateral.ventricle, left.inferior.lateral.ventricle, right.lateral.ventricle, right.accumbens.area	3	190570666	190610218	8.53E-11	1.56E-09	2.89E-13	1.25E-09		
GNA12	ENSG00000146535	X3rd.ventricle, X4th.ventricle, left.lateral.ventricle, right.lateral.ventricle	7	2767746	2883958	1.52E-10	2.20E-09	8.21E-12	6.24E-10		
GNPTAB	ENSG00000111670	right.cerebellum.white.matter	12	102139275	102224716	9.37E-09					
GPATCH1	ENSG00000076650	X3rd.ventricle	19	33571786	33621448	8.87E-12					
HMGA2	ENSG00000149948	total.brain.volume	12	66217911	66360075	4.59E-12					
HRK	ENSG00000135116	left.hippocampus, right.hippocampus	12	117293949	117319246	3.59E-12	7.79E-12				
HYI	ENSG00000178922	left.pallidum	1	43916824	43919660	3.58E-09					
IGF1	ENSG00000017427	left.cerebellum.white.matter, right.cerebellum.exterior, right.cerebellum.white.matter	12	102789645	102874423	1.80E-11	5.05E-09	1.22E-11			
IGF2BP1	ENSG00000159217	right.medial.orbitofrontal	17	47074774	47133012	1.19E-08					
INA	ENSG00000148798	total.brain.volume	10	105036920	105050108	2.10E-11					
KANSL1	ENSG00000120071	total.brain.volume	17	44107282	44302733	1.78E-10					
KATNA1	ENSG00000186625	right.hippocampus	6	149916009	149970108	1.58E-08					
KLF7	ENSG00000118263	cerebellar.vermal.lobules.VIII.X	2	207938861	208031991	1.69E-09					
KRTCAP2	ENSG00000163463	white.matter	1	155141884	155145951	3.15E-09					
LATS1	ENSG00000131023	left.hippocampus, right.hippocampus	6	149979289	150039392	1.38E-08	1.30E-08				
LEMD3	ENSG00000174106	left.hippocampus, right.hippocampus	12	65563351	65642107	4.40E-14	1.96E-14				
LOXL1	ENSG00000129038	right.lateral.orbitofrontal	15	74218330	74244478	3.41E-09					
LRRC37A	ENSG00000176681	total.brain.volume	17	44370099	44415160	6.58E-09					

LRRC37A2	ENSG00000238083	total.brain.volume	17	44588877	44633016	4.91E-11					
LRRC49	ENSG00000137821	white.matter	15	71145578	71342414	4.07E-09					
LYPD6B	ENSG00000150556	right.transverse.temporal	2	149894621	150071776	4.08E-11					
MAPT	ENSG00000186868	total.brain.volume	17	43971748	44105700	1.59E-10					
MCC	ENSG00000171444	cerebellar.vermal.lobules.VIII.X	5	112357796	112824527	2.90E-12					
METTL10	ENSG00000203791	left.isthmus.cingulate, left.hippocampus, right.hippocampus	10	126436718	126480439	4.21E-09	1.16E-08	9.88E-09			
MITF	ENSG00000187098	white.matter	3	69788586	70017488	1.74E-08					
MPL	ENSG00000117400	left.pallidum	1	43803478	43818443	3.95E-09					
MR1	ENSG00000153029	white.matter	1	181003067	181031074	7.37E-10					
MSRB3	ENSG00000174099	right.inferior.parietal, left.hippocampus, right.hippocampus	12	65672423	65882024	4.77E-11	7.67E-23	3.34E-24			
MTX1	ENSG00000173171	white.matter	1	155178490	155183615	9.35E-10					
MYLK2	ENSG00000101306	left.pallidum, right.pallidum, right.pallidum	20	30407111	30422492	2.42E-10	4.10E-09	3.93E-10			
NADK	ENSG00000008130	right.ventral.DC	1	1682671	1711896	4.84E-09					
NINL	ENSG00000101004	left.pallidum	20	25433341	25566153	2.76E-10					
NOL12	ENSG00000100101	X3rd.ventricle, left.lateral.ventricle	22	38077680	38170137	9.00E-09	1.04E-08				
NSF	ENSG00000073969	total.brain.volume	17	44668035	44834830	4.76E-11					
NSRP1	ENSG00000126653	left.insula, right.precuneus, right.insula	17	28442539	28513493	5.76E-11	6.51E-10	1.66E-10			
NUMBL	ENSG00000105245	cerebellar.vermal.lobules.I.V	19	41172596	41196877	5.12E-09					
NUP160	ENSG00000030066	left.pallidum, right.pallidum	11	47799639	47870107	2.98E-10	1.55E-10				
NUP210L	ENSG00000143552	left.inferior.lateral.ventricle	1	153965161	154127592	7.17E-10					
NUP37	ENSG00000075188	left.cerebellum.white.matter, right.cerebellum.exterior, right.cerebellum.white.matter	12	102467967	102513902	5.19E-13	1.42E-10	3.01E-13			
NUP43	ENSG00000120253	left.hippocampus, right.hippocampus	6	150045451	150070801	9.04E-09	6.67E-09				
ORC5	ENSG00000164815	left.entorhinal	7	103766788	103848495	9.01E-10					
P4HA2	ENSG00000072682	CSF	5	131527531	131631008	5.35E-09					
PANX2	ENSG00000073150	left.fusiform	22	50609160	50618723	5.30E-10					
PAPPA	ENSG00000182752	cerebellar.vermal.lobules.I.V, Brain.stem, left.cerebellum.exterior, left.cerebellum.white.matter, right.cerebellum.exterior, right.cerebellum.white.matter	9	118916083	119164601	1.53E-09	2.88E-13	7.38E-22	1.15E-18	1.54E-20	6.52E-18
PARPBP	ENSG00000185480	left.cerebellum.white.matter, right.cerebellum.exterior, right.cerebellum.white.matter	12	102513956	102591298	2.19E-10	1.34E-09	3.42E-11			
PCGF6	ENSG00000156374	total.brain.volume	10	105062553	105110891	1.79E-11					
PDCD11	ENSG00000148843	total.brain.volume	10	105156405	105206049	5.17E-10					
PIGF	ENSG00000151665	cerebellar.vermal.lobules.I.V	2	46808076	46844258	1.25E-08					
PITPNM2	ENSG00000090975	cerebellar.vermal.lobules.I.V	12	123468027	123634562	8.88E-09					
PLEKHM1	ENSG00000225190	total.brain.volume	17	43513266	43568115	1.58E-09					
PRDM16	ENSG00000142611	left.caudate	1	2985732	3355185	3.58E-09					
PRDM5	ENSG00000138738	grey.matter	4	121606074	121844025	1.44E-10					
PYGB	ENSG00000100994	left.pallidum	20	25228705	25278650	3.47E-10					
RAPGEF2	ENSG00000109756	left.caudate	4	160025330	160281321	1.46E-09					
RELN	ENSG00000189056	cerebellar.vermal.lobules.VIII.X	7	103112231	103629963	2.08E-28					
RFX4	ENSG00000111783	Brain.stem, left.ventral.DC, right.ventral.DC	12	106976685	107156581	1.88E-16	1.19E-09	6.91E-10			
RHPN2	ENSG00000131941	X3rd.ventricle	19	33469499	33555794	2.23E-11					
RIC8B	ENSG00000111785	Brain.stem	12	107168373	107283090	1.43E-10					
RNF11	ENSG00000123091	left.cerebellum.exterior, right.cerebellum.exterior	1	51701943	51739127	2.78E-12	2.14E-12				
RP11-12J10.3	ENSG00000258539	left.isthmus.cingulate	10	126305649	126480296	2.82E-09					
RP11-144F15.1	ENSG00000257545	Brain.stem, left.ventral.DC, right.ventral.DC	12	106889736	107168696	7.66E-19	1.26E-10	4.00E-11			
RP11-201K10.3	ENSG00000273088	white.matter	1	155141885	155159748	6.36E-09					
SELO	ENSG00000073169	right.ventral.DC	22	50639408	50656045	1.58E-08					
SLC39A1	ENSG00000143570	left.inferior.lateral.ventricle	1	153931575	153940188	1.26E-08					
SLC44A5	ENSG00000137968	cerebellar.vermal.lobules.I.V, cerebellar.vermal.lobules.VIII.X	1	75667816	76076801	1.06E-09	3.13E-23				
SLC6A4	ENSG00000108576	left.insula, right.precuneus, right.insula	17	28521337	28563020	2.87E-10	4.16E-09	6.26E-10			
SPPL2C	ENSG00000185294	total.brain.volume	17	43922256	43924438	5.68E-11					
SQSTM1	ENSG00000161011	white.matter	5	179233388	179265078	1.91E-08					
SSH2	ENSG00000141298	left.insula, right.precuneus, right.insula	17	27952956	28257294	2.68E-12	1.06E-09	5.60E-12			
STH	ENSG00000256762	total.brain.volume	17	44076616	44077060	1.78E-12					
STK17A	ENSG00000164543	cerebellar.vermal.lobules.I.V	7	43622357	43666385	8.06E-09					
STX6	ENSG00000135823	right.pallidum, white.matter	1	180941861	180992047	1.94E-09	5.24E-13				
SZT2	ENSG00000198198	left.pallidum	1	43855553	43918321	6.89E-10					
TAF5	ENSG00000148835	total.brain.volume	10	105127724	105148822	9.27E-11					
TBC1D9B	ENSG00000197226	white.matter	5	179289066	179334859	2.65E-10					
THBS3	ENSG00000169231	grey.matter, white.matter	1	155165379	155178842	5.84E-09	1.22E-09				
THNSL1	ENSG00000185875	left.caudate	10	25305587	25315593	1.09E-08					
TIE1	ENSG00000066056	left.pallidum	1	43766664	43788779	1.57E-09					
TPX2	ENSG00000088325	left.putamen, left.pallidum, right.putamen, right.pallidum	20	30327074	30389608	6.30E-09	1.24E-11	9.84E-10	2.04E-11		
TRIOBP	ENSG00000100106	X3rd.ventricle, left.lateral.ventricle	22	38093011	38172563	6.01E-09	5.39E-09				
TTC39A	ENSG00000085831	left.cerebellum.exterior, right.cerebellum.exterior	1	51752930	51810788	2.82E-12	7.64E-13				
USMG5	ENSG00000173915	total.brain.volume	10	105148798	105156223	5.48E-10					
VCAN	ENSG00000038427	white.matter	5	82767284	82878122	9.12E-11					
WIF1	ENSG00000156076	left.hippocampus, right.hippocampus	12	65444406	65515346	8.26E-15	5.59E-16				
WNT3	ENSG00000108379	total.brain.volume	17	44839872	44910520	6.82E-11					
ZIC1	ENSG00000152977	left.pars.triangularis, right.pars.triangularis, right.postcentral	3	147111209	147228080	3.42E-09	5.55E-12	7.78E-09			
ZIC4	ENSG00000174963	left.pars.triangularis, left.postcentral, right.pars.triangularis, right.postcentral	3	147103833	147124647	1.55E-09	1.89E-08	2.26E-12	4.98E-09		
ZNF474	ENSG00000164185	left.vessel	5	121465208	121515312	5.17E-09					