A functional role for non-coding variation in schizophrenia genome-wide significant loci

Roussos .. Sklar

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EXTENDED EXPERIMENTAL PROCEDURES

GWAS data sets

A large published SCZ GWAS data set [\(Ripke et al., 2013\)](#page-12-0) (SCZ) in the form of summary statistic *P* values was obtained from public access website (https://pgc.unc.edu/). The SCZ meta-analysis included n=13,833 cases with SCZ and n=18,310 controls and identified nine LD independent regions that met genome-wide significance $(P < 5x10^{-8})$. In this dataset there were 3,538 out of 9,898,078 (9,815,700 after removing the major histocompatibility complex (MHC) locus) imputed SNPs that were genome-wide significant. A total of 22 LD independent regions were identified by replicating the top SNPs in an independent cohort. A previously published GWAS data set in rheumatoid arthritis (RA) [\(Stahl et al., 2010\)](#page-12-1) was obtained through collaboration with investigators in the form of summary statistic *P* values. The RA GWAS data included a total of 5,485 seropositive individuals with RA and 22,609 controls of European descent. In the RA GWAS there were in total 13,798 out of 8,099,406 SNPs that reached genome-wide significance. For all the analyses presented here, the MHC locus (chr6: 25-35Mb) was excluded from both GWAS datasets.

eSNP data sets

Cohorts: Brain eSNPs were generated using the gene expression and genotyping data of Caucasian samples, included in the Braincloud [\(Colantuoni et al., 2011\)](#page-11-0) (GEO accession number: GSE30272), NIA/NIH [\(Gibbs et al., 2010\)](#page-11-1) (GEO accession number: GSE15745) and Harvard Brain Tissue Resource Center (HBTRC) [\(Zhang, 2013\)](#page-12-2) (GEO accession number: GSE44772) datasets (Table S1). The following non-brain eSNP datasets were downloaded from public access websites: LCL [\(Xia et al., 2012\)](#page-12-3) [\(http://www.bios.unc.edu/research/genomic_software/seeQTL/\)](http://www.bios.unc.edu/research/genomic_software/seeQTL/), liver [\(Innocenti et al.,](#page-11-2) [2011\)](#page-11-2) [\(http://www.scandb.org/\)](http://www.scandb.org/), peripheral blood mononuclear cells (PBMC) [\(Westra et](#page-12-4) [al., 2013\)](#page-12-4) [\(http://genenetwork.nl/bloodeqtlbrowser\)](http://genenetwork.nl/bloodeqtlbrowser), skin [\(Grundberg et al., 2012\)](#page-11-3) and adipose tissue [\(Grundberg et al., 2012\)](#page-11-3) [\(http://www.muther.ac.uk/Data.html\)](http://www.muther.ac.uk/Data.html).

DNA Genotyping QC and Imputation of Brain eSNP data sets: The preprocessing of Braincloud, NIA/NIH and HBTRC SNP data was done using plink (v1.07) [\(Purcell et al.,](#page-11-4) [2007\)](#page-11-4). More specifically, individuals were removed if they were outliers with respect to estimated heterozygosity or had missing SNPs > 10%. SNPs were removed if: missing genotype rate > 5%; Hardy–Weinberg equilibrium P value <10⁻⁴; minor allele frequency <5%. Multidimensional scaling (MDS) was applied to detect population stratification and samples demonstrating non-Caucasian ancestry were eliminated. Genotyping data were imputed using the same reference panel that SCZ were imputed (1000 Genomes reference panel - phase I; March 2012). Prephasing and imputation of genotyping data was done with SHAPEIT (v2) [\(Delaneau et al., 2013\)](#page-11-5) and IMPUTE2 (v2.3.0) [\(Howie et](#page-11-6) [al., 2009\)](#page-11-6), respectively. After imputation, we retained only SNPs that have been imputed with high certainty (Info score > 0.8) and SNPs were removed as described above

Microarray preprocessing: Microarray probes that did not map to human genome or mapped to more than a locus were removed using Blat [\(Kent, 2002\)](#page-11-7) from UCSC

browser [\(http://genome.ucsc.edu\)](http://genome.ucsc.edu/). Probes containing SNPs with MAF > 0.01 according to 1000 Genomes European panel or in the current population were removed from the analysis. For preprocessing of the Braincloud and HBTRC data (two-color customspotted microarrays), we performed background correction on the linear scale, log² transformation and loess normalization. For the NIA/NIH dataset (Illumina HumanRef-8 Expression BeadChips), we applied variance-stabilizing transformation and Robust Spline Normalization [\(Schmid et al., 2010\)](#page-12-5). Probes that were not detected reliably in our gene expression data in >50% of samples were completely removed. Missing data for probes that were detected in >50% of samples were imputed to allow downstream processing. Known (age, sex, post-mortem interval (PMI), pH, RNA integrity number (RIN) and batch) and hidden confouders were removed using a Bayesian framework [\(Stegle et al., 2010\)](#page-12-6) for joint modeling of diverse sources of phenotypic variability.

eSNP analyses: Human brain regional eSNP analysis was performed in eight gene expression and genotype datasets from 3 independent studies (Table S1). eSNP analysis was conducted based on linear regression models, adjusted for the first five MDS components. We define a significant *cis* interaction as any SNP that lies within 1Mb upstream or downstream from a gene. For multiple testing corrections we applied a false discovery rate (FDR) at 10%. We found that removal of confounds from gene expression datasets using a Bayesian framework, which accounts for both known and hidden confounds, compared to lineal models, resulted on average in ~2.2-fold greater number of discovered eSNPs (Table S2).

cis **regulatory element (CRE) annotation description**

Multiple CRE annotations were used in the current study (Table S2, S3).

ENCODE and REMC data: ChIP-seq and DHS data generated as part of the ENCODE [\(Maurano et al., 2012\)](#page-11-8) and REMC [\(Zhu et al., 2013\)](#page-13-1) projects for human brain/neuron (Table S2), T - helper cells, liver, skin and adipose tissue were downloaded from the NCBI repository [\(http://www.ncbi.nlm.nih.gov/geo/roadmap/epigenomics/\)](http://www.ncbi.nlm.nih.gov/geo/roadmap/epigenomics/). Additional data for the dorsolateral prefrontal cortex after fluorescence-activated cell sorting (FACS) were generated as described previously [\(Cheung et al., 2010;](#page-11-9) [Shulha et al.,](#page-12-7) [2013;](#page-12-7) [Shulha et al., 2012\)](#page-12-8) (Table S3). The non-brain ENCODE and REMC data that were used in the current study are the following:

1. T-helper: GSM621447; GSM665812; GSM665839; GSM701489; GSM701491; GSM701539; GSM772790; GSM772835; GSM772836; GSM772849; GSM772851; GSM772852; GSM772855; GSM772859; GSM772860; GSM772862; GSM772867; GSM772868; GSM772869; GSM772876; GSM772881; GSM772884; GSM772899; GSM772902; GSM772904; GSM772905; GSM772906; GSM772911; GSM772914; GSM772915; GSM772916; GSM772919; GSM772920; GSM772924; GSM772925; GSM772928; GSM772930; GSM772934; GSM772944; GSM772946; GSM772947; GSM772948; GSM772953; GSM772954; GSM772955; GSM772963; GSM772973; GSM772985; GSM772986; GSM772987; GSM772988; GSM772997; GSM772998; GSM772999; GSM773004; GSM817166; GSM916024; GSM916026; GSM916027; GSM916059; GSM916071.

2. Liver: GSM1059451; GSM1059458; GSM1112808; GSM1112809; GSM1112814; GSM537697; GSM537698; GSM537705; GSM537706; GSM537707; GSM537709; GSM621629; GSM621630; GSM621654; GSM621675; GSM669910; GSM669972; GSM670008

3. Adipose: GSM621401; GSM621418; GSM621420; GSM621425; GSM621435; GSM621443; GSM621458; GSM669904; GSM669908; GSM669922; GSM669925; GSM669930; GSM669934; GSM669938; GSM669940; GSM669975; GSM669984; GSM669988; GSM669998; GSM670017; GSM670020; GSM670027; GSM670035; GSM670041; GSM670043; GSM670045; GSM772739; GSM772740; GSM772741; GSM772744; GSM772745; GSM772746; GSM772747; GSM772748; GSM772757; GSM772760; GSM772761; GSM772762; GSM772764; GSM772765; GSM772771; GSM772802; GSM772812; GSM772816; GSM772817; GSM772819; GSM772821; GSM772822; GSM906394; GSM906416; GSM916055; GSM916066

4. Skin: GSM1127060; GSM1127076; GSM669589; GSM669591; GSM669592; GSM774240; GSM774241; GSM817169; GSM817170; GSM817194; GSM817195; GSM817196; GSM817197; GSM817234; GSM817235; GSM817237; GSM817240; GSM817242; GSM817246; GSM817247; GSM878623; GSM878624; GSM878633; GSM878634; GSM878635; GSM878636; GSM878640; GSM878641; GSM878642; GSM878643; GSM878644; GSM878645; GSM878646; GSM878647; GSM878648; GSM878649; GSM941717; GSM941718; GSM941735; GSM941736; GSM941738; GSM941742; GSM958154; GSM958155; GSM958156; GSM958158; GSM958161; GSM958162; GSM958163; GSM958164; GSM958167; GSM958168

FACS H3K4me3 chromatin mark data: Previously published FACS H3K4me3 chromatin mark data [\(Cheung et al., 2010;](#page-11-9) [Shulha et al., 2013;](#page-12-7) [Shulha et al., 2012\)](#page-12-8) with the addition of 3 samples were used in the current study (Table S3). Freshly frozen (never fixed) tissues from N=38 samples across lifespan, was provided by four independent brain banks. Specimens were obtained from the dorsolateral portion of the prefrontal cortex, primarily from cytoarchitectonic (Brodmann) Area 10 (BA10) and regions that border on BA10, including portions of BA9 and BA46. Tissue aliquots (200–500 mg/subject) were extracted in hypotonic lysis buffer, purified by ultracentrifugation and resuspended in 1x PBS, immunotagged with anti-neuronal nucleus (anti-NeuN, Millipore) antibody and sorted into NeuN(+) and NeuN(-) fractions. Mononucleosomal

preparations from at least 1 \times 10⁶ sorted nuclei were prepared for subsequent chromatin immunoprecipitation with anti-H3K4me3 antibody (Upstate/Millipore), followed by library preparation and sequencing on an Illumina Genome Analyzer II platform, as described previously [\(Cheung et al., 2010;](#page-11-9) [Shulha et al., 2013;](#page-12-7) [Shulha et al., 2012\)](#page-12-8).

Processing of chromatin mark data: The majority of our FACS sequencing libraries has been included in previous publications [\(Cheung et al., 2010;](#page-11-9) [Shulha et al., 2013;](#page-12-7) [Shulha](#page-12-8) [et al., 2012\)](#page-12-8) (see also Table S3). All libraries contained single-end 36-bp reads and were mapped to the human genome with Bowtie2 (version 2.1.0) [\(Langmead and](#page-11-10) [Salzberg, 2012\)](#page-11-10). We allowed up to one mismatch and mapped all sequences to the gender appropriate genome hg19. Reads that mapped to multiple locations were discarded. As previously reported, H3K4me3 levels did not show correlations with postmortem interval and tissue pH [\(Cheung et al., 2010;](#page-11-9) [Shulha et al., 2013;](#page-12-7) [Shulha et](#page-12-8) [al., 2012\)](#page-12-8). We analyzed genome-matching reads with the MACS software [\(Zhang et al.,](#page-13-2) [2008\)](#page-13-2) (version 2.0.10) to identify significant peaks ($P < 1 \times 10^{-5}$) with bw = 230 bp, as defined experimentally by PCR, tSize = 36 bp, and other parameters set at default.

Identical analysis described above was applied for the ENCODE and REMC hg19 mapped sequence reads data. Because control data were available for each ChIP-seq peak data, paired ChIP-seq peak and control data were analyzed jointly using MACS to increase specificity of peak detection.

Definition of functional datasets

All SNP coordinates described here are relative to UCSC hg19. The functional datasets used in the current study were divided into 3 groups: eSNP, CRE and creSNP (eSNP in a *cis* regulatory element).

- 1. *eSNP:* The brain eSNP dataset was generated by including all significant *cis* eSNPs derived from each eSNP analyses. For the non-brain eSNPs (LCL, liver, skin, PBMC and adipose tissue), we used the list of *cis* eSNPs generated as described previously [\(Grundberg et al., 2012;](#page-11-3) [Innocenti et al., 2011;](#page-11-2) [Westra et](#page-12-4) [al., 2013;](#page-12-4) [Xia et al., 2012\)](#page-12-3) at FDR 10%.
- 2. *CRE:* The ChIP-seq and DHS significant peaks were clustered into subgroups based on assay and origin of tissue (Figure S4). For definition of high confidence CRE intervals we divided the genome into 50 bp bins and included only intervals with > 50% overlap among the peaks of each track within the samples in that subgroup. The > 50% threshold is an arbitrary cutoff that was selected based on the following criteria: 1.) the count of identified CRE intervals, 2.) the total and

average CRE interval, 3.) the proportion of the human genome covered, and 4.) the overlap with curated data generated by the ENCODE and REMC. Overall, the > 50% threshold retains high confidence CREs without reducing significantly the size and number of included intervals. Finally, we integrated multiple CRE subgroup annotations for generating functional annotations defining active promoters (overlap of H3K4me3 with H3K9ac or H3K27ac), poised promoters (overlap of H3K4me3 with H3K27me3), active enhancers (overlap of H3K4me3 with H3K9ac or H3K27ac), repressed enhancers (overlap of H3K4me1 with H3K27me3) and open chromatin (DHS).

3. *creSNP:* The creSNP functional dataset was defined as the eSNPs that lie within different CRE subcategories.

GWAS Positional Annotation

LD calculation: Variant call format (VCF) genotype files for the European reference sample provided by the November 2010 release of Phase 1 of the 1000 Genomes Project (1000G) were obtained from the NCBI repository. Additional quality control based on standard GWAS procedures was performed on 1000G data using Plink version 1.07 [\(Purcell et al., 2007\)](#page-11-4). More specifically, individuals were removed if they had missing SNPs > 10%. SNPs were removed if: missing genotype rate > 5%; Hardy– Weinberg equilibrium P value <10⁻⁶; minor allele frequency <1%. The identity by state and identity by descent analysis implemented at Plink were used for estimation of relatedness. The individual with the higher missingness from each related pair was removed. For each 1000G SNP, the \tilde{r} pairwise LD was calculated within 1,000,000 base pairs (1Mb) on either side of the SNP threshold of $r^2 \ge 0.8$.

We used a mixed approach for assigning the GWAS SNPs into functional categories. For the eSNP functional category, we leveraged the eSNP dataset in the densely mapped 1000G data to identify the GWAS studied SNP that was tagged, as a result of linkage disequilibrium (LD). For generating the CRE or creSNP functional categories we used a positional approach (ignoring the annotation categories of SNPs in LD with the tag SNP). This mixed model allows us to capture all possible SNPs that affect gene expression (tag or SNPs in LD), followed by positional selection of SNPs that lie within putative regulatory DNA regions. GWAS SNPs that did not fit into any of the above functional categories were assigned as the FUV category. The Table S7 describes the count of SCZ for each functional brain category.

Quantification of Enrichment

Stratified Q-Q plots were generated for assessment of the similarity or differences between the empirical cumulative distribution function of the functional and FUV datasets. Each empirical null distribution was corrected by applying a control method leveraging only the FUV SNPs that are less enriched for true associations. A similar approach was used by Schork *et al.* and allows correcting for global variance inflation due to effects of cryptic relatedness and population substructure minimizing the deflation due to over-correction of test statistics for polygenic traits by standard genomic control methods (Schork [et al., 2013\)](#page-12-9). The inflation factor, λ_{GC} was computed as the median z-score squared divided by the expected median of a chi-square distribution with one degree of freedom. All test statistics described here were divided by λ_{GC} . In the stratified Q-Q plots, the enrichment of SCZ SNPs for a specific functional category is observed as a horizontal deflection from the FUV category.

The quantification of enrichment for each functional category was done using the categorical enrichment score (CES), as described previously [\(Schork et al., 2013\)](#page-12-9). The CES provides a summary score of category-specific enrichment where the mean is taken over all SNP z-scores in the given category. It is estimated based on as the mean $(z^2 - 1)$ and it is justified as a measure of enrichment based on a simple Bayesian mixture model framework. The CES is a conservative estimate of the variance attributable to non-null SNPs, given a standard normal null distribution and a non-null distribution symmetric around zero.

The statistical significance of CES was evaluated by permutation on a combined 10,000 randomized set of CREs and eSNPs. For CREs, we generated 100 random sets matched for distance from TSS, size and number of intervals. We generated 100 random sets of eSNPs matched for MAF, distance from TSS and local LD (based on sum r^2) with the most significant eSNP from each LD interval (using the clump function implemented in PLINK with r2 < 0.1 within 100-kb window). For each functional category, we estimated the null distribution of CES by combining the 100 random sets of CREs and eSNPs (10,000 total permutations).

Control for potential confounds: Linkage disequilibrium and minor allele frequency

We examined whether differences in the LD (estimated based on the sum r^2) and MAF among functional categories (Table S6), could lead to spurious enrichment. For each SNP, the SCZ GWAS summary value (log of z^2 after FUV inflation control) was included as the dependent variable and the genetic variance (estimated as MAF*(1-MAF)), total LD (sum \hat{r}) and membership for functional annotation categories were included as the independent variables in multiple regression analysis (Table S7). The heteroscedasticity and correlated responses from cluster samples was corrected using the Huber-White method.

Differences in the LD (estimated based on the sum r^2) were observed among the functional categories (Table S6). We examined for enrichment of each functional category compared to the FUV SNPs after retaining for each annotation category the most significant, LD-independent SCZ SNP using the clumping function implemented at Plink (with parameters of $r^2 > 0.1$ and distance 100 kb). The significance values for the curves for each functional annotation were calculated using a two-sample Kolmogorov-Smirnov Test (Table S6).

Recombination hotspot interval calculation

We divided the genome into recombination hotspot intervals based on inferred recombination rates and genetic map positions for 1000 Genomes data (March 2012 allpopulations map data distributed with IMPUTE2 software[\(Howie et al., 2009\)](#page-11-6)). We used a sliding window analysis to identify hotspots defined as regions with greater than 20 cM/Mb average rate and cumulative distance ('width' in cM) greater than 0.01 cM, then conservatively filtered out hotspots less than 100 Kb or 0.1 cM from a longer geneticwidth hotspot, and defined intervals as regions between consecutive hotspot midpoints. This resulted in 4,258 hotspots with mean (standard deviation) rate: 33 cM/Mb (11.6) and widths: 4.1 Kb (6.4) and 0.1 cM (0.098), and 4,280 hotspot intervals.

Regulatory Trait Concordance (RTC) analysis

The likelihood of a shared functional effect between a SCZ genome-wide significant SNP and an eSNP was assessed by the RTC approach [\(Grundberg et al., 2012;](#page-11-3) [Nica et](#page-11-11) [al., 2010\)](#page-11-11). Each SCZ index SNP in Ripke et al. was mapped to recombination hotspot intervals. For each recombination interval we identified all genes with a significant eSNP. We then repeat the eSNP analysis based on linear regression models after conditioning on the SCZ index SNP and estimate the change in the eSNP statistical significance. If the GWAS index SNP and the eSNP tag the same functional variant, then by removing the genetic effect of the GWAS SNP, the eSNP statistical association will be reduced or completely lost. To take in account the LD structure within each recombination hotspot interval, we ranked the impact of SCZ index SNP on eSNP significance (*RankGWAS*) by correcting for all other SNPs within the same interval (*NSNPs*). The rank estimates the number of SNPs which when used to correct the expression data, have a higher impact on the eSNP statistical significance (less significant adjusted *P*) than the GWAS SNP. The RTC is estimated based on the following formula:

$$
RTC = \frac{N_{SNPs} - Rank_{GWAS}}{N_{SNPs}}
$$

If the GWAS SNP is the same with the eSNP then the eSNP adjusted *P* will be 1, the *Rank_{GWAS}* = 0 and RTC = 1. RTC score ranges from 0 to 1, with values \geq 0.9 indicating likely causal regulatory effects, as demonstrated previously [\(Grundberg et al., 2012;](#page-11-3) [Nica et al., 2010\)](#page-11-11).

Chromosome Conformation Capture (3C) analysis

Postmortem prefrontal cortex brain tissue for cases with SCZ and controls was obtained from the University of Maryland and pair-matched for age, sex, PMI, and pH (*N*=3/group) (Table S8). PFC tissue (200mg) was dissected, homogenized, and crosslinked for 10 minutes at 25°C in 1% formaldehyde, 1X protease inhibitor (Sigma), and 2 mL of lysis buffer (10 mmol/L Tris hydrogen chloride pH 8.0 / 10 mmol/L sodium chloride / 0.2% IPEGAL CA-630 (Sigma Aldrich, St. Louis, Missouri)). Crosslinking was stopped by the addition of glycine to a final concentration of 0.125 mol/L for 10 minutes at 4°C and the homogenate was incubated at 4°C for an additional 25 minutes. Cell were lysed by pipetting >50 times and spun at 5000 rpm, the supernatant was removed, and the pellet was washed twice with 1X New England Buffer 4 (NEB4) (New England Biolabs, Boston, Massachusetts). Samples were resuspended in 200 µl of 1X NEB4 and divided into four 50 µl aliquots. An additional 312 µl of 1X NEB4 was added to each aliquot. Samples were incubated at 65°C for 10 minutes in 1X NEB4 buffer and 38 µl of 1% SDS for more efficient digestion and to remove protein not associated with DNA. To quench SDS 10% of Triton X-100 was added to each sample and the samples were digested with *HindIII*-HF or *NcoI*-HF (NEB) at 37°C overnight with gentle shaking.

HindIII-HF or *NcoI*-HF was inactivated by the addition 86 µl of 10% SDS incubated for 30 minutes at 65°C. Ligation mixture (7.61 ml) was added to each sample. The ligation mixture consisted of 745 µl of 10% Triton X-100, 745 µl of 10X ligation buffer (1 M Tris HCL, pH 7.5, 1 M $MgCl₂$, 1 M DTT dithiothreitol (Bio-Rad)), 80 µl of 10 mg/ml bovine serum albumin (NEB), 80 µl of 100 mM ATP (Sigma) and 5960 µl of autoclaved water. 50 µl of T4 DNA ligase (1 U/µl, Invitrogen) was added to three aliquots and one sample was used as a no ligase control. Ligation proceeded for five hours at 16°C and samples were reverse cross-linked at 65°C overnight with 50 uL of 10 mg/ml of proteinase K (Sigma). For improved ligated DNA recovery another 50 µl of proteinase K was added and incubated at 65°C for two hours. DNA was extracted with phenol (pH 8.0, Fisher), and phenol-chloroform (1:1) (pH 8, Fisher). DNA was precipitated using 1/10 the volume of 3M sodium acetate (pH. 5.4) and 2.5 the volume of ice-cold ethanol overnight. The samples were centrifuged at 8000 RPM for 30 minutes and washed with 70% ethanol. The final DNA pellet was dissolved in 1X TE buffer (pH 8.0). Phenol and phenolchloroform extraction and ethanol precipitation was repeated. The final 3C library was washed five times with 70% ethanol. Ligase and no ligase reactions were dissolved in 100 µl and 33 µl of TE buffer (pH 8.0) respectively [\(Mitchell et al., 2013\)](#page-11-12). 3C libraries with and without the critical ligase step were run on a 2% agarose gel to visualize ligation efficiency. Samples ran at a higher molecular weight after ligation, indicated by an upward shift on the gels (Figure S5A) [\(Dekker, Rippe et al. 2002,](#page-11-9) [Dekker 2006\)](#page-11-13).

Physical looping interactions were quantified with PCR. Primers were designed less than 120 bp from the *HindIII* or *NcoI* restriction site (Table S9). The PCR products were resolved on a 2% agarose gel and the level of interaction between two primers was measured semiquantitatively using band intensities normalized with the background (raw 3C interaction) with ImageJ [\(Schneider et al., 2012\)](#page-12-10). Library input was adjusted for each library according the interaction between two neighboring primers (<5000 bp apart) and two distant primers (<30,000 apart) control primer 1 (CCTGGATCATCAGACAGAACTAAAGCTCTT) located at chr13:99113854 and control primer 2 (CTTCAACTGAAAACACACGAACAGGAAGAA) located at chr13:99109553 (Figure S5B). A bacterial artificial control (BAC) 3C library was created for the *CACNA1C* region to normalize for primer efficiency starting with equimolar concentrations of RP11-465I2 and RP11-698B24 (Figure S5C) [\(Dekker, Rippe et al.](#page-11-9) [2002,](#page-11-9) [Dekker 2006\)](#page-11-13). For each library (*HindIII* and *NcoI*) in the human postmortem brain tissue studies, we transformed the raw 3C interaction to *Z* scores, followed by scaling (0 to 1) (Scaled 3C interaction). All 3C PCR products were sequence verified and the interactions were not present in the no ligase and water controls.

For primers #2, #4 and #5, physical looping interactions were further quantified with qPCR using an ABI Prism 7900 (Applied Biosystems). The thermal cycling program consists of 5 min at 95°C, followed by 40 cycles of 30 sec at 95°C, 30 sec at 60°C and 30 sec at 72°C. Only one DNA was amplified in each PCR (monoplex). The reactions were run in triplicate for each sample and DNA PCR product was measured through SYBR Green I (Life Technologies). For more accurate quantification of amplified DNA, we used the Relative Standard Curve Method¹²³. To account for differences in the amount of input material between the samples, the SYBR Green I signal from each 3C interaction region was normalized to the geometric mean of the SYBR Green I signal of two endogenous references [large ribosomal protein (*RPLP0*) and Glucuronidase, Beta (*GUSB*)].

Human induced pluripotent stem cells differentiation into neurons

Human induced pluripotent stem cells (hiPSCs) were derived from fibroblasts of a control sample (GM03651) as described previously [\(Brennand et al., 2011\)](#page-11-13). For neural differentiations, neural progenitor cells (NPCs) were dissociated with Accutase and plated in neural differentiation media (DMEM/F12, 1x N2, 1X B27-RA, 20 ng/ml BDNF (Peprotech), 20 ng/ml GDNF (Peprotech), 1 mM dibutyrl-cyclicAMP (Sigma), 200 nM ascorbic acid (Sigma) onto PORN/Laminin-coated plates. Density is critical: for 6-well plates, 200,000 NPCs were plated per well. hiPSC derived-neurons were differentiated for ~6 weeks in neural differentiation media. The majority of forebrain hiPSC neurons are VGLUT1-positive, and so are presumably glutamatergic, although approximately 30% of neurons are GAD67-positive (GABAergic) [\(Brennand et al., 2011\)](#page-11-13). Gene expression comparisons of 6-week-old neurons to the Allen BrainSpan Atlas indicate that our hiPSC neural cells resemble fetal brain tissue {Brennand, 2014 #2341}.

Transient Transfection and Luciferase Assays

We constructed luciferase reporter plasmids by cloning the regulatory sequence containing rs2159100 into the pGL4.24 vector (Promega) upstream of the minP. The regulatory sequence was introduced at the 5' and 3' by using *KpnI* and *XhoI*. We sequenced the inserted portions of the constructs to verify the nucleic acid sequences and the location of the SNP. Human embryonic kidney 293 (HEK-293) cells or Neuro-2a cells (40-60% confluent) were transfected with each reporter vector (450ng) or construct (500 ng to ensure the same copy number as the empty vector) and the Renilla luciferase expression vector pRL-TK (200 ng, Promega) using Lipofectamine 2000 (3:1, μL Lipofectamine: μgDNA) (Invitrogen) in 200 μL Opti-MEM (Invitrogen) in 12 well

plates. HEK 293 cells or Neuro-2a cells were grown in 1 ml DMEM supplemented with 10% FBS, and the media was not changed after the addition of the transfection reagents. Twenty-four hours after transfection, cells were lysed by the addition of 250 μL of Passive Lysis Buffer (Promega). The luciferase activity in the cell lysates was determined using the Dual Luciferase Reporter System (Promega) in quadruplicates. All experiments for HEK-293 and Neuro-2a were performed in duplicates. Firefly luciferase activities were normalized to that of Renilla luciferase and expression relative to the activity of the rs2159100 C allele was noted.

The regulatory sequence containing the $rs2159100$ C and T alleles \pm 250 bp upstream/downstream (hg19; chr12:2,346,143-2,346,643) is provided below. Six nucleotides (in bold/underlined fonts) were added at the 5' and 3' ends as restriction sites for the *KpnI* and *XhoI,* respectively.

GGTACCACAATGCCTTGTGATACTCTTGTTCTTCTGGTTTGAGTTTTGGTAGATAAG CACATCTGAGTCTTGCTGTGTTAATGTGTCTGTATTTTGGTGTATCTGCTTGCTTGT CGTGTGGGGCATATGCCAAGTCCAGTAGTGGATGGGCTGGGGAAGACCAGACCT TATCACATGGTGCCCTTGGGGGGAAATCTTAATTCCAATGTGTGAAACCAGTGAAA GTATGATTTTCTGGGTCAATTTTAAAAATATA**[C/T]**GTTCAAGCAAAAAGCAACCTGT TATCTCTTCTCTTTCTGCCTCTGCACACAGCAGCCTCCATTGCCTAGGGTATGATA GTGTGGGTTCACTTTGTCCATCTCATTTGGATGACATCAGCGAAGATGCATTCTGT ATCTCTCCACTGAGGCCTGTGACAGGACCTAATGTTTTGTGGAGCTGAGAGAAAAT AAACCAAAATGACCCAATGGAATATAAATGCTGATTTCTGTTCCTGTTGTTTGACAG AAA**CTCGAG**

SUPPLEMENTAL REFERENCES

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Supplementary Tables

Table S1, related to "Methods: eSNP data sets". Characteristics of the brain gene

expression and genotype cohorts

Table S2, related to "Methods: *cis* **regulatory element annotations".** Brain related epigenomic data generated by the ENCODE and REMC projects that were used in the current study

Table S3, related to "Methods: *cis* **regulatory element annotations".** Brain fluorescence-activated cell sorting samples used in this study and sequencing statistics

Brain Banks: HBTRC – Harvard Brain Tissue Resource Center (Dr. Francine Benes); UM-BTB – University of Maryland Brain and Tissue Bank for Developmental Disorders (Dr. Ron Zielke); MPRC – Maryland Psychiatric Research Center (Dr. Rosalinda Roberts and Dr. Andree Lessard); UCI/UCD – University of Irvine/Davis (Dr. Ted Jones and Dr. William E. Bunney Jr.). PLoS = Shulha et al., PLoS Genet 9(4):e1003433 (2013); AGP = Shulha et al., Archives of General Psychiatry 69:314-324 (2012); PNAS = Cheung et al., Proceedings National Acad Sci USA 107:8824-8829 (2010). yr = years; gw = gestational week.

Table S4, related to Figure 1 and Table 1. Counts of SNPs per annotation category

Table S5, related to Table 1. SNP enrichment for different GWAS P values

* SNP enrichment for different GWAS P values illustrates the estimated change in terms of the proportion of SNPs in different functional categories that reach different P values (P < 10^{-3} ; P < 10^{-5} ; P < 5 x 10⁻⁸) in comparison to functionally unannotated variants NA indicates that no SNPs exceed the specified GWAS P value

Table S6, related to Figure 1 and 2 and Table 1. Per functional category average per SNP total tagged LD, MAF and P values of the enrichment for all functional categories compared to FUV estimated with the Kolmogorov-Smirnov statistic after removing SNPs with r^2 > 0.1 and those that were <100 kb from a more strongly associated variant in the SCZ study.

Variables	coefficient	SE	95% CI	P value
Intercept	-1.4218	0.0012	$(-1.424,-1.419)$	$< 2 \times 10^{-16}$
LD (sum r^2)	0.0006	5.72E-06	(0.0006, 0.00063)	$< 2 \times 10^{-16}$
Genetic variance (MAF*(1-MAF))	1.1652	0.0086	(1.1485, 1.182)	$< 2 \times 10^{-16}$
eSNP	0.1260	0.0021	(0.1219, 0.1301)	$< 2 \times 10^{-16}$
DHS	0.0112	0.0046	(0.0021, 0.0203)	0.016
Active promoter	0.0266	0.0067	(0.0135, 0.0398)	7.2×10^{-5}
Active enhancer	0.0395	0.0045	(0.0306, 0.0483)	$< 2 \times 10^{-16}$
Poised promoter	-0.0241	0.0100	$(-0.0437,-0.0046)$	0.016
Repressed enhancer	-0.0128	0.0068	$(-0.0262, 0.0005)$	0.060

Table S7, related to Figure 1 and 2 and Table 1. Multiple regression analysis in the SCZ GWAS

Subject	Gender	Age	PMI (hr)	pH	Condition	Match
35	Male	52	12	6.8	Control	1
28	Male	31	15	6.6	Control	17
11	Male	46	19	6.7	Control	40
1	Male	53	11	6.8	Schizophrenia	35
17	Male	31	14	6.5	Schizophrenia	28
40	Male	44	14	6.6	Schizophrenia	11

Table S8, related to Figure 5. Demographics of postmortem cohort used for the 3C experiments

Table S9, related to Figure 5. Primers used for the *CACNA1C* 3C experiments

Supplementary Figures

Figure S1, related to Figure 1 and Table 1. Stratified Q-Q plots show enrichment for eSNP and CRE or eSNP and creSNP categories in the adult brain-homogenate **(a, b)**, fetal brain-homogenate **(c, d)**, brain-FACS **(e, f)** and Primary Cell Culture/iPS **(g, h)** tissue. All empirical null distributions were corrected for inflation by using the FUV inflation control. The major histocompatibility complex locus (chr6: 25-35Mb) was excluded from the SCZ dataset.

Figure S2, related to Figure 2. Categorical enrichment for the individual functional annotation as measured by the CES. The CES are scaled using the maximum value across functional categories (H3K4me1 in fetal brain tissue). For each functional category, we performed 10,000 permutations to calculate the null distribution of categorical enrichment for comparison to observed categorical enrichment. Functional categories with empirical P values $\leq 10^{-4}$ are indicated with asterisk (*). All summary statistics were corrected for inflation by using the functionally unannotated variant (FUV) inflation control. The major histocompatibility complex locus (chr6: 25-35Mb) was excluded from the SCZ dataset.

Figure S3, related to Figure 1 and Table 1. Stratified Q-Q plots show enrichment for combined functional annotation categories in SCZ. SNPs with $r^2 > 0.1$ and those that were <100 kb from a more strongly associated variant in the SCZ study were removed for each functional category. Higher enrichment was observed across the eSNP and **(a)** combined CRE or **(b)** creSNP functional annotation categories. We computed significance values for the curves for each functional annotation category relative to those for FUV SNPs, using a two-sample Kolmogorov-Smirnov Test. The enrichment was significant for all functional categories when compared with the FUV category. The major histocompatibility complex locus (chr6: 25-35Mb) was excluded from the SCZ dataset.

Figure S4, related to "Methods: *cis* **regulatory element annotations".** Crosscorrelation map generated by clustering histone modification and DHS peaks based on pair-wise correlations. Heat indicates degree of positive (blue) or negative (yellow) correlation between data sets. **(a)** Brain related epigenomic data (n=140) that were used in the current study. Tissue: adult brain -homogenate (red); fetal brain - homogenate (black); brain - FACS: NeuN(-) (green); brain - FACS: NeuN(+) (cyan); primary cell culture/iPS (blue). **(b)** T-helper cell related epigenomic data (n=50). **(c)** Liver related epigenomic data (n=14). **(d)** Skin related epigenomic data (n=46). **(e)** Adipose tissue related epigenomic data (n=41) that were used in the current study. Tissue: adipose nuclei (black); adipose derived mesenchymal stem cells (red).

Figure S5, related to Figure 5. Chromosome conformation capture library controls. A) 3C libraries were made with (+ligase) and without (-ligase) ligase. Two independent aliquots of ligase libraries were created. B) 3C libraries were normalized using a neighboring primer (left) and quantified (right). C) A BAC control was generated for the *CACNA1C* region using equimolar concentrations of RP11-465I2 and RP11-698B24 (left) to normalize for primer efficiency.

