Massively parallel assessment of gene regulatory activity at human cortical structure associated variants

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Abstract

Genetic association studies have identified hundreds of largely non-coding loci associated with inter-individual differences in the structure of the human cortex, though the specific genetic variants that impact regulatory activity are unknown. We implemented a Massively Parallel Reporter Assay (MPRA) to measure the regulatory activity of 9,092 cortical structure associated DNA variants in human neural progenitor cells during Wnt stimulation and at baseline. We identified 918 variants with regulatory potential from 150 cortical structure associated loci (76% of loci studied), of which >50% showed allelic effects. Wnt stimulation modified regulatory activity at a subset of loci that functioned as condition-dependent enhancers. Regulatory activity in MPRA was largely induced by Alu elements that were hypothesized to contribute to cortical expansion. The regionally specific impact of genetic variants that disrupt motifs is likely mediated through the levels of transcription factor expression during development, further clarifying the molecular mechanisms altering cortical structure.

Main

Introduction

The expansion of the cerebral cortex in humans, relative to non-human primates, is thought to contribute to our distinctive cognitive and social abilities[1,2]. Genetic association studies conducted in large consortia, including the Enhancing Neuroimaging Genetics through Meta-analysis (ENIGMA), have identified single nucleotide polymorphisms (SNPs) associated with inter-individual variability in human cortical surface area and thickness measured with magnetic resonance imaging, providing a foothold into the molecular mechanisms affecting human cortical size and shape[3]. However, most cortical structure associated variants are found in non-coding regions of the genome without a clearly defined function.

Cortical structure associations are enriched in regulatory regions of the genome active in primary human neural progenitor cells (phNPCs) and colocalize with genetic variants influencing chromatin accessibility - a proxy for transcription factor (TF) binding - and gene expression specifically in NPCs (chromatin accessibility and expression quantitative trait loci; ca/eQTLs)[3–6]. This suggests cortical structure associated variants alter gene regulatory elements by modifying TF binding and gene expression, impacting fate decisions of phNPCs during prenatal

development, ultimately leading to differences in cortical size in adulthood. TFs exhibit regionally
specific expression patterns[7], and translocate to the nucleus in response to activation of
developmental signaling pathways such as Wnt[8], which may lead to variants having stronger or
weaker effects within specific regions or in stimulated conditions. This hypothesis is supported by
cortical structure associations enriched within Wnt responsive regulatory elements and
associations found near genes involved in the canonical Wnt pathway[3,6].

Gene regulatory elements are marked by accessible chromatin or histone post-translational
modifications that allow TF binding, and through looping with promoters, can influence gene
expression[9–11]. In addition to chromatin loops, enhancer RNAs (eRNAs), transcribed in non-
coding genomic regions, initiate looping through complementarity to upstream antisense promoter
RNAs (uaRNAs) recruiting polymerase to increase transcription[12,13]. The latter mechanism
involves co-opting the expression of Alu elements, a set of transposable elements that are
prevalent across the human genome[14]. Alu elements have been theorized to strongly contribute
to human cortical evolution because they have increased in prevalence together with brain size
along the human lineage[15].

Due to the correlation of nearby genetic variants (linkage disequilibrium or LD), the specific
variant(s) with gene regulatory activity in an associated locus are unknown. We implemented a
Massively Parallel Reporter Assay (MPRA), a high-throughput experiment designed to measure
the regulatory activity of short DNA sequences using barcoded expression constructs to identify
cortical structure associated variants with gene regulatory activity[16,17]. MPRA can measure the
regulatory function of a given element (MPRA activity) as well as the influence of an allele within
the regulatory element (allelic effect or expression modulating variant [emVar]). Here, we tested
the regulatory impact of 9,092 cortical structure associated variants within phNPCs. Considering
the role of the Wnt signaling pathway in brain size, we tested whether regulatory potential of
variants changes in response to Wnt stimulation. We identified 1,258 elements (7% of elements
tested) with regulatory potential from 150 cortical structure associated loci (76% of loci studied).
Among these, >50% of cortical structure associated variants showed allelic effects. Regulatory
activity in MPRA was largely driven by Alu elements, a small subset of which showed condition-
dependent gene regulatory function. Disruption of motifs by emVars highlighted TFs likely
involved in human brain development and suggested that the impact of genetic variants on
different brain regions is mediated by levels of TF expression in those regions. Overall,

characterization of MPRA-validated cortical structure associated variants further clarify molecular
mechanisms altering cortical structure.

Results

MPRA design and reproducibility

Fig.1 Experimental design and reproducibility

a, SNP selection. 502 caQTLs were chosen in chromatin accessible peaks in phNPCs (left). ENIGMA
GWAS data contains SNPs associated with surface area or thickness (global and 34 regions of the cortex)
(right). Cortical regions are color-coded according to the number of loci included in this study, stratified by
surface area and thickness (top) and the number of elements per locus (bottom). **b,** Experimental design.
c, Replicate correlations of MPRA activity were uniformly strong (median = 0.99). **d**, An example

correlation of MPRA activity across two replicates. Each set of MPRA elements is colored along
with a line of best fit. **e**, Histograms of MPRA activity, quantified as transcription rate, for different
sets of MPRA elements, where positive controls showed higher MPRA activity than negative
controls (see also **Supplementary Fig. 4**). For visualization purposes, the distribution of negative
controls are shown upside down. **f**, MPRA activity of caQTL elements above the negative controls
(MPRA active; FDR < 0.1) was detected in either allele of 37 regions from 502 assessed. Allelic
expression imbalances (allelic effects) were observed more frequently in MPRA active elements.
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We designed four sets of 150 base pair elements to understand the gene regulatory and allelic

effects governing human brain development and structure (**Fig. 1a,b**): (1) *positive controls*: 10

constitutively active promoter regions commonly used to drive transgene expression (min-tk

promoter, Ef1-alpha core promoter, CMV promoter) (2) *caQTLs*: 502 chromatin accessibility

peaks that contain bi-allelic SNPs replicably associated with their accessibility in a population of

phNPCs[4,6] (**Supplementary Fig. 1**), (3) *ENIGMA GWAS*: 9,092 regions centered on any SNP

associated ($P < 1 \times 10^{-5}$) with inter-individual differences in human cortical structure from 198

genome-wide significant loci[3] and (4) *negative controls*: 194 scrambled elements from the

ENIGMA GWAS. In total, we tested the regulatory activity of 19,500 elements.

Each element was synthesized and cloned into a plasmid upstream of a minimal promoter to drive

expression of luciferase followed by a random 20 bp barcode[18]. We transfected the pooled MPRA

library into 16 biological replicates of phNPCs. After 24 hours to allow expression of the MPRA

constructs, we activated Wnt signaling using CHIR (CHIR99021, also known as CT99021), a

potent GSK3β inhibitor and Wnt activator, or studied baseline gene regulation using a vehicle

(DMSO), for 48 hours[3,8,19]. For each biological replicate, we measured the expression of barcodes

using RNA-seq and the number of barcodes present in that replicate using DNA-seq. We

calculated the MPRA activity of an element as the summed RNA counts divided by the DNA

counts.

After removing outliers in expression level and filtering poorly represented elements (**Methods**,

Supplementary Fig. 2), we retained 28 (s.d. = 12.5) barcodes on average for each element

across 14 replicates (**Supplementary Fig. 3**). Regulatory activity was strongly correlated between

biological replicates, demonstrating the reproducibility of our assay (median Pearson's correlation

= 0.99; **Fig 1c,d**). At least one element from each positive control region showed significantly

higher regulatory activity than negative controls (Benjamini-Hochberg (BH)-adjusted *P*-value

[FDR] < 0.1; **Fig. 1e, Supplementary Fig. 4**), as expected.

134 We next evaluated the concordance between caQTLs from our previous work[4,6] and MPRA

135 conducted in the same cell type - phNPCs (**Supplementary Tables 1,2)**. Though there was

136 limited concordance between MPRA activity or allelic effects in caQTLs and MPRA, as found in

137 previous work[20–24] (**Supplementary Fig. 5**), the strongest predictor of detecting MPRA allelic

138 effects from caQTLs was MPRA activity (Odds Ratio [O.R] = 8.22; two-sided Fisher's Exact test

139 P = 3.2 x 10^{-6}, **Fig. 1f, Supplementary Table 2**). Our subsequent downstream analyses only

140 evaluated allelic effects when at least one allele showed MPRA activity significantly above

141 negative control samples.

Identifying regulatory elements associated with inter-individual differences in human cortical structure

Fig. 2 MPRA regulatory activity and allelic effects at cortical structure associated loci

a, Summary of MPRA elements at cortical structure associated loci with MPRA activity and allelic effects. Sankey diagram illustrates the composition of elements across tested loci (top), colored by MPRA activity (middle) or emVars (bottom). emVars were tested only for variants where at least one allele showed MPRA activity. **b,** MPRA active cortical structure associated loci ordered by MPRA activity. *Top row:* the level of MPRA activity of the allele with highest activity within a locus. *Middle row:* a barplot showing the number of cortical regions (among 34) that were associated to a locus containing an MPRA active element colored by GWAS significance level, followed by GWAS *P-value* of global surface area and thickness. *Bottom row:* the number of emVars at the locus. **c**, MPRA transcription level of the higher activity allele is associated with the absolute value of allelic difference (logFC).

To identify cortical structure associated regulatory elements, we empirically tested the regulatory activity of genetic variants associated with inter-individual variability in human cortical structure from the ENIGMA GWAS study[3] (**Fig. 1a, Supplementary Table 3a**). Cortical regions varied in

the number of detected GWAS significant loci (e.g., 54 loci associated with pericalcarine surface

area), and the number of associated SNPs within a locus due to LD patterns (e.g., 428 variants

were tested at the large chromosome 17q21 inversion region[25]), influencing the number of

elements tested in our MPRA (**Fig. 1a, Supplementary Table 3a**). The vast majority (88.4%) of

variants tested in the MPRA were located in introns or intergenic regions, with presumed gene

regulatory function. Twenty percent of variants were associated with both cortical surface area

and thickness, suggesting these variants have pleiotropic effects (**Supplementary Fig. 6a,**

Supplementary Table 3a).

Among 17,837 elements (9,052 variants) in 198 loci, we observed 988 elements with MPRA

activity significantly higher than the negative controls in the vehicle condition (FDR < 0.1, **Fig.**

2a,b, **Supplementary Fig. 6**, **Supplementary Tables 1,3a**). While only 6% of tested elements

showed MPRA activity, 70% of loci contained at least one element with detectable regulatory

activity. This suggests that most cortical structure associated loci have at least one SNP with

detectable regulatory activity in phNPCs using MPRA.

More MPRA active elements were discovered at surface area associated loci than at thickness

associated loci, likely due to the larger number of elements tested for surface area

(**Supplementary Fig. 6**). MPRA active elements generally had regionally limited effects, being

significantly associated with only one or two cortical regions (GWAS $P < 1\times10^{-5}$) rather than having

broad impacts across the entire cortex (**Fig. 2b, Supplementary Fig. 6, Supplementary Table**

3a).

We next evaluated allelic effects at cortical structure associated loci that showed MPRA activity

when both alleles were present. We found 386 variants with significant allelic effects (79% of loci,

58% of variants tested; FDR < 0.1; **Fig. 2b, Supplementary Tables 1,3b**), indicating that most

cortical structure associated loci have at least one genetic variant capable of altering transcription

in progenitor cells during brain development. Consistent with the caQTL dataset, emVars

exhibited higher expression levels than non-emVars, further supporting the relationship between

MPRA expression strength and allelic effects (**Fig. 2c**).

Cortical structure associated MPRA active elements are enriched in promoter regions of Alus

Fig. 3 MPRA activity of cortical structure associated loci is driven by Pol III promoter elements within Alus

a, Enrichment of MPRA active elements across classes of transposable elements, **b**, SINE-subfamily elements, and **c,** Alu element orientation, showing that forward Alu elements are major drivers of MPRA activity. **d,** MPRA active elements binned by level of activity were colored based on their overlap to Alu elements in different orientations (FWD/REV) or containing A and B box Pol III promoter elements. **e,** Schematic structure of a typical Alu-element which contains a left arm serving as Pol III promoter containing A and B boxes, followed by a right arm terminating with a polyA repeat. **f,** Frequency of MPRA elements containing sequence corresponding to the positions of observed consensus elements in the Alu element, including A and B boxes (top). Density of MPRA elements stratified by MPRA activity group across position (middle). Activity level of MPRA active elements graphed according to their position within the Alu element, stratified and colored using base mismatches in parts per thousand (milliDiv) (bottom). **g,** Example MPRA elements demonstrating that higher MPRA activity (color) is observed when the elements contain both A

and B boxes. **h,** Elements where A/B matched sequences occur before the Pol III termination sequence
(TTTT) are more often inactive. **i,** Significance of MPRA allelic effects mapped within the Alu element show
that genetic variation near A/B boxes have strong effects on MPRA activity. **j,** Overrepresentation of MPRA
active elements within enhancer RNAs or upstream antisense RNAs assessed in neural progenitor cells
using RIC-seq[12]. **k,** Younger elements with lower milliDiv show higher activity than older elements,
specifically when found in the left arm of the Alu element. Statistical significance of enrichment was
assessed by two-sided Fisher's exact test. *P*-values were corrected for multiple comparisons using the
Benjamini–Hochberg method. FWD: Forward, REV: Reverse, Veh: vehicle, O.R: odds ratio, EPRI:
enhancer-promoter RNA interaction.
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We next aimed to identify genetic features associated with MPRA active elements linked to

cortical structure. We found that MPRA active elements were enriched in short interspersed

nuclear elements (SINEs; 56.7% in active, 11.6% in inactive elements), consistent with previous

MPRA experiments in other cell types[24,26] (**Fig. 3a**). SINEs are a type of the retrotransposons

comprising a set of DNA elements that express themselves and re-integrate into new positions,

resulting in a large number of repeated copies found throughout the genome. Retrotransposons

have been hypothesized to contribute to the rapid evolution of both gene regulation and human

brain structure[15,27]. Among subfamilies of SINEs[28], MPRA active elements were specifically

enriched in forward oriented Alu elements[29] (**Fig. 3b-d**), which aligned with our oligo design that

used only forward sequences.

A typical Alu element contains two consensus sequences in its left arm that function as RNA

polymerase III (Pol III) promoters, the A-Box (TGGCTCACGCC) and the B-Box

(GTTCGAGAC)[30,31] (**Fig. 3e**). We estimated the relative position of MPRA elements within the Alu

sequence, and observed that the most active MPRA elements were located in the left arm of the

Alu element and contained the A or B boxes (**Fig. 3d-g**). Presence of both A and B boxes resulted

in the highest MPRA activity (**Fig. 3d-g**). One of the scrambled sequences (negative control)

unintentionally contained a B-box like sequence (GTTCGAGC) and had strong MPRA activity,

further supporting that these Alu based sequences drive MPRA activity (**Supplementary Fig. 7**).

We also found reduced MPRA activity when elements containing A or B boxes were followed by

a TTTT (T4), the termination signal for Pol III (**Fig. 3h**). Subsequently, we explored allelic effects

within MPRA active elements that contained Alu sequences. We found that emVars were more

often found in Alu elements compared to MPRA active variants without allelic effects (O.R = 1.56,

P = 0.007). Notably, the variants with the most significant allelic effects on transcription were

located within the left arm of the Alu element and overlap or were close to A or B boxes, further

supporting the critical regulatory role of these regions (**Fig. 3i**).

Strong Alu-element driven transcription may be an artifact of non-chromatinized sequences in the episomal MPRA library, that are normally silenced in the endogenous genome. We found that MPRA active elements were enriched with eRNAs and uaRNAs in phNPCs[12] compared to MPRA inactive elements (**Fig. 3j**), suggesting that at least a subset of MPRA active elements exhibit endogenous genome activity.

The age of Alu elements can be estimated based on their accumulated mutations over time, with younger Alu elements having lower sequence divergence to ancestral sequences than older Alu elements. The level of sequence divergence can be measured using mismatches per thousand bases to consensus sequences (milliDiv)[32]. We found that younger cortical structure associated Alu elements (<100 milliDiv) showed higher activity than older elements (>150 milliDiv) (**Fig. 3f,k**), consistent with previous findings[29]. These results suggest that more recent insertion of Alu elements created novel gene regulatory regions impacting neurodevelopmental genes that influenced cortical structure along the human lineage.

Overall, these results demonstrate that MPRA activity of cortical structure associated sequences in neural progenitors are strongly driven by Alu elements, especially younger Alu, that enhance Pol III based transcription.

Transcription factors contributing to regulatory function at cortical structure associated elements

Fig. 4 TFs enriched in cortical structure associated regulatory elements and underlying regional specificity

a, Proportion of elements containing TF binding motifs stratified by MPRA activity. **b,** Enriched TF binding
clusters. Statistical significance of enrichment in **a** and **b** were assessed by a two-sided Fisher's exact test.
c, The level of TF motif disruption (x-axis) is related to MPRA allelic differences (y-axis). *P*-values were
estimated by Pearson's correlation test. **d-e,** TF expression levels during prenatal development across
brain regions (colored) were associated with the impact of variant on cortical structure. Each dot represents
DK-atlas brain area colored by macro-scale brain area (**d**) (**Supplementary Fig. 8**). **e**, Brain areas are
colored by scaled expression values (top) and GWAS Z-scores (middle). Seqlogo showing G allele matches
the Rarg binding motif site (bottom). In **c** and **d**, the lines were fitted using linear regression and shadows
indicate 95% confidence interval. *P*-values were corrected for multiple comparisons using the Benjamini–
Hochberg method (**b,d**). O.R: odds ratio.
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Since TF binding is known to regulate gene expression, we explored whether certain TF binding
motifs contribute to MPRA activity. Elements with MPRA activity had more complex sequences
as evaluated by Shannon entropy[33] ($p = 4.58 \times 10^{-24}$ using a t-test) and were more likely to contain
a TF binding motif (**Fig. 4a**), as has been found in previous MPRA studies[22,34,35]. Specifically,
three separate clusters of similar TF binding motifs (TFBMs) – ZNF135/460, MEF2A-D, and Zfx
– were enriched in human cortical structure associated MPRA active elements in phNPCs (**Fig.**
4b). ZNF460 TFBMs were previously found to be enriched in great ape-specific structural variants
and hypothesized to increase brain size along the human lineage[36]. Conditional knockout of
Mef2c in mice was shown to reduce whole brain size[37]. *MEF2C* is also involved in neuronal
differentiation, and genetic variation within or near this gene creates risk for psychiatric
disorders[35,38]. Alu elements contain sequences similar to ZNF135, ZNF460, and MEF2A binding
motifs which exhibit human-specific activity and may be driving this observed overlap[39,40].
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Previous studies have shown that the level of motif disruption is associated with MPRA allelic
effects[34,35], so we evaluated this relationship in our data. We found that emVars strongly
correlated with the predicted level of motif disruption, whereas elements without allelic effects had
no detectable relationship with motif disruption (**Fig. 4c**). Our results suggest that TF binding
affinity impacts the strength of allelic effects on transcription observed in MPRA. Notably,
disruption of RARA, ZNF135, and NR2F1 motifs led to the largest impacts on MPRA activity. It
was reported that the consensus sequences of evolutionary recent Alu subfamiles contain binding
sites for retinoic acid receptors[41], consistent with our finding of stronger MPRA activity within
younger Alu elements (**Fig. 3f,k**). This indicates TFs that may shape the human cortex during
development.

Spatial relationships between TF expression and allelic effects

Genetic variants often are associated with differences in a limited number of cortical regions rather than having global effects across all cortical regions, despite the fact that the variants are constant in all cells (**Fig. 2b**). We hypothesized that the molecular mechanism underlying regionally specific cortical structure effects is that regions more affected by a given variant would also have higher expression of TFs disrupted by that variant, similar to findings from previous studies[42,43]. To test this hypothesis, we compared predicted TF binding motifs disrupted by emVars to corresponding TF expression levels in 16 macroscale brain regions (**Supplementary Fig. 8**) from 8 post-conception weeks (pcw) to 24 pcw from the BrainSpan Atlas[44]. We then analyzed the spatial correlation between TF expression and the impact of the variant on cortical structure from the ENIGMA GWAS. We found the GWAS effect size of 7 cortical structure associated SNPs was significantly correlated with regional TF expression (**Supplementary Tables 4,5**). As one example, elevation of *RARG* expression level leads to a greater reduction in surface area with rs28735393-G (matched to Rarg binding motif, **Fig. 4d,e**). Our results provide evidence that the combination of TF expression levels with genetic variants disrupting the binding of those motifs, leads to alterations of the impact of genetic variants on cortical structure.

 # Wnt activity modulating MPRA activity

 Fig. 5 Regulatory elements associated with cortical structure show condition-dependent
 function
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a, The percentage of MPRA active elements across brain regions stratified by condition (top: Vehicle,
bottom: Wnt Stimulation via CHIR) and measurement (left: surface area, right: thickness). **b,** The number
of MPRA active elements in each condition (bottom). The top panel shows the percentage of condition-
dependent MPRA elements. **c,** Volcano plot showing condition-dependent MPRA activity in Wnt stimulation
vs vehicle. **d,** Enrichment of TEs within elements exhibiting higher activity during Wnt stimulation compared
to elements without detectable differences in MPRA activity across stimulation conditions. **e,** Schematic
example linking MPRA element to genes at their promoter region or using chromatin interaction (Hi-C) for
distal elements. **f,** Percentage of MPRA elements linked to genes stratified by MPRA activity and condition.
Active elements were more likely to be linked to a gene. **g,** Scatter plot showing gene expression changes
vs MPRA activity changes due to Wnt stimulation. A sign test was used to determine if condition-dependent
MPRA activity occurred in the same direction as condition-dependent gene expression more than expected

by chance. **h**, *CDC42EP3* locus highlighting a condition-dependent MPRA element (rs4670759). Top: -
log10(p-value) from GWAS of inferior temporal cortex surface area. Each dot represents variants in the
locus colored by LD (r^2) to the index SNP (rs9309013, purple diamond). Middle: $-\log_{10}$(FDR) of MPRA
activity in stimulation or vehicle conditions. Bottom: Differential chromatin accessible peaks colored by
shrunken log2FC, followed by a promoter region. **i-l,** Boxplots displaying differences between conditions in
multiple measurements: MPRA activity (**i**), chromatin accessibility (**j**), gene expression (**k**) in phNPCs. **l,**
MPRA allelic effects of rs4670759 in stimulated condition. Dots indicate replicates/donors in **i-l**. Same
replicate was connected by line in **l**. ; *n* of replicates/donors: n_{MPRA} = 14, $n_{chromatin\_accessibility}$ = 72, $n_{gene\_expression}$
= 74. *P*-values were corrected for multiple comparisons using the Benjamini-Hochberg procedure. Stim:
Stimulation, Veh: Vehicle, CP: cortical plate of the developing cortex; GZ: germinal zone of the developing
cortex.
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Wnt stimulation alters gene regulation to modulate proliferation and differentiation of neural

progenitor cells shaping brain structure[6,8,45]. To evaluate the condition-dependent activity of

cortical structure associated variants, we stimulated the Wnt pathway in phNPCs using CHIR

(**Methods**). We detected a similar number of MPRA active elements (927 among 17,848

elements; 6.94% of tested variants and 67.17% of tested loci) in the stimulated condition (**Fig.**

5a,**b**, **Supplementary Fig. 6**, **Supplementary Tables 1,3c**). The same TF clusters were enriched

in MPRA active elements under both the vehicle and Wnt stimulated conditions, likely due to 657

active elements being detected in both conditions (**Supplementary Fig. 9**). While we detected

270 elements that were active only in the Wnt stimulated condition (**Fig. 5b**), many of these

regions are likely just above or below the threshold cutoff. To more stringently evaluate condition

dependency, we further identified MPRA active elements with significantly different levels of

activity across conditions. We found 32 elements with higher activity due to Wnt stimulation and

7 elements with higher activity in the vehicle condition, demonstrating Wnt-dependent gene

regulation of cortical structure associated elements (**Fig. 5b,c, Supplementary Table 3d**). MPRA

active elements with higher activity in simulated conditions more often overlapped with SINEs

compared to elements whose activity did not significantly differ between conditions (**Fig. 5d**).

Condition-dependent regulation of gene expression

In order to determine downstream impacts of condition-specific MPRA active elements, we

mapped them to putative target genes. MPRA active elements in gene promoters were directly

assigned to their target genes, while the rest were mapped to the target genes via chromatin

interaction profiles in the developing brain[10]. We found that MPRA active elements were more

likely to be linked to a gene than MPRA inactive elements (**Fig. 5e, f**). Interestingly, elements with

increased MPRA activity under Wnt stimulation were linked to genes with higher expression under
the same condition[6], suggesting these elements function as condition-dependent enhancers[6] (**Fig.**
5g). As an example, rs4670759, a SNP associated with the surface area of the inferior temporal
cortex, showed higher MPRA activity (**Fig. 5h**) and increased chromatin accessibility (**Fig. 5i,j**)
under Wnt stimulation. This variant is located in the promoter of *CDC42EP3* (**Fig. 5h**), which is
involved in actin cytoskeleton formation[46] and whose expression is increased under Wnt
stimulation[6] (**Fig. 5k**). We note this element is an emVar in the stimulated condition, though we
were unable to test for allelic effects in the vehicle condition because one of the alleles was
missing in the QCed dataset (**Fig. 5l**). In summary, we present evidence that condition-dependent
MPRA activity of cortical structure associated SNPs results in condition-dependent effects on
gene expression.
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Condition-dependent allelic effects at cortical structure associated loci

Fig. 6 Condition-dependent allelic effects in cortical structure associated regulatory elements

a, Percentage of emVar across brain regions stratified by condition (left = vehicle, right = stimulation) and measurements (surface area (top), thickness (bottom)). **b,** Number of emVars in each condition (bottom). The percentage of condition-dependent allelic effects (top). **c,** Disrupted TFs and their Wnt induced expression at condition-dependent allelic MPRA elements. The size of the dots indicates the significance of differential expression due to Wnt stimulation colored by shrunken log$_2$FoldChange. **d**, *DCDC1* locus highlighting condition-dependent allelic MPRA element (rs273576). Top; -log$_{10}$(pvalue) from GWAS of Lingual and Pericalcarine surface area. Each dot represents variants in the locus colored by LD (r^2) to the

index SNP (rs2022130, purple diamond). Middle: $-\log_{10}$(FDR) from MPRA activity in stimulation and vehicle
conditions. Bottom: transposable elements and TFs disrupted by rs273576. **e**, Z score of GWAS regional
surface area. **f,** Boxplot shows normalized RNA/DNA ratio of MPRA element (rs273576) stratified genotype
and conditions. **g**, *FOXO3* locus highlighting condition-dependent allelic MPRA element (rs2802295) (left).
Top: $-\log_{10}$(pvalue) from GWAS of global surface area and $-\log_{10}$(FDR) from *FOXO3* eQTL[47] (FDR < 0.05
are shown). Each dot represents variants in the locus colored by LD (r^2) to the index SNP (rs2802295,
purple diamond). Middle: $-\log_{10}$(FDR) from MPRA allelic effects in stimulation and vehicle conditions.
Bottom: transposable elements disrupted by rs2802295 and Wnt responsive element (WRE) and the
position of the designed gRNAs for the CRISPRi experiment. **h**, Z score of GWAS global surface area. **i**,
Boxplot shows normalized RNA/DNA ratio of MPRA element (rs2802295) stratified genotype and conditions.
j, Experimental design of CRISPRi (top), immunofluorescence image showing expression of gRNA
constructs in phNPCs (middle) and relative *FOXO3* expression as measured by qPCR (2 sample t-test with
Holm post-hoc correction, n = 6 unique donors, mean+/- standard error are shown) (bottom).

Many variants (n = 183) showed MPRA allelic effects regardless of stimulation, but more variants

were identified as emVars in only one condition (n_emVars$_{stimulation\ only}$ = 150, n_emVars$_{vehicle\ only}$ =

203), suggesting Wnt stimulation can drive context-specific variant function (**Fig. 6a,b,**

Supplementary Table 3b,e). By using an interaction test, we detected 12 variants with allelic

effects that differ significantly across conditions (condition-dependent emVars) (**Fig. 6b**,

Supplementary Table 3f). Of those, none disrupted TCF/LEF motifs which are TFs downstream

of canonical Wnt pathway activation[48]. However, TFs disrupted by these stimulus-interacting

SNPs were differentially expressed in phNPCs following Wnt stimulation, suggesting a

mechanism by which cortical structure associated variants exert condition-specific allelic effects

(**Fig. 6c**).

An interesting example of a condition-dependent emVar is rs273576 at the intron of *DCDC1*,

where Wnt activation diminished the allelic effects detected in the MPRA (**Fig. 6c-f**). The A allele

is associated with decreased lingual and pericalcarine surface areas and more closely matches

binding sites for ZEB1 and ASCL1, TFs that were differentially expressed in response to Wnt

stimulation[6] (**Fig. 6c,d**). This variant is an eQTL for *MPPED2* (300 kb downstream) in multiple

tissues in GTEx, though not detected in brain or phNPCs.

Another example is rs2802295 within an intron of *FOXO3* (**Fig. 6g**). The G allele is associated

with increased global surface area and decreased expression levels of *FOXO3* in the adult brain[47]

(**Fig. 6g**). *FOXO3* gene expression was also downregulated under Wnt stimulation[6]. We detected

an allelic effect at this variant under Wnt stimulation that was blunted in vehicle (**Fig. 6i**). This

variant is proximal to a Wnt responsive element, where chromatin is less accessible due to Wnt

stimulation. To validate the effects of this non-coding region on *FOXO3* gene expression, we

performed CRISPR interference (CRISPRi) by designing two gRNAs to target dCas9-KRAB
nearby this region (~1kb from rs2802295) (**Fig. 6j**). We found that silencing of this element by
dCas9-KRAB was associated with decreased expression of *FOXO3*, experimentally linking a
cortical surface area associated emVar to its gene of action. Using MPRA, CRISPRi validation,
and QTLs, we suggest a likely causal variant, gene, and context associated with differences in
human cortical structure.

In summary, we identified emVars associated with cortical structure. A subset of these variants
showed differential regulatory activity in response to Wnt stimulation, suggesting that signaling
pathways critical for brain development may influence the regulatory potential of the genetic
variants. These findings reveal how genetic variants shape cortical structure associated
differences in the human brain.

443   Discussion

Here, we assessed the gene regulatory potential of cortical structure associated variants in
phNPCs. We found that most cortical structure associated loci contained elements with regulatory
activity in phNPCs, and that cortical structure associated genetic variants influence the level of
that regulatory activity. Previous QTL studies using endogenous variation in populations of brain
tissue or phNPCs have also found that genetic variants exert effects during neural development
to influence adult cortical structure traits[4,6,49,50]. Our MPRA results provide further evidence to this
effect using experimental manipulation of alleles, ruling out confounding effects from other
variants that exist within an individual.

Because experimental manipulation allows the separation of alleles in high LD, specific genetic
variants can be nominated that alter gene regulatory function. Those variants that disrupt TF
binding motifs can be used to suggest TFs that lead to changes in cortical structure. Here, we
found that TFs like RARA, RARG, ZNF135, and NR2F1 are critical for determining inter-individual
differences in human cortical structure. Though previous work on these TFs has been conducted
in model systems like mice[51–53], our findings demonstrate that the TFs impact human brain
development and structure. We also provide evidence that the expression level of TFs binding to
cortical structure associated variants within a brain region is associated with the impact of those

variants on cortical structure in that region. These results suggest a mechanism for region-specific
effects detected in cortical structure genetic association studies.

Wnt signaling has been well known in non-human model systems and in vitro human model
systems to impact brain development[45]. Therefore, we reasoned that the regulatory effect of
cortical structure associated variants may vary in response to Wnt stimulation. We find that some
genetic variants exhibit Wnt stimulation-dependent gene regulatory function, which we have also
demonstrated in the endogenous genome using chromatin accessibility and gene expression[6].
Notably, genetic effects can be either accentuated or blunted due to Wnt stimulation. This
suggests that multiple cell types and stimulation states will need to be studied to fully elucidate
gene regulatory effects impacting cortical structure.

Alu elements are primate specific and have increased in prevalence in the genome coincident
with brain size increases along the human lineage[15]. Alu elements transcribe themselves through
highjacking proteins derived from other retrotransposable elements and insert themselves into
other regions of the genome, providing a mechanism for more rapid gene regulatory evolution
than the accumulation of single base pair mutations[14]. Here, we provide evidence that cortical
structure associated variants within Alu elements strongly influence transcription in phNPCs. This
suggests that Alu elements have been critical for creating gene regulatory elements that harbor
RARA, RARG, ZNF135, and NR2F1 motifs, which influence human cortical structure[39,40]. Over
time, these Alu elements have likely accumulated genetic variation that alters their function,
impacts gene expression during development, and contributes to inter-individual differences in
cortical structure.

We found that Pol III promoter regions supplied by Alu elements are strong drivers of MPRA
activity. Pol III is generally associated with transcription of short sequences, less than 500 bp in
length. Interestingly, the barcode element is contained within the 3' UTR of the luciferase gene
which is ~1.8 kb from the MPRA element, suggesting luciferase and the barcode are likely
transcribed by Pol II rather than Pol III. Given this evidence, we propose a mechanism whereby
Pol III transcribes the MPRA element, effectively forming an eRNA that enhances Pol II
recruitment, leading to higher expression of the luciferase gene and the barcode contained within.
This mechanism has in fact been observed in neural cells[13].

We implemented MPRA using transfection of episomal plasmids, so our results do not
recapitulate endogenous genome chromatinization, one of the major mechanisms underlying
gene regulation. Likely because of this, MPRA active elements had limited overlap with chromatin
accessibility peaks and MPRA allelic effects had limited overlap with caQTLs that were discovered
and replicated in the cell type (**Supplementary Fig. 1,5**). Nevertheless, a subset of MPRA active
elements did show overlap with eRNAs or uaRNAs active in phNPCs. Also, condition-dependent
MPRA elements showed evidence of mapping to genes with condition-dependent gene
expression. This provides evidence that MPRA active elements do have at least partial overlap
with activity in the endogenous genome. However, we cannot rule out that MPRA activity from
some elements is only due to episomal activity and gene regulatory activity from individual loci
would not be recapitulated within the endogenous genome[54]. Future studies generating specific
base pair edits in the endogenous genome would be able to disentangle these effects.

506 Methods

507 SNP Selection

Several variant sets were generated for inclusion in the MPRA. (1) *positive controls* - Known
promoter or enhancer sequences that are commonly used to constitutively drive gene expression
(minTk promoter, Ef1-alpha promoter, CMV promoter) were included. If the sequences were >150
bp, arbitrary 150 bp subsets were used. (2) *caQTLs* - SNPs within chromatin accessibility peaks
in primary human neural progenitors were selected for inclusion[4,6]. A 150 bp sequence
surrounding the variant from the human reference genome was included. Both alleles were
synthesized. Indels were removed. (3) *ENIGMA GWAS* - SNPs associated with human cortical
surface area or thickness were selected from a large GWAS of cortical surface area and
thickness[3]. Both alleles from all variants with $P < 1 \times 10^{-5}$ in the association to a cortical structure
phenotype and $r^2 > 0.6$ to the identified index SNPs were included. A 150 bp sequence
surrounding the variant from the human reference genome was included. (4) *negative controls* -
Negative controls were generated by randomly scrambling the cortical structure associated
variant sequences for a subset of those variants. Each allele at the SNP of the unscrambled
sequence was included for analysis. Both alleles were removed when the MPRA library sequence
contained a restriction enzyme site used for cloning.

Creating variant oligo library

The 150 bp library oligos mentioned above were synthesized by Agilent and amplified via PCR using the NEBNext 2X Q5 Hifi HS Mastermix (NEB, cat#M0453S) with primers MPRA-chipprimer-R and MPRA-chipprimer-F. Primer information is available in **Supplementary Table 6.**

We then added random 20bp barcodes and restriction sequences (MluI and SpeI) to each variant as previously described[35] using a pair of primers: MPRA-BC_Primer_R and MPRA-chipprimer-F. Following library preparation, the sample was digested with SpeI-HF (NEB, cat#R3133S) and MluI-HF (NEB, cat#R3198S) for 1 hour at 37°C. This step was immediately followed by a 1-hour rSAP treatment (NEB, cat#M0371S) at 37°C, and then a 5-minute heat inactivation at 65°C. Finally, the digested library was purified using the Zymo Clean and Concentrator Kit-5 (Zymo, cat#D4033).

Creating MPRA Plasmid Library

The plasmid pJD_Donor_eGP2AP_RC was obtained from Addgene (Addgene, cat#133784). The plasmid was digested with restriction enzymes SpeI-HF and MluI-HF for 1 hour at 37°C, then was heat inactivated for 20 minutes at 80°C, brought to room temperature, followed by rSAP treatment for 1 hour at 37°C, and heat inactivation for 5 minutes at 65°C. The digested plasmid was then gel extracted from 1% agarose gel using Zymo Gel DNA Recovery kit (Zymo, cat#D4007). The digested plasmid and oligo library was ligated together using T7 DNA ligase (NEB, cat#M0318S) at room temperature for 30 minutes followed by clean up via Zymo clean and concentrator-5 kit. The ligation product was transformed into NEB 5-alpha Electrocompetent E. coli (NEB, #C3020K), grown for 7 hours at 30°C in 4L of LB, then maxi prepped using ZymoPURE II Plasmid Maxiprep Kit (zymo, cat#D4203).

Barcode Mapping

The barcodes (BC)s were mapped to the variants in a method previously described[35]. Briefly, the barcode and variant region was amplified off the plasmid with primers that contain Illumina P5 and P7 adapters (*Bcmap_P5_AAV_R* and *Bcmap_P7_AAV_F_index*) with NEBNext 2X Q5 Hifi HS Mastermix. The PCR product was purified and sequenced using the NovaSeq 6000 SP (2×250 bp) with custom sequencing primers BCmap_R1Seq_AAV_R, BC_Map_R1seq, and BCmap_R2Seq_AAV_F (**Supplementary Table 7**). The code used to map the barcodes to

553   variants can be found in this github repository:
554   MPRA. There was a mean of 113 unique barcodes per variant prior to transfection.

Adding in minimal promoter and Luciferase

The luciferase and minimal promoter sequences were obtained from the pGL4.23 plasmid
(Promega, cat#E841A). The pGL4.23 plasmid was enzymatically digested with KpnI-HF (NEB,
cat#R3142S) and XbaI (NEB, cat#R0145S) at 37°C for 3 hours, followed by heat inactivation at
65°C for 10 minutes. The digestion was cleaned up via Zymo clean and concentrator-5 kit, and
then gel extracted using Zymo Gel DNA Recovery kit, yielding a band size around 1.780Kb. The
plasmid containing the library and barcodes was digested first with KpnI-HF and rSAP at 37°C for
3 hours, then 65°C for 10 minutes for heat inactivation, cleaned up using Zymo clean and
concentrator-5 kit, and then gel extracted using Zymo Gel DNA Recovery kit. Then the product
was digested with XbaI and rSAP 37°C for 2 hours, then 65°C for 10 minutes for heat inactivation,
and cleaned up using Zymo clean and concentrator-5 kit. The digested plasmid containing the
library was ligated with the fragment carrying the minimal promoter and luciferase sequence using
T7 DNA ligase at room temperature for 30 minutes. The resulting construct was then purified with
the Zymo Clean and Concentrator-5 kit, yielding the final library plasmid, PJD_lib_minP_Luc.
PJD_lib_minP_Luc was transformed via heat shock into Library Efficiency DH5α Competent Cells
(Invitrogen, cat#18263012) and grown in kanamycin LB overnight at 30°C. The
PJD_lib_minP_Luc plasmid was maxi prepped with ZymoPURE II Plasmid Maxiprep Kit.

Transfecting PJD_lib_minP_Luc into phNPCs

Acquisition of phNPCs and cell culture conditions were previously described[49]. Donor number 88
was used for these experiments and is biologically male. 6-well plates were coated with Poly-L-
Ornithine (10 µg/ml; Sigma-Aldrich, cat#P3655-100MG) and Laminin (10ug/mL, Corning,
cat#354232). After coating, phNPCs were plated at a concentration of 400,000 cells per well. The
cells were plated and fed with growth factors as described[35]. 24 hours after plating, the cells were
half fed with the media and growth factors described[35]. 48 hours after plating, the cells were half-
fed with the media and growth factors as described[35], and with nucleic acid transfection enhancer
(NATE) (1:100 final dilution; InvivoGen, cat#lyec-nate). 30 minutes after feeding, the cells were
transfected with PJD_lib_minP_Luc plasmid via Lipofectamine Stem Transfection Reagent
(Invitrogen, cat#STEM00001) according to the instructions, using Opti-Mem (Thermo Scientific,

cat#31985070), 2.5ug of PJD_lib_minP_Luc plasmid, and 1ug of pUC19 plasmid (Thermo

Scientific, cat#SD0061) per single 6-well. Co-transfecting with pUC19 plasmid increased

transfection efficiency. 24 hours after transfection, the cells were half fed and then treated either

with 7uL of 2.5uM CHIR or 7uL DMSO (vehicle). At exactly 48 hours after CHIR or vehicle

treatment, RNA and DNA was extracted from the cells using the Quick-DNA/RNA Miniprep Plus

Kit (Zymo, cat#D7003), with 600 µL of Shield Buffer and 600 µL of Lysis Buffer per well. In total,

there were 16 vehicle replicates and 16 CHIR replicates, where one replicate was one well in a

6-well plate. For each 6-well plate, three wells were vehicle and three wells were CHIR treated.

MPRA   Library   Preparation   For   Sequencing

Extracted RNA was reverse-transcribed into cDNA using SuperScript IV Reverse Transcriptase

(Invitrogen, cat. no. 18090050) and a primer designed to bind downstream of the barcodes

(Lib_Hand_RT). The resulting cDNA was then amplified with Lib_Hand and Lib_seq_Luc_R

primers, utilizing NEBNext 2X Q5 Hifi HS Mastermix, and then purified using the Zymo DNA Clean

and Concentrator kit-25 (Zymo, cat. no. D4033).

Additionally, 4 µL of DNA from each well was PCR amplified in four separate PCR reactions using

Lib_Hand_RT and Lib_Seq_Luc_R primers with NEBNext 2X Q5 Hifi HS Mastermix, followed by

purification with the Zymo Clean and Concentrator kit-25. P5 and P7 adapter sequences and i7

demultiplexing indices were then incorporated into both cDNA and DNA libraries through PCR

using P5_seq_Luc_F and P7_Ind_#_Han primers with NEBNext 2X Q5 Hifi HS Mastermix.

The final PCR products were purified and size-selected using AMPure XP Beads (Beckman

Coulter, cat. no. A63881) at 0.7X and 0.9X ratios to isolate a 273 bp DNA fragment. The libraries

were then sequenced on a NovaSeq SP using a 35 × 8 × 0 cycle format with 20% PhiX and

custom sequencing primers Index_primer and R1_primer (**Supplementary Table 8**).

MPRA data processing

<u>Sequencing alignment and Quantification of barcode counts</u>: Adapter sequences were trimmed

using cutadapt (ver. 4.1)[55] using default parameters. Trimmed sequences were then converted to

reverse complement DNA sequences. We extracted only sequences with 20 bp length and

perfectly matched to the BC information in the barcode-variant table which consists of the unique

barcode-variant pairs.

In each replicate, we first filtered BCs if the DNA and RNA read counts had less than 5 read

counts, and kept elements with at least 5 unique BC. Per each variant and replicate,

$\log_2$(RNA/DNA) were fed to boxplot.stats() to detect BC outliers (**Supplementary Fig. 2**). After

removing BC outliers, RNA and DNA counts were combined for each element. Among 16

replicates, those with lower correlation to other biological replicates (correlation coefficients to

others = 0.3 - 0.7) were removed. Elements available at least 10 replicates were retained for

analysis.

621 MPRA regulatory effects

<u>Regulatory activity</u>: For each dataset (caQTL, ENIGMA-vehicle, ENIGMA-stimulation), α values

(transcription rate adjusted by batch effects including sequencing depth) were estimated using

analyzeQuantification() implemented in MPRAnalyze[56] to assess regulatory activity of each

element compared to the distribution of background (α of negative controls). *P*-values of median

absolute deviation score (*P*-MAD score) were corrected for multiple comparisons by the

Benjamini-Hochberg (BH) procedure [FDR]. Condition-dependent MPRA elements were

identified among MPRA elements that were available in both conditions and were active in at least

one of the conditions. Differential activity was estimated using *analyzeComparative()* followed by

testLtr() in MPRAnalyze. For this analysis we excluded one replicate that had library size

correction factors of DNA < 0.10 which led to the failure to execute *analyzeComparative()*. We set

a statistical significant threshold at FDR < 0.1 to define "Active" elements or condition-dependent

elements. In most subsequent analyses including enrichment analyses, each region was

classified if either allele exhibited MPRA activity and as inactive if both alleles showed no

detectable activity to ensure that each region was counted only once.

<u>Allelic effect</u>: We applied *mpralm()* (mpra v.1.16.0)[57] followed by *topTable()* (limma v3.50.3)[58] to

detect allelic effects in MPRA active elements (MPRA element pairs with one of element was

active in any conditions). Condition-dependent allelic effects were assessed with

~genotype+condition+genotype*condition. Only MPRA element pairs that showed allelic

differences (emVars) in at least one condition were used to test for condition-dependent effects.

We set a statistical significance threshold at FDR < 0.1 to define emVars or condition-dependent

emVars.

Transposable elements

Transposable elements (TEs) overlapping MPRA elements were annotated using the RepeatMasker database obtained from the UCSC Table Browser. SINE, LINE, or LTR classes were extracted since those had at least 500 regions overlapped with MPRA elements used in this study (1,825, 1,367, 746, respectively). Alu elements, a subclass of SINEs, were further classified into three groups based on milliDiv score (<100, 100-150, >150)[32] to interpret evolutionary ages.

Relative position of MPRA elements within the Alu sequence: To estimate the position of the MPRA element within the Alu element, the genomic location of each MPRA element was identified and overlapped with forward oriented Alu element that had length ≥ 250 bp. This position was used to estimate the approximate location of the A-Box (TGGCTCACGCC) and the B-Box (GTTCGAGAC), A_5TACA_6, and the poly-A Tail[30,31]. The relative MPRA element position within the Alu was determined by the distance from the Alu start position to the midpoint of the overlapping regions.

Enrichment analysis: Fisher's exact tests were performed to assess the following enrichment of RE features in target groups against specified background.
(1) Frequency of each TE overlaps (active vs inactive), (2) frequency of Alu overlaps (SINE-active vs SINE-inactive), (3) frequency of forwarded oriented Alu (Alu-active vs Alu-inactive), (4) frequency of A/B box followed by TTTT in 150bp, and (5) frequency of emVars in Alu elements (emVars vs MPRA active elements with non significant allelic effects).

When we tested enrichment in stimulation-enhancing MPRA activity, we used the number of RE in MPRA elements with non-significant activity differences between conditions as a reference while running Fisher's exact test. We considered significant enrichment if FDR < 0.1.

TF binding site motif enrichment test

To avoid testing redundant TF binding site [TFBS] motifs[59], we used pre-clustered TF bind motifs available in JASPAR website (https://jaspar2020.genereg.net/static/clustering/2020/vertebrates/CORE/interactive_trees/JASP

AR_2020_matrix_clustering_vertebrates_archive.zip]. Among 137 clusters, we extracted clusters

in which at least one TF was expressed in phNPCs in either Vehicle or Wnt-stimulated condition[6],

resulting in 123 clusters. TFBSs within the 150 bp MPRA elements were predicted by FIMO

(v5.5.2)[60] with FDR < 0.1.

We evaluated (1) enrichment for any TF clusters overlaps, and (2) specific TF clusters using

Fisher's exact test.

Enrichment analysis in various features for MPRA active elements

Since we tested MPRA activity per element not per variant, we extracted one element per variant

with higher activity (i.e. smaller P value) for enrichment analyses unless specified. Those

elements were then grouped as either active elements (FDR < 0.1) or inactive elements (FDR ≥

0.1). In the inactive group, we ensured that both elements were tested and inactive. Two-tailed

Fisher's Exact tests were used to estimate the enrichment or depletion of MPRA active elements

using MPRA inactive elements as a background set unless specified in the section below. A

threshold of FDR < 0.1 was applied to identify statistically significant results.

MPRA elements regulating their putative target genes

Endogenous gene regulatory information: Data were acquired from our previous study[6]. Protein-

coding genes or lincRNAs expressed in unstimulated (vehicle) or CHIR stimulated phNPCs were

extracted. Differentially expressed genes (DEGs) and differentially accessible chromatin peaks

due to CHIR stimulation were defined if FDR < 0.1.

Mapping MPRA element to genes: We mapped MPRA elements to genes by overlapping either

promoter regions defined as 2kb upstream of transcription start site (TSS) in Gencode v19 or

chromatin interaction (Hi-C loops) from germinal zone (GZ) or cortical plate (CP). The proportion

of MPRA elements linked to their putative target genes were compared between MPRA active

elements vs MPRA inactive elements using Fisher's exact test.

<u>Directionality of condition-dependent effect:</u> *binom.test()* was applied to test whether the
percentage of concordant directionality of effect between MPRA activity and gene expression due
to Wnt stimulation was higher than expected by chance.

Prediction of TFs disrupted by emVars

704 motifbreakR(v2.8.0)[61] were used to predict TFs disrupted by emVars on JASPAR2022 vertebrate
705 from MotifDB with method='ic', threshold = 1e-4. We further extracted TFs predicted as strongly
706 impacted (effect= "strong") by motifBreakR and expressed in unstimulated and/or stimulated
707 condition in phNPCs.

CRISPRi experiments

gRNAs targeting a putative regulatory element near rs2802295 were designed using GT-Scan[62]
for CRISPRi experiments. An initial 4 targeting gRNAs positioned at roughly even spacing along
the length of the regulatory element were selected in addition to 2 non-targeting gRNAs as
controls. All gRNAs sequences tested are listed in **Supplementary Table 9** with the ones used
for statistical analyses and high content screening in bold. gRNAs were cloned into pLV-U6-C-
UbC-DsRed-P2A-Bsr (https://www.addgene.org/83919/) for gene expression experiments. Lenti-
dCas9-KRAB-blast (https://www.addgene.org/89567/) vector was used for dCas9-KRAB
expression. Lentiviruses for each plasmid were generated using the psPax2
(https://www.addgene.org/12260/) packaging plasmid and pMD2.G
(https://www.addgene.org/12259/) envelope plasmid as described previously[5]. 293T cells were
cultured directly in phNPC proliferation media for 48 hours after transfection using TransIT-LT1
Transfection Reagent (Mirus Bio). The viral supernatant was then harvested, purified by syringe
filtering, and frozen at -80°C for direct use in subsequent experiments. Viral titers were calculated
using the qPCR Lentivirus Titer Kit (Applied Biological Materials). A preliminary screen of the 4
gRNAs was performed to identify those with the highest potential for perturbing *FOXO3*
expression. phNPCs were plated in 24 well plates at 80,000 cells per well. The following day, cells
were infected with viruses for Lenti-dCas9-KRAB-blast and pLV-U6-C-UbC-DsRed-P2A-Bsr
containing the targeting gRNAs at a calculated MOI of 2 for each virus. Plates were then spun at
37°C for 1 hour to increase transduction efficiency[63]. phNPCs were cultured for an additional 5
days after infection prior to RNA extraction.

RNA purification was performed using the Direct-Zol RNA Microprep Kit (Zymo Research) and

cDNA from 200ng of purified RNA was generated using the iScript cDNA Synthesis Kit (Bio-Rad).

Gene expression was then measured by qPCR using the SsoAdvanced Universal SYBR Green

Supermix (Bio-Rad) and a QuantStudio 5 Real-Time PCR System (Thermo Fisher). *FOXO3*

expression was normalized to the expression of the *ACTB* (Beta-actin) gene. Gene expression

values for each targeting gRNA were compared to the average of the non-targeting controls.

Primers used for qPCR are provided in **Supplementary Table 10.**

From our initial pilot screen, the 2 gRNAs out of 4 with the highest impact on gene expression

were used for downstream experiments. qPCR experiments for each individual gRNA or a

combination of 2 gRNAs were repeated for 5-6 unique phNPC donor lines with the average

expression of 3 wells per donor per gRNA combination considered a biological replicate.

742 Correlation between regional TF expression and GWAS SNP
743 effects

Normalized gene expressions in fetal brains (8 post-conception weeks (pcw) to 24 pcw) were

obtained   from   BrainSpan   Atlas   of   the   Developing   Human   Brain

https://www.brainspan.org/static/download.html[64]. Gene expression was averaged for each

cortical region defined by Allen Human Brain Atlas (AHBA) (**Supplementary Fig. 8**). GWAS Z

scores were calculated using the effect size of the allele that matched the TF motif sequence

divided by the standard error. Since brain regions used within the ENIGMA GWAS were based

on Desikan-Killiany Atlas (DK-Atlas), we mapped 36 DK regions to 16 AHBA regions for

comparison. Linear regression was used to assess correlation between GWAS Z score and gene

expressions.

754 Data availability

755   Data generated in this study are provided as supplementary tables, and raw data will be available

756   upon publication.

Code availability

758 All code used in analyses will be available before publication.

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