

CHD8 regulates neurodevelopmental pathways associated with autism spectrum disorder in neural progenitors

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Truncating mutations of chromodomain helicase DNA-binding protein 8 (*CHD8*), and of many other genes with diverse functions, are strong-effect risk factors for autism spectrum disorder (ASD), suggesting multiple mechanisms of pathogenesis. We explored the transcriptional networks that *CHD8* regulates in neural progenitor cells (NPCs) by reducing its expression and then integrating transcriptome sequencing (RNA sequencing) with genome-wide *CHD8* binding (ChIP sequencing). Suppressing *CHD8* to levels comparable with the loss of a single allele caused altered expression of 1,756 genes, 64.9% of which were up-regulated. *CHD8* showed widespread binding to chromatin, with 7,324 replicated sites that marked 5,658 genes. Integration of these data suggests that a limited array of direct regulatory effects of *CHD8* produced a much larger network of secondary expression changes. Genes indirectly down-regulated (i.e., without *CHD8*-binding sites) reflect pathways involved in brain development, including synapse formation, neuron differentiation, cell adhesion, and axon guidance, whereas *CHD8*-bound genes are strongly associated with chromatin modification and transcriptional regulation. Genes associated with ASD were strongly enriched among indirectly down-regulated loci ($P < 10^{-8}$) and *CHD8*-bound genes ($P = 0.0043$), which align with previously identified coexpression modules during fetal development. We also find an intriguing enrichment of cancer-related gene sets among *CHD8*-bound genes ($P < 10^{-10}$). In vivo suppression of *chd8* in zebrafish produced macrocephaly comparable to that of humans with inactivating mutations. These data indicate that heterozygous disruption of *CHD8* precipitates a network of gene-expression changes involved in neurodevelopmental pathways in which many ASD-associated genes may converge on shared mechanisms of pathogenesis.

CHD8 | NPCs | RNA-seq | ChIP-seq | autism

The genetic architecture of autism spectrum disorder (ASD) is complex and heterogeneous. A wave of recent discoveries has identified individual genes that contribute to ASD when they suffer heterozygous inactivation by coding mutation, copy number variation, or balanced chromosomal abnormalities (1–7). Many of these genes fit neatly into current biological models of ASD involving altered synaptic structure and glutamatergic neurotransmission, but others have been surprising, with a less ready biological interpretation, including genes involved in chromatin modification, DNA methylation, cell adhesion, and global transcriptional regulation. This diversity of genes predisposing to ASD suggests either that there are many pathways that independently can result in the autism phenotype or that functionally distinct ASD-risk genes can trigger consequences that converge on a limited number of shared pathways of ASD pathogenesis. Because now experimental tools are available to reduce gene expression specifically in human neural

progenitor cells (NPCs), which can mimic the impact of functional hemizygosity, we have explored this question by investigating the functional genomic consequences of suppressing chromodomain helicase DNA-binding protein 8 (*CHD8*), a particularly penetrant ASD gene.

CHD8 is an ATP-dependent chromatin remodeler of the SNF2 family (8). *CHD8* was identified as one of the genes in the minimal region of overlap of de novo 14q11.2 microdeletions in two children with developmental delay and cognitive impairment (9). We previously detected direct disruption of *CHD8* by a de novo balanced translocation, with concomitantly reduced mRNA expression, in a patient diagnosed with ASD, intellectual disability, obsessive-compulsive disorder, precocious puberty, macrocephaly, and mild facial dysmorphism (7). No other clinical abnormalities were observed in a recent follow-up examination. Concurrent

Significance

Truncating mutation of chromodomain helicase DNA-binding protein 8 (*CHD8*) represents one of the strongest known risk factors for autism spectrum disorder (ASD). We mimicked the effects of such heterozygous loss-of-function mutations in neural progenitor cells and integrated RNA sequencing with genome-wide delineation of *CHD8* binding. Our results reveal that the molecular mechanism by which *CHD8* alters neurodevelopmental pathways may involve both direct and indirect effects, the latter involving down-regulation following *CHD8* suppression. We also find that *chd8* suppression in zebrafish results in macrocephaly, consistent with observations in patients harboring loss-of-function mutations. We show that reduced expression of *CHD8* impacts a variety of other functionally distinct ASD-associated genes, suggesting that the diverse functions of ASD risk factors may constitute multiple means of triggering a smaller number of final common pathways.

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Data deposition: The sequences reported in this paper have been deposited in the Gene Expression Omnibus (GEO) database (accession no. [GSE61492](http://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE61492)).

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exome sequencing and targeted mutation screening studies have now confirmed unambiguously that de novo truncating mutations of *CHD8* are among the strongest individual risk factors for ASD (1–4, 10). Interestingly, *CHD8* alterations also have been described as somatic events in gastric, colorectal, skin, and glioblastoma multiforme cancers (11–14). Despite its evident importance, little is known from previous studies about the cellular and molecular consequences of disrupting a single copy of *CHD8* and the regulatory connection between this chromatin-remodeling enzyme and the critical pathways associated with either neurodevelopment or cancer. Therefore we sought to determine the effects of perturbing the network of genes regulated by *CHD8* in early neural development by suppressing its expression in human induced pluripotent stem cell (iPSC)-derived NPCs. We integrated transcriptome sequencing (RNA-seq), to evaluate the consequences of *CHD8* suppression on global gene expression, with delineation of the genome-wide distribution of *CHD8*-binding sites using ChIP sequencing (ChIP-seq). Our findings indicate that *CHD8* regulates many functionally distinct genes associated with ASD and members of pathways important to neurodevelopment, suggesting that apparently diverse genetic lesions actually converge on shared pathways of ASD pathogenesis.

Results

Generation and Characterization of Stable *CHD8* Knockdown NPCs.

Fig. 1A provides an overview of the integrative functional genomic approach. To mimic the ~50% reduction in expression of *CHD8* expected to result from heterozygous inactivating mutation [and actually observed in lymphoblasts from our index translocation case (7)], we used lentiviral delivery of shRNAs into a cell type more relevant to ASD, a previously characterized human iPSC-derived NPC line from a control individual, GM8330-8 (15). We used six independent shRNAs targeting *CHD8* coding sequences to ensure a high number of biological replicates and two controls designed against the coding sequence of GFP and bacterial β -galactosidase (LacZ), respectively. All experiments were performed in duplicate, and, in addition, independent infection in each of two batches was carried out for one *CHD8* hairpin (sh6 and sh6_2), one GFP hairpin (GFP and GFP_2), and one LacZ hairpin (LacZ and LacZ_2) (Fig. 1). We performed RNA-seq on all lines using our previously published customization of the strand-specific dUTP method (Fig. 1B and C) (16) and also carried out Western blotting using three independent commercially available antibodies (Fig. 1D). The knockdown of *CHD8* RNA ranged from 38–69% across lines (Fig. 1C). The degree of *CHD8* suppression did not affect NPC morphology or the expression levels of the neural ectodermal markers paired box 6 (*PAX6*), sex determining region Y-box 1 (*SOX1*), and Musashi homolog 1 (*MSH1*) in comparison with nontargeting controls (GFP and LacZ) (Fig. S1A and Dataset S1).

Transcriptional Consequences of *CHD8* Suppression in NPCs. We generated an average of 40.6 million reads per line by strand-specific RNA-seq to monitor changes in genome-wide gene expression (Dataset S2B). All libraries contained synthetic RNA spike-ins, which we used to determine empirical transcript detection thresholds (Fig. S2B and C); transcripts from 15,903 genes were detectable above thresholds in all lines. We performed analysis of differential expression incorporating batch and treatment as factors in a regression model (17). Overall, 1,756 genes were differentially expressed as a consequence of *CHD8* suppression (nominal $P < 0.05$), 369 of which were significant at Benjamini–Hochberg $q < 0.05$ (see Fig. S2D for the full range of differentially expressed genes from $q < 0.1$ to $q < 0.0001$). Many more genes were up-regulated than down-regulated following *CHD8* suppression (1,140 vs. 616). Pathway and gene ontology (GO) term enrichment analysis revealed a striking difference in the nature of the pathways represented by up-regu-

lated and down-regulated genes (Fig. 2 and Dataset S3). The former, larger set was associated with the terms “chondroitin sulfate biosynthesis,” “cytoplasmic sequestering of protein,” and “RING-type zinc fingers” (Fig. 2 and Dataset S3), whereas the smaller latter group was associated with terms related to neural development and function including “cell adhesion,” “neuron differentiation,” “synapse,” “ion transport,” “axon guidance,” “cadherin signaling pathway,” and “protocadherin gene family” (Fig. 2 and Dataset S3). Weighted gene coexpression network analysis (WGCNA) (18) clustered all 15,903 genes into 21 modules and identified four modules that had very high correlation with *CHD8* expression ($r > 0.7$ and $P < 1 \times 10^{-4}$) (see Figs. S3 and S4 for all coexpression modules and protein–protein interactions and SI Materials and Methods for complete details). Consistent with the pathways associated with down-regulated genes, “cell adhesion” (in addition to “Wnt signaling” and “cell projection”) was among the most enriched annotation terms for coexpression modules with genes whose expression decreased in correlation with *CHD8* suppression (Fig. S3).

These analyses provide a direct link between the chromatin modifier *CHD8* and regulation of genes of critical importance to neural development in humans. Many of the strongest individual effects were detected among genes involved in neuronal function or synaptic regulation [e.g., laminin alpha 4 (*LAMA4*), $P = 5.95 \times 10^{-13}$; neural cell adhesion molecule 1 (*NCAMI*), $P = 2.99 \times 10^{-10}$; *LRR4B*, $P = 9.17 \times 10^{-10}$; *TIMP3*, $P = 2.65 \times 10^{-10}$; multiple EGF-like-domains 10 (*MEGF10*), $P = 1.1 \times 10^{-8}$; discs large homolog 2 (*DLG2*), $P = 1.19 \times 10^{-8}$; *SLIT1*, $P = 1.38 \times 10^{-8}$, to list a few], including genes previously implicated in ASD risk [e.g., sodium channel, voltage-gated, type II, alpha subunit (*SCN2A*), $P = 3.85 \times 10^{-9}$; methyl-CpG binding domain protein 3 (*MBD3*), $P = 4.9 \times 10^{-8}$; SH3 and multiple ankyrin repeat domains 3 (*SHANK3*), $P = 2.4 \times 10^{-4}$]. The majority of the most significant genes involved in neural development were down-regulated. Consequently, we examined the entire set of down-regulated genes and found it to be strongly enriched for 628 genes associated with ASD [$P = 3.25 \times 10^{-8}$, odds ratio (OR) = 2.78] as defined by the SFARI gene 2.0 (19) (574 genes) and AutismKB (20) (171 genes; 117 overlap with SFARI) databases, both of which have established varying levels of support for an association with ASD of each gene included (see SI Materials and Methods for details on supporting evidence for these gene lists).

Based upon the pathways associated with up-regulated genes and the reported association of *CHD8* with multiple cancers, we performed similar analyses using a large list of 5,873 cancer-associated genes (“TCGA cancer”) compiled from a variety of studies by The Cancer Genome Atlas (TCGA) Gene Ranker (Materials and Methods). We found enrichment of these TCGA-defined genes among the up-regulated loci ($P = 9.82 \times 10^{-7}$, OR = 1.37). Notably, these results were specific, because there was no enrichment of ASD genes in the up-regulated set ($P = 0.79$) and no enrichment of TCGA cancer loci in the down-regulated set ($P = 0.51$), prompting us to investigate further the molecular mechanisms driving these differences.

Genome-Wide Targets of *CHD8* Binding. Because the genome-wide binding sites of *CHD8* had not been delineated previously, we performed ChIP-seq in control cells using three independent *CHD8* antibodies to distinguish indirect from potentially direct consequences of changing *CHD8* levels. Very deep sequencing was performed for all antibodies (61–84 million reads per antibody) and input (88 million reads). All three *CHD8* antibodies yielded similar genomic distributions that were highly enriched for peaks at promoter regions (Fig. 3A–C). Overall, we identified widespread binding of *CHD8* throughout the genome, with 7,324 sites that were replicated by all three antibodies at a Benjamini–Hochberg q value < 0.05 using Model-Based Analysis of ChIP-Seq 2

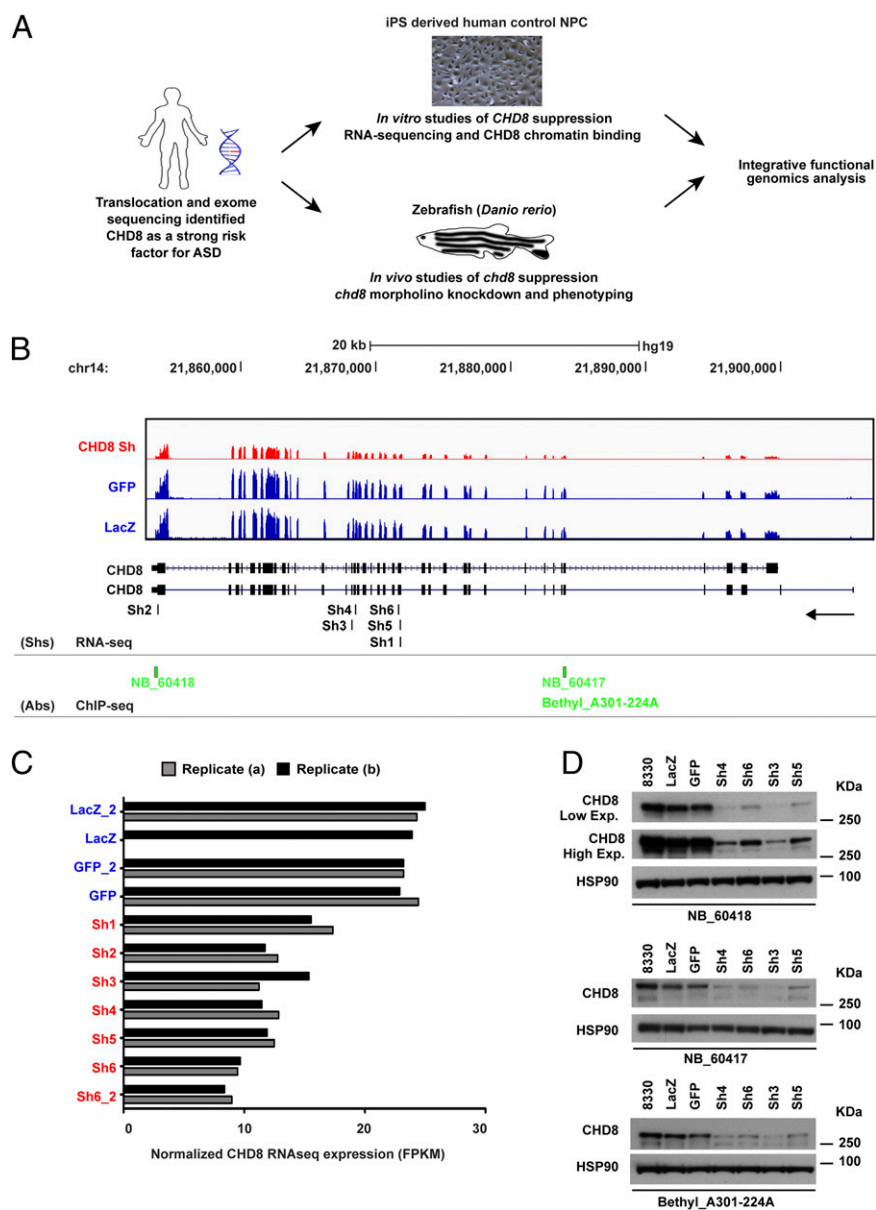


Fig. 1. Generation and characterization of human NPC lines with stable *CHD8* knockdown. (A) Schematic representation of the study design and experimental flowchart presented in the article. Following the identification of *CHD8* as a strong risk factor for ASD, we characterized the transcriptional effects of *CHD8* knockdown in human control NPCs and the genome-wide binding targets of *CHD8*. In parallel, we analyzed the in vivo phenotypes associated with *chd8* suppression in zebrafish. The functional genomics output emerging from integrated analyses of these datasets is discussed in this paper. (B) *CHD8* expression levels measured from RNA-seq are shown for control NPCs (blue) and stable *CHD8* knock-down (KD) clones (red). Pooled tracks for all samples in each condition are presented. Track height is proportional to total library size. The locations of the different shRNA sequences used (Sh1–Sh6) are indicated at the bottom of the graph (RNA-seq row), and the epitope regions of the different *CHD8* antibodies used in the ChIP-seq studies are indicated in green (ChIP-seq row). (C) Normalized expression levels of *CHD8* transcript are plotted for technical (Replicate a) and biological (Replicate b) replicates as normalized expression values for convenience. FPKM, fragments per kilobase per million reads. Reduced *CHD8* expression was observed in all knockdown clones. LacZb is excluded; it was removed because of insufficient reads (SI Materials and Methods). (D) Western blotting analysis of *CHD8* protein levels in *CHD8* stable knockdown clones for each of the three antibodies used (NB_60417, NB_60418, and Bethyl A301-224A). Two different isoforms of *CHD8* protein (~270 kDa and ~290 kDa) were observed in the control lines (8330, GFP, and LacZ), and down-regulation of protein was observed only for *CHD8* knockdown clones (Sh3, Sh4, Sh5, and Sh6). Comparable amounts of total protein were used for different samples, and HSP90 was used as loading control.

(MACS2) (Fig. S5B). We focused all subsequent integration with expression-related analyses on this stringently defined set of replicated sites. Overlaying the *CHD8*-binding sites with genome-wide chromatin states from www.broadinstitute.org/~anshul/projects/roadmap/segmentations/models/coreMarks/parallel/set2/final/ (accessed May 28, 2014) in an ES cell-derived neural progenitor line generated by the Roadmap Epigenomics consortium (21), we found that *CHD8*-binding sites are localized

predominantly to genomic regions marked by histone H3 trimethyl Lys4 (H3K4me3), signifying active transcription start sites (TSSs), with 83% of *CHD8*-binding sites being in an active TSS state, as compared with only 1% of the whole genome (Fig. 3D). *CHD8*-binding sites also are enriched to a lesser extent (twofold) for enhancer status (characterized by the presence of H3K4me1), comprising 4% of sites as compared with 2% of the whole genome (Fig. 3D). In our expression dataset, 5,658 genes

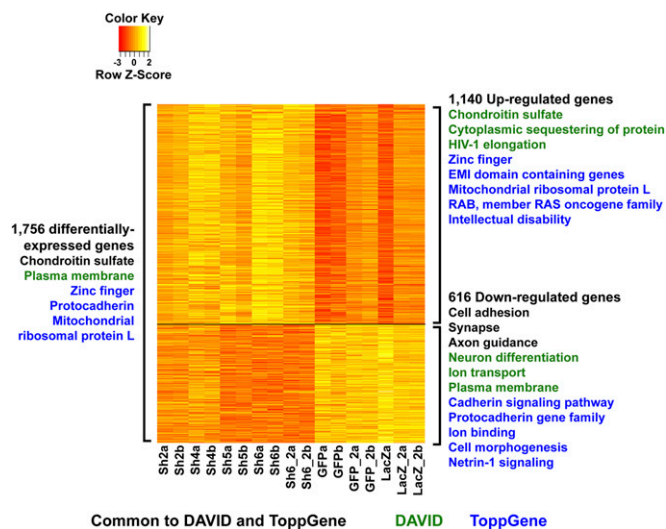


Fig. 2. Differentially expressed genes and associated annotation terms. The heatmap shows gene expression in \log_2 cpm after batch correction for the 1,756 differentially expressed genes, with genes down-regulated by CHD8 knockdown (616 genes) on the bottom and genes up-regulated following CHD8 suppression (1,140 genes) on the top. Values have been centered and scaled for each row. Each row represents a single gene. Statistically significant functional annotation and pathway terms identified using DAVID (FDR < 5%) and ToppGene (Bonferroni-corrected $P < 0.05$) for all 1,756 differentially expressed genes are listed on the left. On the right of the heatmap, significant terms are provided for down- and up-regulated genes separately. Similar terms have been condensed and summarized for simplicity in this figure; the full list of associated terms and P values is provided in [Dataset S3](#). The most significant terms for up-regulated genes were “chondroitin sulfate biosynthesis” ($P = 2.55 \times 10^{-6}$) and “mitochondrial ribosomal protein L genes” ($P = 2.28 \times 10^{-6}$); for down-regulated genes the most significant terms were “plasma membrane” ($P = 4.31 \times 10^{-11}$), “protocadherin genes” ($P = 1.16 \times 10^{-10}$), “calcium ion binding” ($P = 1.35 \times 10^{-7}$), and “single organismal cell-cell adhesion” ($P = 4.78 \times 10^{-7}$). P values for synapse, neuron differentiation, and axon guidance among down-regulated genes ranged from 2.55×10^{-3} to 2.81×10^{-5} (see [Dataset S3](#) for complete results).

contained at least one CHD8-binding site within 10 kb of the TSS, and the proportion of genes that have a CHD8-binding site increased with increasing baseline gene expression ([Fig. S5C](#)). Analysis of the set of all genes with CHD8-binding sites using GREAT (22) yielded many enriched functional annotations, but the top GO molecular functions were related to transcriptional regulation, and enriched terms from pathway databases included “p53 pathway,” “Hedgehog signaling pathway,” and “cell cycle” ([Dataset S4](#)). In this system, 522 of the genes with CHD8-binding sites were differentially expressed, representing only 9.2% of all CHD8-bound genes and 29.7% of all differentially expressed genes, indicating that the majority of the gene-expression changes that we detected upon CHD8 suppression are likely to be caused by indirect regulatory effects. A higher proportion of up-regulated genes than down-regulated genes are bound by CHD8 ([SI Materials and Methods](#) and [Fig. S6 A–C](#)).

To explore sequence motifs at the CHD8-binding sites, we carried out de novo motif analyses separately for the sets of binding sites detected by each of the three antibodies. We found that a motif matching that for CCCTC-binding factor (CTCF) binding was the most significant motif that was replicated between antibodies ([SI Materials and Methods](#) and [Dataset S5](#)), as is consistent with CTCF being a previously known interactor of CHD8 (23). A motif matching the binding site for yin yang 1 (YY1), a ubiquitous transcription factor that interacts with CTCF (24), also was discovered, as the second most significant hit ([Dataset S5](#)). Other

enriched motifs that may represent coactivators or corepressors with CHD8 are listed in [Dataset S5](#).

Distinct Characteristics of CHD8 Regulation Among ASD and Cancer Gene Sets. Integration of CHD8-binding sites with the differentially expressed genes gave further insight into the gene enrichments initially noted and into the diverse functions of genes associated with ASD. It was the set of down-regulated genes that lack CHD8-binding sites (i.e., genes down-regulated indirectly by CHD8 suppression) that was enriched for neurodevelopmental, cadherin/cell adhesion, and axon guidance pathways ([Datasets S6–S8](#); the most significant genes are listed in Table 1). Among these indirectly down-regulated genes there also was a strong enrichment for the broad set of ASD-associated genes as defined by SFARI and AutismKB ($P = 1.09 \times 10^{-9}$, OR = 3.39), which includes genes discovered from previous genetic studies as well as genes investigated based on prior neurobiological hypotheses of ASD. This gene set also was nominally significant for enrichment among genes with CHD8-binding sites ([Fig. 4A](#)). Integration of CHD8-binding sites also provided further specificity of the TCGA cancer-associated gene-set enrichment. These genes were highly enriched among all genes with CHD8-binding sites ($P = 1.55 \times 10^{-58}$, OR = 1.80), a result that was significant regardless of whether the genes were differentially expressed. This same gene set was not enriched significantly among either up-regulated or down-regulated genes that did not possess CHD8-binding sites, indicating that potential for direct regulation by CHD8 was the primary driver of the TCGA enrichments ([Dataset S8A](#)).

We further scrutinized these findings by testing for enrichment of genes from additional ASD and cancer gene sets that were compiled using different criteria. We first evaluated a smaller list of SFARI ASD genes curated using the same criteria applied by Parikshak et al. (25) [235 genes; SFARI gene score = syndromic (S) or evidence level 1 (“high confidence”) to 4 (“minimal evidence”)]. Like the larger ASD gene set, this restricted set was enriched among down-regulated ASD genes lacking CHD8-binding sites ($P = 2.26 \times 10^{-2}$, OR = 2.18; [Dataset S8A](#)) and was not associated with CHD8-bound genes ($P = 0.16$; [Dataset S8A](#)). When we performed analyses with a narrowly defined set of genes associated with ASD based on harboring at least a single de novo loss-of-function (LoF) mutation from exome sequencing studies [131 genes were tested by combining nine statistically significant “high-confidence ASD” genes with another 122 “probable ASD” genes defined by Willsey et al. (26)], we observed enrichment among CHD8-bound genes ($P = 4.34 \times 10^{-3}$, OR = 1.69) but not among down-regulated genes ([Fig. 4A](#)), in contrast with the SFARI/AutismKB results. When we interrogated the genes and pathways underlying the differences among these ASD gene sets, we found a consistent pattern: The genes identified from de novo LoF mutation were enriched for CHD8 binding and consistently were associated with chromatin modification and transcriptional regulation ([Dataset S8B](#)), whereas the SFARI/AutismKB dataset was enriched for genes without CHD8-binding sites that were down-regulated following CHD8 suppression, and these genes were associated with cell adhesion and neurotransmitter/axon-related pathways ([Dataset S8B](#)).

These disparate enrichment patterns for ASD-associated genes based on the mode of discovery and pathways implicated also shed light on the patterns observed in network analyses performed by Parikshak et al. (25) and Willsey et al. (26), who generated coexpression networks across brain regions and development from BrainSpan ([www.brainspan.org](#)). Of the five coexpression modules found to be enriched for ASD genes in the study by Parikshak et al., two (M2 and M3) consisted of genes expressed early in fetal neocortical development and were enriched for transcriptional and chromatin regulators and genes harboring rare de novo LoF mutations from previous exome

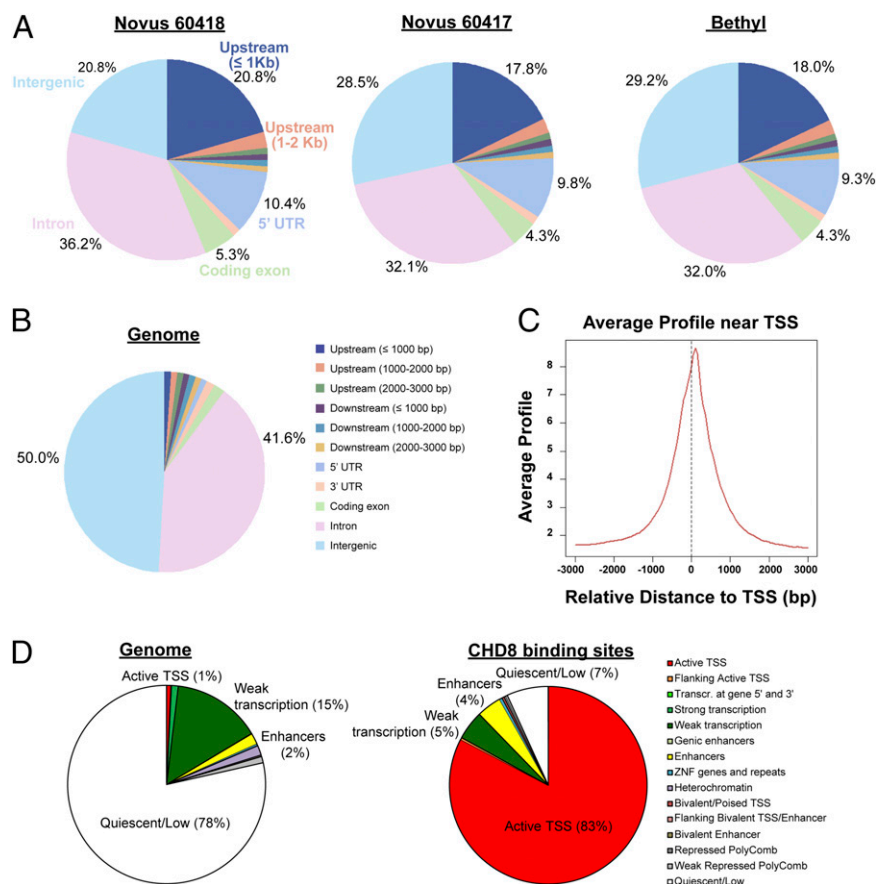


Fig. 3. Distribution of ChIP-seq peaks from three CHD8 antibodies. (A) Genomic distribution of sequence peaks captured by each of the three antibodies, compared with the whole genome. Upstream regions are defined as regions upstream of the TSS; the 5' UTR is the region between the TSS and the coding start site. (B) Whole-genome distribution of the genomic features in A. "Intergenic" refers to anything that does not fall into any of the preceding categories in the legend shown on the right. (C) ChIP-seq read density relative to TSSs for one representative antibody (Novus 60417). We found 7,324 peaks that were detected by all three antibodies. These peaks were mapped to 5,658 genes. (D) Distribution of chromatin states identified by the Roadmap Epigenomics consortium (21) in an ES cell-derived NPC for the whole genome (Left) and the 7,324 CHD8-binding sites detected by all three antibodies (Right).

studies. Three modules (M13, M16, and M17), consisting of genes expressed in late fetal to early postnatal stages, were enriched for synaptic proteins and SFARI ASD genes. We observed a clear separation in the overlap between our gene categories defined by *CHD8* regulation and these five modules: Down-regulated genes without CHD8-binding sites were enriched for genes belonging to M13, M16, and M17 but not M2 or M3, whereas CHD8-bound genes were enriched for genes belonging to M2 and M3 but not M13, M16, or M17 (Fig. 4B). We also obtained concordant patterns with the complementary study carried out by Willsey et al. (26), in which three co-expression modules were found to be enriched for de novo LoF ASD genes. All three of those modules were enriched among the set of all CHD8-bound genes in our study (Fig. 4B) but not the indirectly down-regulated genes.

In cancer, the TCGA gene set is both large and broadly defined, so we similarly turned to two gene sets compiled using narrower criteria: one much smaller set of 224 genes based upon somatic point mutations in 21 tumor types from a recent publication from Lawrence et al. (13) and a second manually curated Catalog of Somatic Mutations in Cancer (COSMIC) cancer gene census of 513 genes based on causally implicated mutations (27). Both smaller cancer gene sets also were highly enriched among genes with CHD8-binding sites and most strongly among the subset of CHD8 targets that do not show differential expression in NPCs because of CHD8 knockdown ($P < 1.9 \times 10^{-11}$ for all three gene sets; Fig. 4A and Dataset S84). Notably, when we considered

the 302 genes present in the SFARI/AutismKB ASD and TCGA cancer genes sets (the two most broadly defined gene sets), we found the same pattern of enrichment as in the SFARI/AutismKB genes overall, with enrichment only among down-regulated and unbound genes ($P = 0.023$, OR = 2.01; Dataset S8C).

Given the significance of the findings in ASD and cancer gene sets, we sought to establish their specificity. We tested all the 184 gene lists available from other complex human diseases and traits obtained from the National Human Genome Research Institute (NHGRI) Genome-Wide Association Study (GWAS) Catalog (28) (minimum gene-set size = 10 genes). No results approached the significance levels of the ASD and cancer enrichments, because the most significant result was human trait-related enrichment "height" ($P = 4.46 \times 10^{-6}$ among the set of CHD8-bound genes; Dataset S9). In comparison, the most significant ASD and cancer enrichments were 1.01×10^{-9} and 1.55×10^{-58} , respectively (Fig. 4A). However, we did find enrichment of curated gene sets associated with schizophrenia (29) and bipolar disorder (30) among genes down-regulated but not bound by CHD8, following the same pattern as the SFARI/AutismKB ASD genes (2.36×10^{-2} and 1.70×10^{-3} , OR = 2.45 and 3.28, respectively). Notably, we did not find enrichment of the genes in proximity to the 108 common polymorphisms that were significant genome-wide in a recently published study of schizophrenia (31). Genes associated with intellectual disability (25) also were enriched among down-regulated genes but, like the ASD genes defined by truncating mutations, were enriched more significantly among

Table 1. Differentially expressed and CHD8-bound genes associated with ASD and neurodevelopmental pathways

Gene	ASD list	FC, knockdown/control	P value
Selected genes that are indirectly regulated by CHD8*			
LAMA4		-4.44	5.95×10^{-13}
TIMP3		-3.23	2.65×10^{-10}
KCNJ10	S/A	-4.95	3.44×10^{-10}
SCN2A	Both	-7.31	3.85×10^{-9}
SLIT1		-4.88	1.38×10^{-8}
MBD3	S/A	2.72	4.90×10^{-8}
BAI1		-3.25	2.40×10^{-7}
SYTL4		-3.45	2.79×10^{-7}
GPX1	S/A	1.99	2.58×10^{-5}
SOX9		-2.08	3.33×10^{-5}
HS3ST5	S/A	6.91	4.05×10^{-5}
ACSBG1		-2.16	6.88×10^{-5}
SHANK3	S/A	1.87	2.39×10^{-4}
EFHD1		-2.87	2.66×10^{-4}
LIFR		-1.78	3.02×10^{-4}
TESK2		-2.41	4.03×10^{-4}
HYDIN	S/A	-3.30	7.09×10^{-4}
PLXNA2		-2.20	8.01×10^{-4}
ANO5	W	-1.93	9.12×10^{-4}
RIMS3	S/A	-1.73	9.64×10^{-4}
KANK1	S/A	-1.74	9.95×10^{-4}
Selected ASD-associated genes that are bound by CHD8†			
CHD8‡	Both	-2.56	3.62×10^{-9}
NFKBIL1‡	W	2.02	7.61×10^{-5}
CBX4‡	W	1.85	3.79×10^{-4}
TCF3	W	1.69	1.17×10^{-3}
SDC2	S/A	1.79	1.20×10^{-3}
RAI1	S/A	1.69	1.57×10^{-3}
SLITRK5	S/A	1.69	2.03×10^{-3}
PTEN	S/A	-1.81	3.00×10^{-3}
ARHGAP2	S/A	2.67	5.33×10^{-3}
OGT	S/A	-1.49	0.0111
RPS6KA2	S/A	-1.51	0.0114
TRIO	S/A	1.44	0.0424
ADK	S/A	-1.40	0.0430
LZTS2	S/A	1.39	0.0432
CBS	S/A	1.38	0.0437
POGZ	Both	-1.30	0.100
ARID1B	Both	-1.20	0.264
MBD5	Both	-1.19	0.307
TRIP12	Both	-1.17	0.337
ADNP	Both	-1.13	0.436
SETD2	Both	-1.05	0.757
SYNGAP1	Both	1.04	0.818
CUL3	Both	1.01	0.951
SUV420H	Both	1.00	0.978

Both, genes in both the SFARI/AutismKB and Willsey et al. (26) lists; FC, fold change of gene in shRNA knockdowns compared with controls: A positive fold-change corresponds to up-regulation when CHD8 is knocked down; a negative-fold change corresponds to down-regulation (*P* values for differential expression are listed for each gene); S/A, genes in the ASD gene list from SFARI and AutismKB, as described in *SI Materials and Methods*; W, genes in the Willsey et al. (26) pASD or hcASD gene list. See *Dataset S1* for all detectable genes and associated data.

*Genes that are strongly differentially expressed following *CHD8* suppression (Benjamin-Hochberg $q < 0.05$) and lack *CHD8*-binding (i.e., that are indirectly regulated by *CHD8*). Genes shown are either ASD-associated genes or down-regulated genes associated with cell adhesion or neurodevelopmental pathways.

†ASD-associated genes in which *CHD8*-binding sites were detected by all three antibodies ($q < 0.05$). Genes either were differentially expressed with a *CHD8*-binding site or were bound by *CHD8* and found in both ASD gene sets described.

‡These genes met $q < 0.05$ for differential expression. FC, fold change of gene in shRNA knockdowns compared with controls. A positive fold-change

CHD8-bound genes (*Dataset S8A*). Only “targets of fragile X mental retardation protein 1 (FMRP),” a gene set defined by molecular analysis (high-throughput sequencing together with UV-crosslinking and immunoprecipitation, HITS-CLIP) rather than by disease association (32), was enriched among up-regulated genes (as well as among *CHD8*-bound genes) (*Dataset S8A*). For *CHD8*-bound genes, one of the most significant human phenotypes from the Human Phenotype Ontology (HPO) database was “abnormality of skull size” ($P = 1.04 \times 10^{-23}$, OR = 2.49), following “abnormality of the cerebrum” and “abnormality of the forebrain” (*Datasets S4* and *S8A*). All these are consistent with the clinical phenotype of our index case with translocation interrupting *CHD8*, who exhibits ASD, intellectual disability, and macrocephaly, and with the phenotypes of other subjects heterozygous for inactivating *CHD8* mutation.

In Vivo Analysis of *CHD8* in Zebrafish Embryos. Given that virtually all patients reported with truncating mutations in *CHD8* have a macrocephalic phenotype (33), we asked whether suppression of *chd8* might lead to increased head size in *Danio rerio* (zebrafish) embryos by acting on early neurogenesis. We have shown previously that head-size evaluations in zebrafish embryos can serve as a surrogate for the evaluation of candidate genes for neurocognitive traits (34). Therefore we suppressed the sole zebrafish ortholog of *CHD8* and evaluated the gross morphometric and cell-specific characteristics of morphants.

Using reciprocal BLAST, we first identified a single zebrafish *CHD8* ortholog (*chd8* on chromosome 2; 62% amino acid identity). Next, we designed two splice-blocking morpholinos (sb-MO), targeting the splice donor site of exons 7 and 8 respectively, which we injected into embryos at the one- to two-cell stage. Masked quantitative scoring of embryos at 4.5 d postfertilization (dpf) injected with *chd8* MO1 showed a reproducible macrocephaly phenotype (12% increase in morphants compared with controls, $P < 0.0001$) (Fig. 5 A–C). This phenotype was paralleled by the efficiency of splice blocking of the two sb-MOs, which led to the retention of introns 7 and 8, respectively, and the presence of a premature stop codon, as established by RT-PCR and Sanger sequencing (Fig. S7 A–E). We further characterized the *chd8* transcripts by quantitative PCR and RNA-seq to confirm the impact of the morpholino on *chd8* transcription (*SI Materials and Methods* and Fig. S7 F–H). A scrambled morpholino induced no phenotypes (34). Finally, macrocephaly was unlikely to be driven by overall developmental delay; morphants had a normal appearance with regard to their pigment cells, there was no apparent pathology in other external organs, such as the heart or the swim bladder, and their body length was indistinguishable from that of control embryos from the same clutch.

To probe further the underlying cause(s) of the macrocephalic phenotype, we stained embryos at 2 dpf with an anti-HuC/D (human neuronal protein Hu antigen, a marker for newborn neurons). We selected this time point because it precedes the development of macrocephaly and therefore allowed us to evaluate the forebrain before the appearance of gross anatomical defects. We observed a striking increase in HuC/D expression, which appeared ectopic in 65% of embryos injected with *chd8* MO1, as compared with controls (Fig. 5 D–F). Next, we stained the embryos with a phospho-histone H3 antibody, which is an M-phase marker, and quantitatively scored the number of proliferating cells in *chd8* morphants and controls. We counted an average of 408 p-histone H3⁺ cells for controls compared with 518 p-histone H3⁺ cells in embryos injected with *chd8* MO1 ($P = 0.0018$; Fig. 5 G–I), indicating that the macrocephaly phenotype is likely to be caused by disturbed neuronal proliferation at early developmental stages.

corresponds to up-regulation when *CHD8* is knocked down; a negative-fold change corresponds to down-regulation. *P* values for differential expression are listed for each gene.

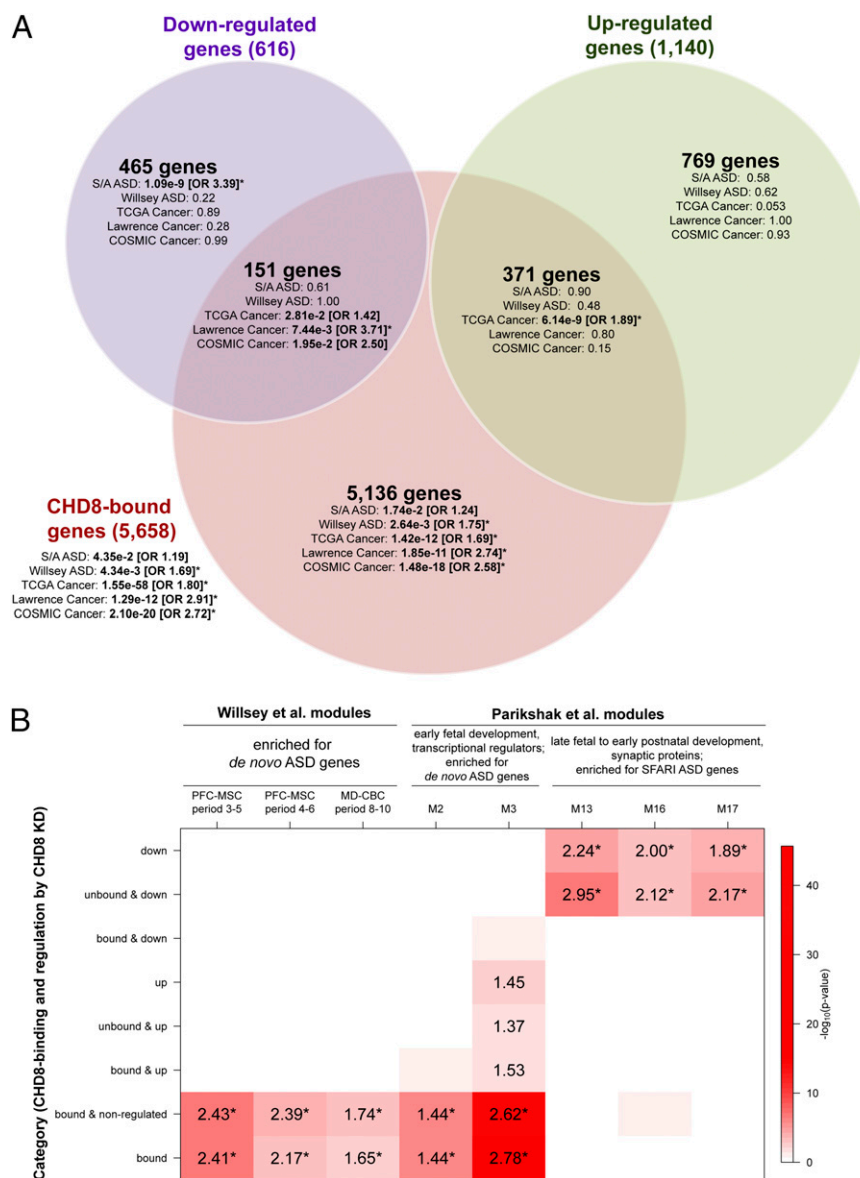


Fig. 4. Enrichments for ASD and cancer gene sets and published BrainSpan coexpression networks among CHD8-regulated and CHD8-bound genes. (A) Gene-set enrichments are shown for the sets of genes that are both differentially expressed and bound by CHD8, differentially expressed only, or CHD8-bound only, shown as a Venn diagram. Enrichments for the set of all genes bound by CHD8, independent of differential expression, are shown outside the Venn diagram. For each set of genes in the Venn diagram, enrichment P values are shown for five disease gene lists. ORs are shown for enrichments that met $P < 0.05$, with asterisks indicating enrichments that met $q < 0.05$. Disease gene lists were obtained from SFARI and AutismKB (S/A ASD), *de novo* LoF mutations in ASD are from Willsey et al. (26) (Willsey ASD), TCGA gene ranker (TCGA Cancer), the pan-cancer exome sequencing study by Lawrence et al. (13) (Lawrence Cancer), and the Wellcome trust ("COSMIC Cancer"), as described in *SI Materials and Methods*. All gene-set analyses and Benjamin-Hochberg-corrected P values are provided in [Dataset S8A](#). (B) For each set of CHD8-regulated or CHD8-bound genes, enrichments are shown for overlap with BrainSpan coexpression modules generated by Willsey et al. (26) and Parikshak et al. (25) that had been found to be enriched for ASD genes. The red shading in each cell corresponds to the $\log_{10} P$ value for enrichment, as shown in the color scale on the right. The number in each cell is the OR for enrichment, shown only if the enrichment met $P < 0.05$. Enrichments that met $q < 0.05$ are indicated by an asterisk next to the OR. Names of coexpression modules are as reported in the respective publications.

Discussion

CHD8 is a prominent example of several genes, including other chromodomain helicases (*CHD7*, *CHD3*, *CHD2*), histone demethylases (*ARID1B*, *KDM6B*, *KDM6A*), methylases (*MLL5*, *EHMT1*, *METTL2B*), and methylated DNA-binding proteins (*MBD5*, *MBD3*), that have implicated the disruption of chromatin regulation as a precipitating factor in ASD. However, regulation of chromatin is but one of the many cellular processes that has been proposed by genetic and biological studies in ASD. Investigation of ASD pathogenesis also has focused on RNA

surveillance, cell adhesion, synaptic proteins, glutamate neurotransmission, ion transport, and other functions, suggesting that ASD may involve related phenotypes caused by quite different pathogenic mechanisms. Previous studies have sought insight from curated protein-protein interaction (PPI) databases to connect the regulatory networks associated with genes involved in these biological functions (3). Recently, weighted gene coexpression network analysis (18) has been used in elegant studies to explore directly the coexpression of genes that drive neurodevelopment, measured either by microarray or RNA-seq, in human brains and

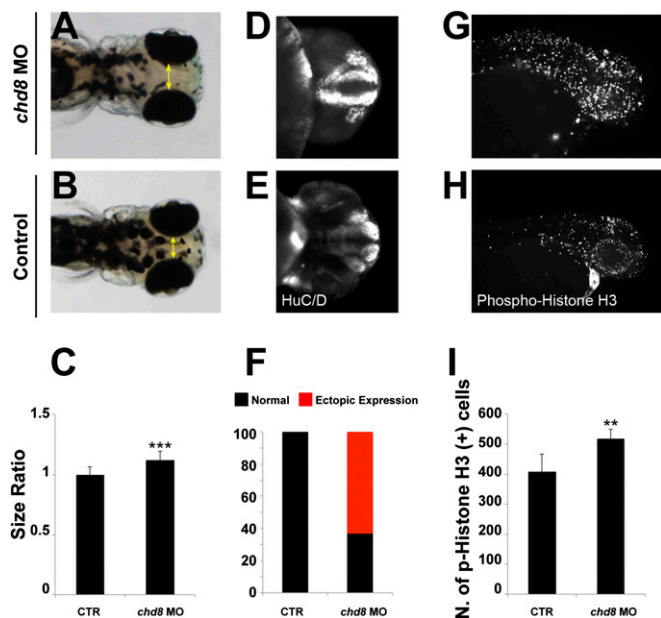


Fig. 5. Injection of *chd8* MO leads to macrocephaly, ectopic expression of HuC/D, and increased numbers of proliferating cells. (A and B) Representative images show dorsal views of an embryo injected with *chd8* MO (A) and a sham-injected control (B). (C) Quantification of macrocephaly was performed in embryo batches by measuring the distance across the convex tip of the eye cups (yellow arrows) at 4.5 dpf ($n = 70$ embryos; repeated three times). The macrocephaly phenotype represents a 12% increase compared with controls. *** $P < 0.0001$ (Student *t* test). (D and E) Suppression of *chd8* leads to increased ectopic expression of HuC/D at 2 dpf. Representative images (with HuC/D-antibody staining) show the ventral views of an embryo injected with *chd8* MO and a sham-injected control. HuC/D levels in the anterior forebrain of the embryos injected with the *chd8* MO are significantly higher than in controls. (F) Percentage of embryos with normal (black) or ectopic (red) HuC/D protein levels in the anterior forebrain in embryo batches injected with *chd8* MO exhibit an ectopic expression of HuC/D compared with controls. (G and H) Phospho-histone H3 staining for proliferating cells in the zebrafish brain at 2 dpf. Representative images (with p-histone H3-antibody staining) show the lateral views of an embryo injected with *chd8* MO and a sham-injected control. (I) Quantification of p-histone H3-positive cells from control embryos or embryos injected with *chd8* MO ($n = 20$ embryos per group). Data are presented as the mean \pm SEM. ** $P = 0.0018$ (two-tailed *t* test comparisons between MO-injected and controls).

lymphoblastoid cell lines (25, 35). These studies have revealed coexpression modules containing multiple ASD genes and provided insights into developmental timing and regional specificity of transcriptional coexpression signatures (25, 26, 35). However, such correlation studies are not designed to examine directly the cause-and-effect relationships involving ASD mutations and their consequences. Therefore, we sought to examine the role of *CHD8* by establishing the functional genomic effects in human NPCs of reducing its expression to a level comparable to that expected from the heterozygous inactivating mutations seen in ASD.

This study shows that *CHD8* has a broad impact on the regulation of gene expression and reveals an intriguing contrast in the nature of the pathways altered by its suppression based on the directionality and direct/indirect nature of the effect. First, we find that *CHD8* mutation plays an indirect role in down-regulating gene expression in pathways involved in neurodevelopment, supporting a role for chromodomain helicases in neuronal differentiation (36–38). This mechanism connects *CHD8* to many other ASD-associated genes and canonical pathways thought to act in ASD pathogenesis (36–38). Mediators of this regulatory effect could represent high-priority targets for probing probe this mechanism further. Several genes involved in

neurodevelopmental pathways are among the most significantly affected by *CHD8* suppression, most notably *SCN2A*, *DLG2*, *SHANK3*, and a number of cell-adhesion genes (*LAMA4*, *NCAMI*, *MEGF10*) ($P < 1.5 \times 10^{-8}$ for all genes), most of which are down-regulated. These observations suggest that *CHD8* mutation may precipitate abnormal neurodevelopment through its indirect regulatory effect on a network of neurodevelopmental genes, many of which are associated with ASD.

These data also show that the enrichments of various gene sets associated with ASD among genes regulated by *CHD8* are sensitive to the underlying molecular and physiological functions of the genes. ASD-associated genes that are annotated as functioning in neuronal development are enriched more significantly among genes that are indirectly down-regulated following *CHD8* suppression, a finding derived exclusively from the SFARI and AutismKb gene sets, which include genes implicated in a range of genetic and neurobiological studies. Indeed, analysis of the combined SFARI/AutismKb gene set, irrespective of *CHD8* regulation, reveals strongest association with pathways involved in synaptic transmission, cell–cell signaling, neuron differentiation, and neuronal development. On the other hand, the overall gene set defined by the presence of a de novo LoF mutation is associated more significantly with pathways involved in chromatin modification and protein methyltransferase activity, with less significant enrichment for pathways such as axon guidance, and is enriched among genes with *CHD8*-binding sites, regardless of expression. Overall, the nature of the ASD genes discovered to date indicates that pathogenesis may be precipitated by quite different triggers, but our data show that reduced *CHD8* function is a trigger that produces altered expression of a number of other functionally distinct ASD genes, suggesting that these ASD genes may converge on common final pathways. The respective enrichments of *CHD8*-bound genes and down-regulated, unbound genes among the modules of chromatin/transcriptional regulators expressed in early fetal development (modules M2 and M3 in ref. 25) and of synaptic proteins expressed later in development (M13, M16, and M17 in ref. 25) further suggest that *CHD8* may influence abnormal neurodevelopment by both direct and indirect molecular mechanisms. These data indicate that early in fetal development *CHD8* mutation may alter the functioning of transcriptional regulators sensitive to direct regulation by *CHD8*, whereas later in fetal development indirect mechanisms may regulate genes important for synaptic function, perhaps in conjunction with some of the early transcriptional/chromatin regulators.

Finally, we note an enrichment of cancer-associated loci among genes with *CHD8*-binding sites, regardless of whether these genes are differentially expressed in these NPCs. *CHD8* has been implicated in multiple cancers in several studies (11–14), most recently a pan-cancer deep sequencing study from Lawrence et al. (13), but the mechanistic link between *CHD8* and cancer pathways is unclear. There is strong enrichment among all genes with *CHD8* ChIP-determined binding sites for genes in each of the three cancer datasets that we evaluated, indicating that *CHD8* has a direct role in their transcriptional regulation. That most such genes in this NPC system did not show altered expression caused by *CHD8* suppression suggests that cell specificity and other cooperating factors may be important in determining *CHD8* regulation of particular cancer pathways in specific tumor types. Notably, in a study published during review of this paper, Bernier et al. (39) performed extensive phenotyping of 15 individuals harboring truncating *CHD8* mutations and found that both of the subjects that were assessed after the age of 40 years developed tumors, including one subject who was diagnosed with rectum carcinoma at age 42 and died from complications of metastases. Clearly, the impact of *CHD8* on the development of primary tumors and metastatic disease warrants further exploration.

The integration of ChIP-sequencing with transcriptome sequencing enabled a search for binding motifs that propose

cofactors with the potential to act in concert with *CHD8*. Motif analysis implicates YY1, a cofactor of CTCF, CTCF itself, and other factors that may act as coactivators or corepressors with CHD8. However, these analyses have limitations, because the cofactors implicated by motif analysis are necessarily restricted by the available database of known motifs, and in many cases they do not distinguish between family members that recognize similar sequences. Also, this study was performed in a single cell type, albeit one that is highly relevant to neural development. Additional studies performed directly on relevant brain tissue and on peripheral cell types would be of interest, and such studies are in progress. For example, studies of differentiating neurons or cells of origin of specific tumor types are likely to reveal additional functions of CHD8, including differences in gene expression that reflect cell type and stage-specific direct regulatory effects and consequent differences in the networks of indirect effects. In particular, investigation of other cell types could elucidate the regulatory function of CHD8 for the many disease-associated genes with CHD8-binding sites that did not show altered expression in NPCs as the result of CHD8 suppression.

In conclusion, these studies identify a strong association between *CHD8* and ASD pathogenesis and also support a role for the gene in cancer formation through a distinct set of genes. The connection uncovered between *CHD8* and a network of diverse ASD-associated genes supports the hope that targeting therapeutic intervention to a limited number of shared pathways of pathogenesis eventually could provide effective treatment for ASD individuals with quite different genetic defects. It also points to the need to identify the direct target(s) of CHD8 that mediates this indirect effect as one or more additional players in the ASD transcriptional network.

Materials and Methods

Cell Culture, Viral Transduction, and Stable Cell Line Generation. Human control NPCs GM8330-8 were kindly provided by S.J.H. They originated from a control patient (not an affected individual) and were derived from iPSC clones through a neuronal differentiation protocol as described in ref. 15. High-efficiency pLKO.1 HIV-based lentiviral vectors carrying six different shRNAs targeting *CHD8* (Dataset S2A) as well as against nontargeting controls (Sh against GFP and LacZ) were developed by the RNAi consortium (TRC-Hs1.0, Human) at the Broad Institute (Cambridge, MA). Further details are provided in *SI Materials and Methods*.

Functional Genomic Studies. ChIP was performed as previously described (40) in control NPCs infected with the GFP hairpin. Three independent anti-CHD8 antibodies were used (Novus Biological NB100-60417 and NB100-60418 and Bethyl A301-224A) (see Fig. 1B for epitope location). Complexes were precipitated with Dynabeads Protein A beads (Invitrogen), and immunoprecipitated chromatin was eluted in elution buffer de-crosslinked at 65 °C for 8 h (or overnight) and treated with proteinase K (Roche). DNA was purified by extracting with phenol and chloroform and precipitating in ethanol, followed by library preparation for Illumina HiSeq 2000 sequencing. For RNA-seq, libraries were prepared using a customized version of the originally published, strand-specific dUTP method (41, 42). Libraries were generated for all shRNA knockdowns and controls. Libraries were multiplexed, pooled, and sequenced on multiple lanes of an Illumina HiSeq 2000 to a targeted depth generating an average of 40 M paired-end 50-cycle reads for each sample (average final depth ~45 M total reads). See *SI Materials and Methods* for detailed procedures.

Computational Analyses and Statistical Methods. RNA-seq data were aligned to the human genome (GrCh37, Ensembl build 71) using Gsnap (43) version 2012-07-207. Only reads with unique alignments were retained, and only genes that met the threshold for detection in all the samples were included in analyses [more than three reads, as determined by analysis of External RNA Controls Consortium (ERCC) spike-ins as described in ref. 16] (Fig. S2 B and C), resulting in 15,903 genes. For differential expression analysis, sh1

was excluded because of the low level of knockdown of *CHD8*. Differential expression analysis was carried out using a two-factor model that incorporated batch effects using differential expression sequencing (DESeq) (17) version 1.12.1 (*SI Materials and Methods* and Fig. S8 B and C). Functional and pathway enrichments were assessed using DAVID (44) and ToppGene (45). Only functional/pathway enrichments meeting a false-discovery rate (FDR) < 5% (DAVID) or a Bonferroni-corrected *P* value < 0.05 (ToppGene) are presented. Disease gene-set enrichments were assessed using Fisher's exact test, and all results, including Benjamini-Hochberg (46) corrected *P* values, are provided in Dataset S8A. The complete disease gene sets used are described in *SI Materials and Methods*. ChIP-seq libraries were aligned to GrCh37 (Ensembl build 71) using Burrows-Wheeler Alignment (BWA) version 0.7.5a (47). CHD8-binding peaks were detected separately for each antibody using MACS2 (version 2.0.10.2013.9.13) (48) with a cutoff of a Benjamini-Hochberg corrected *P* value < 0.05. We used only peaks that were detected by at all three antibodies (7,324 peaks; Fig. S5B). We used CEAS (49) to obtain genomic distributions of peaks relative to the hg19 refGene track, as shown in Fig. 3, and GREAT (22) to map peaks to genes, allowing peaks up to 10 kb from a gene's TSS in either direction to be mapped to that gene, and to determine enriched GO terms and pathways (Dataset S4). We obtained genome segmentations by chromatin state, based on five histone modifications (H3K4me3, H3K4me1, H3K36me3, H3K9me3, and H3K27me3), for an ES cell-derived neural progenitor line from the NIH Roadmap Epigenomics consortium (<http://nihroadmap.nih.gov/epigenomics/>) (21). De novo motif analysis was performed using Homer version 4.2 (50) on the peak list for each antibody separately, which obtained the nine replicated motifs listed in Dataset S5. The full peak length was used, and sequences were masked for repetitive sequences.

WGCNA (18) was performed using signed correlation on all 15,903 genes. Module-trait relationships (correlation between the module eigengene and the trait of interest) and gene significance (correlation between gene expression and the trait of interest) were computed for each module and each gene, respectively, using *CHD8* expression level as the trait. All analyses are provided in *SI Materials and Methods*, and all 21 modules are shown in Fig. S3. PPI network analyses were performed for differentially expressed genes using DAPPLE (51), which assesses significant interactions for given genes based on permutation statistics over the manually curated InWeb database (52). To visualize the network we used Cytoscape 3.0.2 (53) and subnetworks generated using the reactomeFI plugin (54). Based on the topological overlap matrix (TOM), 699 genes were involved in the top 0.5% of coexpression interactions. The PPI subnetwork with the largest overlap with CHD8-coexpression hub genes, defined as genes in the top 10% of genes in each of the four *CHD8*-correlated modules by intramodular connectivity, is shown in Fig. S4B.

Morpholino, Immunostaining, and Embryo Manipulations. Zebrafish embryos were raised and maintained as previously described (55). Splice-blocking MOs against *chd8* (*chd8*-MO1, 5'-GAGAATGGAATCATAACTTACTTGA-3', and *chd8*-MO2, 5'-GCAAATGTGCAAGCAAGTAACACCT-3') were obtained from Gene Tools, LLC. We injected 10 ng of *chd8*-MO1 and *chd8*-MO2 into wild-type zebrafish embryos at the one- to two-cell stage. Suppression of endogenous message was shown by PCR amplification of cDNA reverse transcribed from extracted total mRNA (primers are available upon request). All experiments shown in this study were performed using *chd8*-MO1 and replicated with *chd8*-MO2. Injected embryos were either fixed at 2 dpf for immunostaining or fixed at 4.5 dpf for head-size measurement; the distance across the convex tips of the eye cups was measured and compared with an age-matched control group from the same clutch. Further details on methods in zebrafish, including whole-mount immunostaining and characterization of the *chd8* transcript by RNA-seq, are given in *SI Materials and Methods*.

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