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Transcriptional determinism and stochasticity contribute to the complexity of autism-associated *SHANK* family genes

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SUMMARY

Precision of transcription is critical because transcriptional dysregulation is disease causing. Traditional methods of transcriptional profiling are inadequate to elucidate the full spectrum of

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AUTHOR CONTRIBUTIONS

X.L. and Y.-h.J. conceived and designed the project. X.L. performed most of the data collection and data analysis. P.S.-M., Y.M., Y.W., and A.Q.-H. prepared and processed human brain tissues. G.W. assisted the long-read sequencing production. M.G. and P.N. assisted in the data analysis. X.L. and Y.-h.J. wrote the manuscript together with all co-authors.

DECLARATION OF INTERESTS

Y.-h.J. is a scientific co-founder of Couragene, Inc., but this study is unrelated to his role. The project was supported initially by a sponsored research project by Taysha Gene Therapies. Taysha Gene Therapies did not have any direct role in the conceptualization, design, data collection, analysis, decision to publish, or preparation of the manuscript.

SUPPLEMENTAL INFORMATION

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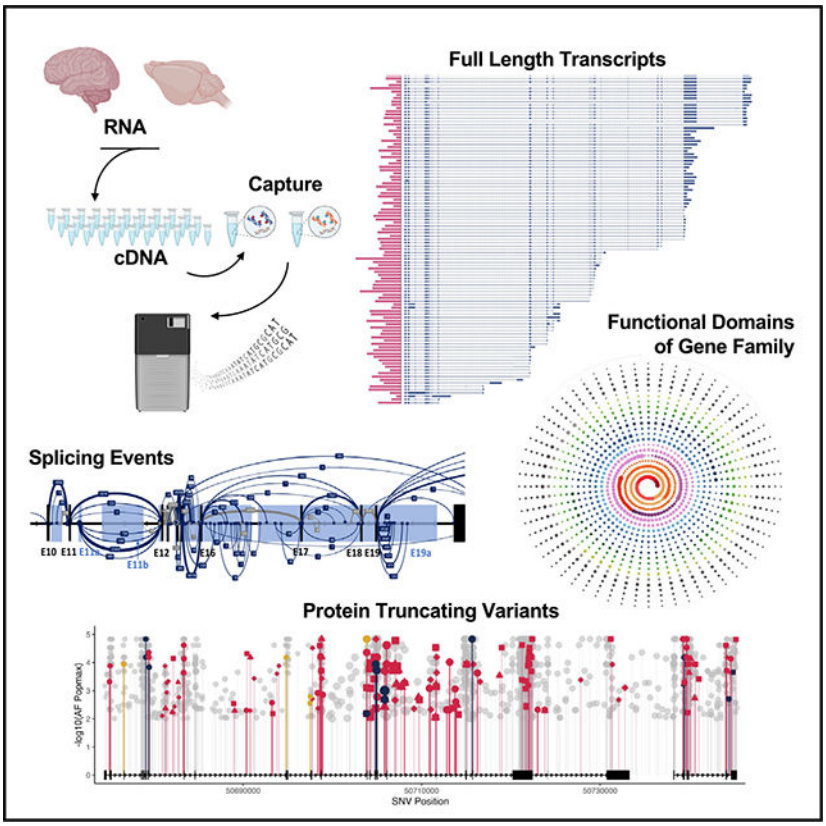
the transcriptome, particularly for longer and less abundant mRNAs. *SHANK3* is one of the most common autism causative genes. Twenty-four *Shank3*-mutant animal lines have been developed for autism modeling. However, their preclinical validity has been questioned due to incomplete *Shank3* transcript structure. We apply an integrative approach combining cDNA-capture and long-read sequencing to profile the *SHANK3* transcriptome in humans and mice. We unexpectedly discover an extremely complex *SHANK3* transcriptome. Specific *SHANK3* transcripts are altered in *Shank3*-mutant mice and postmortem brain tissues from individuals with autism spectrum disorder. The enhanced *SHANK3* transcriptome significantly improves the detection rate for potential deleterious variants from genomics studies of neuropsychiatric disorders. Our findings suggest that both deterministic and stochastic transcription of the genome is associated with *SHANK* family genes.

In brief

Lu et al. revealed the sophisticated transcriptional landscape of SHANK family genes with an integrative approach combining targeted capture and long-read sequencing. The findings elucidated a full spectrum of transcriptome, illustrated the deterministic and potentially stochastic nature of transcription, and sharpened the profiling of autism-related genetic variations.

Graphical Abstract



INTRODUCTION

In the central dogma of molecular biology, RNA transcription acts as a rheostat, orchestrating the cellular functions of the genes in response to intrinsic and extrinsic signals. The complex functions in the organs, such as the brain, require a diverse proteome from a relatively small gene pool. This diversity is facilitated by transcriptional regulation involving alternative promoter usage and splicing occurring in >90% of neuronal genes in mammalian brains.[1–4] Disruption of transcript-specific regulatory elements due to DNA mutations can lead to diseases. Transcriptome-wide changes are implicated in neuropsychiatric conditions, including autism spectrum disorder (ASD).[5–9] Accurate annotation and interpretation of these changes relies on a comprehensive transcriptomic profile, either for a given gene or on a genome-wide scale. However, popular short-read sequencing is suboptimal for delineating longer transcripts and discovering novel exons and splicing events.[10] Standard long-read sequencing techniques are not sufficiently sensitive to detect transcripts with lower abundance. A theoretical solution lies in the combination of mRNA/cDNA-capture methods[11] and long-read sequencing, which could identify both long and low-abundance transcripts. However, this approach has been sparingly reported, probably due to the technical challenge of preserving the mRNA integrity. The current inability to construct a complete transcriptome fuels a continuing debate over the extent of pervasive transcription across the genome and the significance of endogenous transcriptional "dark matter."[12–16] The incomplete transcriptome impedes accurate annotation of disease-linked variants and interpretation of transcriptomic data. This shortfall affects the validation of genetically modified disease models used in preclinical research to develop molecular therapies. Previous studies have indicated specific functions of *SHANK3* mRNA transcripts at synapses.[17–30] An incomplete human *SHANK3* transcriptome could underestimate the contribution of the genetic risk for ASD and other neuropsychiatric disorders. Similarly, the incomplete mouse transcriptome complicates the interpretation of its relevance to human *SHANK3* disorders from studies of more than 24 lines of genetically modified animal models.[19,31,32] To bridge these substantial gaps in knowledge, we performed standard Iso-Seq (SIS) for whole-transcriptome analysis and paired it with targeted cDNA capture and long-read sequencing techniques (capture-Iso-Seq, CIS) to specifically investigate the *SHANK* family genes in human and mouse brain. We discovered a drastically intricate *SHANK3* transcript structure and a broad transcriptomic diversity across the human and mouse genomes. We identified unexpected extensive fusion transcripts and atypical patterns of transcripts in *Shank3*-mutant mice. The enhanced *SHANK3* transcriptome has significantly improved the discovery rate of deleterious variants in genomic and transcriptomic studies of neuropsychiatric disorders. Our study advocates for a paradigm shift in experimental design and evaluation of genetic disease models using genetically modified animals, emphasizing the need to carefully evaluate the molecular validity of these mutant animal models in preclinical research.

RESULTS

Dataset overview and experimental strategy evolution and optimization

We sequenced 56 single-molecule real-time (SMRT) libraries of human and mouse brains using the PacBio Sequel II System (Figures 1A and 1B). Sixteen libraries proceeded using the SIS method. Forty libraries were constructed following the CIS method, which employed targeted capture enrichment with specific oligonucleotide probe panels that covered the full genomic regions of *SHANK/Shank* family genes (*SHANK1–3*, Tables S1 and S2). A non-neuronal gene, *TP53*, was included as a comparison. Twenty libraries were synthesized from cerebral cortex of neurotypical children ages 5–6 years and young adults ages 24–30 years. For mice, 35 libraries were derived from striatum (ST) and prefrontal cortices (PFCs) of 21-day-old wild-type (WT) C57BL/6J and *Shank3* mutants (*Shank3*$^{\Delta e4-9}$, *Shank3*$^{\Delta e21}$, and *Shank3*$^{\Delta e4-22}$).[20,21,33–35] We processed only the RNA with an integrity number (RIN) above 7 for human and above 8 for mouse samples for subsequent sequencing. The quality and reproducibility of the SIS and CIS platforms were optimized (Figures S1A–S1I). For experimental validation, RT-PCR and Sanger sequencing was used to confirm novel *SHANK3* transcripts from CIS. We performed *in silico* transcriptome analyses using short-read bulk RNA sequencing (srRNA-seq) and single-cell RNA-seq (scRNA-seq) data and gene discovery analyses of exome sequencing (ES) and whole-genome sequencing (WGS) data from the PsychENCODE project along with other genomics studies.[5,36–40]

SIS uncovered more diverse transcriptomes genome-wide in mouse and human brains

From the SIS of 12 SMRT libraries of the human brain, we uncovered 131,585 unique transcripts across 15,308 annotated genes, including 311 novel transcripts (UCSC Track 1). The distribution of unique transcripts and sequencing reads per gene are shown in Figure 1C. The number of unique transcripts for a given gene was significantly correlated (Pearson r = 0.8871, $p < 0.001$) with its abundance (Figure 1D). From four SISs of mouse ST and PFC, we uncovered 154,492 unique transcripts from 16,556 annotated genes, with 1,570 being novel (Figure 1E and UCSC Tracks 2 and 4).

In human brains, the average number of isoforms per gene was 19, with an average sequence read count of 63. Notably, 595 genes exhibited over 100 isoforms (Figure 1E; Table S3a). *SEPTIN4* has the highest number of isoforms at 692; it is a gene encoding a presynaptic scaffold and GTP-binding protein involved in exocytosis and which interacts with alpha-synuclein, implicated in Parkinson's disease.[41] In mouse brains, the average number of unique transcripts per gene was found to be 8, with an average of 17 sequence reads per transcript. *Sorbs1* had the highest number of isoforms at 158; this gene encodes a Sorbin and Src homology 3 (SH3) domain-containing protein involved in insulin signaling and stimulation[42] (Table S3b). We identified 182 genes with more than 50 isoforms and 19 genes with over 100 isoforms in mouse brains. Of these, 7 have human orthologs that also exhibit more than 100 isoforms. Our studies revealed a greater transcript diversity than other studies using the same sequencing platform and analytic algorithm.[8,43] We examined the transcript diversity of 213 highly confident ASD-risk genes consolidated from three recent extensive ASD genomics studies using our SIS data[44–46] (Figure 1F; Table S4). On average, individual ASD-risk genes exhibited 56 transcripts, with a 90% confidence interval

(CI) in the range 8–140. *ANK2* was noted for having the highest number of transcripts at 372. Remarkably, the expression level of *SHANK3* was one of the lowest, ranking 212 of 213 ASD-risk genes (Figure 1F). Genes associated with brain disorders, especially ASD and neurodevelopmental disorders (NDDs), have significantly greater numbers of transcripts compared to genes implicated in disorders not related to the brain (Figures 1G and 1H).

A complex mouse *Shank3* transcriptome from CIS

We noted that the longest annotated *SHANK3/Shank3* transcripts in humans (NM_001372044.2, 7,691 bp, hg38) and mice (NM_021423.4, 7,380 bp, mm39) have not been detected in any published long-read RNA-seq datasets.[6,8,43] From four SISs of mouse ST and PFC, we identified only five *Shank3* transcripts (ranging from 5,625 to 6,463 bp) in ST, with none detected in PFC upon validation. The discrepancy in transcript number and the variation between ST and PFC were consistent with the highest expression level of *Shank3* in ST and lower expression in neocortex at P21 days.[26] The failure to detect longer *Shank3* mRNAs by SIS was most likely due to their low abundance, as transcripts up to 14.5 kb were successfully sequenced in our libraries (Figures S1F and S1G).

With CIS, we detected 545 *Shank3* transcripts in the mouse ST (Figure 2A) and 345 in PFC (Figure 3A), including the longest annotated transcript (NM_021423.4). We successfully validated 51 (85%) of 60 representative novel transcripts by RT-PCR and sequencing (Figures 2E–2H; Table S5). To evaluate the quality of each transcript, we employed a confidence metric that integrates the transcript abundance, the length of predicted open reading frame (ORF), and validation with srRNA-seq data (Figure S2A). In ST, 223 (41%) of *Shank3* transcripts were classified as high confidence, while 382 (59%) were in moderate confidence. In PFC, 168 (49%) transcripts were in high confidence, with the remaining 176 (51%) of moderate confidence. Analysis revealed 36 and 26 potential transcription start sites (TSSs) in ST and PFC, respectively. In the ST, 142 *Shank3* transcripts originated at exon 1 of the annotated referenced transcript (NM_021423.4) and terminated at 26 different sites (Figure 2B). Thirty-five transcripts terminated within exon 21, each presenting a different ORF. Exon 21, the largest coding exon of 2,257 bp, was spliced out in many transcripts. Over 90% of transcripts terminated within 100–500 bp of an annotated transcription termination site (TTS) and poly(A) signal (Figure S3). This indicates that the early terminations are not artifacts of RNA degradation or cDNA synthesis errors. Intron retentions were observed in introns 1, 2, 11, 12, and 19, leading to altered ORFs and earlier stop codons. While some transcript structure variations were subtle, they are predicted to encode different ORFs (Figures 2C and 2D).

In the PFC, we identified 59 *Shank3* transcripts initiating from 19 different exons and terminating within the last coding exon 22 (Figures 3A and 3B). Notably, 28 of these transcripts started within exon 21 with different ATG codons. This finding aligns with our prior results obtained from 5′ rapid amplification of cDNA ends (RACE) experiments.[26] We discovered 12 new exons in ST and 17 in PFC, with 11 being shared in both regions (Figures 2A and 3A). In addition, we discovered 4 new untranslated exons, U1–4, located 5′ upstream of the annotated *Shank3* exon 1 (Figure 2A). Six new and alternative spliced exons, E9a–f, were identified between exons 9 and 10. The spliced variants between exons 9

and 10 were the most abundant, with 4,326 reads in ST and 641 in PFC, while exon 12e was exclusive to the PFC (Figure 2E).

Surprisingly, we observed a considerable number of novel fusion transcripts, in which different *Shank3* exons were joined to downstream exons 2–5 of the *Acr* gene, which encodes the acrosin protein in the acrosome of spermatozoa[47] (Figures 2A and 3A). These fusion transcripts were validated by PCR and sequencing (Figures 2E and 2F). We noted that splice events linking *Shank3* exons 17 and 21 to *Acr* exons occurred more frequently than others. Specifically, fusions from *Shank3* exon 21 to *Acr* exon 2 (208 reads) and exon 3 (243 reads) were the most abundant. Western blot using antibody against the C-terminus epitope of ACR detected significantly increased SHANK3-ACR fusion protein in $Shank3^{\Delta e4-22-/-}$ mutant mice compared to WT (Figure 2I). The protein bands of SHANK3-ACR fusion product no. 1 and SHANK3-ACR fusion product no. 2 were extracted for liquid chromatography-tandem mass spectrometry (LC-MS/MS) protein identification analysis. The peptide sequences identified from LC-MS/MS match to the protein sequence encoded by exon 2 or exon 3 of *SHANK3*. We also identified splice products from *Shank3* exons 17 and 21 to three novel exons/transcripts (T1–3) situated downstream of *Acr* (Figure 2A). These transcripts in ST and PFC are predicted to yield five ORFs, extending the SHANK3 protein by an additional 64 aa (NP_001358973).

The transcriptomic architecture of *Shank3* revealed by CIS in ST and PFC displayed both shared and unique characteristics. Overall, 230 transcripts (42% of ST, 67% of PFC) were common to both brain regions (Figure 3C). We analyzed the tissue-specific usage of TSSs and coding sequence starting sites (CDSs). Transcripts were categorized as follows: overlapping with the annotated *Shank3* mRNA, U1–4 to *Shank3*, *Shank3-Acr* fusion, and *Shank3*-T1–3. In ST, 75% of transcripts belonged to the category overlapping with the annotated *Shank3*, and 24% fell within the *Shank3-Acr* fusion category (Figure 3E). In PFC, 52% of the transcripts were overlapping with annotated *Shank3*, while 43% were classified as *Shank3-Acr* fusion transcripts (Figure 3F).

Protein-domain-specific mouse SHANK3 proteome

SHANK3 and its family encode proteins possessing six domains: ubiquitin-like (Ubl), ankyrin repeats (ANKYR), postsynaptic density (PSD) protein 95/discs large homolog 1/zonula occludens 1 (PDZ), SH3, a proline-rich region containing Homer and Cortactin-binding sites (Pro), and a sterile alpha motif (SAM).[48–50] As a scaffold protein in the PSD of synapses, SHANK3 protein interacts with various synaptic proteins via these domains, contributing to synaptic architecture and function. There are 474 ORFs predicted from 545 *Shank3* transcripts in ST and 270 ORFs in PFC using GeneMarkS-T,[51] with 261 ORFs being common to both brain regions (Figure 3D). ORFs of novel transcripts were further corroborated by proteome data derived from various *in silico* datasets, utilizing graded criteria for sequence identity and overlap (Figure S2B).

Among the 125 ORFs predicted from 140 *Shank3* transcripts starting from exon 1 in ST, only 4 encompassed all six protein domains (Figure 3G). Among the 270 ORFs predicted from 345 *Shank3* transcripts in PFC, only 1 contained the complete set of six protein domains, while 37 ORFs had more than three protein domains (Figures 3H–3K). One

hundred nineteen SHANK3 ORFs (30%) in PFC comprised only a single protein domain, typically the Pro domain. Approximately 15% of the predicted ORFs lacked recognized protein domains. The protein domain combinations were found to be non-random and tissue specific; for instance, no predicted ORFs included the SAM-SH3 combination. The SAM-Pro-SH3 and SAM-SH3-ANKYR domain combinations were exclusive to PFC, while the Ubl-ANKYR-Pro-SAM and ANKYR-SH3-PDZ-Pro combinations were identified only in ST (Figure 3L).

Uniquely altered *Shank3* transcriptome in *Shank3*-mutant mice

Sixteen *Shank3*-mutant mouse lines and eight mutant rat, dog, and non-human primate lines featuring various exonic deletions or point mutations have been generated to model *SHANK3*-associated ASD[31] (Figure 4A). Using the same *Shank3* probe design, we conducted CIS on *Shank3*-mutant mice: those with deletions of exons 4–9 (*Shank3*$^{\Delta e4-9}$), exons 4–22 (*Shank3*$^{\Delta e4-22}$), and exon 21 (*Shank3*$^{\Delta e21}$).[20,21,33,34] In *Shank3*$^{\Delta e4-9}$ homozygous mice, we detected 69 *Shank3* transcripts in ST and 56 in PFC. Representative mutant and residual transcripts are diagrammed in Figure 4B, with details provided in Figures S4A and S4B. In ST and PFC of *Shank3*$^{\Delta e4-9}$ mice, we identified three long transcripts (~7.3 kb), harboring a deletion of exons 4–9. Interestingly, the first exon of these transcripts, with the exon 4–9 deletion, was in intron 1 of the annotated *Shank3*, a TSS not utilized in WT mice, suggesting an alternative TSS due to the exon 4–9 deletion. These transcripts also lacked coding exon 22 and exhibited fusions between exon 21 of *Shank3* and exon 2 of *Acr*. ORF prediction suggests that the resultant SHANK3-ACR fusion proteins for these mutant transcripts are 1,254 aa for PB.6361.147, 1,073 aa for PB.6623.114, and 833 aa for PB.6623.199. Approximately 70% of the residual transcripts are initiated from intron 16/exon 17 and terminate within exon 21/intron 21 of *Shank3* or exon 5 of *Acr*. Transcripts starting at exon 11 were exclusively detected in ST. The proportion of transcripts initiated from intron 16/exon 17 was increased in *Shank3*$^{\Delta e4-9}$ mice compared to WT. A total of 54 ORFs (ranging from 113 to 1,327 aa) were predicted in ST, with a similar pattern observed in PFC from residual transcripts of *Shank3*$^{\Delta e4-9}$ mice.

In *Shank3*$^{\Delta e21}$ homozygous mice, we identified 401 *Shank3* transcripts in ST and 148 in PFC (Figures 4C, S4C, and S4D). In *Shank3*$^{\Delta e4-22}$ homozygous mice, the numbers were 436 in ST and 792 in PFC (Figures 4D, S4E, and S4F). Remarkably, over 99% of these transcripts were *Shank3-ACR* fusion events in both brain regions of *Shank3*$^{\Delta e21}$ and *Shank3*$^{\Delta e4-22}$ mice. The predominant transcripts in *Shank3*$^{\Delta e21}$ mice were from the intron 16/exon 17 region in both ST and PFC. Conversely, in *Shank3*$^{\Delta e4-22}$ mice, transcription primarily initiated from intron 1/exon 2. We also detected multiple novel exons interposed between *Shank3* and *Acr* genes (Figures S4C* and S4D*), exclusive to these *Shank3* mutant lines and absent in WT. Fusion transcripts of *Shank3-Acr* were more prevalent in *Shank3*$^{\Delta e4-22}$, *Shank3*$^{\Delta e4-9}$, and *Shank3*$^{\Delta e21}$ mutants. Moreover, a significant overexpression of *Acr* transcripts was found in neocortex and hippocampus of *Shank3*$^{\Delta e4-22-/-}$ mice (Figures 4E and 4F). Bulk RNA-seq data analysis from ST of *Shank3*$^{\Delta e4-22-/-}$ mice also indicated a compensatory expression from *Shank1* and *Shank2*, which was protein domain specific (Figures 4H–4K).

The *Shank3* transcriptomic findings from CIS prompted us to extend our approach to include all *Shank* family genes (*Shank1–3*) using a joint capture strategy. This joint CIS for the *Shank* family genes identified 664 *Shank1* and 495 *Shank2* transcripts in PFC and 320 *Shank1* and 326 *Shank2* transcripts in ST (UCSC Tracks 4 and 5). The overall transcript structures and patterns of *Shank3* from both single-gene and joint CIS were similar. We discovered seven novel exons upstream of the annotated exon 1 of *SHANK1* (Figure S5A). Fusion transcripts involving *Shank1* and *Shank2* with adjacent genes were also detected. The most upstream novel exon of *Shank1* overlapped with the last exon of the *Clec11a* gene (NM_009131.3), which is transcribed in the reverse direction relative to *Shank1* (Figure S5B). The fusion transcripts between *Shank1* and *Josd2*, a gene located approximately 100 kb downstream, were exclusively detected in PFC. Two new untranslated exons, U1 and U2, were found about 24 kb upstream of the annotated 5′ exon 1 of *SHANK2* (Figure S5C).

Transcript diversity of *SHANK* family genes in human brains

In the current reference genome (hg38), an annotated human *SHANK3* mRNA (7,691 bp, NM_001372044) is displayed, yet it has not been experimentally validated. With CIS on *SHANK* family genes, we discovered 472 unique *SHANK3* transcripts (Figures 5A–5C, UCSC Track 6), with the longest being 6,824 bp. Notably, the annotated 7,691 bp *SHANK3* transcript (NM_001372044) was absent. The absence of the longest *SHANK3* transcript is unlikely to be a result of RNA degradation, because a 10.8 kb *SHANK2* transcript was detected in the same captured sample, which is much longer than the 3 kb peak in the study by Shimada et al.[52] and the 1 kb in that of Yang et al.[9] Instead, it appeared to be due to extremely low or no expression of the full-length *SHANK3* transcript in adult frontal and temporal cortices. Most of the 472 unique *SHANK3* transcripts clustered within regions spanning exons 1–9 and 10–22. None incorporated splicing between exons 9 and 10, a region characterized by high GC content (77% of GC) and a CpG island (hg38). The failure to detect exon 9–10 transcripts was probably not due to high GC content, because we could not detect them by regular RT-PCR using DNA polymerase that has been optimized for efficiently amplifying up to 90% of GC-rich templates. Similarly, the failure to detect the exon 9–10 splicing transcripts was not likely due to RNA quality, because the RIN for these tissues was between 7 and 9.5. We noted that 43 unique transcripts initiated from this CpG island, implying a TSS within intron 9. *In silico* analysis using a parameter-free assembly approach (Cufflinks-Cuffmerge)[53] applied to srRNA-seq data and a published region-specific long-read transcriptome profiling on different postmortem human brain regions[52] also failed to detect any transcripts connecting exons 9 and 10. Taken together, our data support that the lack of exon 9–10 splicing transcripts in these analyzed tissues is a biological phenomenon and less likely to be due to technical reasons.

Similar to mouse *Shank3*, we detected 66 fusion transcripts between *SHANK3* and *ACR* (Figure 5C). These fusion transcripts, intron retention, and novel exons were validated by RT-PCR and sequencing (Figure 5D). Fifty-eight of them were fusion transcripts comprising exon 19/exon 20 of *SHANK3* (exon 20 is the largest exon in human, equivalent to exon 21 in mouse) to exons 2–5 of *ACR*. Nine transcripts that started within *SHANK3* exon 20 were found to be fused with *ACR*. We observed splicing events connecting *SHANK3* exons 19–20 to uncharacterized downstream exons, T1–2, of *ACR*. We also detected three

novel untranslated exons (U1–U3) upstream of exon 1 of *SHANK3* mRNA (Figure 5E). The sequence of U2 is highly conserved in mouse.

With the joint capture for *SHANK* family genes, we detected 86 *SHANK1* and 277 *SHANK2* transcripts (UCSC Track 6), from which 69 ORFs for *SHANK1* and 165 ORFs for *SHANK2* were predicted. Across these *SHANK* family ORFs, we observed 17 different combinations of the six functional domains, with the PDZ domain appearing most frequently (Figure 5F). A complete set of all six functional domains (Ubl, ANKYR, SH3, PDZ, Pro, and SAM) was predicted only in one *SHANK2* transcript.

The unexpected discovery of extensive fusion transcripts between *SHANK3* and *ACR* in human brain tissue led to a comprehensive genome-wide analysis for fusion transcripts in SIS data. We detected 2,265 fusion transcripts (1.7% of the total transcripts) associated with 3,499 genes in the brains of children and adults, with 963 fusion transcripts common to both groups. About 98% of fusion transcripts are between two adjected genes. A small number of fusion transcripts are among three adjacent genes. No fusion transcript is from distant genes or genes from two chromosomes. Gene Ontology enrichment analysis revealed a significant enrichment of fusion transcripts in genes associated with ASD (Figures 5G and 5H).

To access the functional constraint of novel *SHANK3/Shank3* exons in humans and mice identified by CIS, we utilized evolutionary rate profiling (GERP)[54,55] and PhyloP[56] conservation scores. In mice, GERP and PhyloP scores for most *Shank3* novel exons were significantly higher than those of a non-transcribed region, but they were lower than scores for known coding exons in both PFC and ST (Figures 5I and 5J; Tables S6A and S6B). A concordant pattern was observed in human *SHANK3* (Figures 5K and 5L; Tables S6C and S6D). These results suggest that the novel exons of *SHANK3/Shank3* uncovered by CIS are evolutionarily constrained elements, underscoring their potential functional significance.

Transcript diversity and novel transcripts of the *TP53* gene in human and mouse

To examine whether the transcriptional complexity is exclusively associated with synaptic genes, we applied SIS and CIS to *TP53* in human brain and to *Trp53* in mouse brain and thymus, where *Trp53* expression is the highest. SIS detected only five *Trp53* transcripts in mouse ST and three in mouse PFC, which is consistent with the data in the literature,[57,58]. In contrast, CIS identified a comprehensive set of 243 transcripts from thymus, 164 from PFC, and 188 from ST (Figures S6A–S6C, UCSC Track 7). The patterns of unique *Trp53* transcripts are similar among the three tissues, with 18 alternative TSSs deduced from thymus transcripts. A significantly higher percentage of transcripts exhibited intron retention in *Trp53* compared to *Shank3*. In addition, novel tissue-specific 5′ exons unique to brain (bU1) and thymus (tU1/tU2) were discovered.

In human brain, CIS detected 106 *TP53* transcripts, which predicted 60 ORFs, 18 TSSs, and three 3′ transcriptional ends (Figure S6D, UCSC tracks). We also discovered three novel exons (hT1–3) at the 3′ end, which extended the C terminus of the TP53 ORF by 72 aa and is conserved with the mouse *TRP53* (77% identical). These observations underscore the diversity of the *TP53/Trp53* transcriptome, which is complex but less heterogeneous than that of *SHANK* family genes.

Developmental, tissue, and cell-type specificity of *SHANK3*/*Shank3* transcripts from CIS

To investigate the developmental specificity of *Shank3* transcriptome, we aligned mouse srRNA-seq data of the cerebral cortex at different ages from day E14.5 to P180[59–61] to *Shank3* transcripts from CIS (Figure 6A). The E14.5 embryos exhibited the least diversity of *Shank3* transcripts. As development progressed, the number of unique *Shank3* transcripts increased, reaching a maximum at day P56 before declining at day P180. Further analysis on cell-type specificity aligning scRNA-seq data from the anterior cingulate area (ACA) of 8-week-old mice[40] to *Shank3* transcripts identified by CIS demonstrated a significantly higher abundance of *Shank3* transcripts in glutamatergic neurons compared to GABAergic neurons. The *Shank3* transcripts including exon 18 were exclusively found in endothelial cells (Figure 6B).

To investigate tissue specificity, we analyzed the exon usage in mouse *Shank3* transcripts from CIS against scRNA-seq data from five cerebral cortex subregions.[40] The exon usage patterns of *Shank3* CIS transcripts within the same cell type exhibited unique variations across different brain subregions (Figure S7). This tissue-specific exon usage was also observed in other long-read transcriptome sequencing and scRNA-seq studies.[52,62] The pattern of human *SHANK3* transcripts in infants and children was distinct from that of adults when we aligned human srRNA-seq data to *SHANK3* transcripts from CIS. The *SHANK3* exon usage also changed with age.

We mapped *Shank3* transcripts to 103 Genomics Visium spatial transcriptome of the mouse to visualize the expression pattern *in situ.*[63] Two probes targeting *Shank3* exons 11 and 22, and one for *Acr* exon 5, facilitated this analysis. Three *Shank3* transcripts identified by CIS were enriched to distinct anatomical regions (Figures 6C–6F). *Shank3* transcript TALONT000202476 containing exon 11 and TALONT000200721 incorporating exon 22 have similar cell-specific expression patterns, albeit at different levels of abundance. Transcript TALONT000200852, a fusion transcript connecting *Shank3* exon 21 and *Acr* exon 5, displayed a cell-type-specific expression pattern. Furthermore, we found a cellular-compartment-specific preference for the *Shank3* transcripts. The inclusion of *Shank3* largest exon 21 is significantly more common in synapses than in nuclei from mouse brain scRNA-seq data[64] (Figure 6G). Exon 2 of the *Acr* gene, frequently fused with *Shank3* exons, was significantly less present in the nucleus of AD models compared to WT. The splicing events involving a 5′ segment of *Acr* exon 5 were more common across both nucleus and synapses in AD mice, while splicing involving the latter 2/3 of *Acr* exon 5 was more frequent in the nucleus of WT (Figure 6H).

Applications of the *SHANK3* transcriptome from CIS to genome sequencing and transcriptome analyses of ASD and other neuropsychiatric disorders

Human *SHANK3* transcripts identified through CIS exhibit expression patterns that are specific to developmental stages and brain regions, such as the cerebral cortex and cerebellum (Figures 7A–7D and S8). We extended the *in silico* transcriptome diversity analysis to 213 highly confident ASD-risk genes consolidated from three recent extensive ASD genomics studies (Table S4).[44–46] The transcriptome diversity of ASD-risk genes was significantly greater than that of non-ASD-associated genes (Pearson r = 0.386, *p*

< 0.001). Specifically, ASD-risk genes associated with gene expression regulation and neuronal communication showed a significantly higher level of transcriptome complexity compared to genes in other functional categories (Pearson r = 0.825 and Pearson r = 0.793, respectively, both $p < 0.001$). *SHANK3*, consistently reported as one of the top 5 ASD-causing genes in these studies,[44–46] is also implicated in schizophrenia (SCZ),[65] bipolar disorder (BPD),[66] and major depressive disorder (MDD).[67] To investigate alterations in *SHANK3* transcriptomes across these disorders, we analyzed srRNA-seq data from the PsychENCODE project.[36] Principal-component analysis (PCA) revealed unique transcript patterns for each disorder, especially for ASD and SCZ (Figure 7E). The expression of a subset of *SHANK3* transcripts varied across ASD, MDD, BD, SCZ, and controls (Figures 7E–7I). Brain-region- and age-specific expression of *SHANK3* transcripts formed a distinct cluster in PCA (Figure 7J). Exons 12, 15, 20, and 22 of *SHANK3* transcripts in BA7 were significantly more represented in ASD brains than in controls (Figure 7K), and exon 10 showed a higher expression in BA38 of ASD brains (Figure 7L).

While *SHANK3* genetic mutations are implicated in 1%–2% of ASD cases and to a lesser extent in other neuropsychiatric disorders,[44–46,68,69] we sought to examine whether incorporating the enhanced *SHANK3* transcript structure from CIS into publicly available ES and WGS of ASD/SCZ/BPD datasets could uncover additional disease-associated single nucleotide variants (SNVs).[46,70–73] We reanalyzed sequence variants on a large cohort of 177,000 samples of both controls and disease subjects, including ES data from the Autism Sequencing Consortium,[46] BPD Exomes,[71] and SCZ Exome Meta-analysis Consortium[70] as well as WGS of ASD, SCZ, and BP cohorts from BrainVar[72] and BrainGVEX.[73] Variant identifications and annotations were previously based on the mRNA reference NM_001372044.2 and hg38 genome assembly. We used Variant Effect Predictor (VEP; release 107)[74] and Genome Aggregation Database (gnomAD; v.3.1.2)[75] for annotation and filtering, including variants with a population allele frequency of ≤0.01 for protein-truncating variants (PTVs), and excluding missense and synonymous variants for further analysis. SpliceAI[76] and SnpEff[77] were used to analyze splice variants and evaluate the pathogenic potential of stop-loss, stop-gain, and frameshift variants. This reannotation identified 1,530 new SNVs across 55,000 cases pooled from ASD (11,986 ES, 923 WGS), BP (14,210 ES), and SCZ (27,648 ES) cohorts (Figure 7M), resulting in the discovery of 27 stop-loss, 60 stop-gain, 52 frameshift, and 53 splice variants in *SHANK3* considered potentially deleterious or PTVs using CIS annotation in disease subjects but not in controls. This was a marked contrast to the variants analyzed using the current reference (0 stop-loss, 1 stop-gain, 4 frameshift, and 16 splice variants). Accordingly, the detection rate for potential deleterious SNVs of *SHANK3* increased from 1.3% when using the current reference (NM_001372044) to 12.5% when annotated with the *SHANK3* CIS transcripts, highlighting the significance of comprehensive transcriptome annotation in uncovering genetic contributions to neuropsychiatric disorders (Figure 7N).

DISCUSSION

Diverse transcription is crucial for generating proteomic diversity and facilitating complex cellular functions. Precision of transcription is critical because mutations in the transcriptional regulatory DNA elements can cause numerous single-gene disorders. Despite

the recent report of the completed human genome,[78] the transcriptome remains largely uncharted. Our work applying SIS on human and mouse brains discovered unprecedented transcriptome diversity.[8,43] Using the similar SIS protocol, Glinos et al.[8] reported a maximum of 178 isoforms for a single gene, with only five genes exhibiting more than 100 isoforms. Both Glinos et al.[8] and Shimada et al.[52] detected a median of 2 isoforms per gene across various tissues and cell lines. Leung et al.'s study[43] noted a peak of 40 isoforms per gene in the human cortex. Furthermore, Chau et al.[79] assembled an average of 4 isoforms per gene from bulk RNA-seq of human developing brains. Significantly, these studies uncovered only a few incomplete *SHANK3* mRNA isoforms. However, our study identified as many as 692 isoforms for a single gene, with 595 genes having more than 100 isoforms, and an average of 19 isoforms per gene in the human cerebral cortex. Our results suggest that the extent of transcript complexity described in existing literature is significantly underestimated, particularly for genes like *SHANK3*.

Our targeted capture and long-read sequencing have mapped the *SHANK* family transcriptomes in detail, with the majority of novel transcripts likely endogenously expressed. This is supported by our strict identification process, validation through RT-PCR and Sanger sequencing, consistency across experiments and brain regions, and conservation between species. In addition, the specificity of these transcripts was confirmed in *Shank3*-mutant mice. Despite the high confidence, it remains a possibility that a small fraction might not be expressed endogenously. The discovery of a substantial number of fusion transcripts for *SHANK3/Shank3* in our study was unexpected, with a prevalence that surpassed the findings of other studies.[8,43] Until recently, fusion transcripts have been largely investigated in cancer-related studies because of their oncogenic properities.[80,81] Yet, their presence in normal cells has only recently been acknowledged.[8,43,82] Two recent studies using the SIS method[8,43] reported a mere 136 fusion transcripts (0.41% of total transcripts) in human brains. In contrast, our study identified 2,265 fusion transcripts in human brains, constituting 1.7% of total transcripts. Interestingly, these fusion transcripts were found to be particularly more enriched in the human ASD-associated transcriptome.

The enhanced *SHANK3* transcript structure from CIS has significantly increased the detection rate of PTVs or predicted loss-of-function (LOF) variants in ES and WGS data for neuropsychiatric disorders. Further functional validations are warranted to determine the pathogenicity of these new identified PTVs. Our findings highlight the significance of employing fully characterized transcript structures in genomics studies of disease gene discovery. Transcriptional dysregulation in the brain has been implicated in neuropsychiatric disorders.[5,83] By integrating the *SHANK3* transcriptome data from CIS and the transcriptome data from PsychENCODE, we discovered brain-region-specific dysregulation in the *SHANK3* transcriptome associated with ASD and other neuropsychiatric disorders. Notably, brain-region-specific DNA methylation in intragenic CpG islands, which show altered methylation in ASD brains,[52,84,85] suggests that epigenetic changes could be instrumental in *SHANK3* transcript variations. Our findings add the *SHANK* family to the gene families such as *CACNA1C* and *NEUREXIN* that are examples of extreme transcriptional diversity.[86–88]

In *Shank3*-mutant mice, stable transcripts with exonic deletions indicated truncated protein production or upregulated non-mutant isoforms.[31,89] Cryptic promoters, especially within intron 16/exon 17, suggest alternative initiation and potential novel protein isoforms. These could perturb the PSD protein interactome, indicating possible loss and gain of function in *Shank3* mutants. Such complexities question the molecular and phenotypic consistency of *Shank3* mouse models.[17–19,31,90,91] For example, differential behavioral phenotypes and receptor subunit alterations are noted across different mutant lines.[20,31,35] Specifically, *Shank3*$^{\Delta e21}$ mutants show unique upregulation of alternative transcripts and fusion transcripts, diverging behaviorally from *Shank3*$^{\Delta e4-22}$ mutants.[31,92] These molecular nuances challenge the translational fidelity of *Shank3* mouse models for preclinical studies and necessitate reevaluation, particularly for models in therapeutic development.

Our study's detailed alignment of *SHANK3/Shank3* transcripts underscores its proteomic diversity at the PSD, essential for complex synaptic functions.[49,50,93] However, about 15% of the transcripts, possibly arising from cryptic promoters or alternative splicing, lack substantial ORFs or are poorly expressed, hinting at stochastic transcription events previously noted in other species.[94–103] Challenges to the ENCODE projects' findings on genome transcription by subsequent short-read RNA-seq studies[12–14,16,104–106] align with our discovery that *SHANK3/Shank3* and *TP53* transcription involves intragenic promoters and frequent intron retention. These regions, less conserved evolutionarily, affirm pervasive transcription and suggest a more deterministic transcriptional landscape for these genes in humans and mice.

Limitations of the study

Several limitations of the study warrant discussion. We will not be able quantify the extent of stochasticity of transcription from the current analysis. The extensive functional validation of transcripts at the protein level remains a challenge, as some transcripts may function uniquely at the RNA level, eluding protein-interaction analyses. Also, our capture-based method trades sensitivity for efficiency when scaling up, as increased gene targets reduce sequence depth, necessitating careful experimental design for quality data.

STAR★METHODS

RESOURCE AVAILABILITY

Lead contact—Further information and requests for resources and reagents should be directed to and will be fulfilled by the lead contact, Yong-Hui Jiang (yong-hui.jiang@yale.edu).

Materials availability—Oligonucleotide probe panels were synthesized by Integrated DNA Technologies (IDT). The probe coverage and design are provided in Tables S1 and S2.

Data and code availability

- Both human and mouse raw sequencing data have been deposited at SRA under BioProject: PRJNA1066952 and are publicly available. Accession numbers are

listed in the key resources table. All UCSC tracks described in manuscript, and raw Tandem Mass Spectrometry data have been deposited at Mendeley and are publicly available. The DOI is listed in the key resources table.

- This paper does not report original code.
- Any additional information required to reanalyze the data reported in this paper is available from the lead contact upon request.

EXPERIMENTAL MODEL AND STUDY PARTICIPANT DETAILS

Human brain tissues—Adult human cortex tissues (n=4, 24–33 years old; frontal cortex, n=2; temporal cortex, n=2) were obtained from Mayo Clinic Florida Biospecimen Bank and processed at Yale University School of Medicine. Children cortex tissues (n=4, 5–12 years old; temporal cortex, n=3; amygdala, n=1) were obtained and processed from the Children's Hospital of Fudan University in Shanghai, followed the same RNA extraction, library preparation and sequencing protocols as Yale site. The IRB protocols were approved both at Mayo Clinic Florida and the Children's Hospital of Fudan University in Shanghai.

Mice—Wild type C57BL/6J mice were obtained from the Jackson Laboratory. *Shank3* mutant mice of Shank*3* exons 4–9 deletion ($Shank3^{\Delta e4-9}$)[35] and *Shank3* exons 4–22 ($Shank3^{\Delta e4-22}$)[20] were generated and maintained in Jiang's lab. *Shank3* exon 21 deletion ($Shank3^{\Delta e21}$) *was obtained from Jackson Laboratory (*$Shank3^{tm1.1Pfw}$/J and Strain #:018398).[116] Mice were housed of 4–5 per cage in pathogen-free mouse facility with free access to food and water on a 12-hour light: dark cycle at the ambient temperature of 20–22°C and humidity of 30–70%. An equal number of male and female mice were used for all experiments. All procedures were performed following the approved animal protocol by Yale University School of Medicine Animal Care and Use Committee.

METHOD DETAILS

RNA isolation and quality control—Mouse brain tissues were snap-frozen in liquid nitrogen immediately after dissection. Human brain tissues were snap-frozen in liquid nitrogen within an hour after dissection. All tissues were stored in liquid nitrogen thereafter. Total RNA was isolated from 20 mg frozen tissues, using NucleoZOL™ (Takara Bio, 740404.200) and NucleoSpin® RNA set for NucleoZOL™ (Takara Bio, 740406.50) following the manufactures specifications, followed by rDNase Set (Takara Bio, 740963) to digest DNA, and NucleoSpin® RNA Clean-up XS (Takara Bio, 740903) for RNA repurification. RNA purity (260/280, 260/230) and concentration were measured on NanoDrop™ 2000/2000c Spectrophotometers. RNA integrity number (RIN) was assessed using Agilent 2100 Bioanalyzer system.

Generation of standard and captured Iso-seq libraries—The Iso-seq libraries were prepared by following the manufacturer's instructions for each step (Iso-Seq™ Express Template Preparation for Sequel® and Sequel II Systems for standard Iso-seq; Customer Collaboration – Iso-Seq® Express Capture Using IDT xGen® Lockdown® Probes for capture Iso-seq). The 600 ng of total RNA was used as input. Only the RNA with RIN

higher than 7 of human samples, and 8 of mouse samples were processed for reverse transcription, amplification, enrichment, and library preparations.

Hybridization capture panel design—Hybridization capture panel design was assisted by IDT (Integrated DAN Technologies). Briefly, after extracted as 120-base-length sequence of interested gene, xGen Lockdown probes were aligned to the genome and calculated the number of possible enrichment sites. A “perfect” probe was considered as only has 1 hit (the target of interest) with genome, but most of the sequences returned more than 1 hit. Following IDT proprietary xGen Off-Target QC Method, any probes with more than 50 hits were removed because of non-specific targets in genome. The specifics and details of each probe panel are presented in Table S3.

Hybridization protocol—300 ng of total RNA in less than 5.4 μL of volume mixed with 2 μL of NEBNext Single Cell RT Primer Mix. The final volume was brought up to 9 μL with nuclease-free water. The reaction was placed in a thermocycler and run for 5 minutes at 70°C, followed by holding at 4°C for primer annealing and first-strand synthesis. Reverse transcription template switching reaction was then performed by adding 5 μL of NEBNext Single Cell RT Buffer, 3 μL of nuclease-free water, and 2 μL of NEBNext Single cell RT Enzyme Mix to the first-strand cDNA. The reaction was incubated in a thermocycler at 42°C with the lid at 52°C for 75 minutes, followed by holding at 4°C. After adding 1 μL of Iso-Seq Express Template Switching oligo to the 19 μL reaction for a final volume of 20 μL, the reaction was incubated again in a thermocycler at 42°C with the lid at 52°C for 15 minutes, followed by holding at 4°C.

The Reverse Transcription and Template Switching reaction product was then purified using ProNex Beads before proceeding with cDNA amplification. For amplification, 50 μL of NEBNext Single Cell cDNA PCR master Mix, 2 μL of NEBNext Single Cell cDNA PCR Primer, 2 μL of Iso-Seq Express cDNA PCR primer, and 0.5 μL of NEBNext Cell Lysis Buffer were added to the purified product. The reaction was incubated in a thermocycler and run for 45 seconds at 98°C, followed by 14 cycles of the following steps: 10 seconds at 98°C, 15 seconds at 62°C, and 3 minutes at 72°C. The reaction was then held for 5 minutes at 72°C, followed by holding at 4°C. Finally, the product was purified again using ProNex Beads before proceeding with either the library preparation for standard Iso-Seq (SIS) or the capture steps for capture-based Iso-Seq (CIS).

As for the capture steps, first concentrate a total of 500ng cDNA in a 1.5 mL LoBind tube along with 7.5 μL of Cot DNA. To this mixture, add 1.8X volume of ProNex beads and gently pipette mix 10 times, followed by incubation for 10 min at room temperature. Place the tube on a magnet stand and wait until supernatant is clear. Remove the supernatant and wash twice with 200μL of freshly prepared 80% ethanol while on the magnet stand. Spin the tube strip briefly after removing the second wash, return to magnetic stand, and remove residual ethanol. Next, immediately add the hybridization reaction mix (which comprises 2X Hybridization Buffer, Hybridization Buffer Enhancer, xGen Asym TSO block, xGen RT-primer-barcode block, and 1X xGen Lockdown Panel) to elute the cDNA. Gently pipette mix 10 times and incubate for 5 min at room temperature. Then, place the tube on the magnetic stand to separate the beads from the supernatant. Transfer 17 μL of the supernatant

to a new 0.2 mL PCR tube and briefly centrifuge it. Ensure that the tube is tightly sealed to prevent evaporation. Finally, place the sample tube in the thermal cycler and start the hybridization program: HYB program (lid set at 100°C), 95°C for 30 sec, 65°C for 4 hr, and lastly hold at 65°C.

During the incubation, prepare 1X working buffers and beads for capture. Preheat the wash buffers to +65°C in a heat block or water bath. To prepare the capture beads, allow the Dynabeads M-270 Streptavidin to warm to room temperature for 30 minutes prior to use. Thoroughly vortex the beads for 15 seconds to mix them, then aliquot 50 μL of beads into a 0.2 mL PCR tube, followed by adding 100 μL of 1X Bead Wash Buffer per capture, and pipette the mixture 10 times. Place the PCR tube on a magnetic rack. When the supernatant is clear, carefully remove and discard it without disturbing the beads. Note: Allow the Dynabeads to settle for at least 1 minute before removing the supernatant. Thereafter, two washes are performed as follows: Add 100 μL of 1X Bead Wash Buffer, pipette 10 times to mix, then place the PCR tube on a magnetic rack, allowing the beads to fully separate from the supernatant. Carefully remove and discard the clear supernatant. Repeat this process for a total of two washes. Finally, resuspend the beads in 17 μL of Bead Resuspension Mix per capture. The Bead Resuspension Mix includes xGen 2X Hybridization Buffer (8.5 μL), xGen Hybridization Buffer Enhancer (2.7 μL), and Nuclease-Free Water (5.8 μL). By following these steps carefully, you can ensure that the buffers and beads are prepared correctly for the capture step and obtain reliable results.

Then Bind cDNA to the capture beads, by incubating the samples in a thermocycler set to +65C for 45 minutes. Then Wash the captured cDNA with 1X wash buffers and elute the cDNA with 46ul elution buffer. To amplify the captured DNA sample, NEBNext High-Fidelity 2X PCR Master Mix is recommended, and the NEBNext Single Cell cDNA PCR Master Mix is alternative for post capture amplification. Assemble the following PCR reaction: 50 μL of NEBNext High-Fidelity 2X PCR Master Mix, 2 μL of NEBNext Single Cell cDNA PCR Primer, 2 μL of Iso-Seq Express cDNA PCR Primer, 0.5 μL of NEBNext Cell Lysis Buffer, and 45.5 μL of the captured library. Amplify the PCR reaction mix using the following PCR protocol: Denature the DNA at 98°C for 45 seconds. Perform 14 cycles of the following steps: a. Denature the DNA at 98°C for 10 seconds. b. Anneal the primers at 62°C for 15 seconds. c. Extend the DNA at 72°C for 3 minutes. Final extension at 72°C for 5 minutes, and hold at 4°C. Finally perform the post amplification clean up steps with ProNex brands and ethanol. Use 1 μL of sample to quantifiy with Qubit dsDNA HS kit and dilute 1 μL of sample to 1.5ng/μL and run 1 μL on an Agilent Bioanalyzer using the High Sensitivity DNA kit. We used 500ng cDNA for library construction as Sequel II sequence platform required. After DNA damage repair, end repair/A-Tailing, overhang adapter ligation, and purification with ProNex Beads, the cDNA library is ready for sequencing

Sequencing platform—To load the cDNA library onto the PacBio Sequel II System, the diffusion method was applied and followed by a 24-hour movie time and a 2-hour pre-extension time. The samples were cleaned up using ProNex beads and loaded onto the plate at a concentration of 50–100 pM.

Real-time quantitative polymerase chain reaction (RT-qPCR)—Two μg of total RNA was reverse transcribed into cDNA templates using RNA to cDNA EcoDry™ Premix kit including both random hexamer and oligo(dT)$_{18}$ primers (Takara Bio, 639548). KAPA SYBR® FAST qPCR Master Mix (2X) Universal (Kapa Biosystems, KK4602) was used for qPCR reactions with 18 ng of cDNA as template input. The following program on CFX96 Touch Real-Time PCR Detection System (BIO-RAD) was used: 3 minutes at 95°C for enzyme activation, followed by 40 cycles of denaturation (95°C, 3 seconds) and annealing, extension, data acquisition (60°C, 30 seconds), followed by dissociation and holding at 4°C. The PCR primers are shown in Table S3.

Western Blot—Whole cell lysates were extracted from mouse brain tissue using the NucleoSpin® RNA/Protein Kit (Takara Bio, 740933.50). Protein concentrations were quantified using the Protein Quantification Assay (Takara Bio, 740967.250). The samples were then mixed with 4x Laemmli buffer (Bio-Rad, 1610747) and heated at 98 °C for 5 minutes to denature the proteins. Subsequently, proteins were loaded onto 4–20% Mini-PROTEAN® TGX Stain-Free™ Protein Gels (Bio-Rad, 4568094) for electrophoresis. For immunodetection, the gels were incubated with antibodies targeting the C-terminus (Invitrogen, PA5–114207, 1:500) and N-terminus (Invitrogen, PA5–99580, 1:500) of ACR overnight at 4°C.

QUANTIFICATION AND STATISTICAL ANALYSIS

Sequence data filtering algorithm—The following pipeline was diagramed in Figure S12. Sequencing reads were screened initially with Lima (v2.5.0) and IsoSeq (v3). A transcript with both cDNA primers and the poly(A) was identified and called Full-length reads.[117] The Full-length reads which had less than 100 base pairs 5’ end overhang, less than 30 bases pairs 3’ end overhang, and less than 10 base pairs gaps in the middle are considered as the same transcript. Clustering using hierarchical alignment, and iterative cluster merging, generate polished sequence, with quality scores. The output further filtered with SQANTI3 (v4.3)[107] after cluster and collapse to generate unique transcripts. SQANTI3 filtered the transcripts as below: If a transcript is Full-Splice Match (FSM), then it was retained unless the 3’ end was unreliable (intrapriming). If a transcript was not Full-Splice Match, then it was retained only if all below were met: (1) 3’ end is reliable. (2) did not have a junction that was labeled as RT-Switching. (3) all intro-exon junctions were canonical.[117] Further criteria included a transcript had to include at least 2 exons, and in the sense orientation and predicted open reading frame (ORF) had longer than 100 amino acids for the given transcript.

Iso-seq data analysis pipeline—The flow chart (Figure S12) described the analytic pipeline for ISO-Seq sequence dat. The subreads.bam file of an Iso-Seq SMRT cell was a raw input. The number of the SMRT cells, instead of the number of multiplex samples sequenced on a SMRT cell, regardless of the library preparation methods [Sta-Iso-Seq (SIS) or Cap-Iso-Seq (CIS)], dictated the direction of the analysis flow.

Transcript confidence score—To assess the quality of individual transcript, transcripts after filtering steps were scored by the following scoring metrics: (1) Score of 3 point: If the

exons of transcript were presented in the sequences of by either Illumina short read methods of the bulk RNAseq (human dataset: UCLA-ASD, BrainGVEX, CMC, CommonMind and LIDB) and SMART scRNAseq. (2) Score of 2 points: If a transcript had predicted ORF longer than 100AA. (3) If the abundance of a transcript were higher than 20 percentage of the rank of the abundance of all transcripts. The summation of scores was confidence score to define each transcript: high confidence (≥ 4 points), moderate confidence (2–3 points), and low confidence (0–1 point).

RNA-seq data processing—Illumina bulk RNA-Seq raw data in FASTQ format after quality control and filtering with fastp,[108] and SMART scRNAseq FASTQ data, were aligned to hg38 for human sequences and mm39 for mouse sequences using HISAT 2.2.1.[109] Aligned RNA-Seq data (aligned to hg37/38) in BAM format were converted to FASTQ format using SAMtools[110] when the raw FASTQ was not available, followed by the same process as above. Gene expression counts and DEXSeq-counts were calculated using FeatureCount[111] for further gene expression and exon usage analysis. Detailed RNAseq datasets information summarized in Table S4.

Differential transcript usage—Transcript-level quantification of the processed RNA-Seq data was performed using the software Salmon 1.4.0.[112] The transcriptome index used for quantification was built from the reference genome annotation (in GTF format), along with the reference genome FASTA file. Transcript abundances were estimated using the quasi-mapping algorithm (–quasiMAP) mode, which performs a light-weight alignment-free estimation of abundances based on k-mer matching. The output files were generated in TPM (transcripts per million) format.

Differential exon usage (DEU)—DEXSeq-counts tables were imported into R, analysis with R package DEXSeq.[113] Normalization and filtering were performed to remove lowly expressed exons. DexSeq uses a binomial generalized linear model to estimate exon expression, accounting for the variability in exon-exon junction usage across samples. DEU was then tested using the DEXSeq function, which fits a statistical model to test for differences in exon usage between two or more groups of samples. Exons with an adjusted p-value ≤ 0.05 and a log2 fold change ≥ 1 or ≤ −1 were considered significantly differentially used and visualized with built-in function of DEXSeq.

Whole genome sequencing and exome analysis—DNA variation data post variation calling in VCF format were downloaded from Autism Sequencing Consortium (ASC), Bipolar Exomes (BipEx), whole-exome sequencing case-control study of epilepsy (Epi25), Schizophrenia exome meta-analysis consortium (SCHEMA), and PsychENCODE. VCFs initially aligned to hg38 (BipEx and Epi25) and the datasets (ASC, SCHEMA and PsychENCODE) after alignment lift over from hg37 to hg38 with UCSC LiftOver tool and chain file, were subsetted to the region of interest (SHANK3, chr22:50670000–50770000) using BCFtools (v 1.16).[110] The data format was modified using HTSlib (v 1.16)[114] and TAB-delimited file InderXer (Tabix, v 0.2.5).[115] Then the data were annotated with Ensembl Variant Effect Predictor (VEP, release 107)[74] and filtered with Genome Aggregation Database (gnomAD, v3.1.2)[75] by INFO/AF_popmax<=0.01. Filtered DNA variation were

aligned to novel exons detected in SIS and CIS with SpliceAI[76] for splicing event analysis, and with SnpEff [77] to evaluate other deleterious SNV (stop lost, stop gain and frameshift).

Spatial transcriptional analysis—An open access Visium dataset of mouse brain coronal section from 10x Genomics[63] in FASTQ format was analyzed using customized references and annotation generated from mouse *Shank3* CIS transcripts using Cell Ranger,[118] followed by quantitation with customized probe-set (probe-transcripts relation spreadsheet) using 10x Genomics Space Ranger v2.0. The output cloupe file was visualized using 10x Genomics Loupe Visualization Software v6.5.

Data visualization—Visualization was performed using ggplot2 (version 3.3.2) in R (version 4.2.2) for plotting gene expression, transcript and exon usage profiles and heatmaps.

Tandem mass spectrometry—Trypsin digests were subjected to liquid chromatography-tandem mass spectrometry (LC-MS/MS) analysis using the ThermoFisher Scientific LTQ-Orbitrap XL mass spectrometer. Raw mass spectrometry data were processed using MaxQuant v2.5.2.0. Standard label-free quantification parameters were applied as group-specific parameters. Additionally, a pooled Mouse Proteome dataset sourced from UniProt and predicted Shank3 open reading frames (ORFs) from CIS were included as global parameters. Identified protein IDs were subsequently reported as results.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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Highlights

- Revealed the transcriptional complexity of SHANK family genes in human and mouse brains
- Improved detection rate for potential deleterious variants of neuropsychiatric disorders
- Evidence for the debate over transcriptional determinism and stochasticity

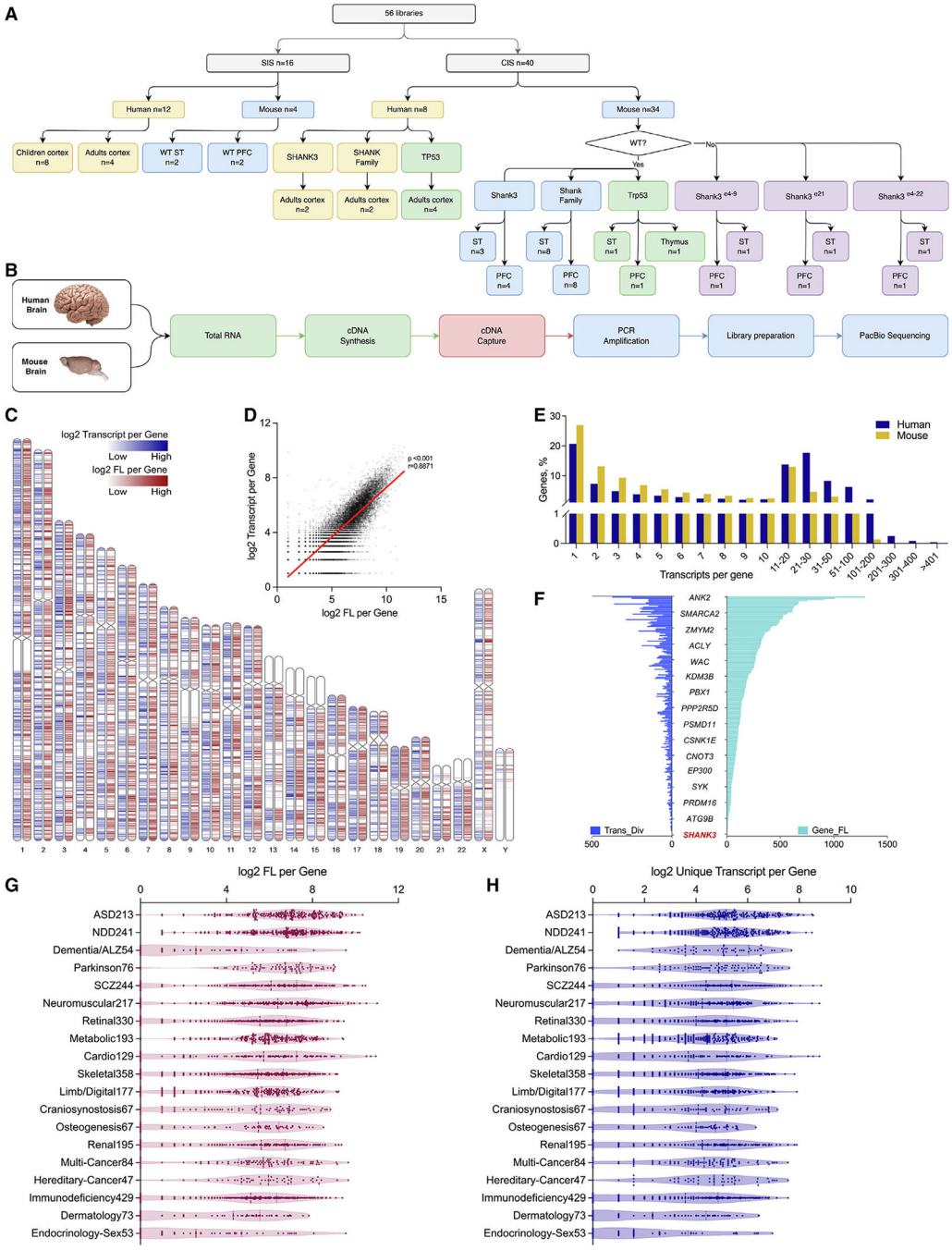

Figure 1. Genome-wide transcript diversity and abundance in brains detected by SIS

(A) Experimental design of SIS and CIS of human and mouse tissues.

(B) Schematic of experimental procedure for RNA capture and long-read sequencing.

(C) Number of unique transcripts (transcript diversity) for individual genes (blue) and the number of sequence reads (abundance) (red) for an individual transcript detected in human cerebral cortex by SIS with projected chromosome coordinates and idiograms.

(D) Transcript diversity was significantly correlated with the sequence reads (abundance) of the transcripts.

(E) Number of transcripts per gene genome wide from SIS in human and mouse brains.
(F) Number of unique transcripts (Trans_Div) and abundance (Gene_FL) for 213 ASD risk genes, shown as an average of 56 transcripts per gene and a median of 35.
(G and H) Human SIS data show heightened transcript diversity in genes associated with brain disorders, especially ASD and NDD, compared to other diseases. We observed a strong correlation between transcript diversity and abundance in all gene clusters except for those related to dementia/Alzheimer's disease.

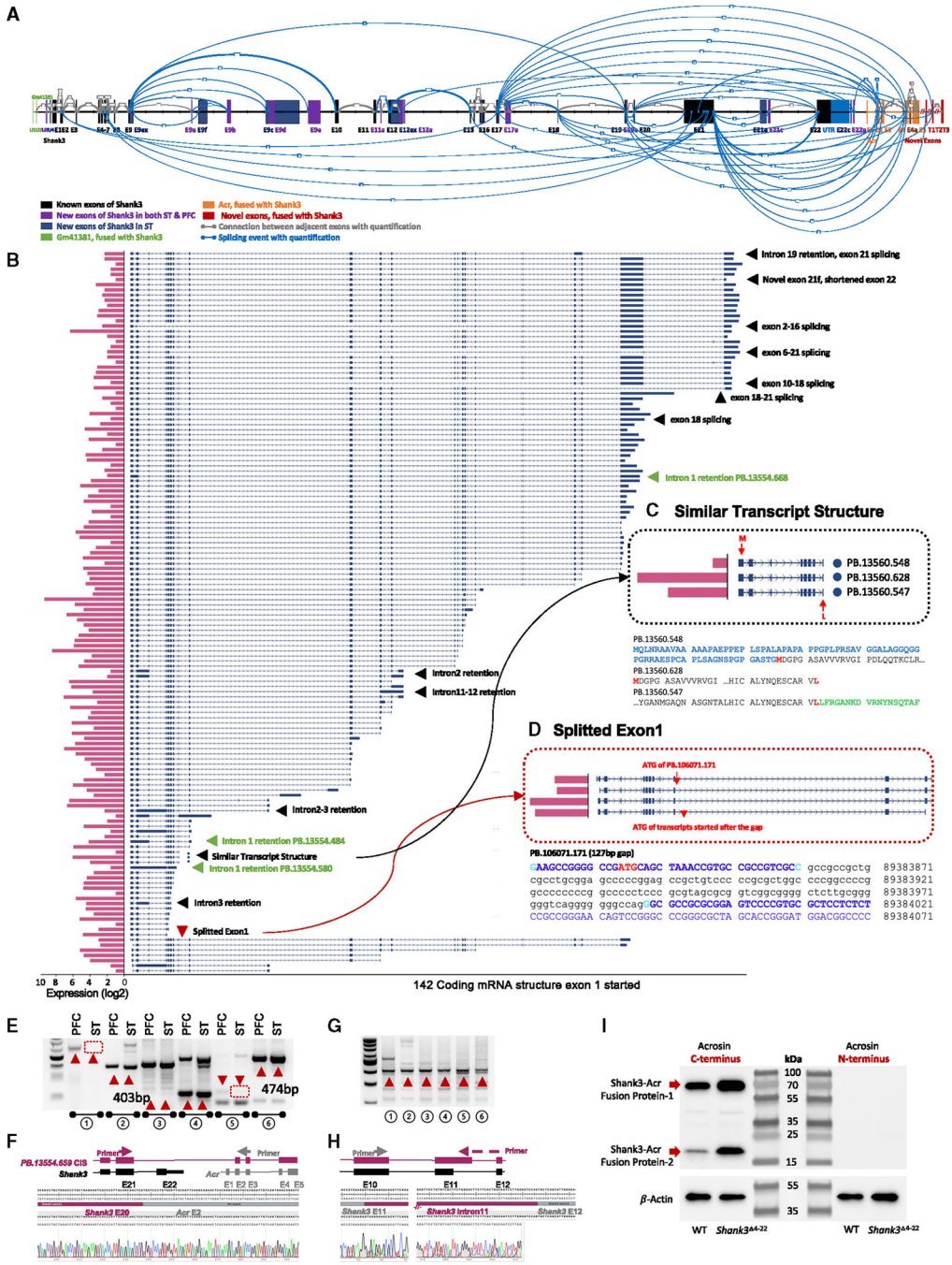

Figure 2. Novel *Shank3* transcriptome in mouse striatum (ST) by CIS

(A) CIS revealed a refined *Shank3* gene structure and splicing patterns in WT mouse ST. The established *Shank3* structure (NM_001034115, mm39) is expanded with newly detected exons shared between ST and PFC, depicted in purple. Unique splicing events, represented by gray lines with thickness indicating read quantity, include novel ST-specific exons in dark blue and alternative splices in light blue. Fusion transcript exons near Gm41381 and *Acr*, shown in green and orange, respectively, feature unique splicing with newly identified red exons (T1–T3) exclusive to *Shank3*. New exon U3 is shared between ST and PFC. U4 is

linked to Gm4138 and ST specific. Exon 21e is a new in-frame exon and 21c is a new exon harboring a stop codon (enlarged view in Figure S9).

(B) One hundred forty-two unique transcripts started with the canonical exon 1 of annotated *Shank3* (NM_001034115) in ST and terminated at different positions. Pink bar plots on the left are the abundance (log2 counts). Arrows describe the features of given transcripts.

(C) Example of transcripts with similar structures in panorama but different at the sequence level with predicted ORFs and ATG codons. The transcripts of PB.13560.548, PB.13560.628, and PB.13560.547 are similar, but the predicted ORFs show different ATG codons and protein domains.

(D) Details of the split exon 1. There is a cryptic splicing of 127 bp (non-capitalized sequence in black) within the annotated exon 1 of transcript PB.106071.171, which resulted in a predicted upstream ATG codon and an additional 134 aa. Other transcripts have transcriptional start sites (TSSs) in exon 1 but a predicted ATG codon in exon 2. Variability in TSS and intron 1 retention, as seen in transcripts PB.13554.484, PB.13554.580, and PB.13554.668, leads to ORFs of 304, 106, and 1,290 aa, respectively.

(E) Validation of new transcripts from paired mouse PFC and ST samples. Pair 1, novel exon U1; pair 2, fusion transcript between *Shank3* exon 21 and *Acr* exon 2; pair 3, splicing event between *Shank3* exon 9 and exon 19; pair 4, splicing event between *Shank3* exon 5 and exon 21; pair 5, novel exon 9b of *Shank3*; pair 6, *Shank3* exon 11 extension/intron 11 retention. The red arrows indicate the novel products confirmed by Sanger sequencing. Other bands are products from known transcripts.

(F) Sanger sequencing confirmation of a fusion transcript between *Shank3* exon 21 and *Acr* exon 2 in mouse brain (pair 2 of E).

(G) Fusion transcripts in other tissues. Forward and reverse primers were from exon 20 of *Shank3* and exon 5 of mouse *Acr*, respectively. Lane 1, liver in P21 mouse; lane 2, thymus in P21 mouse; lane 3, ovary in P21 mouse; lane 4, ovary in 3-month-old mouse; lane 5, testis in P21 mouse; lane 6, testis in 3-month-old mouse. The red arrows show the novel products confirmed by Sanger sequencing as indicated. Other bands are known products.

(H) Sanger sequencing of *Shank3* exon 11 extension/intron 11 retention in mouse brain (lane 6 of G).

(I) Western blot shows the upregulation of SHANK3-ACR fusion protein in mouse PFC of $Shank3^{\Delta e4-22-/-}$ mutant mice compared to WT.

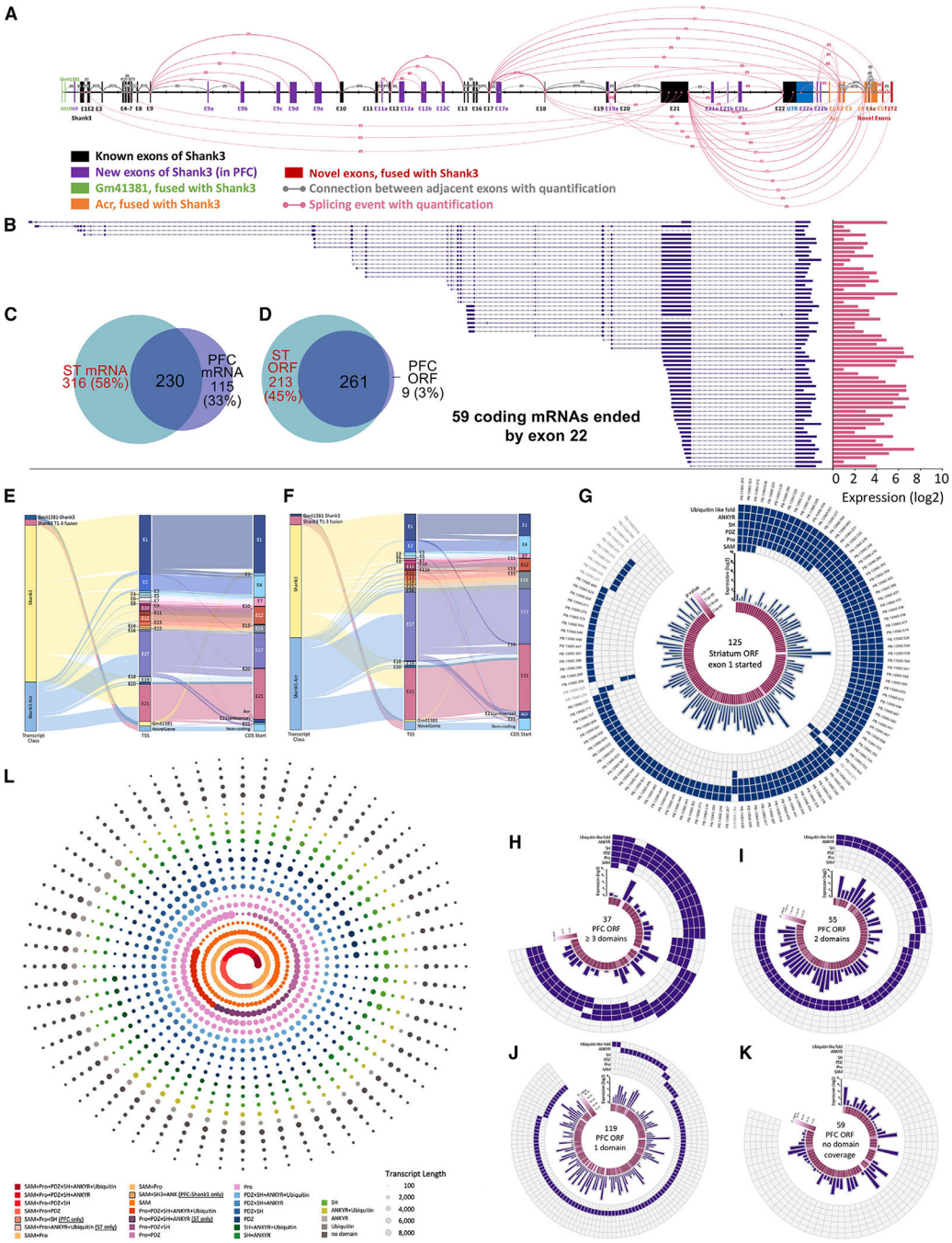

Figure 3. Novel *Shank3* transcriptome in mouse PFC by CIS and predicted domain structures of ORFs

(A) New *Shank3* transcript structure and conch plot of splicing events discovered in WT mouse PFC by CIS. Color code is the same as in Figure 2A. The novel exon 9a (chr15: 89394416–89394465, mm39) is shared between PFC and ST. Other novel exons such as exon 12e (chr15: 89414330–89414640, mm39) were unique to PFC. Novel exons 21a, 21b, and 21c are predicted to result in an early stop codon and shorter ORFs (chr15: 89394416–89394465, chr15: 89408698–89408784, chr15: 89418571–89418609, mm39) (enlarged view in Figure S10).

(B) Structure of 59 transcripts with different TSSs but terminating at annotated exon 22 of *Shank3*. Pink bar plot represents the abundance (log2 counts) of each transcript.
(C and D) The comparison of transcripts and predicted ORFs between mouse ST and PFC.
(E and F) The pattern of deduced TSSs and predicted starting sites of the coding sequence (CDS) for all *Shank3* transcripts, including new 5′ and 3′ fusion transcripts from CIS in mouse ST (E) and PFC (F). Each filament represents an individual transcript in different classes of GM41381 (U1–U2)-*Shank3*, *Shank3*-T1–3, *Shank3*, Shank3-Acr (first column), deduced TSS (middle column), and predicted starting sites of CDS (third column).
(G) A total of 125 unique ORFs are predicted from 142 transcripts starting with exon 1 in ST. The pattern of the combination of six protein domains is shown in the outermost ring of the windmill plot. The middle layer shows the abundance of each RNA transcript and the *p* value of its expression level compared to other transcripts. Only four ORFs of transcripts contained all six protein domains.
(H–K) Four windmill plots showing 270 predicted ORFs from all 345 transcripts detected in PFC classified by the combination of functional domains.
(L) Spiral plot showing an aggregated functional domain coverage of the transcripts captured by the *Shank1–3* joint probe panel by CIS of mouse PFC and ST. Each dot represents a unique transcript. Each color represents a unique combination of functional domains. The dots are ordered from the longest to the shortest transcript, while the colors are arranged from the SAM to the Ubl domain.

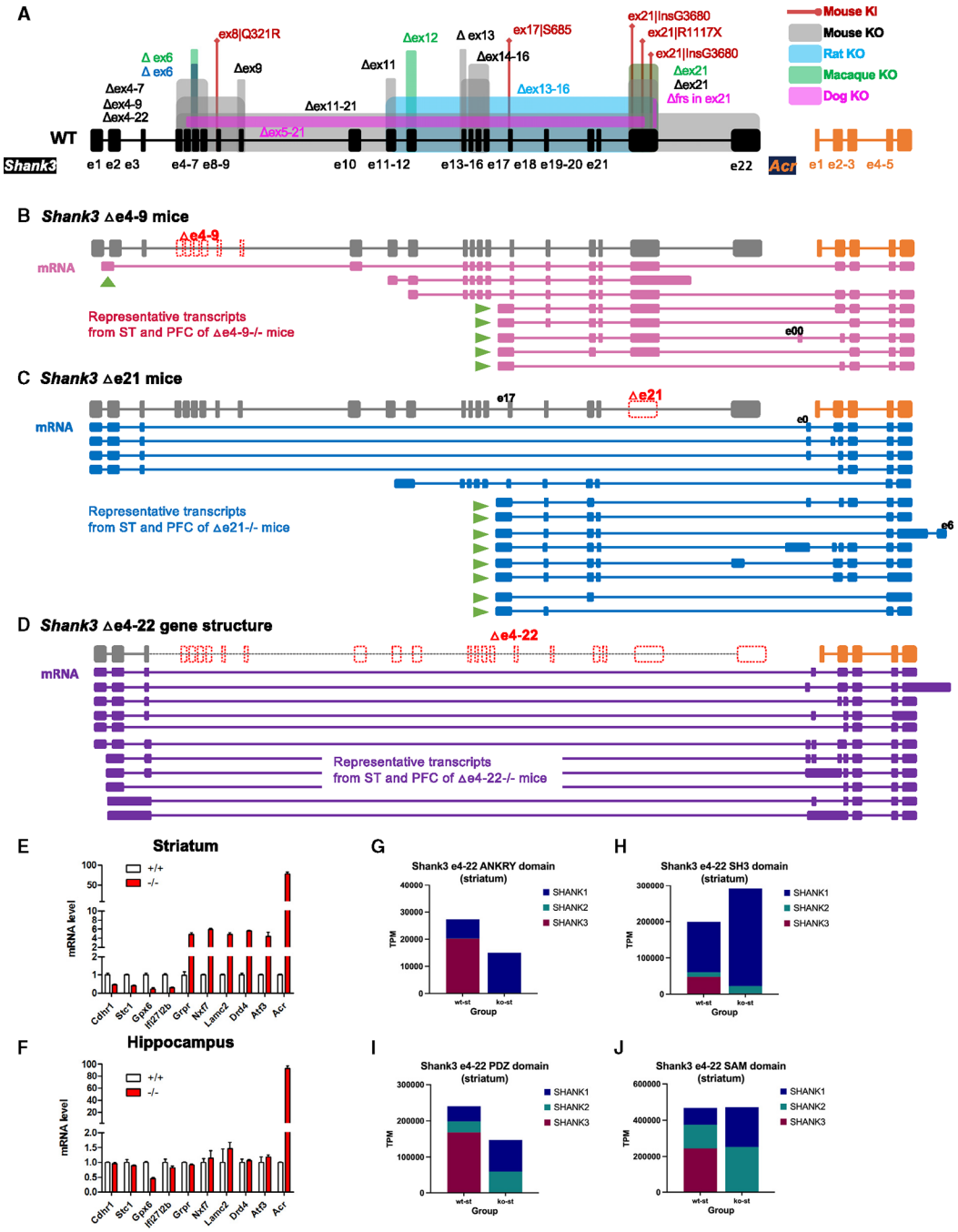

Figure 4. The summary and illustration of altered *Shank3* transcripts in *Shank3*$^{\Delta e4–9}$, *Shank3*$^{\Delta e21}$, and *Shank3*$^{\Delta e4–22}$ mutant mice from CIS

(A) Current annotated mouse *Shank3* and *Acr* (NM_013455, mm39) gene structures. The annotations of genetically targeted mutations in mouse, rat, monkey, and dog are shown (KO, exonic deletions; KI, knockin mutation).

(B) The gene structure of *Shank3*$^{\Delta e4–9}$ mutant mice is in gray, and representative mRNA transcripts, according to structural uniqueness, from *Shank3*$^{\Delta e4–9–/–}$ mice are in pink. No transcript using first annotated exon 1 was detected. Instead, the first exon, presumably a cryptic TSS (arrow), was detected in intron 1. The exon 4–9 deleted transcript missed

exons 11, 12, and 22, but has a fusion between *Shank3* and *Acr*. The transcripts starting at intron 16/exon 17 (arrows) as the first exon were most abundant. Extensive fusion transcripts between *Shank3* exon 21 and *Acr* exon 2 were observed. The last coding exon 22 was not detected in any transcripts.

(C) The gene structure of *Shank3*$^{\Delta e21}$ mutant mice and *Acr* gene, in gray, and representative mRNA transcripts, from a structural uniqueness perspective, from *Shank3*$^{\Delta e21-/-}$ mice in blue. Splicing between exon 4 of *Shank3* and exons of *Acr* that resulted in fusion transcripts was observed. The transcripts starting at intron 16/exon 17 (arrows) as first exon and fusion between *Shank3* and *Acr* were most common. The coding exon 22 was not detected in any transcript.

(D) The gene structure of *Shank3*$^{\Delta e4-22}$ mutant mice and the *Acr* gene, in gray, and representative mRNA transcripts in purple to reflect structural uniqueness. The number of fusion transcripts between *Shank3* and *Acr* is significantly increased in *Shank3*$^{\Delta e4-22-/-}$ mutant mice.

(E and F) Increased expression of the *Acr* transcript in *Shank3*$^{\Delta e4-22-/-}$ mutant mouse by RT-qPCR. The expression of the *Acr* gene was significantly increased in both striatum and hippocampus by >100-fold.

(G–J) Compensatory expression of the functional domains of *SHANK* family proteins in the striatum of *Shank3*$^{\Delta e4-22}$ mutant mice. The bulk RNA-seq data of *Shank3*$^{\Delta e4-22}$ were analyzed for the compensatory expression of other functional domains of *Shank1* and *Shank2* genes. The deficiency of the ANKYR and SH3 domains of SHANK3 was compensated for by SHANK1, but the deficiency of the PDZ and SAM domains was compensated for by both SHANK1 and SHANK2. The deficiency of the SAM and SH3 domains was fully compensated for, but the deficiency of the ANKYR and PDZ domains was partially compensated for.

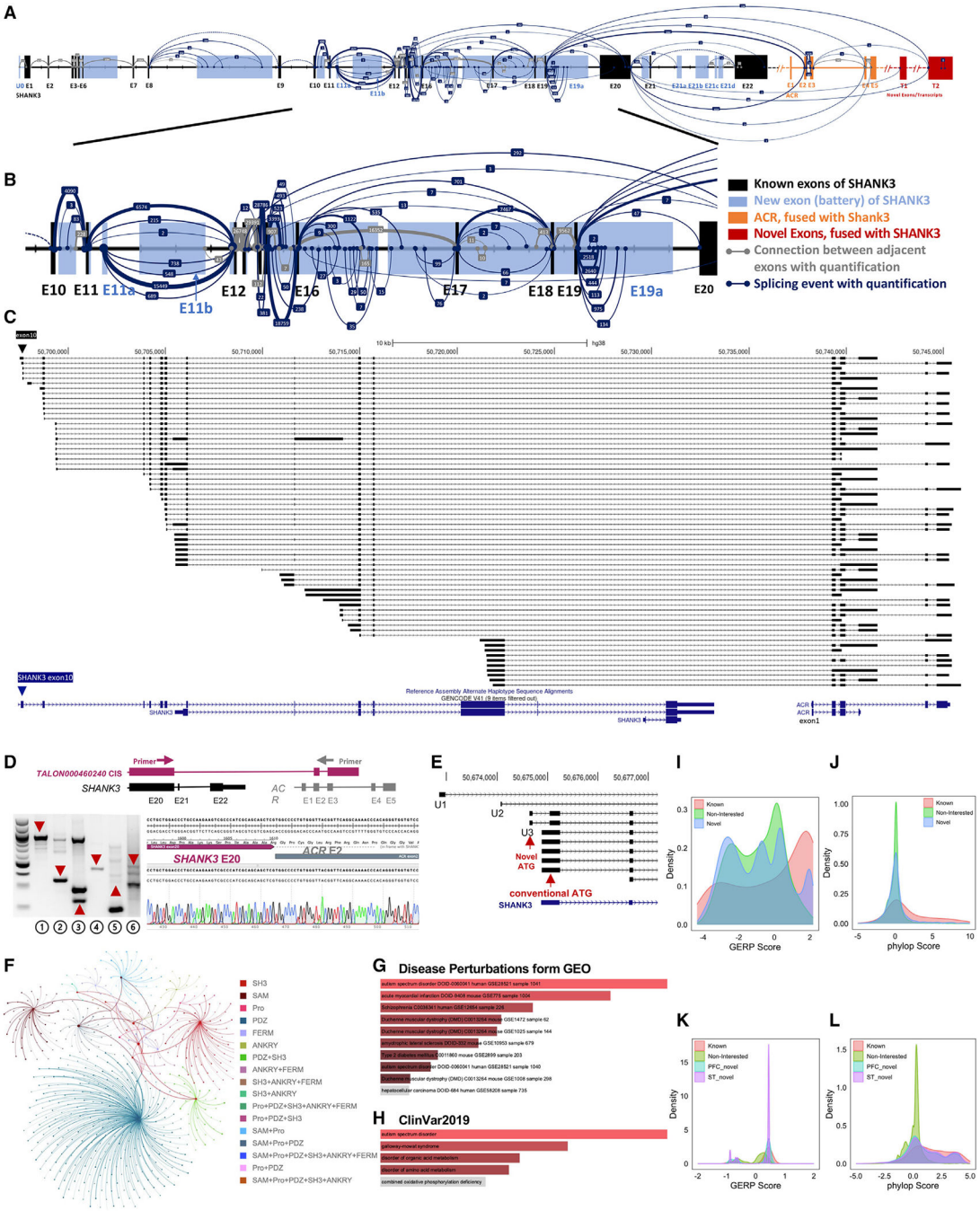

Figure 5. The novel transcripts of human *SHANK3* genes detected by CIS and predicted ORFs

(A) New *SHANK3* transcript structure and conch plot of *SHANK3* transcripts discovered by CIS in normal human cortex. Black backbone is the annotated *SHANK3* transcript of NM_001372044 (hg38). Blue rectangles represent novel exons of *SHANK3*. The exons of *ACR* are shown as orange rectangles. The new and uncharacterized exons distal to *ACR* are red rectangles. The gray line connects adjacent exons, while the light blue line illustrates alternative splicing events. The number of sequence reads for the splicing event is shown

in the middle of connecting lines and reflected in the thickness of the connecting lines (enlarged view in Figure S9).

(B) Zoomed-in view of the splicing events between exons 10 and 20 in the human cortex. Exons 16 and 20 of *SHANK3* in humans correspond to exons 17 and 21 of *Shank3* in mice.

(C) Structure and abundance of the fusion transcripts between *SHANK3* and *ACR* in the human cortex. A majority of fusion transcripts are initiated after exon 10, mainly from introns 16 and 17 and exon 21. The fusion transcripts are notably skipping exon 20 (the largest exon) of *SHANK3* and exon 1 of *ACR*.

(D) Validation of novel *SHANK3* transcripts in human brain tissue by RT-PCR and Sanger sequencing. Diagram of the primer design of L1 is shown. RT-PCR gel: L1, fusion transcript between *SHANK3* exon 20 and *ACR* exon 2; L2, fusion transcript between *SHANK3* exon 20 and *ACR* exon 4; L3, fusion transcript between SHANK3 exon 19 and ACR exon 2; L4, novel exon U3; L5, intron 14 retention; and L6, intron 15 retention. M, DNA marker. The Sanger sequence of the RT-PCR product of *SHANK3* exon 20 and *ACR* exon 2 fusion from L1 is shown.

(E) Three new exons upstream of the annotated exon 1 of *SHANK3* mRNA (NM_001372044) (U1, chr22: 50672853–50672979; U2, chr22: 50674076–50674097; U3, chr22: 50674642–50674705, hg38). A new ATG codon is in U2.

(F) Dandelion plot shows functional domain combinations of the *SHANK1*, *SHANK2*, and *SHANK3* transcripts from CIS. Each dot represents a unique transcript, and each color is a unique combination of functional domains. There are 17 combinations of functional domains of human *SHANK* family genes. The PDZ domain was significantly more present (~70%) in predicted ORFs.

(G and H) Significant enrichment of fusion transcripts in transcriptome data of ASD and schizophrenia. Gene Ontology enrichment analysis with Enrichr95 in 41 disease-related datasets is shown. The fusion transcripts were significantly enriched in ASD and schizophrenia in disease perturbations from the GEO dataset (G) and the ClinVar2019 dataset (H).

(I and J) Distribution of GERP (G) and PhyloP (H) scores across human *SHANK3* genomic regions of known coding exons, novel exons from CIS, and a non-transcribed region in cerebral cortex. (I) The GERP score for novel exons from CIS in cerebral cortex is significantly higher than in a non-transcribed region (D = 0.097; $p < 0.001$) but significantly lower than that of *SHANK3* known exons (D = 0.299; $p < 0.001$). (J) The PhyloP score for novel exons from CIS in cerebral cortex is significantly higher than in a non-transcribed region (D = 0.133, $p < 0.001$) but significantly lower than that of *SHANK3* known coding exons (D = 0.296, $p < 0.001$).

(K and L) Distribution of GERP and PhyloP scores across mouse *Shank3* genomic regions of known coding exons, novel exons from CIS, and a non-transcribed region in PFC and ST. (K) The GERP score for novel exons from CIS in PFC and ST is significantly higher than that of a non-transcribed region (PFC, D = 0.548, $p < 0.001$; ST, D = 0.602, $p < 0.001$) but significantly lower than that of known *Shank3* coding exons (PFC, 0.15, $p < 0.001$; ST, D = 0.0960; $p < 0.001$). (L) The PhyloP score for novel exons from CIS in PFC and ST is significantly higher than that of a non-transcribed region (PFC, D = 0.385, $p < 0.001$; ST, D = 0.439, $p < 0.001$) but significantly lower than that of known *Shank3* coding exons (PFC, D = 0.184, $p < 0.001$; ST, D = 0.128, $p < 0.001$).

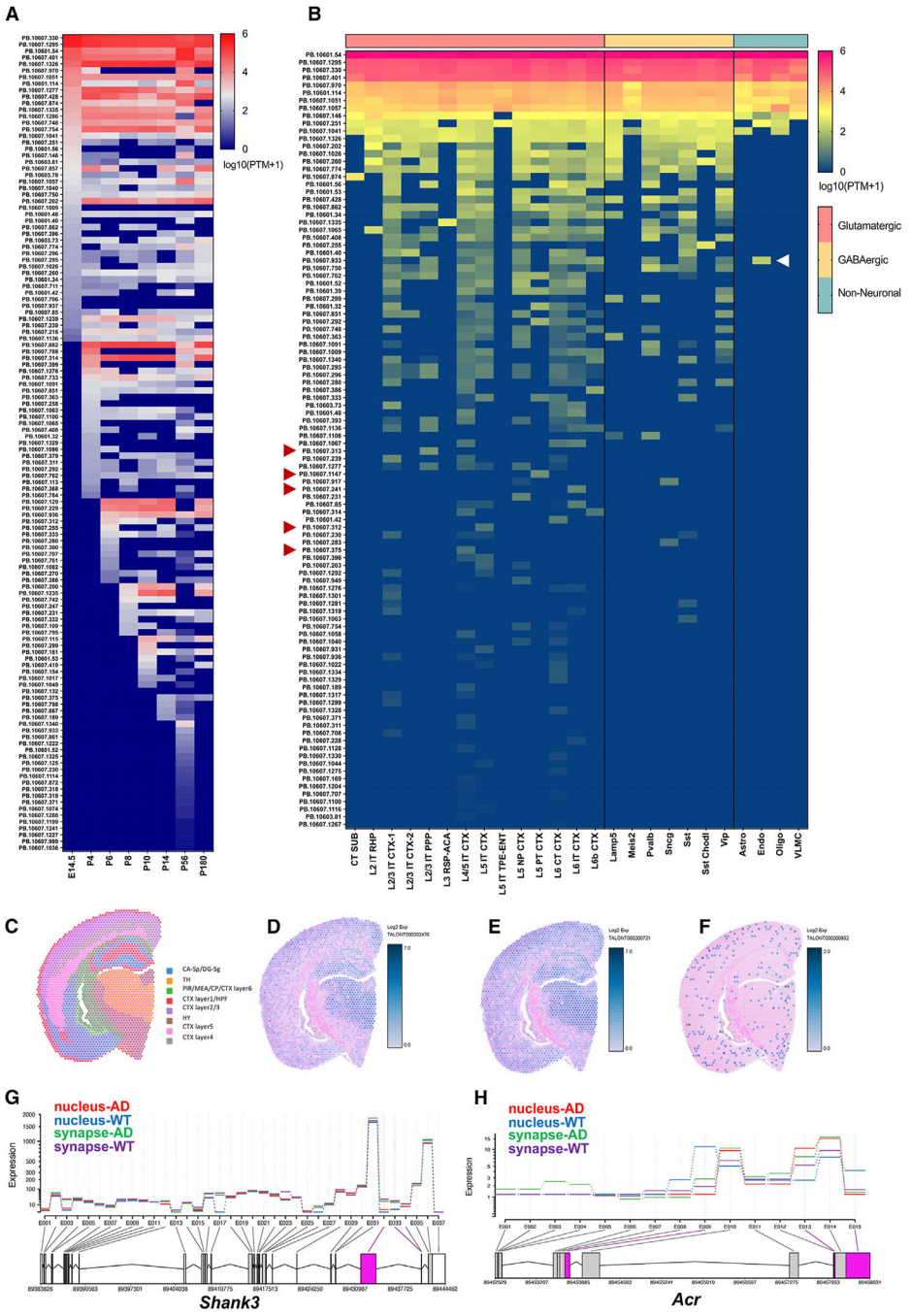

Figure 6. Development-, cell-type-, and cell-compartment-specific and spatial transcriptome of *Shank3* in mouse brains

(A) Development-specific *Shank3* transcripts in mouse cerebral cortex.

(B) Cell-type-specific *Shank3* transcripts in mouse brains. The scRNA-seq of the anterior cingulate cortex (ACA)[5] was aligned to *Shank3* transcripts detected by CIS. Glutamatergic neurons, especially the L2/3, L4/5, and L6 CTX, have more diverse *Shank3* transcripts compared to GABAergic neurons and non-neuronal cells. Certain transcripts were cell-type specific. The *Shank3* transcript (PB.10607.933) including exon 18 was detected only in endothelial cells.

(C–F) Mouse *Shank3* transcripts in the Visium spatial transcriptome. (C) Visium spatial anatomy (CA, cornu ammonis; DG, dentate gyrus; TH, thalamus; PIR, piriform cortex; MEA, medial amygdala; CP, choroid plexus; CTX, cortex; HPF, hippocampal formation; HY, hypothalamus).

(G) Cellular compartment-specific changes in *Shank3* exon usage in the hippocampus of an Alzheimer's disease (AD) mouse model from scRNA-seq data from different cellular compartments. The nucleus, compared to synapses, expressed significantly fewer splicing events of 32 and 33 that correspond to exon 21, the largest exon of mouse *Shank3*.

(H) Different patterns of *Shank3-Acr* fusion transcripts in nucleus and synapse between WT and AD mice.

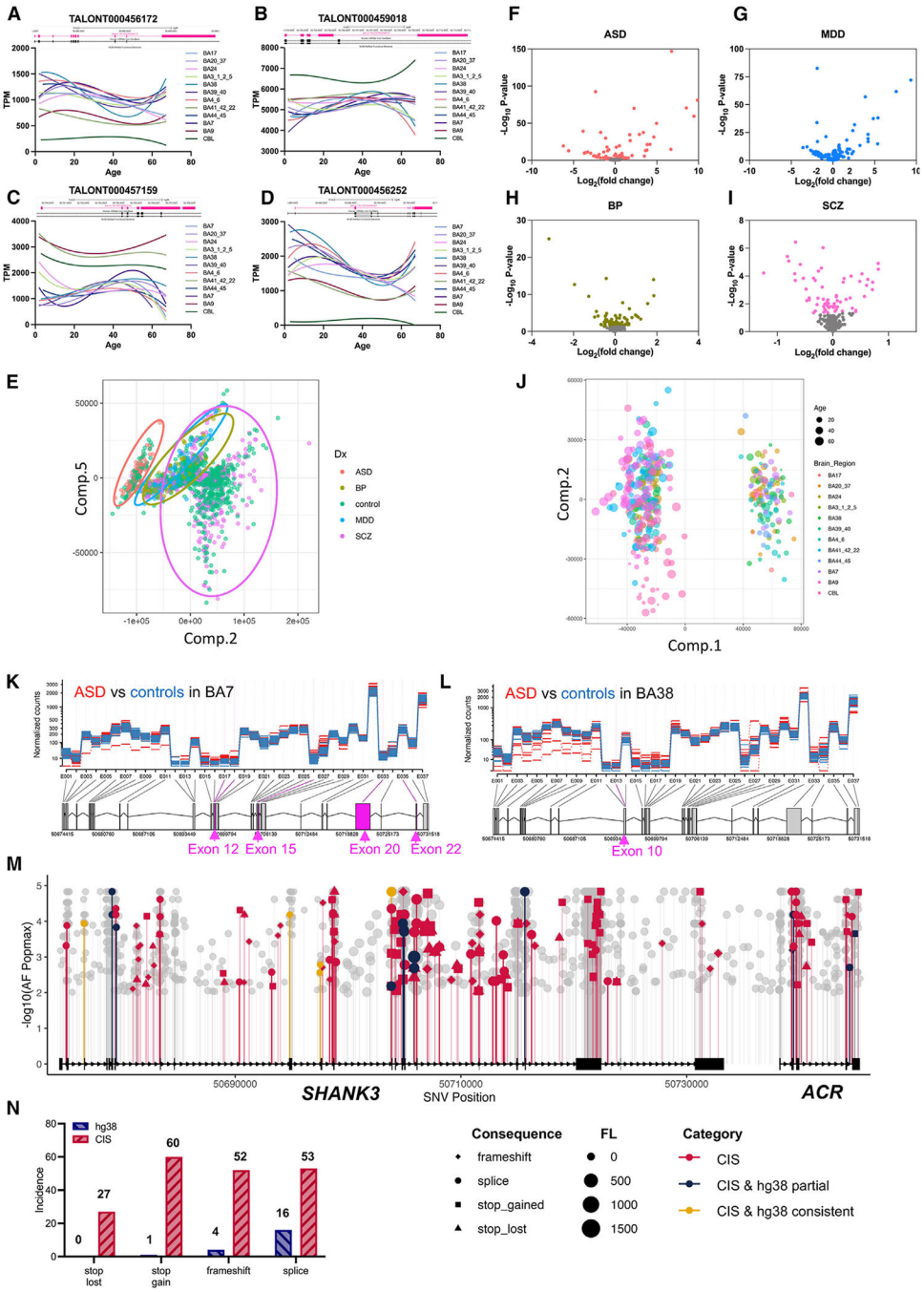

Figure 7. Improved transcriptome analysis of ASD transcriptome and sequence variant annotations of genome sequence data using the *SHANK3* transcript structure from CIS

(A–D) The patterns of human *SHANK3* transcripts from CIS changed at different ages and brain regions. Bulk RNA-seq data of normal controls was aligned to *SHANK3* transcripts detected using CIS (BA, Brodmann area; CBL, cerebellum).

(E–I) PCA of human *SHANK3* transcripts from CIS and bulk RNA-seq data of 2,474 cases with ASD, BPD, MDD, or SCZ, and normal controls from PsychENCODE (only data from prefrontal cortex are included). The clusters of MDD and BPD overlapped but are separate

from ASD and SCZ. (F–I) Volcano plots for individual disorders ASD (n = 68), MDD (n = 87), BPD (n = 297), and SCZ (n = 736) compared to controls (n = 1,286).
(J) PCA of *SHANK3* transcripts in different brain regions and ages (BA, Brodmann area; CBL, cerebellum).
(K and L) Brain-region-specific change in *SHANK3* transcripts in ASD brains. Bulk RNA-seq data of subregions of the brain from ASD and controls were aligned to *SHANK3* transcripts from CIS. (K) Exons 11, 15, 20, and 22 of *SHANK3* transcripts were significantly more represented in the BA7 region of ASD. (L) Exon 10 of *SHANK3* transcripts is significantly more represented in BA38 of ASD brain.
(M) Utilizing the updated *SHANK3* transcript structure from CIS enhanced PTV detection in ASD, SCZ, and BPD exome and genome sequencing data. From 55,000 cases, we identified 1,530 new PTVs, a significant increase from previous annotations using the *SHANK3* transcript NM_001372044.2 in hg38. Of these, 192 variants were likely deleterious, including 27 stop-loss, 60 stop-gain, 52 frameshift, and 53 splice variants, compared to the earlier finding of 22 such variants.
(N) The discovery rate of PTVs for *SHANK3* is increased from 1.3% using NM_001372044.2/hg38 as a reference to 12.5% using the transcript structure from CIS in this study.

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
Acrosin Polyclonal Antibody (C-terminus)	Invitrogen	Cat #: PA5–114207; RRID: AB_2884721
Acrosin Polyclonal Antibody (N-terminus)	Invitrogen	Cat #: PA5–99580; RRID: AB_2818513
Biological samples		
Human Brain Tissue	Mayo Clinic, Jacksonville, FL, 32224 USA	Adults, *n* = 4
Human Brain Tissue	Children's Hospital of Fudan University, Shanghai, 201102 China	Children, *n* = 4
Critical commercial assays		
NucleoZOL™	Takara Bio	Cat #:740404.200
NucleoSpin® RNA set for NucleoZOL™	Takara Bio	Cat #: 740406.50
rDNase Set	Takara Bio	Cat #: 740963
NucleoSpin® RNA Clean-up XS	Takara Bio	Cat #: 740903.50
RNA to cDNA EcoDry™ Premix kit	Takara Bio	Cat #: 639548
KAPA SYBR® FAST qPCR Master Mix (2X) Universal	Kapa Biosystems	Cat #: KK4602
xGen® Lockdown® Probes	Integrated DNA Technologies	customized
Iso-Seq Express Oligo Kit	PACBIO	PN 101–737–500
SMRTbell Express Template Prep Kit 2.0	PACBIO	PN 100–938–900
Elution Buffer	PACBIO	PN 101–633–500
NEBNext® Single Cell/Low Input cDNA Synthesis & Amplification Module	NEB	Cat#: E6421S
NEBNext® High-Fidelity 2X PCR Master Mix	NEB	Cat#: M0541S
ProNex® Beads	Promega	Cat #: NG2001
Qubit® dsDNA HS Assay Kit	Invitrogen	Cat #: Q32851
NucleoSpin® RNA/Protein Kit	Takara Bio	Cat #: 740933.50
Protein Quantification Assay	Takara Bio	Cat #: 740967.250
4x Laemmli buffer	Bio-Rad	Cat#: 1610747
4–20% Mini-PROTEAN® TGX Stain-Free™ Protein Gels	Bio-Rad	Cat #: 4568094

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Deposited data		
Iso-Seq/RNA-Seq	This study	BioProject: PRJNA1066952
UCSC tracks	This study	https://doi.org/10.17632/n5vjjmwb9p.1
Tandem Mass Spectrometry	This study	https://doi.org/10.17632/n5vjjmwb9p.1
UCLA-ASD RNA-Seq	PsychENCODE	syn4921369
BrainVar RNA-Seq	PsychENCODE	syn21557948
BrainGVEX RNA-Seq	PsychENCODE	syn4590909
CMC RNA-Seq	PsychENCODE	syn22344687
LIDB RNA-Seq	PsychENCODE	syn12299750
CommonMind RNA-Seq	PsychENCODE	syn2759792
SMART scRNA-seq	Yao et al.[40]	https://assets.nemoarchive.org/dat-jb2f34y
ASC VCF	Autism Sequencing Consortium	https://asc.broadinstitute.org/downloads
BipEx VCF	Bipolar Exomes	https://bipex.broadinstitute.org/downloads
Epi25 VCF	whole-exome sequencing case- control study of epilepsy	https://epi25.broadinstitute.org/downloads
SCHEMA VCF	Schizophrenia exome meta-analysis consortium	https://schema.broadinstitute.org/downloads
Mouse brain Visium	10x Genomics	FFPE Mouse Brain Coronal Section 1
Experimental models: Organisms/strains		
C57BL/6J	Jackson Laboratory	Strain #:000664
Shank3$^{\Delta e21}$	Jackson Laboratory	Shank3$^{tm1.1Pfw}$/J, Strain #:018398
Shank3$^{\Delta e4–9}$	Jiang Lab	Wang et al.[35]
Shank3$^{\Delta e4–22}$	Jiang Lab	Wang et al.[20]
Software and algorithms		
Lima v2.5.0	PACBIO	https://lima.how/
IsoSeq v3	PACBIO	https://github.com/PacificBiosciences/IsoSeq
pbmm2	PACBIO	https://github.com/PacificBiosciences/pbmm2
SQANTI3 v4.3	Pardo-Palacios et al.[107]	https://github.com/ConesaLab/SQANTI3
fastp	Chen et al.[108]	https://github.com/OpenGene/fastp

REAGENT or RESOURCE	SOURCE	IDENTIFIER
HISAT 2.2.1	Kim et al.[109]	https://daehwankimlab.github.io/hisat2/download/
SAMtools	Danecek et al.[110]	https://www.htslib.org/
FeatureCount	Liao et al.[111]	https://mnh.github.io/bioinfo-notebook/docs/featureCounts.html
Salmon 1.4.0	Patro et al.[112]	https://github.com/COMBINE-lab/salmon/releases
DEXSeq	Anders et al.[113]	https://bioconductor.org/packages/release/bioc/html/DEXSeq.html
BCFtools v 1.16	Danecek et al.[110]	https://bioconda.github.io/recipes/bcftools/README.html
HTSlib v 1.16	Bonfield et al.[114]	https://www.htslib.org/download/
Tabix v 0.2.5	Li et al.[115]	https://github.com/samtools/tabix
VEP, release 107	McLaren et al.[74]	https://github.com/Ensembl/ensembl-vep
gnomAD, v3.1.2	Koch et al.[75]	https://gnomad.broadinstitute.org/news/2021-10-gnomad-v3-1-2-minor-release/
SpliceAI	Jaganathan et al.[76]	https://github.com/Illumina/SpliceAI
SnpEff	Cingolani et al.[77]	https://pcingola.github.io/SnpEff/
Cell Ranger	10x Genomics	https://www.10xgenomics.com/support/software/cell-ranger/latest
Space Ranger v2.0	10x Genomics	https://www.10xgenomics.com/support/software/space-ranger/latest
Genomics Loupe Visualization Software v6.5	10x Genomics	https://www.10xgenomics.com/support/software/loupe-browser/latest