

GENETICS

Disruption of the autism gene and chromatin regulator KDM5A alters hippocampal cell identity

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Chromatin regulation plays a pivotal role in establishing and maintaining cellular identity and is one of the top pathways disrupted in autism spectrum disorder (ASD). The hippocampus, composed of distinct cell types, is often affected in patients with ASD. However, the specific hippocampal cell types and their transcriptional programs that are dysregulated in ASD are unknown. Using single-nucleus RNA sequencing, we show that the ASD gene, lysine demethylase 5A (*KDM5A*), regulates the development of specific subtypes of excitatory and inhibitory neurons. We found that *KDM5A* is essential for establishing hippocampal cell identity by controlling a differentiation switch early in development. Our findings define a role for the chromatin regulator *KDM5A* in establishing hippocampal cell identity and contribute to the emerging convergent mechanisms across ASD.

INTRODUCTION

Epigenetic chromatin remodeling is an essential step in cellular differentiation, and it is believed to dictate and maintain cellular identity (1). Studies have shown that undifferentiated cells have a globally “open” chromatin state (2, 3), which transitions to a more compact state as cells differentiate and acquire their identity (4). Chromatin regulation is required for proper gene expression and for brain development (5), with critical functions in establishing cell identities and the wiring of neuronal circuits (1, 6). It is also one of the top pathways disrupted in autism spectrum disorder (ASD) with causative mutations identified in several genes encoding chromatin remodelers (e.g., *ARID1B*, *CHD8*, and *SETD5*) (7–9).

Recent studies have shown that ASD genes are enriched in specific cell types in the brain during development (10, 11) and that changes in gene regulatory networks in ASD occur in a cell type-specific manner in the brain (12). The cellular landscape of the brain is extremely diverse, allowing for complex cognitive functions and behaviors. The diversity of the neuronal and nonneuronal cell types in the brain has been demonstrated at several levels including morphological (13), transcriptomic (14–17), epigenomic (18), as well as circuitry and electrophysiological properties (19). The cellular complexity of the brain increases within specific brain regions and reflects the specific functions of each region. The hippocampus, known for its role in memory formation (20), is characterized by a distinct cellular diversity mediating specific hippocampal functions (21, 22). It is composed of different regions that include the dentate gyrus and the cornu ammonis 1 (CA1), CA2, and CA3. Each of these regions has its own identity, unique properties, and specific cell types expressing region-specific markers. Single-cell sequencing studies have elucidated the different cellular subtypes in the hippocampus and identified their specific markers (16, 23, 24).

Multiple brain regions have been implicated in the pathogenesis of ASD, including the hippocampus (25). It has been shown that the hippocampus is impaired in individuals with ASD, both structurally (26, 27) and functionally (28, 29), including an abnormal pattern of hippocampal development with larger hippocampal volume and abnormal functional connectivity during learning in individuals with ASD compared to controls (26–29). However, the specific cell types in the hippocampus that are particularly affected in ASD are not known. We recently identified a disease gene, lysine demethylase 5A (*KDM5A*), with pathogenic variants in patients with ASD, lack of speech, and intellectual disability, among other phenotypes. In addition, we showed that complete loss of *KDM5A* in the constitutive knockout mouse model (*Kdm5a*^{-/-}) results in transcriptional dysregulation in the hippocampus and in neurobehavioral abnormalities, including severe deficits in social behavior, vocalization, learning, and memory, and increased repetitive behaviors, in addition to abnormal dendritic morphogenesis (30). *KDM5A* is a chromatin regulator that belongs to the *KDM5* family of lysine-specific histone H3 demethylases. Evidence from our studies demonstrates that *KDM5A* is critical for brain development and is essential for mediating complex behaviors including learning and memory (30). Murine *Kdm5a* is first detected at embryonic day 9.5 (E9.5) in the auditory pit (31) and at E10.5 in the brain (31). *KDM5A* is ubiquitously expressed across cell types in the brain (32), its expression pattern throughout development is dynamic, and it is differentially expressed at different stages of cell lineage birth and maturation (33). To investigate whether *KDM5A* has a cell type-specific function in the brain, we performed single-nucleus RNA sequencing (snRNA-seq) on hippocampal tissue from wild-type (WT) and *Kdm5a* knockout (*Kdm5a*^{-/-}) mice. We found a role for *KDM5A* in establishing identities of specific subtypes of hippocampal cells. Furthermore, we found that loss of *KDM5A* alters the developmental trajectory of hippocampal neurons and their transcriptional networks by altering the expression of common and unique sets of genes in the different cellular subtypes.

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RESULTS

Loss of KDM5A alters hippocampal cell composition

To investigate the effect of losing the chromatin regulator and ASD gene *KDM5A* on the development of the hippocampus, we profiled hippocampi from 20-week-old WT and *Kdm5a* knockout (*Kdm5a*^{-/-}) mice using snRNA-seq (Fig. 1A). We had previously generated and characterized the *Kdm5a* constitutive knockout mouse model (*Kdm5a*^{-/-}) and confirmed the complete loss of KDM5A transcript and protein in the *Kdm5a*^{-/-} (30). We sequenced a total of 105,086 nuclei that met our quality control measures (fig. S1), and we recapitulated all known major cell types in the hippocampus (Fig. 1, B and C). We assigned identities to the cell clusters by assessing the expression of known markers of the different hippocampal cell types (16, 23, 24, 34) and identified subclusters of hippocampal CA1, CA2, CA3, and dentate gyrus excitatory neurons, as well as inhibitory neurons, glia, neural stem cells, and endothelial clusters (Fig. 1D and fig. S2, A and C). We assessed the cellular composition of the sequenced cells and determined the percentage of each cell type. As expected, excitatory neurons represented the predominant cell type (73%) of all the sequenced cells (Fig. 1C). In addition, we confirmed that *Kdm5a* is ubiquitously expressed across all cell types in the hippocampus (fig. S2B).

Next, we compared the cell clusters between WT and *Kdm5a*^{-/-} hippocampi. We sequenced similar numbers of nuclei from WT (51,628) and *Kdm5a*^{-/-} (53,458) ($P = 0.9159$) hippocampi and found that the cell proportions of the major cell types (excitatory, inhibitory, glia, and endothelial) were similar among all replicates of the same genotype: four WT and four *Kdm5a*^{-/-} samples (Fig. 2A and fig. S3, A and B). We analyzed the proportions of WT and *Kdm5a*^{-/-} nuclei for each cluster and identified six distinct clusters that are most vulnerable to the loss of KDM5A: two inhibitory subclusters (Inh1 *Pvalb* positive and Inh2 *Sst* positive) and four excitatory subclusters (CA1.1, CA1.4, CA2, and CA3.4) (Fig. 2, B and C, and fig. S3C). We found that the two inhibitory subclusters (Inh1 and Inh2) and excitatory subclusters CA1.4 and CA3.4 have lower cell proportions from *Kdm5a*^{-/-} compared to WT. On the other hand, we found that subcluster CA1.1 has more cells from *Kdm5a*^{-/-} compared to WT (Fig. 2C and fig. S3C). Subcluster CA2 is composed of two spatially separated populations on the uniform manifold approximation and projection (UMAP) plot (Fig. 2B). Although we did not find a change in cell proportions for CA2 between genotypes (Fig. 2C and fig. S3C), the composition of the subcluster shifts in the *Kdm5a*^{-/-} compared to WT (Fig. 2, A and B). The six clusters that are vulnerable to the loss of KDM5A each express known markers of hippocampal excitatory neurons in the CA1 (*Man1a*, *Mpped1*, *Satb2*, and *Calb1*), CA2 (*Dcn*, *Prss23*, and *Scgn*), or CA3 (*Rnf182*) or inhibitory neurons (*Pvalb* and *Sst*) confirming their assigned identities (Fig. 2D).

KDM5A regulates common and unique genes in the vulnerable hippocampal cells

We assessed the transcriptional changes between WT and *Kdm5a*^{-/-} in the six affected clusters to understand the unique sensitivity of these clusters to the loss of KDM5A. Thousands of differentially expressed genes (DEGs) were identified in the six clusters of interest [table S1; false discovery rate (FDR)-corrected $P \leq 0.05$ and \log_2 fold change $\geq |0.3|$], some of which overlap between clusters indicating a common function of KDM5A in the different

hippocampal cell types (table S1). We defined a signature for KDM5A that is common across affected cell types, by combining the expression in the four affected excitatory clusters together (CA1.1, CA1.4, CA2, and CA3.4) and the two affected inhibitory clusters together (Inh1 and Inh2) and then comparing each to the expression of the combined unaffected excitatory and inhibitory clusters, respectively. We found that of the 2688 DEGs in the combined affected excitatory cluster, 196 are mouse homologs of known ASD genes [Fisher's exact test, $P < 2.2 \times 10^{-16}$, odds ratio (OR) = 2.48], and of the 2385 DEGs in the combined affected inhibitory cluster, 177 are mouse homologs of known ASD genes (Fisher's exact test, $P < 2.2 \times 10^{-16}$, OR = 2.50) (table S2). We identified 767 DEGs that are common among all four excitatory clusters irrespective of their direction of dysregulation (Fig. 3A, top). Of these, 24 genes are commonly up-regulated in all four excitatory clusters, 115 are commonly down-regulated in all four excitatory clusters, and the remaining 628 genes are dysregulated in opposite directions in at least one of the four clusters (Fig. 3A, bottom). We also identified 1428 DEGs that overlap between the two inhibitory clusters irrespective of their direction of dysregulation (Fig. 3A, top). Of these, 336 genes are commonly up-regulated in both inhibitory clusters, 529 genes are commonly down-regulated in both inhibitory clusters, and the remaining 563 genes are dysregulated in opposite directions in both clusters (Fig. 3A, bottom). Several DEGs in the six clusters of interest are mouse homologs of known ASD genes, many of which are common among all six clusters (e.g., *Astn2*, *Cntn4*, and *Gfap*) (table S1). Gene ontology (GO) analysis showed that these common DEGs are enriched for neurogenesis, neuron differentiation, axon guidance, dendritic development, and synaptic signaling (Fig. 3B). The data suggest a common role for KDM5A in regulating pathways underlying neuronal development in all six clusters of excitatory and inhibitory neurons. In addition to the common DEGs, we identified DEGs that are unique to each of the six clusters irrespective of their direction of dysregulation; there are 56 exclusive DEGs for CA1.1 that do not overlap with DEGs from any of the other clusters, 50 for CA1.4, 303 for CA2, and 75 for CA3.4. For the inhibitory clusters, 308 DEGs are unique to the parvalbumin positive (PVALB⁺) cluster, and 1046 are unique to the somatostatin positive (SST⁺) cluster (Fig. 3A, top). Considering directionality, we identified 568 genes that are exclusively up-regulated in CA1.1, 132 in CA1.4, 460 in CA2, 126 in CA3.4, 492 in PVALB⁺, and 506 in SST⁺ (Fig. 3A, bottom). In addition, we found 59 genes that are exclusively down-regulated in CA1.1, 338 in CA1.4, 497 in CA2, 346 in CA3.4, 379 in PVALB⁺, and 1103 in SST⁺. We performed GO analysis on the DEGs that are exclusive per cluster and found similar GO terms between the unique and the common DEGs (fig. S4). Thus, KDM5A regulates common and unique genes in the six subtypes of neurons that are most vulnerable to its loss. Although some of the dysregulated genes may be different in each of the cell types, the molecular pathways they regulate converge on essential neurodevelopmental processes, highlighting the importance of KDM5A in mediating neuronal development and function.

We examined the transcriptional changes between WT and *Kdm5a*^{-/-} in all the cells in order to compare the DEGs with previously published bulk RNA-seq data and publicly available chromatin immunoprecipitation-sequencing (ChIP-seq) data. We identified a total of 3099 DEGs between WT and *Kdm5a*^{-/-}, including 1881 up-regulated and 1218 down-regulated genes (table S1;

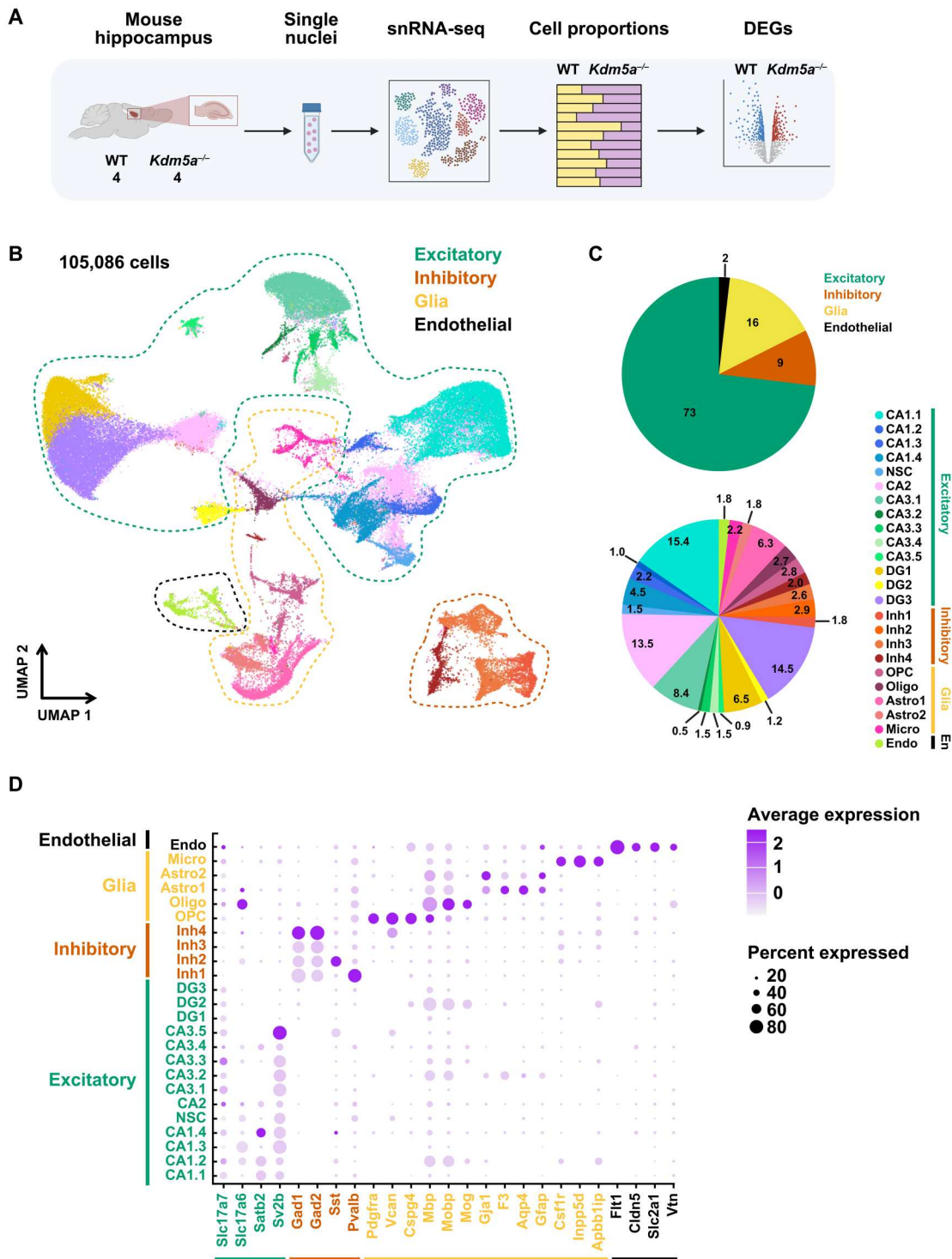


Fig. 1. snRNA-seq from WT and *Kdm5a*^{-/-} mice identifies 24 clusters of hippocampal cell types. (A) Schematic summary of the experimental and analytical workflow. The figure was created using BioRender. (B) UMAP plot of 105,086 nuclei from hippocampi of four WT and four *Kdm5a*^{-/-} mice (20 weeks old) colored by cluster identity and annotated according to known cell types. (C) Cell type composition across all cells from WT and *Kdm5a*^{-/-} hippocampi. Pie chart depicting the percentage of all cells belonging to each major cell type group (top). Pie chart depicting the percentage of all cells belonging to each cell type cluster (bottom). (D) Dot plot showing the relative expression of marker genes for major cell types: excitatory, inhibitory, glia, and endothelial. The dot size represents the percentage of nuclei expressing each marker gene, and the dot color intensity represents the average expression of the gene (light color, low expression; dark color, high expression).

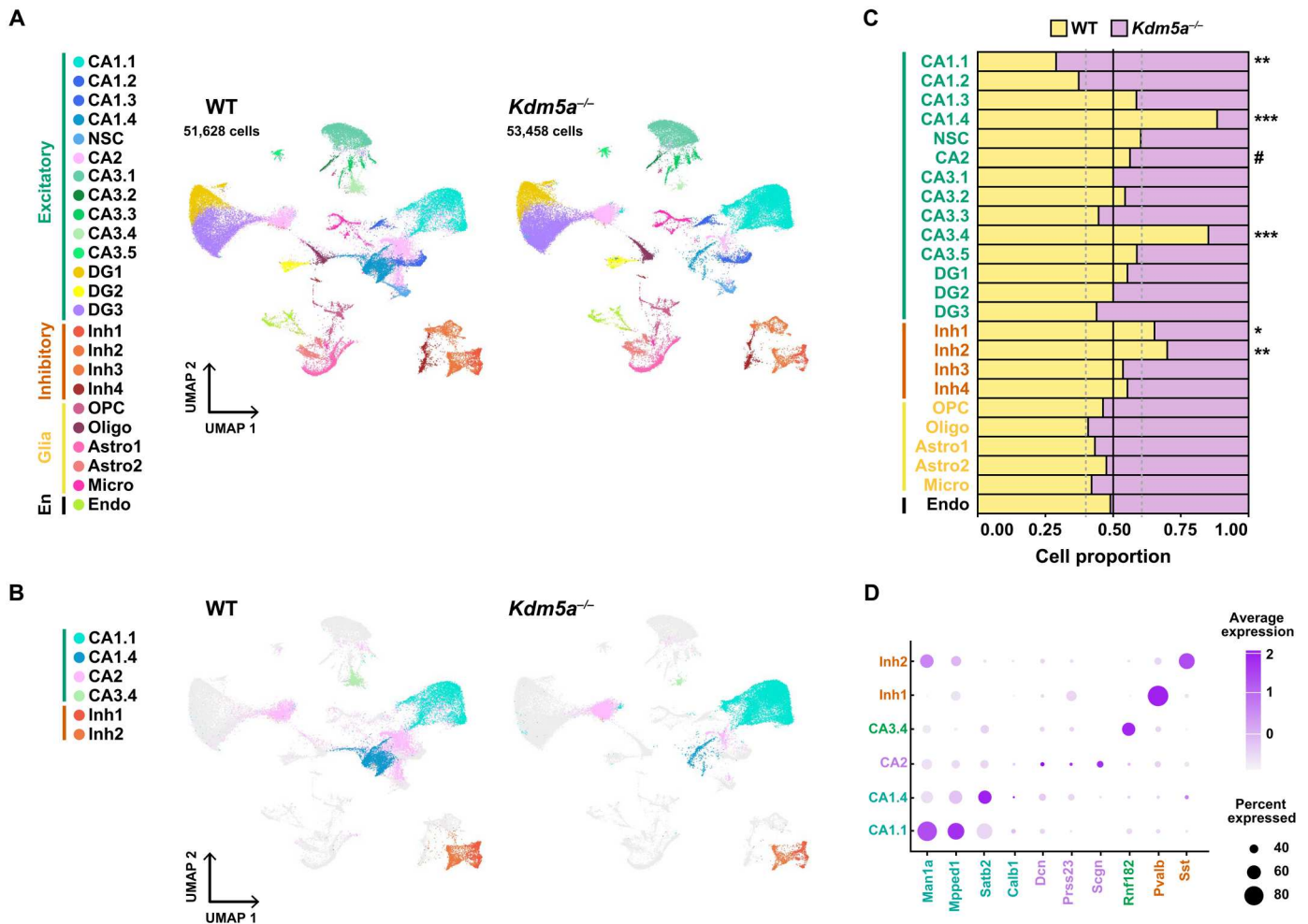


Fig. 2. Loss of KDM5A alters the composition of specific subclusters of hippocampal CA1, CA2, CA3, and inhibitory cell types. (A) UMAP plots of 51,628 nuclei from WT and 53,458 nuclei from *Kdm5a*^{-/-} hippocampi colored by cluster identity and annotated according to known cell identity markers. (B) UMAP plots of nuclei from WT and *Kdm5a*^{-/-} hippocampi colored by cluster identity and annotated according to known cell identity markers. Clusters with changing cell proportions in WT and *Kdm5a*^{-/-} hippocampi are colored. Clusters not changing between WT and *Kdm5a*^{-/-} hippocampi are in gray. (C) Proportion of cells in each cluster by genotype (CA1.1, ***P* = 0.0023; CA1.4, ****P* < 0.0001; CA2, #*P* = 0.3216; CA3.4, ****P* < 0.0001; Inh1, **P* = 0.0223; Inh2, ***P* = 0.004). Data were analyzed using chi-square test. (D) Dot plot showing the relative expression of marker genes for specific neuronal subtypes. The dot size represents the percentage of nuclei expressing each marker gene, and the dot color intensity represents the average expression of the gene (light color, low expression; dark color, high expression).

FDR-corrected $P \leq 0.05$ and \log_2 fold change $\geq |0.3|$). Of these DEGs, 200 are mouse homologs of known ASD genes (Fisher's exact test, $P < 2.2 \times 10^{-16}$, OR = 2.15) (table S1). We compared the DEGs to data from hippocampal bulk RNA-seq in WT and *Kdm5a*^{-/-} mice that we had previously generated (30) and found 104 common DEGs irrespective of their direction of dysregulation (Fisher's exact test, $P < 2.2 \times 10^{-16}$, OR = 2.96). Of these, 15 are commonly up-regulated genes (Fisher's exact test, $P = 0.2078$, OR = 1.41), 25 are commonly down-regulated genes (Fisher's exact test, $P = 1.33 \times 10^{-5}$, OR = 2.82), and the remaining 64 are dysregulated in opposite directions (fig. S5A and table S3). Furthermore, to identify potential direct targets of KDM5A, we compared the hippocampal bulk RNA-seq data to publicly available anti-KDM5A ChIP-seq data generated from mouse embryonic stem cells (35). We found 75 DEGs (59 up-regulated and 16 down-regulated) that overlap with KDM5A ChIP-seq peaks, suggesting that KDM5A directly binds and regulates the expression of these genes (fig. S5B and table S3).

Comparing these two datasets to the snRNA-seq dataset generated here identified three genes as high-priority possible direct targets of KDM5A: *Hapln1* and *Nptx2* are commonly up-regulated in the bulk RNA-seq and snRNA-seq datasets and have KDM5A ChIP-seq peaks; *Cmss1* is commonly down-regulated in the bulk RNA-seq and snRNA-seq datasets and has KDM5A ChIP-seq peaks (fig. S5C and table S3).

Kdm5a^{-/-} shows accelerated hippocampal development

We evaluated the developmental trajectories between WT and *Kdm5a*^{-/-} hippocampi through pseudotime analysis using Monocle (36). Integrating Monocle clustering with our cell type annotations identified two main clusters (Fig. 4, top and bottom) that had cells with a "neural stem cell" expression profile. The cells in these two clusters were used as starting points of the pseudotime trajectories. In both clusters, the WT cells are distributed throughout the entire trajectory, from early to later time points in

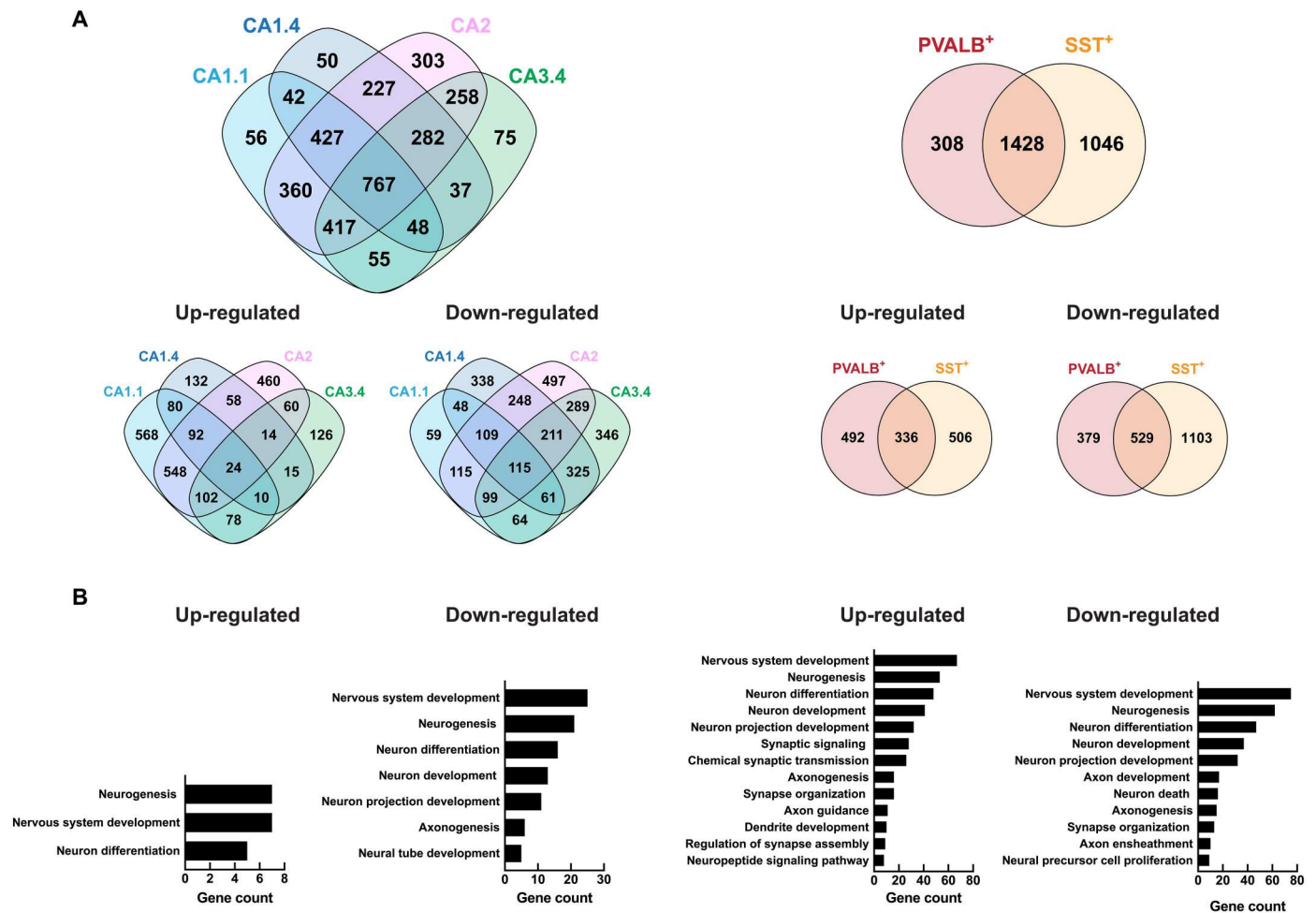


Fig. 3. Differential gene expression analysis reveals common function of KDM5A in the six vulnerable clusters. (A) Venn diagrams showing overlap of DEGs between WT and *Kdm5a*^{-/-} in excitatory (left) and inhibitory (right) clusters, irrespective of their direction of dysregulation (top). Overlap between commonly up-regulated DEGs and commonly down-regulated DEGs in all four excitatory (left) and inhibitory (right) clusters (bottom). (B) GO analysis of the DEGs that are commonly up-regulated or commonly down-regulated in excitatory (left) and inhibitory (right) clusters. Data are from DEGs with an FDR-corrected $P \leq 0.05$ and \log_2 fold change $\geq |0.3|$.

development (Fig. 4). However, for the *Kdm5a*^{-/-} cells, we identified a shift in the distribution of cells toward the end of the pseudotime trajectory compared to WT cells (Fig. 4). The shift of *Kdm5a*^{-/-} cells and their accumulation at the end of the developmental trajectories suggests an accelerated development and a more mature cellular identity in the *Kdm5a*^{-/-} hippocampus.

Given the accelerated developmental trajectory identified by pseudotime analysis, we analyzed hippocampal neuronal morphology to determine whether there are any differentiation defects following loss of KDM5A. We measured dendritic complexity and length in vivo by Golgi-Cox staining of brains from WT and *Kdm5a*^{-/-} mice. Sholl analysis revealed a significant increase in dendritic complexity (~116%) of hippocampal neurons from *Kdm5a*^{-/-} mice (Fig. 5). Furthermore, hippocampal neurons from *Kdm5a*^{-/-} mice have a significant increase in total dendritic length (~67%), while the mean dendritic length is unchanged compared with neurons from WT littermates (Fig. 5B). The data suggest an accelerated differentiation of hippocampal neurons upon loss of KDM5A.

Distinct CA2, CA3, and inhibitory cell populations are sensitive to KDM5A

CA2 is spatially divided into two subclusters on the UMAP (fig. S2A): One of them is marked by the expression of *Rps6*, encoding the 40S ribosomal protein S6, whose phosphorylation indicates neuronal activity during synaptic plasticity (37, 38); and the other is marked by the expression of *Nme2*, which encodes a nucleoside diphosphate kinase involved in neurite outgrowth inhibition (39). Loss of KDM5A showed a decrease in the *Rps6* cluster accompanied by an increase in the *Nme2* cluster (Fig. 2B and fig. S2A), suggesting a role for KDM5A in mediating CA2 cell identity. To validate our finding, we performed in situ hybridization (ISH) on brain sections from WT and *Kdm5a*^{-/-} mice for *Rps6* and *Nme2*. We found that in the CA2 region of the hippocampus, there is a decrease in the percentage of cells expressing *Rps6* (fig. S6, A and B) and an increase in the percentage of cells expressing *Nme2* (fig. S6, C and D) in the *Kdm5a*^{-/-} compared to WT, confirming the shift in CA2 cellular identity revealed by the snRNA-seq data.

Cluster CA3.4 is marked by the expression of *Shisa6* and *Smad3*, both genes encode molecules involved in regulating synaptic

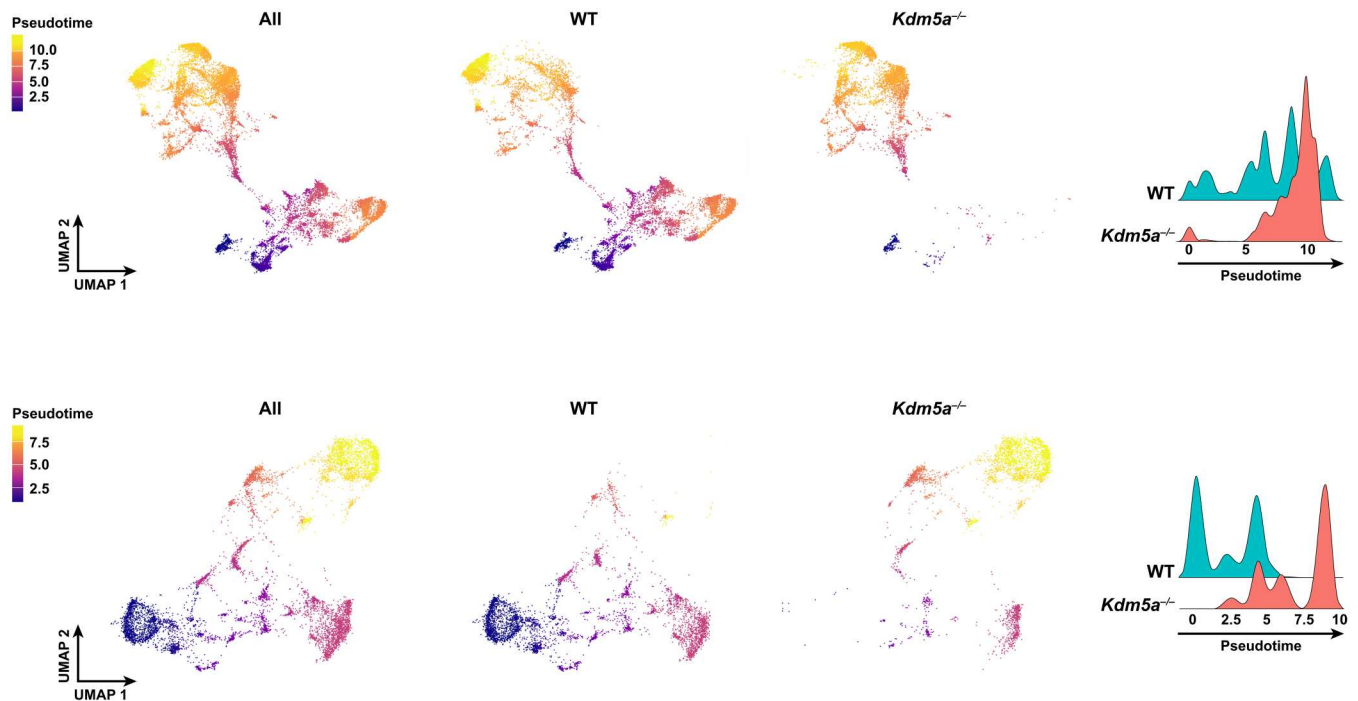


Fig. 4. Developmental trajectory tracing using pseudotime analysis reveals accelerated development and more mature cellular identity in *Kdm5a*^{-/-} compared to WT hippocampus. Pseudotime UMAP analysis of all (left), WT (middle), and *Kdm5a*^{-/-} (right) cells colored by pseudotime (early, blue; late, yellow) in the two Monocle clusters that contain neural stem cells. These cells were used as starting points of the pseudotime trajectory. Top and bottom: The developmental trajectory of cells in each of the two Monocle clusters.

transmission in the hippocampus (40, 41). CA3.4 is distinct from other CA3 clusters in that it expresses markers of mossy fibers like *Calb2*, *Thbs2*, *Drd2*, *Prrx1*, *Ttn*, *Crispld1*, and *Hk2* (24). In *Kdm5a*^{-/-}, there is a decrease in the CA3.4 cell proportion (Fig. 2C), suggesting that KDM5A is important in establishing the specific mossy-like CA3 cell identity. To confirm the decrease in *Shisa6*-marked CA3 cells, we performed ISH on brain sections from WT and *Kdm5a*^{-/-} mice and found that in the CA3 region of the hippocampus, the percentage of cells expressing *Shisa6* is decreased in *Kdm5a*^{-/-} compared to WT (fig. S6, E and F), validating the decrease in CA3.4 cluster in *Kdm5a*^{-/-} hippocampus.

The two inhibitory clusters sensitive to KDM5A loss express *Pvalb* or *Sst*, markers of parvalbumin and somatostatin interneurons, respectively (Fig. 2, C and D). These two interneuron subtypes play an essential role in regulating network activity in the hippocampus, and it has been recently shown that they undergo long-term inhibitory synaptic plasticity with a specific impact on CA1 pyramidal neurons (42). To validate the decrease in cell proportions of these two inhibitory subtypes in the *Kdm5a*^{-/-}, we performed single-molecule fluorescence ISH (smFISH) on brain sections from WT and *Kdm5a*^{-/-} mice. We found that in the hippocampus, the number of cells coexpressing the pan-inhibitory marker *Gad1* and *Pvalb* or coexpressing *Gad1* and *Sst* is decreased in the *Kdm5a*^{-/-} compared to WT, confirming our snRNA-seq data (fig. S7). This suggests an essential role for KDM5A in mediating the somatostatin and parvalbumin cellular identities in the hippocampus.

KDM5A functions early in development to specify proper CA1 cell identity

The unique increase in cell proportion in subcluster CA1.1 in the *Kdm5a*^{-/-} prompted us to further investigate the different CA1 subcluster identities. The CA1 layer of the hippocampus is divided into two sublayers, the deep CA1 and the superficial CA1, that are distinct in location (43), density (44), gene expression (45, 46), and functions (47, 48). The deep CA1 neurons develop at E12.5, while the superficial CA1 begins forming at E16.5 (49). To determine whether the two CA1 subclusters CA1.1 and CA1.4, which are sensitive to loss of KDM5A, correspond to either of the deep or superficial CA1 layers, we examined their developmental trajectories with pseudotime analysis. We found that CA1.1 has a more mature cellular identity compared to CA1.4 (fig. S8), suggesting that CA1.1 may correspond to deep CA1 cells, formed earlier in development, whereas CA1.4 may correspond to superficial CA1 cells, formed later in development. The shift in developmental trajectories of *Kdm5a*^{-/-} cells toward a more mature cell identity compared to WT (Fig. 4 and fig. S8), combined with the identified changes in cell proportions where CA1.1 is increased and CA1.4 is decreased in the *Kdm5a*^{-/-} compared to WT (Fig. 2C), indicates that the CA1.1 subcluster has a more mature identity and corresponds to deep CA1 cells that are increased in the *Kdm5a*^{-/-}, while the CA1.4 subcluster has a younger identity and corresponds to superficial CA1 cells that are decreased in the *Kdm5a*^{-/-}.

To better understand the changes in deep and superficial CA1 identities upon loss of KDM5A, we assessed the expression of known markers of superficial and deep cells in the two clusters, CA1.4 and CA1.1. *Satb2*, which encodes the transcriptional

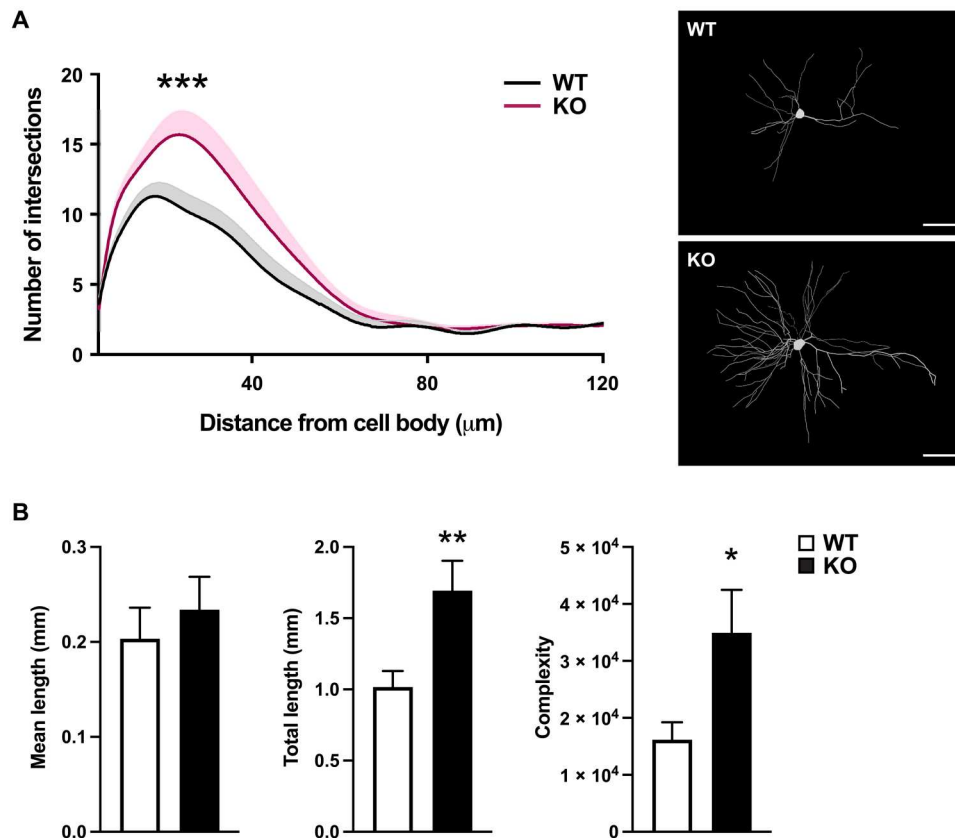


Fig. 5. *Kdm5a* knockout mice have impaired dendritic morphogenesis in the hippocampus. (A) Sholl analysis from Golgi-Cox–stained neurons revealed an increase in dendritic complexity of hippocampal CA1 neurons from *Kdm5a*^{−/−} (KO) mice (pink) compared to WT (black) littermates (***P* < 0.0001). Data were analyzed using two-way analysis of variance (ANOVA), followed by Tukey’s multiple comparisons test. Representative tracings of Golgi-Cox–stained hippocampal CA1 neurons (right). Scale bars, 20 μm . (B) Golgi-Cox staining showed significantly increased dendritic length (***P* = 0.0095) and complexity (**P* = 0.0338) of hippocampal CA1 neurons from *Kdm5a*^{−/−} mice compared to WT littermates. Mean dendritic length is unchanged (*P* = 0.5255). Data were obtained from mice at 14 to 16 weeks of age. Data were analyzed using unpaired *t* test. All values are means \pm SEM (WT, *n* = 3; *Kdm5a*^{−/−}, *n* = 3; WT neurons, *n* = 15; *Kdm5a*^{−/−} neurons, *n* = 15).

regulator special AT-rich sequence binding protein 2 (SATB2), known for its expression in the hippocampal CA1 neurons (50) and for its role in synaptic plasticity and memory formation (51), is also a known marker of superficial CA1 neurons (52). By filtering for genes unique to a cluster, we found that *Satb2* is exclusively expressed in the CA1.4 cluster, pointing to a superficial identity for CA1.4. To validate our findings, we performed ISH on brain sections from WT and *Kdm5a*^{−/−} for *Satb2* and found that in the *Kdm5a*^{−/−}, there is a decrease in the percentage of cells expressing *Satb2* in the CA1 compared to WT, confirming the superficial identity of the CA1.4 cluster (Fig. 6, A and B). In addition, we performed Western blot analysis on hippocampal tissue from WT and *Kdm5a*^{−/−} mice and found that SATB2 expression is decreased in the *Kdm5a*^{−/−} compared to WT (fig. S9). *Hs6st3*, which encodes a heparan sulfate sulfotransferase known to be involved in cellular proliferation and differentiation (53), was exclusively expressed in CA1.1. ISH on brain sections from WT and *Kdm5a*^{−/−} confirmed an increase in the percentage of cells expressing *Hs6st3* in the CA1 compared to WT (fig. S6, G and H). We further analyzed DEGs between WT and *Kdm5a*^{−/−} in cluster CA1.1 and found that markers of deep CA1 cells are up-regulated in *Kdm5a*^{−/−} compared to WT, including *Col11a1*, *Astn2* (52), and *Pvalb* (49) (table S1).

Col11a1 has been shown to distinctly mark the hippocampal CA1 deep layer (52), so to validate the up-regulation of *Col11a1* and confirm the CA1.1 deep cellular identity, we performed ISH on brain sections from WT and *Kdm5a*^{−/−} for *Col11a1*. We found an increase in the percentage of cells expressing *Col11a1* in the *Kdm5a*^{−/−} CA1 compared to WT (Fig. 6, C and D), which is in line with its up-regulation as a DEG and with the increased cellular proportion in CA1.1 in the *Kdm5a*^{−/−} compared to WT.

Together, the data demonstrate an increase in deep CA1 neurons and a decrease in superficial CA1 neurons in *Kdm5a*^{−/−} hippocampi compared to WT, consistent with the identified cell proportion differences in CA1.1 and CA1.4 between WT and *Kdm5a*^{−/−} (Fig. 2C) and the more mature and younger identities of CA1.1 and CA1.4, respectively (fig. S8). These findings are further supported by the accelerated differentiation phenotype of hippocampal CA1 neurons from the *Kdm5a*^{−/−} (Fig. 5). Our results indicate that KDM5A is required for appropriate development of CA1 deep and superficial neurons (Fig. 6E).

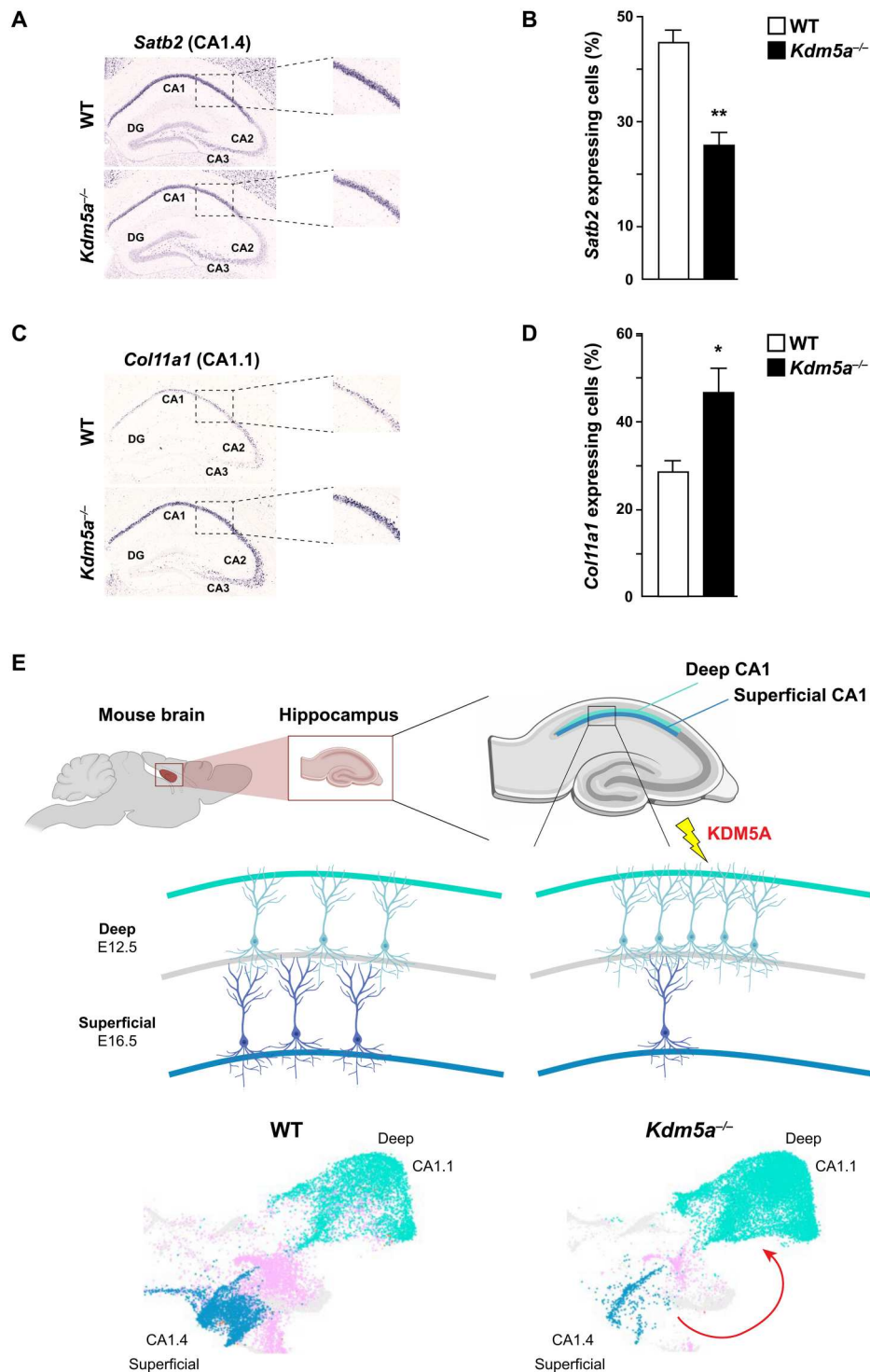


Fig. 6. KDM5A regulates the development of hippocampal CA1 superficial and deep neurons. (A to D) ISH on brain tissue for *Satb2* and *Col11a1* shows differential expression between WT and *Kdm5a*^{-/-}. Hippocampal sections from WT (top) and *Kdm5a*^{-/-} (bottom) mice show decreased expression of *Satb2* (A) and increased expression of *Col11a1* (C) in *Kdm5a*^{-/-} compared to WT, with the data quantified in (B) and (D), respectively [WT, *n* = 3; *Kdm5a*^{-/-}, *n* = 3; (B) ***P* = 0.0028 and (D) **P* = 0.0388]. Data were analyzed using unpaired *t* test. All values are means ± SEM. (E) Schematic model of how loss of KDM5A affects the development and appropriate cell identity specification of hippocampal CA1 superficial and deep neurons. Loss of KDM5A results in an increased cell proportion of deep CA1 neurons (CA1.1) and a decreased cell proportion of superficial CA1 neurons (CA1.4). The figure was created using BioRender.

DISCUSSION

The hippocampus, known for its essential roles in mediating cognitive abilities, is impaired in individuals with ASD, both structurally (26, 27) and functionally (28, 29), but the underlying defects at the single-cell resolution are unknown. Using snRNA-seq, we unraveled the role of the ASD gene and chromatin regulator KDM5A in mediating the development of specific excitatory and inhibitory neurons essential for proper hippocampal function. Subtypes of excitatory CA1 neurons, marked by the expression of *Satb2* (CA1.4), CA3 neurons expressing *Shisa6* and *Smad3* (CA3.4), and inhibitory neurons expressing *Pvalb* (Inh1) and *Sst* (Inh2) are decreased in mice lacking KDM5A. A subtype of CA1 neurons (CA1.1) is increased in the *Kdm5a*^{-/-} compared to WT, and a subtype of CA2 neurons changes identity from *Rps6* positive to *Nme2* positive.

Although all six cellular subtypes are sensitive to KDM5A levels and require its function for proper development, DEG analysis identified transcriptomic networks that are unique to each cluster, in addition to ones that are common between these clusters. Thus, the transcriptomic changes that these cells undergo when KDM5A is lost are in part similar indicating a common, broader role for KDM5A in these cell types and in part unique indicating a different specialized role for KDM5A in each of these subtypes. In addition, the DEGs common among these six clusters include homologs of known ASD genes and are enriched for neuron development, maturation, and differentiation terms, which is in line with the pseudotime analysis that suggests a role for KDM5A in neuronal development. By comparing the DEGs from the current snRNA-seq data to those from our previously published hippocampal bulk RNA-seq data and to KDM5A-bound genomic loci from published ChIP-seq data, we identified three genes that are common among these three datasets and that could be potential direct targets of KDM5A: *Cmss1*, *Hapln1*, and *Nptx2*. *Cmss1*, *cms1* ribosomal small subunit homolog, is expressed in the brain and has not been extensively studied, while *Hapln1*, hyaluronan and proteoglycan link protein 1, has recently been shown, alongside lumican and collagen I, to mediate folding of the developing human neocortex (54). The third high-priority potential KDM5A target is *Nptx2*, neuronal pentraxin 2, also known as *Narp*. It is an immediate early gene known for its roles in promoting neurite outgrowth (55), organizing excitatory synapses on hippocampal neurons, clustering AMPA-type glutamate receptors (56), and accumulating at excitatory synapses specifically on gamma-aminobutyric acid (GABA)ergic interneurons expressing parvalbumin (57). The human homolog has been linked to several disorders including epilepsy (55) and Alzheimer's disease (58). Given these functions, we hypothesize that the increase in dendritic complexity in the *Kdm5a*^{-/-} hippocampal neurons could be mediated by the up-regulation of *Nptx2*. Additional studies are needed to demonstrate whether KDM5A mediates a subset of its functions through its direct regulation of *Nptx2*.

Pseudotime analysis revealed that loss of KDM5A results in a shift toward a more mature cellular identity and a decrease in later developed hippocampal cells compared to WT. The data are in line with the increase in CA1.1 cells (earlier born) and the decrease in the CA1.4 cells (later born) that we identified as deep and superficial cells of the CA1, respectively, indicating a role for KDM5A in regulating neuronal development and differentiation in the hippocampus. In addition, assessing the neuronal arborization of hippocampal neurons showed a remarkable increase in

dendritic complexity and branching in the *Kdm5a*^{-/-} compared to WT, in line with the pseudotime results showing a more mature identity upon loss of KDM5A. Abnormal dendritic maturation has been a recurrent phenotype observed in ASD (59), and changes in either direction—more (60, 61) or less branching (62, 63)—are observed in ASD and can alter proper circuitry and brain function. We previously showed that upon loss of KDM5A, layer II/III pyramidal cortical neurons show a severe reduction in dendritic complexity (30). The two opposing effects induced by the loss of KDM5A in different brain regions suggest that although KDM5A is ubiquitously expressed in the brain, it could have not only common but also distinct functions that are brain region specific. Additional studies investigating the specific role of KDM5A in different brain regions are necessary to pinpoint the region-specific molecular and phenotypic defects observed in this subtype of ASD.

When KDM5A is lost, there is a shift in identity of the CA2 cellular subclusters, with a decrease in CA2 neurons expressing *Rps6*, a gene known for its function in synaptic plasticity (37, 38), and an increase in CA2 neurons expressing *Nme2*, a gene known for its role in neurite outgrowth inhibition (39). We also identified a role for KDM5A in mediating the development of a subtype of CA3 neurons; we found a decrease in the *Kdm5a*^{-/-} of a CA3 subtype that expresses *Shisa6* and markers of mossy fibers. This unique CA3 subtype affected by the loss of KDM5A shows a specificity for KDM5A function in the development of hippocampal *Shisa6*-expressing mossy-like CA3 neurons. CA2 neurons begin forming at around E10, and they are followed later by CA3 neurons that first develop at E11 (64, 65). Our data suggest a role for KDM5A at very early stages in hippocampal development, as early as E10, mediating the development of specific CA2 and CA3 neuronal subtypes. Our data also show a role for KDM5A in regulating the development of *Pvalb* and *Sst* expressing interneurons. In the hippocampus, these two subtypes of interneurons originate from the medial ganglionic eminence between E9 and E12, compared to other interneuron subtypes that originate from the caudal ganglionic eminence (66). This places the function of KDM5A at this time window in development because these two specific subtypes of interneurons are sensitive to its loss compared to other subtypes. In addition, synapses formed by these two subtypes of interneurons in the hippocampus are regulated through two distinct pathways, and they each undergo distinct synaptic plasticity mechanisms (42, 67). These interneuron subtype-specific synaptic plasticity mechanisms have a role in regulating output from the CA1 pyramidal neurons (42, 67), so a decrease in the proportion of these cells in the *Kdm5a*^{-/-} hippocampus can potentially have a damaging impact on hippocampal circuits and their function. Last, changes in cell proportions of excitatory and inhibitory neurons can result in a dysregulation of the excitatory-inhibitory balance, a neurodevelopmental disease mechanism frequently observed in ASD (68).

The CA1 principal neurons (CA1PNs) give rise to the CA1 layer of the hippocampus, which is divided into two functionally distinct sublayers, the deep and the superficial (49). Here, we identified a role for KDM5A in establishing deep and superficial CA1 cellular identities. Our data show that upon loss of KDM5A, the superficial CA1 neurons that are born at E16.5 do not develop properly. We find that in the *Kdm5a*^{-/-}, there is a decrease in the proportion of superficial CA1 neurons and that more CA1PNs develop into deep CA1 neurons instead. This suggests that KDM5A is essential at E16.5 for the establishment of the superficial CA1 cellular

identity. The superficial CA1 neurons are marked by the expression of *Satb2* (52), which encodes a transcriptional and chromatin regulator widely expressed in the brain and specifically in CA1 neurons of the hippocampus (50). It has essential roles in long-term memory formation in the adult hippocampus (51), short-term working memory, hippocampus-dependent spatial memory, and synaptic plasticity (69). *SATB2* haploinsufficiency results in a neurodevelopmental disorder characterized by intellectual disability, limited to absent speech, and abnormal behaviors (70, 71). It is worth noting that in addition to ASD, patients with pathogenic *KDM5A* mutations have intellectual disability and absent speech (30). Our previous work has demonstrated that loss of *Kdm5a* leads to severe impairments in spatial learning and memory (30). The deep and superficial CA1 neurons have distinct functions (47–49, 72, 73), and recent studies have expanded on the unique roles of these sublayers by showing that the CA1 deep neurons are involved in reward configuration, while the CA1 superficial neurons drive memory formation (74). Given all the aforementioned data, the loss of CA1 superficial neurons marked by the expression of *Satb2* in *Kdm5a*^{-/-} is predicted to be one major factor leading to the severe cognitive deficits observed upon loss of KDM5A (30). This finding uncovers a key role for KDM5A in regulating the superficial CA1 neurons and in mediating learning and memory.

Our study identifies a critical role for the ASD gene and chromatin regulator KDM5A in regulating hippocampal cell type identity of specific excitatory and inhibitory neurons. Charting the cell type-specific changes that ensue upon loss of this neurodevelopmental disease gene enabled the dissection of molecular changes that underly disease pathogenesis and instructed us on hippocampal development and how chromatin regulation dictates cellular identity. We identify multiple transcriptionally distinct neuronal subtypes of the CA1, CA2, CA3, and inhibitory neurons, further highlighting the rich cellular diversity of these hippocampal regions, in line with recent studies (16, 23, 24, 34). Although much remains to be learned about the different CA1, CA2, and CA3 neuronal subtypes, our results identify the role of an ASD gene in the development of superficial and deep CA1 neurons, in addition to other roles in specific excitatory and inhibitory neuronal populations of the hippocampus. We show that different neuronal subtypes of the CA are differentially affected by loss of KDM5A, which demonstrates that these different subtypes are not only transcriptionally distinct, but that they respond differently to the same genetic defect. This suggests that the different identified subtypes are likely to mediate different functions in the hippocampus, as in the case of the deep and superficial CA1 neurons (47, 48). Further studies to characterize the function of KDM5A in other brain regions will identify how different cellular subtypes respond to the loss of this chromatin regulator and will highlight points of convergence and divergence among the downstream molecular networks it regulates in different cell types.

Loss of KDM5A leads to a more mature cellular identity in the hippocampus, which provides *in vivo* evidence and complements recent studies showing that chromatin regulators, notably KDM1A, function as an “epigenetic barrier” during neuronal maturation, and inhibiting them leads to an accelerated neuronal maturation (75, 76). Our data suggest that KDM5A may be a player in this epigenetic barrier, regulating transcriptional programs essential for neuronal development and proper maturation. Further studies aimed to identify the molecular function of KDM5A in regulating neuronal maturation will be crucial to our understanding of how

chromatin regulators contribute to neuronal differentiation and maturation. In addition, investigating the specific cells, neuronal networks, and molecular pathways that are affected in this genetic subtype of ASD and uncovering the specific cell types that mediate the different ASD phenotype(s) will inform future studies aimed at targeting these cell types for therapies. The information gained will enable the identification of mechanisms that are convergent among different genetic subtypes of ASD where other chromatin regulators and remodelers are mutated.

MATERIALS AND METHODS

Animals

All animal care and use procedures were approved by the University of Texas Southwestern Medical Center (UTSW) Institutional Animal Care and Use Committee (protocol number 2017-102300) and were compliant with U.S. government principles about the care and use of animals, Public Health Service Policy on Humane Care and Use of Laboratory Animals, Guide for the Care and Use of Laboratory Animals, and the Animal Welfare Act. Animal husbandry was performed in the UTSW animal facility, accredited by the Association for Assessment and Accreditation of Laboratory Animal Care International. The *Kdm5a* constitutive knockout mouse model (*Kdm5a*^{-/-}) was previously generated and characterized in El Hayek *et al.* (30). We confirmed the complete loss of KDM5A transcript and protein in the *Kdm5a*^{-/-} (30).

Tissue collection and nucleus isolation

The WT and *Kdm5a*^{-/-} hippocampi from 20-week-old mice were processed together and on the same day. Nucleus isolation was adopted and modified from Habib *et al.* (77). Mice were euthanized with CO₂, and the brains were dissected on ice. First, the brain was oriented so that its posterior side is closer to the experimenter. Next, the brain was cut sagittally into its two hemispheres, and the hippocampi were isolated from each hemisphere. The hippocampus was oriented like the brain, where the posterior end was closer to the experimenter. Using a 1-mm-diameter tissue puncher adapted to a syringe and placed at the center of the hippocampus, a tissue punch was collected capturing the entirety of the hippocampus. The punches were immediately stored at -80°C. For each animal, tissue punches from the right and left hippocampi were combined and homogenized using a glass Dounce homogenizer in 500 μ l of ice-cold nucleus EZ lysis buffer (NLB) (NUC-101, Sigma-Aldrich) 25 times with pestle A and then 25 times with pestle B. The mixture was transferred to a prechilled Eppendorf tube, and 1000 μ l of NLB was added to the tube and incubated on ice for 5 min. The tubes were centrifuged at 800g for 5 min at 4°C. The nuclei were resuspended in 75 μ l of nucleus suspension buffer [consisting of 1 \times phosphate-buffered saline (PBS), 1% bovine serum albumin (AM2618, Thermo Fisher Scientific), and 1% ribonuclease inhibitor (0.2 U/ μ l; AM2694, Thermo Fisher Scientific)]. The nuclei were then filtered through a 40- μ m Flowmi cell strainer (H13680-0040, Bel-Art) to remove cellular debris. Concentration of nuclei for each sample was calculated using Trypan Blue (T8154, Sigma-Aldrich) on a hemocytometer, aiming for a final concentration of ~1000 to 2000 nuclei/ μ l.

Single-nucleus RNA library preparation and sequencing

Single nuclei were processed using the 10X Genomics Next GEM Single Cell 3' Reagent Kit v3.1 following the manufacturer's user guide. All mRNA samples were examined for quantity and quality by NanoDrop and Bioanalyzer 2100 (Agilent). The libraries were constructed following TruSeq Stranded mRNA Sample preparation guide (Illumina), and paired-end sequencing was performed on the Illumina NovaSeq 6000 platform at the McDermott Center Next Generation Sequencing Core at UTSW.

snRNA-seq data analysis

Data were stored, and analyses were performed on the Texas Advanced Computing Center high-performance computing servers, a resource of the University of Texas at Austin (Austin, TX). Raw sequencing data were converted to fastq format using the Cell Ranger (78) mkfastq command (10X Genomics, v1.1.0). Sequencing reads were aligned to the mm10 mouse reference genome and quantified using Cell Ranger (78) count (10X Genomics, v1.1.0). We used the R package Seurat (v4.1.0) (79) for further filtering and analysis. Nuclei with greater than 300 genes expressed, greater than or equal to 500 unique molecular identifiers, and less than or equal to 0.4% of counts corresponding to mitochondrial genes were retained for downstream analysis.

SCTransform, with the glmGamPoi package (80), was applied to each sample to normalize and stabilize the variance of molecular count data. Results were saved in a new assay (SCT) with corrected counts, log1p counts, and Pearson residuals. SelectIntegrationFeatures was used to identify features that are repeatedly variable across samples for integration. The number of features requested was 3500. These features were passed to PrepSCTIntegration to prepare for integration of the samples. Anchors between the samples were identified, and these anchors were used to integrate the samples using IntegrateData, creating the "integrated" assay. Principal components analysis dimensionality reduction and UMAP dimensionality reduction were performed on this assay. The top 30 principal components were retained for further analysis. A shared nearest-neighbor graph was constructed using the FindNeighbors function with $k = 20$. Twenty-four clusters were identified using Louvain clustering in the FindClusters function at a resolution of 0.3. Two of these clusters were merged because they had similar marker genes and no distinctive markers unique to each cluster. One cluster was split into two unique clusters because each exclusively expressed either *Pvalb* or *Sst*, markers of parvalbumin and somatostatin cells, respectively.

Differentially expressed genes

FindAllMarkers was used to calculate the expression level in a single cluster versus the average expression in all other clusters, expression level of *Kdm5a*^{-/-} cells versus WT cells across all clusters, and expression level of *Kdm5a*^{-/-} cells versus WT cells within each cluster. We used MAST v1.16.0 (81) with a generalized linear model framework and selected genes with a log₂ fold change of expression of $\geq|0.3|$ and FDR-corrected $P \leq 0.05$ as DEGs. The list of known ASD genes was obtained from the Simons Foundation Autism Research Initiative Gene 2018 database (82) (using the 2022 Q2 release). The genes that were differentially expressed in *Kdm5a*^{-/-} cells versus WT cells across all cells were overlapped with DEGs identified through bulk RNA-seq on WT and *Kdm5a*^{-/-} hippocampi from El Hayek *et al.* (30).

To define a KDM5A signature that is common across affected cell types from the six vulnerable clusters (CA1.1, CA1.4, CA2, CA3.4, Inh1, and Inh2), the expression in all affected excitatory clusters (CA1.1, CA1.4, CA2, and CA3.4) was combined. Similarly, the expression in both affected inhibitory clusters (Inh1 and Inh2) was combined. The FindMarkers function was used to compare the expression level in the combined excitatory clusters versus the average expression in all other excitatory clusters [except the neural stem cell (NSC) cluster]. It was also used to compare the expression level in the combined inhibitory clusters versus the average expression in all other inhibitory clusters. Genes with a log₂ fold change of expression of $\geq|0.3|$ and FDR-corrected $P \leq 0.05$ were selected as the KDM5A signature genes.

Analysis of candidate KDM5A targets

To nominate possible direct and indirect KDM5A targets, anti-KDM5A ChIP-seq data generated from mouse embryonic stem cells from Beshiri *et al.* (35) were downloaded from the National Center for Biotechnology Information (NCBI) Gene Expression Omnibus (GEO) repository (accession number GSE28343). The LiftOver tool from the University of California Santa Cruz Genome Browser (83) was used to convert the peak bed file from the mouse reference genome mm9 to the mm10 assembly. ChIP-seeker (84, 85) was used to annotate the genomic region of the peaks. The nearest genes around the peaks were overlapped with DEGs identified through hippocampal bulk RNA-seq from El Hayek *et al.* (30) and the identified snRNA-seq DEGs from this study. DEGs from the bulk and snRNA-seq data were overlapped with the annotated ChIP-seq peaks to nominate possible direct targets. The remaining DEGs were considered indirect targets. Common genes between the bulk and snRNA-seq data that overlapped with the annotated ChIP-seq peaks were considered high-priority candidate direct targets.

GO analysis

We used the functional annotation tool DAVID (86) to find biological process terms enriched in the DEG sets.

Pseudotime analysis

Monocle3 (v1.2.7) (36) was used for pseudotime analysis. A cell dataset object was created using RNA counts. Cells were clustered using Leiden community detection. Partitions were also calculated using a kNN pruning method. Genes most specifically expressed in each partition were identified, and trajectories were built from the reduced dimension space using reversed graph embedding. Identification of neural stem cells from our Seurat (79) clustering was used to locate neural stem cells in the Monocle partitions. Cells that had an identity of neural stem cells were primarily located in two Monocle clusters (Fig. 4). Cells in these two different clusters were used as trajectory starting points or nodes (Fig. 4). The pseudotime data from one of the Monocle clusters that contained most of the CA1 cells were added back to the Seurat object and visualized on the original UMAP (fig. S8).

Golgi-Cox staining

Mice at 14 to 16 weeks of age were euthanized, and brains were dissected out. Golgi-Cox staining was performed as described in Zaout and Kaindl (87). The brains were cut into two hemispheres and impregnated with Golgi-Cox solution (1% K₂Cr₂O₇, 1% HgCl₂,

and 0.8% K_2CrO_4) at room temperature for 7 days in the dark. Subsequently, they were protected with tissue protectant solution [30% sucrose, 1% PVP40 (polyvinylpyrrolidone), and 30% ethylene glycol in 0.05 M phosphate buffer (pH 7.2)] at 4°C for 24 hours after which the solution was replaced with a fresh one and kept at 4°C for 5 days. Brains were embedded in optimal cutting temperature (OCT) medium. Frozen sagittal sections (150 μ m in thickness) were prepared using a microtome (Leica) and loaded on 3% gelatin-coated glass slides. The slides were dried for 5 days at room temperature, dehydrated, and developed as described in Zaqout and Kaindl (87) and mounted with Cytoseal (Thermo Fisher Scientific). Bright-field images were acquired using Nikon Fi3 camera with 60 \times objective at 0.9- μ m steps in *z* axis. Dendritic and spine analyses were performed using Neurolucida 360 software (MBF Bioscience) at the UTSW Whole Brain Microscopy Facility. For Sholl analysis, images were processed and analyzed in Fiji (ImageJ v2.14.0/1.54f) (88).

Western blot analysis and antibodies

To detect endogenous proteins, hippocampal tissue was isolated from male and female mice at 18 weeks of age and immediately frozen in liquid nitrogen. From each animal, both hippocampi, the left and right, were isolated and lysed in SDS lysis buffer [250 mM tris (pH 6.8), 4% SDS, 3.2% glycerol, 10 mM *N*-ethylmaleimide, and 1 mM phenylmethylsulfonyl fluoride] supplemented with 2 mM OPT, and cOmplete mini EDTA-free protease inhibitor and phosSTOP phosphatase inhibitor tablets (Roche). Samples were boiled for 10 min and then centrifuged at 14,000 rpm for 15 min, and the supernatant was collected. Protein concentrations were determined using the DC protein assay (Bio-Rad). Seventy micrograms of protein was loaded per lane, with β -mercaptoethanol and bromophenol blue, onto 8% polyacrylamide gels. Gels were run, and protein was transferred to polyvinylidene difluoride membranes (Millipore) for 2 hours on ice. Membranes were blocked in 5% milk in TBS-T [20 mM tris (pH 7.5), 150 mM NaCl, and 0.1% Tween-20] for 1 hour at room temperature and incubated with primary antibody overnight at 4°C. Membranes were washed in 5% TBS-T, followed by a 1-hour and 30-min incubation in secondary antibody (1:10,000) at room temperature using donkey anti-rabbit (711-035-152, Jackson ImmunoResearch). Following antibody incubation, signal was detected with enhanced chemiluminescence (SuperSignal West Pico chemiluminescent substrate, Thermo Fisher Scientific). Primary antibodies used were SATB2 (ab92446, Abcam; 1:1000) and β -actin (13E5, Cell Signaling Technology; 1:1000). Bands were quantified using Fiji (ImageJ v2.14.0/1.54f) (88).

Single-molecule fluorescent ISH

Mice (30 to 35 weeks of age) were euthanized with a ketamine/xylozine cocktail, followed by perfusion with cold 1 \times PBS for 3 min and then with cold 4% paraformaldehyde (PFA) in PBS for 15 min at a rate of 8 ml/min. Brains were then dissected out and washed in 10 ml of 1 \times PBS, followed by an overnight incubation in 4% PFA at 4°C. The next day, the brains were cryoprotected by immersing them in sucrose solutions with gradually increasing concentrations (from 5 to 25%) until they sink. They were next embedded in OCT and allowed to freeze overnight at -80°C . The brains were sectioned at -20°C using a cryostat at 16 μ m in thickness by the UTSW Histo Pathology Core. The sections were mounted on

Superfrost Plus microscope slides. smFISH was performed using RNAscope Multiplex Fluorescent Reagent Kit v2 (323100) following the manufacturer's instructions for fixed-frozen tissue samples. The target retrieval was performed manually following the manufacturer's instructions. Three probes were used: *Gad1* mouse 400951-C1, *Pvalb* mouse 421931-C2, and *Sst* mouse 404631-C3. Three Opal fluorophores were used: Opal 520 (FP1487001KT, Akoya Biosciences; 1:750) for C1, Opal 570 (FP1488001KT, Akoya Biosciences; 1:750) for C2, and Opal 620 (FP1495001KT, Akoya Biosciences; 1:750) for C3. Mouse-positive (RNAscope 3-plex 320881) and mouse-negative (RNAscope 3-plex 320871) control probes were also used. Slides were imaged at $\times 20$ magnification in *z* stacks using Zeiss LSM 880 confocal microscope at the UTSW Neuroscience Microscopy Facility. Images of hippocampal sections from $n = 3$ mice per genotype were analyzed using Fiji (ImageJ v2.14.0/1.54f) (88). For each slice, we quantified the number of fluorescent puncta that are *Gad1*⁺, *Gad1*⁺*Pvalb*⁺, and *Gad1*⁺*Sst*⁺. The data were plotted as the number of *Gad1*⁺*Pvalb*⁺ or *Gad1*⁺*Sst*⁺ puncta over the total number of *Gad1*⁺ puncta.

ISH, microscopy, and image analysis

Mice (8 to 10 weeks of age) were euthanized with CO₂, and the brains were collected on ice. The brains were immediately placed in freezing blocks filled with OCT and left on dry ice to freeze and then stored in clear plastic wrap at -80°C . The OCT blocks were equilibrated at -20°C before sectioning using Leica CM3050S cryostat. Sagittal sections of 25 μ m in thickness were collected on slides then fixed in 4% PFA (15 to 30 min), followed by fresh 0.25% (v/v) acetic anhydride in 0.1 M triethanolamine (pH 8.0) (5 min, $\times 2$), and dehydrated in the Leica autostainer XL in an ethanol sequence reaching 100% ethanol and then air-dried.

ISH was performed at the Baylor College of Medicine RNA ISH Core facility as previously described (89). Sequences for the ISH probes used (*Col11a1*, *Hs6st3*, *Nme2*, *Rps6*, *Satb2*, and *Shisa6*) are available through the Allen Brain Atlas Mouse Brain Data Portal (<https://mouse.brain-map.org/>). Images were collected using the slide scanner Zeiss Axioscan.Z1, 20 \times 0.8 numerical aperture lens, and automated quantification was performed as previously described (90). Briefly, a Python script was used to identify the location and signal strength of each gene within the area of interest. Total cells and cells expressing the gene of interest were quantified, and for each gene, the percentage of cells expressing the gene out of the total cells within the corresponding hippocampal region was plotted.

Supplementary Materials

This PDF file includes:

Figs. S1 to S9

Legends for tables S1 to S3

Other Supplementary Material for this manuscript includes the following:

Tables S1 to S3

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