# A gain-of-function recurrent missense variant leads to a GABAergic/glutamatergic imbalance in a forebrain organoid model of PACS1 syndrome

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**Abstract**

PACS1 syndrome is a neurodevelopmental disorder characterized by intellectual disability and craniofacial abnormalities resulting from a *de novo* p.R203W variant in phosphofurin acidic cluster sorting protein 1 (PACS1). PACS1 is known to play roles in both the endosomal pathway and nucleus, but little is known about how this variant affects the developing nervous system and patients have few therapeutic options. Here, we used stem cell-derived forebrain organoids to show that PACS1$^{(+/R203W)}$ cells share convergent molecular features of autism spectrum disorder (ASD) through a gain-of-function mechanism. These features include an aberrant propensity towards GABAergic differentiation, increased inhibitory synaptic density, and impaired expression of an ASD-specific gene network enriched at synapses. This work is the first to investigate the impact of the p.R203W variant on the developing brain and suggests a therapy that either clears p.R203W PACS1 or targets convergent pathological mechanisms of ASD could be beneficial for patients.

## Main

Neurodevelopmental disorders (NDDs) including autism spectrum disorder (ASD) and intellectual disability (ID) are highly prevalent in the population, both affecting 1-2% of individuals. Despite the pervasiveness of these disorders, few therapies exist due to a myriad of challenges including the genetic heterogeneity, historic inaccessibility of nervous tissue, and diagnosis usually occurring after the brain has developed. In certain instances, the genetic etiology of an NDD can be traced to a single gene, providing a more constrained genetic background to study these disorders and work towards therapy development. PACS1 syndrome, an NDD caused by a recurrent missense variant, represents one such condition.

PACS1 syndrome (MIM615009) was first reported in 2012 when two patients in an intellectual disability cohort were noted to have strikingly similar craniofacial abnormalities and comparable developmental delays. Upon sequencing, both individuals were found to have the same *de novo* c.607C>T variant in the gene encoding phosphofurin acidic cluster sorting protein 1 (PACS1), which results in an arginine to tryptophan (R203W) substitution. Expression of the variant in a zebrafish model also produced craniofacial abnormalities, supporting the hypothesis that this was the variant responsible for patient symptoms.[1] Nearly 200 patients have since been identified with the same p.R203W pathogenic variant along with one report of a patient with an R203Q substitution.[2–6] All patients have ID and are also frequently afflicted with seizures, ASD, feeding dysfunction, and abnormal formation of other limbs and organs.[3]

Most previous research on PACS1 has focused on its endogenous function. From these studies, it is known the 963 amino acid-long protein contains four domains: an N-terminal atrophin-1 related region (ARR); the furin binding region (FBR), which binds to targets expressing a specific acidic cluster sorting motif;[7] a middle region with autoregulatory properties and a nuclear localization signal (NLS);[8] and a C-terminal region with unknown function. The FBR is where the PACS1 syndrome-causing variant is located, potentially interfering with its binding ability. These domains enable PACS1 to perform multifunctional roles in the endosomal pathway and nucleus. PACS1 was first discovered for its role in the endosomal pathway as a cytosolic sorting protein which binds to a phosphorylated acidic cluster motif and traffics targets to the *trans*-Golgi Network (tGN).[8,9] PACS1 was initially reported in the context of the endoprotease furin, but has since been shown to bind to dozens of other targets with roles in vesicle trafficking, ion transport, gene expression, and other processes.[10–19] Additionally, PACS1 has shuttling functions between the nucleus and the cytoplasm.[20] Roles of PACS1 in the nucleus - where it accumulates during the S phase of the cell cycle - include mediating DNA replication repair,[21,22] exporting of viral transcripts,[23] RNA binding,[20] and regulation of chromatin stability through interaction with histone deacetylase (HDAC) proteins.[21]

PACS1 expression in the brain is high during embryonic development and is maintained at lower levels after birth, suggesting it plays an important role during neurodevelopment and continues to maintain cellular processes in adult tissues. Genome-wide association studies (GWAS) also suggest a broader role in neural development and function for *PACS1* as it has been identified as a susceptibility gene in severe early-onset obesity,[24] developmental delay,[25] and bipolar disorder.[26] However, the majority of research on PACS1's function in neurons has been in the context of Alzheimer's disease (AD), where it has been shown to impact the onset of AD by facilitating localization of the β-secretase BACE1 and the Vps10 sorting receptor homolog SORLA to the tGN.[27,28]

The role of PACS1 during human neurodevelopment and the impact of the R203W variant have never been investigated, and very few therapeutic options exist for patients.[29] In this study, we used induced pluripotent stem cell (iPSC)-derived forebrain organoids to assess the impact of the R203W variant in cortical development. We found the p.R203W variant acts in a gain-of-function manner (GOF) to produce disease phenotypes with strong overlap to convergent features of ASD. This includes ectopic GABAergic expression driven by known autism risk genes, an excitatory/inhibitory synaptic imbalance, and an upregulated gene network reported in many studies of ASD. Altogether these results revealed a striking similarity to previously proposed convergent pathologies of ASD - despite PACS1 syndrome being primarily associated with ID –

highlighting the overlap between etiologies of the two disorders. This is the first study to investigate the impact of the PACS1 variant on a knock-in model of the developing brain and suggests that therapies which clear the variant-containing PACS1 protein or target convergent mechanisms of ASD could be beneficial for PACS1 syndrome patients.

## Results

### PACS1 p.R203W variant does not impact maturation and laminar organization of forebrain organoids

To investigate how the p.R203W variant in PACS1 impacts developing neural cell types, we generated forebrain organoids from two control and two patient lines **(Fig. 1a)**. In addition, we introduced the heterozygous variant which leads to p.R203W into one control line using CRISPR/Cas9 **(Supplementary Fig. 1a-f)**. Cultures were directed towards excitatory forebrain neural cell types through exposure to an established sequence of growth factors **(Supplementary Fig. 2a)**.[30,31] Organoids of both genotypes expressed expected markers such as SOX2+PAX6+ neural progenitors and CTIP2+TBR1+ excitatory neurons **(Supplementary Fig. 2b-d).** Both genotypes also exhibited structures such as clear layering between proliferative neural rosettes surrounded by excitatory neurons **(Fig. 1b)**. No differences in the proportion of cells expressing progenitor and postmitotic neuron markers were found at day 40 **(Fig. 1c)**, a time point chosen based on the onset of high PACS1 expression shown in Allen Brain Atlas developing human transcriptomic data **(Supplementary Fig. 3a)** and subsequent correlation to organoid day.[32]

Since forebrain organoids are composed of heterogeneous cell types, we performed single-cell RNA sequencing (scRNAseq) to investigate cell type-specific transcriptomic changes in PACS1$^{(+/R203W)}$ developing neural cells (see methods and **Supplementary Fig. 3b-c**). ScRNAseq at day 40 revealed six distinct groups in both genotypes of early and late radial glia, intermediate progenitors, newborn neurons, excitatory neurons, and leptomeningeal cells **(Fig. 1d-e)**. Feature plots showed segregation of proliferative and more mature neuron markers to the respective clusters as expected, while *PACS1* is expressed broadly across cell groups **(Fig. 1f-g, Supplementary Fig 3d)**. Transcriptomic analysis also showed no differences in the proportions of cell types between PACS1$^{(+/R203W)}$ and control organoids, suggesting they are developing similarly on a gross cytoarchitectural level **(Fig. 1e; Supplementary Fig. 3e)**.

### PACS1 p.R203W variant results in upregulation of risk genes for other neurodevelopmental disorders by a gain-of-function mechanism

To uncover molecular pathways dysregulated by the PACS1 p.R203W variant in distinct developing neural cell types, we performed differential gene expression analysis of forebrain organoid transcriptomes. Our conservative approach (see Methods; **Supplementary Fig.4a-b**) yielded a total of 637 combined unique differentially expressed genes (DEGs). These included genes with known involvement in neuronal differentiation such as *ZIC1, ZIC2,* and *LHX9* in the newborn group and *DLX1* and *ARX* in the excitatory neuron group **(Fig. 2a)**. A complete list can be found in **Supplementary Table 1.** Top DEGs in the mature neuron group were strongly enriched for risk genes of NDDs with overlapping phenotypes to PACS1 syndrome including ID, epilepsy, cerebellar hypoplasia (CH), and ASD **(Fig. 2b)**.

To determine if the observed DEGs in PACS1$^{(+/R203W)}$ cells were the result of PACS1 loss-of-function (LOF) or GOF - information critical for therapy development - we generated PACS1$^{(-/-)}$ iPSC lines by introducing a frameshift variant in exon 4 **(Supplementary Fig. 1c)** and differentiating them to forebrain organoids **(Supplementary Fig. 2a).** PACS1$^{(-/-)}$ organoids developed as expected and did not show a difference in the ratio of CTIP2+/SOX2+ cells when compared to control and in PACS1$^{(+/R203W)}$ organoids at day 40 **(Supplementary Fig. 2c-d).** Since all organoids developed similarly regardless of genotype, we then asked if the observed DEGs in PACS1$^{(+/R203W)}$ cells followed the same patterns in the PACS1$^{(-/-)}$ lines. We searched for candidates from the DEG analysis that met three criteria: 1) they were differentially expressed across

pseudobulk to eliminate the need for a specific double-stain marker; 2) a difference in expression between genotypes was strongly observed in a feature plot; and 3) the protein had a nuclear localization to increase the accuracy of quantifications. Of three top candidates shown in **Fig. 2c** ZIC1, a zinc finger protein with known roles in neuronal differentiation, was chosen for validation because it is also an OMIM risk gene for ASD, ID, and cerebellar hypoplasia. Immunohistochemistry analysis confirmed upregulation of the ZIC1 protein in PACS1$^{(+/R203W)}$ organoids and showed PACS1$^{(-/-)}$ organoids had the same levels as controls, suggesting the R203W variant operates under a GOF mechanism **(Fig. 2d-e)**. The GOF hypothesis is supported by previous research suggesting that lack of PACS1 does not impact gross brain development in mice.[33] This finding is critically important for therapy development as it suggests clearing the variant-containing PACS1 could be therapeutic for patients.

## Neurons in PACS1$^{(+/R203W)}$ organoids have a propensity towards GABAergic differentiation

To obtain a more comprehensive understanding of what pathways are affected in PACS1$^{(+/R203W)}$ cells, we performed gene ontology (GO) analysis of DEGs in each cell type **(Supplementary Fig. 5a-b)**. Downregulated genes in PACS1$^{(+/R203W)}$ organoids were not enriched for GO processes. However, GO analysis of upregulated transcripts in PACS1$^{(+/R203W)}$ showed strong patterns: while the newborn neuron group had differentially expressed transcripts related to neuron migration and neurite outgrowth **(Supplementary Fig. 5a-b)**, the most mature neuron group showed enrichment for terms related to GABAergic differentiation, even though we utilized a directed forebrain organoid protocol **(Fig. 3a)**. The presence of a few GABAergic cells arising from dorsal forebrain lineages was not unexpected,[34] but they were much more prevalent in PACS1$^{(+/R203W)}$ organoids. GABAergic-related GO terms were enriched due to upregulation of genes such as *ARX, DLX1, DLX2, DLX5, DLX6-AS1,* and *GAD1* in PACS1$^{(+/R203W)}$ neurons **(Fig. 3b-c)**. To validate these findings, we quantified the number of cells expressing GAD67, the protein product of *GAD1,* in day 40 organoids. As suggested by the transcriptomic data, we found that PACS1$^{(+/R203W)}$ organoids had a higher number of GAD67+ cells when compared to control organoids. We did not observe this increased number of GABAergic cells in the PACS1$^{(-/-)}$ organoids, further supporting the hypothesis that the p.R203W variant has a GOF effect **(Fig. 3d-e)**.

## Pseudotime analysis reveals a gene module driving ectopic GABAergic fate in excitatory neurons from PACS1$^{(+/R203W)}$ organoids

In feature plots, all GABAergic markers segregated to the same branch of the neuron cluster that had simultaneous expression of excitatory markers **(Fig. 4a)**. We next asked what genes might be driving this aberrant GABAergic-like trajectory in PACS1$^{(+/R203W)}$ organoids using pseudotime analysis. Since an increase in GABAergic cells was not found in PACS1$^{(-/-)}$ organoids **(Fig. 3d-e)**, we did not include this genotype in subsequent investigations. Construction of the pseudotime trajectory with PACS1$^{(+/+)}$ and PACS1$^{(+/R203W)}$ cells showed a clear path from radial glia to neurons as pseudoage increased **(Fig. 4b)**. A single trajectory was generally followed until a divergence into hydra-like branches at the neuron cluster. To look at which genes might be driving the fate decision of cells in the GABAergic-like branch, we first grouped genes that varied highly over pseudotime into modules of similar expression patterns. Then, we enhanced the granularity of the neuron group by increasing the cluster resolution parameter **(Fig. 4c)** and plotted aggregate expression of each module in these subclusters **(Fig. 4d**). This method allowed us to isolate genes specifically expressed at the GABAergic fate point. Intriguingly, one module of 78 genes – arbitrarily numbered module 21 - was specifically and highly expressed just prior to the GABAergic fate point (boxed subclusters in **Fig. 4d**) and was distinct from classically GABAergic genes in module 51. This gene set was enriched for transcripts known to play a role in suppression of excitatory or promotion of inhibitory differentiation, such as *ZIC1, ZIC2, MEF2C,* and *FOXG1.*[35–38] Module 21 was drastically more expressed in PACS1$^{(+/R203W)}$ organoids compared to controls **(Fig. 4e-f)** and was in fact the most differentially expressed gene set between genotypes in cells just entering

the GABAergic branch **(Fig. 4g)**. These results suggest that module 21 genes are responsible for driving the increased number of GABAergic cells in PACS1$^{(+/R203W)}$ neurons.

## Neurons in PACS1$^{(+/R203W)}$ organoids have a higher density of GABAergic but not glutamatergic synapses

In addition to promoting GABAergic and suppressing glutamatergic differentiation, module 21 genes such as *MEF2C* and *FOXG1* have previously been shown to regulate excitatory/inhibitory synaptic density.[36,37,39] Since this balance plays a critical role in disease mechanisms of neurodevelopmental disorders such as ASD and epilepsy,[40] we next investigated if excitatory/inhibitory balance might be impaired in PACS1 syndrome. As a proxy for number of synapses, we plotted collective expression of transcripts encoding glutamatergic or GABAergic receptors from our scRNAseq data, a complete list of which can be found in **Supplementary Fig. 6a.** Expression patterns of glutamatergic receptor transcripts in neurons was unaffected between genotypes **(Fig. 5a)** but expression of GABAergic receptor transcripts was drastically upregulated **(Fig. 5b)**. We next looked at the relationship of glutamatergic and GABAergic receptor expression within the same cell. A small correlation was found in control neurons **(Supplementary Fig. 6b)**, suggesting that neurons with more synapses – perhaps more mature cells – are the ones expressing GABAergic transcripts. In contrast, PACS1$^{(+/R203W)}$ neurons had high GABAergic receptor expression regardless of total receptor expression **(Supplementary Fig. 6c)**. This suggests PACS1$^{(+/R203W)}$ neurons expresses a higher percentage of GABAergic receptors, regardless of glutamatergic receptor expression.

To directly look at synapses, we performed immunostaining for pre- and post-synaptic markers in day 79 organoids. Excitatory synapses were defined as colocalization of pre-synaptic marker SYN1 and post-synaptic marker PSD95, while inhibitory synapses were defined as colocalization of pre-synaptic marker vGAT and post-synaptic marker Gephyrin **(Fig. 5c)**. As hypothesized, we found an increase in the density of inhibitory but not excitatory synapses in PACS1$^{(+/R203W)}$ organoids **(Fig. 5d)**, suggesting that the ectopic GABAergic cells are maintained and able to form synapses, further contributing to an excitatory/inhibitory (E/I) signaling imbalance. This is an interesting result as an E/I imbalance has been proposed as a common feature of ASD from diverse genetic etiologies,[37,40–43] indicating that the p.R203W variant may confer toxicity in a manner similar to ASD-causing variants.

## Gene co-expression analysis reveals an ASD-specific gene network upregulated in PACS1$^{(+/R203W)}$ neurons

Previous results such as ectopic GABAergic expression and an impaired E/I synaptic balance were remarkably similar to hypothesized convergent mechanisms of ASD. This is intriguing as PACS1 syndrome is caused by a single variant with best known roles in the tGN and nucleus, while ASD is typically caused by a collective burden of many variants, often in synaptic genes. To further investigate the shared molecular signature of PACS1 syndrome and ASD, we employed weighted gene co-expression network analysis (WGCNA).[44] This powerful tool for analyzing RNA sequencing data enables identification of modules of genes with common expression patterns and has the ability to correlate them to external traits such as genotype. It is frequently used in studies of disorders with diverse etiologies to identify convergent networks of misregulated genes.

We employed WGCNA on the transcriptome of equal numbers of PACS1$^{(+/R203W)}$ and control cells **(Fig. 6a)**. Ten distinct modules of gene expression were identified, five of which were significantly associated with genotype **(Fig. 6b-c)**. The arbitrarily named turquoise, brown, and blue modules **(Fig. 6d)** were selected for further investigation because these modules had a positive correlation between a gene's significance to the genotype and its interconnectedness within a module **(Supplementary Fig. 7a-d)**. To determine if any gene modules correlated to those identified in other studies of autism, we extracted ASD neuronal-specific modules in eight landmark studies and performed a gene overlap test.[37,45–51] Intriguingly, we found that the blue module

overlapped specifically with at least one module from all these studies **(Fig. 6e)**. No other correlations were found. This module was specific to the excitatory neuron group - when the same analysis was performed in newborn neurons, a module correlating to the blue module in neurons was not identified, and there was no overlap with previously identified ASD-specific modules **(Supplementary Fig.7e-f)**.

To determine the functional significance of this blue module, we performed gene ontology cellular component analysis. To reduce redundancy of terms and help visualize results, we generated a treemap.[52] A large portion of this treemap was devoted to synaptic terms, suggesting that members of this upregulated module of genes were enriched for synaptic functions, a structure known to be specifically affected in ASD **(Fig. 6f)**. This suggests a previously uncharacterized role for either PACS1 or its binding partners at synapses and is an intriguing area of future investigation.

## Discussion

Here we present some of the first work investigating the impact of the PACS1 R203W variant on the developing brain. A common thread through our results was the striking similarity to proposed convergent mechanisms of autism, such as aberrant GABAergic differentiation driven by ASD risk genes, an excitatory/inhibitory synaptic imbalance, and an upregulated network of genes found in other studies of ASD. Suppressed GABAergic inhibition was one of the first proposed convergent phenotypes of ASD in 2001 and continues to be a well-established hypothesis, particularly a hypofunction in parvalbumin-expressing interneurons.[42,43,53–55] However, other studies have also reported an increased density of GABAergic cells in ASD models, including a recent paper showing pathogenic mechanisms of *ARIDB*$^{+/-}$ , *SUV420H1*$^{+/-}$, and *CHD8*$^{+/-}$ organoids converged toward early GABAergic induction.[37,56–59] Of course, certain discrepancies may arise from limitations in model systems in addition to genetic background, such as *in vitro* cultures lacking the circuitry and diversity of a functioning brain. The relative simplicity of an *in vitro* model is likely the reason for the apparent discrepancy between increased GABAergic cells in our PACS1$^{(+/R203W)}$ organoids and the epilepsy experienced by patients. While much more is left to understand, an accumulation of evidence suggests the excitation/inhibition balance is in some way impaired in the majority (if not all) of ASD cases.[36,40–43,53–56,60–64]

In this study, though we used an excitatory glutamatergic forebrain differentiation protocol, upregulation of GABAergic markers on the transcriptomic and protein level were found in PACS1$^{(+/R203W)}$ organoids. These cells co-expressed excitatory markers such as Glutaminase, suggesting an aberrant co-occurrence of excitatory and inhibitory properties in PACS1$^{(+/R203W)}$ glutamatergic neurons. We used pseudotime analysis to show these GABAergic characteristics were driven by a module of known ASD risk genes including *MEF2C* and *FOXG1.* This is in line with previous reports showing a FOXG1-dependent GABAergic/glutamatergic imbalance in ASD organoid models[37] and MEF2C regulation of excitatory/inhibitory synaptic density.[36] This cluster also included many *ZIC* proteins, which have been shown to induce the generation of calretinin-positive GABAergic neurons[35] in addition to their more well-known roles in the cerebellar development. Intriguingly, a prevalent brain malformation in PACS1 patients is cerebellar hypoplasia,[3] which could have a mechanistic etiology in ZIC1.

Perhaps the strongest support for PACS1 syndrome's parallel to ASD was found with our WGCNA analysis, which identified an upregulated module of genes in PACS1$^{(+/R203W)}$ neurons that highly correlated to ASD-specific modules found in eight other high-profile studies. The degree of specificity was also notable as no other module - even those significantly associated with genotype - ever overlapped. No genes were present in every single one of these modules, but highly recurrent genes were strongly enriched for synapses. Members of the blue module itself were also enriched at synapses, which are considered a convergent affected structure in ASD and were found to have an impaired E/I ratio in this study.[65]

An additional key insight from this study is evidence for the PACS1 R203W variant operating in a GOF manner. ZIC1 and GAD67 were both increased in the PACS1$^{(+/R203W)}$ organoids, but PACS1$^{(-/-)}$ knockout organoids had the same levels as the controls. Our results implicate a GOF over a LOF since PACS1$^{(+/R203W)}$

and PACS1$^{(-/-)}$ organoids had different phenotypes. However, we do not yet know by what molecular mechanism the tryptophan substitution confers toxicity. Future studies should involve crystallography to try and resolve the structure of the FBR in the mutant and control PACS1, immunoprecipitation assays in neurons to establish if binding partners are lost or acquired upon introduction of the variant, binding affinity assays to determine if p.R203W binds variably to targets, and immunohistochemistry of PACS1 and its targets to assess how the variant impacts localization in the cell.

The GOF hypothesis supported by this study is reinforced in previous literature. The only study to date that directly looked at the p.R203W PACS1 showed zebrafish exhibited craniofacial deficits when overexpressing p.R203W - but not wild-type - PACS1 mRNA.[1] The presence of endogenous PACS1 in this study suggests a GOF effect. Evidence against a LOF mechanism is further supported by a report of a family with a multiexonic deletion in PACS1 having little to no cognitive impairment, craniofacial abnormalities, or structural brain abnormalities.[66] Moreover, Pacs1$^{(-/-)}$ mice present with a deficiency of circulating B cells, CD4$^{+}$ T cells, and CD8$^{+}$ T cells, but an absence of gross neurological phenotypes.[33] These findings implicate that an effective therapy for PACS1 syndrome patients could be clearing of the variant-containing protein. One potential approach to achieve this is with anti-sense oligonucleotides (ASOs) designed to bind in a complementary manner to the variant-containing mRNA and block translation. More studies should be done to determine if clearing PACS1 R203W with ASOs would be a safe and effective treatment.

Therapies such as ASOs provide much promise for PACS1 patients. Unfortunately, most other genetic causes of ASD and ID are much more complex and developing therapies for each genetic origin is very challenging. A momentous leap forward for the field would be to develop a way to target common phenotypes across a multitude of etiologies, such as restoring the excitation/inhibition balance. This is a monumental task, but the potential provides hope for millions of people suffering from ID and ASD.

## Methods

### Culture of iPSCs

Induced pluripotent stem cells were maintained in mTeSR+ medium (Stem Cell Technologies, 05825) on plates pre-coated with Matrigel (Corning, 354234) dissolved in DMEM/F12 at a concentration of 10 µg/mL. Stem cells were passaged as needed using ReLeSR medium (Stem Cell Technologies, 05872). One control line GM03651 was purchased from Coriell. A second control line GM27160 and two patient lines GM27159 and GM27161 were purchased from WiCell. An isogenic counterpart of each of these lines (except GM27160) was generated using CRISPR/Cas9. Cell lines that had undergone CRISPR were validated with whole exome sequencing analysis. All cell lines were regularly tested for mycoplasma contamination.

### CRISPR

Generation of cell lines was carried out utilizing the CRISPR/Cas9 system. Briefly, the Cas9 was delivered to cells as a px459-derived plasmid (Addgene, 62988) with a cloned guide that was designed using crispr.mit.edu software. The forward sequence was: caccgTTACAAGAATCGGACCATCT and reverse: aaacAGATGGTCCGATTCTTGTAAc (lowercase letters are sticky ends to clone into the plasmid). 5 µg of plasmid and 5 µg of donor DNA (sequence TGCAGATCATGCTGCAAAGGAGAAAACGTTACAAGAAT**T**GGACTATCTTAGGCTATAAGACCTTGGCCGTGGGACTCATCAAC) were delivered to 1 million cells using the NEON transfection system (ThermoFisher, MPK5000). 24 hours after electroporation, cells were exposed to 0.5 µg/mL puromycin for 24 hours. Cells were left to expand for roughly 10 days. When visible colonies had formed, they were manually isolated with a pipet and sanger sequenced. Clones of interest were sent for whole exome sequencing at Novogene to check for off-target effects.

## Generation of forebrain organoids

Forebrain organoids were generated according to the previously described protocol by Sloan and coworkers, with slight modifications.[31] In brief, iPSCs were grown on matrigel-coated plates in mTeSR medium until colonies were ~ 5 mm across. Colonies were detached from the plate by incubating at 37ºC with 0.35 mg/mL dispase for 20 minutes. Detached colonies were then transferred to an ultra-low attachment dish (Corning, cat. no. 3262) with DMEM/F12 (1:1, Life Technologies, 1130-032) containing 20% vol/vol knockout serum replacement (Life Technologies, cat. no. 10828-028), 1% vol/vol non-essential amino acids (Life Technologies, 1140-050), 0.1 mM 2-mercaptoethanol (Sigma-Aldrich, M3148), 5 μM of the SMAD inhibitor dorsomorphin (Sigma-Aldrich, P5499), 10 μM SMAD inhibitor SB-4321542 (R&D Systems/Tocris, cat. no. 1614), and 10 μM rock inhibitor Y-27632 (Selleckchem, S1049). Media was changed daily for the next five days with the same reagents as described above, minus the Y-27632. At day 6, media was changed to a Neurobasal-A base containing 1x B-27, 1x Glutamax, 20 ng/mL FGF2, and 20 ng/mL EGF (R&D Systems, 236-EG). Shaking at 95 rpm was begun on day 13, at which point media changes began to occur every other day. This was continued until day 25. At day 25, growth factors were switched to 20 ng/mL BDNF (PeproTech 450-02) and 20 ng/mL Neurotrophin 3 (PeproTech, cat. no. 450-03). Shaking and media changes continued as described. Organoids were changed to neurobasal-A medium supplemented with 1x B-27 beginning at day 43 and continuing until the desired endpoint.

## Dissociation of forebrain organoids for scRNAseq

Dissociation was done using the Worthington Papain Dissociation System. Steps were done with sterile materials in a tissue culture hood. At day 40-42, two organoids between 3-4mm were selected per genotype. Organoids were pooled and washed with PBS + 0.2% BSA, then chopped into ~0.5mm pieces using a #10 blade. Organoids were incubated in a sterile-filtered 5 mL solution containing 20 U/mL papain in 1x EBSS, 0.46% sucrose, 26mM NaHCO3 (included in kit), 0.5mM EDTA (included in kit), and DNase (included in kit) equilibrated with 95%O2:5%CO2. Organoid pieces were incubated for 75 minutes on a 95 RPM orbital shaker in a 37ºC incubator. Following incubation, organoids were triturated using wide bore tips (Rainin, 30389218) and transferred to a 15 mL conical vial. Cells were spun 300xg for 5 minutes. The pellet was resuspended in a 3 mL papain inhibitor solution containing 2 mg/mL ovomucoid inhibitor, 2 mg/mL albumin, 0.46% sucrose, and DNase in 1x EBSS (all from Worthington kit). This cell suspension was gently applied on top of a 5 mL solution of 10 mg/mL ovomucoid inhibitor, 10 mg/mL albumin, and 0.46% sucrose to create a density gradient. The gradient was spun at 70xg for 6 minutes. The pellet was washed with 5 mL PBS + 0.2% BSA and spun again at 300xg for 5 minutes. The pellet was resuspended in 130 μL PBS + 0.2% BSA and filtered to remove cell clumps (Bel-Art, H13680-0040), resulting in the final single-cell suspension. In total, seven samples were processed in three distinct batches (pictured in **Supplementary Fig. 3b**).

## ScRNAseq analysis

Preparation of samples was carried out as described above. Samples only passed if viability was >85%. The library prep was conducted by Northwestern's NUseq core using the 10x Chromium library prep system. Sequencing was done using 100bp, paired-end reads on an Illumina HiSeq 4000 with one sequencing lane per sample. Raw data was demultiplexed and FASTQ files were processed using the CellRanger pipeline.

After sequencing, we analyzed the data using Seurat and Monocle3 packages in R.[72–74] In Seurat, data was filtered for nFeature_RNA > 500, nFeature_RNA < 6500, and percent.mt < 15. For each sample, doublets were first identified by performing processing steps (NormalizeData, FindVariableFeatures, ScaleData, RunPCA, FindNeighbors, FindClusters, RunTSNE) and then running two separate doublet finding packages: DoubletFinder and DoubletDecon.[67,68] Barcodes were extracted for consensus cells that both packages identified as doublets (roughly 5% of each sample). Following doublet identification (a schematic is shown in **Supplementary Fig. 3c**), analysis was re-started, and doublets were removed in the initial filtering steps.

Samples were merged and then NormalizeData and FindVariableFeatures was performed on each sample. A FindIntegrationAnchors step was done to enable comparative visualization across samples and remove batch effects, followed by IntegrateData. After integration, the ScaleData, RunPCA, FindNeighbors, FindClusters, and RunTSNE steps were performed. For differential expression analysis, public data (GSM3885398, GSM3885387, ERS4649261, ERS4649262, and ERS4649263) was incorporated by projecting onto pre-existing data using the TransferAnchors step. Cluster identification was bioinformatically aided with the Clustifyr package[69] using a fetal human tissue dataset generated in the Kriegstein lab as a reference.[70] Cluster names were simplified to represent the broad cell type classification.

Differential expression analysis was performed using the FindMarkers function. To increase the power for this analysis, five additional public controls (GSM3885398, GSM3885387, ERS4649261, ERS4649262, and ERS4649263) generated with the same differentiation protocol and nearly identical timepoint were downloaded from the NCBI GEO repository **(Supplementary Fig. 4a)**.[75,76] To reduce the number of false positives caused by inter-organoid variability, FindMarkers was performed six ways: between three different groups of controls (organoids generated in the Guemez-Gamboa lab, Kriegstein lab, and Testa labs) and two different batches of PACS1$^{(+/R203W)}$ organoids. Only hits up or downregulated in all five+ tests with an average adjusted *p* value of < 0.05 after Bonferroni correction for all genes were included as DEGs **(Supplementary Fig. 4b; Supplementary Table 1)**. Top DEGs were significant for all six tests. Mitochondrial (MT-) and ribosomal (RPS/RPL) proteins were removed for GO and DAVID analysis.

## Pseudotime analysis

Pseudotime analysis was performed by closely following tutorials developed by Cole Trapnell at cole-trapnell-lab.github.io/monocle3/ and integrated with Seurat using the SeuratWrappers function. First, a cell data set object was created from the fully processed Seurat object containing all seven samples. Cluster_cells, preprocess_cds, estimate_size_factors, and learn_graph functions were used to generate a pseudotime trajectory graph. The root node was manually chosen and only the largest partition was used, which contained all the neurons. Graph_test, find_gene_modules, and aggregate_gene_expression functions were used to group modules of gene expression over pseudotime. Downstream analysis was performed with base Seurat and Monocle3 functions.

## WGCNA

To perform weighted co-expression network analysis, 4,884 control and PACS1$^{(+/R203W)}$ neurons were randomly subsetted from the Seurat object. Data was processed closely following tutorials developed by Peter Langelder and Steve Horvath from horvath.genetics.ucla.edu. First, a data frame of each cell's expression of all genes was made. Genes and samples with too many missing values were removed with the function goodSamplesGenes. The blockwiseModules function was run using a scale free topology power of 3, a signed TOM type, minimum module size of 30, and a max block size of 50,000 so genes were not divided in blocks. This resulted in 10 distinct modules, 5 of which were significantly correlated with genotype. The abs and intramodularConnectivity functions were used to determine which modules had a positive correlation between a gene's interconnectedness within a module and significance to genotype. To test if modules overlapped with previously identified ASD modules, the geneOverlap package was used and *p* values were adjusted using a Benjamini-Hochberg correction.

## Cryopreservation and sectioning

Organoids were fixed with 4% paraformaldehyde in PBS at 4ºC overnight. They were then washed with PBS and transferred to a 30% sucrose solution in PBS and incubated at 4ºC for at least 16 hours. To cryopreserve, organoids were embedded in OCT medium using a mold and snap frozen in dry ice and ethanol. Organoids

were stored at -80ºC indefinitely. Frozen organoids were sectioned 15-20 µm thick at -16ºC in a cryostat and applied to Thermofisher slides. Slides were left to dry overnight and then stored at -80ºC.

### Immunohistochemistry

Organoid sections were dried for at least one hour either at room temperature or 37ºC. OCT was removed in a series of 3 x 10 min washes in PBS. An antigen retrieval step was performed by steaming slices at 99ºC for 10 minutes in a 10 mM citrate solution at pH 6.0. To remove the citrate buffer, slides were washed 1 x 5 min in $dH_20$ and 2 x 5 min in PBS. Tissue was blocked in a 10% NDS, 3% Triton X-100 solution in 1x PBS for 1-2 hours at room temperature. Following blocking, tissue was incubated in a 5% NDS, 1.5% Triton X-100 solution in 1x PBS containing primary antibodies at 4ºC overnight. The overnight incubation was followed by 3 x 10 min PBS wash, then secondary antibody incubation in the same solution as the primary antibodies (CTIP2 Abcam ab18465; PAX6 BioLegend 901301; TBR1 Proteintech 66564-1-Ig; SOX2 EMD Millipore AB5603; ZIC1 Invitrogen PA5-47681, GAD67 EMD Millipore MAB5406; Syn1 Cell Signaling Technology 5297; PSD95 NeuroMab K28/43; vGAT ThermoFisher PA5-27569; Gephyrin Synaptic Systems 147-021; TRA1-60 Sigma MAB4360; OCT4A Cell Signaling Technology C30A3). Excess antibodies were removed with three more 10-minute PBS washes and a 20-minute incubation in 2 µg/mL DAPI. Slides mounted with Fluoromount medium and imaged on a Nikon W1 spinning disk confocal microscope or BZX710 Keyence epifluorescent microscope. Images were analyzed using ImageJ (Fiji).

### Synaptic quantifications

Day 79 organoid sections were stained for MAP2 and either SYN1/PSD95 or vGAT/GPHN primary antibodies as described in the above section. Confocal images were taken on a Nikon W1 Dual CAM Spinning Disk microscope at 60x resolution with a 1 µm step size with a Z stack of 15 images, then collapsed to one plane in ImageJ based on maximal intensity. Pre- and post- synaptic puncta channels were converted to binary and superimposed, then the number of colocalizations above 10 pixels in area was automatically counted using the Analyze Particles function.

## Data Availability

ScRNAseq data will be uploaded to the NCBI Gene Expression Omnibus repository. For DEG analysis, samples GSM3885398, GSM3885387, ERS4649261, ERS4649262, and ERS4649263 were downloaded from GEO.

## Code Availability

All code is available upon request. Code used in this study is a compilation from the following sources. Reads were aligned using the 10X Genomics Cell Ranger pipeline found at https://support.10xgenomics.com/single-cell-gene-expression/software/pipelines/latest/using/tutorial_ov. The output matrix files were first processed with Seurat version 4 following tutorials produced by the Satija lab found at https://satijalab.org/seurat/. Pseudotime analysis was done with Monocle version 3 following the workflow produced by the Trapnell lab found at https://cole-trapnell-lab.github.io/monocle3/. WGCNA version 1.70-3 was performed by following tutorials produced by the Langfelder and Horvath labs found at https://horvath.genetics.ucla.edu/html/CoexpressionNetwork/Rpackages/WGCNA/.

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### Contributions

L.R. performed all experiments, contributed to project guidance, wrote the manuscript, and made the figures. A.G.G. provided continual guidance at every stage, technical support, and edited the manuscript and figures.

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## Ethics declaration

### Competing interests

The authors declare no competing interests.

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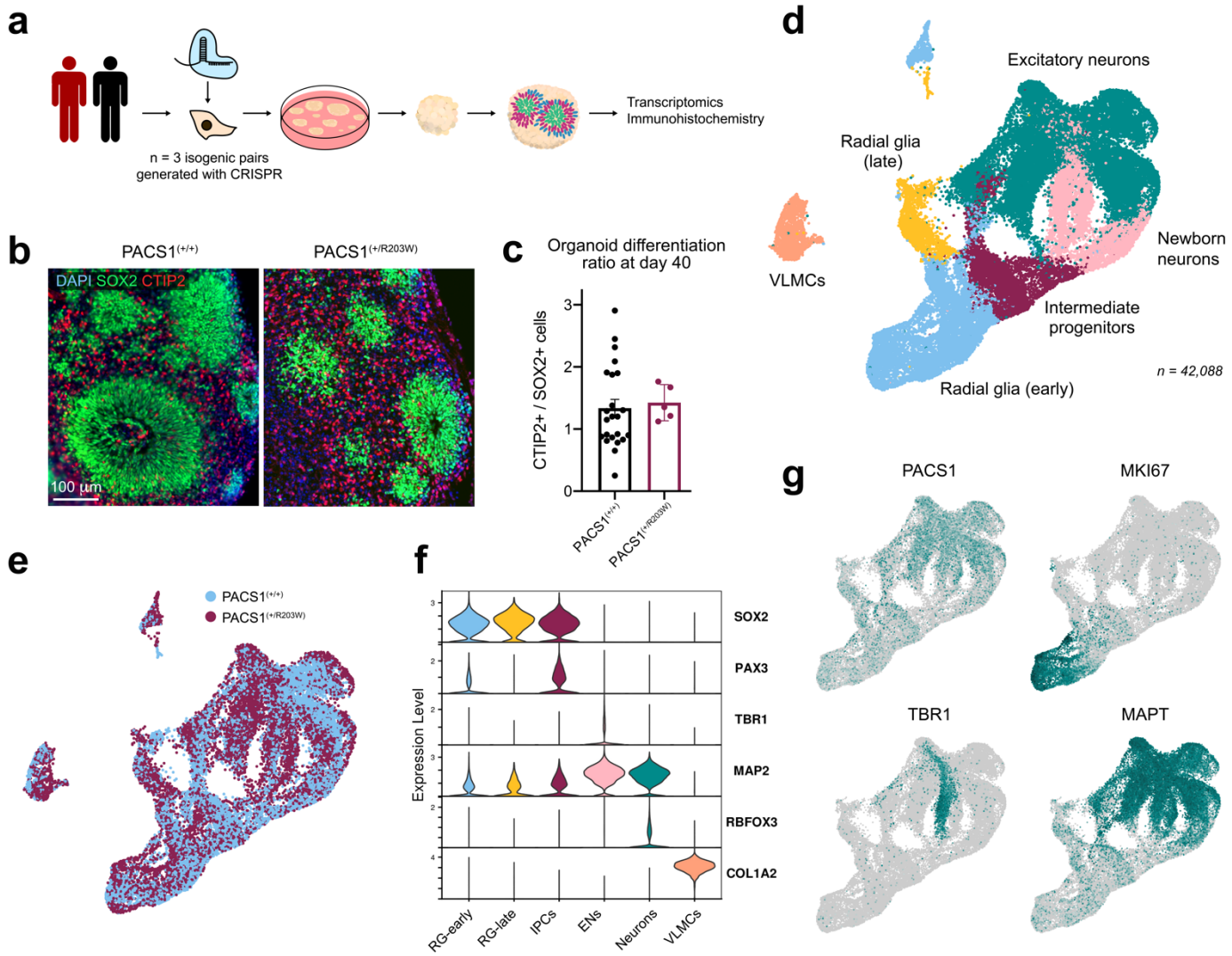
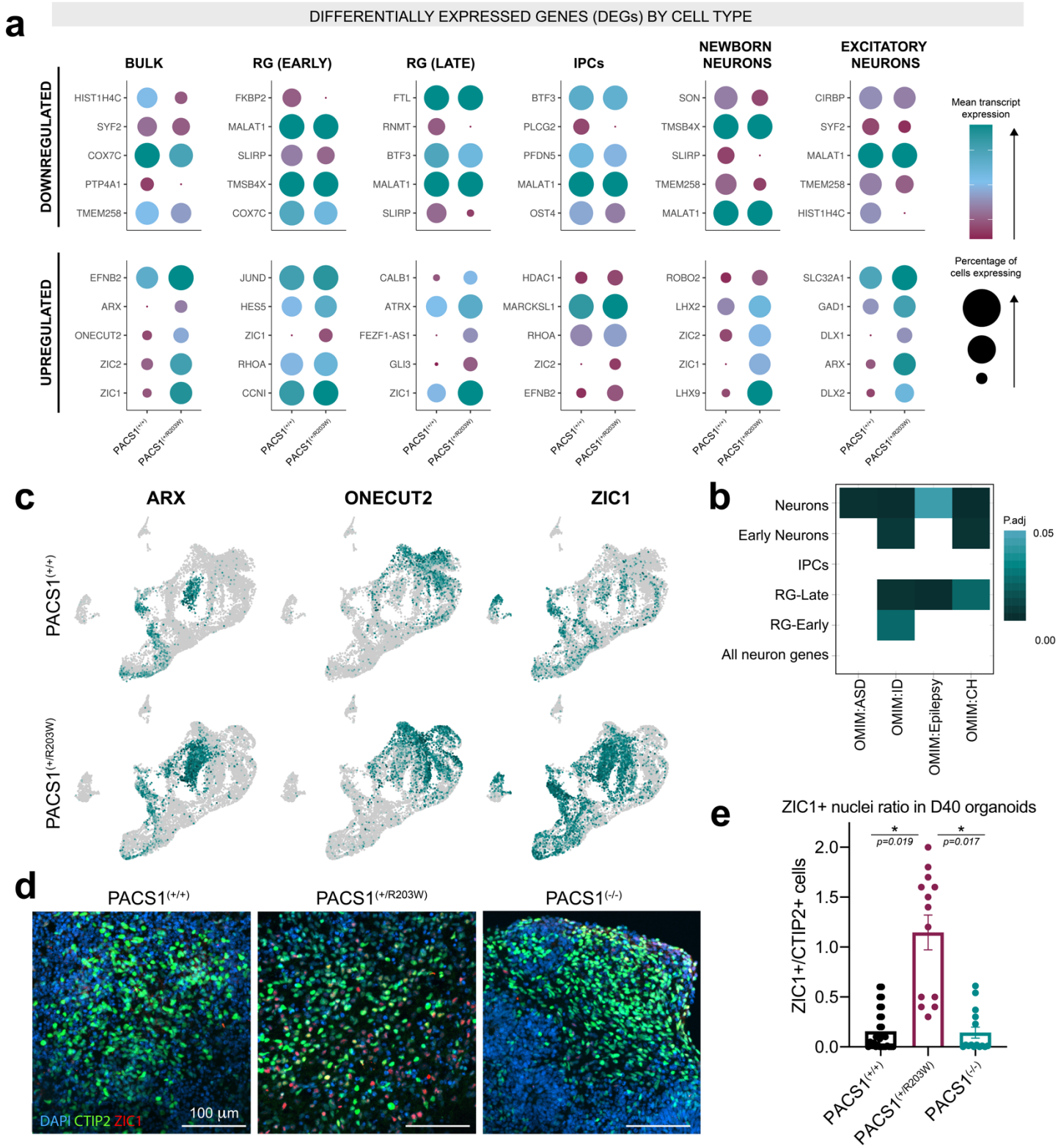
## FIGURES

**Fig. 1: PACS1 p.R203W variant does not impact maturation and laminar organization of forebrain organoids.**




**a**, Experimental overview. Isogenic cell lines are generated using CRISPR and differentiated to forebrain organoids. **b**, Organoids are developing expected structures such as SOX2+ rosettes surrounded by CTIP2+ neurons. Scale bar = 100 µm. **c**, The ratio of CTIP2+ neurons to SOX2+ progenitors in day 40 organoids, showing the mean + SEM. **d,** Single-cell RNA sequencing reveals six distinct cell types. **e,** Distribution of cell groups is not different between genotypes. PACS1$^{(+/R203W)}$ cells have been down-sampled to match the number of control cells for visualization purposes. **f,** Cell types express the expected markers: proliferative types are *SOX2*+; intermediate progenitors express *PAX3*; early neurons are *TBR1*+; neurons are *RBFOX3*+*MAP2*+; ependymal cells specifically express collagen markers such as *COL1A2*. **g,** Feature plots show expression of a transcript in each cell. *PACS1* is expressed in all cell groups, but segregation is observed of proliferative marker *MKI67,* newborn neuron marker *TBR1,* and neuron marker *MAPT.*

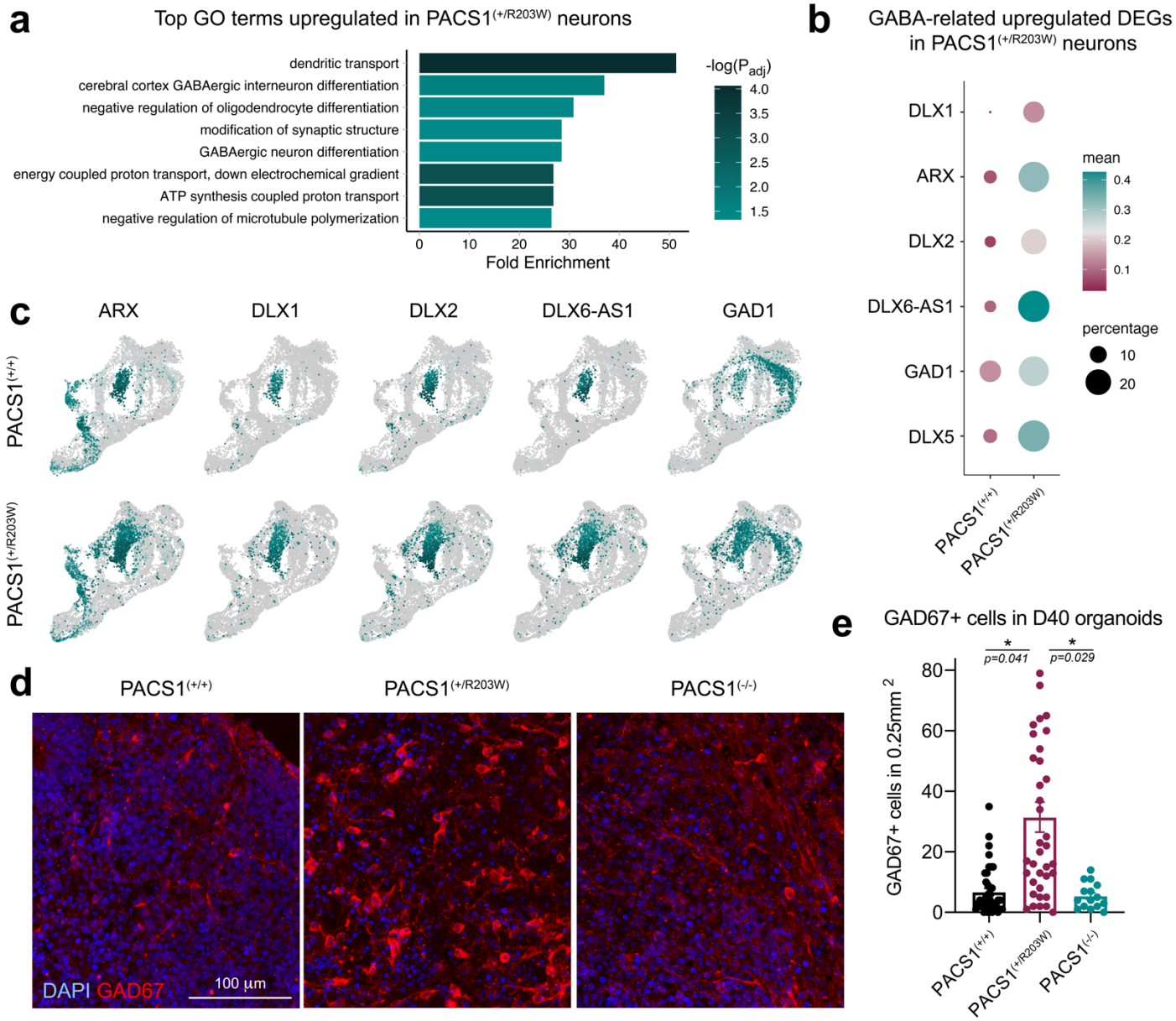
**Fig. 2: PACS1 p.R203W variant results in upregulation of risk genes for other neurodevelopmental disorders by a gain-of-function mechanism**




**a,** Top differentially expressed genes between PACS1$^{(+/R203W)}$ and PACS1$^{(+/+)}$ organoids. Ranges of figure legends have been simplified for readability. Figure legends for each graph can be found in Supplementary Fig. 4c. **b,** *p* values of a gene overlap test between top DEGs in each group and OMIM risk genes for neurodevelopmental disorders with converging phenotypes to PACS1 syndrome. *p* values were calculated using a Fisher's exact test with a Benjamini-Hochberg correction for multiple testing. **c,** Feature plots of top DEGs upregulated in PACS1 syndrome. n=11,851 cells. **d,** Immunohistochemistry analysis in day 40 organoid

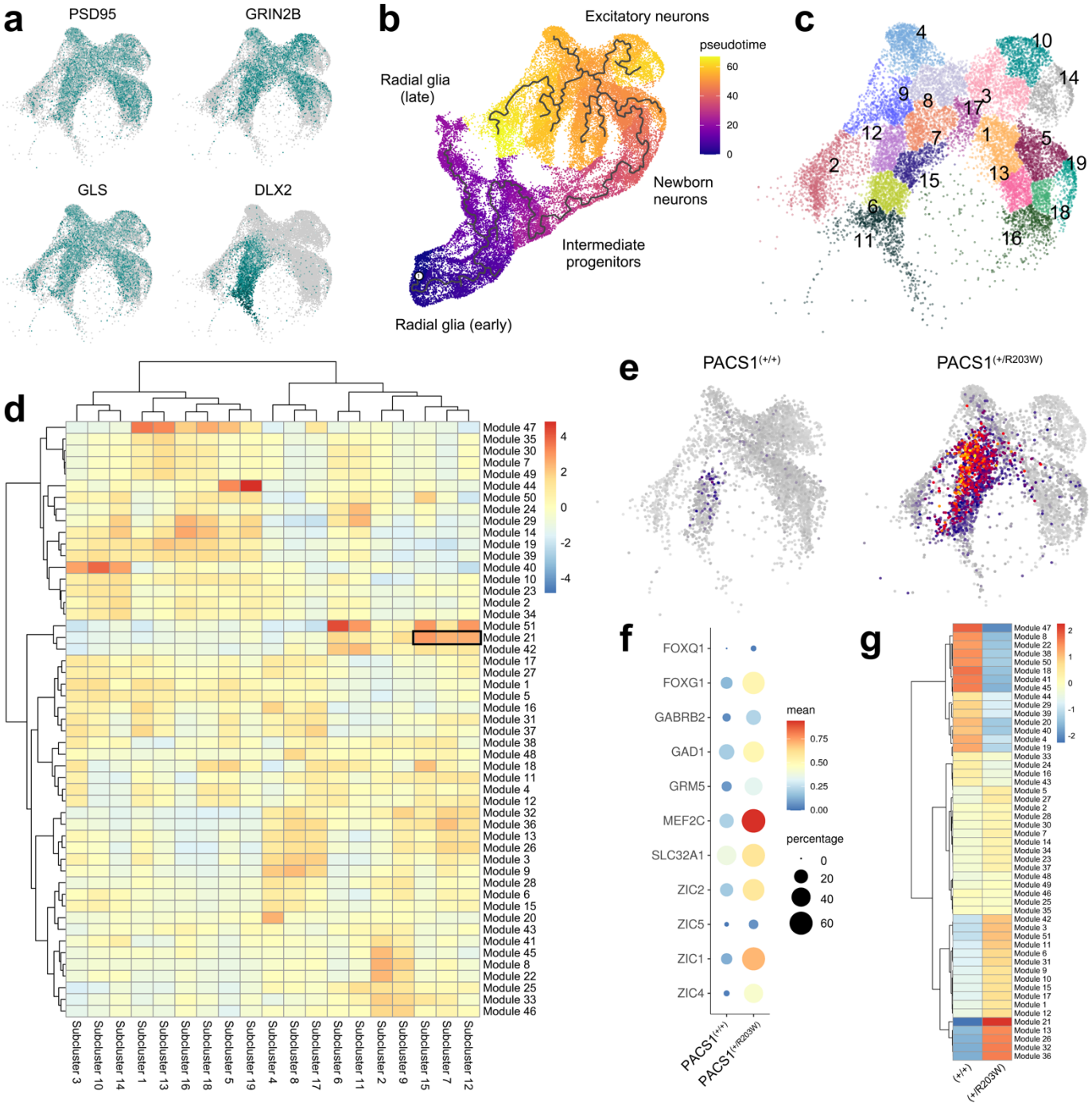
sections of CTIP2 and ZIC1, a top DEG and OMIM risk gene for ASD, ID, and CH. A PACS1$^{(-/-)}$ organoid was included to investigate the mechanism of the R203W variant. Scale bar = 100 µm. **e,** Quantifications of immunohistochemistry analysis in day 40 organoids shown in D, showing mean + SEM. Data is from 28 PACS1$^{(+/+)}$ organoid sections, 13 PACS1$^{(+/R203W)}$ sections, and 16 PACS1$^{(-/-)}$ sections. *p* values are calculated from n = 6 PACS1$^{(+/+)}$ organoids, n = 3 PACS1$^{(+/R203W)}$ organoids, and n = 3 PACS1$^{(-/-)}$ organoids. Values were calculated using a two-tailed t test assuming unequal variance.

**Fig. 3: PACS1$^{(+/R203W)}$ neurons have ectopic GABAergic expression.**



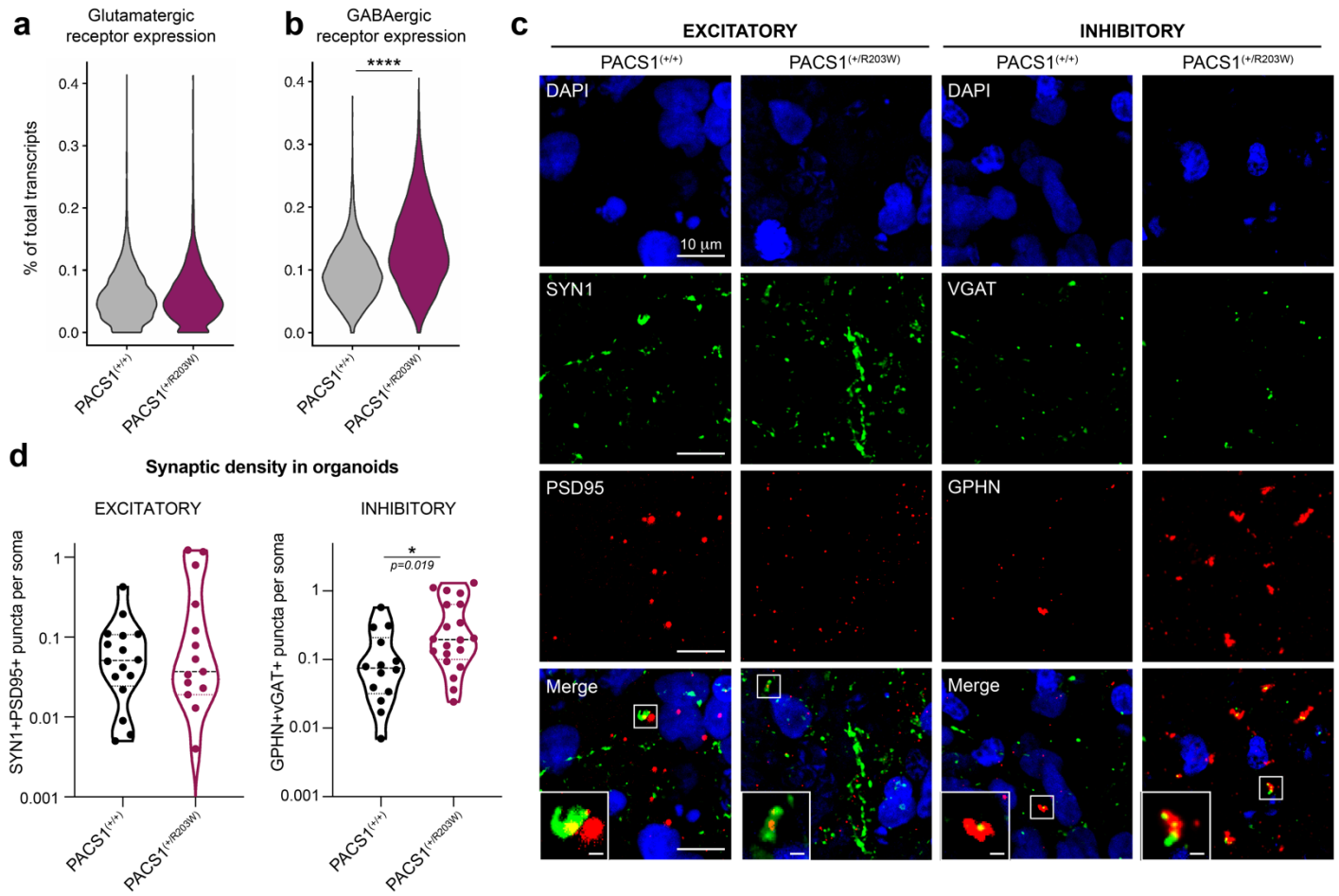

**a,** Top GO terms for upregulated genes in PACS1$^{(+/R203W)}$ newborn and excitatory neurons, arranged by fold enrichment. *p* value was calculated using a Fisher's exact test with FDR adjustment. **b,** Dot plots of DEGs involved in GABAergic differentiation in PACS1$^{(+/+)}$ and PACS1$^{(+/R203W)}$ neurons. **c,** Feature plots of DEGs involved in GABAergic differentiation in PACS1$^{(+/+)}$ and PACS1$^{(+/R203W)}$ neurons. n=11,851. **d,** Immunohistochemistry analysis of GAD67 expression, the protein product of the gene *GAD1*, in day 40 organoids. Scale bar = 100 µm. **e,** Quantification of GAD67+ soma per 500x500 micron area in day 40 organoids showing mean + SEM. Data shown is from 40 PACS1$^{(+/+)}$ organoid sections, 46 PACS1$^{(+/R203W)}$ sections, and 15 PACS1$^{(-/-)}$ sections. *p* values are calculated from n = 10 PACS1$^{(+/+)}$ organoids, n = 9 PACS1$^{(+/R203W)}$ organoids, and n = 10 PACS1$^{(-/-)}$ organoids. Values were calculated using a two-tailed t test assuming unequal variance.

**Fig. 4: Ectopic GABAergic expression is driven by a module including *ZIC* genes.**



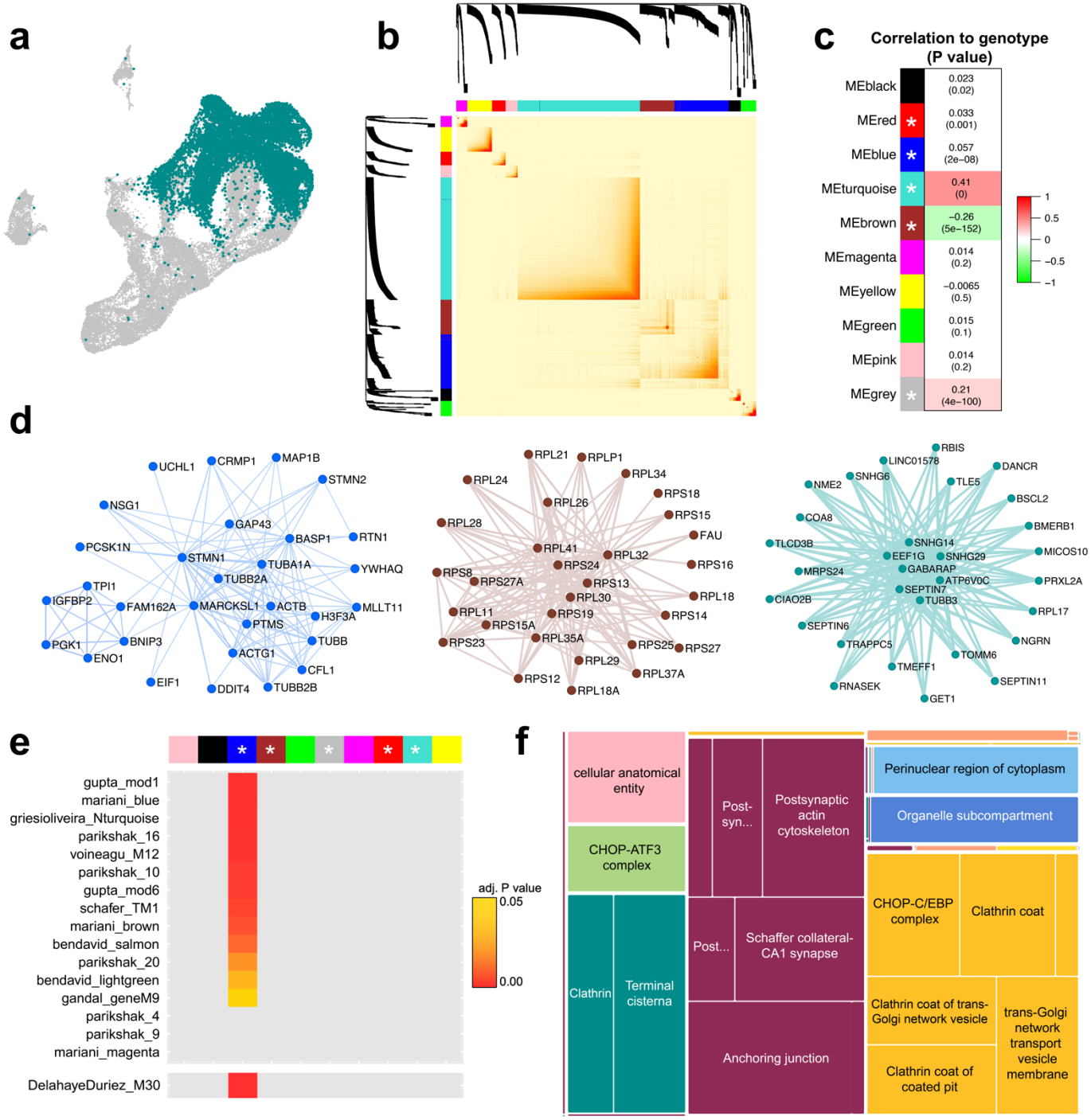

**a,** Feature plots show even expression in the neuron group of glutamatergic markers such as *PSD95*, *GRIN2B*, and *GLS*. A subset of neurons simultaneously expresses GABAergic markers such as *DLX2*. **b,** A pseudotime trajectory shows progression in pseudoage from radial glia to neurons. **c,** The neuron group is subclustered at a higher resolution in order to investigate drivers of GABAergic fate. **d,** Heatmap of aggregate expression of distinct gene clusters that vary over pseudotime in each subcluster shown in c. Note module 21 is specifically expressed in subclusters 7, 12, and 15, which are just prior to and entering the *DLX2*+ branch. **e,** Feature plots showing aggregate expression of transcripts in module 21 segregated by genotype. **f,** Dot plots of select module 21 genes show enhanced expression in PACS1$^{(+/R203W)}$ cells. **g,** Heatmap of aggregate expression of gene modules in cells just entering the GABAergic branch (clusters 7, 12, and 15 in panel c). Note module 21 is the most differentially expressed between genotypes.

**Fig. 5: PACS1$^{(+/R203W)}$ neurons have an increased density of inhibitory synapses.**




**a,** Percentage of total transcripts that comprise glutamatergic receptors or transporters. A full list of genes used is in Supplementary Fig. 5a. $p$ = 1 was calculated using a two-sided Wilcoxon rank-sum test with Bonferroni correction for all cells. **b,** Percentage of total transcripts that comprise GABAergic receptors or transporters. A full list of genes used is in Supplementary Fig. 5a. $p$= 8.07e-258 was calculated using a two-sided Wilcoxon rank-sum test with Bonferroni correction for all cells. **c,** Expression of excitatory and inhibitory synapse markers in day 79 organoid sections. Co-localization of pre-synaptic marker SYN1 and post-synaptic marker PSD95 was used to identify excitatory synapses, while co-localization of post-synaptic marker GPHN and pre-synaptic marker vGAT was used to define inhibitory synapses. Scale bar = 10 µm in the main panels and 1 µm in the zoomed panel. **d,** Quantification of colocalized synaptic puncta as shown in c. The dashed lines show the median and the dotted lines indicates quartiles. Data is from n=14-20 organoid sections across 4 PACS1$^{(+/+)}$ and 5 PACS1$^{(+/R203W)}$ organoids. $p$ = 0.15 for excitatory and 0.019 for inhibitory synapses. Values were calculated using a two-tailed t test assuming unequal variance.

**Fig. 6: Gene co-expression analysis reveals an ASD-specific gene network upregulated in PACS1$^{(+/R203W)}$ neurons**




**a,** Input for WGCNA was 4,884 control and PACS1$^{(+/R203W)}$ cells highlighted in teal. **b,** A topological overlap matrix shows hierarchical clustering of genes into modules based on their relative interconnectedness. **c,** 10 distinct gene modules were identified with WGCNA. Correlation was calculated using the Pearson method. *p* value was calculated using a student asymptotic *p* value test. White stars indicate *p* value significance after Bonferroni correction using all cells. **d,** Top 30 gene members of the modules with a positive relationship between a gene's association with genotype and importance within a module. Mitochondrial transcripts were removed from the blue module. **e,** Heatmap showing *p* values from a gene overlap test between members of the modules identified in this study, and neuronal ASD-specific modules from eight other studies and one epilepsy-specific module. *p* values were calculated using Fisher's exact test and corrected for multiple testing using the Benjamini-Hochberg procedure. **f,** A tree map of gene ontology cellular component analysis of the blue module members.[52] The large magenta area suggests enrichment at synapses.