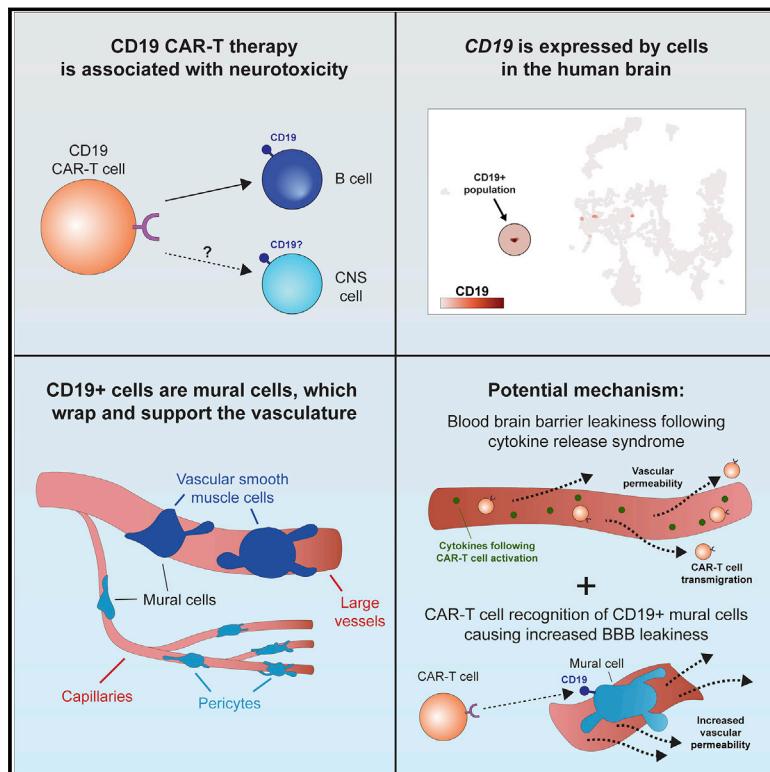


# Single-Cell Analyses Identify Brain Mural Cells Expressing CD19 as Potential Off-Tumor Targets for CAR-T Immunotherapies

## Graphical Abstract



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## In Brief

Single-cell RNA sequencing analysis shows that CD19, primarily considered as a B cell-specific surface antigen, is expressed in human brain mural cells that are critical for blood-brain-barrier integrity, suggesting that this cell population may contribute to the neurotoxicity of CD19-directed immunotherapy including CAR-T.

## Highlights

- Single-cell RNA-seq reveals CD19 expression in human brain mural cells
- Mural cells line blood vessels and maintain blood-brain barrier integrity
- Brain mural cell CD19 expression is present across brain regions and human age
- Targeting CD19<sup>+</sup> mural cells may contribute to neurotoxicity of CAR-T therapy

Article

# Single-Cell Analyses Identify Brain Mural Cells Expressing CD19 as Potential Off-Tumor Targets for CAR-T Immunotherapies

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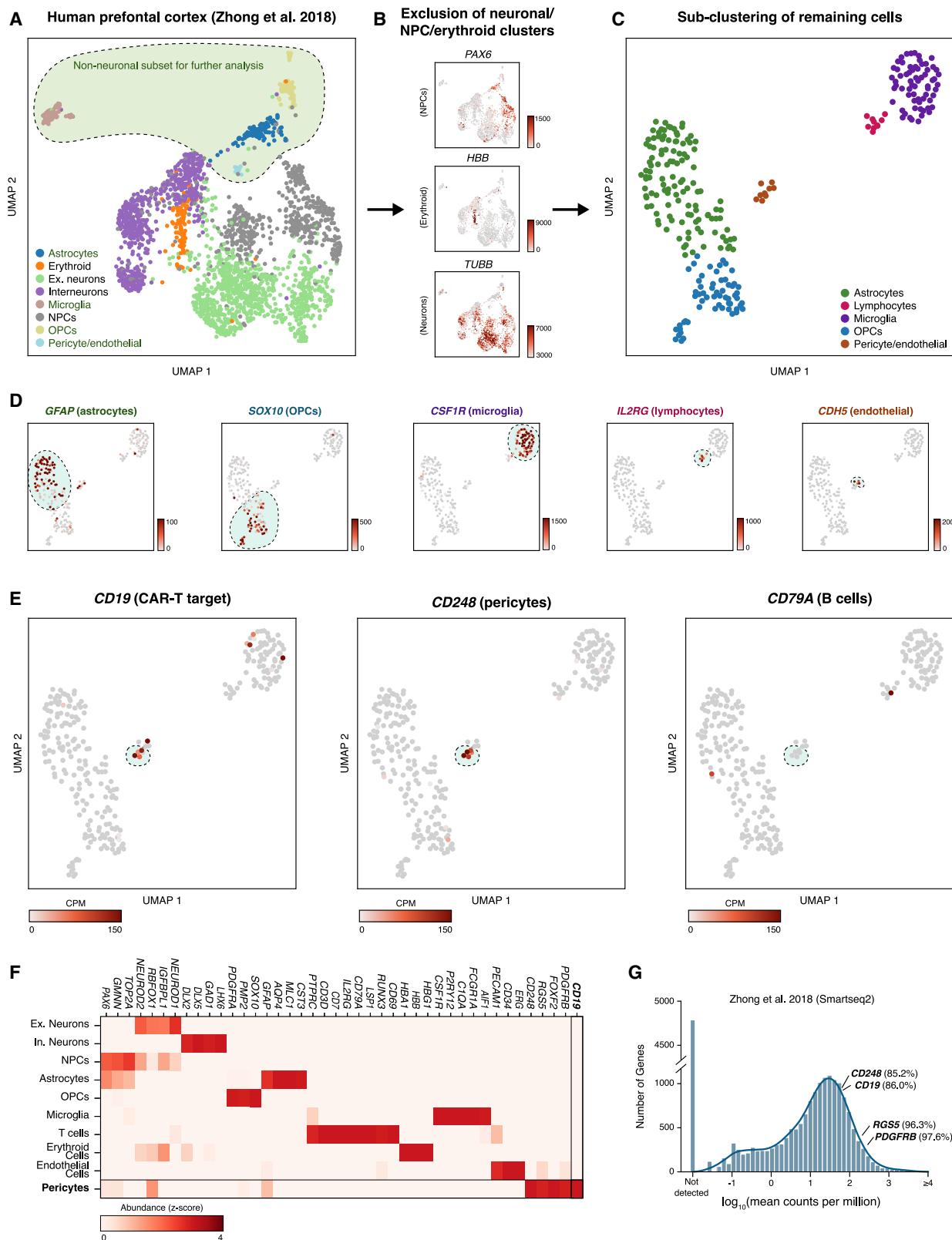
## SUMMARY

CD19-directed immunotherapies are clinically effective for treating B cell malignancies but also cause a high incidence of neurotoxicity. A subset of patients treated with chimeric antigen receptor (CAR) T cells or bispecific T cell engager (BiTE) antibodies display severe neurotoxicity, including fatal cerebral edema associated with T cell infiltration into the brain. Here, we report that mural cells, which surround the endothelium and are critical for blood-brain-barrier integrity, express CD19. We identify CD19 expression in brain mural cells using single-cell RNA sequencing data and confirm perivascular staining at the protein level. CD19 expression in the brain begins early in development alongside the emergence of mural cell lineages and persists throughout adulthood across brain regions. Mouse mural cells demonstrate lower levels of Cd19 expression, suggesting limitations in preclinical animal models of neurotoxicity. These data suggest an on-target mechanism for neurotoxicity in CD19-directed therapies and highlight the utility of human single-cell atlases for designing immunotherapies.

## INTRODUCTION

CD19-targeting CAR-T cells have shown tremendous clinical efficacy in patients with B cell leukemia and lymphoma, including

those who have relapsed after receiving traditional chemotherapy regimens (Brentjens et al., 2003; Kochenderfer et al., 2010; Porter et al., 2011; Brentjens et al., 2013; Grupp et al., 2013). For example, in a recent phase II study of 111 patients



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with refractory B cell lymphoma, of whom 101 were administered CD19 CAR-T cell therapy, 40% of patients showed a complete remission of disease 15 months after treatment (Neelapu et al., 2017). Similar results were observed in a separate study, with complete remission in 43% and 71% of patients with diffuse large B cell and follicular lymphoma, respectively (Schuster et al., 2017). However, in addition to adverse effects related to cytokine release syndrome (CRS), a systemic inflammatory response caused by the massive activation of CAR-T cells after infusion, both studies reported a high incidence of neurotoxicity (64% and 39%, respectively). This is in agreement with previously reported rates of neurotoxicity in patients receiving CD19 CAR-T cell therapy as well as CD19/CD3 BiTE therapy, which uses bi-specific antibodies to recruit T cells to CD19-positive cells (Goebeler et al., 2016; Gust et al., 2017; Nagorsen et al., 2012).

Analyses of patients who received CD19 CAR-T therapy and displayed neurologic adverse reactions revealed edema, multifocal hemorrhage, and vascular disruption of the brain (Gust et al., 2017). Neuropathologic evaluation of brain tissue from a patient who developed fatal neurotoxicity revealed CAR-T cell infiltration into the brain, indicating that CAR-T cells are able to traffic to and ultimately infiltrate the brain (Gust et al., 2017). This is in agreement with a recent study that showed that in human patients, blinatumomab, a CD19/CD3 BiTE, leads to T cell adhesion and perivascular transmigration across brain endothelium (Klinger et al., 2020). However, the underlying mechanisms behind why neurotoxicity is observed with CD19-targeting immunotherapies have been harder to resolve (Ruellá and June, 2018). One possible mechanism could be the unanticipated expression of CD19 on non-B cells in the brain. This mechanism could explain the higher incidence of neurotoxicity in CD19-directed immunotherapies, compared to treatments targeting other B cell proteins, such as CD20. Notably, CD19 CAR-T cells are sensitive to even low levels of CD19 antigen density, emphasizing the importance of identifying any potential reservoir of CD19 other than B cells (Majzner et al., 2020). We therefore sought to investigate whether a population of CD19-expressing cells is present in the human brain.

Using single-cell RNA sequencing (RNA-seq) data generated from human brain, we show that CD19 is expressed in mural cells. Mural cells are an integral part of the neurovascular unit (NVU), which surround endothelial cells and are critical for regulating the integrity of the blood-brain barrier (BBB) (Sweeney et al., 2019). Mural cells include pericytes and vascular smooth

muscle cells (vSMCs), which are closely related cell types that differ anatomically: pericytes localize along capillaries, while vSMCs are found along larger vessels. Genetic mouse models have shown that pericytes are crucial components of the BBB, which does not properly develop in their absence (Armulik et al., 2010; Daneman et al., 2010). We found that CD19 expression in human mural cells was observed across multiple datasets, brain regions, and developmental time points. Importantly, the expressed isoform of CD19 in the brain contains the epitope targeted by clinical CAR-T cells and BiTEs (Sommermeyer et al., 2017; Sotillo et al., 2015). More broadly, this study highlights the utility of efforts like the Human Cell Atlas (Regev et al., 2017), whose large-scale generation of tissue-wide single-cell RNA sequencing (scRNA-seq) data can be used to inform target selection in cancer immunotherapy.

## RESULTS

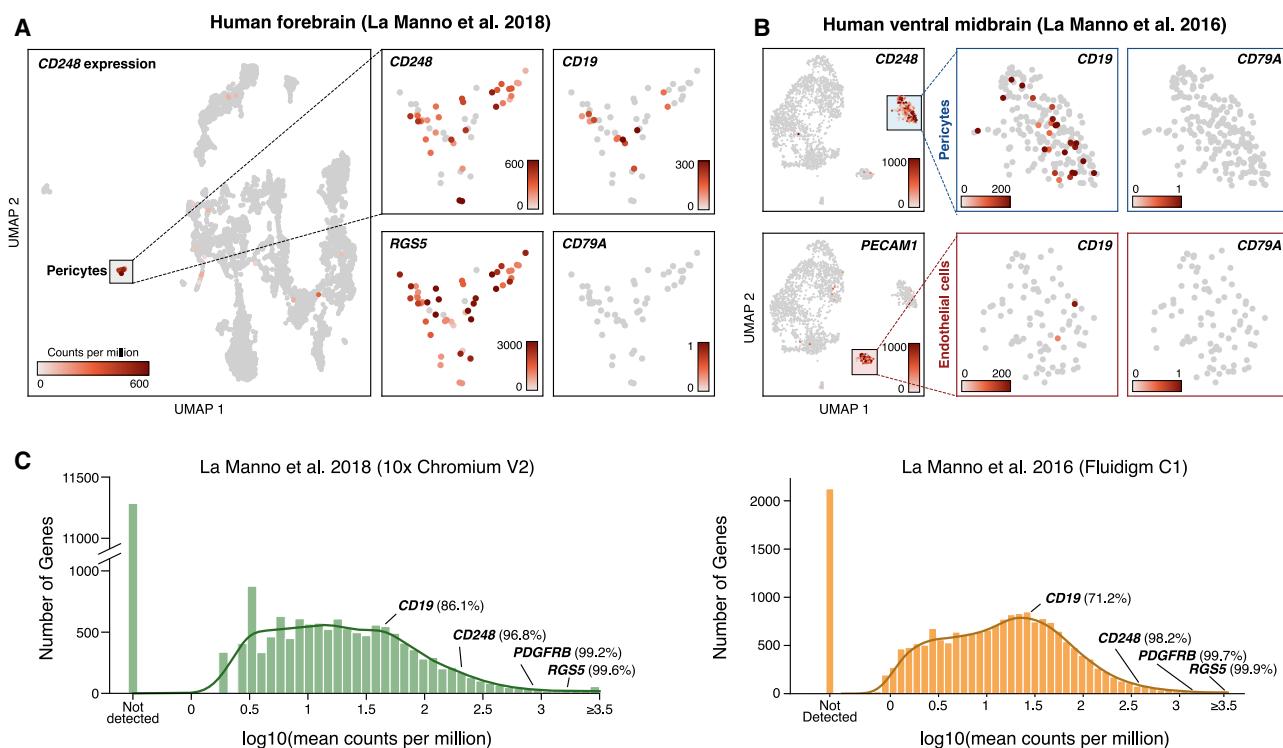
### CD19-Positive Mural Cells Identified by scRNA-Seq in Human Brain

We hypothesized that if a population of CD19-expressing cells is present in the human brain, it might be identifiable in data generated from recent efforts to map the human brain transcriptome with single-cell resolution. We first analyzed scRNA-seq data from 2,364 human prefrontal cortex cells (Zhong et al., 2018) Figure 1A). We clustered cells and identified broad populations, focusing subsequent analyses on non-neuronal, non-erythroid cells. These further segregated into astrocyte, lymphocyte, microglial, oligodendrocyte precursor, endothelial, and pericyte populations (Figures 1B, 1C, and S1A). These populations were identified based on the expression canonical marker genes; mural cells specifically expressed expected marker genes, such as *PDGFRB*, *FOXF2*, *RGS5*, and *CD248*, while endothelial cells expressed a distinct set of markers, including *CDH5* and *PECAM1* (*CD31*; Figure 1D). Surprisingly, this analysis revealed a small population of cells (~1.5% of non-neuronal cells; ~0.2% including neuronal cells) that expressed CD19 and co-expressed the mural cell marker *CD248* (Figures 1D and S1B). These cells were negative for the vSMC marker *ACTA2*, indicating that these cells were pericytes (Figure S1C). Importantly, this population was negative for the B cell marker *CD79A*, arguing against the possibility of artifactual CD19 expression due to B cell-pericyte doublets (Figure 1E). Additionally, these cells were positive for the tetraspanin *CD81*, which chaperones CD19 through secretory pathways to the plasma

**Figure 1. Identification of CD19 Expressing Pericytes in Single-Cell RNA Sequencing Data**

- (A) Uniform manifold approximation and projection (UMAP) of single-cell prefrontal cortex RNA-seq data.
- (B) The neuronal, neural progenitor cell, and erythroid clusters were identified as shown and subsequently excluded from further analysis. Unit, counts per million (CPM).
- (C) Non-neuronal cells are re-clustered and embedded using UMAP to more clearly distinguish populations.
- (D) Cell populations are identified by shown marker genes. Note that pericytes and endothelial cells are called as a single cluster but identifiable as separate populations by marker gene expression. Scale bar indicates CPM.
- (E) *CD248*-positive pericytes are also positive for the CAR-T target *CD19* but negative for the B cell marker *CD79A*.
- (F) Heatmap showing relative gene abundances for specific marker genes across identified cell clusters. Pericytes are identified by *PDGFRB*, *CD248*, *RGS5*, and *FOXF2* expression and specifically express *CD19*.
- (G) Histogram of mean abundance of all genes in the human prefrontal cortex. Percentiles indicate the expression rank of a particular relative to all detected genes.

See also Figure S1.



**Figure 2. Confirmation of Mural Cell CD19 Expression in Two Independent Datasets**

(A and B) UMAP plots showing single-cell RNA-seq data from (A) human forebrain (La Manno et al., 2018) and (B) human ventral midbrain (La Manno et al., 2016), colored by gene-expression value, showing CD19 expression in pericytes. (C) Histogram of mean gene-expression values ( $\log_{10}$  counts per million) in identified pericyte cells in La Manno et al. (2018) and La Manno et al. (2016). Relative gene-expression percentiles are shown for indicated genes.

membrane and is required for surface expression of CD19 in B cells (Figure S1D; (Maecker and Levy, 1997; Miyazaki et al., 1997; Shoham et al., 2003; Tsitsikov et al., 1997; van Zelm et al., 2010; Braig et al., 2017)). We also observed a separate population of lymphocytes that expressed the marker genes *CD45* (*PTPRC*), *CD3*, *CD7*, and *IL2RG* (*CD132*), representing a population of T cells (Figures 1E and 1F). Notably, the expression of lymphocyte markers and pericyte markers was mutually exclusive in the observed clusters, and the expression of CD19 was specific to the pericyte cluster (Figure 1F).

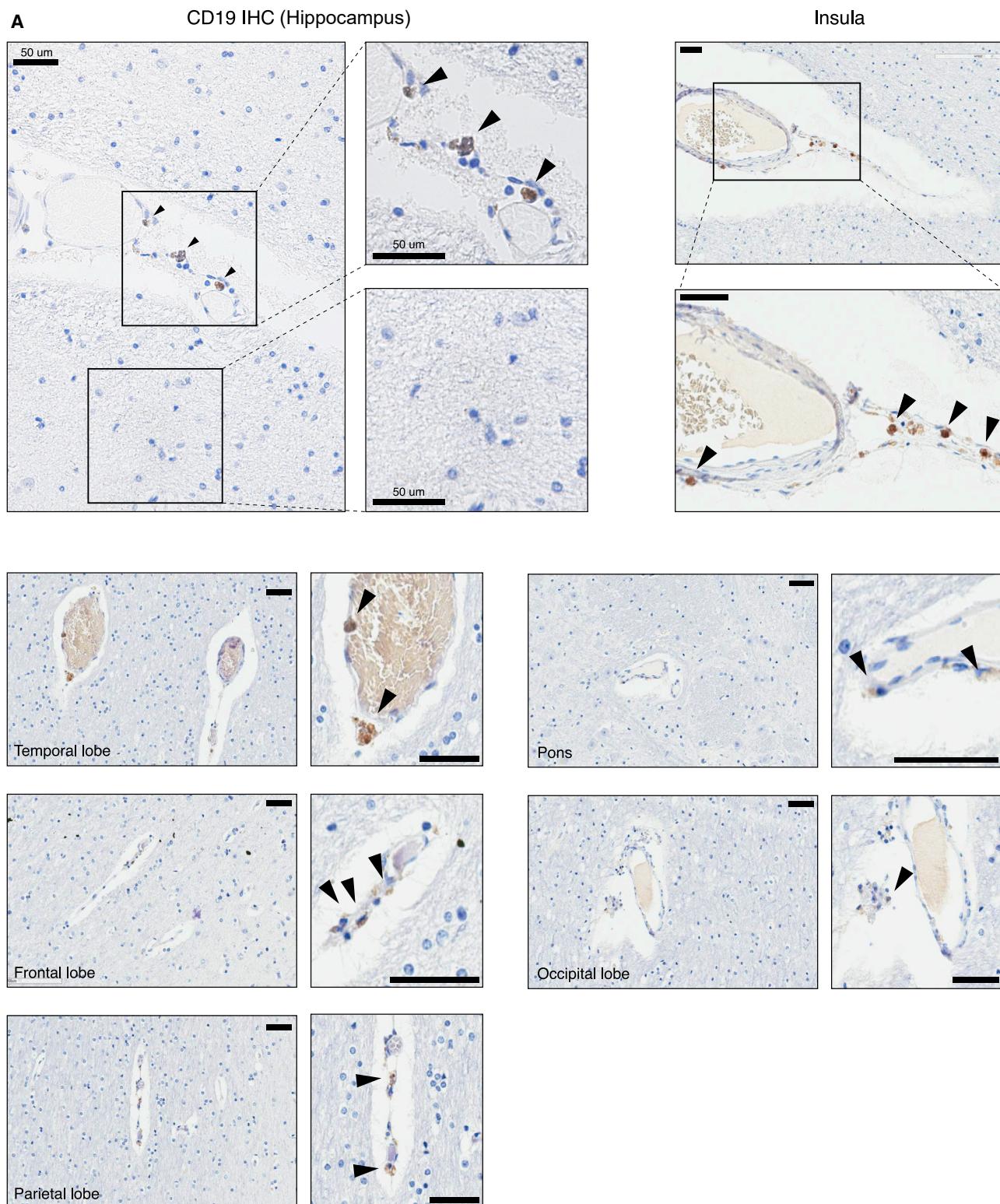
Due to the inherent sparsity of single-cell data, even highly expressed genes will not be detected in every cell, causing single-cell data to be zero-inflated (Butler et al., 2018). To ask at what level CD19 is expressed in pericytes, we compared the expression of CD19 to known pericyte marker genes, relative to the overall distribution of gene expression in pericytes. Mean-normalized gene-expression values can provide a rough estimate of the relative expression of a given gene in a cell population of interest, despite the inherent sparsity of single-cell data. As expected, pericyte markers such as *CD248*, *RGS5*, and *PDGFRB* ranked in the top percentiles of gene expression in pericytes from the human prefrontal cortex (85<sup>th</sup>, 96<sup>th</sup>, and 98<sup>th</sup>, respectively) (Figure 1G; STAR Methods). CD19 was similarly expressed highly, in the 86<sup>th</sup> percentile for gene expression.

We next sought to reproduce this observation in additional independent scRNA-seq datasets. In a dataset from human fore-

brain (La Manno et al., 2018), a mural cell population (48/7906 cells) positive for the markers *CD248* and *RGS5* also displayed CD19 expression (12/48 cells had measurable CD19) and the absence of the B cell marker *CD79A* (Figure 2A). In a third dataset derived from human ventral forebrain (La Manno et al., 2016), we again observed expression of CD19 in a population of *CD248*-positive mural cells (195 mural cells/1977 cells, of which 25/195 had measurable CD19; Figure 2B). As with other datasets, no RNA sequencing counts for *CD79A* were identified in any mural cells. Although spurious non-mural CD19-positive cells were observable, these were likely the result of cell-cell doublets, such as mural-endothelial cell doublets observed due to the adhesion of mural to endothelial cells in the brain. In these datasets, CD19 ranked in the 86<sup>th</sup> and 71<sup>st</sup> percentiles for gene expression, suggesting relatively robust expression (Figure 2C).

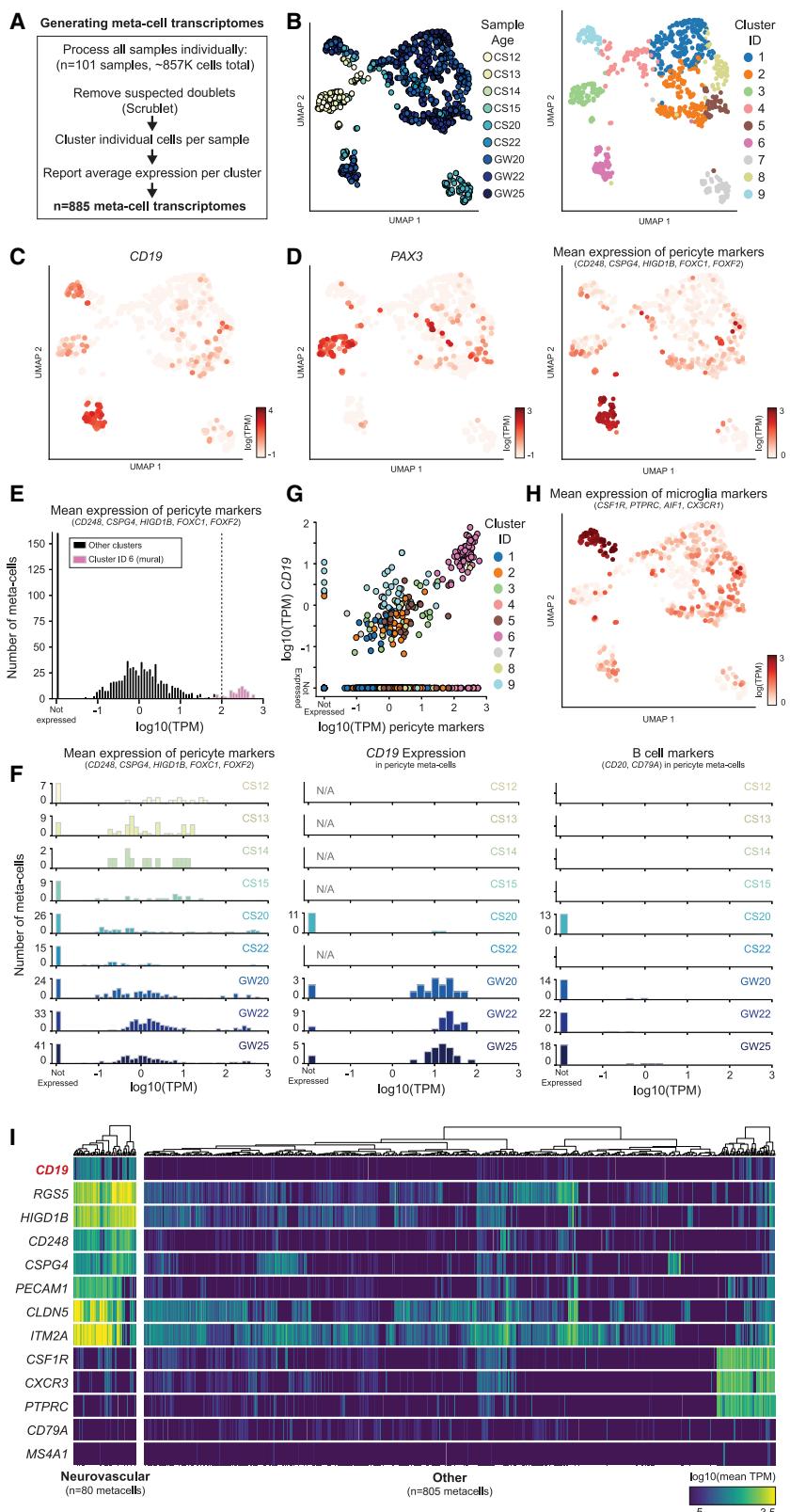
### CD19 Expression in Adult Human Brain by IHC

To assess the expression of CD19 protein in human mural cells, we performed immunohistochemistry on several regions of the human brain using a clinically validated anti-human CD19 antibody (clone BT51E) on samples from healthy deceased subjects. This antibody recognizes the C terminus of the CD19 protein, which is cytoplasmically localized. We found CD19 expression on cells present adjacent to the vessel basement membrane walls in perivascular areas (Figure 3). Abluminal CD19



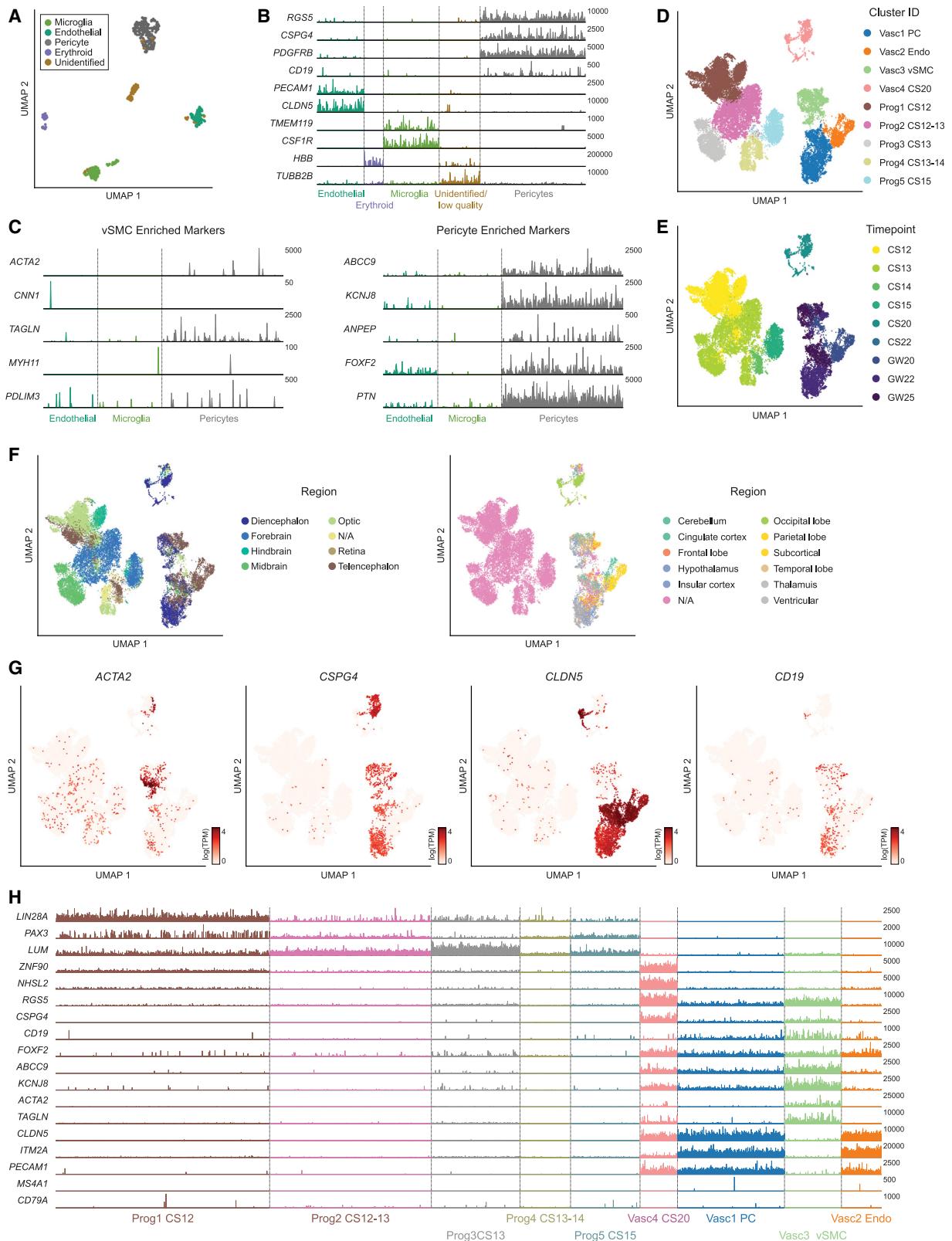
**Figure 3. Perivascular Staining of CD19 in Human Brain**

Representative immunohistochemistry staining for CD19 in human brain tissue. Formalin fixed paraffin embedded (FFPE) samples were stained for CD19 with a clinical protocol. Representative staining is shown for the hippocampus, insula, temporal lobe, frontal lobe, parietal lobe, pons, and occipital lobe. Scale bar, 50 μm. 5 slides were stained for the hippocampus, and 10 slides were stained for other brain regions.



**Figure 4. Meta-Cell Clustering Identifies CD19 Expression in Human Neurovascular Meta-Cells**

- (A) Schematic showing how samples were processed (see *Figure S2* for an example of a single dataset).
- (B) UMAP projection of meta-cells, colored by either sample age or cluster.
- (C and D) Expression of (C) CD19 and (D) PAX3, marking undifferentiated progenitors, and mean expression of indicated pericyte marker genes.
- (E) Histogram of mean mural cell marker expression across all samples, showing separation of the identified neurovascular meta-cell cluster.
- (F) Histogram as in (E) but separated by sample age, showing (left) pericyte marker gene expression, (middle) CD19 expression, (right) B cell marker gene expression. Note that CD19, but not B cell markers, are expressed in neurovascular meta-cells.
- (G) Scatterplot showing the correlation of pericyte marker genes with CD19 expression. Note the separation of the neurovascular cluster.
- (H) UMAP projection of meta-cells, colored by mean expression of microglia markers.
- (I) Heatmap showing log10 average TPM values for selected genes across meta-cells. Dendograms indicate Ward's hierarchical clustering of each of the two populations shown.
- See also *Figures S2*.



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expression was observed across multiple brain regions, with particular regions, such as the hippocampus, insula, temporal lobe, frontal lobe, and parietal lobe displaying a comparatively higher, albeit still rare, incidence of CD19-positive cells. In contrast, regions such as the pons and occipital lobe displayed lower rates of CD19-positive cells (Figure 3). Notably, CD19-positive cells were found along smaller capillaries ( $<8\ \mu\text{m}$ ) as well as larger vessels ( $>8\ \mu\text{m}$ ; majority of cells depicted), suggesting that, in addition to pericytes, CD19 may also be expressed in vSMCs. It is possible that the CD19 staining observed by immunohistochemistry (IHC) may also be due to other cell types, such as passing B cells or glia. However, the abluminal localization along the vasculature of CD19<sup>+</sup> cells is most consistent with staining of mural cells. While we found perivascular CD19 staining across multiple brain regions, abluminal CD19 staining was not found along all vessels, suggesting that there may be heterogeneity in the frequency of CD19<sup>+</sup> mural cells. This may contribute to the inter-patient differences in severity and outcome of neurotoxicity observed clinically.

### Analysis of Pericytes and vSMCs Shows CD19 Expression across Mural Cells

Based on the perivascular staining of CD19 along both larger and smaller vessels, which suggested that CD19 expression might be a more general feature of mural cells, we sought to analyze the transcriptome of pericytes and vSMCs. We analyzed a large scRNA-seq dataset generated by the BRAIN Initiative Cell Census Network (BICCN, Kriegstein/UCSF), comprising scRNA-seq data from diverse human brain regions over many developmental time points spanning Carnegie Stage 12 (CS12), corresponding to gestational week (GW) 4, through GW25. This dataset included roughly 857,000 cells across 101 individual samples, affording us the power to perform a detailed analysis of the cell types within the NVU.

We first processed all 101 samples individually, using the same pipeline, removing any cells identified as putative doublets (Wolock et al., 2019) and then identifying distinct clusters of transcriptionally similar cells (Figures 4A and S2A). These clusters were aggregated into “meta-cell” transcriptomes (STAR Methods), representing the average gene expression across the cells in that cluster, allowing us to efficiently identify populations of interest within the larger dataset. Together, this analysis identified 885 meta-cells, which were then analyzed together. We identified highly variable genes across the 885 meta-cells and then used these genes to perform dimensionality reduction and clustering (Figures 4B and S2B). Importantly, the clusters of

meta-cells were largely similar on quality measures, such as the number of single cells per meta-cell, and mean number of counts per meta-cell (Figure S2C).

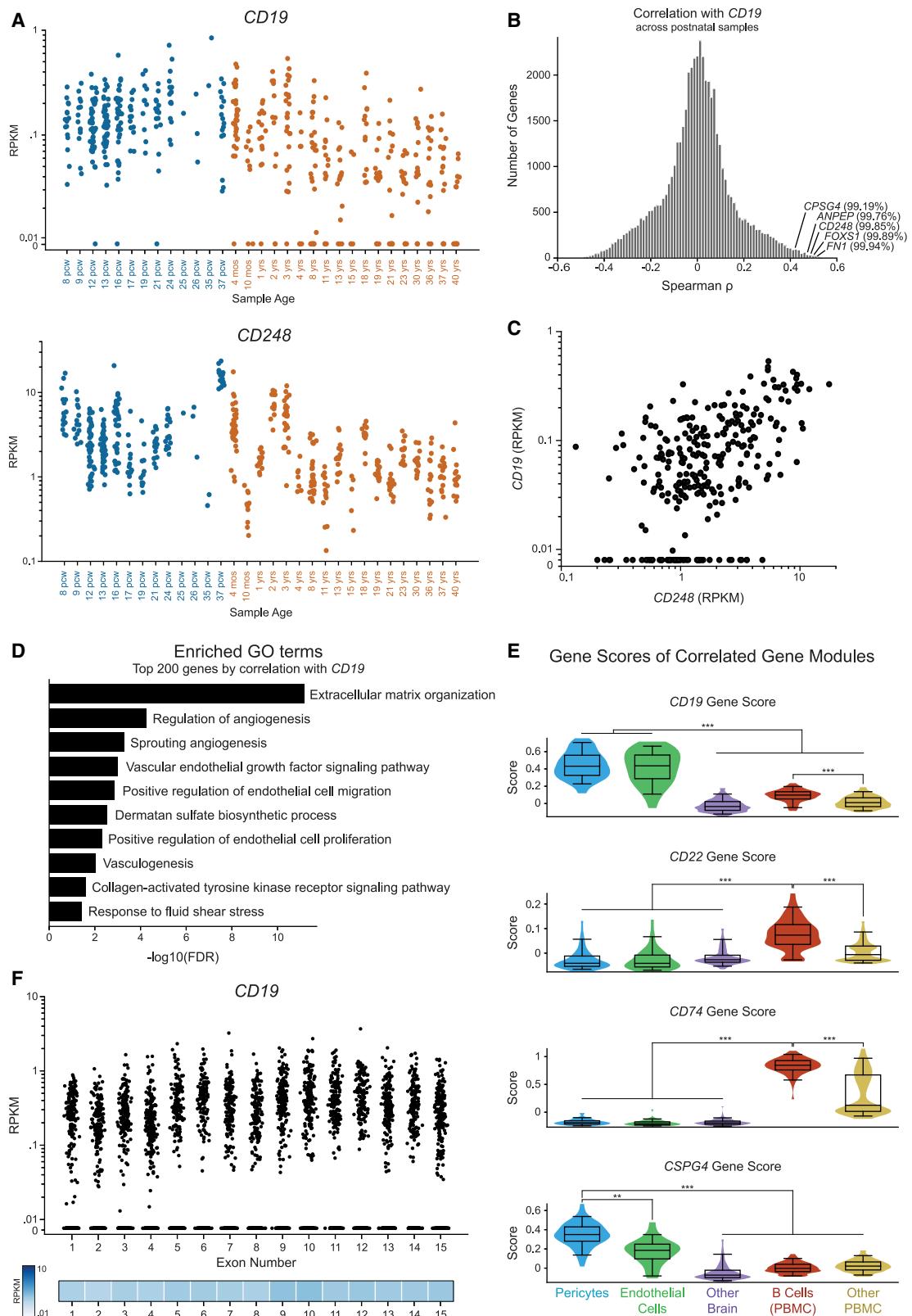
This analysis revealed that CD19 is highly expressed in neurovascular meta-cell clusters, consistent with our prior findings (Figures 4C and 4D). The meta-cell analysis does not show a clear separation between mural and endothelial cells (or other NVU cell types), since meta-cells may include closely clustering cell types in sparse individual samples. However, this clustering shows clear separation of meta-cells expressing mural markers, and the expression of CD19 and absence of B cell markers in these meta-cells (Figures 4E and 4F). Additionally, we observed a strong correlation between expression of mural cell marker genes and CD19 across the data (Figure 4G). Low-level CD19 expression was also observed in microglia clusters (Figures 4G and 4H). However, as mural cells and other cells of the NVU, which include perivascular macrophages, are tightly connected, this expression may also be the result of contamination from mural cells in microglia meta-cells.

Notably, mural meta-cells were not present in early time points (CS12–CS15), which had high expression of early developmental markers such as *PAX3* and *LIN28A* (Figures 4D and S2E). In contrast, mural meta-cells at GW25 displayed high and specific expression of mural cell marker genes, such as *CD248*, *RGS5*, and *FOXF2*, in addition to specific expression of CD19 (Figure S2F). In these data, we are able to clearly separate undifferentiated progenitors and differentiated mural cells, which demonstrates that, once mural cells emerge, they express CD19 (Figures 4E, 4F, S2E, and S2F). Collectively, there was clear separation of meta-cells expressing neurovascular markers from all other meta-cells (Figure 4I), which allowed us to identify cells of interest for further analysis.

Pericytes are closely related to vSMCs, which together are the mural cells that line the brain vasculature (Uemura et al., 2020). These cells differ based on anatomical position; pericytes localize around capillaries and vSMCs localize around larger vessels, including arteries, arterioles, and venules. However, these cells are transcriptionally similar, sharing the identity of many marker genes and appearing to exist on a transcriptional lineage continuum (Vanlandewijck et al., 2018). *ACTA2*, encoding alpha-smooth muscle actin, is a canonical marker used to distinguish these two populations, which is significantly upregulated in vSMCs. Many pericyte markers, however, such as *CSPG4* and *RGS5*, are also highly expressed in vSMCs, causing brain vSMCs to often be annotated as pericytes (Uemura et al., 2020).

### Figure 5. CD19 Is Expressed in Both Pericytes and vSMCs

- (A) Subset of non-neuronal cells from Zhong et al. (2018), La Manno et al. (2016), and La Manno et al. (2018).
- (B) Expression of marker genes used for clustering. Max TPM per gene (y axis) is indicated.
- (C) (Left) Low expression of vSMC marker genes and (right) high expression of pericyte marker genes. y axis labels indicate maximum TPM value shown.
- (D–F) Neurovascular and progenitor subset of BICCN data annotated by (D) cluster ID, (E) time point, or (F) region. Note that samples were annotated at different levels of regional granularity, so the resulting annotations are sometimes overlapping.
- (G) Expression of vSMC marker (*ACTA2*) as well as *CSPG4* (pericyte) and *CLDN5* (endothelial) markers. CD19 is expressed primarily in the vSMC and pericyte clusters.
- (H) Track plot showing the lack of early developmental marker expression and distinguishing markers between vSMC and pericyte clusters. y axis labels indicate maximum TPM value shown.
- See also Figure S3.



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We first asked whether the cells we had originally identified as *CD19*<sup>+</sup> mural cells represented pericytes or vSMCs. We aggregated the three datasets from [Zhong et al. \(2018\)](#) and [La Manno et al. \(2016\), \(2018\)](#) and analyzed them as a single integrated dataset ([Figure S3A](#)). We identified a subset of non-neuronal clusters enriched for mural marker genes (*CD248*, *CSPG4*) and confirmed that this population expressed both *CD19* and *CD81* ([Figures S3B](#) and [S3C](#)). We also identified subsets expressing the endothelial and microglial markers *CD248* and *CSF1R*, respectively ([Figure S3B](#)). The non-neuronal subset (mural cells, endothelial cells, and microglia) was then re-analyzed and clustered to distinguish between transcriptional differences in cell types of the NVU ([Figures 5A](#) and [5B](#)). We observed that these cells showed strong enrichment of reported pericyte markers, such as *ABCC9* and *KCNJ8*, without enrichment of *ACTA2* or other vSMC marker genes, suggesting that they represented bona-fide pericytes ([Figure 5C](#); [Chasseigneaux et al., 2018](#); [Vanlandewijck et al., 2018](#)).

We repeated this analysis using the cells from the BICCN dataset, performing a detailed analysis of neurovascular cells and related progenitors. Using the meta-cells as a guide, we identified ~92K single cells representing the NVU as well as early progenitors ([Figures 4B](#) and [S3D](#)). We noticed a bifurcation of progenitors from the early time points that appeared to be neural lineage-biased and others that seemed biased toward non-neuronal lineages. The non-neuronal-biased progenitors, as well as the pericyte and endothelial cell clusters were then subset for further analysis ([Figure S3E](#)). These ~26K cells showed a clear separation between progenitors from CS12-CS15 and differentiated NVU cells from GW20-GW25 ([Figures 5D](#) and [5E](#)). They were also representative of cells from many brain regions ([Figure 5F](#)). Note that samples were annotated at different levels of granularity, such as forebrain versus diencephalon or telencephalon, and only later samples had finer-grained regional resolution. Notably, we identified a smaller vSMC population in this dataset, likely due to the high number of cells in the BICCN data, as pericytes were more abundant than vSMCs ([Figure 5G](#)). These cells showed high expression of *ACTA2* and *TAGLN*, as well as other pericyte markers such as *RGS5* ([Figure 5H](#)). Importantly, *CD19* was expressed in the vSMC population as well, consistent with the high transcriptional similarity between these two populations. This suggests that *CD19* expression is not unique to pericytes but a common feature in human brain mural cells, consistent with the staining pattern observed by IHC ([Figure 3](#)). We note that further assigning the identity of this vSMC population to venule, arterial, and arteriole sub-populations was challenging since canonical markers identified in mice did

not perfectly align with the sub-populations identified in this human dataset ([Vanlandewijck et al., 2018](#)), suggesting transcriptional differences between mouse and human mural cells ([Figure S3F](#)).

Within the BICCN dataset, we observed a higher degree of endothelial marker gene expression in mural cell clusters and vice versa. For example, while endothelial and mural cell clusters are distinguishable, the expression of *CLDN5*, an endothelial cell marker, and *CSPG4*, a mural cell marker, is more overlapping than would be expected ([Figure 5G](#)). We postulate that this is likely due to incomplete separation of these two cell types during tissue dissociation, as cells within the tightly interacting NVU are challenging to fully dissociate, and cross-contamination is common ([Vanlandewijck et al., 2018](#)). However, the analysis of previous datasets, as well as analyses of individual BICCN datasets, showed greater separation of neurovascular cells and clearly demonstrated that *CD19* expression is primarily in mural cells and not endothelial cells ([Figures 1F](#), [5B](#), and [S2A](#)).

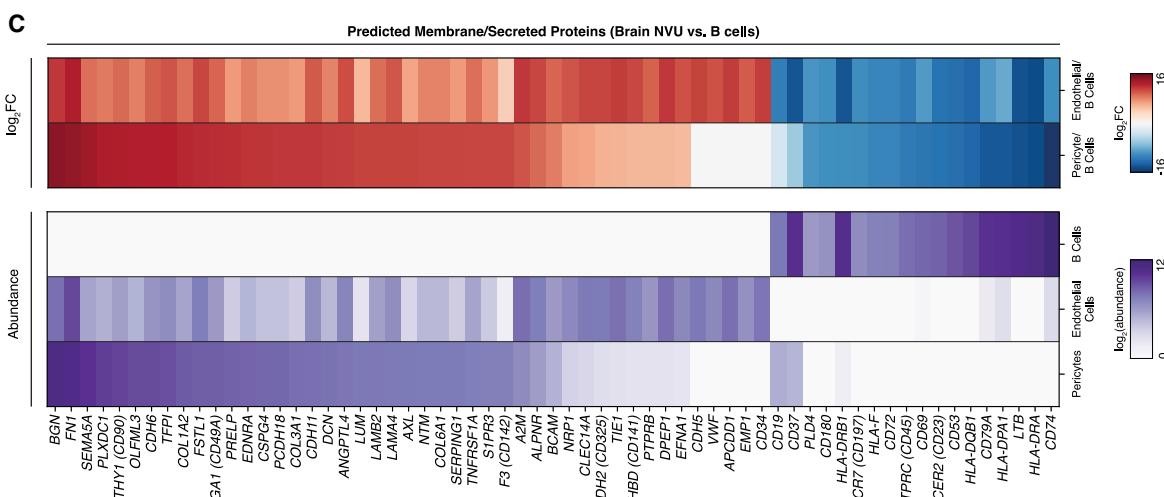
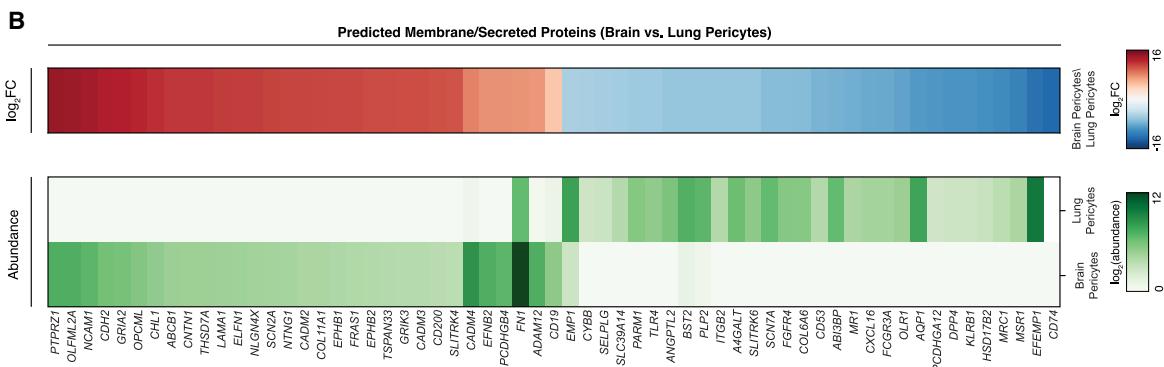
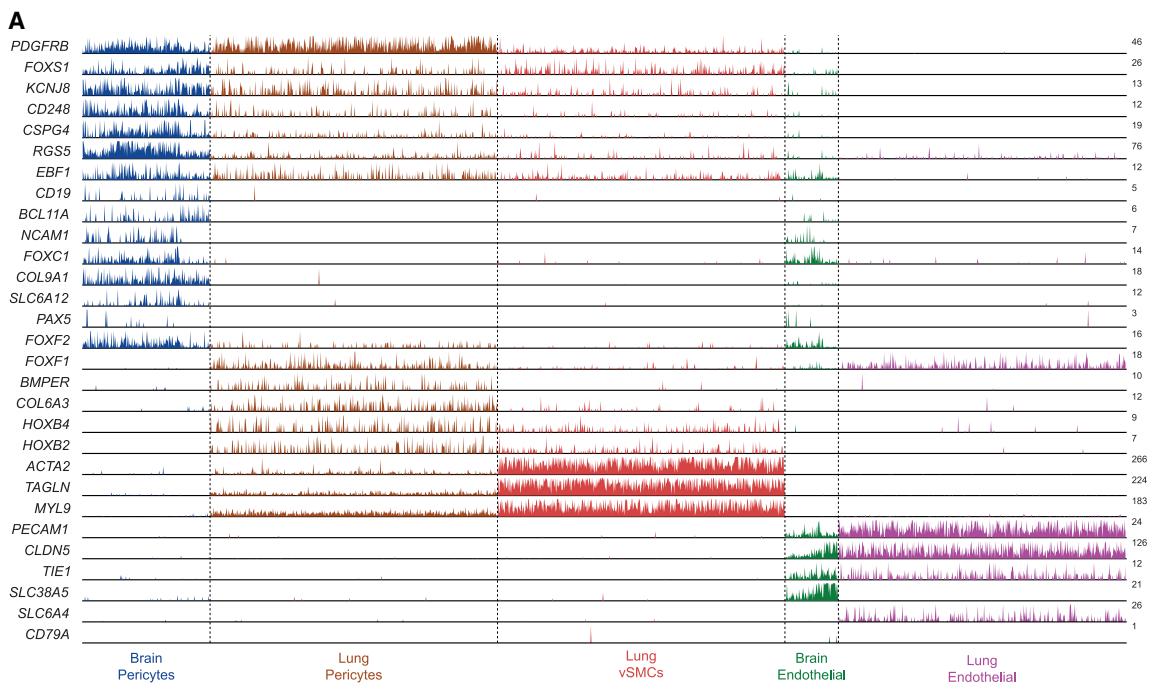
### The CD19 Isoform Recognized by CAR-T Cells Is Expressed in the Adult Human Brain

We next sought to analyze the expression of *CD19* in the adult human brain to complement the results observed previously by immunohistochemistry. scRNA-seq analysis of mural cells in adult human brain samples is challenging, due to difficulties in tissue acquisition and low numbers of neurovascular cells in the brain, relative to neuronal or glial populations ([Figure S3A](#)). Additionally, the majority of existing datasets are prospectively enriched for neuronal populations of interest. Therefore, to ask whether *CD19* expression in pericytes is also present in adult samples, we utilized bulk RNA-seq data across human age and brain regions generated by the Allen Institute Brainspan Project ([Miller et al., 2014](#)). These data contain more than 500 prenatal and postnatal samples from diverse brain regions (n = 237 prenatal; n = 287 postnatal). Since bulk tissue analyzed in different samples will have varying proportions of mural/vascular cells, the relative expression of mural genes across samples is a proxy for the underlying proportion of mural cells in the bulk tissue. For example, *CD248* and *ANPEP*, two pericyte markers, are highly correlated in these data ([Figure S4A](#)).

We first confirmed that *CD19* is expressed in both prenatal and postnatal samples at similar levels, and also expressed in samples from different brain regions ([Figure 6A](#); [Figure S6B](#)). We next performed a genome-wide correlation analysis across only the postnatal samples to identify genes that are co-expressed with *CD19*. Strikingly, this analysis showed that *CD248*, *CSPG4*, *ANPEP*, *FOXS1*, and *FN1* were in the top 1%

### Figure 6. The CAR-T-Recognized CD19 Isoform Is Expressed in the Adult Human Brain

- (A) Expression of *CD19* (top) and *CD248* (bottom) in bulk RNA-seq data from human brain. Colors indicate prenatal and postnatal samples, and the different samples from distinct regions but the same age are plotted on the same x coordinate.
- (B) Histogram of the distribution of spearman correlation values for all genes with *CD19* expression in only postnatal samples. The indicated percentiles indicate the percentile of that gene's correlation.
- (C) Scatterplot of *CD248* against *CD19* RPKM values in only postnatal samples.
- (D) Enriched GO terms in the top 200 genes by spearman correlation with *CD19*.
- (E) Gene score distribution in single cells belonging to pericyte or endothelial clusters, as well as other brain cells along with B cells and other PBMCs. Gene score was calculated with the top 30 genes by spearman correlation.
- (F) RPKM values per exon of *CD19* in the Brainspan data, showing expression of the key exons 2 and 4 for CAR-T cell recognition.  
See also [Figures S4](#), [S5](#), and [S6](#).



(legend on next page)

of genes correlated with *CD19* in these data (Figures 6B and 6C). Additionally, the top 200 genes most correlated with *CD19* were enriched for gene ontology (GO) terms associated with the NVU, such as angiogenesis, vasculogenesis, response to fluid shear stress, and multiple extracellular matrix related terms (Figure 6D). This suggests that the expression of *CD19* in the adult brain is primarily the result of mural cell abundance, rather than B cell abundance. We then integrated the *CD19*-correlated gene module in adult brain with scRNA-seq data from the human brain and peripheral blood mononuclear cells (PBMCs) (STAR Methods). This allowed for the comparison of the gene module score (i.e., the enrichment of genes correlated with *CD19* or another target) in brain pericytes, endothelial cells, and B cells, in addition to other brain cells and PBMCs (Figure S4C). As a control, the gene modules associated with *CD22* and *CD74*, two highly expressed B cell-enriched genes, were enriched in B cells, and the *CSPG4* gene module was highly enriched in pericytes (Figures 6E, S4D, and S4E). In contrast, the expression of the *CD19*-correlated gene module is highest in pericytes (Figure 6E). We noted that *CD19* expression seemed to decrease with age (Figure 6A). This could be explained by two potential mechanisms: either (1) the proportion of *CD19*<sup>+</sup> cells changes over time, or (2) the level of *CD19* expression in mural cells changes over time. However, the concomitant decrease in *CD248* expression with age, as well as prior studies (Erdő et al., 2017; Yang et al., 2020), support the former mechanism.

Last, we asked whether the *CD19* isoform that is expressed in the adult brain contains the specific *CD19* epitope that is recognized by clinical CAR-T cells and BiTEs. The FMC63 scFv in clinical use recognizes an epitope encoded by exon 4 of *CD19* (Sommermeyer et al., 2017). Additionally, variants skipping exon 2 may also result in a lack of *CD19* trafficking to the cell surface, also allowing evasion of FMC63 detection (Sotillo et al., 2015). An analysis of the mean expression of each *CD19* exon in the adult human brain showed an even distribution of expression across exons, which clear expression of the key exons 2 and 4 (Figure 6F).

### Analysis of Neurotoxicity in Mouse Models of *CD19* CAR-T Cell Therapy

We next asked whether the expression of *CD19* in human mural cells is conserved in mice, which would allow for the use of mouse models to study mechanisms of human neurotoxicity. First, we extracted and dissociated whole brains from healthy C57BL/6J mice following a previously described protocol and analyzed the presence of *CD19* by flow cytometry (Boroujerdi et al., 2014). This revealed the presence of a CD45-high *CD19*<sup>+</sup> B cell population, as expected, and also revealed a rare population of CD45-*CD19*<sup>+</sup> cells, which expressed *CD19* at a similar

level to B cells but did not express the B cell marker B220 (Figures S5A–S5C). Pericytes, identified as a *CD13*<sup>+</sup>*CD45*<sup>−</sup> population, were indeed positive for *CD19*, albeit rarely, in contrast with what was observed in human data (Figure S5D). To examine this at the transcriptional level, we performed scRNA-seq on cells isolated from dissociated whole mouse brain, followed by enrichment for *CD19*<sup>+</sup>, *CD13*<sup>+</sup>, and *CD31*<sup>+</sup> cells. This analysis again confirmed low *Cd19* expression in pericytes relative to *CD19*<sup>+</sup> B cells, and in fewer cells compared to scRNA-seq of human mural cells (Figure S5E). Together, these data suggest that *CD19*<sup>+</sup> mural cells are relatively less abundant in the mouse brain, compared to human brain.

Given the presence of non-B *CD19*<sup>+</sup> cells in mice, we asked whether an infusion of *CD19*-directed CAR-T cells into immunodeficient, non-tumor bearing NOD.Cg-Prkdc<sup>scid</sup> /I2rg<sup>tm1Wjl</sup>/SzJ (NSG) mice would result in an observable neurologic phenotype. Since NSG mice do not develop B cells, any phenotype observed in this model is B cell independent and cannot be attributed to global effects related to CRS following B cell targeting. We generated either CD28 (28)-based or 4-1BB (BB)-based CAR-T cells, which differ in the signaling domain but are both used clinically, targeting either human (FMC63 scFv) or mouse *CD19* (1D3 scFv, the same as used for flow cytometry analysis above). This comprised three conditions: hCD19BBz, representing a negative control that should not target murine *CD19*, mCD19BBz, and mCD1928z. We confirmed species-specific activation using human and murine *CD19*<sup>+</sup> B-ALL cell lines (Figure S6A).

Seven days after infusion of CAR-T cells, NSG mice that had received CAR-T cells targeting murine *CD19*, but not those mice receiving CAR-T cells targeting human *CD19*, displayed increased BBB permeability as measured by Evans Blue dye (EBD; Figures S6B–S6D). EBD binds to albumin, which normally remains in the bloodstream under physiologic conditions; when the BBB is disrupted, albumin is able to cross into the brain parenchyma. A fourth group of mice receiving no transfusion displayed no EBD staining, and a fifth group that received mannitol at the time of EBD infusion, representing a positive control, displayed high EBD staining, as expected (Figures S6C and S6D). The effect was higher in the mCD1928z compared to mCD19BBz CAR-T cell group, which is likely the result of stronger antigen receptor signaling provided by the CD28 domain (Salter et al., 2018). Indeed, prior studies have shown that the choice of costimulatory domain (CD28, 4-1BB, or others) strongly impacts sensitivity to antigen density with CD28-based CARs particularly sensitive to low levels of antigen density (Majzner et al., 2020). This experiment was repeated using a syngeneic, immunocompetent C57Bl/7J model with pretreatment with cyclophosphamide as a lymphodepleting preconditioning regimen. The

**Figure 7. Brain Pericyte-Specific Expression of *CD19***

(A) Track plot showing expression of selected marker genes for each population. Note that *CD19* expression is limited to brain pericytes but not lung pericytes. Additionally, brain pericytes express certain transcription factors, such as *BCL11A*, that are enriched in B cells. y axis labels indicate maximum TPM value.  
(B) (Top) Heatmap of log2 fold change in gene expression of surface/secreted genes between brain and lung pericytes. (Bottom) Heatmap of expression of the same genes. Abundance data have been quantile-normalized to improve comparison of relative expression between the two populations.  
(C) Heatmaps showing (top) log2 fold change and (bottom) quantile-normalized abundance, as in (B). Comparisons are made between brain pericytes, brain endothelial cells, and B cells (from PBMCs), and genes are ordered by log2 fold change between pericytes and B cells.  
See also Figure S7 and Table S1.

syngeneic study recapitulated the pattern of BBB permeability observed in the NSG model, suggesting that the presence of murine CD19<sup>+</sup> B cells in the syngeneic model did not affect the specific disruption of the BBB observed only in the murine-targeting conditions (Figures S6E and S6F). We also measured BBB integrity 4 days after CAR-T cell transfusion in NSG mice by measuring gadodiamide uptake using 9.4 tesla magnetic resonance imaging. This showed an increase in uptake in the mice that had received murine targeting but not human targeting, CD19 CAR-T cells (Figures S6G and S6H). Finally, we dissociated brains from treated mice and performed flow cytometry to measure CD19<sup>+</sup> pericyte depletion following CAR infusion. Mice receiving mCD1928z, but not hCD1928z, showed a decrease in both CD45<sup>+</sup>CD19<sup>+</sup> B cells as well as CD45<sup>-</sup>CD13<sup>+</sup>CD19<sup>+</sup> mural cells (Figure S6I). We did not specifically analyze any behavioral changes, but no overt behavioral phenotype was observed.

In summary, these experiments suggest that despite a lower frequency of CD19 expression in mouse mural cells, compared to human mural cells, administration of CD19-directed CAR-T cells can cause BBB leakiness and pericyte depletion in mice lacking B cells. It is possible that some of this effect may be mediated by targeting of microglia, or other CD19-expressing cell types not yet identified. Microglia were previously demonstrated to be depleted following CD19-directed CAR-T cell treatment in mice (Pennell et al., 2018) and also display low expression of *Cd19* in our scRNA-seq data, as well as the Tabula Muris database (Tabula Muris Consortium, 2018). Therefore, these experiments demonstrate that, while some aspects of CAR-T function and toxicity may be measured in preclinical mouse models, there are important limitations, since species-specific differences in cell type and transcriptional state may not perfectly match human-specific pathophysiology. Indeed, initial CAR-T cell studies in mouse models did not predict the degree of neurotoxicity that was later observed in human clinical trials (Giavridis et al., 2018; Ruella and June, 2018).

### Human Brain Mural-Specific Expression of CD19

Since pericytes are present in multiple organs, we performed a comparative analysis of brain pericytes with pericytes and vSMCs from the lung, a highly vascularized tissue with high numbers of pericytes and endothelial cells (Travaglini et al., 2020). Although all mural cell populations showed shared expression of a core transcriptional identity, such as *PDGFRB*, *RGS5*, *FOXS1*, and *KCNJ8*, we identified numerous transcriptional differences between brain and lung pericytes (Figures 7A and S7A). Notably, *CD19* is specifically expressed in brain, but not lung mural cells. Only 6/2,724 lung mural cells had any detectable counts for *CD19*, showing no apparent enrichment over nonspecific counts found in all non-B cell clusters. This organ specificity may be explained by broader differences in lineage-specific transcription factors between the two cell types, which agrees with the distinct developmental origins of brain pericytes from neural crest cells (Etchevers et al., 2001). Namely, brain pericytes express *BCL11A*, and lowly express *PAX5*, two key developmental factors in brain development but also B cell development (Kozmik et al., 1992; Urbánek et al., 1997; Liu et al., 2003; Dias et al., 2016). Interestingly, *EBF1*, another tran-

scription factor important for B cell development, was expressed by both populations. The expression of these B cell factors in brain mural cells may explain the specific expression of *CD19* expression in these cells.

Finally, we asked whether genes predicted to localize to the membrane or cell surface were differentially expressed in the NVU relative to lung pericytes as well as B cells and could be used to improve the safety of CD19-directed CAR-T (Figures 7B and 7C). For example, recent studies have shown the utility of “AND” logic gates to improve the specificity of CAR-T cells by requiring recognition of two distinct antigens (Roybal et al., 2016). We identified the most highly differentially expressed genes among those annotated as being secreted or located on the cell surface, in brain versus lung pericytes, as well as B cells versus brain pericytes and endothelial cells (Bausch-Fluck et al., 2015; Zheng et al., 2017). These could improve the specificity of B cell-directed CAR-T cells by requiring dual recognition of two antigens. This analysis revealed additional genes, such as *CD74*, *HLA-DRA*, and *LTP*, which are both highly expressed and highly enriched on B cells relative to brain pericytes (Figure 7C). An alternative approach to improving CAR-T cell specificity is to use “NOT” gates, where inhibitory CAR constructs recognize non-specific target proteins and prevent T cell activation (Fedorov et al., 2013). We identified numerous genes that are enriched in pericytes over B cells, such as *BGN*, *FN1*, and *SEMA5A* (Figure 7C).

### DISCUSSION

CD19-directed immunotherapies, such as CAR-T cells and BiTEs, have been highly effective in patients with B cell malignancies, achieving clinical remission in up to 90% of patients in some indications (June et al., 2018). However, neurotoxicity remains a major adverse effect experienced by these patients, which in some cases can lead to cerebral edema and death. Here, we describe a possible on-target mechanism for CD19 CAR-T cell-mediated neurotoxicity, which may be caused by the previously unrecognized expression of *CD19* in mural cells in the human brain. We show that *CD19* expression is present across multiple independent datasets and present in pericytes as well as vSMCs. This finding was confirmed by IHC in adult tissue and by an analysis of a large database of RNA-seq data from adult samples from different ages and brain regions. Importantly, the *FMC63* epitope that is recognized by clinical-grade CD19 CAR-T cells and BiTEs. Surprisingly, we found that *CD19* expression in mural cells is significantly higher in human than in mouse, highlighting the importance of studying these adverse effects in the context of human tissues. It is important to state that we are not advocating for the discontinuation of CD19-directed therapy use in cancer patients, particularly since acute neurotoxic adverse effects can be effectively managed in most cases, and CD19-directed immunotherapies are tremendously effective treatments for B cell cancers (Neelapu, 2019). However, these findings do provide a strong impetus for further investigation of mural cell targeting in the context of human clinical trials, as well as potential strategies to develop the next generation of therapies with improved safety profiles.

Previous studies in patients who received CD19 CAR-T therapy and had neurologic adverse reactions demonstrated edema, multifocal hemorrhage, and vascular disruption in the brain, and neuropathologic evaluation of the brain from patients who developed fatal CRS and/or neurotoxicity revealed CAR-T cell infiltration (Gust et al., 2017; Torre et al., 2018). These findings are consistent with a role of mural cells in mediating neurotoxicity; namely, CD19 CAR-T cell-mediated depletion of pericytes and vSMCs may lead to endothelial activation, disruption of tight junctions, increased BBB permeability, and cerebral edema. However, a caveat of the present study is that we have not determined the precise contributions or synergistic effects of mural cell death and CRS in this process. For example, vascular permeability changes related to CRS might increase permeability across the BBB and increase or enable CAR-T cell access to CD19-expressing mural cells. Along these lines, the CAR signaling motif also appears to play an important role in CAR-T cell-induced neurotoxicity; fatal cases of cerebral edemas have occurred predominantly in patients receiving CAR-T cells with the CD28 signaling domain (Table S1; Gardner et al., 2017; Gofshteyn et al., 2018; Lee et al., 2015; Locke et al., 2017; Maude et al., 2018; Park et al., 2018; Juno Therapeutics a Subsidiary of Celgene, 2020; Turtle et al., 2016a; Turtle et al., 2016b; Turtle et al., 2017). This may be due to (1) higher levels of cytokine release and CRS due to stronger antigen-receptor signaling (Ying et al., 2019), (2) increased sensitivity to low CD19 antigen density (Majzner et al., 2020), (3) higher cytolytic activity, or (4) a combination of these factors.

In addition to CAR-T cell-intrinsic effects, the clinical manifestation of severe neurotoxicity is likely a complex multifactorial process, where the lymphodepleting or chemotherapy regimen, scFv specificity (e.g., SJ25C1 of JCAR015 versus FMC63 of tisagenlecleucel and axicabtagene ciloleucel), and co-localization of on- and off-target cells may all play a contributing role. For example, these findings may caution the use of intrathecal or intraventricular administration of anti-CD19 CAR-T cells for refractory primary CNS high-grade lymphoma where CAR-T cell activation could be triggered not only by the tumor itself but also by neighboring brain mural cells. Similarly, patient populations that are already at risk for encephalopathy after chemotherapy or whole-brain radiotherapy may also be particularly vulnerable to neurotoxicity (Korfel and Schlegel, 2013). Future studies should investigate the connection between neurotoxicity associated with CD19 CAR-T cells and with chemotherapy agents such as fludarabine (Lowe et al., 2018), as well as any potential effects that systemic inflammation observed during CRS has on CD19 expression in mural cells.

There are several important limitations to our study. First, we have not demonstrated that mural cell CD19 expression is the cause of clinical neurotoxicity, and further studies are needed to determine whether CD19 CAR-T cells target and deplete human mural cells in patients, and whether this effect is independent or correlated with CRS. Second, it will be important to investigate whether there are inter-patient differences in mural cell frequency and/or CD19 expression in mural cells (for example, age related) that could lead to differences in the incidence of neurotoxicity. Third, the IHC data suggest some degree

of heterogeneity in CD19 expression, as not all vessels were stained positive, and as mentioned above, inter-patient differences in CD19 mural cell expression may contribute to differences in patient outcome. Neurovascular cell heterogeneity in humans remains poorly understood and will be important to study further (Faal et al., 2019; Uemura et al., 2020). Finally, while we have focused our discussion on the immediate clinical effects of mural cell depletion, there may also be long-term effects associated with targeting of these cells, which will require further studies in longitudinal patient cohorts.

The scRNA-seq data that we have presented here were obtained from prenatal human samples. We focused primarily on human data due to observed differences in CD19 expression between human and mouse (La Manno et al., 2016). Current data of the human brain are largely biased toward early or elderly time points, and there is a vast gulf of information missing on pediatric and middle-aged individuals. Thus, while our analysis clearly shows CD19<sup>+</sup> mural cells in prenatal brain samples, and the analysis of bulk RNA-seq data show that CD19<sup>+</sup> expression persists throughout adulthood, future studies are needed to fully interrogate whether CD19 expression changes as a function of age or region. The majority of publicly available adult human scRNA-seq data were sort-enriched for neuronal or other specific populations and therefore lack abundant mural cell populations to conclusively analyze. Two recently published adult datasets do show a small mural cell population and include CD19<sup>+</sup> mural cells (Hodge et al., 2019; Kim et al., 2020), but further datasets are needed due to the low numbers of mural cells and/or sparse expression of mural cell marker genes in existing adult data. Whether the comparatively low numbers of mural cells in adult datasets are due to technical challenges, such as sampling frequencies and/or loss of cells during dissociation, or biological differences associated with aging, and how this relates to the levels of CD19 expression observed in prenatal samples, is an important question for future study. Nonetheless, there is clear evidence that CD19 expression is present in brain mural cells in at least some time points and cases, which has thus far been overlooked. We hope that this finding will inform future studies in human samples, including in patients experiencing neurotoxicity.

More broadly, this research highlights one of many utilities for developing a comprehensive human single-cell atlas for clinical medicine (Regev et al., 2017). scRNA-seq is an unbiased, genome-wide measurement of gene expression that can capture even rare populations of cells. These rare cell types might otherwise be missed in measurements of bulk tissue due to their low frequency, but, as we now demonstrate, could be critically important in the clinical effects of targeted therapy. While current CAR-T cells recognize only a single antigen, future generations of CAR-T cells may be able to discriminate between unique combinations of target antigens to improve their cell-type specificity. This may be particularly useful as CAR-T cells are increasingly deployed against solid tumors, which share most antigens with normal tissues. We envision that a comprehensive database of gene expression across all human cell types will enable the precise identification of cell-type-specific target antigens that can be used to design safe and effective cellular immunotherapies.

We hope that the analytical framework of comparing on-target and potential off-target cell types aids the future development of immunotherapies.

## STAR★METHODS

Detailed methods are provided in the online version of this paper and include the following:

- KEY RESOURCES TABLE
- RESOURCE AVAILABILITY

- Lead Contact
- Materials Availability
- Data and Code Availability

- EXPERIMENTAL MODEL AND SUBJECT DETAILS

- Cell lines and cultures
- *In vitro* T cell transduction and cultures
- Mouse models

- METHOD DETAILS

- Processing of human brain single-cell RNA-sequencing data and identification of CD19+ mural cells
- Generation of human brain meta-cells from Kriegstein/BICCN data
- Analysis of vSMC versus pericyte identity
- Analysis of Brainspan bulk RNA-sequencing data and integration of correlated gene sets with single-cell data
- Processing of human lung single-cell RNA-sequencing data and comparison of gene expression with human brain mural cells
- Comparison of gene expression profiles of human brain mural cells with peripheral blood B cells
- Generation of mouse single-cell RNA-sequencing data
- Analysis of mouse single-cell RNA-sequencing data
- Flow cytometry of mouse brain
- Cytotoxicity assays
- BBB permeability assays
- Immunohistochemistry of adult human brain
- High resolution mouse brain MRI

- QUANTIFICATION AND STATISTICAL ANALYSIS

## SUPPLEMENTAL INFORMATION

Supplemental Information can be found online at <https://doi.org/10.1016/j.cell.2020.08.022>.

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## AUTHOR CONTRIBUTIONS

K.R.P., D.M., C.H.J., H.Y.C., A.D.P., and A.T.S. conceived the study. K.R.P., K.E.Y., and A.T.S. performed RNA sequencing experiments and analysis. K.R.P., D.M., A.T.S., T.J.M., and V.G.B. performed and reviewed human IHC analysis. D.M. and F.L. performed T cell manufacturing. D.M. performed *in vitro* T cell experiments. D.M., E.P., P.B., M.H., F.L., K.G., and J.S. performed animal experiments. E.P. analyzed mouse flow cytometry data. P.B., M.H., and N.E.W. performed animal imaging experiments. T.J.M., R.R., S.M., I.M., C.H.J., H.Y.C., A.D.P., and A.T.S. oversaw and guided experiments and analysis. K.R.P., D.M., A.D.P., and A.T.S. wrote the manuscript, and all authors reviewed and provided comments on the manuscript.

## DECLARATION OF INTERESTS

C.H.J. and A.D.P. report intellectual property licensed to Novartis and Tmunity Therapeutics related to CAR-T cells. K.R.P., A.T.S., H.Y.C., D.M., A.D.P., and C.H.J. are listed as inventors on patent applications filed by Stanford University and the University of Pennsylvania. K.R.P. is a consultant for Maze Therapeutics. A.T.S. is a scientific founder of Immunai and receives research funding from Arsenal Biosciences. H.Y.C. is a co-founder of Accent Therapeutics and Boundless Bio and an advisor to 10x Genomics, Arsenal Biosciences, and Spring Discovery.

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## STAR★METHODS

### KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
CD19 (IHC)	Abnova	Cat# MAB9625; RRID:AB_10754868
CD19 (mouse)	Biolegend	Cat# 115541; RRID:AB_11204087
CD45 (mouse)	Biolegend	Cat# 103154; RRID:AB_2572116
CD31 (mouse)	Biolegend	Cat# 102420; RRID:AB_10613644
B220 (mouse)	Biolegend	Cat# 103258; RRID:AB_2564053
CD31 (mouse)	BD Biosciences	Cat# 561410; RRID:AB_10612003
CD41 (mouse)	Biolegend	Cat# 133914; RRID:AB_11125581
CD19 (mouse)	BD Biosciences	Cat# 562701; RRID:AB_2737731
CD19 isotype control	BD Biosciences	Cat# 562602; RRID:AB_11153860
7-ADD	Biolegened	Cat# 420404
Deposited Data		
Human brain single-cell RNA-seq data (prefrontal cortex)	Zhong et al., 2018	GEO: GSE104276
Human brain single-cell RNA-seq data (ventral midbrain)	La Manno et al., 2016	GEO: GSE76381
Human brain single-cell RNA-seq data (forebrain)	La Manno et al., 2018	SRA: SRP129388; <a href="https://panglaodb.se/view_data.php?sra=SRA646572&amp;srs=SRS2833947">https://panglaodb.se/view_data.php?sra=SRA646572&amp;srs=SRS2833947</a> ; <a href="https://panglaodb.se/view_data.php?sra=SRA646572&amp;srs=SRS2833946">https://panglaodb.se/view_data.php?sra=SRA646572&amp;srs=SRS2833946</a>
Human brain single-cell RNA-seq data (BICCN/Kriegstein)	NeMO	RRID:SCR_017266; <a href="http://data.nemoarchive.org/biccn/grant/devhu/kriegstein/transcriptome/scell/processed/counts/">http://data.nemoarchive.org/biccn/grant/devhu/kriegstein/transcriptome/scell/processed/counts/</a>
Human brain <b>bulk RNA-seq data</b> (Brainspan)	Allen Brain Atlas	RRID:SCR_008083; <a href="https://www.brainspan.org/static/download.html">https://www.brainspan.org/static/download.html</a>
Human <b>peripheral blood mononuclear cell</b> single-cell RNA-seq data	10x Genomics 3' Single Cell Gene Expression 3.0	<a href="https://support.10xgenomics.com/single-cell-gene-expression/datasets/3.0.0/pbmc_10k_v3">https://support.10xgenomics.com/single-cell-gene-expression/datasets/3.0.0/pbmc_10k_v3</a>
Human lung single-cell RNA-seq data (epithelial, endothelial, stromal, immune cell)	Travaglini et al., 2020	Synapse: syn21041850
Mouse whole dissociated brain single-cell RNA-seq data	This paper	GEO: GSE153647
Experimental Models: Cell Lines		
A20	University of Pennsylvania, Dr. Johg Lee	RRID: CVCL_1940
Nalm6	University of Pennsylvania, Dr. Johg Lee	RRID: CVCL_0092
Experimental Models: Organisms/Strains		
NOD/scid/IL2rg	University of Pennsylvania Stem Cell and Xenograft Core	RRID:IMSR_JAX:005557
Software and Algorithms		
Seurat v3.1.4	<a href="https://satijalab.org/seurat/">https://satijalab.org/seurat/</a>	RRID:SCR_007322
Scanpy v1.4.3	<a href="https://scanpy.readthedocs.io/en/stable/">https://scanpy.readthedocs.io/en/stable/</a>	RRID:SCR_018139

## RESOURCE AVAILABILITY

### Lead Contact

Further information and requests for resources, data, and details of analysis should be directed to and will be fulfilled by the Lead Contact, Ansuman T. Satpathy ([satpathy@stanford.edu](mailto:satpathy@stanford.edu)).

### Materials Availability

The anti-hCD19-BBz CAR was constructed as described previously ([Milone et al., 2009](#)). The mCD19-BBz and mCD19-28z CARs were constructed by ligating the mCD19 scFv (1D3) into the CAR backbone sequences of pTRPE –BBz and pTRPE –28z. Third generation lentiviral vectors were produced as previously described ([Milone et al., 2009](#)).

### Data and Code Availability

The accession number for the sequence data for mouse single-cell RNA-sequencing reported in this paper is GEO: GSE153647. Details of analysis are provided below in the [STAR Methods](#), and any further questions should be directed to and will be fulfilled by the Lead Contact.

Processed sequencing data (gene counts per cell) were downloaded as follows: [Zhong et al., 2018](#) (GEO GSE104276), [La Manno et al., 2016](#) (GEO GSE76381), [La Manno et al., 2018](#) (PanglaoDB database, Karolinska Institute, <https://panglaodb.se/samples.html>), Kriegstein/BICCN (NeMO archive, <https://nemoarchive.org/>), Brainspan (Allen Institute, <https://www.brainspan.org/static/download.html>). PBMC data was downloaded from the 10x Genomics reference datasets available on their website ([https://support.10xgenomics.com/single-cell-gene-expression/datasets/3.0.0/pbmc\\_10k\\_v3](https://support.10xgenomics.com/single-cell-gene-expression/datasets/3.0.0/pbmc_10k_v3)). UMI counts for human lung ([Travaglini et al., 2020](#)) were downloaded from Synapse, accession syn21041850.

## EXPERIMENTAL MODEL AND SUBJECT DETAILS

### Cell lines and cultures

The murine A-20 B cell leukemia cell line and human Nalm6 line were kindly provided by Dr. Jong Lee (University of Pennsylvania). Nalm6 and A20 tumor cell lines were cultured in RPMI-1640 media supplemented with 10% FBS, penicillin (100 U/mL; GIBCO), streptomycin sulfate (100 mg/mL; GIBCO), and 10 mmol/L HEPES (GIBCO) (R10 media). 50 µM of 2-mercaptoethanol was added in the culture media of A20 cell line.

### In vitro T cell transduction and cultures

Human CAR-T cells were produced from normal donor T cells provided by the University of Pennsylvania Human Immunology Core, as previously described ([Milone et al., 2009](#)). Cells were transduced with lentiviral vector encoding anti-human (h) or murine (m) CD19 scFv fused to CAR backbones containing either human 4-1BB or human CD28 and CD3zeta (CD247) signaling domains, as described previously, and were expanded ex vivo for 11 days ([Milone et al., 2009](#)). Two cell expansions were produced from 2 different healthy human donors. The transduction efficiencies ranged from 20 to 40%.

### Mouse models

All animal studies adhered to the NIH Guide for the Care and Use of Laboratory Animals and in accordance with protocols approved by The University of Pennsylvania Institutional Animal Care and Use Committee. NOD/scid/IL2rg (NSG) mice were obtained from the University of Pennsylvania Stem Cell and Xenograft Core and were housed in the vivarium at the University of Pennsylvania under BSL2 and pathogen-free conditions.  $1 \times 10^7$  T cells with 20% normalized CAR+ expression were administered to non-tumor bearing mice by one intravenous tail vein injection in 100 µL of PBS. Adult mice of both sexes were used for experiments. For other experiments not using NSG mice, adult healthy C57BL/6J mice of both sexes were obtained from Jackson Laboratory and housed in a BSL1 mouse facility in accordance with University protocols.

## METHOD DETAILS

### Processing of human brain single-cell RNA-sequencing data and identification of CD19+ mural cells

Sequence data were downloaded as described in Data and Code Availability. For data shown in [Figures 1 and 2](#), samples were processed using Seurat ([Butler et al., 2018](#)) version 2.3.4. For data shown in [Figures 4, 5, 6, and 7](#), samples were processed using Scanpy version 1.4.3 ([Wolf et al., 2018](#)) or Seurat version 3.1.4.

To process data from Zhong et al., 2018, La Manno et al., 2016, and La Manno et al., 2018, cells with fewer than 500 detected genes or UMI counts were excluded, and cell counts were normalized per cell using log(depth-normalized counts + 1). The ~1500–2500 most variable genes were used for clustering based on the variance to mean ratio. As the datasets include both post-mitotic and actively cycling cells, the cell cycle status was computed using the CellCycleScoring function and subsequently regressed out using the ScaleData function in Seurat. Principle component analysis was performed using the genes identified as highly variable for each dataset, and the top ~25–50 principle components were used for subsequent dimensionality reduction using the UMAP algorithm.

Clusters were called using the FindClusters function in Seurat, and marker genes for each cluster were identified using the FindMarkers function. Clusters were subsequently manually annotated by comparing highly enriched genes to known cell-type markers. For the analysis of Zhong et al., 2018 data, neuronal precursor cells, erythroid cells, and neuronal cells were identified and excluded, and the remaining cells were subsequently re-clustered. Gene expression data shown in Figures 1G and 2C represents mean expression across all cells with a given cluster label. Percentages and cell counts of a given label (e.g., pericytes) represent the total number of cells in a given cluster, not necessarily only those cells positive for an individual marker, unless otherwise indicated.

### Generation of human brain meta-cells from Kriegstein/BICCN data

To identify meta-cells in the data from the Kriegstein lab/BICCN data, the processed UMI counts were first analyzed using Scrublet on a per-sample basis to identify potential doublets. The following parameters were used: expected\_doublet\_ratio = 0.08, sim\_doublet\_ratio = 2, min\_counts = 3, min\_cells = 3, min\_gene\_variability\_pct1 = 75, n\_prin\_comps = 30. A consistent threshold for doublet calling was set at 0.35 across all samples. The results of this were then saved and imported alongside the UMI counts into Seurat. Samples were all processed individually with the same pipeline, using the same parameters and clustering resolution. As a result, some samples may be under-clustered, as the clustering resolution was intentionally set to err on the side of merging similar cells rather than creating false positive clusters. Only cells with > 250 and < 4000 features were included, as well as less than 20% mitochondrial reads. A cell cycle difference score was calculated following the Seurat vignette. The SCTransform workflow was used, and percent mitochondrial reads, number of genes detected, and cell cycle difference were included as variables to regress out. 30 components were used for UMAP dimensionality reduction, with a random seed of 1, and n.neighbors = 100. 30 components were used for FindNeighbors, and a resolution of 0.2 was used for FindClusters. Subsequently, predicted doublets were excluded, and the average depth-normalized expression of each cluster was calculated using the AverageExpression function. This information, representing the meta-cell transcriptome, alongside the individual cell metadata (e.g., cell barcode, cluster ID, UMAP coordinates, etc.), and the meta-cell metadata (e.g., number of cells per meta-cell, mean number of genes per cell for each meta-cell, etc.) was saved for each sample. A representative example of this output is shown in Figure S2A.

This information was then imported into Python and merged, to create a metacell x genes matrix, with each value representing mean(depth-normalized UMI counts). Variable genes were identified by fitting an exponential function to the log(mean) versus log(-coefficient of variation) for each gene across meta-cells. The top 6000 genes, representing ~20%, based on distance from the fit line, were used for PCA. Data was log transformed, then z-normalized, prior to PCA. The scikit-learn decomposition.PCA() function was used with the svd\_solver = 'arpack'. Then, the top 30 PCs were used as input to the UMAP reducer in the umap-learn package with the following parameters: n\_neighbors = 80, n\_components = 2, min\_dist = 0.25, transform\_seed = 50, n\_epochs = 500. Last, the top 30 PCs were used as input to KMeans clustering in the scikit-learn package, with k = 9.

### Analysis of vSMC versus pericyte identity

To generate the integrated dataset shown in Figures 5A–5C and S3A–S3C, the data from Zhong et al., 2018; La Manno et al., 2016, and La Manno et al., 2018 was first processed using Scrublet to identify doublets with the following parameters: expected\_doublet\_rate = 0.06, min\_counts = 2, min\_cells = 3, min\_gene\_variability\_pct1 = 85, n\_prin\_comps = 30. A threshold of 0.25 was used for doublet calling. The doublet information along with expression information was then imported into Seurat. Individual Seurat objects were created, using only the union of genes annotated in all three datasets (to account for small differences in upstream sample processing), and only including cells with at least 500 features detected. Cell cycle scoring was performed as described above, and each sample was processed using SCTransform with vars.to.regress = nFeature\_RNA and percent.mt. 3000 features for integration were then chosen using the SelectIntegrationFeatures function, followed by PrepSCTIntegration, FindIntegrationAnchors, and IntegrateData using default parameters. For RunUMAP, the following parameters were used: dims = 1:30, seed.use = 20, min.dist = 0.3, spread = 1. 50 PCs were used as input for FindNeighbors and resolution = 1 was used for FindClusters. This output is shown in Figure S3A.

The subset of cells that were in clusters expressing pericyte/vSMC genes (*RGS5*, *CSPG4*, *FOXF2*, *ACTA2*), endothelial genes (*PECAM1*) or microglial genes (*CSF1R*) were included for downstream analysis. Predicted doublets were not included. These samples were re-processed using the same parameters for RunUMAP, FindNeighbors, and FindClusters as above, with the exception of dims = 1:30 for FindNeighbors and resolution = 1 for FindClusters.

To generate the dataset shown in Figures 5D–5H and S3D–S3F, meta-cells from cluster ID = 3, 4, 6, or 9 were included. The ~92K single cells that comprised these meta-cells were identified and imported into Scanpy. These were first processed, excluding cells with fewer than 500 genes, genes detected in fewer than 3 cells, and cells with more than 10% mitochondrial reads. Doublets had already been excluded in the generation of meta-cells, and so all cells analyzed here were not predicted doublets. Variable genes were identified with min\_mean = 0.0125, max\_mean = 3, min\_dist = 0.5. n\_counts was regressed out, and max\_value = 10 was used for pp.scale(). 50 PCs were calculated and used for the pp.neighbors() calculation; default parameters were used for tl.umap() and resolution = 0.65 was used for tl.leiden(). This output is shown in Figure S3D.

Then, clusters identified as pericytes/vSMCs/endothelial cells, as well as the subset of progenitor (early time points) cells that were SOX2 low and showed scattered expression of pericyte markers were analyzed. This comprised ~26K cells. These samples were re-processed using the same workflow, unless noted here: maximum percent\_mito = 8%, resolution = 0.25 for tl.leiden().

### Analysis of Brainspan bulk RNA-sequencing data and integration of correlated gene sets with single-cell data

RPKM data processed at the gene- and exon-level was downloaded from the Allen Institute portal. For correlation analysis, spearman correlation was used. The top 200 genes by spearman correlation were analyzed using the PANTHER web tool.

PBMC data was integrated with the data from the Zhong et al., 2018; La Manno et al., 2016, and La Manno et al., 2018 datasets. The BICCN data was not included due to the less clear separation between pericytes and endothelial cells observed in this data, presumably due to incomplete dissociation of the tightly interacting cell types leading to a high rate of false-positive endothelial transcripts being present in pericyte cells and vice-versa. These samples were merged and processed using Scanpy using join = outer. Cells with greater than 20% mitochondrial genes were excluded. To calculate a gene score, the top 30 genes identified based on spearman correlation in the Brainspan data were used as inputs to the tl.score\_genes() function in Scanpy, with the following parameters: ctrl\_size = 50, random\_state = 1, n\_bins = 10. The ttest\_ind() function in scipy.stats was used to calculate p values for gene scores between the different clusters.

### Processing of human lung single-cell RNA-sequencing data and comparison of gene expression with human brain mural cells

Lung data was first processed using Seurat to identify clusters with high expression of pericyte and endothelial marker genes. These clusters, which included ~19K cells, were then merged with the brain data from Zhong 2018, La Manno 2016, and La Manno 2018. The BICCN data was excluded for the same reason as above (less clear separation of mural and endothelial cells). These cells were then processed together to allow identification of the vSMC, pericyte, and endothelial cell clusters in the lung data. The annotations of pericyte/endothelial cell identity were carried over from the above merged analysis of the three datasets described in the analysis of vSMC versus pericyte identity of these datasets (Figures 5A–5C). Diffxpy was used to perform differential gene expression testing between the different pairwise comparisons of interest (e.g., brain versus lung pericytes). Due to the vast difference in the number of lung pericytes and endothelial cells versus brain cells, the lung data was sub-sampled to match the same number of cells in the respective brain cluster being analyzed in order to avoid confounding effects due to differences in cell number (i.e., the lung data had greater power to detect lowly-expressed genes).

### Comparison of gene expression profiles of human brain mural cells with peripheral blood B cells

Single-cell gene expression data from B cells was generated from the PBMC data described in Data and Code Availability. A database of extracellular proteins based on mass spectrometry data was used for the analysis of putative cell-surface proteins (Bausch-Fluck et al., 2015). This database includes proteins that may be secreted and thus not strongly enriched at the cell surface, but in the interest of not excluding proteins that might be at the cell surface, this more comprehensive database was used. Gene expression was computed as the mean across all single B cells. To account for differences in the expected distribution of mean gene expression due to the inherent sparsity of single-cell RNA expression data, which is biased based on cell capture and library preparation technique, as well as sequencing depth, the distribution of gene expression values were quantile normalized using the normalize.quantiles function in R to facilitate comparisons across cell types. As such, the absolute gene expression values are pseudo-arbitrary, in that they have been transformed to be relatively comparable across different cell types from distinct single-cell sequencing techniques.

### Generation of mouse single-cell RNA-sequencing data

Four mouse brains were dissected and dissociated following an established protocol (Lin and Monje, 2018). Briefly, whole brain was minced using a razor blade, and all tissue (not following the note to remove vessels as in the protocol) was pooled and transferred into a tube filled with an enzymatic dissociation solution containing 250 µg/mL STEMzyme I (Worthington #LS004107). This was incubated at 37 degrees for 30 minutes, then triturated, and filtered using a sucrose gradient to remove myelin debris. Red blood cell lysis was performed with ACK lysis buffer (Gibco #A1049201). The resulting cells were then resuspended in FACS buffer (HBSS without Ca/Mg, supplemented with 1% BSA), blocked at RT for 10 minutes using BioLegend TruStain FcX (#101320), then stained for 20 minutes on ice with the following antibodies: CD19 (BioLegend 115512, diluted 1:100), CD13 (BD 558745, diluted 1:20), CD31 (BioLegend 102421, diluted 1:50), and CD45 (BioLegend 103113, diluted 1:100). CD19+, CD31+, and CD13+ cells were enriched by flow sorting prior to capture using the 10x Genomics 5' scRNA kit following the manufacturer protocol. Libraries were sequenced using a HiSeq 2500.

### Analysis of mouse single-cell RNA-sequencing data

Data was processed using CellRanger v3.0.2 using the prebuilt mm10 index v3.0.0. The resulting files were then processed using Scrublet to remove suspected doublets with the following settings: min\_counts = 2, min\_cells = 3, min\_gene\_variability\_pct1 = 85, n\_prin\_comps = 30, threshold = 0.25. Doublets were then excluded, and the remaining data was then processed using Scanpy, excluding cells with less than 250 genes or more than 10% mitochondrial reads. Variable genes were selected with the following parameters: min\_mean = 0.0125, max\_mean = 3, min\_disp = 0.5. 30 PCs were used for pp.neighbors(), and data was clustered using tl.leiden() with resolution = 0.15.

### Flow cytometry of mouse brain

For the mouse CNS neurovascular fraction characterization experiments, the following antibodies against mouse molecules were used: anti-mouse CD19 (clone 6D5 0.1mg/ml, BV650, Biolegend, dilution 1 to 200), CD45 (clone 30-F11, APC-Fire750 Biolegend, dilution 1 to 100), CD31 (clone 390, PerCP-Cy5.5, Biolegend, dilution 1 to 400), B220 (clone RA3-6B2, PE/Dazzle 594 Biolegend, dilution 1 to 200), Zombie Aqua Fixable Viability kit (Biolegend #423102, dilution 1 to 400). Isolation and flow cytometric analysis of CNS perivascular cells was performed as previously described (Crouch and Doetsch, 2018) with modifications to digestion volumes to accommodate whole murine brain rather than microregions. Antibody staining was done using CD13-PE (BD Biosciences, cat. No. 558745), CD31-PECy7 (BD Biosciences, clone 390, cat no. 561410), CD41-APC (Biolegend, cat. no. 133914), CD45-APC (Biolegend, clone 30-F11, cat no. 103112), CD19-BV421 (BD Biosciences, clone 1D3, cat. no. 562701), and BV421 Rat IgG2a κ isotype control (BD Biosciences, cat. no. 562602). Dead cells were excluded with 7-AAD (Biolegend, #420404). Fluorescence minus one (FMO) controls were used for each marker to define gates for each population. Species-matched isotype controls were performed for each primary antibody to ensure the absence of non-specific binding. Samples were acquired on a BD 5-laser LSRFortessa or 4-laser BD Aria II. Note that epitope cleavage of CD13 and CD19 and loss of cell viability were observed with harsher digestion protocols using papain or higher concentrations of collagenase/dispase. An important challenge for these perivascular cell suspension preparations is the cell stress during dissociation due to trituration and digestion by enzymes, both known to impact cell viability and cause the low cell numbers we observed in our experiments. The dissociation itself is also known to cause significant cell surface marker epitope cleavage. We observed that CD13 and CD19 expression was present but variable according to the enzyme used (papain, collagenase, or hyaluronidase) for digestion.

Human T cell viability was determined by staining with LIVE/DEAD Fixable Violet Dead Cell Stain Kit (Invitrogen), followed by surface antibody staining. CAR expression was detected using biotinylated protein L (Genscript) followed by R-phycoerythrin-conjugated streptavidin (BD cat. #554061). All experiments were acquired on a BD LSR Fortessa flow cytometer (Becton Dickinson) and data was analyzed with FlowJo v10.

### Cytotoxicity assays

All cytotoxicity assays were flow cytometry-based and adapted from the Quah et al. protocol (Quah et al., 2007). Target cells were stained with CFSE (Thermo Fisher) and T cell effectors were stained with Cell Trace Far Red (Thermo Fisher) prior to co-culture. 24 hours after co-culture, cells were harvested, stained for viability with LIVE/DEAD Fixable Violet Dead Cell Stain Kit and analyzed by flow cytometry.

### BBB permeability assays

Evans Blue dye (EBD) infusion was performed as previously described (Radu and Chernoff, 2013). Seven days after CAR-T cell infusion, mice were injected intravenously with EBD and/or mannitol. Thirty minutes after EBD infusion, mice were euthanized; brain tissues were excised and fixed in 10% formalin solution for at least 24 h. After washing and dehydrating, brains were embedded in paraffin and cross-sectioned at a thickness of 5 $\mu$ m. Sections were deparaffinized, mounted with DAPI-containing Vectashield, and analyzed by fluorescence microscopy.

C57BL/6J mice were obtained from Jackson Laboratory and housed in a BSL1 mouse facility at the University of Pennsylvania. Murine splenocytes from donor mice were harvested, activated with murine-specific anti-CD3/anti-CD28 magnetic beads (Thermo Fisher), and transduced with retroviral vector encoding mCD19mBBz CAR, mCD19mCD28z CAR, or hCD19mBBz CAR transgenes. Four days after transduction,  $5 \times 10^6$  murine CAR-T cells were injected intravenously in recipient mice  $\pm$  50mg/kg cyclophosphamide for lymphodepletion.

### Immunohistochemistry of adult human brain

Human anti-CD19 antibody (clone BT51E) was used to stain normal human brain. Automated immunohistochemistry was performed with Ventana Benchmark XT following a clinically validated protocol for CD19. 4-micron FFPE tissue sections were deparaffinized and rehydrated. Antigen retrieval was performed using Standard Cell Condition 1 (pH 8.5) for 60 minutes (Ventana Medical Systems). Slides were incubated with anti-human CD19 (1:50, monoclonal, Abnova BT51E; or Leica #NCL-L-CD19-163, clone BT51E) at 37° for 30 minutes. The ultraView Universal DAB Detection system (Ventana) was used with 3,3'diaminobenzidine chromogen.

### High resolution mouse brain MRI

$3 \times 10^7$  T cells with 32% normalized CAR+ expression were counted using the ADAM-CellIT (NanoEnTek, Seoul, Korea) and the Multi-sizer 4e (Beckman Coulter). T cells were administered to non-tumor bearing NSG mice ( $n = 4$  per group) by one intravenous tail vein injection in 100 $\mu$ L of PBS. All the MRI experiments were performed 4 days post CAR-T cell infusions. Mice were anesthetized by using isoflurane maintained at 1%–1.5% in 1 L/min air. Lateral tail vein was cannulated for administration of gadodiamide (287 mg/ml) or mannitol (25% w/v). Mice were placed in a 20 mm diameter commercial quadrature proton coil (m2m Imaging Corp., Cleveland, OH) and the probe was transferred into a 9.4T horizontal bore small animal MRI scanner (Bruker, Billerica, MA). During the MRI scans, animal body temperature was maintained at 37°C with the air generated and blowing through a heater (SA Instruments, Inc., Stony Brook, NY). Respiration and body temperature were continuously monitored using an MRI compatible small animal monitoring system (SA Instruments, Inc., Stony Brook, NY).

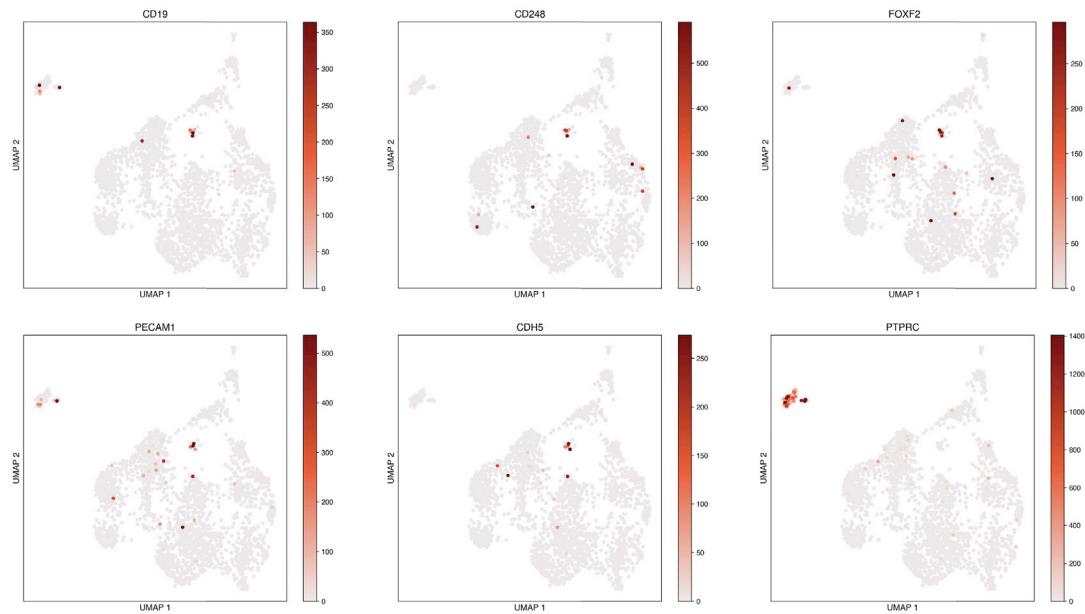
T1-mapping was performed pre- and post-gadodiamide injection on 0.7 mm thick coronal slice in the mid-brain using saturation recovery protocol with following parameters: number of averages = 2, field of view =  $20 \times 20 \text{ mm}^2$ , matrix size =  $192 \times 192$ , echo time = 8 ms, repetition times = 200, 500, 800, 1200, 1500, 4000 and 9000 ms, scan time 30 minutes. Following the baseline T1 acquisition, gadodiamide was administered via tail vein as described previously (Ku et al., 2018; van Vliet et al., 2014). T1 acquisition scans were run consecutively after one another. Pre- and post-gadodiamide administration T1 maps were generated using the image sequence analysis tool in Paravision 6.0.1 by exponential fitting of the signal recovery. An in-house MATLAB script was used to quantify the  $\Delta T1$  ( $T1_{\text{Post}} - T1_{\text{Pre}}$ ) caused by gadodiamide leakage in the brain.

#### QUANTIFICATION AND STATISTICAL ANALYSIS

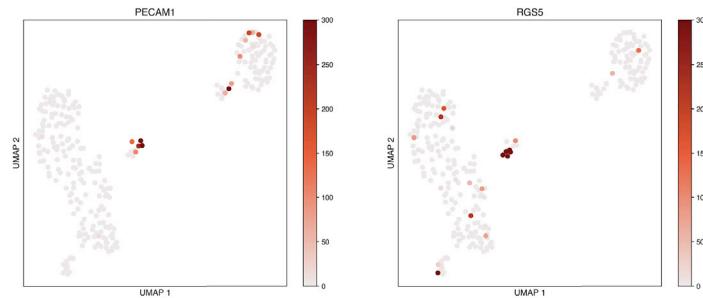
Information on specific statistical tests used is provided in the figure legends and/or [Method Details](#). Quantification of MRI  $\Delta T1$  was performed using a MATLAB script. For single-cell RNA-sequencing analysis, specific details regarding pre-processing steps and parameters are provided in the relevant section of the [Method Details](#).

# Supplemental Figures

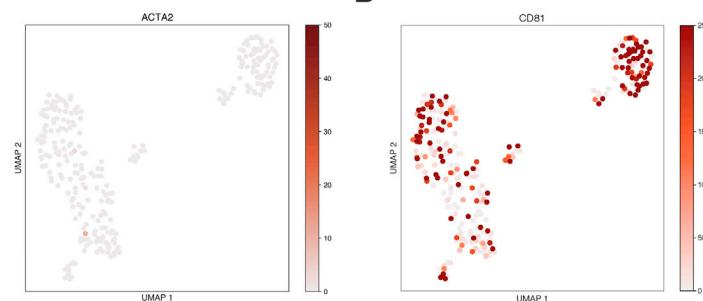
**A**



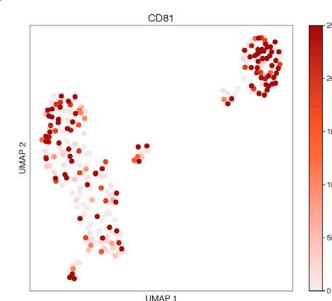
**B**



**C**

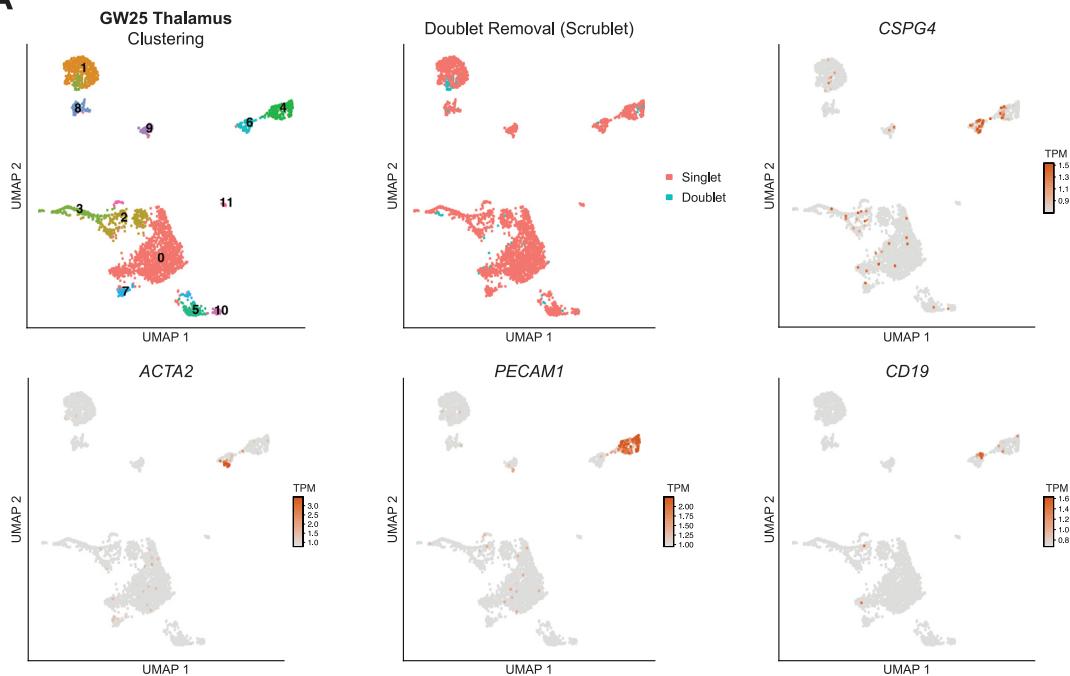
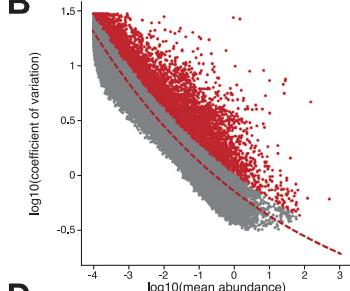
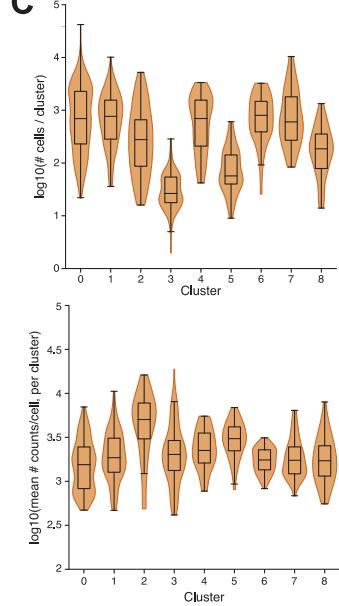
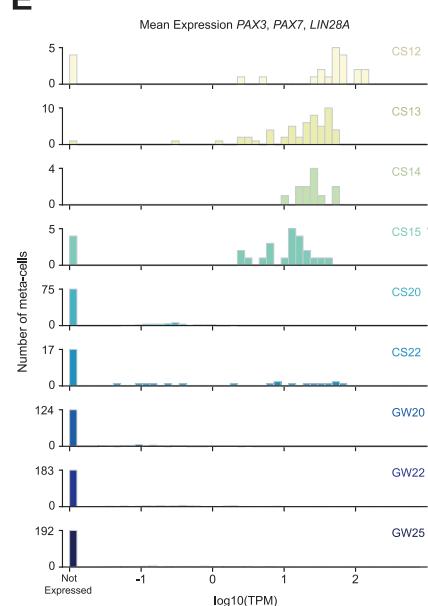
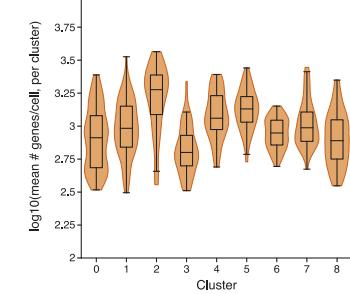
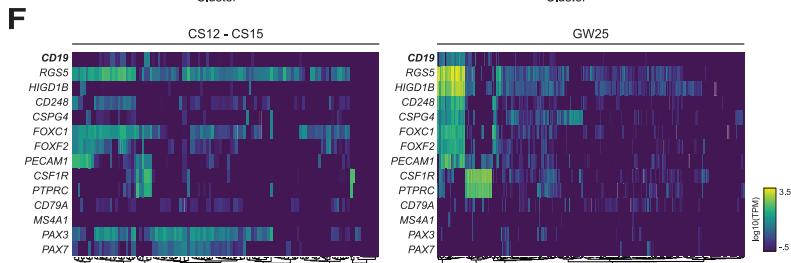


**D**



**Figure S1. Identification of Non-neuronal Clusters of Interest, Related to Figure 1**

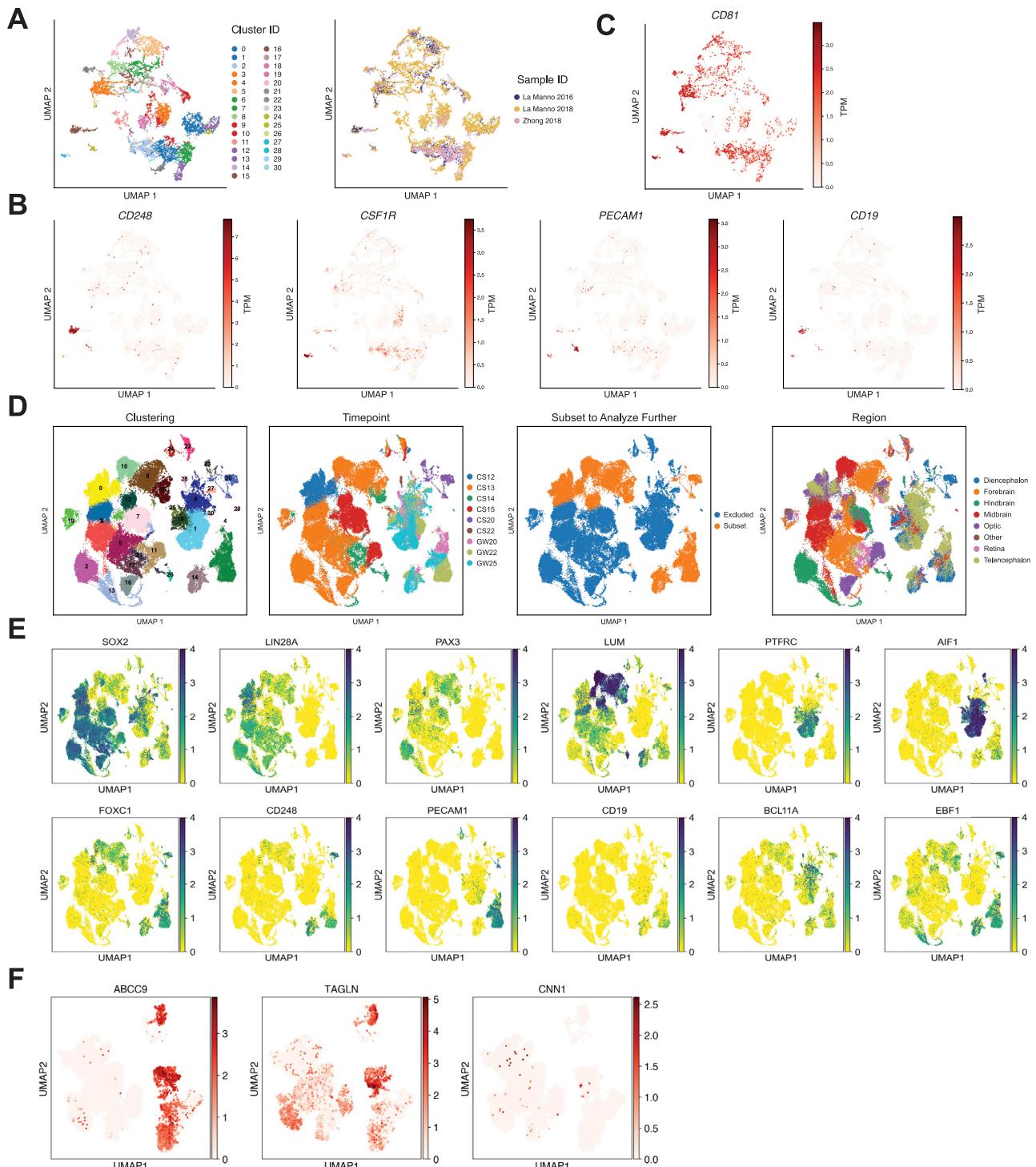
(A) UMAP plots showing expression of selected marker genes for mural cells, endothelial cells, and microglia/immune cells. The color represents counts per million of each gene for a given cell in all panels (a-d). (B) Separate clusters of endothelial cells (*CD31/PECAM1*+) and mural cells (*RGS5*+ ) within the central cluster of neurovascular cells. (C) Lack of *ACTA2* expression indicating pericyte identity. (D) *CD81* is expressed in mural cells.

**A****B****C****E****D****F**

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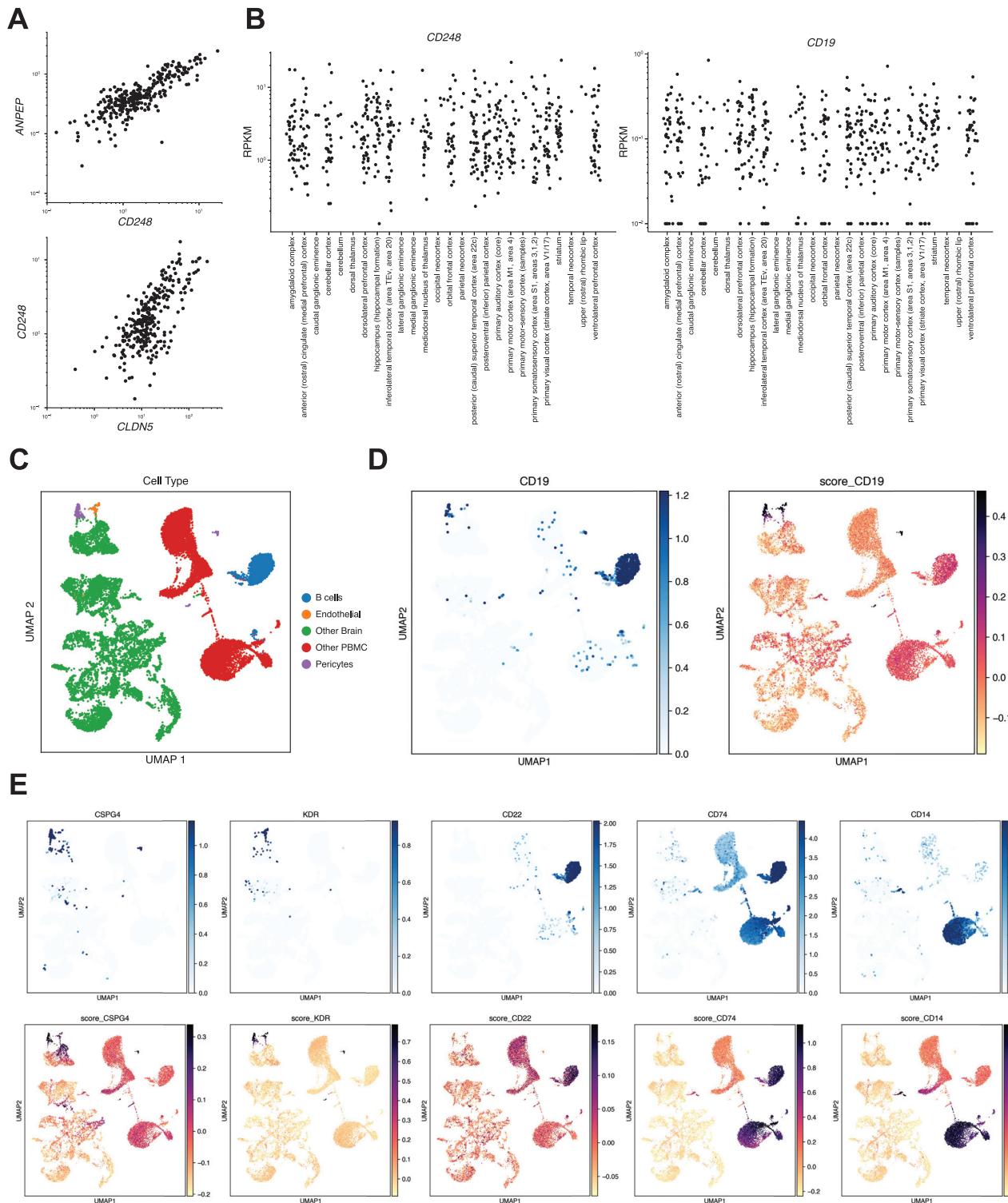
**Figure S2. Meta-Cell Clustering of BICCN Data, Related to Figure 4**

(A) Example processing of one sample (GW25 Thalamus). Clustering, doublet identification, as well as expression of selected marker genes is shown. Note that vSMCs, pericytes, and endothelial cells are clustered closely together, but are clearly distinguished by marker gene expression. (B) Identification of variable genes across meta-cells. Genes used for downstream analysis are shown in red. (C) Number of cells per metacell, grouped by cluster. (D) Mean number of genes (left) or counts (right) per cell in each metacell, grouped by cluster. (E) Mean expression of *PAX3*, *PAX7*, and *LIN28A* per meta-cell, showing the high expression at early time points and progressive loss throughout development. (F) Heatmap of selected marker genes for each meta-cell in the indicated ranges of sample ages, showing the absence of *CD19* and mural cell gene expression at CS12-15 but the presence at GW25.



**Figure S3. Identification of Sub-clusters of Interest across Brain Datasets, Related to Figure 5**

(A) Integrated clustering of Zhong 2018, La Manno 2016, and La Manno 2018. (B) Expression of *CD248* (pericyte), *CSF1R* (microglia), *PECAM1* (endothelial), and *CD19* in these datasets. (C) *CD81* is expressed by *CD19*+ mural cells. (D) Analysis of cells comprising progenitors, vascular cells, and microglia from BICCN data. The cells that were included for subsequent analysis are indicated in the third panel. (E) Expression of selected marker genes that were used to identify clusters of cells. Cells are colored by TPM of indicated gene. (F) Expression of vascular SMC marker genes distinguishing venous, arterial, and arteriole SMCs. Cells are colored by TPM of indicated gene.



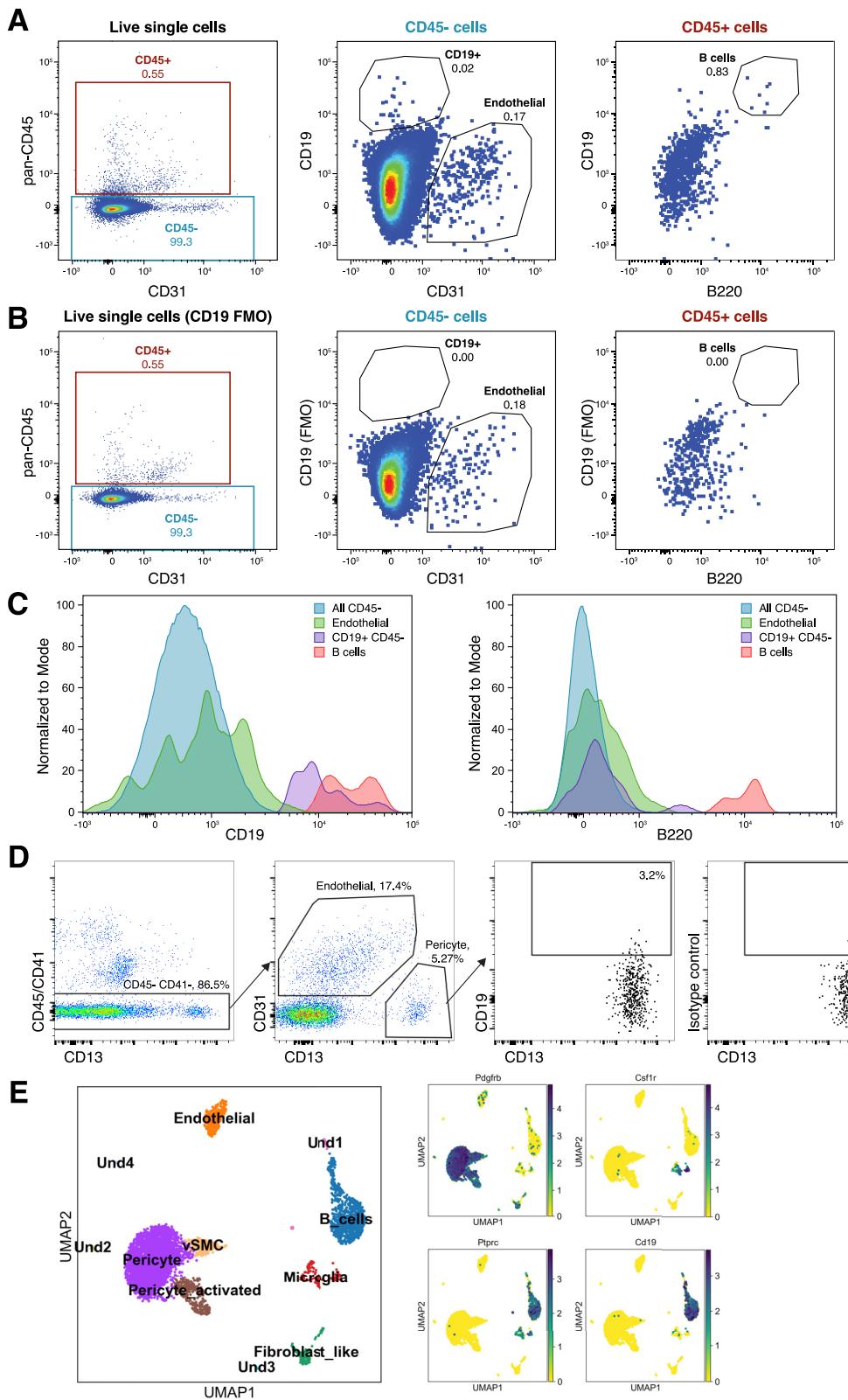
**Figure S4. Analysis of Allen Institute Brainspan Data, Related to Figure 6**

(A) (top) Scatterplots showing correlation of CD248 and ANPEP (CD13), two pericyte markers and (bottom) CD248 and CLDN5, an endothelial cell marker gene, showing overall correlation of neurovascular-associated genes. (B) Expression of CD19 and CD248 by brain region. (C) Integrated clustering of brain and PBMC

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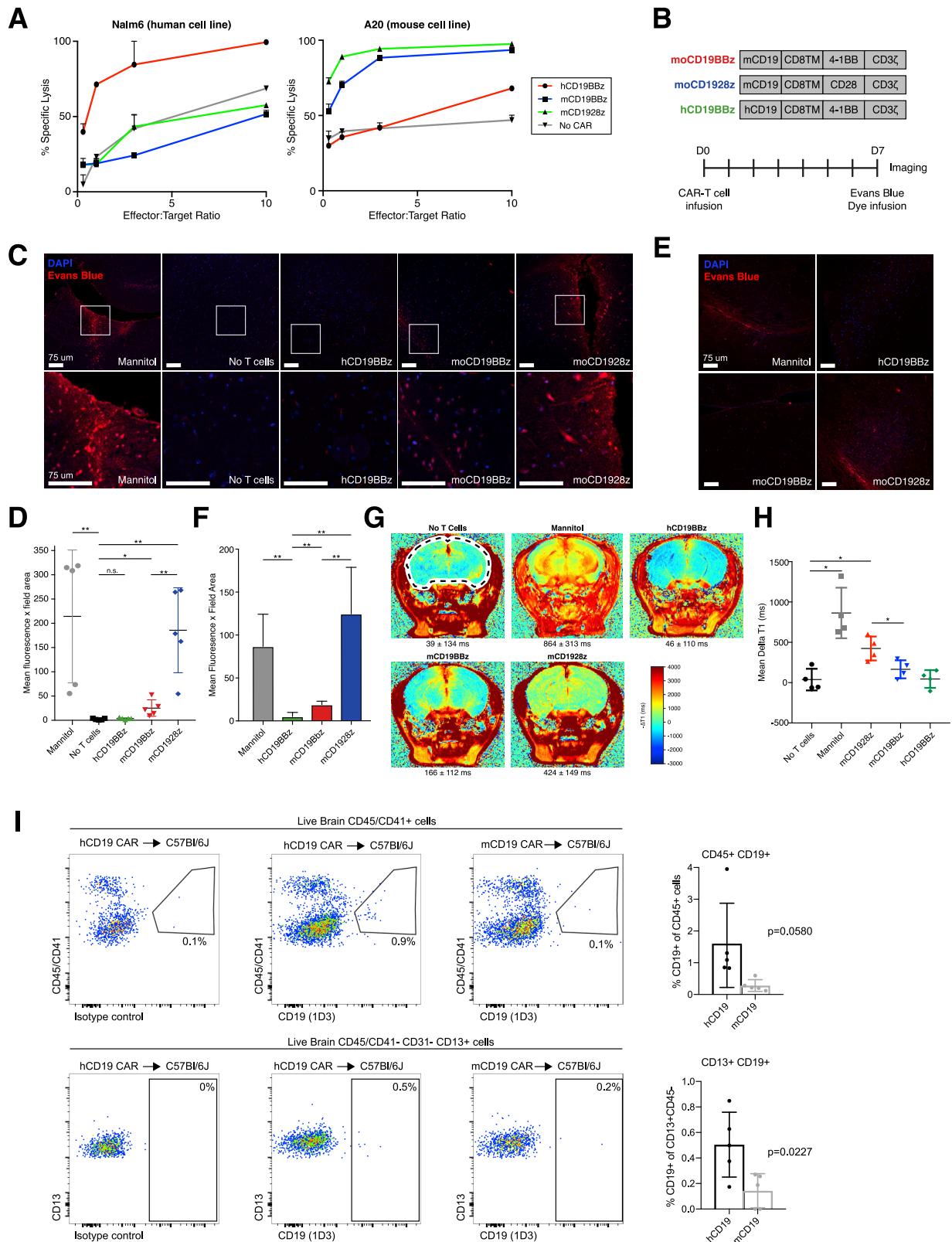
data, with B cell, pericyte, and endothelial cell clusters identified. (D) Comparison of *CD19* expression and *CD19*-correlated gene score, showing high score expression in pericytes and endothelial cells, and smaller enrichment in B cell cluster, indicating that pericytes/vasculature is the primary driver of *CD19* expression in the brain. Expression for genes is in counts per 10,000; unit for gene score is arbitrary unit. (E) Comparisons of additional genes/correlated gene scores, as controls, showing ability of the approach to selectively enrich for expected populations of cells. Units are as in (d).



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**Figure S5. Analysis of CD19 Expression in Mouse Brain, Related to Figure 6**

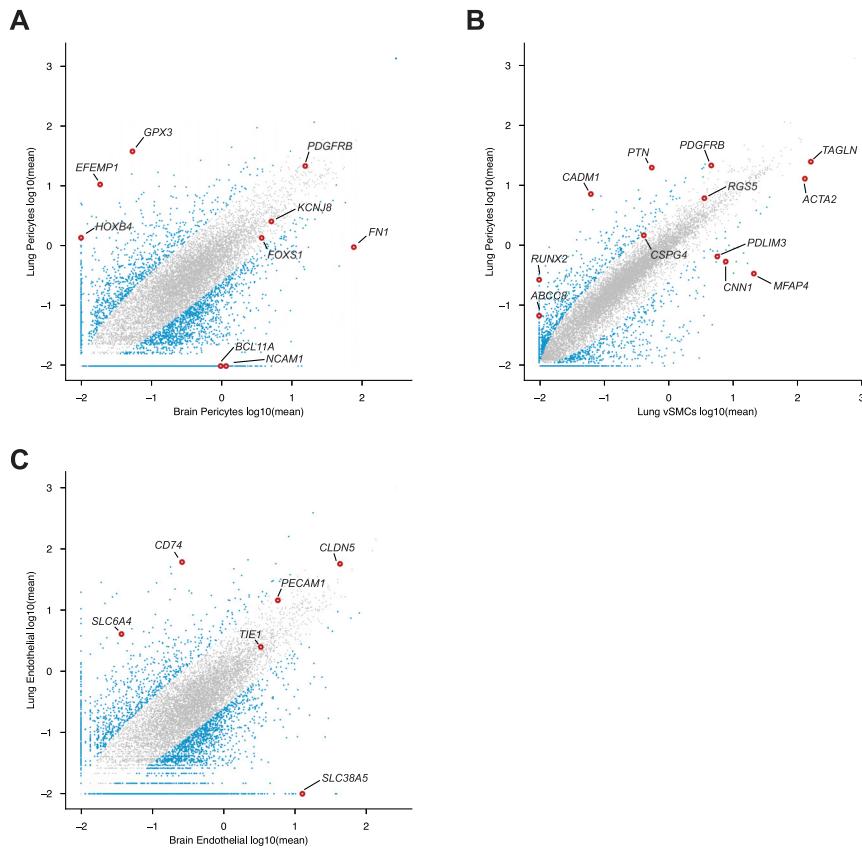
(A) Flow cytometric analysis of CNS neurovascular cells isolated from healthy C57BL/6J mice. The live cell fraction is shown with gating for the CD45+ and CD45- populations. CD19+ cells are present in the CD45- population as shown, along with CD31+ endothelial cells. CD19+ B220+ B cells are found in the CD45+ fraction. (B) Fluorescence minus one (FMO) control staining for CD19, with the same gating as in (a). (C) (left) CD19 is expressed at roughly similar levels in the CD45- fraction as in the CD45+ fraction (B cells); (right) B220 is expressed only by the CD45+ CD19+ B cell fraction, but not in the CD45- CD19+ fraction. (D) Flow cytometry analysis of live, single cells, showing gating for CD45- CD41- CD13+ CD31- pericytes, and expression of CD19 against CD13. An isotype control for CD19 is shown. Data in (d) represents a separate experiment from (a-c). (E) scRNA-seq of dissociated mouse brain after enrichment for CD19+, CD31+, and CD19+ cells by FACS. (left) UMAP plot showing identified clusters, (right) expression of marker genes showing low CD19 expression in pericytes and moderate CD19 expression in microglia.



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**Figure S6. Infusion of CD19-Directed CAR-T Cells into Mice, Related to Figure 6**

(A) Flow cytometry based killing assays in human B-ALL cell line Nalm6 and murine CD19+ A20 B-ALL cell line in presence of human 4-1BB or CD28 based anti-human or murine CD19 CAR-T cells. (B) Representation of CAR constructs and animal study design using NSG mice. (C) Representative immunofluorescence imaging from the cortex showing nuclear staining (DAPI, blue) and Evans Blue dye (EBD) extravasation (red) for BBB disruption in NSG mice following murine-targeting CD19 CAR-T treatment ( $n = 5$  mice per treatment). Scale bar, 75  $\mu$ m. (D) Quantification of EBD staining in (c) in each group of mice as determined by the product of the mean fluorescence intensity per field area. Representative sections ( $n = 3$  per mouse) were analyzed and graphed. \* $p < 0.05$ , \*\* $p < 0.005$ , 2 tailed unpaired t test. (E) Syngeneic C57BL/6J mouse model recapitulates the murine CD19-specific neurotoxicity pattern observed in the immunodeficient model, showing staining for DAPI and EBD in cortical sections ( $n = 5$  mice per group). Scale bar, 75  $\mu$ m. (F) Quantification of EBD staining for each group shown in (e). \* $p < 0.05$ , \*\* $p < 0.005$ , 2 tailed unpaired t test. (G) *In vivo* imaging using high-resolution MRI with gadodiamide contrast shows BBB leakage after murine-targeting CD19 CAR-T cell therapy. All experiments were performed 4 days after CAR-T cell infusion. Anatomical reference image with selected region of interest (ROI) contour used to calculate delta T1 and T2 between each group is outlined.  $n = 4$  mice per group, mean and standard deviation across mice provided. (H) Quantification of mean delta T1 of MRI within ROI shown in (g),  $n = 4$  mice per group. \* $p < 0.05$ , two-tailed t test. (I) (top) Decrease in CD45+CD19+ cells ( $n = 5$  mice per group) in mCD19-specific CAR-T treated mice as opposed to the hCD19 CAR-T group. (bottom) CD19 versus CD13 ( $n = 5$  mice per group) showing decrease in live brain CD45-CD41-CD31-CD13+CD19+ cells in the mCD19 specific CAR-T as opposed to the hCD19 CAR-T treated group. Error bars show the SEM, p value using two-tailed t test.



**Figure S7. Comparison of Mural Cell Transcriptomes from Brain and Lung, Related to Figure 7**

(A) Scatterplot comparing gene expression in brain pericytes and lung pericytes. Genes with at least an absolute log<sub>2</sub> fold change of 2 are colored in blue, and selected genes are annotated, showing similar expression of markers such as *PDGFRB*, but distinct expression of other genes such as *BCL11A* and *HOXB4*. (B) as in (a) but comparing lung vSMCs with lung pericytes. (C) as in (a) but comparing brain endothelial cells and lung endothelial cells.