

Supplementary Materials for
Single-cell multi-omics sequencing uncovers region-specific plasticity of glioblastoma for complementary therapeutic targeting

Xin Wang *et al.*

Corresponding author: Liang Chen, chenliang8@genomics.cn; Baohui Liu, bliu666@whu.edu.cn;
Ying Gu, guying@genomics.cn

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Materials and Methods

Patient sample collection

The tumor core and paired peritumoral brain tissues were collected from patients with GBM during the surgery performed in the Department of Neurosurgery, Renmin Hospital of Wuhan University, China. The detailed information of each patient was provided in table S1. To ensure the high quality for downstream library construction, collected tissues were placed in liquid nitrogen within 15 minutes after resection. GBM was preliminarily diagnosed by computed tomography (CT) or MRI, and later confirmed independently by three neuropathologists in the Department of Pathology at Renmin Hospital of Wuhan University via the post-operative histopathological examination. Particularly, the mutational status of isocitrate dehydrogenase (*IDH*) for each tumor sample were assessed via immunohistochemistry and/or DNA sequencing.

Hematoxylin-eosin (HE) staining

By following the standard protocol of HE staining, tissue sections were stained in the hematoxylin solution for 5 minutes and then washed by running water at room temperature for 5 minutes. Subsequently, the tissue slides were dyed with the eosin solution for 2 minutes, and were then gradually dehydrated by treatment of an increasing concentration of ethanol (75%, 95%, and 100%). Finally, the slices of HE staining were visualized using a slide scanning system (Axio Scan.Z1, Zeiss, Germany).

Preparation of single cell nucleus suspension

Single nuclei were prepared as previously described (115). All the following processes were carried out on the ice except for other description. Briefly, tissue was thawed and cut into small pieces in a culture dish. Then, those small pieces were transferred to a 2 mL Dounce tissue grinder (Sigma) with 2 mL of homogenization buffer containing 500 mM sucrose (Sigma), 20 mM Tris pH 8.0 (Thermo Fisher Scientific), 10 mM MgCl₂ (Thermo Fisher Scientific), 50 mM KCl (Thermo Fisher Scientific), 1× protease inhibitor cocktail (Roche), 0.1% NP-40 (Roche), 0.1 mM DTT, and 1% nuclease-free BSA. Five minutes later, tissue was first homogenized by 15-20 strokes of the dounce pestle A (loose), and then the 15-20 strokes of the dounce pestle B. The homogenate was subsequently collected and filtered into a 15 mL centrifuge tube through a 70 μM cell strainer (BD) and a 30 μM cell strainer (SYSMEX PARTEC). Then, the filtered homogenate was centrifuging at 500 g for 5 minutes at 4°C. The supernatant was discarded and the pellets were resuspended and counted with DAPI.

Construction and sequencing of snRNA-seq library

snRNA-seq library was prepared with the kit of the DNBelab C Series Single-Cell Library Prep Set (MGI) as described previously (34). Briefly, snRNA-seq libraries with the barcode were constructed from the single-cell suspensions through following steps, including encapsulation of droplet, emulsion breakage, collection of mRNA captured bead, reverse transcription-polymerase chain reaction (RT-PCR), amplification and purification of cDNA. The protocol of manufacturer was also applied to construct indexed sequencing libraries. Then, Qubit ssDNA Assay Kit (Thermo Fisher Scientific) was used to assess the quality of the sequencing libraries before they were sequenced by the machine of DIPSEQ T1 sequencer (BGI). Paired reads were sequenced. Read 1

contained 10 bp unique molecular identifier (UMI), 10 bp cell barcode 1, and 10 bp cell barcode 2. Read 2 contained 10 bp sample index and 100 bp transcript sequence.

Construction and sequencing of snATAC-seq library

snATAC-seq library construction was performed as described by Yu et al (116). Briefly, the kit of DNBelab C Series Single-Cell ATAC Library Prep Set was used to construct single-nucleus ATAC-seq library. snATAC-seq libraries with the barcode were converted from transposed single-nucleus suspensions, through steps as follows: encapsulation of droplet, pre-amplification polymerase chain reaction, beads emulsion breakage, beads collection, DNA amplification, and purification. Indexed sequencing libraries were prepared according to the protocol. Qubit ssDNA Assay Kit (Thermo Fisher Scientific) was used to assess the quality of sequencing libraries before sequencing, and a paired-end 50 sequencing scheme was used by DIPSEQ T1 sequencer (BGI).

Analysis of snRNA-seq data

snRNA-seq data preprocessing

The raw reads after sequencing were parsed and cell barcodes were corrected by PISA software (version 0.7) (117). Then the clean reads were aligned to GRCh38 (hg38) genome by STAR software (version 2.7.9a) (118), and the obtained SAM files were transformed to BAM format files. After sorting of BAM files, empty droplets were filtered by R package DropletUtils (119), and the resulted cell expression matrixes were generated by PISA.

Then, Seurat (version 4.0.5) (120) and DoubletFinder (version 2.0.3) (121) were applied in the following analysis. First, the Seurat objects were generated based on the above matrixes and high-quality cells were acquired through three criteria: 1) genes expressed in more than 5 cells and cells with detected genes more than 500; 2) cells with mitochondrial genes expression proportion less than 5%; 3) cells without extreme outliers of the number of detected genes. Next, doublets were filtered out by DoubletFinder according to the threshold of 5%. After the above preprocessing, matrixes from different samples were merged by Seurat for cross-sample comparison in the following analysis. Particularly, the merged count matrixes were log-normalized by 'NormalizeData' function and scaled by 'ScaleData' function. Then, the 3,000 most variable genes were identified by 'FindVariableFeatures' function and principal components (PCs) were calculated by 'RunPCA' function. Nearest neighbors were computed based on the top 30 PCs, and then 19 clusters were identified by 'FindClusters' function (resolution = 0.6). Finally, uniform manifold approximation and projection (122) implemented in Seurat was used to embedding cells onto two-dimensional graphic.

Cell type annotation

Annotation for each cluster was manually performed based on the expression of canonical marker genes. The marker genes used were *EGFR*, *PTPRZ1* (GBM cell); *TPD52L1*, *ALDOC*, *CLDN10*, *ETNPPL* (Astrocyte); *PLP1*, *MOG*, *MOBP*, *CLDN11* (Oligodendrocyte); *GAD2*, *GAD1*, *SLC32A1* (Inhibitory neuron); *SLC17A7*, *NRGN*, *NEUROD2*, *NEUROD6* (Excitatory neuron); *APBB1IP*, *CD74*, *CIQA* (Myeloid); *CD2*, *CD3D*, *CD3G* (T cell); *MCAM*, *CLDN5* (Pericyte/endothelial cell).

Differential expressed genes analysis

DEGs for each cell type were identified by ‘FindAllMarkers’ function in Seurat, the expression of one cell type was compared with the rest cell types. Wilcoxon statistical test was employed. Particularly, $\text{avg_log2FC} > 0.5$ and $p \text{ value} < 0.05$ were used as the criteria to filter the DEGs.

CNVs and phylogenetic tree analyses

To identify malignant cells from the samples, the CNVs of each cell were estimated with the inferCNV package (version 1.8.1) (28). To obtain accurate CNV results accurately, we implemented the inference process for every patient separately and used their normal cells (microglia and oligodendrocytes) from corresponding peritumoral brain as reference cells. In this analysis, the parameters used include: $\text{cutoff} = 0.1$, $\text{cluster_by_groups} = \text{T}$, $\text{analysis_mode} = \text{‘subclusters’}$, $\text{HMM} = \text{T}$, $\text{tumor_subcluster_partition_method} = \text{‘random_trees’}$. After CNVs were obtained, we employed k -means clustering to group cells based on the CNV pattern and visualized the CNV results on heatmap generated by the ComplexHeatmap package (version 2.11.1) (123). To guarantee the accuracy for identifying malignant cells, only cells containing chr7 amplification and chr10 deletion simultaneously were judged as malignant cells.

Based on the CNV location on chromosome, each p- or q-arm level change was converted to equivalent CNV by consulting the information of the genomic cytoband, and each CNV was further annotated as gain or loss. Finally, the evolutionary tree was constructed by the arm level of CNV and subclones containing identical ones were collapsed. UPhyloplot (version 2.3) (<https://github.com/harbourlab/uphyloplot2>) (47) was used for data visualization.

GO enrichment analysis

The clusterProfiler (version 4.0.5) (124) R package was applied to obtain the enriched GO terms (biological processes). The enriched terms were filtered by setting $\text{qvalueCutoff} = 0.05$, and were then visualized by ggplot2 (version 3.3.5) (<https://github.com/tidyverse/ggplot2>) R package.

Prediction of the transcription factor regulatory network

The pyScenic (versions 0.11.2) (62) software was used to identify the transcription factor regulatory networks. The GRCh38 version of reference files were used, and default parameters were applied. This prediction process consists of three steps: 1) ‘pyscenic grn’, which identifies TF and its target genes, named a regulon, based only on the expression profile of the cell; 2) ‘pyscenic ctx’, which simplifies the regulon and removes the target genes without the motif corresponding to the TF. 3) ‘pyscenic aucell’, the Area Under the recovery Curve (AUC) of the genes was measured to define the regulons based on the enrichment of the previously identified regulons.

Subpopulation analysis of GBM cells

After annotation for cell types, GBM cells were extracted and then subdivided into sample-specific objects by ‘SplitObject’ function in Seurat. For the GBM cell objects, the ‘SelectIntegrationFeatures’ function was used to select the genes that are repeatedly variable across datasets, and then anchors among the objects were called for integration by ‘FindIntegrationAnchors’ function. Subsequently, the different GBM cell datasets were integrated by ‘IntegrateData’ function and subclustered by “FindClusters” function.

Pseudo-time trajectory analysis

To construct the pseudo-time trajectory for GBM cells, the Monocle package (version 2.18.0) (82) was applied. The highly dispersion genes among the extracted cells were identified to arrange the

cells on two-dimensional plot. After the trajectory was constructed, genes changing with the trajectory were identified by ‘differentialGeneTest’ function and the significantly altered genes (q value $< 1e-4$) were selected for downstream analyses. The changes of selected genes along different branches were analyzed by “BEAM” function, and patterns of these genes were further visualized by “plot_genes_branched_heatmap” function.

RNA velocity analysis

RNA velocity (transcriptional dynamics of splicing kinetics) of selected cells was used to deduce the transition status of cells. First, we used velocity (version 0.17.17) (125) to obtain the ratio of its spliced- and unspliced- messenger RNA (mRNA). Then, scVelo (version v0.2.4) (83) was used to analyze the state transition. In details, velocities were calculated by setting the parameter: mode = ‘deterministic’, and the obtained velocities are vectors in gene expression space, representing the direction and speed of movement of the individual cells. Then, the velocities were projected onto the embedding graphic.

Analyses of snATAC-seq data

snATAC-seq data preprocessing

The raw paired-end sequencing data of snATAC-seq were parsed and adapter sequence were trimmed by PISA, and the processed data were then aligned to GRCh38 (hg38) genome by bwa (126). Bap2 (<https://github.com/caleblareau/bap>) was used to deconvolute the beads, beads from the same droplet were combined according to their index. After the alignment and combination, fragment files were obtained.

ArchR (version 1.0.1) (127) was used to perform the following analysis with the obtained fragment files. First, for each sample, low-quality cells were filtered out. Particularly, the parameter minTSS was set as 4, minFrag as 1000, and maxFrag as 100000. To remove the doublets from the samples, the built-in function ‘addDoubletScores’ was applied and the parameter filterRatio was set as 2. After the data were preprocessed, ‘addIterativeLSI’ was performed for dimension reduction, and the key parameters were: resolution = 0.2, sampleCells = 10000, varFeatures = 25000, dimsToUse = 1:30. We did ‘addClusters’ function to cluster cells and ‘addUMAP’ function was used for embedding.

Cell annotation and peak analysis

To annotate the cell clusters in snATAC-seq, we considered the results from three strategies: 1) the chromatin accessibility around the canonical marker genes mentioned in above section; 2) the gene activity score inferred from the openness of the marker genes and the corresponding promoter region; 3) the cell types predicted from the integration with the gene expression matrix of snRNA-seq.

Based on ArchR’s built-in functions, peak identification, motif annotation, peak-to-gene analysis, and co-accessibility analysis were accomplished by ‘getMarkerFeatures’, ‘addMotifAnnotations’, ‘addPeak2GeneLinks’, and ‘getCoAccessibility’ function respectively. In peak-to-gene analysis, the key parameters include: maxDist = 250000, scaleTo = 10^4 , log2Norm = TRUE, and predictionCutoff = 0.4, which are specified in ‘addPeak2GeneLinks’. Similarly, the parameters corCutoff = 0.45, FDRCutOff = $1e-04$, varCutOffATAC = 0.25, and varCutOffRNA = 0.25, pertaining to peak-to-gene analysis, are set within the ‘plotPeak2GeneHeatmap’ function.

Identification of malignant cells

In snATAC-seq, the R package Copy-scAT(35) was employed to identify malignant cells. The input was the fragment format file. The procedure includes three steps: 1) fragment pileup and normalization; 2) chromosome arm CNV analysis; 3) detection of amplifications. Particularly, only barcodes with a minimum number of 5000 fragments were retained, and a pileup of total coverage (number of reads \times read lengths) over bins of 1 Mbp was generated. To account for differences in read depth, binned read counts were linearly normalized over the total signal in each cell, and chromosomal bins that consist predominantly of zero were filtered out for further analysis.

Function enrichment of cis-regulatory elements

GREAT (40) was applied to perform the functional enrichment analysis for cis-regulatory regions identified in snATAC-seq. Particularly, differentially accessible distal peaks for each cell type were supplied to GREAT and the curated GO terms (biological process) were chosen for enrichment with the binomial FDR value. The resulted terms were visualized by heatmap generated using scanpy(128).

Assignment for GBM cells to cellular states

The cellular states of GBM cell were identified by the expression score for four state modules (MES-like, NPC-like, AC-like, and OPC-like) according to the method described in previous study(22). In detail, the signatures of each cellular state were obtained from Neftel et al. (22). Particularly, signatures for MES1-like and MES2-like were grouped into MES-like. Similarly, the NPC1-like and NPC2-like were grouped into NPC-like. The average expression of the obtained signatures for each GBM cell were calculated for assigning the cellular states. The state results were visualized using ggplot2, with each position representing their relative expression score and each quadrant corresponds to specific cellular states.

Cell lines and cell culture

Human glioblastoma cell lines U251, and mouse glioma cell line GL261 were purchased from the Cell bank of Chinese Academy of Science (Shanghai, China). Cells were cultured in DMEM supplemented with 10% fetal bovine serum ([Gibco](#), USA) and 1% penicillin-streptomycin and were kept in a humidified 5% CO₂ incubator at 37°C.

Cell transfection

Plasmids of pLVX-CMV-fluc-puro, pMD2.G and psPAX2, pCDNA3.1-Flag-BACH1, and specific shRNA targeting of human were obtained from Miaoling Biology (Wuhan, China). shRNA targeting of mouse *Bach1* was purchased from Genechem (Shanghai, China). For stable cell line (U251-shBACH1 and GL261-shBach1-Luc) construction, lentivirus was produced in 293T cells helped with lentivirus packaging plasmids (pMD2.G and psPAX2). Transfections were helped with Lipofectamine 3000 (L3000015, Thermo Fisher Scientific) in accordance with the manufacturer's protocol. The specific sequence of shRNA targeting of U251 (shBACH1) is GATGTGCTGTGCGATGTCACC and the sequence of shRNA targeting GL261 (shBach1) is GCTCGACTGTATCCATGACAT.

Reagents and antibodies

Ulixertinib (an inhibitor of ERK1/2, S7854) and hemin (prothemin, an inhibitor of BACH1, S5645) were purchased from Selleck (USA).

The antibodies and their dilutions used in this research were as follows: anti-GAPDH (1:3000, 5174, Cell Signaling Technology), anti-BACH1 (1:1000, 14018-1-AP, Proteintech), anti-ERK1/2 (1:1000, 4695, Cell Signaling Technology), anti-Phospho-ERK1/2 (1:1000, 4370, Cell Signaling Technology), anti-Vimentin (1:2000, 10366-1-AP, Proteintech), anti-N-cadherin (1:1000, 22018-1-AP, Proteintech), and anti-cleaved caspase 3 (1:1000, ab32042, Abcam).

Western blot analysis

Cells were lysed in RIPA Buffer (Servicebio, China) supplemented with a phosphatase inhibitor Cocktail (Roche, Switzerland). Protein concentration was quantified using a BCA Protein Assay Kit (Beyotime, China). 10 µg of protein was separated on SDS-page and transferred to PVDF membrane (Millipore, USA). After blocking in nonfat milk, membranes were incubated with primary antibodies at 4°C overnight and then incubated with HRP-conjugated secondary antibodies for 1 hour at room temperature. Protein signals were visualized using a ChemiDoc Imaging System (Bio-Rad, USA).

RNA isolation and RT-qPCR

Total RNA was extracted using TRIzol reagent (Invitrogen, USA). RNA was converted to cDNA using a PrimeScript RT reagent kit (Takara, Japan) and RT-qPCR was performed with SYBR Green Supermix (Takara, Japan). The comparative Ct method ($\Delta\Delta Ct$) was used to analyze the relative expression from RT-qPCR experiments and *GAPDH* was used for normalization. Specific primer pairs were used: *GAPDH* (Homo sapiens), 5'-GGAAGCTTGTCATCAATGGAAATC-3'(Forward), 5'-TGATGACCCTTTTGGCTCCC-3'(Reverse); *BACH1* (Homo sapiens), 5'-GAACAGGGCTACTCGCAAAG-3'(Forward), 5'-AAAGGGCAGTTGACGGAAC-3'(Reverse); *Gapdh* (Mus musculus), 5'-TCAACAGCAACTCCCCTCTTCCA-3' (Forward), 5'-ACCCTGTTGCTGTAGCCGTATTCA-3' (Reverse); *Bach1* (Mus musculus), 5'-CACCTGACCGCCGCTTG-3'(Forward), 5'-TCGTAGGCAAATACCGCACT-3' (Reverse).

Colony formation assay

In colony formation assay, cells (500 cells/well) were seeded in six-well plates and incubated for 2 weeks. Then, the resulting colonies were washed with PBS, fixed with 4% polyoxymethylene, and stained with 0.1% crystal violet. The assay was performed in triplicate for each sample.

EdU-DNA Synthesis Assay

The cell proliferation was evaluated using Cell-Light EdU Apollo567 In Vitro Kit (RiboBio, China). After incubated with 50 µM EdU for 1 h, glioma cells were fixed with 4% paraformaldehyde and incubated with Apollo® reaction mix. The nuclei were stained by Hoechst 33342. Fluorescence was visualized and captured using an inverted microscope (Olympus BX51, Japan).

Transwell assay

5×10^4 cells were seeded into the upper chamber of polycarbonate transwell filters (Millipore, Millicell, USA) containing 200 µL serum-free DMEM. 600 µL of 10% FBS in DMEM was added in lower chambers as a chemotactic agent. Cells were incubated at 37°C for 24 h. The cells on the

lower chamber were fixed with 4% paraformaldehyde and dyed with crystal violet (Beyotime, China). Images were captured by an inverted microscope (Olympus BX51, Japan) and the assay was performed in triplicate for each sample.

Flow cytometry of apoptosis

Flow cytometry of apoptosis was conducted by dual staining using Annexin V-FITC and propidium iodide (PI) (Thermo Fisher Scientific, USA). Cells were collected, resuspended in 1× binding solution, and stained with Annexin V-FITC and PI. The apoptotic cells were analyzed using a flow cytometry (CytoFLEX, Beckman Coulter, USA) and analyzed by a FlowJo software (version 7.6.1).

***In vivo* experiments**

Four-week-old male BALB/c nude mice from Vital River Laboratory Animal Technology (Beijing, China) were used for the construction of GBM xenograft model. U251-shNC and U251-shBACH1 cells were intracranially injected into the brain (1 mm anterior and 2 mm lateral to bregma at a depth of 3mm) of BALB/c nude mice via stereotactic system (51725, Stoelting, USA). 14 days after tumor injection, ulixertinib (80 mg/kg) was applied orally twice daily (every 12 h) for five consecutive days. Tumor-bearing mice were sacrificed when they showed severe neurological symptoms and the brains were fixed with 4% paraformaldehyde for immunohistochemical and HE staining.

Four-week-old male C57BL/6 mice from Vital River Laboratory Animal Technology (Beijing, China) were used for the construction of GL261 murine glioma models. GL261 cells labelled with luciferase were intracranially injected into the brain (1 mm anterior and 2 mm lateral to bregma at a depth of 3mm) of C57BL/6 mice. 12 days after tumor injection, ulixertinib (80 mg/kg) was applied orally twice daily (every 12 h) for five consecutive days and hemin was intraperitoneally injected (50 mg/kg) daily (every 24 h) for five consecutive days. For the construction of another GL261 murine glioma model, GL261-shNC-Luc and GL261-shBach1-Luc cells were intracranially injected. Starting from day 12, ulixertinib (80 mg/kg) was applied orally twice daily (every 12 h) for five consecutive days. All tumor-bearing mice were captured by IVIS 10 days (day 10), 21days (day 21), and 28 days (day 28) after tumor injection and sacrificed when they showed severe neurological symptoms.

Animal experiments in this research were approved by the institutional animal care and use committee of Renmin Hospital of Wuhan University.

***In vivo* imaging**

In vivo Imaging System (IVIS) was applied for the visualization of GL261-luc in C57BL/6 mice. 10 mg D-Luciferin potassium was injected (ST196, Beyotime) for visualization in each mouse and the tumor was captured using an IVIS (PerkinElmer IVIS Lumina, USA).

Immunohistochemical staining

Tumor tissues of animal models were embedded with paraffin and sectioned. The sections were deparaffinized, hydrated, and antigen repaired with 10mM sodium citrate (pH, 6.0). The samples were incubated in 3% H₂O₂ to remove endogenous peroxidase and blocked by 1% Albumin Bovine V (Beyotime, China). Then, the slides were incubated with primary antibody overnight at 4°C and

visualized by a 3,3'-diaminobenzidine (DAB) kit (Solarbio, China). The images were captured under an inverted microscope (Olympus BX51, Japan)

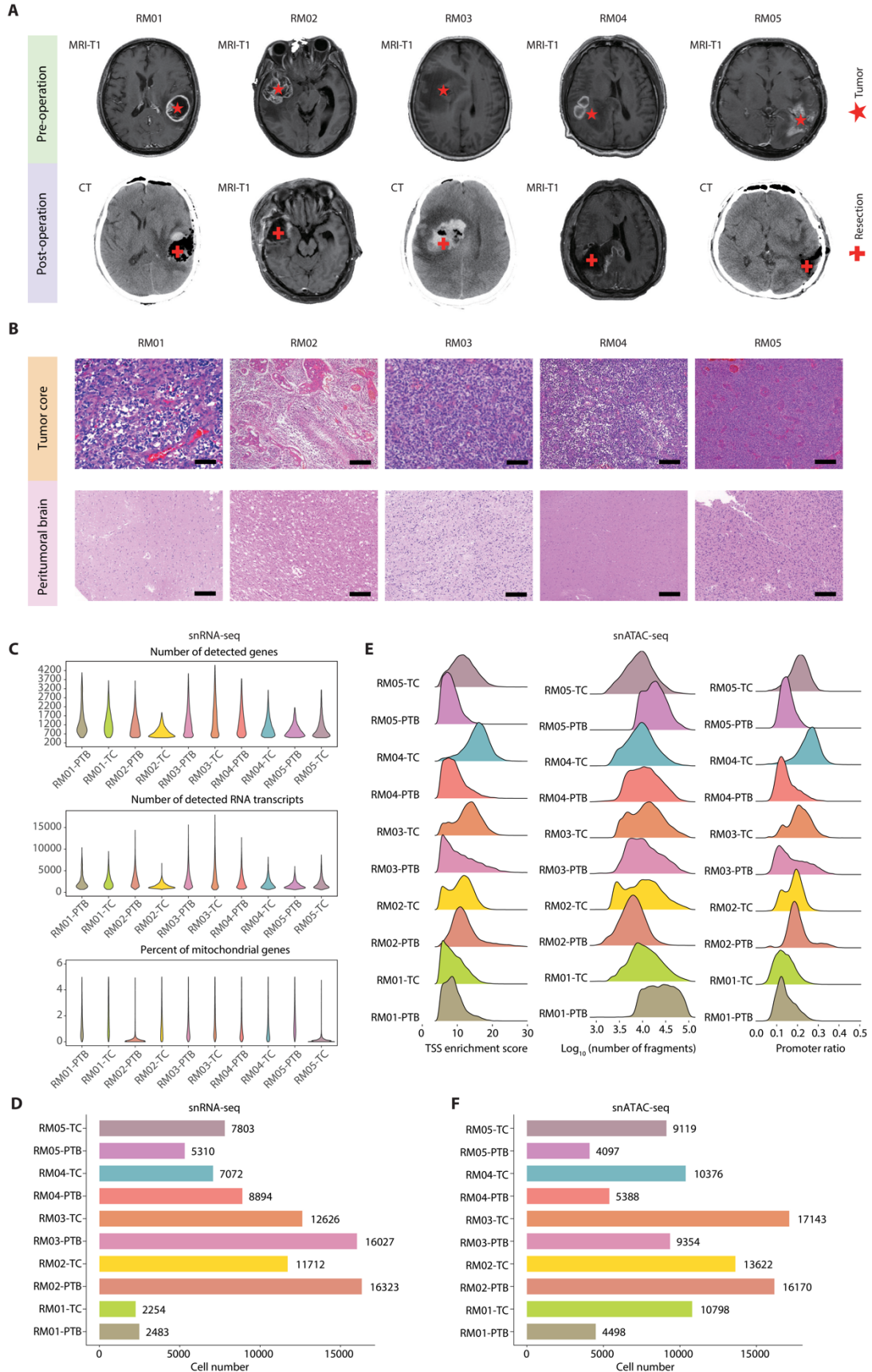


Fig. S1. Sample profile and quality control of single-cell data. (A) Pre-operation and post-operation brain imaging. Asterisks mark the original tumor site. Pluses mark the resection site. (B) Representative HE staining on each sample. Scale bars: 100 μm . (C) Violin diagrams showing the quality for snRNA-seq. The gene number, RNA count, and mitochondrial gene ratio per cell after quality control are shown. (D) Bar plot of cell number after quality control in snRNA-seq. (E) Ridge diagrams showing the quality for snATAC-seq. The TSS enrichment score, $\log_{10}(\text{unique nuclear fragments})$, and promoter ratio per cell after quality control are shown. (F) Bar plot of cell number after quality control in snATAC-seq.

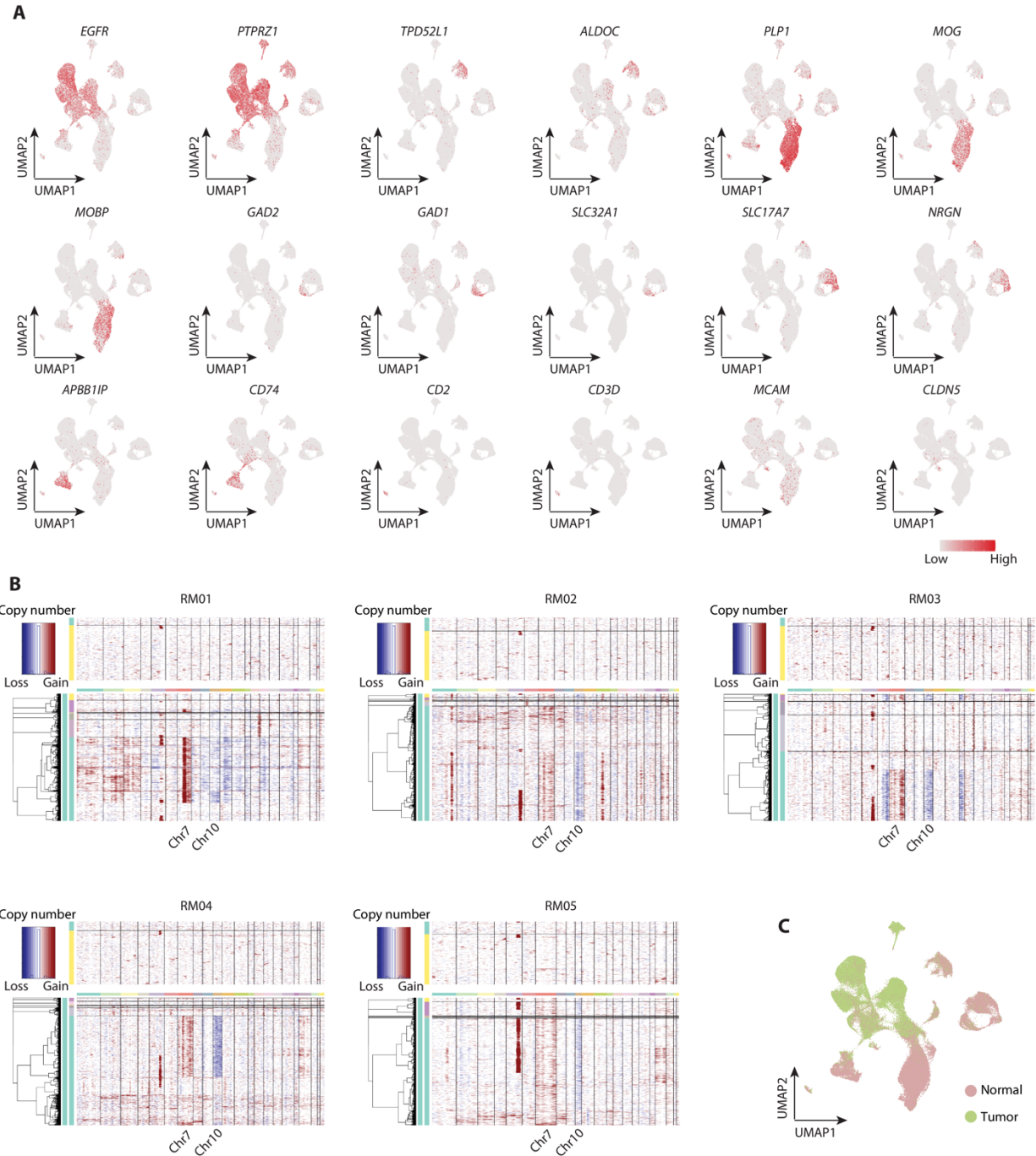


Fig. S2. Basis for cell type identification in snRNA-seq. (A) UMAP showing the expression of canonical marker genes for cell type identification. The redder the color, the higher the expression. (B) CNV analysis of putative malignant cells with InferCNV. The upper part of each heatmap represents normal cells and are considered as reference cells, containing myeloid cells and oligodendrocytes from the peritumoral brain of each patient. The lower part represents observation cells. (C) UMAP colored by malignant status (tumor or normal) inferred by inferCNV.

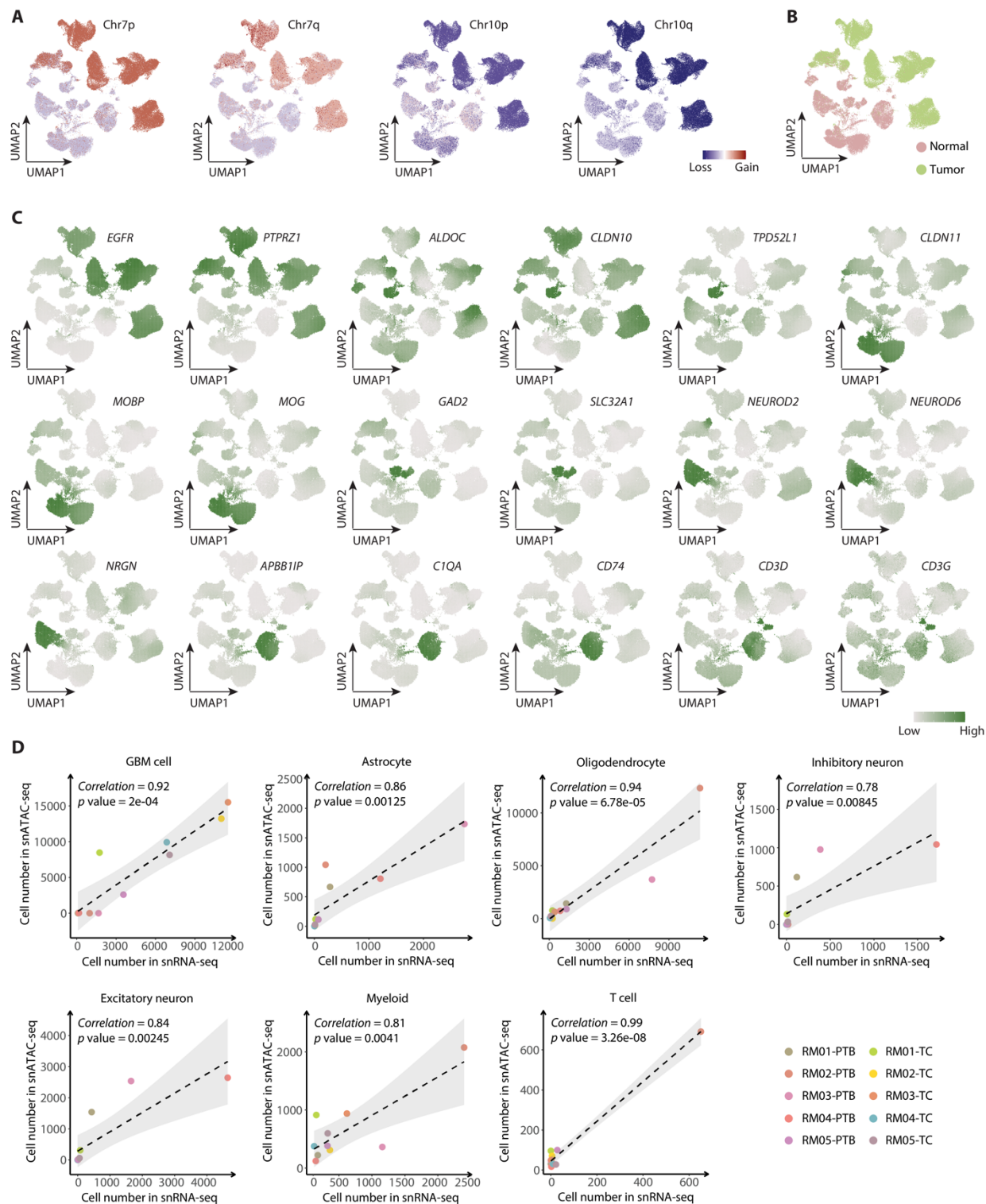


Fig. S3. Basis for cell type identification in snATAC-seq. (A) UMAP colored by canonical CNVs of GBM cells, including chr7p (+), chr7q (+), chr10p (-), and chr10q (-). (B) UMAP colored by malignant status (tumor or normal) inferred by Copy-scAT. (C) UMAP of gene activity score of canonical marker genes for cell type identification. (D) Scatter plots showing

the correlation of captured numbers of each cell type between snRNA-seq and snATAC-seq.
Dots are colored by cell types.

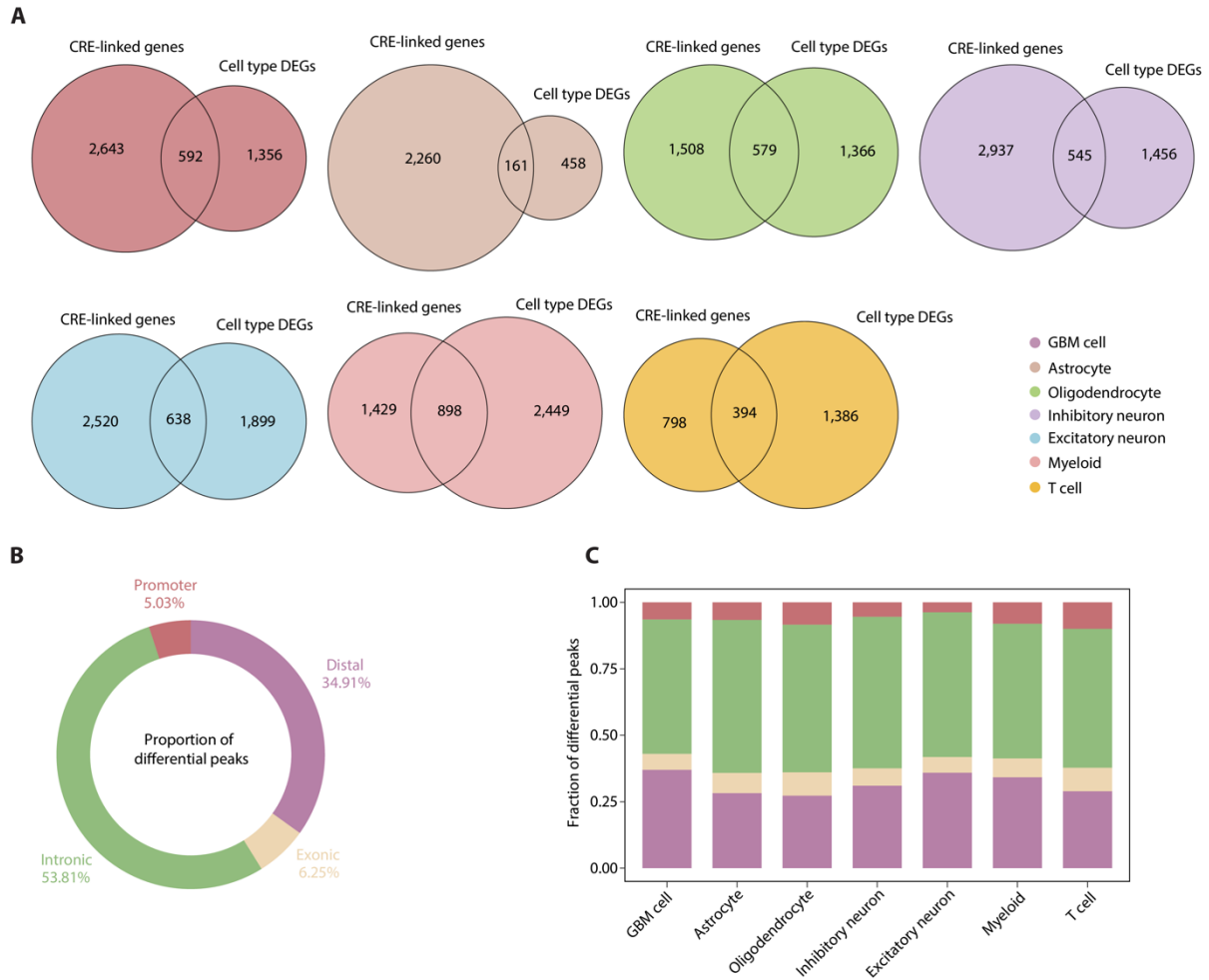


Fig. S4. Gene-linked CREs showing cell type specificity. (A) Venn diagrams for each cell type denoting the overlaps between the CRE-linked genes and upregulated genes in that cell type (cell type DEGs). (B) Donut diagram showing the genomic distribution of all differential peaks. (C) Stacked bar chart showing the genomic distribution of differential peaks across different cell types.

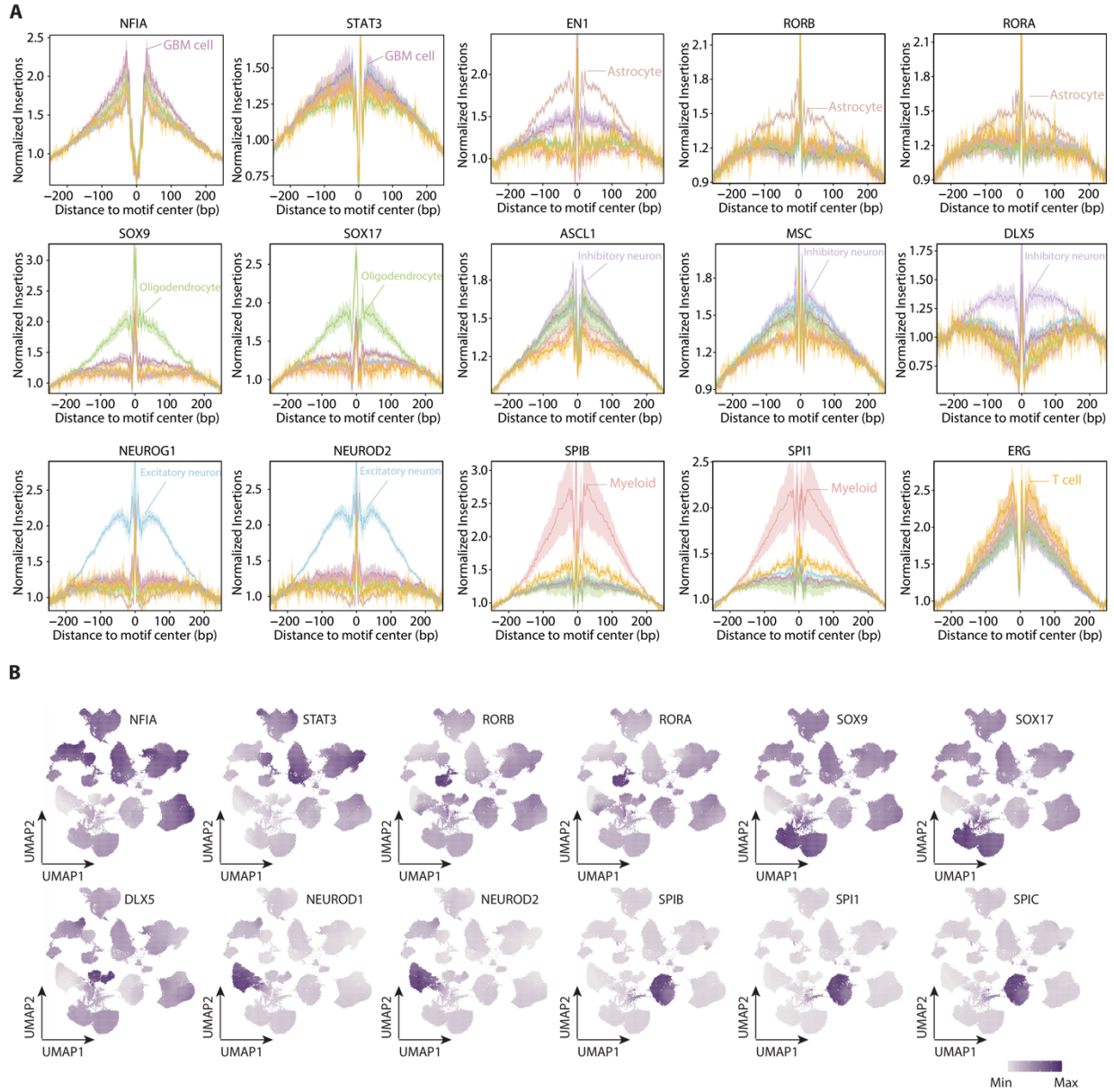


Fig. S5. TF regulation for each cell type. (A) Tn5 bias-adjusted TF footprints for cell-type specific regulators in snATAC-seq. **(B)** UMAP showing the cell type specific motif variability from snATAC-seq.

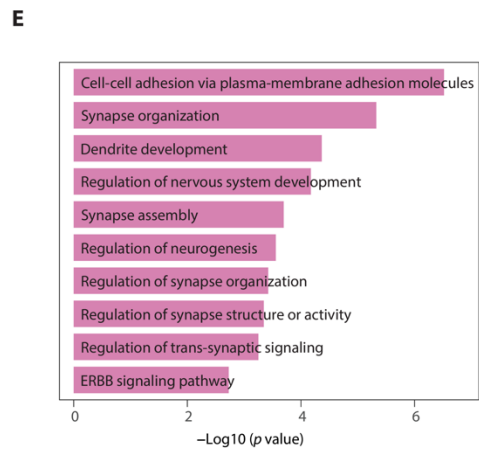
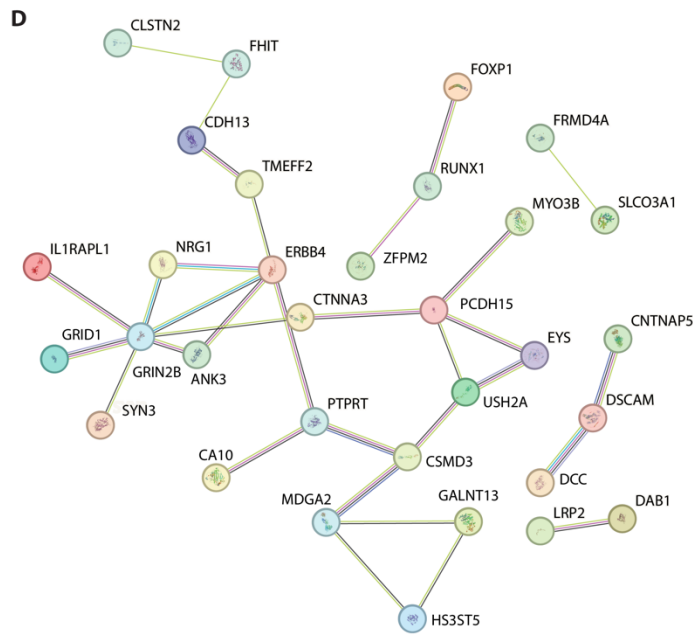
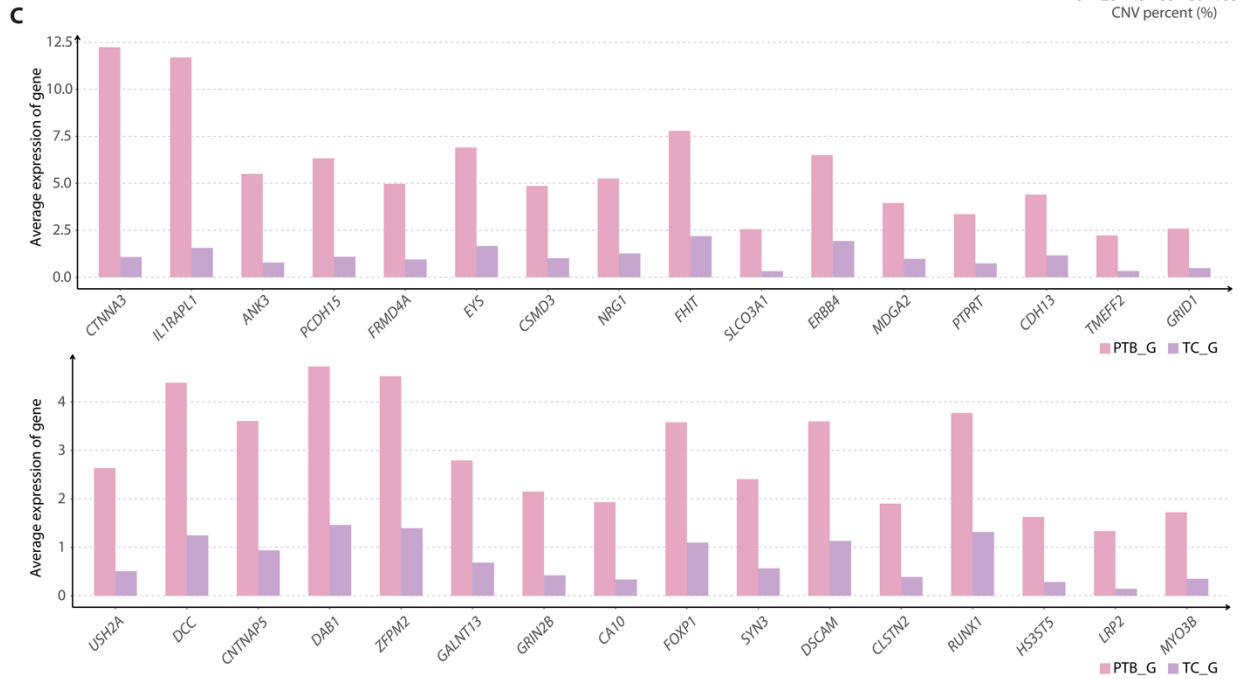
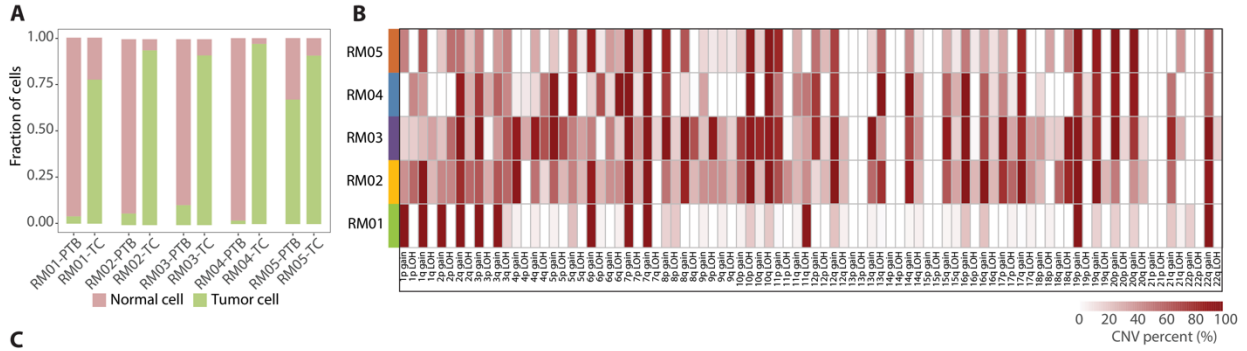


Fig. S6. GBM cells from the two regions harboring distinct transcription characteristics. (A) Bar plot showing the percentage of GBM cells and non-GBM cells in each sample. (B) The summary CNV profiles of GBM cells for each patient. The CNV levels were categorized by the chromosome arm and labeled as gain or loss. Color in the heatmap represents the proportion of cells harboring the CNV events in each individual patient. (C) Bar plots showing the expression of PTB_G signature genes in PTB_G and TC_G. (D) Network showing the PPI in PTB_G signature identified by STRING (53). (E) Bar plot showing the enriched GO terms of PTB_G signature genes.

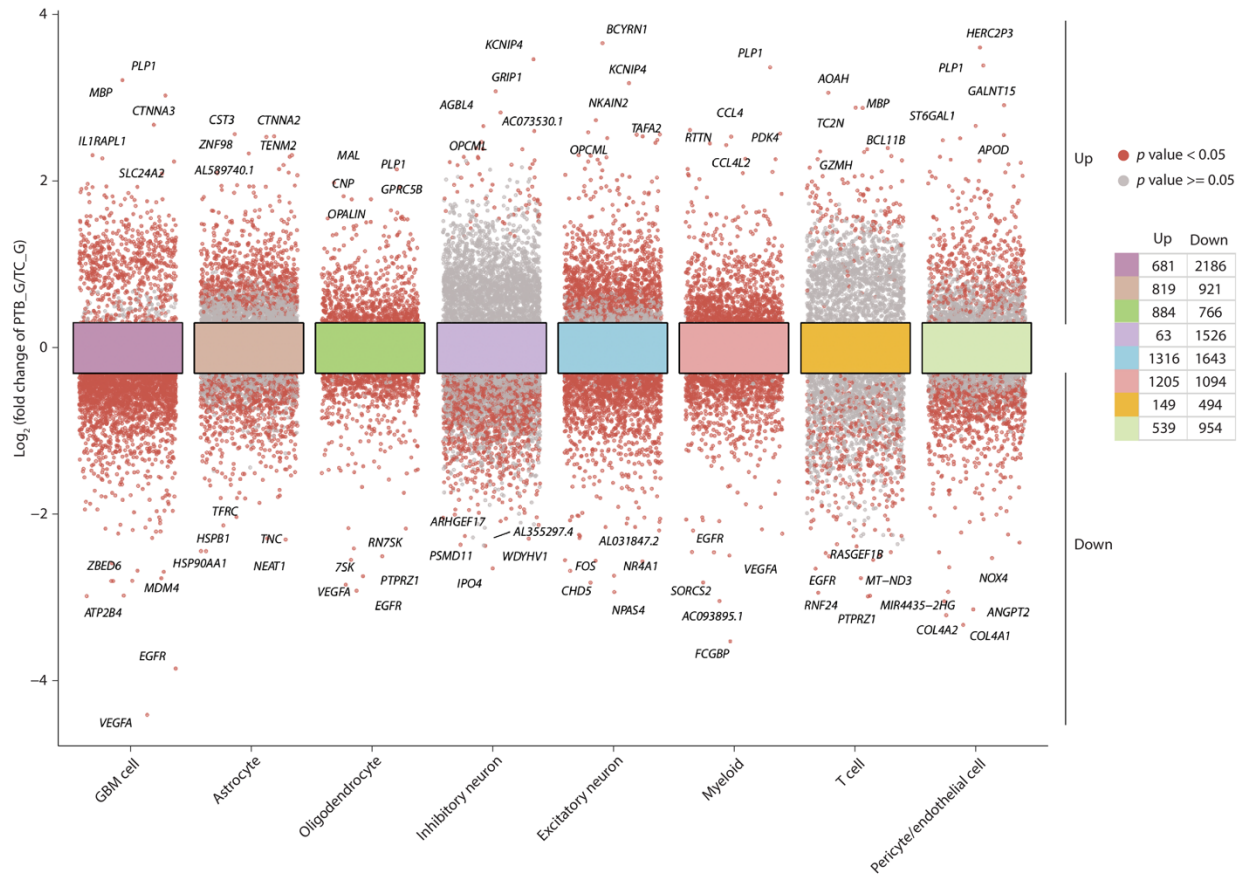


Fig. S7. The altered microenvironment across the two regions. Scatter plot showing the DEGs across the cell types.

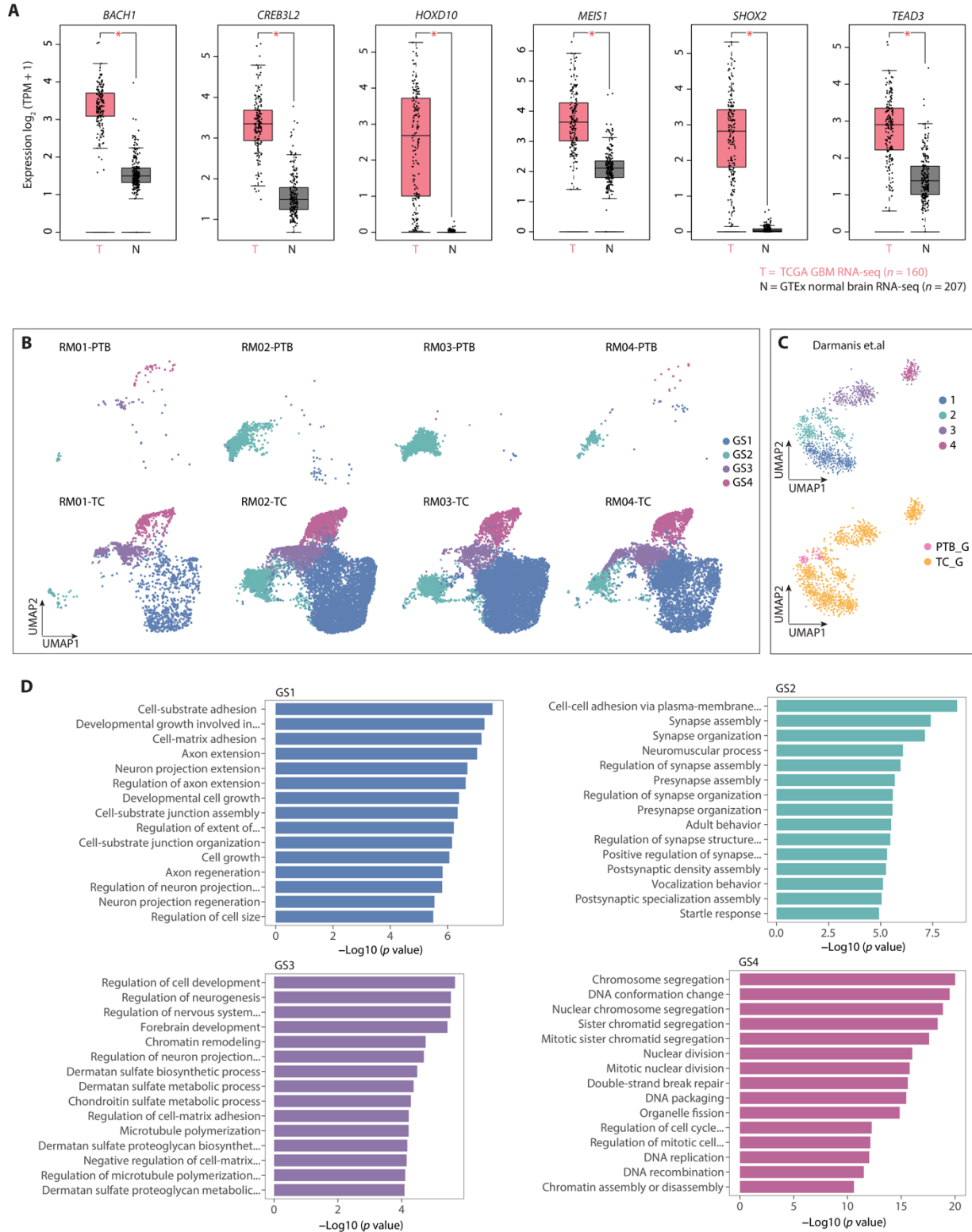


Fig. S8. Subclustering for GBM cells. (A) Boxplots showing the expression of the identified TFs in normal brain and GBM. (B) UMAP of GBM cells subclusters (i.e., GS1, GS2, GS3, and GS4) split by samples. The top panel represents samples from PTB, and the bottom panel

represents samples from the TC. The plot is colored by subclusters. **(C)** UMAP showing GBM cell populations from Darmanis et.al. **(D)** Bar plot of GO terms (biological process) enriched in each subcluster of GBM cells from Fig. 5C.

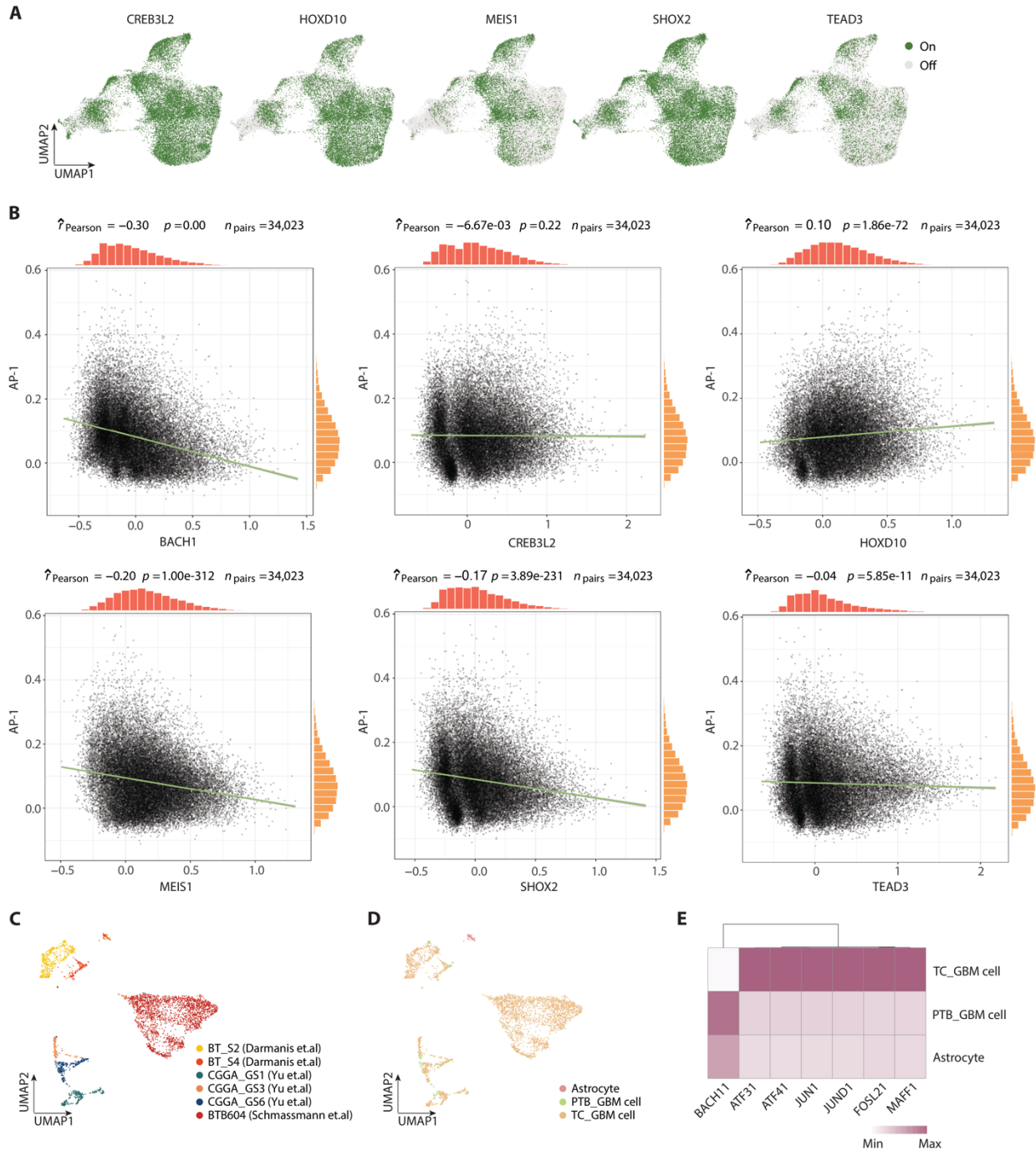


Fig. S9. The correlation analysis of candidate TFs with AP-1. (A) UMAP diagrams showing the activation of candidate regulons in GBM cells. (B) Scatter plots showing the expression correlation of the six candidate regulons with AP-1 regulons. (C and D) UMAP plots of cells collected from public datasets, colored by sample origins (C) and cell populations (D), respectively. (E) Heatmap showing the activity of BACH1 and AP-1 in three groups.

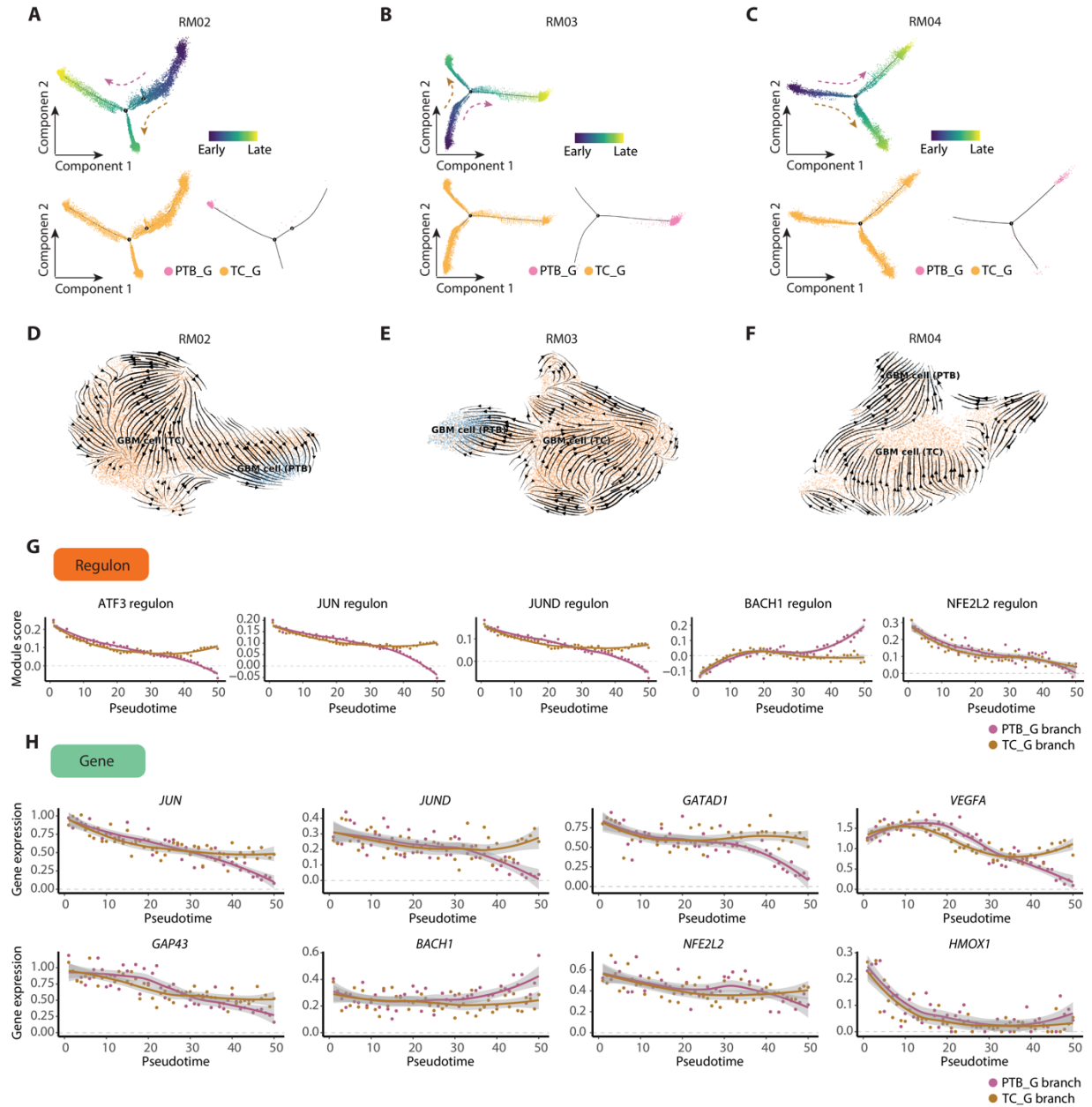


Fig. S10. Pseudo-time trajectory for GBM cells. (A-C) Monocle trajectory for GBM cells from the two regions in RM02 (A), RM03 (B), and RM04 (C). The top image indicates the differentiation time. The bottom two images show the location for different regions of GBM cells. (D-F) RNA velocities for GBM cells in RM02 (D), RM03 (E), and RM04 (F). The associated streamlines are represented in the plots and dots are colored by regions of GBM cells. (G and H) Scatter plots for scores of selected regulons (G) and expression of genes (H). The colors represent different branches.

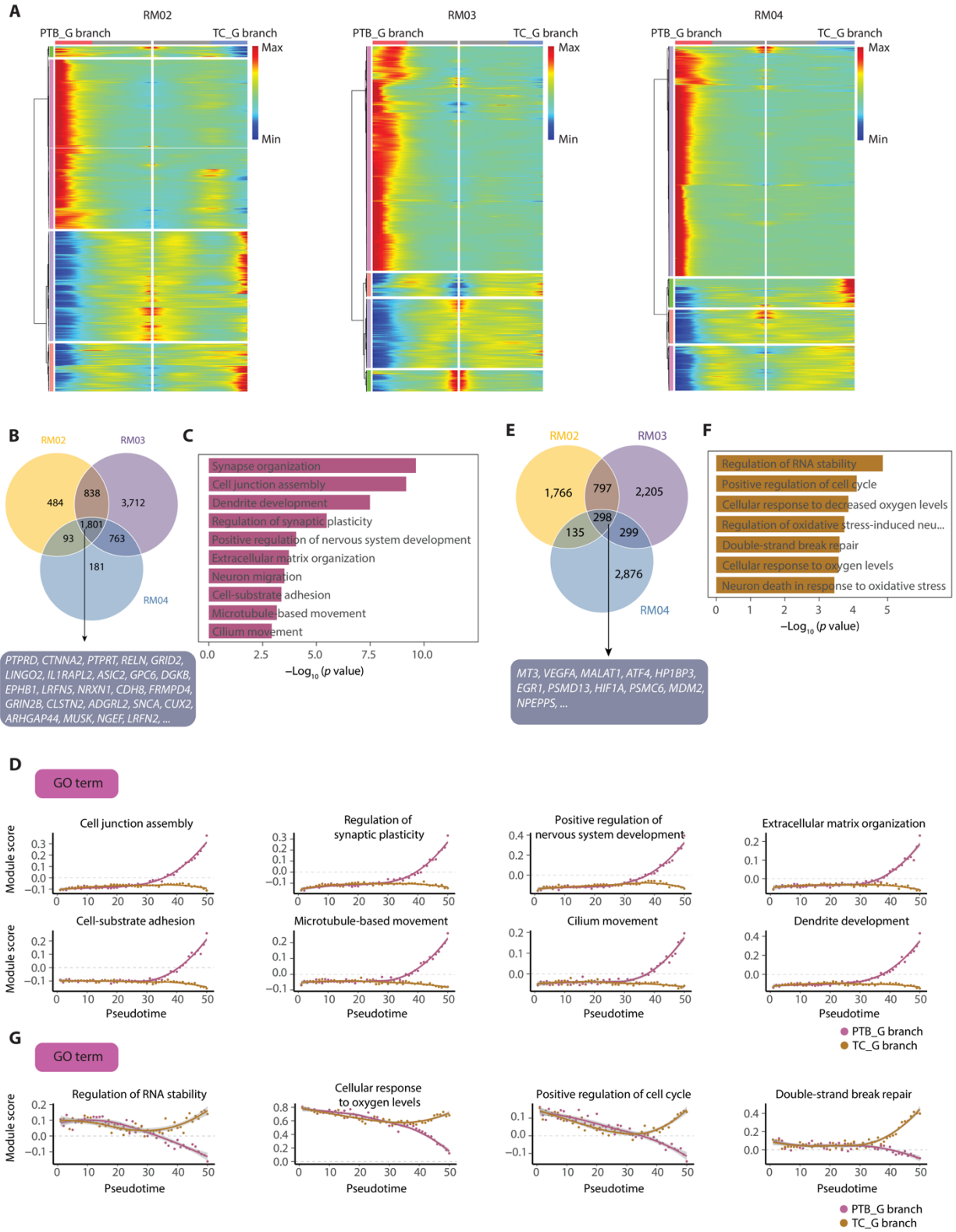


Fig. S11. Changes of different characteristics along the monocle trajectory. (A) Heatmap showing the kinetic clusters of branch-dependent genes for RM02 (left), RM03 (mid), and RM04

(right). **(B)** Venn diagram showing the intersection of genes from three patients, with their expression increasing along the branch located by GBM cells from tumor core. **(C)** Bar plot of GO terms (biological process) enriched by intersection genes for fig. S11B. **(D)** Scatter plots of module scores for GO terms from fig. S11C. **(E)** Venn diagram showing the intersection of genes from three patients, with their expression increasing along the branch located by GBM cells from tumor core. **(F)** Bar plot of GO terms (biological process) enriched by intersection genes for fig. S11E. **(G)** Scatter plots of module scores for GO terms from fig. S11F.

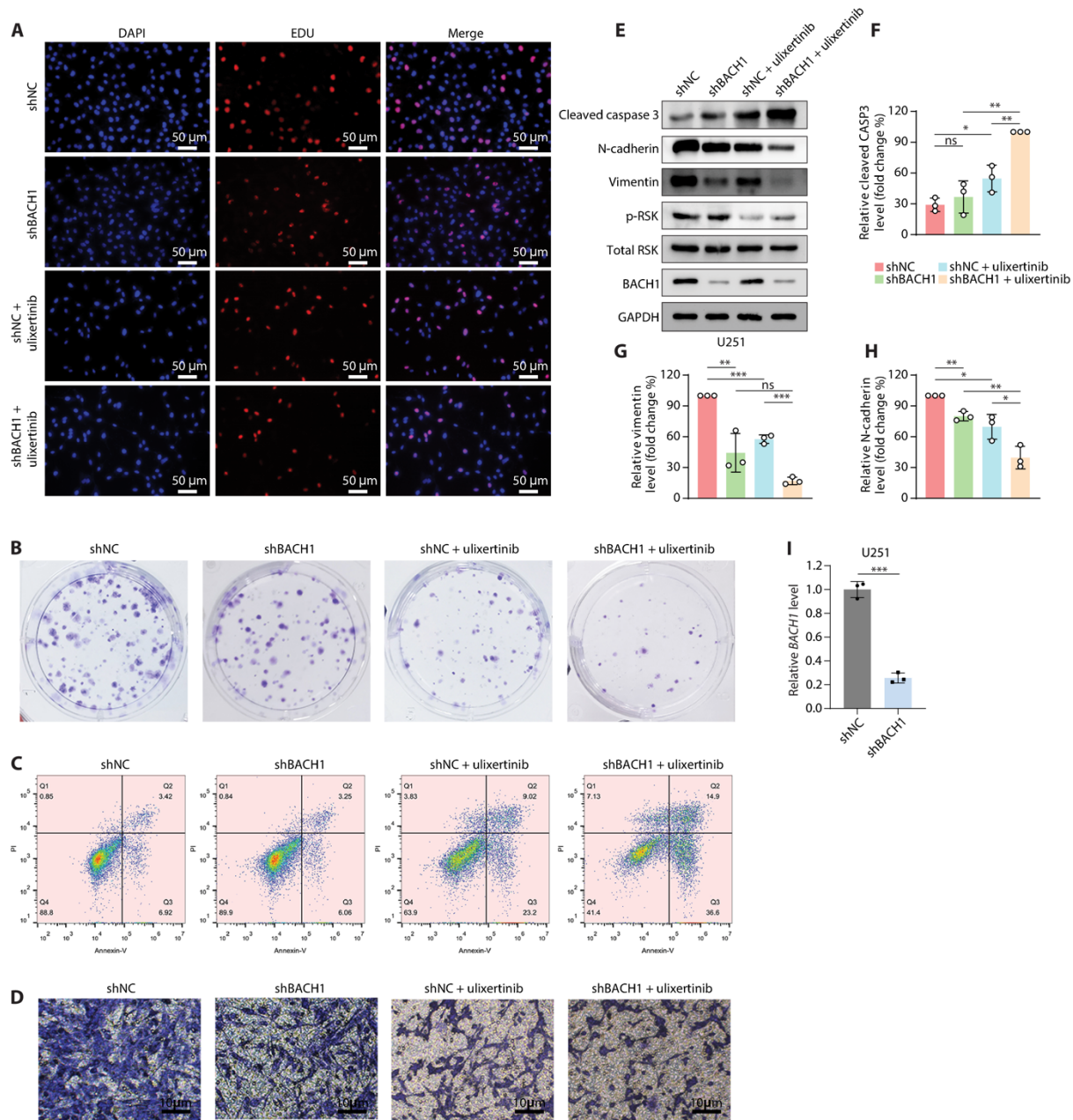


Fig. S12. Combination therapy in GBM suppression. (A-D) Representative images showing BrdU incorporation (A), colony-formation (B), flow cytometry (C), and transwell migration (D) assays of the indicated groups in Fig. 6A. (E-H) Western blot (E) of indicated proteins in U251 cells with multiple treatments, and the quantification were shown (F-H) ($n = 3$ biological replicates). (I) qRT-PCR of BACH1 mRNA expression in U251 cells with or without BACH1 depletion. $n = 3$ technical replicates. * $p < 0.05$, ** $p < 0.01$, and *** $p < 0.001$.

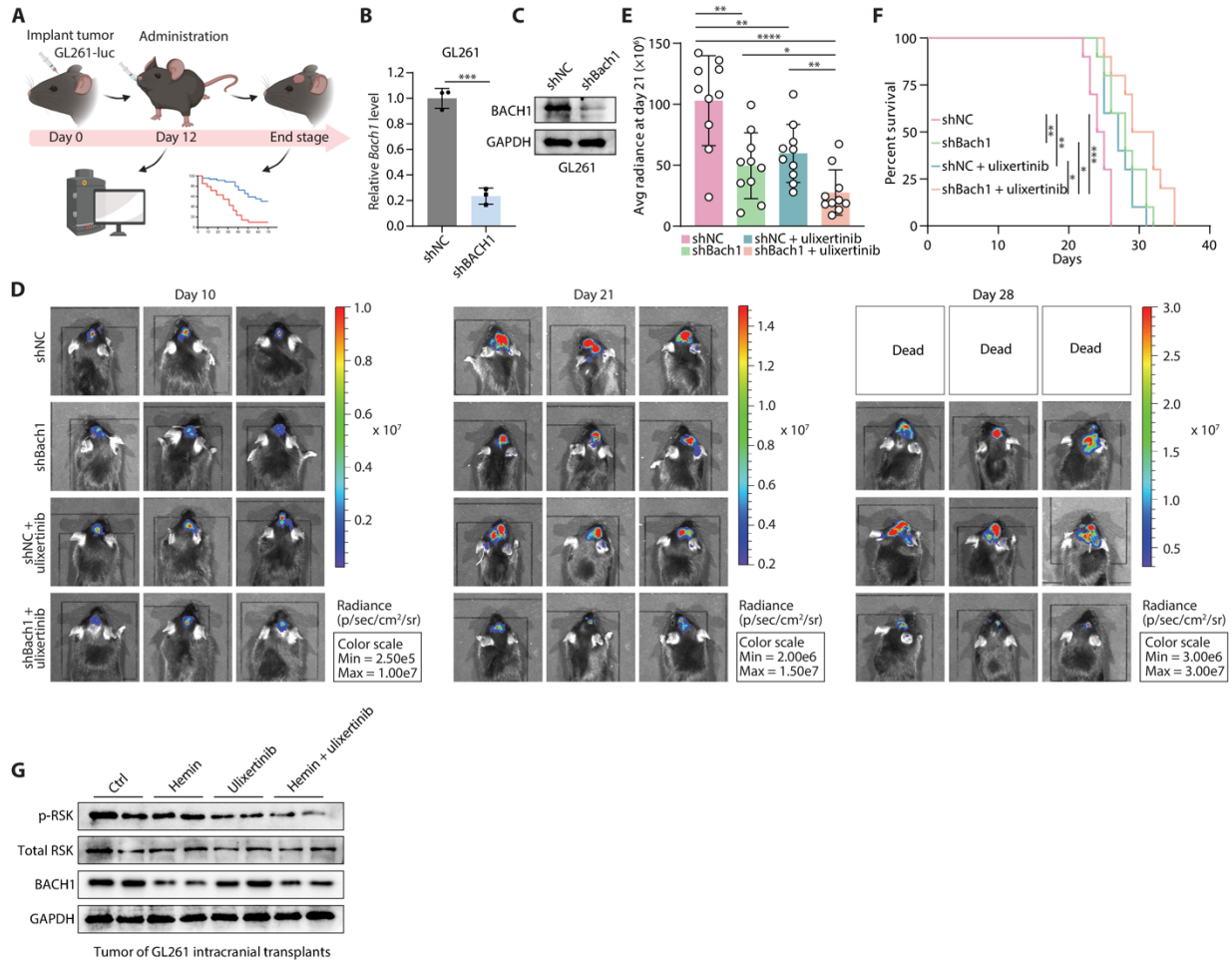


Fig. S13. Combination therapy suppresses tumor growth and improves survival in GL261-derived mouse models. (A) Illustration depicting the therapy schedule. Luciferase-expressing GL261 cells (with or without BACH1 depletion) were intracranially injected into nude mice on day 0. Treatment started on day 12, with or without ulixertinib (80 mg/kg) orally twice daily (every 12 h) for 5 days. ($n = 10$ for each group). (B) qRT-PCR of BACH1 mRNA expression in GL261 cells with or without BACH1 depletion. $n = 3$ technical replicates. (C) Western blot showing the BACH1 protein in GL261 cells with or without BACH1 depletion. (D) Luminescence intensity showing tumors in representative mice over time. (E) Quantitative radiance in the experimental group on day 21. (F) The survival time of the indicated groups of mice in (A). (G) Western blot showing the indicated proteins in mouse tumor lysates from Fig. 6F. * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, and **** $p < 0.0001$.

Table S1. Clinical and experimental information for surgical samples.

Table S2. The summary statistics for snRNA-seq data.

Table S3. The summary statistics for snATAC-seq data.

Table S4. List of DEGs in cell types from snRNA-seq.

Table S5. The compositions of cell types in snRNA-seq and snATAC-seq.

Table S6. ChIP-seq datasets employed in genome track visualization.

Table S7. Distribution of CNVs on the chromosome arms for patients in snRNA-seq.

Table S8. Pan cancer signatures for feature scoring in GBM cells.

Table S9. List of DEGs between GBM cells from two regions and astrocyte from peritumoral brain.

Table S10. List of DEGs of GBM cells between two regions.

Table S11. The percentages of composition for the four GBM cellular states.

Table S12. Lists of motifs enriched in four GBM cellular states in snATAC-seq.

Table S13. List of potential regulons for PTB_G.

Table S14. GO terms (biological processes) enriched in GBM cell subclusters.

Table S15. The fractions of GBM cells with activated regulon for AP1 or/and other candidates.

Table S16. The numbers of cells activated for AP1 or/and BACH1 regulons.

Table S17. Lists of genes with elevated expression along the two monocle branches separately across the three patients.

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