

Supplemental Information

**Human Cortical Organoids Expose a Differential Function of GSK3 on
Cortical Neurogenesis**

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Supplemental information

1. Supplemental figures S1 to S5 legends
2. Supplemental experimental procedures
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4. Supplemental tables. List of differentially expressed genes. List of GSEA categories (Excel files).

1. Supplemental figure legends

Figure S1. Expression profiles for gene signatures defining specific developmental stages. Related to Figure 3.

Stripcharts for expression levels (in log₂cpm) from bulk RNA sequencing at day 18 (A), day 50 (B) and day 100 (C). Unexposed samples are visualized in turquoise, CHIR-exposed samples in orange. Blue and red boxes highlight respectively genes down- or up-regulated by CHIR treatment, considering the following thresholds from RNASeq bulk differential expression analysis: *: nominal p-value < 0.05; **: nominal p-value < 0.01; ***: FDR < 0.05 and absolute log₂FC > 1.

Figure S2. Expression profiles for genes related to dorsalising/ventralizing fate.

Stripcharts for expression levels (in log₂cpm) from bulk RNA sequencing at day 18 (A), day 50 (B) and day 100 (C). Unexposed samples are visualized in turquoise, CHIR-exposed samples in orange. Blue and red boxes highlight respectively genes down- or up-regulated by CHIR treatment, considering the following thresholds from RNASeq bulk differential expression analysis: *: nominal p-value < 0.05; **: nominal p-value < 0.01.

Figure S3. Breakdown of stage-wise differential expression analysis. Related to Figure 3.

A. Top-100 genes according to absolute loading value for the first component from Principal Component Analysis. **B.** Results of gene ontology enrichment analysis performed for Biological Process GO domain on the 100 genes selected as detailed in A; p-values for the top-10 categories are shown. **C.** Gene expression profiles (as z-scores) for the genes identified in bulk RNASeq as modulated by GSK3 treatment (FDR < 0.05, absolute log₂FC > 1) at day 18 (A), day 50 (B) and day 100 (C). **D.** Results of gene ontology enrichment analysis performed on the DEGs identified at day 50, divided for up-regulated and down-regulated genes. Bar plots depict the p-values for the top-10 Cellular Component GO terms.

Figure S4. Overlap of Louvain clusters to cell identities within human fetal data single cell transcriptomes.

Related to Figure 4. A. For each sub-panel, cells (represented as dots) are colored according to the expression levels of paired combinations of representative cell type markers (DCX, STMN2 = mature neurons), (ENO2, ST18 = early neural progenitors), (FAM107A, HOPX = outer radial glia), (SMOC1, S100B = intermediate progenitors) (CDK1, CDC20, MKI67 = proliferating progenitors) (CXCL14 = choroid/Astro). **B.** Heatmap of the overlaps between marker genes characterizing the internal organoids clusters and external gene signatures of the relevant single cell clusters from human fetal brains (Nowakowski et al., 2017). Colors are based on the log2Enrichment computed for each pair of comparisons. * indicates an overlap with p-value < 0.05.

Figure S5. UMAP plots for proliferation/maturity distribution and pseudotime decomposition. Related to

Figure 6. A. Visualization of expression levels of proliferation/progenitor markers (in red: CDK1, MKI67) coupled with maturity markers (in green: DCX, STMN2), (in yellow: overlapping expression). UMAPS are visualized separately for each timepoint and condition. **B.** First and second components inferred from condition-wise diffusion components and visualized as color gradient on UMAP. The two extremes of the color scale (origin-blue and dark red-outcome) represent the distance in the relative diffusion component. CTL D100 is the only condition with an end point in oRG area.

2. Supplemental experimental procedures

hPSC culture. hPSC lines were cultured under feeder-free conditions on Matrigel (BD Biosciences) coated dishes (diluted 1:40 matrigel:DMEM/F12) and grown in TesrTM E8TM medium (Stem Cell Technologies). Cells were passaged upon treatment with ReLeSR (Stem Cell Technologies). All differentiation procedures were performed on iPSC lines with at least 15 passages after reprogramming. Pluripotent lines came from different individuals representing either human iPSC or ESC lines previously described (Adamo et al., 2015; González et al., 2014). All the cultures were regularly tested and maintained mycoplasma free.

Table 1. Distribution of hPSC lines used in this study.

Experiment/Line	809-1-5 (iPSC1)	KOLF2CL (iPSC2)	MIFF1-3 (iPSC3)	3391B (iPSC4)	HUES8 (ESC)
ICC		X	X	X	X
Rosette formation assay		X	X	X	

Growth curve			X	X	X
Bulk RNAseq	X	X	X	X	X
Single cell RNAseq	X	X	X	X	X

Neural induction and lumen quantification. hPSC cells were plated at a density of 0.7×10^5 cells cm^2 on Matrigel (BD Biosciences) coated dishes in grow medium supplemented with 10 μM ROCK inhibitor (Y-2763221, Cell Guidance System). Cell cultures were expanded for two days until they were 70% confluent. The starting differentiation medium includes DMEM/F12 (Life Technologies) with N2 and B27 without retinoic acid (Life Technologies), supplemented with 500 nM LDN193189 (Sigma) and 10 μM SB431542 (Tocris). Cells were fixed at day 20 in 4% (wt/vol) paraformaldehyde (PFA) for 15 minutes at room temperature (RT) and washed 3 times with phosphate-buffered saline (PBS). Cells were then permeabilized with PBS containing 0.5% Triton X-100 (Sigma) and blocked with 10% (vol/vol) normal goat serum (NGS; Vector) for 1 hour at RT. Next, cells were incubated overnight at 4°C. The following primary antibodies and dilutions were used: Anti-PALS1, (Proteintech 117710-1-AP). Object identification module of Cell Profiler software (v.2.1.1) was used to automatically quantify lumens (number and size of Pals1 positive areas) of neural rosettes.

Cortical organoid protocol. hPSCs were plated onto cell cycle-arrested mouse embryonic feeders (MEFs)(Millipore) for one passage, colonies grown for at least 48h and then enzymatically detached by incubation with 0.7 mg/ml dispase (Invitrogen: 17105-041) for approx. 30 min. Suspended colonies were subsequently transferred into ultra-low-attachment 100 mm plastic plates (Corning) in FGF2-free knockout serum medium. For the first 24 h (day 0), the medium was supplemented with the ROCK inhibitor Y-27632 (EMD Chemicals). For neural induction, dorsomorphin (Merck, 5 μM) and SB-431542 (Tocris, 10 μM) were added to the medium until day 5. From day 6 onward, organoids were moved to neural medium (NM) containing Neurobasal (Invitrogen 10888), B-27 serum substitute without vitamin A (Invitrogen 12587), GlutaMax 1:100 (Fisher 35050071), 100 U/ml penicillin and streptomycin (Invitrogen) and 50 mM b-Mercaptoethanol (Gibco 31350010). The NM was supplemented with 20 ng/ml FGF2 (Thermo) and 20 ng/ml EGF (Tocris) for 19 days with daily medium change in the first 10 days, and every other day for the subsequent 9 days. On day 12, floating organoids were moved to orbital shaker (VWR Standard Orbital Shaker, Model 1000) and kept on constant shaking at 50 rpm to promote nutrient and oxygen exchange. To induce neurogenesis, FGF2 and EGF were replaced with 20 ng/ml BDNF (Peprotech) and 20 ng/ml NT3 (Peprotech) starting at day 25, while from day 43 onwards only NM without growth factors was used for medium changes every other day.

Primary and secondary antibodies used and quantifications. Primary antibodies were prepared in PBS + 5% normal donkey serum (Jackson Immuno Research) overnight at 4°C. The following primary antibodies and dilutions were used: anti-PAX6, 1:200 (Biolegend PRB-278B); anti-KI67, 1:200 (Abcam ab15580); anti-NESTIN, 1:500 (Millipore MAB5326); anti-DCX, 1:1000 (BD Biosciences 611706); anti-TBR1, 1:200 (Abcam ab31940), anti-HOPX, 1:50 (Sigma-Aldrich HPA030180), Anti CTIP2 1:200 (Abcam ab18465). After primary incubation, sections were washed three times with PBS and then incubated with appropriate secondary antibodies: anti-rabbit Alexa 488-conjugated (Bethyl A090-516D2) anti-mouse, Alexa 594-conjugated (Bethyl A110-305B4) (Molecular Probes, Invitrogen) diluted 1:500 in blocking solution and incubated for 2 h at RT. Before mounting, sections were incubated with Hoechst 33258 (5 µg/mL; Molecular Probes, Invitrogen) or DAPI 1:5000 (Merck), as indicated on each caption. Quantification of nuclear markers was done by using the automatic cluster counter ITCN plugin from FIJI (v.1.49 NIH-USA) for at least 3 organoids from 3 independent lines. Images were RGB converted and cluster counts were done over pre-defined concentric grids from VLS. The relative number of positive cells was calculated as a percentage of total DAPI+ cells. Images were acquired with a Leica DMI 6000B microscope (10x, 20x and 40x objectives) and analyzed with LAS-AF imaging software and then processed using Image J (v1.49 NIH, USA) to adjust contrast for optimal RGB rendering. Semi-quantitative measurements were made in imageJ, the background noise was removed with a sliding paraboloid filter of dimension 500 px, then a gaussian filter was applied. A positive cell was considered the local maxima with a minimum threshold of signal to noise. To evaluate the density of cells we consider organoid slice stained with DAPI, from the images of the whole organoid we remove, with a sliding paraboloid filter, the background and then apply a common threshold on all the conditions. The area of the resulting mask is considered as area occupied by nuclei and used as normalization.

Bulk transcriptome analysis. Differential gene expression. Gene expression quantification at the gene level was performed by Salmon (version 0.8.2) (Patro et al., 2017), using hg38 RefSeq annotation. To estimate differential expression, the matrix of gene counts was analyzed by edgeR (version 3.20.9) (Robinson et al., 2009). For each time point, genes with an expression level of at least 2 cpm (count per million) in at least 3 samples were selected for the analysis. Small genes, ribosomal genes and fusion genes were excluded. After TMM normalization, differential expression analysis comparing treated to untreated samples was performed using a likelihood ratio test on the coefficients of a negative binomial model. Significantly modulated genes were selected setting an absolute value of log2 fold change (Log2FC) higher than 1 and a false discovery rate (FDR) lower than 5%. Log2 cpm values, were

used for heatmap representation of gene expression profiles (visualized as z-scores). Heatmaps were produced with pheatmap R package (version 1.0.10, Raivo Kolde (2018). pheatmap: Pretty Heatmaps.). Analyses were performed in R version 3.4.4. Functional annotation of biological functions was performed by Gene ontology analysis and Gene set enrichment analysis (GSEA) using as set source H1 collection from the Molecular Signature Database (Liberzon et al., 2015). *Gene Ontology Enrichment Analysis*. Gene ontology enrichment analysis for the Cellular Component domain of the ontology was performed on the 898 DEGs identified at day 50, split in up-regulated and down-regulated genes. The pool of tested genes (as selected for differential expression analysis) was used as background. The analysis was performed by topGO (version 2.30.1) (Adrian Alexa and Jorg Rahnenfuhrer 2016). topGO: Enrichment Analysis for Gene Ontology), relying on Fisher test and Weight01 method to take into account ontology hierarchy; minimum node size was set at 15. After imposing an enrichment cut-off of 2, a 0.01 p-value cut off was applied to select significantly enriched GO terms. Barplot in supplementary figure 4 shows the top-10 categories ranked for p-value.

Gene set enrichment analysis. GSEA was applied to each developmental stage with GSAA software, version 1.2 (Xiong et al., 2014). Raw reads for the same genes tested for differential expression were analyzed by GSAASeqSP (permutation type 'gene set').

Single cell library preparation and sequencing. Briefly, a small volume (6 - 8 μ l) of single-cell suspension at a density of 1000 cells/ μ l was mixed with RT-PCR master mix and immediately loaded together with Single-Cell 3' gel beads and partitioning oil into a single-cell 3' Chip. The gel beads were coated with unique primers bearing 10 \times cell barcodes, unique molecular identifiers (UMI) and poly(dT) sequences. The chip was then loaded onto a Chromium instrument (10 \times Genomics) for single-cell GEM generation and barcoding. RNA transcripts from single cells were reverse-transcribed within droplets to generate barcoded full-length cDNA using Clontech SMART technology. After emulsion disruption, cDNA molecules from one sample were pooled and preamplified. Finally, amplified cDNAs were fragmented, and adapter and sample indices were incorporated into finished libraries which were compatible with Illumina sequencing. The final libraries were quantified by Qubit system (Thermo) and calibrated with an in-house control sequencing library. The size profiles of the pre-amplified cDNA and sequencing libraries were examined by Agilent Bioanalyzer 2100 using a High Sensitivity DNA chip (Agilent). Two indexed libraries were equimolarly pooled and sequenced on Illumina NOVAseq 6000 platform using the v2 Kit (Illumina, San Diego, CA) with a customized paired-end, dual indexing format according to the recommendation by 10 \times Genomics. Using proper cluster

density, a coverage around 250 M reads per sample (2000–5000 cells) were obtained corresponding to at least 50,000 reads/cell.

Single cell transcriptome analysis. Before downstream analyses, data deriving from the 11 samples was integrated by Seurat v3.0-alpha analytical framework (Stuart et al., 2018). After normalization, anchors for data integration were identified considering 3000 anchor points (genes) and 40 dimensions. For data reduction, UMAP was applied with 50 nearest neighbors (nn); cluster initial positions were set considering PAGA node position (Scanpy v1.3.1) (Wolf et al., 2018). On the integrated dataset, clusters were identified by applying Louvain with Multilevel Refinement from Seurat with resolution parameter at 0.7. This resulted in the identification of 15 clusters. For cluster annotation, we applied the FindMarker Seurat function, using MAST as test and filtering for up-regulated genes with adjusted P value < 0.05 . The obtained lists were compared in an overlap analysis with gene lists derived from two published single cell datasets: from the WGCNA analysis of single cell clusters of human fetal brains (paper by Nowakowski et al, DOI: 10.1126/science.aap8809), data were downloaded from the web portal at <https://cells.ucsc.edu/?ds=cortex-dev> and upregulated genes ($vg_diff.float > 1.5$) were selected for each relevant cluster, from the differential expression analysis of single nuclei clusters (Table S6 of the paper by Amiri et al, DOI: 10.1126/science.aat6720), upregulated genes (fold change > 0 and adjusted P value < 0.05) were selected for each relevant cluster. P values and enrichment coefficients were computed relative to a universe of 3000 genes used for all the other single cell analysis. Cluster-specific expression levels of biologically-relevant genes identified among the top dysregulated were visualized by violin plots, stratified for stage and treatment. To test for unbalance in the number of highly expressing cells for representative genes per cluster or cluster set, a threshold of expression at the 90th percentile was fixed, and the number of cells above or below the threshold in treated and untreated cells was compared by Fisher test. The identification of the threshold and sub-sequent analysis was performed on the sub-sampled dataset, separately for Day 50 and Day 100. Cell cycle analysis were performed using Scanpy function `score_genes_cell_cycle`, relying on the genes from (Kowalczyk et al., 2015). Diffusion map algorithm for dimensionality reduction was performed with Scanpy with 50 nn. Pseudotime analysis for lineage branching reconstruction was applied using wishbone algorithm (Setty et al., 2016). The analysis was performed on the complete dataset, as well as separately for each of the four biological conditions in order to infer stage or treatment-selective trajectories; the origin was identified with the same method applied on complete dataset. Trajectories were reproduced defining using at least 3 different markers. Partition-based

graph abstraction (PAGA) algorithm was applied on the complete dataset, as well as separately for each of the four biological conditions and plotted with layout Reingold Tilford. The position of the nodes identified on the complete dataset was exploited in the graph for each biological condition.

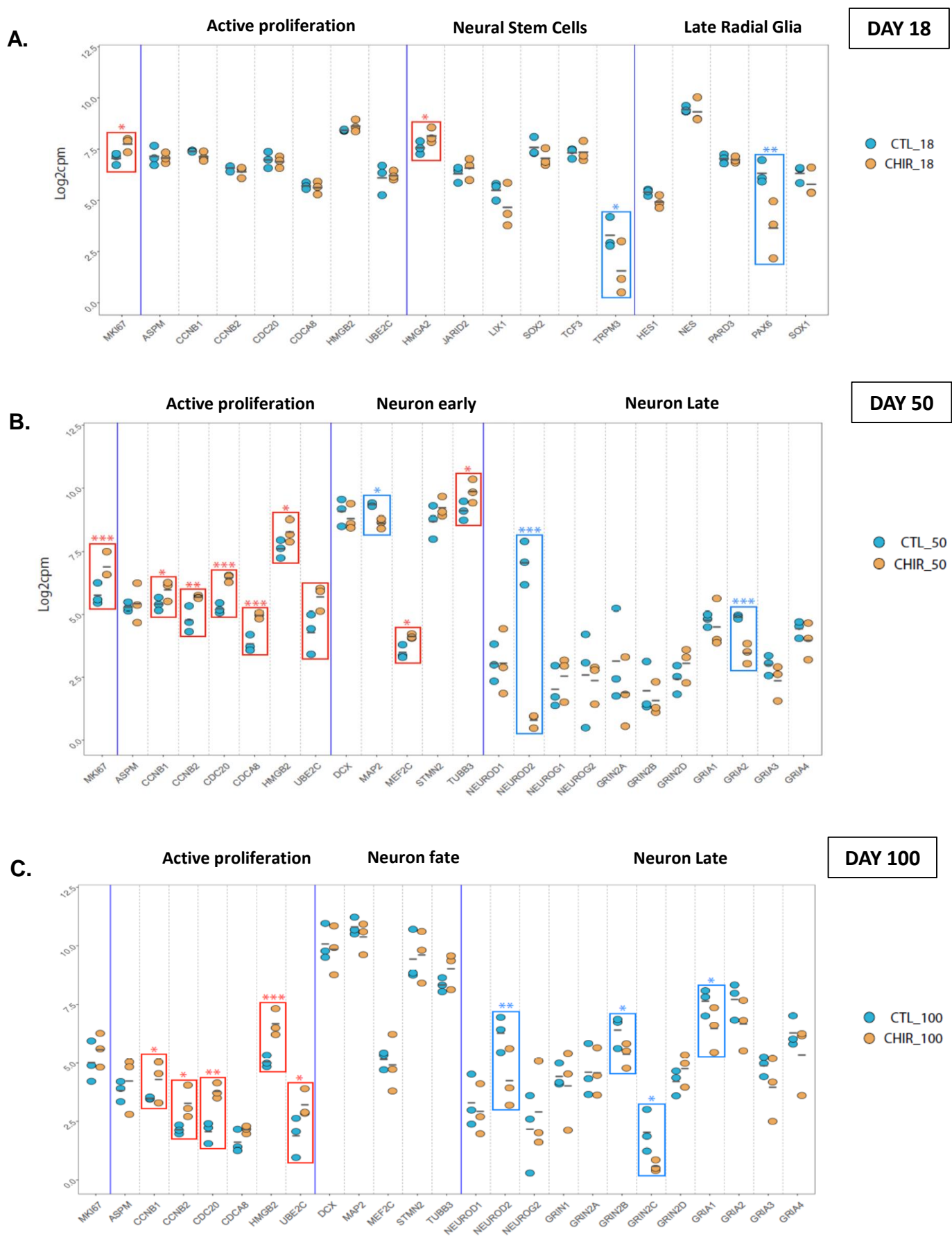
3. References related to experimental procedures

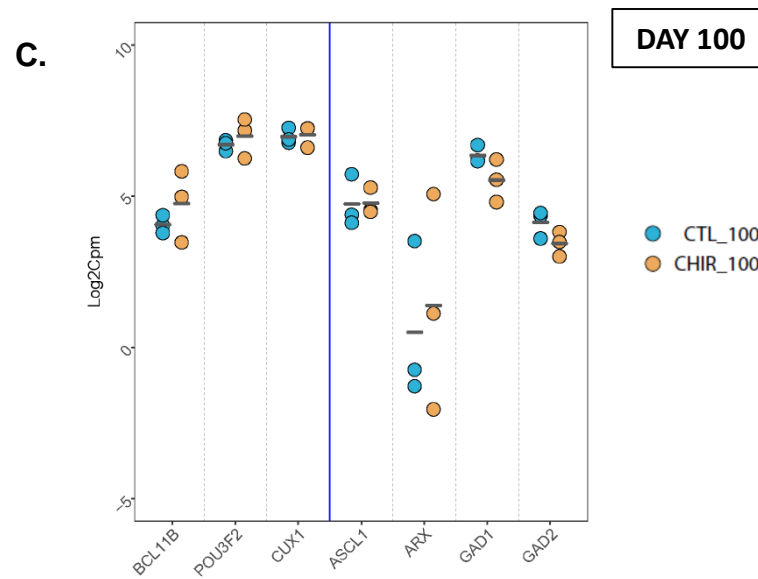
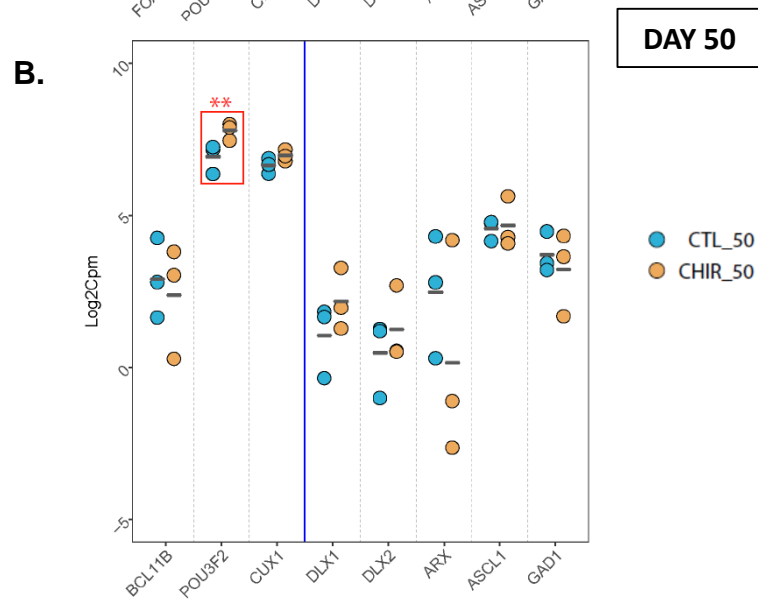
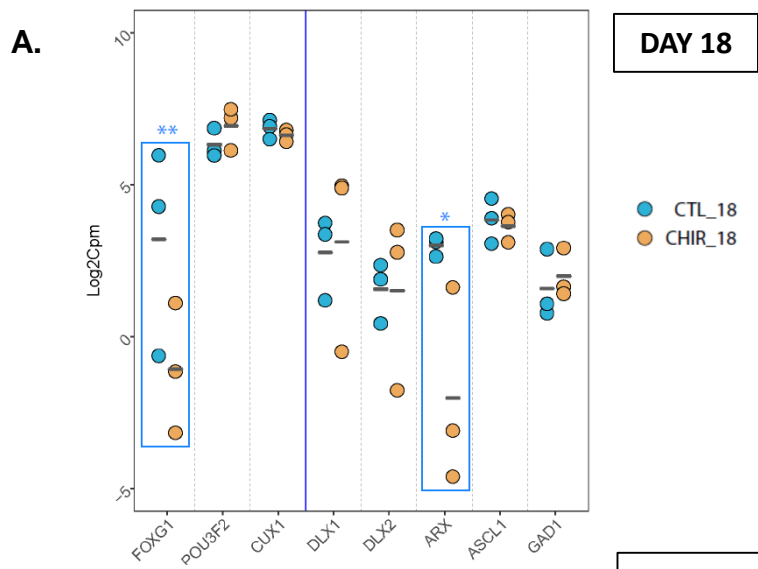
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4. Supplemental tables.

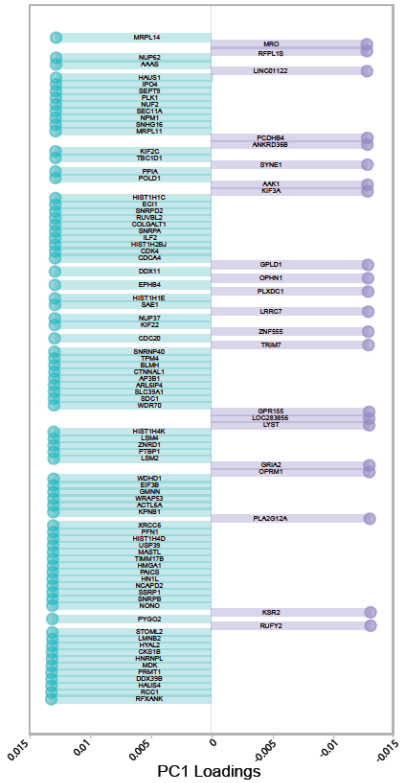
Table 2. List of differentially expressed genes.

Table 3. List of GSEA categories (Excel files).

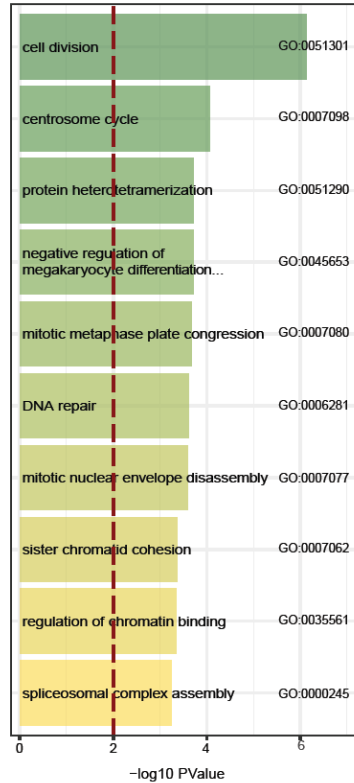




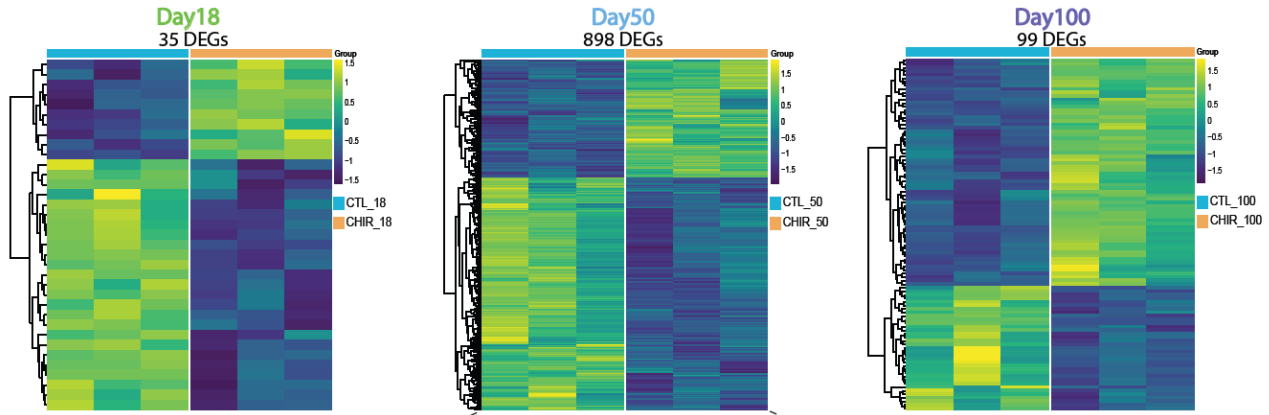
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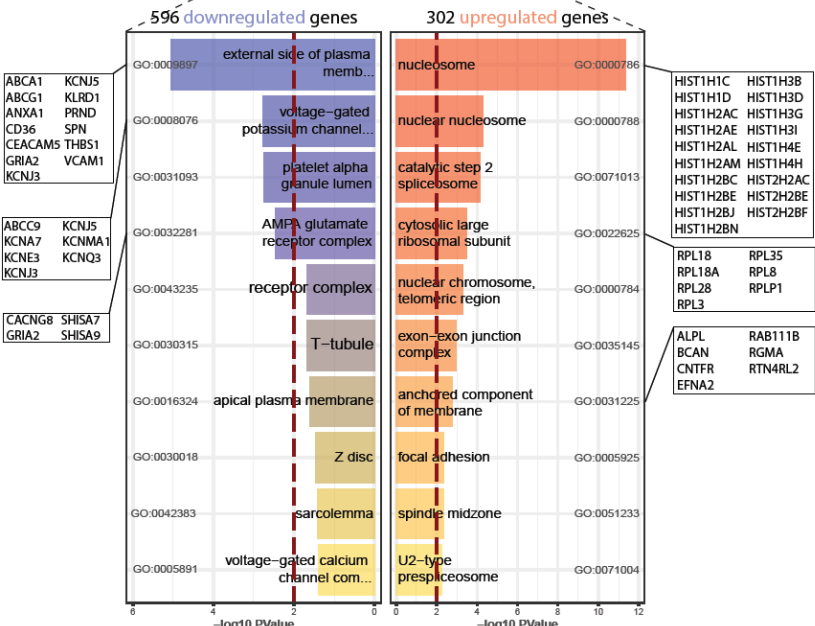
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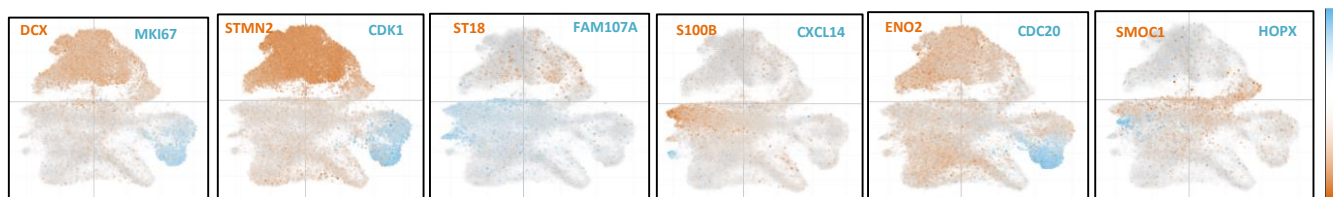
C.



D.



A



B

