UC Berkeley UC Berkeley Previously Published Works

Title

Simultaneous single-nucleus RNA sequencing and single-nucleus ATAC sequencing of neuroblastoma cell lines.

Permalink <https://escholarship.org/uc/item/54w8c5sf>

Journal Scientific Data, 11(1)

Authors

Guyer, Richard Mueller, Jessica Picard, Nicole [et al.](https://escholarship.org/uc/item/54w8c5sf#author)

Publication Date

2024-11-07

DOI

10.1038/s41597-024-04061-x

Peer reviewed

scientific data

Check for updates

Simultaneous single-nucleus RNA OPENDATA DESCRIPTOR Sequencing and single-nucleus ATAC sequencing of neuroblastoma cell lines

Richard A.Guyer 1,2 ✉**, Jessica L. Mueller2, Nicole Picard² & Allan M.Goldstein²** ✉

Neuroblastoma is the most common extracranial solid tumor in children, and a leading cause of childhood cancer deaths. All neuroblastomas arise from neural crest-derived sympathetic neuronal progenitors, but numerous mutations, the most common of which is MYCN amplifcation, give rise to these lesions. Epigenetic aberrations also play a role in oncogenesis and tumor progression. To better understand biologic diversity of neuroblastomas, we performed joint single-nucleus ATAC sequencing and single-nucleus RNA sequencing on six neuroblastoma cell lines, three of which are MYCN amplifed. After standard fltering for high-quality nuclei, we obtained chromatin accessibility and transcript abundance data from 41,733 neuroblastoma tumor cells. Preliminary analysis reveals signifcant diversity in chromatin landscape and gene expression across neuroblastoma cell lines. This dataset is a valuable resource for studying the transcriptional and epigenetic mechanisms of this deadly childhood disease.

Background & Summary

Neuroblastoma is the most common extracranial solid tumor in children, with an especially high incidence among children under age 4 years¹. The disease is stratified into low-risk, intermediate-risk, and high-risk cat-egories, based on clinical and biological features. High-risk disease is fatal in over 60% of cases^{[2](#page-5-1)}. MYCN amplification is the most common mutation in neuroblastoma, and the presence of MYCN amplification is sufficient to designate a tumor as high-risk^{[3](#page-5-2)}. However, over half of high-risk lesions lack this mutation^{[4](#page-5-3)[,5](#page-5-4)}, indicating signifcant biological diversity between neuroblastoma cases. Due to the challenges of obtaining primary tumor samples, various neuroblastoma cell lines have been established from high-risk lesions and are widely used for studying tumor biology.

There are numerous publicly-available transcriptional and epigenetic datasets derived from neuroblastoma cell lines, but information regarding heterogeneity within cell lines is not captured by the bulk methods used to generate these data^{6-[8](#page-5-6)}. The advent of single-cell sequencing tools has permitted high-parameter profiling of transcript abundance and epigenetic features in many cancers⁹. Several studies have reported single-cell RNA sequencing on neuroblastoma tissue, and have demonstrated considerable transcriptional heterogeneity within tumors as well as between cases^{[10](#page-5-8)-13}. Single-cell multiomic tools have been developed that measure two or more features (such as transcript abundance and chromatin accessibility) simultaneously from individual cell[s14](#page-5-10). To our knowledge, however, multiome technology has not yet been applied to neuroblastoma.

We undertook the present study to better understand transcriptional and epigenetic diversity of human neuroblastoma cell lines, both within and between samples. We utilized six widely-studied neuroblastoma cell lines, features of which are displayed in Table [1,](#page-2-0) including MYCN amplifcation status and whether each line is known to have adrenergic or mesenchymal transcriptional circuitry^{[15](#page-5-11)[,16](#page-5-12)}. One of these lines, SH-SY5Y, is a subclone derivative of another, SK-N-SH. By jointly quantifying transcript abundance and chromatin accessibility in individual nuclei, we provide a resource for studying the complex regulation of tumor cell phenotype. We anticipate

1 Division of Pediatric Surgery, Department of Surgery, Johns Hopkins University School of Medicine, Baltimore, MD, USA. ²Department of Pediatric Surgery, Massachusetts General Hospital, Boston, MA, USA. [⊠]e-mail: [rguyer2@](mailto:rguyer2@jh.edu) [jh.edu;](mailto:rguyer2@jh.edu) agoldstein@mgb.org

Table 1. Characteristics of cell lines studied.

Fig. 1 Quality control analysis of single-nucleus datasets. (**A**) Violin plot showing RNA counts in each cell in the snRNA-seq dataset, with the cell line indicated on the X-axis. (**B**) Violin plot showing unique RNA features identifed in each cell in the snRNA-seq dataset, with the cell line indicated on the X-axis. (**C**) Violin plot showing transcriptional start site enrichment in each cell in the snATAC-seq dataset, with the cell lines indicated on the X-axis. (**D**) Violin plot showing nucleosome signal in each cell in the snATAC-seq dataset, with the cell lines indicated on the X-axis. (**E**) Violin plot showing ATAC counts in MACS2-identifed peaks in each cell in the snATAC-seq dataset, with the cell lines indicated on the X-axis.

these data will be extremely useful for identifying the driver genes, transcriptional regulatory networks, and transcription factor-chromatin interactions underlying neuroblastoma cell states.

Methods

Cell lines and culture. Cell lines used in this study were purchased from ATCC (Manassas, VA). All cell growth and preparation for sequencing was done at Massachusetts General Hospital. Cells were cultured under conditions recommended by ATCC. All cells were maintained in incubators a 37 °C and 5% CO_2 . Cells were passaged when they reached approximately 80% confuency. In all cases, cells were prepared for sequencing at a passage number less than 10.

Fig. 2 Basic analysis of snRNA-seq data. (**A**) UMAP projection showing cells from each cell line largely cluster together based on transcript abundance. (**B**) Heatmap showing mean expression in each cell line of the top 250 marker genes identifed for each cell line. (**C**) Transcript abundance of *MYCN* within each cell, overlayed on the UMAP projection shown in (**A**). As anticipated, *MYCN*-amplifed cell lines have markedly higher *MYCN* transcript levels. (**D**) Transcript abundance of *PHOX2B* within each cell, overlayed on the UMAP projection shown in (**A**). In contrast with *MYCN*, *PHOX2B* transcripts are abundant in all cell lines.

Sample preparation. Nuclei were isolated from individual cells using the 10X Genomics (Pleasanton, CA) Demonstrated Protocol CG000365: Nuclei Isolation for Single Cell Multiome ATAC + Gene Expression Sequencing. Briefy, cells were washed with cold PBS and trypsinized to a single-cell suspension. Afer pelleting at 300 rcf in a tabletop centrifuge in 15mL conical tubes, cells were washed twice in 1mL cold PBS supplemented with 0.04% BSA. Cells were then passed through a 40μm strainer and counted using a standard hemocytometer. A total of 1,000,000 cells were transferred to 2 mL microcentrifuge tubes. Cells were resuspended in 100 mL of ice-cold Lysis Buffer (10 mM Tris-HCl at pH 7.4, 10 mM NaCl, 3 mM MgCl₂, 0.1% Tween-20, 0.1% NP-40, 0.01% digitonin, 1% BSA, 1mM DTT, 1U/mL RNase inhibitors, all in nuclease-free water) and incubated for 4minutes on ice, followed by addition of 1 mL of Wash Buffer (10 mM Tris-HCl at pH 7.4, 10 mM NaCl, 3 mM MgCl₂, 1%) BSA, 0.1% Tween-20, 1mM DTT, and 1U/mL RNase inhibitors in nuclease-free water). Cells were centrifuged at 500 rcf for 5 minutes at 4 °C to repellet. A total of 3 washes in Wash Bufer were performed. Cells were then suspended in 1mL of 1x Nuclei Bufer (provided at 20x concentration by 10X Genomics in Chromium Next GEM Single Cell Multiome ATAC Kit A, PN-1000280) supplemented with 1mM DTT and 1U/mL RNase inhibitors. GEM generation and cell barcoding were immediately performed using 10X Genomics Chromium Controller and 10X Genomics Next GEM Chip J. ATAC and gene expression library construction was performed per the 10X Genomics Next GEM Single Cell Multiome ATAC+Gene Expression User Guide, with reagents purchased from 10X Genomics. Sequencing was performed on the Illumina NovaSeq platform at the Harvard University Bauer Core facility.

Data analysis. ATAC and gene expression FASTQ output files were demultiplexed and mapped to the Genome Refence Consortium hg38 reference genome with Cell Ranger ARC (10X Genomics) sofware on the Harvard University Bauer Core's computing cluster. The resulting fragment files and count matrices were processed further using Signac version 1.1.0 and Seurat version 4.3.0.1, and implemented in R version 4.3.2 in the Rstudio computing environment version 2023.03.1+446. High-quality nuclei were selected based on the following criteria: ATAC counts>2500 and <1000000, RNA counts>2000 and <10000, nucleosome signal <2, transcriptional start site enrichment >1 , and mitochondrial RNA $<$ 5% of total transcripts. Table [1](#page-2-0) shows the number of cells from each line retained afer qualiy fltering. ATAC peak calling was performed using the Signac "CallPeaks" function and MACS2 version 2.2.7.1. The standard Seurat and Signac workflows were then used to analyze data. Marker peaks for each cell line with the ATAC data were identifed using the Seurat "FindAllMarkers" function with default arguments, except for the following: only.pos = "TRUE", test.use = "LR", and latent.vars = "nCount_ATAC". Similarly, marker genes for each cell line were identifed with the same function, with the following adjustments to default arguments: only.pos = "TRUE", logfc.threshold = 0.5 , and min.pct = 0.2 .

Fig. 3 Basic analysis of snATAC-seq data. (**A**) UMAP projection showing cells from each cell line largely cluster together based on chromatin accessibility. (**B**) Heatmap showing mean expression in each cell line of the top 1000 marker peaks identifed for each cell line. (**C,E**) Track plots showing ATAC counts in and around the *MYCN* and *PHOX2B* coding sequences in the indicated cell lines. (**D,F**) ATAC count abundance within each cell for the indicated MACS2 peaks, which lie within the MYCN (**D**) and PHOX2B (**F**) coding regions, overlayed on the UMAP projection shown in (**A**).

Data Records

Raw FASTQ fles, count matrices in CSV fomat, and key Cell Ranger output fles (fltered gene expression feature marices and ATAC fragment fles) have been uploaded to the NCBI Gene Expression Omnibus, with accession number GSE262189[17.](#page-5-13) RDS fles containing Seurat objects derived from analyzing each cell line as described here have been uploaded to Mendeley Data, with the following DOIs and corresponding URLs: 10.17632/ wvzz6hbttg.2 ([https://data.mendeley.com/datasets/wvzz6hbttg/1\)](https://data.mendeley.com/datasets/wvzz6hbttg/1)[18,](#page-5-14)<https://doi.org/10.17632/29g4826npf.1> (<https://data.mendeley.com/datasets/29g4826npf/1>[\)19](#page-5-15), <https://doi.org/10.17632/9yc8d8bnss.1>[\(https://data.](https://data.mendeley.com/datasets/9yc8d8bnss/1) [mendeley.com/datasets/9yc8d8bnss/1](https://data.mendeley.com/datasets/9yc8d8bnss/1)[\)20](#page-5-16),<https://doi.org/10.17632/cp4d7t74vb.1>[\(https://data.mendeley.com/](https://data.mendeley.com/datasets/cp4d7t74vb/1) [datasets/cp4d7t74vb/1\)](https://data.mendeley.com/datasets/cp4d7t74vb/1)^{[21](#page-5-17)}.

Technical Validation

We evaluated standard quality control metrics for snRNA-seq and snATAC-seq datasets. Afer fltering for high-quality nuclei, as described above, our dataset consistent of 41,733 nuclei. The gene expression data showed consistency of counts per nucleus and unique features per nucleus across the six cell lines studied (Fig. [1A,B](#page-2-1)). Similarly, ATAC data was consistent across cell lines with respect to transcriptional start site enrichment, nucleosome signal, and the number of counts in peaks (Fig. [1C–E](#page-2-1)).

We performed basic dimensional reduction, clustering, and analysis using transcriptional data. As expected, there was more variation between cell lines that within cell lines, as demonstrated on a UMAP projection (Fig. [2A](#page-3-0)). A subset of SK-N-SH cells does appear closely related to SH-SY5Y cells (Fig. [2A\)](#page-3-0), which may be expected since SH-SY5Y cells are a subclose of the SK-N-SH line. We identifed the top 250 marker genes for each cell line, relative to all other cell lines. Hierarchical clustering based on scaled expression of these 250 genes shows that the non-*MYCN*-amplifed cell lines (SHSY5Y, SK-N-SH, and SK-N-AS) were most similar to one another, with SK-N-SH and SH-SY5Y being most similar, while the *MYCN*-amplifed cell lines (CHP134, Be2c, and SK-N-DZ) diverged from the non-amplifed lines (Fig. [2B\)](#page-3-0). Next, we assessed gene expression of two important neuroblastoma driver genes: *MYCN* and *PHOX2B*. As anticipated, our data show markedly higher expression of *MYCN* transcripts in the *MYCN*-amplifed cell lines (Fig. [2C\)](#page-3-0), while *PHOX2B* transcripts are more uniformly distributed across all six cell lines (Fig. [2D](#page-3-0)).

Similarly, we assessed the chromatin accessibility dataset in each cell line. Just as for gene expression data, we found that the greatest variability was between cell lines, as visualized on a UMAP projection (Fig. [3A\)](#page-4-0). The top 1000 marker peaks for each cell line were identifed and used to generate a heatmap (Fig. [3B](#page-4-0)). Hierarchical clustering of this data shows that the three non-MYCN-amplifed lines cluster together and maintain open chromatin at similar genomic regions. Again, both the UMAP projection and hierarchical clustering of the heatmap suggest a close relationship between the SH-SY5Y and SK-N-SH lines (Fig. [3A,B\)](#page-4-0). In contrast, MYCN-amplifed lines show more divergence in their chromatin accessibility patterns. Due to MYCN amplifcation, we expect the CHP134, Be2c, and SK-N-DZ cell lines to have far more counts mapped to the MYCN locus than the non-amplifed lines. A tracks plot of the MYCN gene shows these results (Fig. [3C](#page-4-0)). Similarly, counts within the ATAC peak found at chromosome 2, bases 15,939,296-15,943,697, which lies within the *MYCN* coding sequence, are seen exclusively in the MYCN-amplifed lines (Fig. [3D\)](#page-4-0). In contrast, there is a similar chromatin signal in both the coding sequence and upstream promoter region of PHOX2B, including at the peak at chromosome 4, bases 41,748,364-41,749,127 (Fig. [3E,F](#page-4-0)).

Code availability

R scripts use for processing and analysis of this data are publicly available at a Mendeley Data repository with https://doi.org/10.17632/s2fcfb8phh.1 (https://data.mendeley.com/datasets/s2fcfb8phh/1)^{[22](#page-5-18)}.

Received: 10 April 2024; Accepted: 1 November 2024; Published online: 07 November 2024

References

- 1. Li, J., Tompson, T. D., Miller, J. W., Pollack, L. A. & Stewart, S. L. Cancer Incidence Among Children and Adolescents in the United States, 2001–2003. *Pediatrics* **121**, e1470–e1477 (2008).
- 2. Newman, E. A. *et al*. Update on neuroblastoma. *Journal of Pediatric Surgery* **54**, 383–389 (2019).
- 3. Cohn, S. L. *et al*. Te International Neuroblastoma Risk Group (INRG) Classifcation System: An INRG Task Force Report. *JCO* **27**, 289–297 (2009).
- 4. Lee, J. W. *et al*. Clinical signifcance of *MYCN* amplifcation in patients with high-risk neuroblastoma. *Pediatr Blood Cancer* **65**, e27257 (2018).
- 5. Yanishevski, D. *et al*. Impact of MYCN status on response of high-risk neuroblastoma to neoadjuvant chemotherapy. *Journal of Pediatric Surgery* **55**, 130–134 (2020).
- 6. Guyer, R. A. *et al*. Diferentiated neuroblastoma cells remain epigenetically poised for de-diferentiation to an immature state. *Disease Models & Mechanisms* **16**, dmm049754 (2023).
- 7. Upton, K. *et al*. Epigenomic profling of neuroblastoma cell lines. *Sci Data* **7**, 116 (2020).
- 8. Harenza, J. L. *et al*. Transcriptomic profling of 39 commonly-used neuroblastoma cell lines. *Sci Data* **4**, 170033 (2017).
- 9. Kashima, Y. *et al*. Single-cell sequencing techniques from individual to multiomics analyses. *Exp Mol Med* **52**, 1419–1427 (2020).
- 10. Dong, R. *et al*. Single-Cell Characterization of Malignant Phenotypes and Developmental Trajectories of Adrenal Neuroblastoma. *Cancer Cell* **38**, 716–733.e6 (2020).
- 11. Mercatelli, D. *et al*. Single-Cell Gene Network Analysis and Transcriptional Landscape of MYCN-Amplifed Neuroblastoma Cell Lines. *Biomolecules* **11**, 177 (2021).
- 12. Jansky, S. *et al*. Single-cell transcriptomic analyses provide insights into the developmental origins of neuroblastoma. *Nature Genetics* **53**, 683–693,<https://doi.org/10.1038/s41588-021-00806-1>(2021).
- 13. Kildisiute, G. *et al*. Tumor to normal single-cell mRNA comparisons reveal a pan-neuroblastoma cancer cell. *Science Advances* **7**, eabd3311 (2021).
- 14. Ma, S. *et al*. Chromatin Potential Identifed by Shared Single-Cell Profling of RNA and Chromatin. *Cell* **183**, 1103–1116.e20 (2020). 15. Boeva, V. *et al*. Heterogeneity of neuroblastoma cell identity defned by transcriptional circuitries. *Nature Genetics* **49**, 1408–1413, <https://doi.org/10.1038/ng.3921>(2017).
- 16. van Groningen, T. *et al*. Neuroblastoma is composed of two super-enhancerassociated diferentiation states. *Nature Genetics* **49**, 1261–1266, <https://doi.org/10.1038/ng.3899>(2017).
- 17. Guyer, R. A. & Goldstein, A. M. *GEO* [https://identifers.org/geo/GSE262189](https://identifiers.org/geo/GSE262189) (2024).
- 18. Guyer, R. A. & Goldstein, A. M. Seurat objects for multiome analysis of neuroblastoma cell lines - 1/4. *Mendeley Data* V1 [https://doi.](https://doi.org/10.17632/wvzz6hbttg.1) [org/10.17632/wvzz6hbttg.1](https://doi.org/10.17632/wvzz6hbttg.1) (2024).
- 19. Guyer, R. A. & Goldstein, A. M. Seurat objects for multiome analysis of neuroblastoma cell lines - 2/4. *Mendeley Data* V1 [https://doi.](https://doi.org/10.17632/29g4826npf.1) [org/10.17632/29g4826npf.1](https://doi.org/10.17632/29g4826npf.1) (2024).
- 20. Guyer, R. A. & Goldstein, A. M. Seurat objects for multiome analysis of neuroblastoma cell lines - 3/4. *Mendeley Data* V1 [https://doi.](https://doi.org/10.17632/9yc8d8bnss.1) [org/10.17632/9yc8d8bnss.1](https://doi.org/10.17632/9yc8d8bnss.1) (2024).
- 21. Guyer, R. A. & Goldstein, A. M. Seurat objects for multiome analysis of neuroblastoma cell lines - 4/4. *Mendeley Data* V1 [https://doi.](https://doi.org/10.17632/cp4d7t74vb.1) [org/10.17632/cp4d7t74vb.1](https://doi.org/10.17632/cp4d7t74vb.1) (2024).
- 22. Guyer, R. A. & Goldstein, A. M. Analysis of multiome sequencing data from neuroblastoma cell lines. *Mendeley Data* V1 [https://doi.](https://doi.org/10.17632/s2fcfb8phh.1) [org/10.17632/s2fcf8phh.1](https://doi.org/10.17632/s2fcfb8phh.1) (2024).

Acknowledgements

This work was supported by NIDDK F32DK121440 to RAG, NIDDK F32DK131792 to JLM, NIDDK R01DK119210 to AMG, and by the Massachusetts General Hospital Department of Surgery Patricia K. Donahoe Resident Research Catalyst Award to RAG. The authors thank the staff at the Harvard University Bauer Core for their expertise and assistance.

Author contributions

R.A.G. and A.M.G. conceived and designed this project. R.A.G., J.L.M. and N.B. performed cell culture, sample preparation, and library construction. R.A.G. did all data analysis and processing. R.A.G. wrote all code used for data analysis. Funding for the project was obtained by R.A.G. and A.M.G. R.A.G. drafed the manuscript and prepared fgures. All authors participated in manuscript editing, and all authors have approved the version of the manuscript submitted for publication.

Competing interests

The authors declare no competing interests.

Additional information

Correspondence and requests for materials should be addressed to R.A.G. or A.M.G.

Reprints and permissions information is available at [www.nature.com/reprints.](http://www.nature.com/reprints)

Publisher's note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional afliations.

Open Access Tis article is licensed under a Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 International License, which permits any non-commercial use, sharing, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if you modifed the licensed material. You do not have permission under this licence to share adapted material derived from this article or parts of it. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit [http://creativecommons.org/licenses/by-nc-nd/4.0/.](http://creativecommons.org/licenses/by-nc-nd/4.0/)

 $© The Author(s) 2024$