

Metacells facilitate the analysis of single-cell multiomics data

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Introduction

- Single-cell multiomics: measurement of different modalities (e.g., ATAC, RNA, proteins) in the same cell (Fig.3A)
- Precise analysis of cell-type specific transcriptional regulation (Fig.3B).

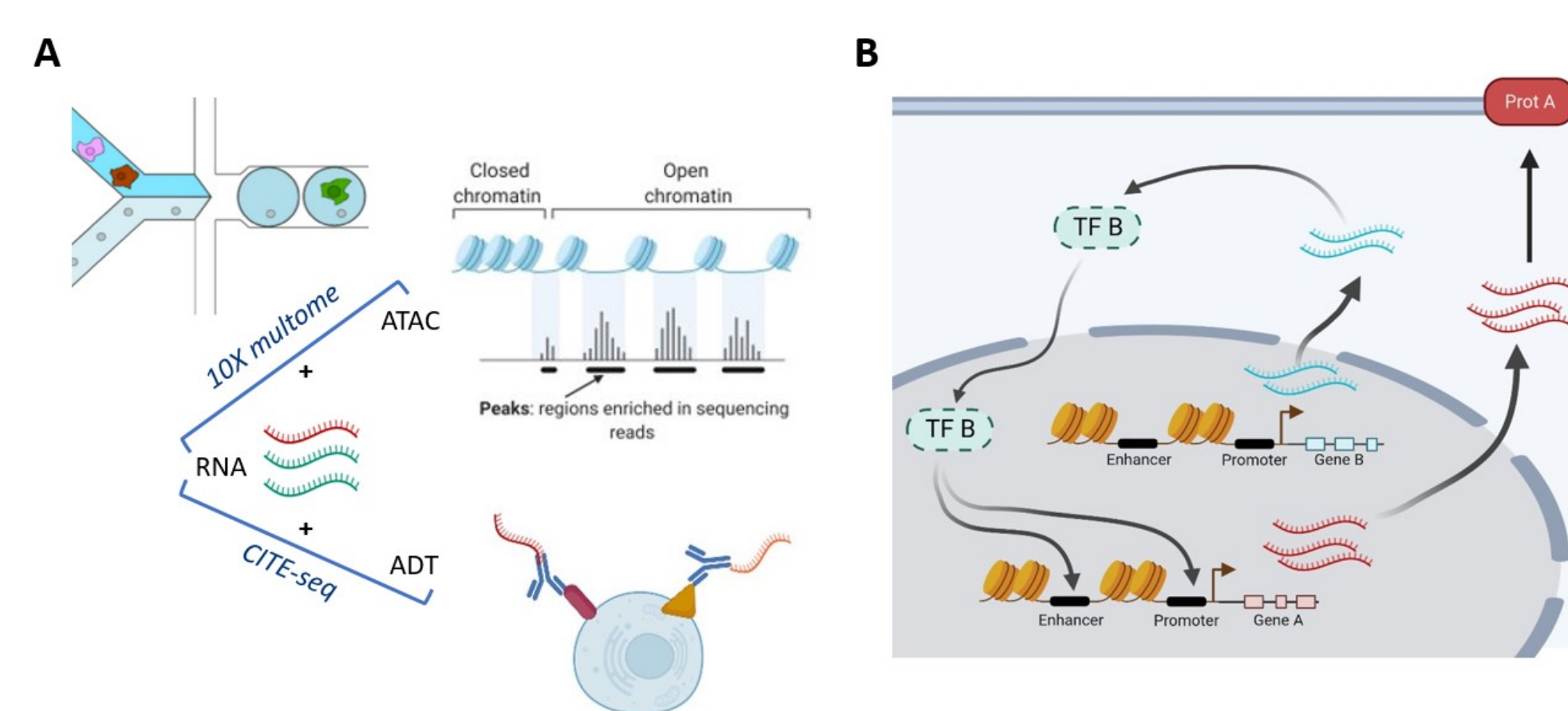


Figure 1: Single-cell multiomics (A) can be used to study cell-type specific transcriptional regulation (B)

- Limitations:** Large size, high sparsity of the data.
- Solution:** Merging highly similar cells in metacells, proposed for scRNA-seq (Baran et al. 2019).
- Aim:** Extension of SuperCell (Bilous et al. 2022), to single-cell multiomics. (Fig.2B).

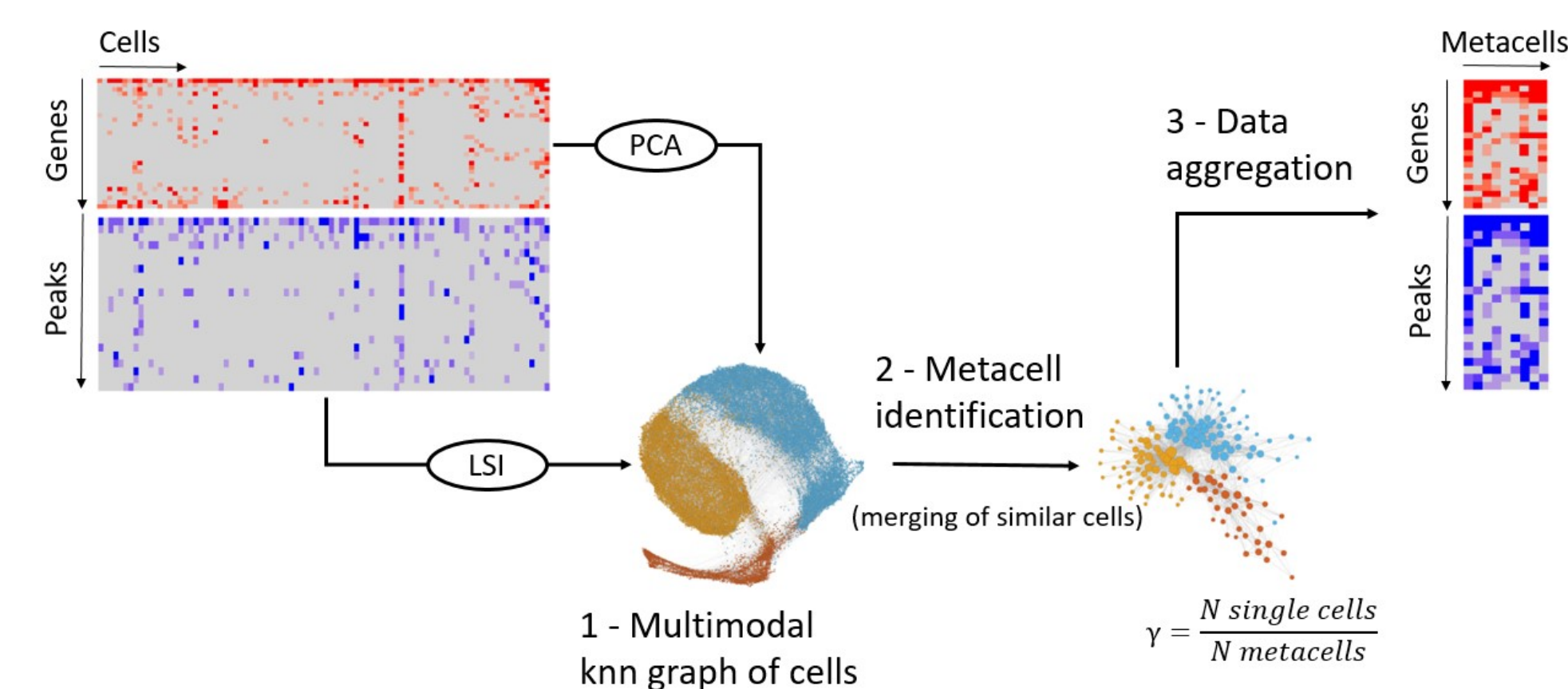


Figure 2: SuperCell workflow to identify multiomics metacells at a graining level γ . The multimodal knn graph is computed using the WNN method from Seurat (Hao et al. 2021).

Benchmark

Multimodal version of SuperCell versus unimodal tools (Fig.3):

- Purer metacells
- Compact and separated metacells in both modalities
- Faster

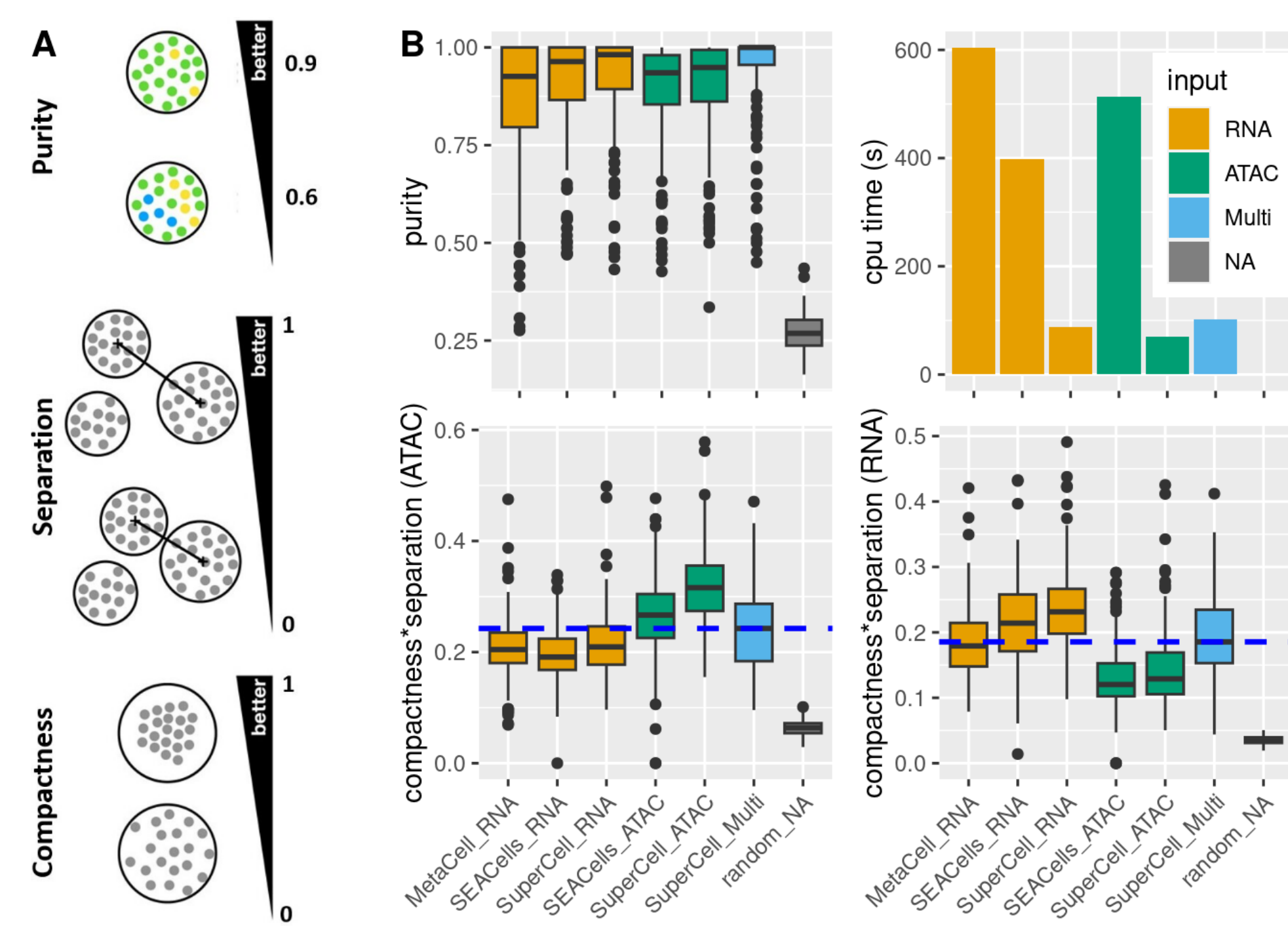


Figure 3: **A** Benchmark metrics. **B** Benchmark results of metacell tools on a 10x multiome (RNA + ATAC) dataset of PBMC, graining level $\gamma=75$. Tools: new version of SuperCell, SEACells (Persad et al. 2022), MetaCell2 (Ben-Kiki et al. 2021).

SuperCell Analyses

10X multiome dataset of PBMCs

SuperCell identifies robust metacells in the PBMC multiomic space (Fig.4).

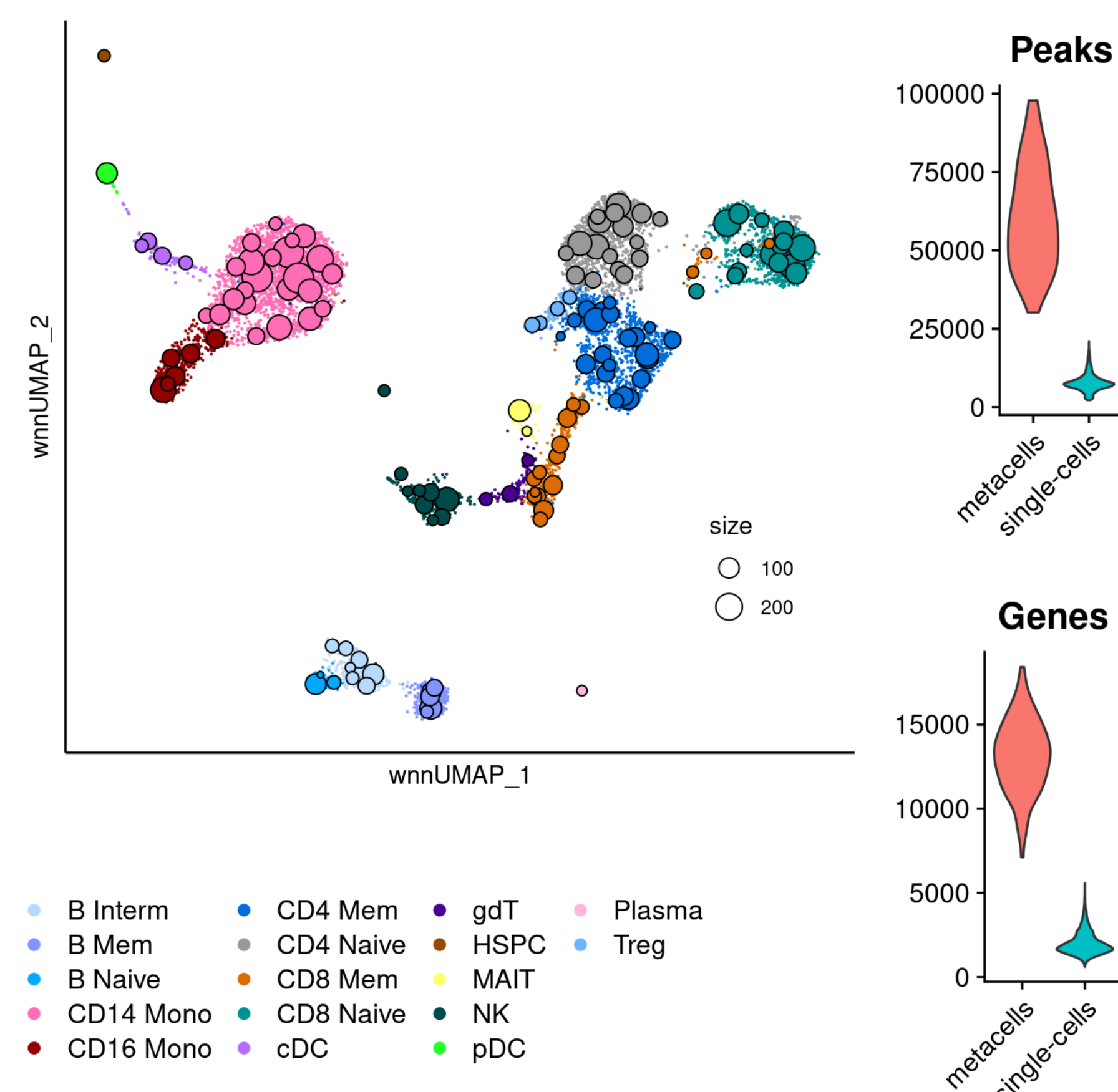


Figure 4: identified metacells in the multiomic space of PBMCs, graining level $\gamma = 75$

Gene accessibility and expression appear more correlated at the metacell level (Fig.5).

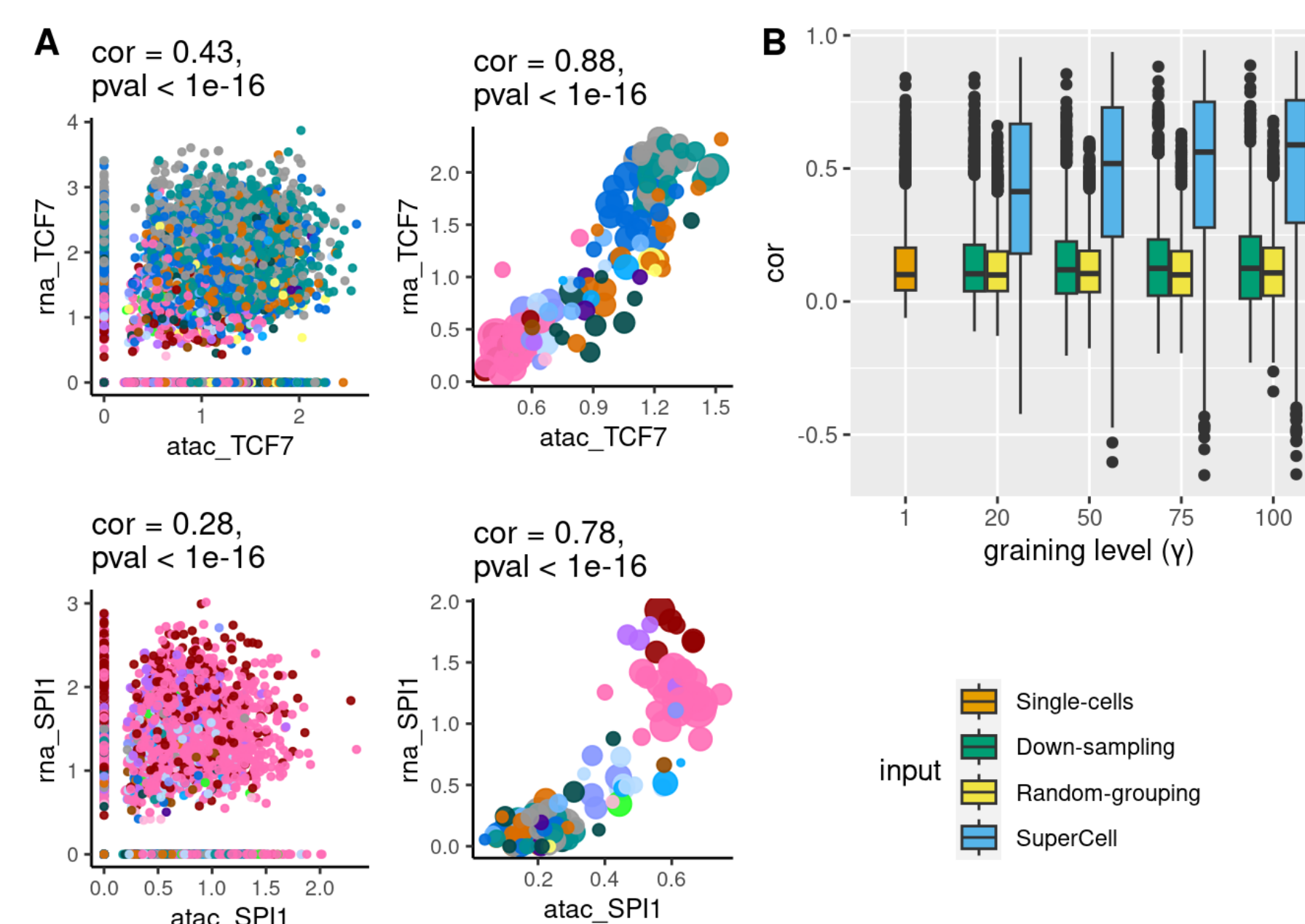


Figure 5: **A.** gene accessibility - gene expression correlation for TCF7 and SPI1. Left: Single-cell level, right: metacell $\gamma = 75$. Same color legend as in Fig.4. **B.** Same correlations for the 2000 highly variable genes (on RNA) with increasing γ .

Correlation between transcription factor (TF) expression and corresponding motif accessibility also becomes clearer using metacells (Fig.6).

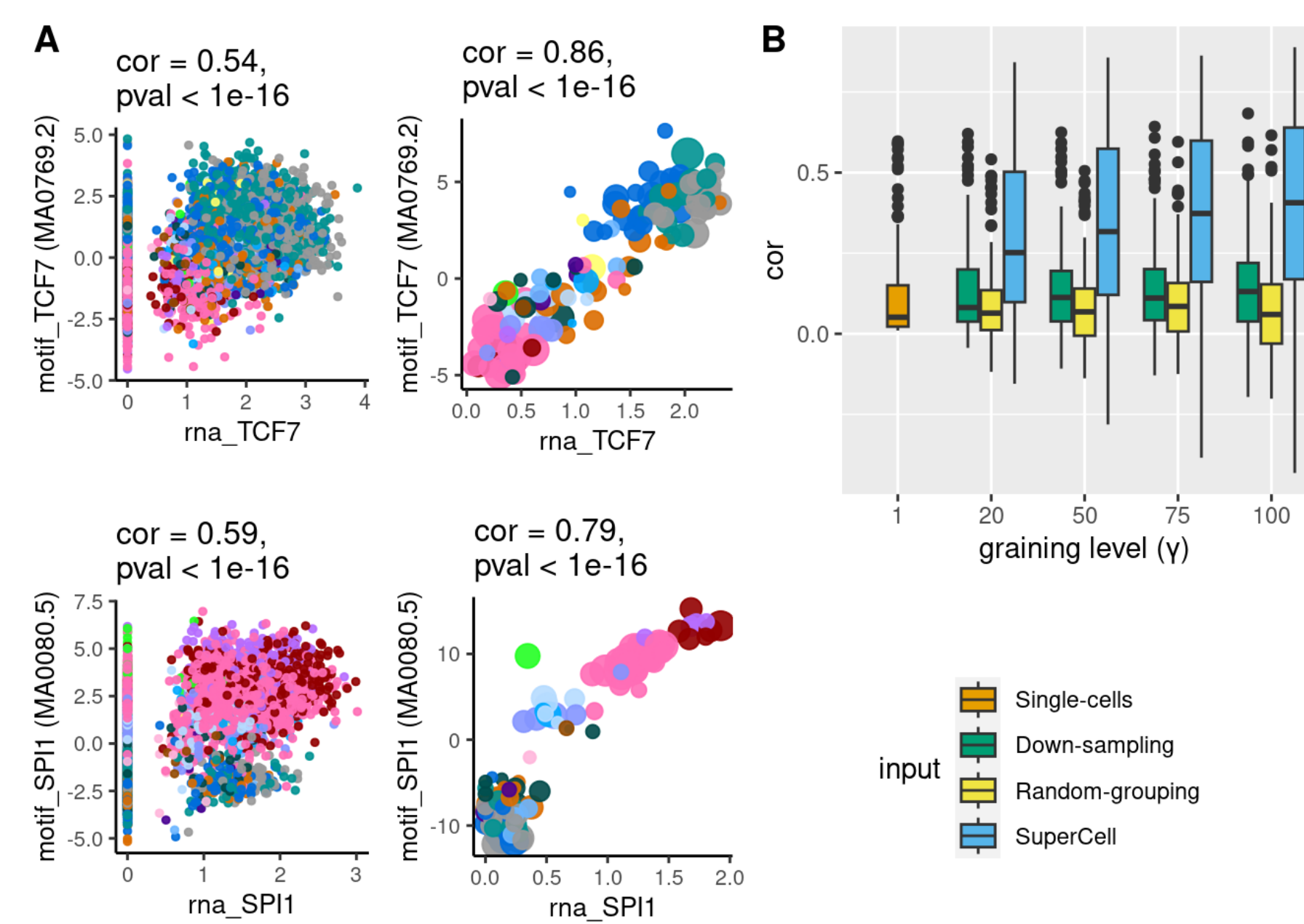


Figure 6: **A.** TF expression (RNA) - motif accessibility (ATAC) correlation for TCF7 and SPI1. Left: single-cell, right: metacells, $\gamma = 75$. Same color legend as in Fig.4. **B.** Same correlations for 200 TFs (with $\text{cor} > 0.01$ in single-cells) with increasing γ .

CITE-seq atlas of 160,000 PBMCs

- Identification of metacells by sample ($\gamma = 20$)
- Correction of the batch effect at the metacell level (Fig.7A&B).
- This workflow runs on a standard laptop (Fig.7C).

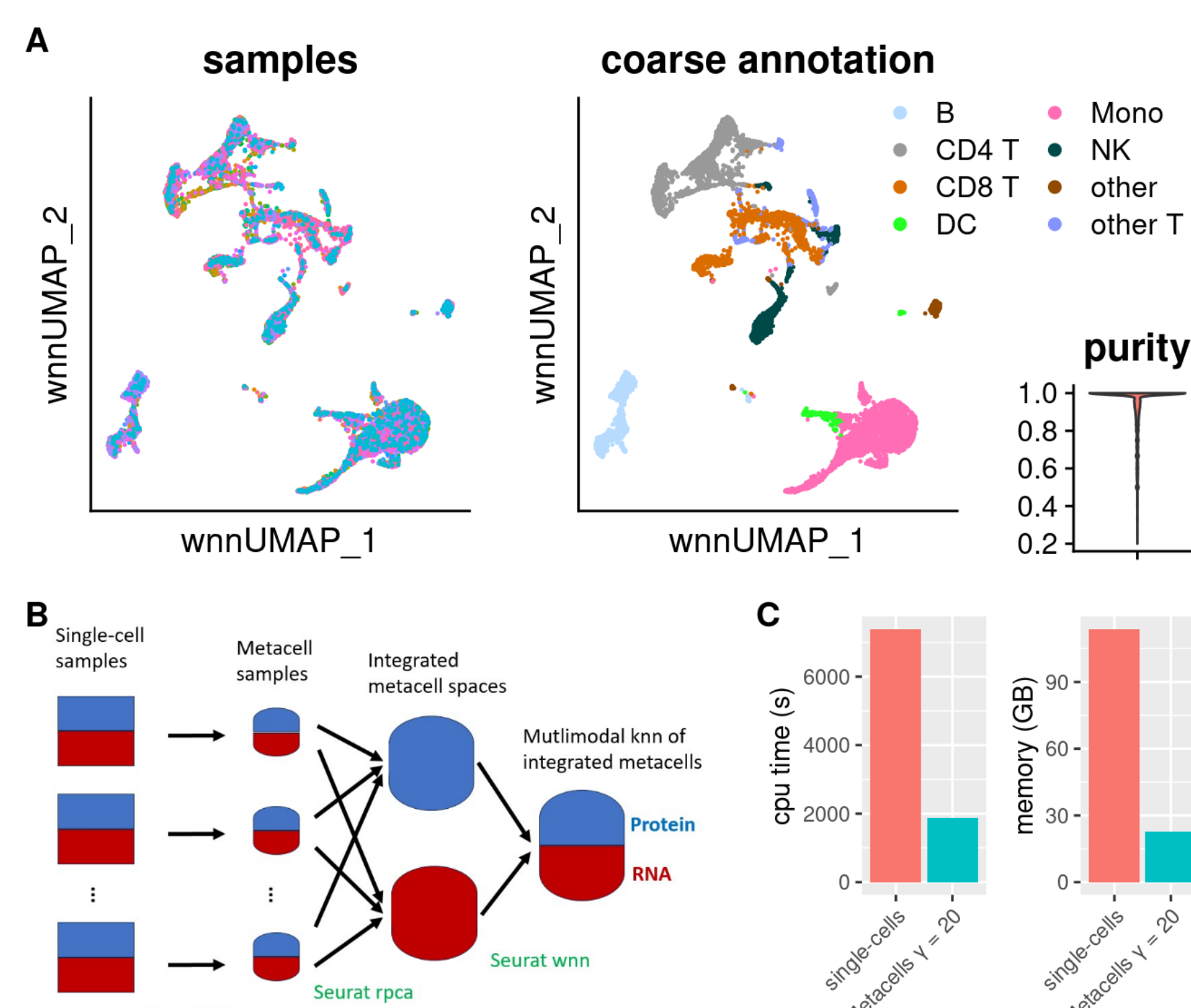


Figure 7: **A.** UMAP visualizations of 8,000 multiomic metacells, metacell purities with respect to original annotation from (Hao et al. 2021). **B.** Metacell workflow. **C.** Computational resources used by single-cell and metacell workflows.

RNA-protein correlation in the CITE-seq atlas is increased with metacells (Fig.8).

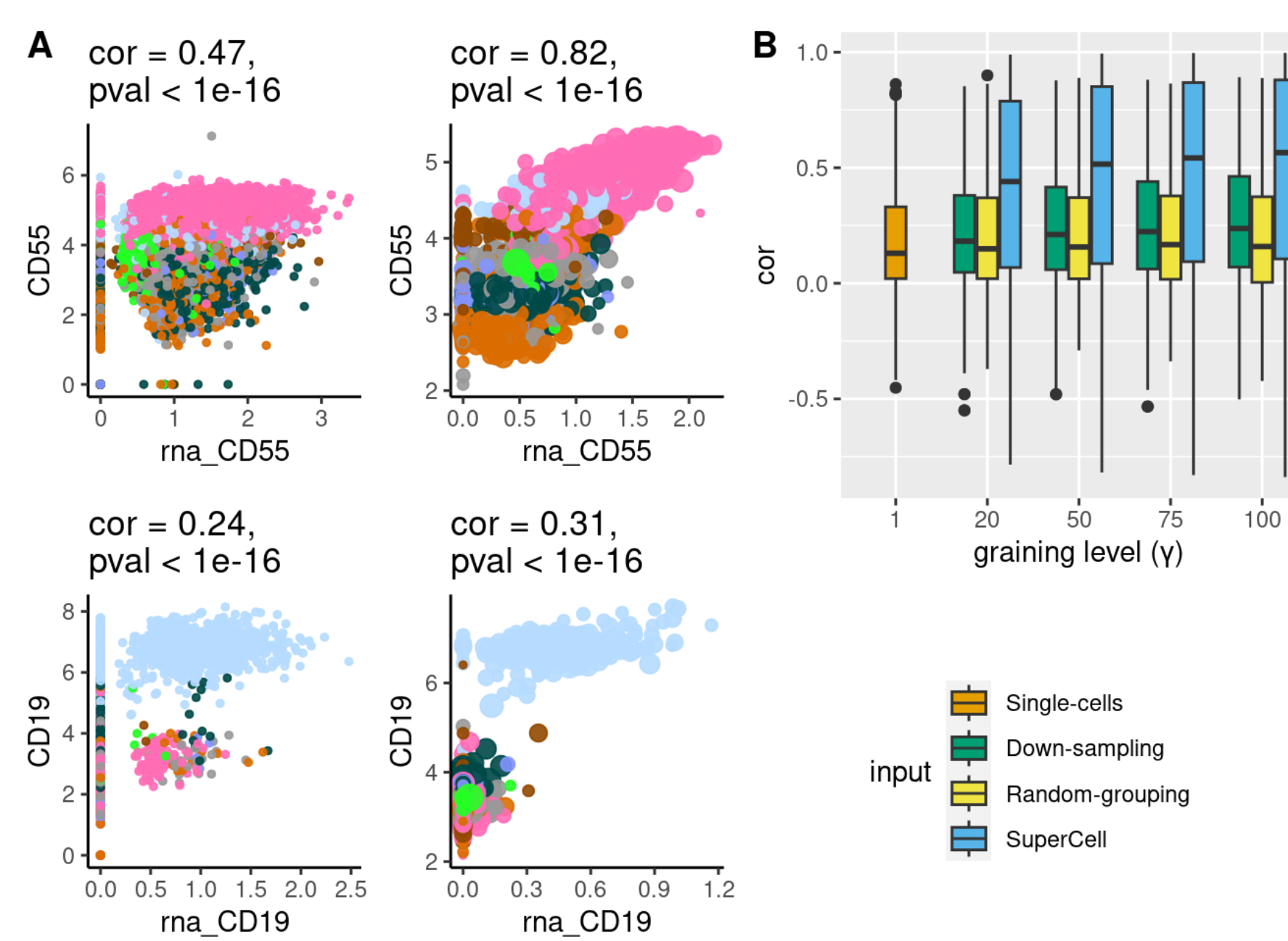
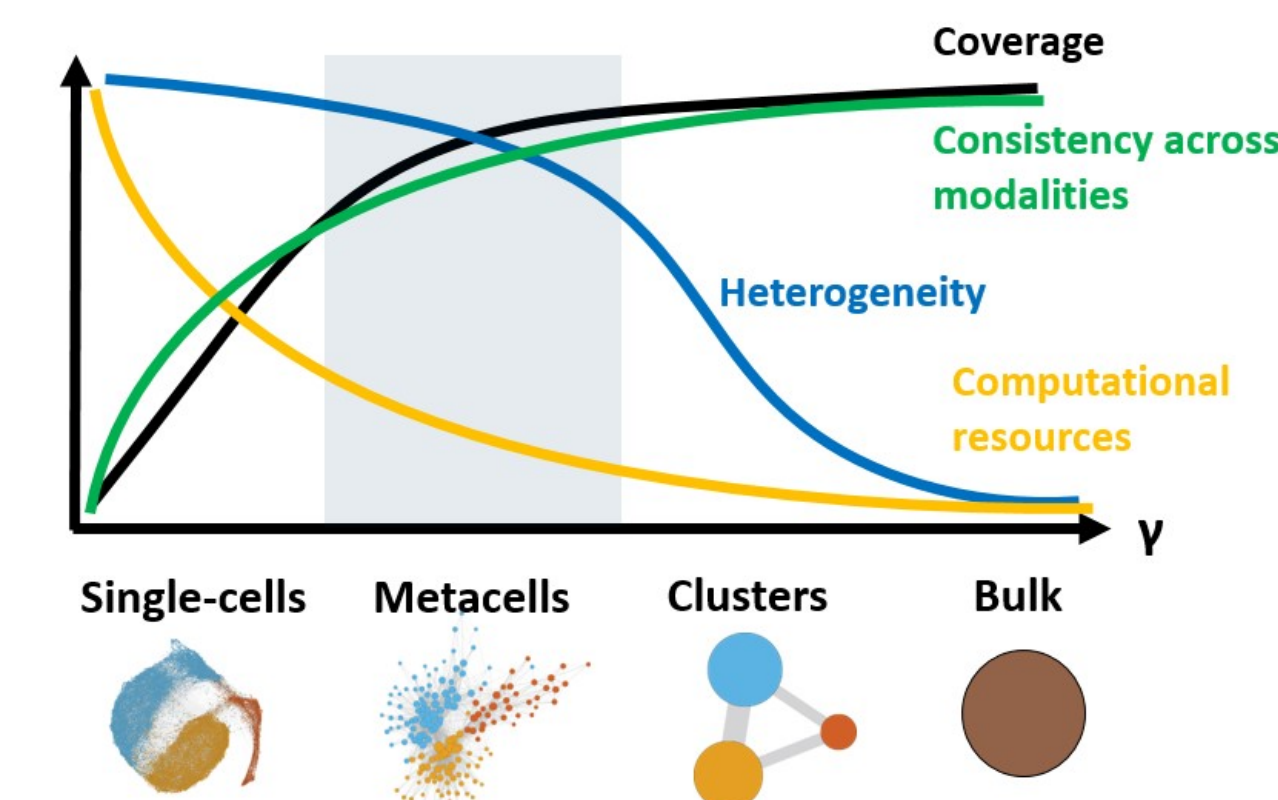


Figure 8: **A.** RNA-protein correlation for CD55 and CD19. Left: Single-cell level, right: metacell $\gamma = 75$. Same color legend as in Fig.7. **B.** RNA-Protein correlation for 213 gene-protein pairs with increasing γ

Conclusion



Perspectives

- Downstream analyses** of sc-multiomics:
 - gene regulatory network inference
 - multiomic velocity
- Metaspots** for spatial-omics

References

Baran, Yael, Akhmad Berovich, Arnan Sebe-Pelros, Yaniv Lubling, Amir Giladi, Elad Chomsky, Zohar Meir, Michael Heichman, Avitaz Lifshitz, and Amos Tanay. 2019. "Metacell: Analysis of Single-Cell RNA-Seq Data Using K-Nearest Graph Partitions." *Genome Biology* 20 (3): 206. <https://doi.org/10.1186/s13059-019-1812-2>.

Ben-Kiki, Oren, Akhmad Berovich, Avitaz Lifshitz, and Amos Tanay. 2021. "A Divide and Conquer Metacell Algorithm for Scalable scRNA-Seq Analysis." Preprint. *Bioinformatics*. <https://doi.org/10.1093/bioinformatics/btab134>.

Bilous, Mariia, Luc Tran, Chiara Giannarini, Aurélien Gabriel, Hugo Michel, Santiago J. Carmona, Mikael J. Pittet, and David Gfeller. 2022. "Metacells Untangle Large and Complex Single-Cell Transcriptome Networks." *BMC Bioinformatics* 23 (1): 336. <https://doi.org/10.1186/s12859-022-04681-1>.

Hao, Yuhai, Stephanie Hao, Erica Andersen-Nissen, William M. Mauck, Shihui Zheng, Andrew Butler, Maddy J. Lee, et al. 2021. "Integrated Analysis of Multimodal Single-Cell Data." *Cell* 184 (13): 3573–3586.e29. <https://doi.org/10.1016/j.cell.2021.04.048>.

Persad, Sitara, Zi-Ning Choo, Christine Dien, Ignas Masilionis, Ronan Chaligné, Tal Naway, Chrysostheos C. Brown, Itzik Pe'er, Manu Setty, and Dana Pe'er. 2022. "SLACell: Inference of Transcriptional and Epigenomic Cellular States from Single-Cell Genomics Data." *bioRxiv*. <https://doi.org/10.1101/2022.06.02.486764>.

