

# Machine Learning In Computational Biology

A comprehensive single cell transcriptional landscape of human hematopoietic progenitors.

Final Project Presentation



Presented by

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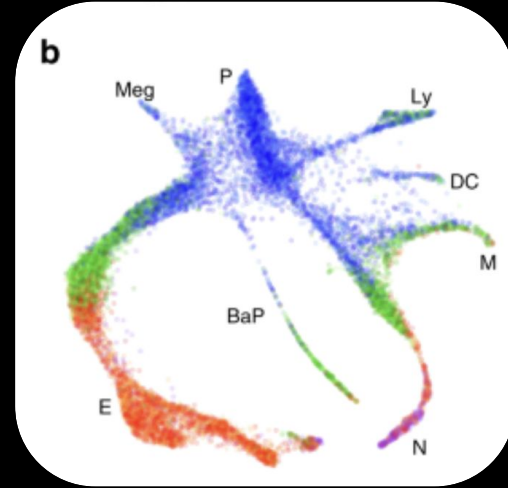
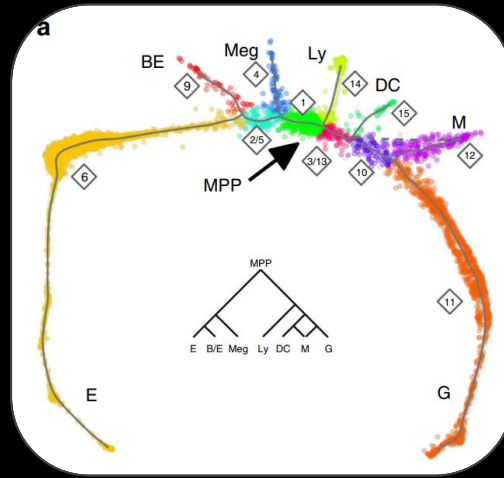
# Biological problem:

Define a comprehensive differentiation trajectory for progenitor cells of the hematopoietic lineage

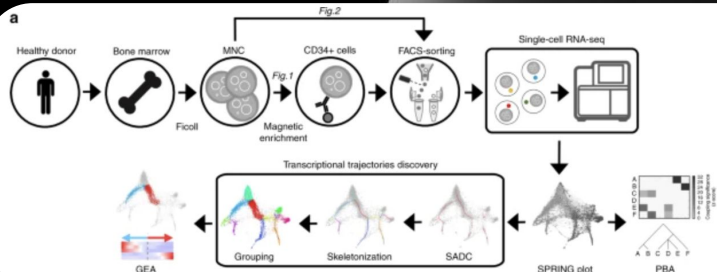
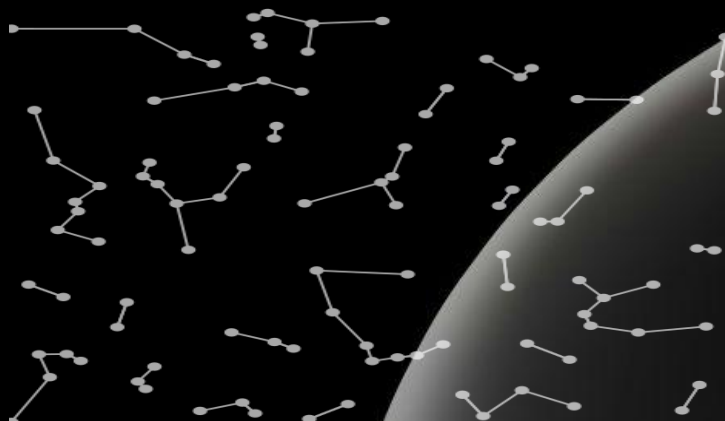
- Many different progenitors
- Focus on CD34 is not enough

## Importance:

Increase resolution regarding early differentiation by analysing whole bone marrow fraction



# Method used



Single-cell RNA-Seq on cells from the entire Bone Marrow + fractionation based on CD164

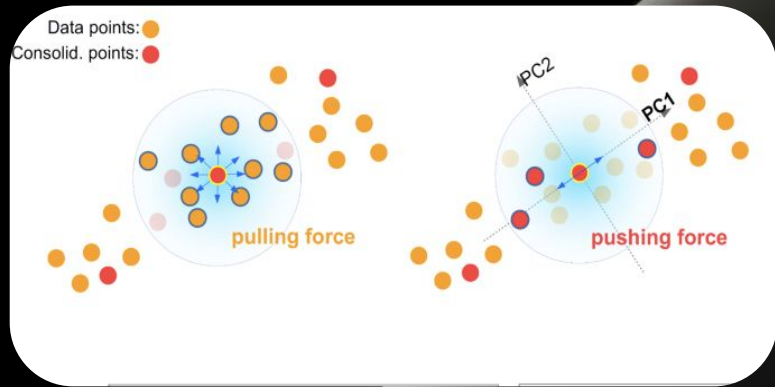


Filtering out cells based on mitochondrial gene expression threshold and normalizing expression values



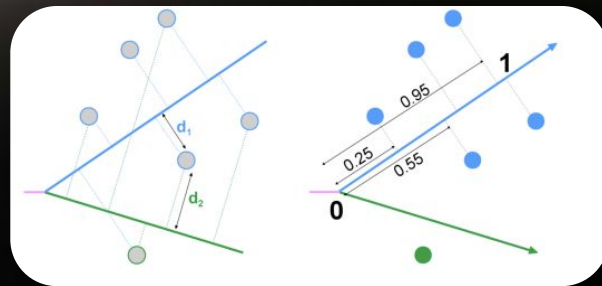
KNN graph ( $k=4$ ), branching reconstruction and ordering of cells based on the structure

# Structure-Aware filtering

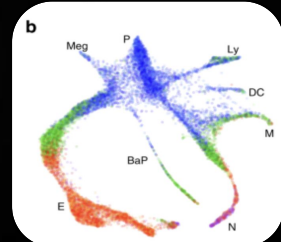
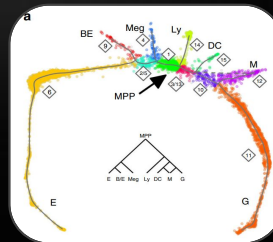


## Branch reconstruction: Minimum spanning tree

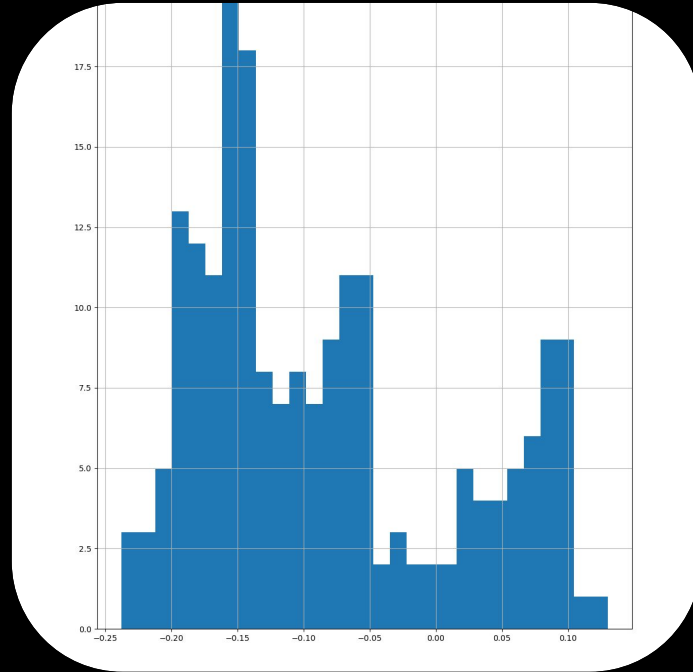
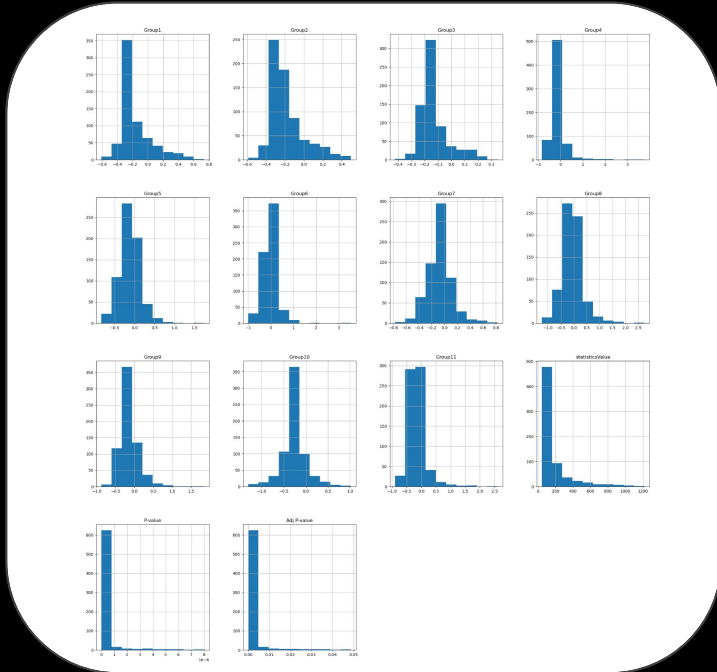
# Branch association



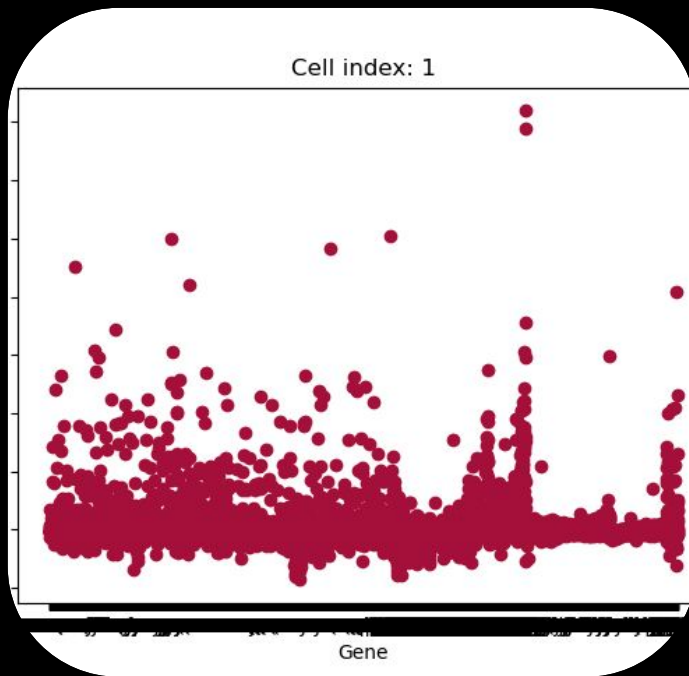
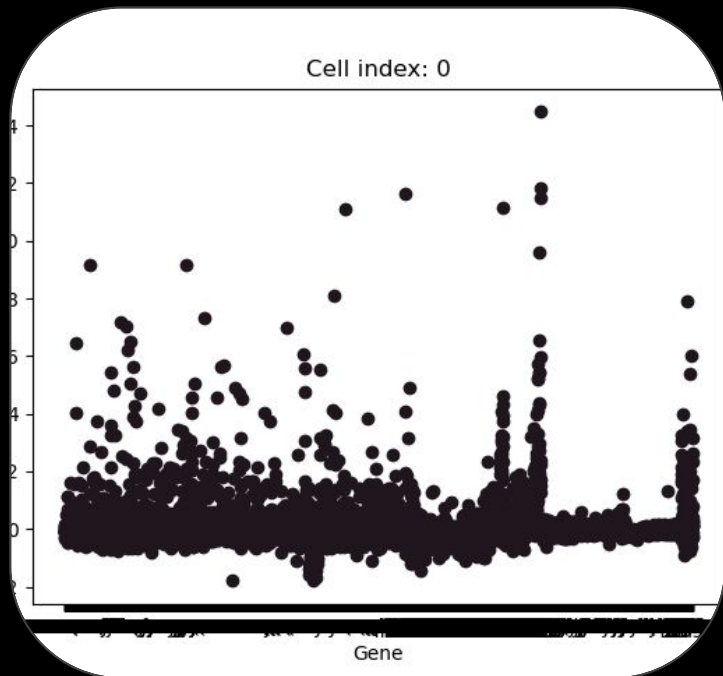
## Measure distances between branches for ordering



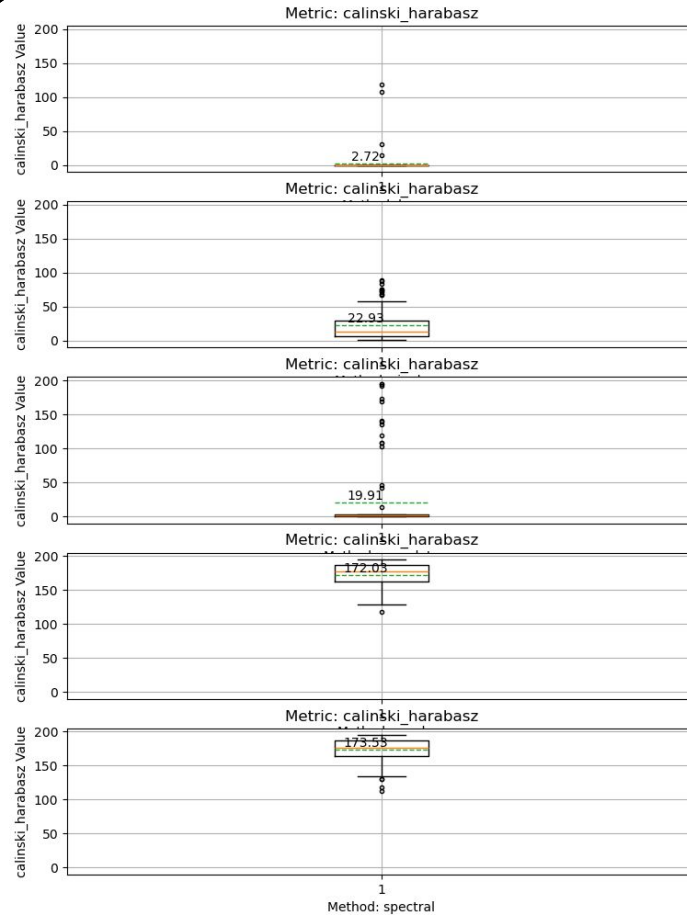
# Representing expression domains



# What the new cells look like



# Significance



KNN

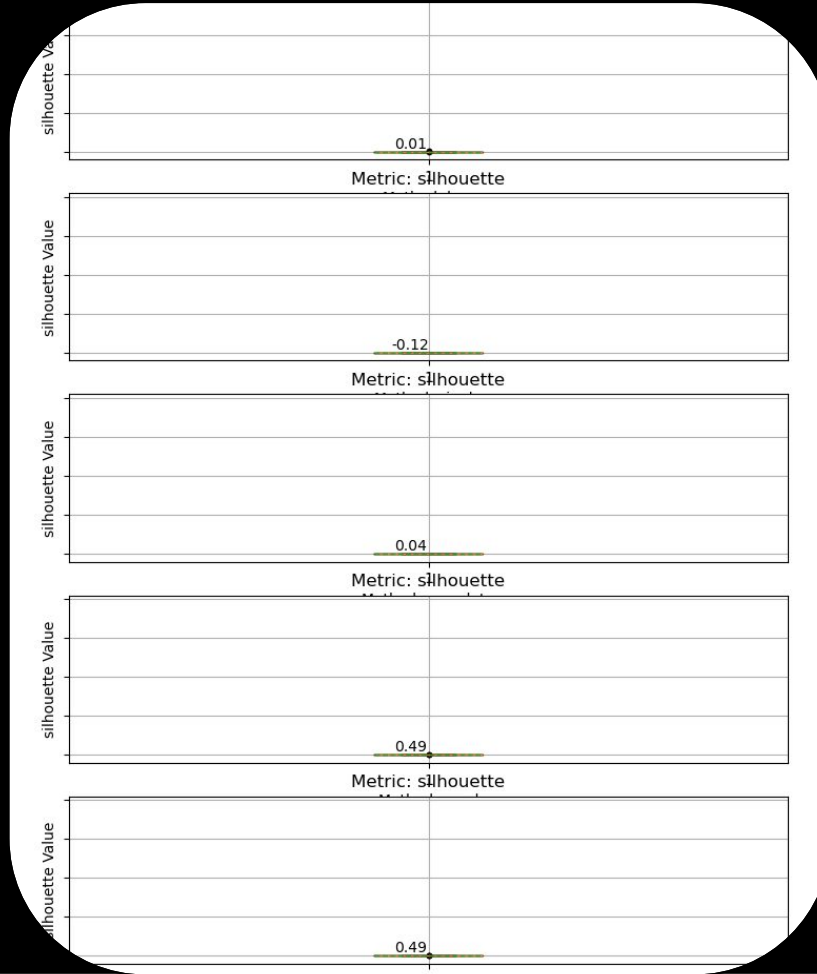
Single Link

Complete Link

Ward Algorithm

Spectral

# Significance



KNN

Single Link

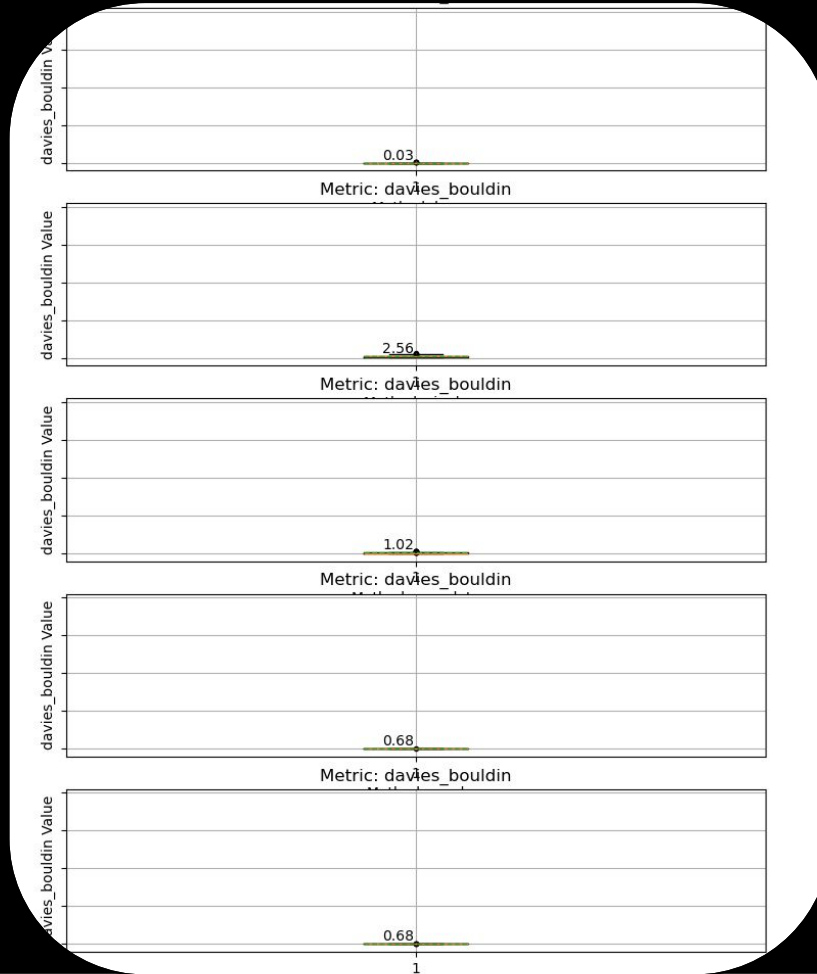
Complete Link

Ward Algorithm

Spectral



# Significance



KNN

Single Link

Complete Link

Ward Algorithm

Spectral

# Still pending

## Altering the KNN

Testing more values of K for the KNN (especially odd values)

Pending 1

## Emulating PCA

The original citation used PCA with 40 components in the final data. This is a significant modifications since the original data has more than 5800 genes and reducing this matrix would yield large benefits in runtime

Pending 2

## Gene Hierarchy

Using LDA to get the feature significance to know what genes contributed to formation of clusters the most

Pending 3

# Citations

- [1] D. Pellin *et al.*, “A comprehensive single cell transcriptional landscape of human hematopoietic progenitors,” *Nat. Commun.*, vol. 10, no. 1, p. 2395, Jun. 2019, doi: 10.1038/s41467-019-10291-0.
- [2] S. Wu, P. Bertholet, H. Huang, D. Cohen-Or, M. Gong, and M. Zwicker, “Structure-Aware Data Consolidation,” *IEEE Trans. Pattern Anal. Mach. Intell.*, vol. 40, no. 10, pp. 2529–2537, Oct. 2018, doi: 10.1109/TPAMI.2017.2754254.
- [3] F. Pedregosa *et al.*, “Scikit-learn: Machine Learning in Python,” *J. Mach. Learn. Res.*, vol. 12, no. 85, pp. 2825–2830, 2011.
- [4] A. P. Chatzigeorgiou and E. S. Manolakos, “MLscAN A flexible tool for single-cell data analysis pipelines and model selection using unsupervised machine learning methods”.