




# 1 Lazy matrix evaluation in R

2 Eric Kernfeld  <sup>1</sup>

3 <sup>1</sup> Department of Biomedical Engineering, Johns Hopkins University, USA

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

## Software

- [Review](#) 
- [Repository](#) 
- [Archive](#) 

Editor: [Open Journals](#) 

## Reviewers:

- [@openjournals](#)

Submitted: 01 January 1970

Published: unpublished

## License

Authors of papers retain  
copyright and release the work  
under a Creative Commons  
Attribution 4.0 International  
License ([CC BY 4.0](#)).

## 4 Summary

5 Modern single-cell genomics technologies hold tremendous promise for cancer genomics,  
6 developmental biology, and cellular engineering ([Griffiths et al., 2018](#); [Lei et al., 2021](#); [Rackham  
et al., 2021](#); [Teichmann & others, 2017](#)). Due to the breadth of applications for single-cell  
7 genomics and due to the idiosyncrasy of individual biological systems, single-cell genomics  
8 admits no one-size-fits-all approach to quality control or analysis. Rather, the rapid adoption of  
9 these measurement devices ([Svensson et al., 2020](#)) has suddenly conscripted an entire cohort of  
10 computational biologists into labor-intensive exploration and bespoke analysis of large, sparse  
11 count matrices.

12  
13 Individual datasets with minimal biological replication easily reach tens of thousands of  
14 observations across tens of thousands of genes or hundreds of thousands of functional genome  
15 elements, and RAM is a ubiquitous bottleneck. Preserving the initial sparsity of the data is a  
16 natural solution. Alas, this sparsity is often annihilated within the first few steps of a typical  
17 analysis – for example, by centering each gene to have a mean of 0. There is a need for  
18 software that can perform common statistical analysis of matrices, such as centering, scaling,  
19 and PCA, while preserving sparsity ([Booeshaghi et al., 2022](#)).

## 20 Statement of need

21 MatrixLazyEval is an R package for sparse matrix operations that prioritizes simplicity and  
22 ease of use, anticipating and avoiding typical misunderstandings. For ease of use, it is  
23 compatible with R's native matrix class as well as the popular Matrix and DelayedArray  
24 packages (see <https://cran.r-project.org/web/packages/Matrix/index.html> and <https://bioconductor.org/packages/release/bioc/html/DelayedArray.html>). Its workhorse objects use  
25 overloaded versions of typical R matrix operations such as `%*%`, `t`, and `tcrossprod`. Since this  
26 leaves some ambiguity about whether a given operation will be performed immediately or put  
27 off, MatrixLazyEval explicitly documents every function as lazy or eager.

28  
29 The package provides implementations of:

- 30 ▪ centering,
- 31 ▪ scaling,
- 32 ▪ replacing a matrix with its residuals after regression, and
- 33 ▪ randomized SVD or PCA.

34 These operations can be easily chained/combined without loss of sparsity. The result is a  
35 viable drop-in replacement for matrix objects in single-cell genomics workflows.

## 36 Acknowledgements

37 The author is grateful to Dr. René Maehr for access to single-cell genomics datasets at  
38 cutting-edge scale and complexity.

## References

- Booeshaghi, A. S., Hallgrimsdottir, I. B., Galvez-Merchan, A., & Pachter, L. (2022). Depth normalization for single-cell genomics count data. *bioRxiv*.
- Griffiths, J. A., Scialdone, A., & Marioni, J. C. (2018). Using single-cell genomics to understand developmental processes and cell fate decisions. *Molecular Systems Biology*, 14(4), e8046.
- Lei, Y., Tang, R., Xu, J., Wang, W., Zhang, B., Liu, J., Yu, X., & Shi, S. (2021). Applications of single-cell sequencing in cancer research: Progress and perspectives. *Journal of Hematology & Oncology*, 14(1), 1–26.
- Rackham, O., Cahan, P., Mah, N., Morris, S., Ouyang, J. F., Plant, A. L., Tanaka, Y., & Wells, C. A. (2021). Challenges for computational stem cell biology: A discussion for the field. *Stem Cell Reports*, 16(1), 3–9.
- Svensson, V., Veiga Beltrame, E. da, & Pachter, L. (2020). A curated database reveals trends in single-cell transcriptomics. *Database*, 2020.
- Teichmann, S., & others. (2017). The human cell atlas. *eLife*, 6, e27041.

DRAFT