

scRNAseq analysis

Abinet Joan

2024-10-18 11:10:20 +0200

Loading packages

```
suppressMessages(library(dplyr))
suppressMessages(library(Seurat))
suppressMessages(library(patchwork))
suppressMessages(library(ggplot2))
suppressMessages(library(SingleR))
suppressMessages(library(formatR))
```

Each sample correspond to one experimental conditions of lung myeloid cells, defined as living singlet CD45+, F4/80+, and/or CD11b+ cells from lung single-cell suspensions pooled from five C57BL/6 male WT mice, either mock-infected (Myeloid-Ctrl) or PR8-infected at day 10 after IAV (2 biological replicates: Myeloid-PR81 and Myeloid-PR82).

The Mouse1, Mouse2 and Mouse3 from GEO (GSE244727) correspond respectively to Myeloid-Ctrl, Myeloid-PR81 and Myeloid-PR82

It is required to create 3 folders, one per mouse, and to rename the files in each folder matrix.mtx, features.tsv and barcodes.tsv.

Loading data

```
myeloid_ctrl.data <- Read10X(data.dir = "Counts/mouse1/") # it is require to specify 10x
directory
cell_MyeloidCtrl <- CreateSeuratObject(counts = myeloid_ctrl.data$`Gene Expression`,
  project = "cell_myeloid_ctrl", min.cells = 3, min.features = 200)

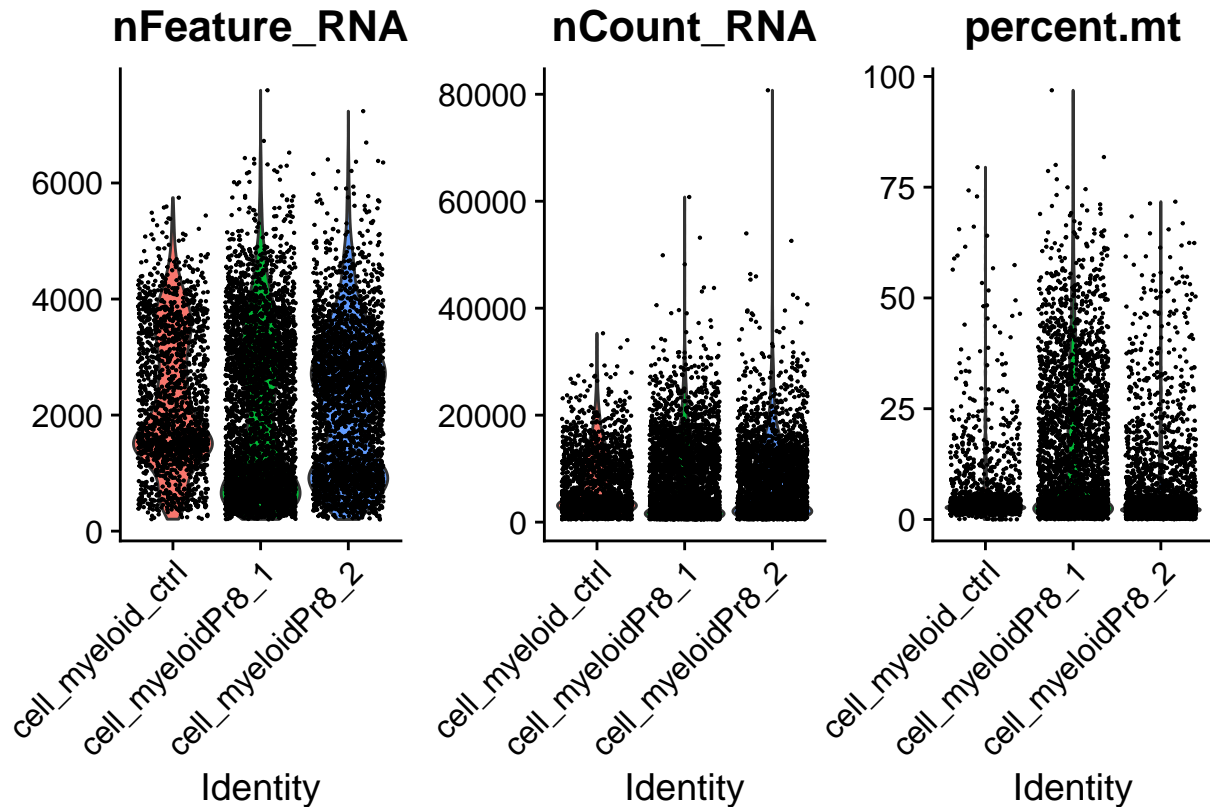
myeloid_PR81.data <- Read10X(data.dir = "Counts/mouse2/")
cell_myeloidPr8_1 <- CreateSeuratObject(counts = myeloid_PR81.data$`Gene Expression`,
  project = "cell_myeloidPr8_1", min.cells = 3, min.features = 200)

myeloid_PR82.data <- Read10X(data.dir = "Counts/mouse3/")
cell_myeloidPr8_2 <- CreateSeuratObject(counts = myeloid_PR82.data$`Gene Expression`,
  project = "cell_myeloidPr8_2", min.cells = 3, min.features = 200)

# Merge the 3 seurat object
myeloid_cell <- merge(cell_MyeloidCtrl, y = c(cell_myeloidPr8_1, cell_myeloidPr8_2),
  add.cell.ids = c("1", "2", "3"), project = "Myeloid_cells")
```

Quality control

```
myeloid_cell[["percent.mt"]] <- PercentageFeatureSet(myeloid_cell, pattern = "^mt-")
VlnPlot(myeloid_cell, features = c("nFeature_RNA", "nCount_RNA", "percent.mt"), ncol = 3,
        pt.size = 0.1)
```



Pre-processing workflow

```
# removing low quality cells
myeloid_cell <- subset(myeloid_cell, subset = nFeature_RNA > 200 & nFeature_RNA <
  5000 & nCount_RNA < 20000 & percent.mt < 10)

# Normalizing the data
myeloid_cell <- NormalizeData(myeloid_cell, normalization.method = "LogNormalize",
  scale.factor = 10000)

# Feature selection
myeloid_cell <- FindVariableFeatures(myeloid_cell, selection.method = "vst", nfeatures =
  2000)

# Identify the 10 most highly variable genes
top10 <- head(VariableFeatures(myeloid_cell), 10)

# plot variable features without labels
plot1 <- VariableFeaturePlot(myeloid_cell)
plot1 <- LabelPoints(plot = plot1, points = top10, repel = TRUE)
```

```

# Scaling the data
all.genes <- rownames(myeloid_cell)
myeloid_cell <- ScaleData(myeloid_cell, features = all.genes)

# Linear dimensional reduction
myeloid_cell <- RunPCA(myeloid_cell, features = VariableFeatures(object = myeloid_cell))
plot2 <- DimPlot(myeloid_cell, reduction = "pca")

# Determine the 'dimensionality' of the dataset
myeloid_cell <- JackStraw(myeloid_cell, num.replicate = 100)
myeloid_cell <- ScoreJackStraw(myeloid_cell, dims = 1:20)
plot3 <- JackStrawPlot(myeloid_cell, dims = 1:20)
plot4 <- ElbowPlot(myeloid_cell, ndims = 20)

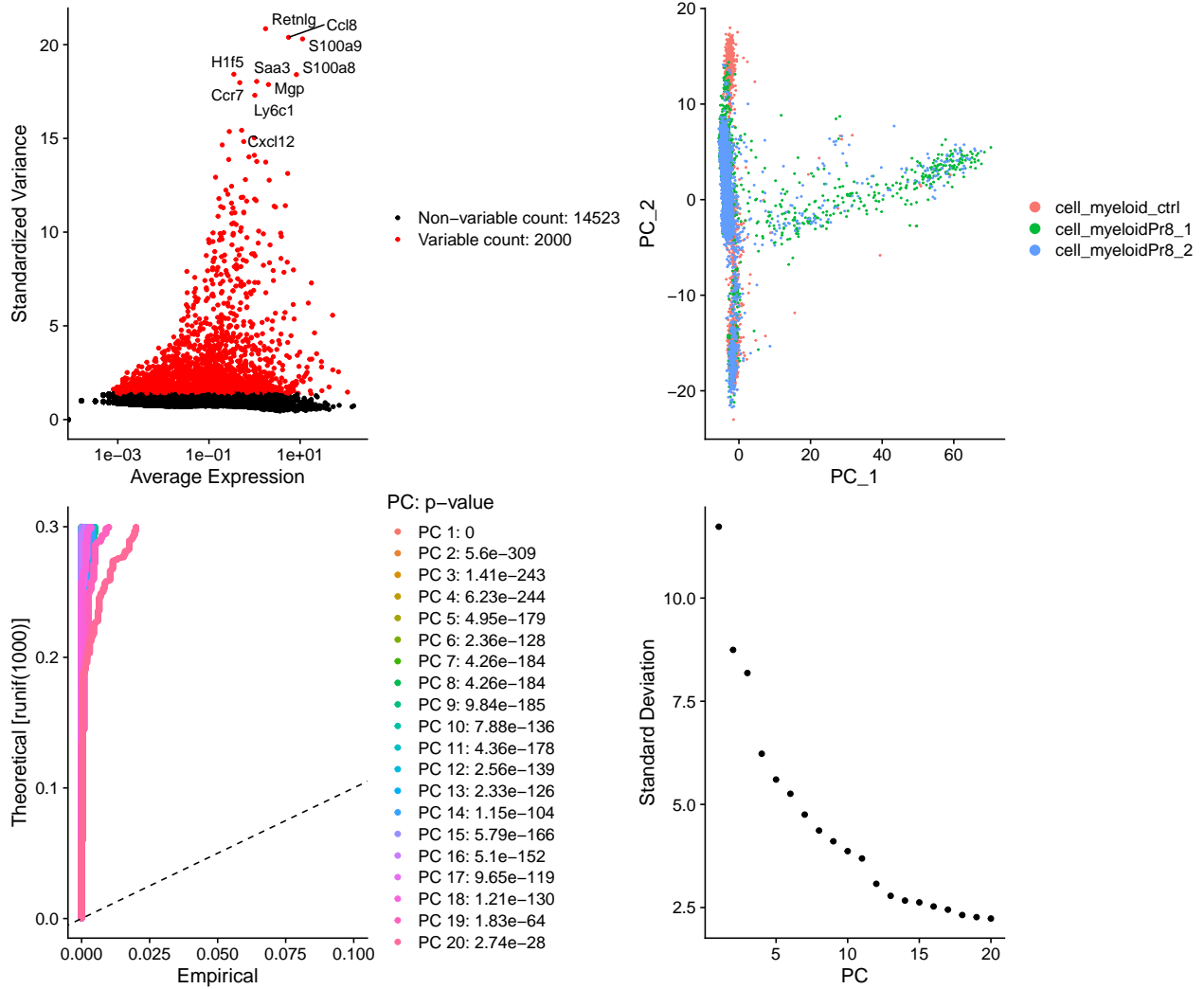
# Cluster cells in umap
myeloid_cell <- FindNeighbors(myeloid_cell, dims = 1:15)
myeloid_cell <- FindClusters(myeloid_cell, resolution = 0.25)
myeloid_cell <- RunUMAP(myeloid_cell, dims = 1:15)

plot1 + plot2 + plot3 + plot4

```

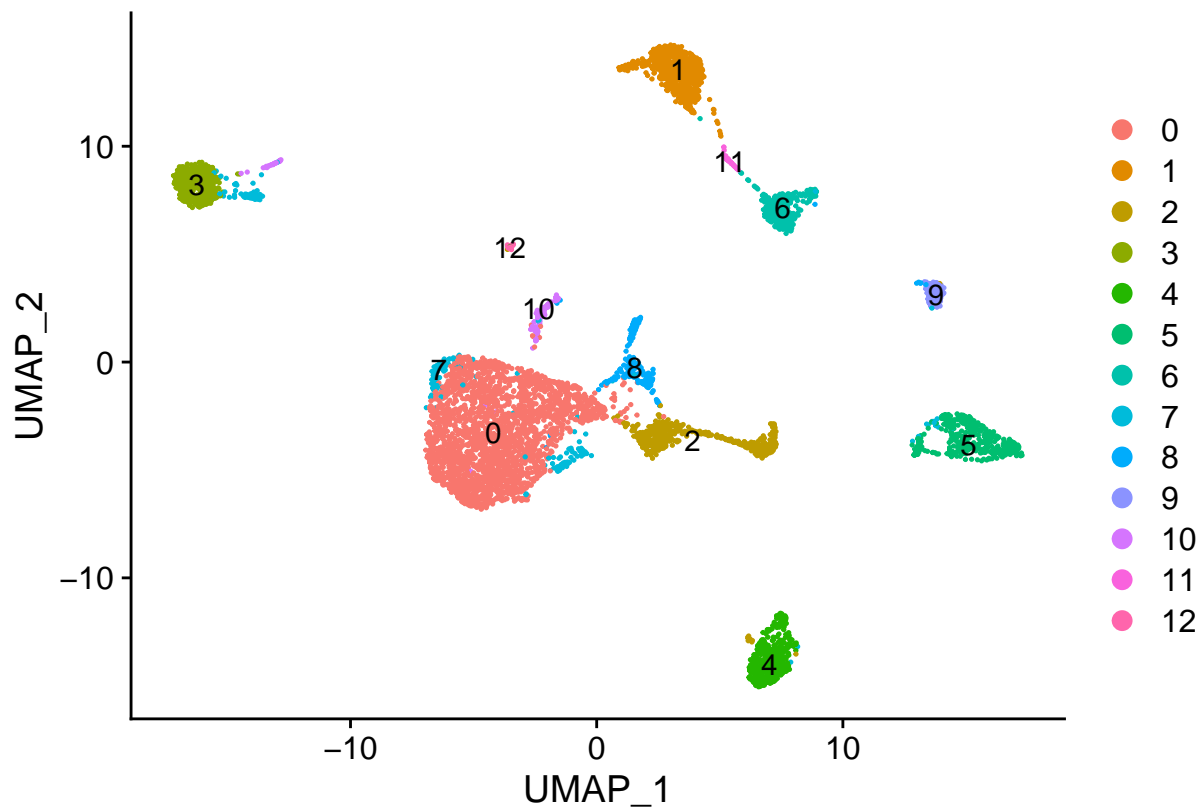
```
## Warning: Transformation introduced infinite values in continuous x-axis
```

```
## Warning: Removed 28000 rows containing missing values (`geom_point()`).
```



Visualizing all clusters

```
DimPlot(myeloid_cell, reduction = "umap", label = T)
```



Cell annotation

```
library(SingleR)
library(ExperimentHub)
library(scuttle)

eh <- ExperimentHub()
query(eh, "TabulaMurisData")

ref <- eh[["EH1617"]]
myeloid_ref <- ref[, !is.na(ref$cell_ontology_class)]
myeloid_ref <- myeloid_ref[, myeloid_ref$tissue == "Lung"]

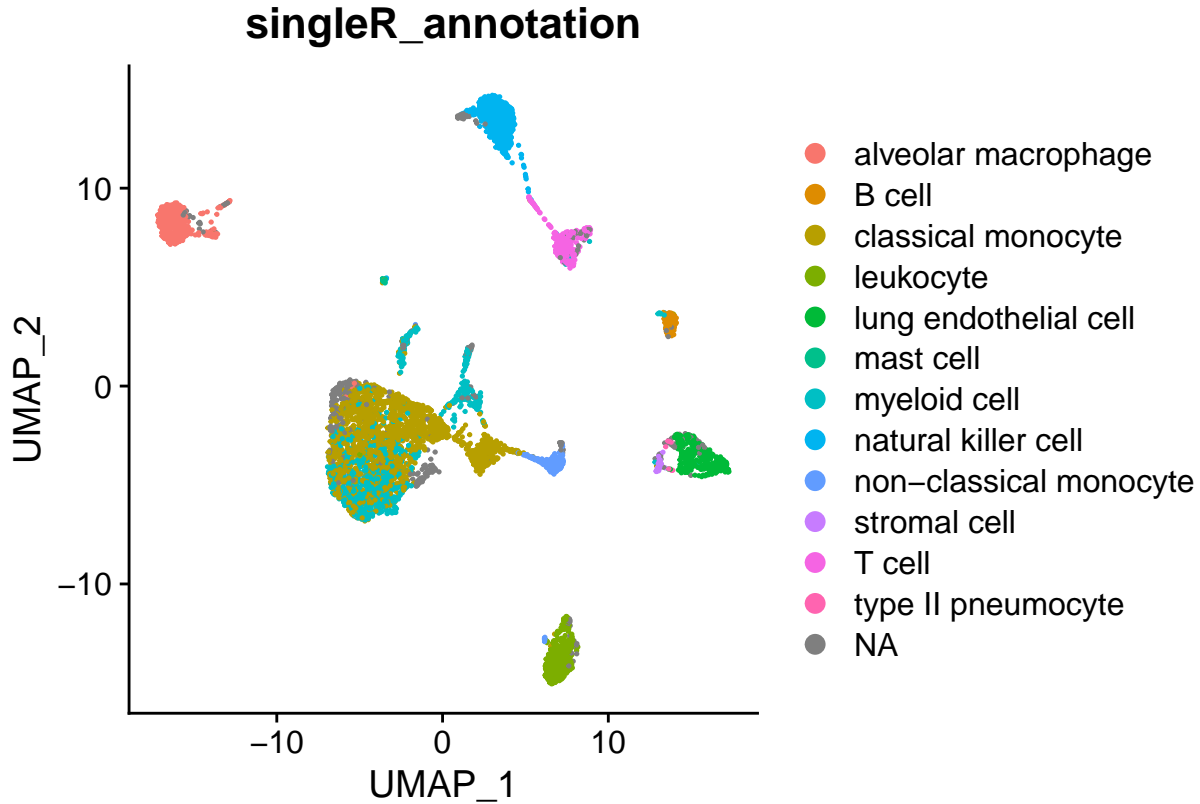
myeloid_ref <- logNormCounts(myeloid_ref)

tested_data <- as.SingleCellExperiment(myeloid_cell)
tested_data <- logNormCounts(tested_data)

results <- SingleR(test = tested_data, ref = myeloid_ref, labels =
myeloid_ref$cell_ontology_class)
cell_annotations <- results

myeloid_cell[["singleR_annotation"]] <- cell_annotations[, c(4)]
```

```
DimPlot(myeloid_cell, reduction = "umap", group.by = "singleR_annotation")
```



Removing the variable used for annotation

```
rm(tested_data)
rm(myeloid_ref)
```

The next step is to subset the clusters of myeloid cells (Macropages, neutrophils, DCs). The others clusters will be removed.

Removing contamination

```
myeloid_cells <- subset(myeloid_cell, seurat_clusters %in% c(0, 2, 3, 4, 7, 8, 10))
```

Saving file

```
saveRDS(myeloid_cells, "Myeloid_cells_Part1.rds")
```

```
sessionInfo()
```

```
## R version 4.3.3 (2024-02-29)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 22.04.4 LTS
##
```

```

## Matrix products: default
## BLAS: /usr/lib/x86_64-linux-gnu/openblas-pthread/libblas.so.3
## LAPACK: /usr/lib/x86_64-linux-gnu/openblas-pthread/libopenblas-p-r0.3.20.so; LAPACK version 3.10.0
##
## locale:
## [1] LC_CTYPE=en_US.UTF-8 LC_NUMERIC=C
## [3] LC_TIME=fr_BE.UTF-8 LC_COLLATE=en_US.UTF-8
## [5] LC_MONETARY=fr_BE.UTF-8 LC_MESSAGES=en_US.UTF-8
## [7] LC_PAPER=fr_BE.UTF-8 LC_NAME=C
## [9] LC_ADDRESS=C LC_TELEPHONE=C
## [11] LC_MEASUREMENT=fr_BE.UTF-8 LC_IDENTIFICATION=C
##
## time zone: Europe/Brussels
## tzcode source: system (glibc)
##
## attached base packages:
## [1] stats4 stats graphics grDevices utils datasets methods
## [8] base
##
## other attached packages:
## [1] TabulaMurisData_1.18.0 scuttle_1.10.1
## [3] SingleCellExperiment_1.22.0 ExperimentHub_2.8.0
## [5] AnnotationHub_3.8.0 BiocFileCache_2.8.0
## [7] dbplyr_2.3.2 formatR_1.14
## [9] SingleR_2.2.0 SummarizedExperiment_1.30.2
## [11] Biobase_2.60.0 GenomicRanges_1.52.0
## [13] GenomeInfoDb_1.36.0 IRanges_2.34.0
## [15] S4Vectors_0.38.1 BiocGenerics_0.46.0
## [17] MatrixGenerics_1.12.2 matrixStats_1.0.0
## [19] ggplot2_3.4.2 patchwork_1.1.2
## [21] SeuratObject_4.1.3 Seurat_4.3.0
## [23] dplyr_1.1.2
##
## loaded via a namespace (and not attached):
## [1] RcppAnnoy_0.0.21 splines_4.3.3
## [3] later_1.3.1 filelock_1.0.2
## [5] bitops_1.0-7 tibble_3.2.1
## [7] R.oo_1.25.0 polyclip_1.10-4
## [9] lifecycle_1.0.3 globals_0.16.2
## [11] lattice_0.22-5 MASS_7.3-60.0.1
## [13] magrittr_2.0.3 plotly_4.10.2
## [15] rmarkdown_2.23 yaml_2.3.7
## [17] httpuv_1.6.11 sctransform_0.3.5
## [19] spam_2.9-1 sp_2.0-0
## [21] spatstat.sparse_3.0-2 reticulate_1.30
## [23] cowplot_1.1.1 pbapply_1.7-2
## [25] DBI_1.1.3 RColorBrewer_1.1-3
## [27] abind_1.4-5 zlibbioc_1.46.0
## [29] Rtsne_0.16 purrr_1.0.1
## [31] R.utils_2.12.2 RCurl_1.98-1.12
## [33] rappdirs_0.3.3 GenomeInfoDbData_1.2.10
## [35] ggrepel_0.9.3 irlba_2.3.5.1
## [37] listenv_0.9.0 spatstat.utils_3.0-3
## [39] goftest_1.2-3 spatstat.random_3.1-5

```

## [41] fitdistrplus_1.1-11	parallelly_1.36.0
## [43] DelayedMatrixStats_1.22.1	leiden_0.4.3
## [45] codetools_0.2-19	DelayedArray_0.26.3
## [47] tidysselect_1.2.0	farver_2.1.1
## [49] ScaledMatrix_1.8.1	spatstat.explore_3.2-1
## [51] jsonlite_1.8.7	ellipsis_0.3.2
## [53] progressr_0.13.0	ggirdges_0.5.4
## [55] survival_3.5-8	tools_4.3.3
## [57] ica_1.0-3	Rcpp_1.0.11
## [59] glue_1.6.2	gridExtra_2.3
## [61] xfun_0.39	withr_2.5.0
## [63] BiocManager_1.30.21	fastmap_1.1.1
## [65] fansi_1.0.4	digest_0.6.33
## [67] rsvd_1.0.5	R6_2.5.1
## [69] mime_0.12	colorspace_2.1-0
## [71] scattermore_1.2	tensor_1.5
## [73] RSQLite_2.3.1	spatstat.data_3.0-1
## [75] R.methodsS3_1.8.2	utf8_1.2.3
## [77] tidyr_1.3.0	generics_0.1.3
## [79] data.table_1.14.8	httr_1.4.6
## [81] htmlwidgets_1.6.2	S4Arrays_1.2.1
## [83] uwot_0.1.16	pkgconfig_2.0.3
## [85] gtable_0.3.3	blob_1.2.4
## [87] lmtest_0.9-40	XVector_0.40.0
## [89] htmltools_0.5.5	dotCall64_1.0-2
## [91] scales_1.2.1	png_0.1-8
## [93] knitr_1.43	rstudioapi_0.14
## [95] reshape2_1.4.4	curl_5.0.1
## [97] nlme_3.1-164	cachem_1.0.8
## [99] zoo_1.8-12	stringr_1.5.0
## [101] BiocVersion_3.17.1	KernSmooth_2.23-22
## [103] parallel_4.3.3	miniUI_0.1.1.1
## [105] vipor_0.4.5	AnnotationDbi_1.62.1
## [107] ggrrastr_1.0.2	pillar_1.9.0
## [109] grid_4.3.3	vctrs_0.6.3
## [111] RANN_2.6.1	promises_1.2.0.1
## [113] BiocSingular_1.16.0	beachmat_2.16.0
## [115] xtable_1.8-4	cluster_2.1.6
## [117] beeswarm_0.4.0	evaluate_0.21
## [119] cli_3.6.1	compiler_4.3.3
## [121] rlang_1.1.1	crayon_1.5.2
## [123] future.apply_1.11.0	labeling_0.4.2
## [125] plyr_1.8.8	ggbeeswarm_0.7.2
## [127] stringi_1.7.12	viridisLite_0.4.2
## [129] deldir_1.0-9	BiocParallel_1.34.2
## [131] Biostrings_2.68.1	munsell_0.5.0
## [133] lazyeval_0.2.2	spatstat.geom_3.2-4
## [135] Matrix_1.6-1	bit64_4.0.5
## [137] sparseMatrixStats_1.12.0	future_1.33.0
## [139] KEGGREST_1.40.0	shiny_1.7.4.1
## [141] interactiveDisplayBase_1.38.0	highr_0.10
## [143] ROCR_1.0-11	memoise_2.0.1
## [145] igraph_1.5.0.1	bit_4.0.5