

PAPER TITLE TO BE DEFINED (in common.yaml)

7-RNAvelocity and scVelo analysis

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Abstract

Lung interstitium macrophages (IMs) are non-alveolar resident tissue macrophages which contribute to the lung homeostasis. These cells were reported to be heterogeneous by our group and other teams, which contains two main distinct subpopulations: CD206+ IMs and CD206- IMs. However, the exact origin of IMs and the transcriptional programs that control IM differentiation remains unclear. In recent report, we analyzed the refilled IMs in the course of time after induced IM depletion with single-cell RNA sequencing (10X Genomics Chromium) and bulk RNA sequencing.

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1 Description

2 Load data

```
# packages
suppressMessages(library(Seurat))
suppressMessages(library(tidyverse))
suppressMessages(library(RColorBrewer))
```

3 Prepare data from Seurat object for RNAvelocity

```
list.name.so <- unique(obj3d.combined$treatment)
list.name.sample <- list.name.so

for (i in 1:length(list.name.so)) {
  so <- obj3d.combined[, obj3d.combined$treatment == list.name.so[i]]
  assign(paste(list.name.sample[i], "seuratObject", sep = "."), so)
}

list3d.name.so <- paste(list.name.sample, "seuratObject", sep = ".")

obj.list <- list()
for (name.so in list3d.name.so) {
  obj.list <- c(obj.list, get(name.so))
}

names(obj.list) <- list.name.so
```

4 Generate loom files

The intermediate loom files were too big to be uploaded to the platform but they can be produced by the following steps.

We counted spliced, unspliced and ambiguous transcripts using velocity command-line tool (<http://velocity.org/>)[@LaManno2018].

For each sample, the following code was used to generate the loom file:

```
velocity run -b "${sampleID}/outs/filtered_feature_bc_matrix/barcodes.
tsv.gz" \
-o "outputDir/${sampleID}.loom" \
"${sampleID}/outs/possorted_genome_bam.bam" \
/GRCm38/genes/genes.gtf
```

- `${sampleID}` is the sample ID.
- `${sampleID}/outs` is the output directory of CellRanger.
- `${sampleID}/outs/possorted_genome_bam.bam` is the BAM file generated from CellRanger.
- `/GRCm38/genes/genes.gtf` is the gene reference used for Cellranger counts.

5 Add cell spliced/unspliced counts to Seurat objects

```
suppressMessages(library("velocity.R"))
```

5.1 load IM-DTR1 loom data

```
list.path.loom <- list.dirs("../IM-DTR1/counts/loom")
list.path.loom <- list.path.loom[-1] # remove the first parent directory:
    Counts/loom/
```

```
list.name.loom <- basename(list.path.loom)
list.name.loom <- str_replace(list.name.loom, pattern = "-", replacement =
    "_")
list.path.loom <- list.files(list.path.loom, pattern = "\\\\.loom$", full.
    names = TRUE)

for (i in 1:length(list.name.loom)) {
  assign(make.names(list.name.loom[i]), read.loom.matrices(list.path.loom[
    i]))
}
```

```
## reading loom file via hdf5r...
## reading loom file via hdf5r...
```

5.2 load IM-DTR2 loom data

```
list.path.loom <- list.dirs("../IM-DTR2/counts/loom")
list.path.loom <- list.path.loom[-1] # remove the first parent directory:
    Counts/loom/
```

```
list.name.loom <- basename(list.path.loom)
list.name.loom <- str_replace(list.name.loom, pattern = "-", replacement =
    "_")
list.path.loom <- list.files(list.path.loom, pattern = "\\\\.loom$", full.
    names = TRUE)

for (i in 1:length(list.name.loom)) {
  assign(make.names(list.name.loom[i]), read.loom.matrices(list.path.loom[
    i]))
}
```

```
## reading loom file via hdf5r...
```

6 Make consistency of cell names in Loom and Seurat object

6.1 for 4-day and 0-day data.

for the 4d and 0d data, we added "Cplus_" and "Plusplus_":

```
colnames(`4d.seuratObject`)[1:2]
```

```
## [1] "CPlus_AAACCCAAGAGGTCAC-1" "CPlus_AAACCCATCATCCTAT-1"
```

```
colnames(`0d.seuratObject`)[1:2]
```

```
## [1] "Plusplus_AAACCCAAGACATAAC-1" "Plusplus_AAACCCAAGGCATTTC-1"
```

Remove the "-1" from cell names:

```
obj.list$`4d` <- RenameCells(obj.list$`4d`,  
                             new.names = sub(colnames(obj.list$`4d`  
                             `), pattern = "-1", replacement = "  
                             "))  
  
obj.list$`0d` <- RenameCells(obj.list$`0d`,  
                             new.names = sub(colnames(obj.list$`0d`  
                             `), pattern = "-1", replacement = "  
                             "))
```

add the prefix to loom file:

```
prefix <- c("CPlus", "Plusplus")  
  
source("../R/aggregateLoom.R")  
for (i in 1:length(list.name.loom)) {  
  loom <- get(list.name.loom[i])  
  assign(list.name.loom[i], value = aggregateLoom(loom, Ori.ID = prefix[i  
  ]))  
}
```

Filter loom with Seurat gene/cell list

```
`0d.loom` <- Plusplus_NGS20_Q148.loom  
`4d.loom` <- CPlus_NGS20_Q147.loom  
  
list.name.sample2 <- c("0d", "4d")
```

```
source("../R/filterLoom.R")  
  
# list.name.ldat <- paste(list.name.sample, "ldat", "filtered", sep = ".")  
ldat.list <- list()  
  
for (name.sample in list.name.sample2) {  
  obj.name <- paste(name.sample, "seuratObject", sep = ".")  
  loom.name <- paste(name.sample, "loom", sep = ".")  
  ldat.name <- paste(name.sample, "ldat", "filtered", sep = ".")  
  
  assign(ldat.name,  
         value = filterLoom(loomObj = get(loom.name),  
                             geneList = rownames(obj.list[[obj.name]]),  
                             cellList = colnames(obj.list[[obj.name]]))  
  )  
}
```

6.2 For IM-DTR2 data

For the 12h, 48h and 24h data, we don't have prefix in Seurat object (no integration done)

```
colnames(obj.list$`2d`)[1:2]
```

```
## [1] "AAACCCAAGAGATTCA-1" "AAACCCAAGGACTTCT-1"
```

```
colnames(obj.list$`0.5d`)[1:2]
```

```
## [1] "AAACCCAAGCGTCGAA-1" "AAACCCAGTGTCCGTG-1"
```

```
colnames(obj.list$`1d`)[1:2]
```

```
## [1] "AAACCCAAGGGCTAAC-1" "AAACCCACACACAGCC-1"
```

Remove the "-1" from cell names:

```
obj.list$`0.5d` <- RenameCells(obj.list$`0.5d`,  
                               new.names = sub(colnames(obj.list$`0.5d`),  
                                               pattern = "-1", replacement =  
                                               ""))  
  
obj.list$`1d` <- RenameCells(obj.list$`1d`,  
                             new.names = sub(colnames(obj.list$`1d`),  
                                             pattern = "-1", replacement =  
                                             ""))  
  
obj.list$`2d` <- RenameCells(obj.list$`2d`,  
                             new.names = sub(colnames(obj.list$`2d`),  
                                             pattern = "-1", replacement =  
                                             ""))
```

Loom file name check:

```
colnames(CD45Plus_NGS21_S229.loom$spliced)[1:2]
```

```
## [1] "CPlus_AAAGGATCACGTCATA" "CPlus_AAACGCTAGACTCTTG"
```

```
CD45Plus_NGS21_S229.loom <- aggregateLoom(CD45Plus_NGS21_S229.loom, Ori.ID  
= "")
```

Filter loom with Seurat gene/cell list

```
list.name.sample3 <- c("0.5d", "1d", "2d")
```

```
ldat.list <- list()  
  
for (name.sample in list.name.sample3) {  
  obj.name <- paste(name.sample, "seuratObject", sep = ".")  
  # loom.name <- paste(name.sample, "loom", sep = ".")  
  ldat.name <- paste(name.sample, "ldat", "filtered", sep = ".")
```

```

assign(lmat.name,
      value = filterLoom(loomObj = CD45Plus_NGS21_S229.loom, # here we
                        use only this object.
                        geneList = rownames(obj.list[[obj.name]]),
                        cellList = colnames(obj.list[[obj.name]]))
      )
}

```

7 Prepare data by group all the Seurat and Loom objects

7.1 Make list

```

list.name.sample <- c("0d", "0.5d", "1d", "2d", "4d")
obj.all <- list()

for (name.sample in list.name.sample) {
  obj.name <- name.sample
  ldat.name <- paste(name.sample, "ldat", "filtered", sep = ".")

  tmp <- list(ldat = get(ldat.name), seurat = obj.list[[obj.name]] )

  obj.all[[name.sample]] <- tmp
}

```

Save to file

```

saveRDS(obj.all, file = "./obj.list.loom_surat.Rds")

```

7.2 Create merged seurat object and loom data.

```

# 1. merged seurat object.
list.name.sample <- names(obj.all)

seurat.all <- list()
ldat.all <- list()

for (sample.name in list.name.sample) {
  obj <- obj.all[[sample.name]]

  seurat.all[[sample.name]] <- obj[["seurat"]]
  ldat.all[[sample.name]] <- obj[["ldat"]]
}

# the following way doesn't conserve the dimension reduction.
# seurat.merge <- merge(x = seurat.all[[1]],
#                       y = unlist(seurat.all[2:length(seurat.all)]),
#                       merge.data = TRUE)

```

```

# try to extract cellnames from the integrated seurat object:
# seurat.combined <- obj.combined # which is a seurat object.
#
# cellnames <- character()
# for (sample.name in list.name.sample) {
#   obj <- seurat.all[[sample.name]]
#
#   cellnames <- append(cellnames, colnames(obj))
# }

# seurat.merge <- seurat.combined[ , cellnames]
seurat.merge <- obj3d.combined

```

As the names were not changed in the original obj.combined, we change that manually:

```

# -1
seurat.merge <- RenameCells(seurat.merge, new.names = sub(colnames(
  seurat.merge), pattern = "-1", replacement = ""))

```

Now merge the ldat:

```

# 2. merged loom data;
# source("~/Desktop/velocityto/Script/aggregateLoom.R")

for (sample.name in list.name.sample) {

  obj <- ldat.all[[sample.name]]
  if (sample.name == list.name.sample[1]) {
    spliced <- obj$spliced
    unspliced <- obj$unspliced
    ambiguous <- obj$ambiguous
  } else {
    spliced <- cbind(spliced, obj$spliced)
    unspliced <- cbind(unspliced, obj$unspliced)
    ambiguous <- cbind(ambiguous, obj$ambiguous)
  }
}

ldat.merge <- list(spliced=spliced,
                  unspliced=unspliced,
                  ambiguous=ambiguous)

```

```
dim(ldat.all[[1]]$spliced)
```

```
## [1] 22597 2039
```

```
dim(ldat.merge$spliced)
```

```
## [1] 22597 47072
```

```
dim(ldat.merge$unspliced)
```

```
## [1] 22597 47072
```

```
dim(ldat.merge$ambiguous)
```

```
## [1] 22597 47072
```

7.3 Regroup data by treatment

```
groupBy <- "treatment"
sample.groupBy <- unique(seurat.merge@meta.data[[groupBy]])

obj <- list()
for (sample.name in sample.groupBy) {
  seurat <- seurat.merge[, seurat.merge@meta.data[[groupBy]] == sample.name]
  cellnames <- colnames(seurat)
  ldat <- list(spliced = ldat.merge$spliced[, cellnames],
              unspliced = ldat.merge$unspliced[, cellnames],
              ambiguous = ldat.merge$ambiguous[, cellnames])
  obj[[sample.name]] <- list(ldat=ldat,
                             seurat=seurat)
}
```

save the merged obj:

```
saveRDS(seurat.merge, file = "seurat3d.merge.seuratObject.Rds")
saveRDS(ldat.merge, file = "ldat.merge.ldat.Rds")
```

8 scVelo analysis with dynamical model

For the details in the estimation of single-cell RNA velocity using dynamical model, refer to the original report[@Bergen2020]:

Bergen, V., Lange, M., Peidli, S., Wolf, F. A. & Theis, F. J. Generalizing RNA velocity to transient cell states through dynamical modeling. Nat. Biotechnol. (2020) doi:10.1038/s41587-020-0591-3.

The following codes were used to calculate scRNA velocity and presenting with the existing embedding and labels.

```
# python below
import scvelo as scv
scv.settings.verbosity = 3 # show errors(0), warnings(1), info(2), hints(3)
scv.settings.presenter_view = True # set max width size for presenter view
scv.set_figure_params('scvelo') # for beautified visualization

# load data
ldata_basal = scv.read("./No_Fumer.loom")

# Preprocess the Data
```



```

scv.pp.filter_and_normalize(ldata_basal, min_shared_counts=20, n_top_genes
=2000)
scv.pp.moments(ldata_basal, n_pcs=30, n_neighbors=30)

# Estimate RNA velocity with dynamical model
scv.tl.recover_dynamics(ldata_basal)
scv.tl.velocity(ldata_basal, mode='dynamical')
scv.tl.velocity_graph(ldata_basal)
scv.pl.velocity_embedding_stream(ldata_basal,
    basis='umap_cell_embeddings', color='cell.type2', figsize=(10,10),
    palette=["#FDBF6F", "#33A02C", "#A6CEE3", "#B2DF8A", "#1F78B4", "#E31A1C", "
    #FB9A99"],
    linewidth=3, arrow_color="black",
    title="pc_1,2", alpha = 0.9, legend_fontsize = 0,
    add_outline=True,
    components='1,2'
)

```

```

scv.pl.velocity_embedding_stream(ldata_basal,
    basis='umap_cell_embeddings', color='cell.type2', figsize=(10,10),
    palette=["#FDBF6F", "#33A02C", "#A6CEE3", "#B2DF8A", "#1F78B4", "#E31A1C", "
    #FB9A99"],
    linewidth=3, arrow_color="black",
    title="pc_1,3", alpha = 0.9, legend_fontsize = 0,
    add_outline=True,
    components='1,3'
)

```

9 Session information

```

sessionInfo()

## R version 4.0.3 (2020-10-10)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 20.04.3 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/liblapack.so.3
##
## locale:
##  [1] LC_CTYPE=en_US.UTF-8      LC_NUMERIC=C
##  [3] LC_TIME=en_GB.UTF-8      LC_COLLATE=en_US.UTF-8
##  [5] LC_MONETARY=en_GB.UTF-8  LC_MESSAGES=en_US.UTF-8
##  [7] LC_PAPER=en_GB.UTF-8     LC_NAME=C
##  [9] LC_ADDRESS=C             LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_GB.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods   base
##
## other attached packages:

```

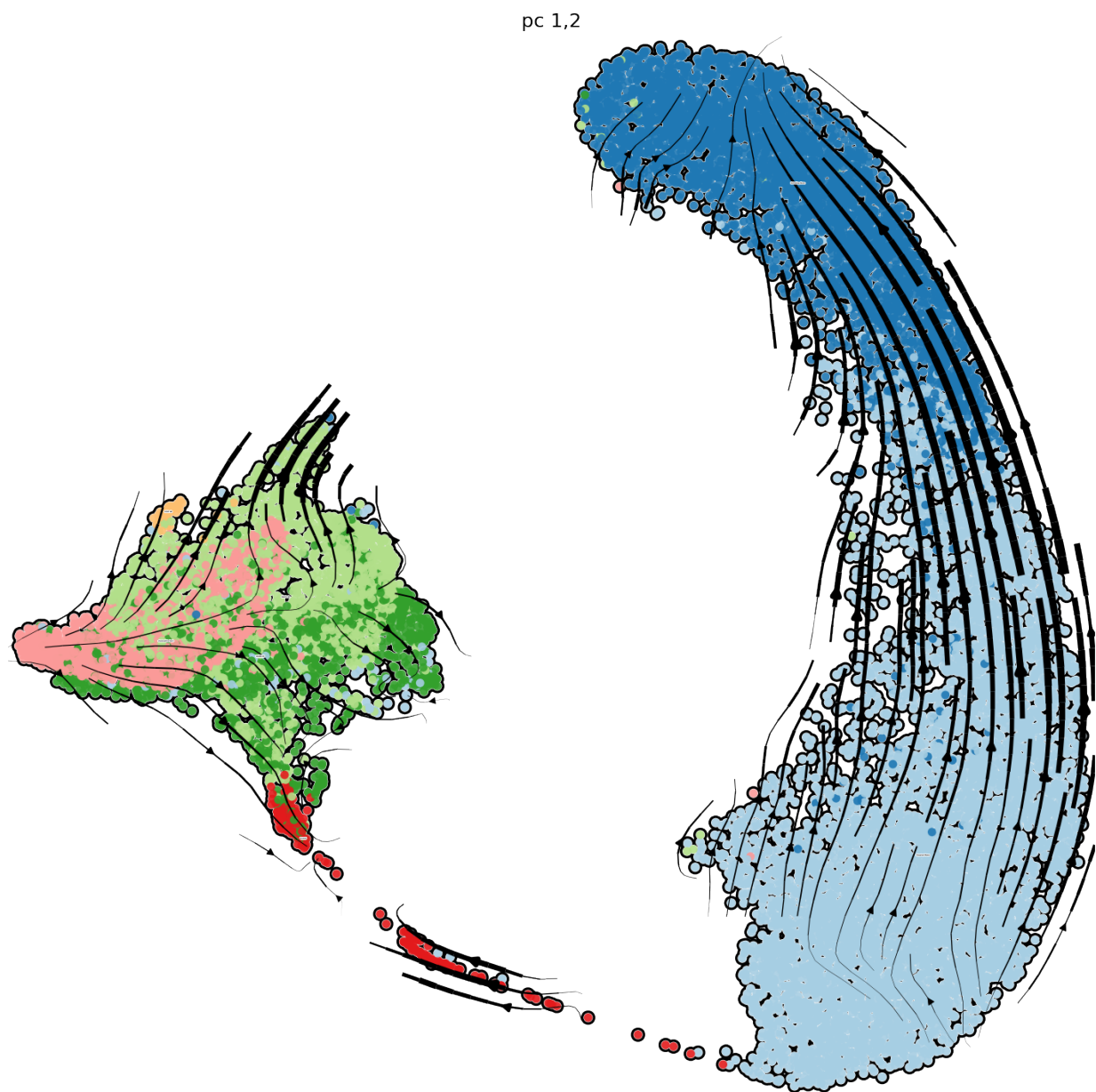


Figure 1: scVelo results for merged sample

pc 1,3

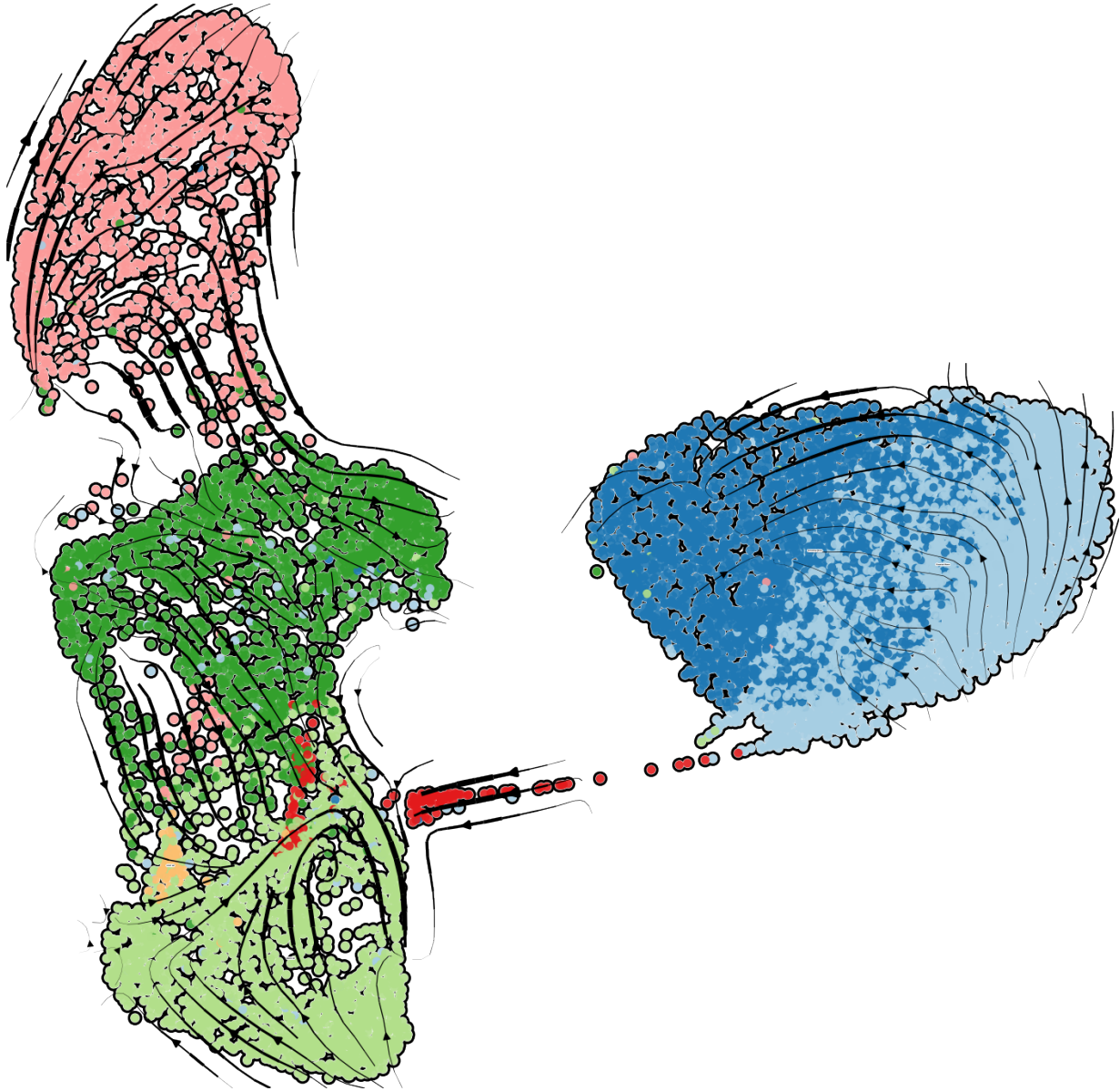


Figure 2: scVelo results for merged sample

##	[1]	velocity.R_0.6	Matrix_1.3-4	RColorBrewer_1.1-2	forcats_0	21
##		.5.1				
##	[5]	stringr_1.4.0	dplyr_1.0.7	purrr_0.3.4	readr_2	22
##		.0.0				
##	[9]	tidyr_1.1.3	tibble_3.1.3	ggplot2_3.3.5		23
##		tidyverse_1.3.1				
##	[13]	SeuratObject_4.0.2	Seurat_4.0.3			24
##						25
##		loaded via a namespace (and not attached):				26
##	[1]	Rtsne_0.15	colorspace_2.0-2	deldir_0.2-10		27
##	[4]	ellipsis_0.3.2	ggribes_0.5.3	fs_1.5.0		28
##	[7]	rstudioapi_0.13	spatstat.data_2.1-0	leiden_0.3.9		29
##	[10]	listenv_0.8.0	bit64_4.0.5	ggrepel_0.9.1		30
##	[13]	lubridate_1.7.10	fansi_0.5.0	xml2_1.3.2		31
##	[16]	codetools_0.2-18	splines_4.0.3	knitr_1.33		32
##	[19]	polyclip_1.10-0	jsonlite_1.7.2	broom_0.7.9		33
##	[22]	ica_1.0-2	cluster_2.1.0	dbplyr_2.1.1		34
##	[25]	png_0.1-7	uwot_0.1.10.9000	shiny_1.6.0		35
##	[28]	sctransform_0.3.2	spatstat.sparse_2.0-0	compiler_4.0.3		36
##	[31]	httr_1.4.2	backports_1.2.1	assertthat_0.2.1		37
##	[34]	fastmap_1.1.0	lazyeval_0.2.2	cli_3.0.1		38
##	[37]	later_1.2.0	htmltools_0.5.1.1	tools_4.0.3		39
##	[40]	igraph_1.2.6	gtable_0.3.0	glue_1.4.2		40
##	[43]	RANN_2.6.1	reshape2_1.4.4	Rcpp_1.0.7		41
##	[46]	Biobase_2.50.0	scattermore_0.7	cellranger_1.1.0		42
##	[49]	vctr_0.3.8	nlme_3.1-152	lmtest_0.9-38		43
##	[52]	xfun_0.24	globals_0.14.0	rvest_1.0.1		44
##	[55]	mime_0.11	miniUI_0.1.1.1	lifecycle_1.0.0		45
##	[58]	irlba_2.3.3	goftest_1.2-2	future_1.21.0		46
##	[61]	MASS_7.3-53	zoo_1.8-9	scales_1.1.1		47
##	[64]	spatstat.core_2.3-0	pcaMethods_1.82.0	hms_1.1.0		48
##	[67]	promises_1.2.0.1	spatstat.utils_2.2-0	parallel_4.0.3		49
##	[70]	yaml_2.2.1	reticulate_1.20	pbapply_1.4-3		50
##	[73]	gridExtra_2.3	rpart_4.1-15	stringi_1.7.3		51
##	[76]	BiocGenerics_0.36.1	rlang_0.4.11	pkgconfig_2.0.3		52
##	[79]	matrixStats_0.60.0	evaluate_0.14	lattice_0.20-41		53
##	[82]	ROCR_1.0-11	tensor_1.5	patchwork_1.1.1		54
##	[85]	htmlwidgets_1.5.3	bit_4.0.4	cowplot_1.1.1		55
##	[88]	tidyselect_1.1.1	parallelly_1.27.0	RcppAnnoy_0.0.19		56
##	[91]	plyr_1.8.6	magrittr_2.0.1	R6_2.5.0		57
##	[94]	generics_0.1.0	DBI_1.1.1	withr_2.4.2		58
##	[97]	pillar_1.6.2	haven_2.4.3	mgcv_1.8-33		59
##	[100]	fitdistrplus_1.1-5	survival_3.2-7	abind_1.4-5		60
##	[103]	future.apply_1.7.0	hdf5r_1.3.3	modelr_0.1.8		61
##	[106]	crayon_1.4.1	KernSmooth_2.23-20	utf8_1.2.2		62
##	[109]	spatstat.geom_2.2-2	plotly_4.9.4.1	tzdb_0.1.2		63
##	[112]	rmarkdown_2.9	readxl_1.3.1	grid_4.0.3		64
##	[115]	data.table_1.14.0	reprex_2.0.0	digest_0.6.27		65
##	[118]	xtable_1.8-4	httpuv_1.6.1	munsell_0.5.0		66
##	[121]	viridisLite_0.4.0				67

10 References