Simulation sceanrio 1

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Load functions:

```
## set wd
library(ggplot2); theme_set(theme_bw())
library(dplyr)
library(Seurat)
library(mclust)
library(MCMCpack)
library(mvtnorm)
library(salso)
source("./SASC_func.R")
source("./comparison_func.R")
data_dir <- "../data/simulation_sc1"</pre>
output_dir <- "../output/simulation_sc1"</pre>
output_data_dir <- "../output/simulation_sc1/data"</pre>
output_figure_dir <- "../output/simulation_sc1/figures"</pre>
pal <- c("#ffb6db", "#33A02C", "#b66dff", "#FEC44F", "#41B6C4", "#8E0152", "#0868AC", "#807DBA", "#E729
         "#00441B", "#525252", "#4D9221", "#8B5742", "#D8DAEB", "#7cdd2d", "#980043", "#8C96C6", "#EC70
         "#FDAE61", "#1D91C0", "#A6DBA0", "#4292C6", "#BF812D", "#01665E", "#41AB5D", "#FE9929", "#2525
names(pal) <- 1:30
set.seed(1999)
```

Load data set:

```
# load simulated data
sim_data <- readRDS(pasteO(data_dir, "/sim_data.rds"))

count <- as.matrix(sim_data@assays$RNA@counts)
gene <- sim_data@reductions$pca@cell.embeddings
gene_umap <- sim_data@reductions$umap@cell.embeddings
condition <- sim_data@meta.data$condition_sim
cell_names <- colnames(count)
celltype_sim <- factor(as.numeric(sim_data@meta.data$celltype_sim))
celltype_type <- sim_data@meta.data$celltype_CLE
celltype_type <- factor(celltype_type, levels = c("C", "A", "B"))

method <- "Truth"</pre>
```

Seurat

```
method <- "Seurat"
### Seurat clustering
t0 <- Sys.time()
sim_data_seurat <- sim_data</pre>
sim_data_seurat <- FindNeighbors(sim_data_seurat, dims = 1:10)</pre>
sim_data_seurat <- FindClusters(object = sim_data_seurat)</pre>
## Modularity Optimizer version 1.3.0 by Ludo Waltman and Nees Jan van Eck
## Number of nodes: 1601
## Number of edges: 58036
## Running Louvain algorithm...
## Maximum modularity in 10 random starts: 0.7187
## Number of communities: 7
## Elapsed time: 0 seconds
cluster <- as.numeric(sim_data_seurat@meta.data$seurat_clusters)</pre>
cluster <- factor(cluster, levels=as.character(1:max(cluster)))</pre>
runtime <- as.numeric(difftime(Sys.time(), t0, units = "secs"))</pre>
### cluster plot
plotcluster_func(gene_umap, cluster, method,
                  filename=pasteO(output_figure_dir, "/cluster_", method, ".pdf"))
### cluster weight, type
cluster_num <- cluster_num_func(cluster, condition)</pre>
cluster_type <- cluster_type_func(cluster, cluster_num, wdiff_tol)</pre>
#### acc
acc_list <- acc_func(predicted=cluster_type, truth=celltype_type)</pre>
cat("\n Balanced accuracy: \n", round(acc_list$balanced_acc, 2))
##
## Balanced accuracy:
## 1 0.96 0.95
# store outputs
results_seurat <- list(cluster = cluster, cluster_type = cluster_type,
                        acc = acc_list$acc,
                        balanced_acc = acc_list$balanced_acc,
                        runtime = runtime)
```

SASC

```
method <- "SASC"
if (file.exists(paste0(output_data_dir, "/results_SASC.rds"))) {
  results_SASC <- readRDS(paste0(output_data_dir, "/results_SASC.rds"))</pre>
```

```
} else {
  #### DMM clustering
  LOCRC_cells <- cell_names[which(condition == "B")]
  y0 <- as.matrix(gene[LOCRC_cells, ])</pre>
  YOCRC_cells <- cell_names[which(condition == "A")]
  y1 <- as.matrix(gene[YOCRC_cells, ])</pre>
  dim(y0)
  dim(y1)
  n_{\text{feat}} \leftarrow \dim(y0)[2]
  ncells <- dim(gene)[1]</pre>
  H <- 2
  seed <- 1999
  beta0_low <- 1</pre>
  beta0_high <- ncells/H/3/2
  t0 <- Sys.time()
  out <- func888(features = gene, y0 = y0, y1 = y1, H = H, seed = seed,
                   cell_names = cell_names, LOCRC_cells = LOCRC_cells, YOCRC_cells = YOCRC_cells,
                  beta0_low = beta0_low, beta0_high = beta0_high, iden_eps = TRUE,
                  niter = 5000, niter_extra = 1000)
  z0_post <- out$z0_post</pre>
  z1_post <- out$z1_post</pre>
  weights <- out$weights
  weights_mat <- out$weights_mat</pre>
  cluster_type_CLE <- out$cluster_type_CLE</pre>
  cluster <- rep(NA, nrow(gene))</pre>
  names(cluster) <- rownames(gene)</pre>
  cluster[LOCRC_cells] <- as.character(z0_post)</pre>
  cluster[YOCRC_cells] <- as.character(z1_post)</pre>
  cluster <- as.numeric(cluster)</pre>
  cluster <- factor(cluster, levels=as.character(1:max(cluster)))</pre>
  runtime <- as.numeric(difftime(Sys.time(), t0, units = "secs"))</pre>
  ####
  ### cluster plot
  plotcluster_func(gene_umap, cluster, method,
                     filename=pasteO(output_figure_dir, "/cluster_", method, ".pdf"))
  # ### cluster type
  Cind <- which(cluster_type_CLE == "C")</pre>
  Bind <- which(cluster_type_CLE == "L")</pre>
  Aind <- which(cluster_type_CLE == "E")
  cluster_type <- rep(NA, length(cluster))</pre>
  cluster_type[which(cluster %in% Cind)] <- "C"</pre>
  cluster_type[which(cluster %in% Bind)] <- "B"</pre>
  cluster_type[which(cluster %in% Aind)] <- "A"</pre>
```

ZINB-WaVE

```
method <- "ZINB-WaVE"
if (file.exists(pasteO(output_data_dir, "/results_zinb.rds"))) {
 results_zinb <- readRDS(paste0(output_data_dir, "/results_zinb.rds"))</pre>
} else {
  library("zinbwave")
  library("SummarizedExperiment")
  # construct SummarizedExperiment object
  t0 <- Sys.time()
  colData <- DataFrame(Condition=condition,</pre>
                        row.names=colnames(count))
  simdata_zinb <- SummarizedExperiment(assays=list(counts=count),</pre>
                                         colData=colData)
  # we filter out the lowly expressed genes,
  # by removing those genes that do not have at least 5 reads in at least 5 samples
  filter <- rowSums(assay(simdata zinb)>5)>5
  table(filter)
  simdata_zinb <- simdata_zinb[filter,]</pre>
  # We next identify the 100 most variable genes
  assay(simdata_zinb) %>% log1p %>% rowVars -> vars
  names(vars) <- rownames(simdata_zinb)</pre>
  vars <- sort(vars, decreasing = TRUE)</pre>
  head(vars)
  simdata_zinb <- simdata_zinb[names(vars)[1:100],]</pre>
  assayNames(simdata_zinb)[1] <- "counts"</pre>
  # run ZINB-WaVE
  zinb_out <- zinbwave(simdata_zinb, K = 10, X="~Condition", epsilon=1000)
  W_zinb <- reducedDim(zinb_out)</pre>
  zinb_out_seurat <- as.Seurat(x = zinb_out, counts = "counts", data = "counts")</pre>
  zinb_out_seurat <- FindNeighbors(zinb_out_seurat, reduction = "zinbwave",</pre>
```

```
dims = 1:10 #this should match K
  )
  zinb out seurat <- FindClusters(object = zinb out seurat)</pre>
  cluster <- as.numeric(zinb_out_seurat@meta.data$seurat_clusters)</pre>
  cluster <- factor(cluster, levels=as.character(1:max(cluster)))</pre>
  runtime <- as.numeric(difftime(Sys.time(), t0, units = "secs"))</pre>
  ### cluster plot
  plotcluster_func(gene_umap, cluster, method,
                    filename=pasteO(output_figure_dir, "/cluster_", method, ".pdf"))
  ### cluster weight, type
  cluster_num <- cluster_num_func(cluster, condition)</pre>
  cluster_type <- cluster_type_func(cluster, cluster_num, wdiff_tol)</pre>
  #### acc
  acc_list <- acc_func(predicted=cluster_type, truth=celltype_type)</pre>
  cat("\n Balanced accuracy: \n", round(acc_list$balanced_acc, 2))
  # store outputs
  results_zinb <- list(cluster = cluster, cluster_type = cluster_type,
                        acc = acc_list$acc,
                        balanced_acc = acc_list$balanced_acc,
                        runtime = runtime)
  saveRDS(results_zinb, file = paste0(output_data_dir, "/results_zinb.rds"))
}
```

GMM

```
acc_list <- acc_func(predicted=cluster_type, truth=celltype_type)</pre>
cat("\n Balanced accuracy: \n", round(acc_list$balanced_acc, 2))
##
## Balanced accuracy:
## 0.96 0.83 0.84
# store outputs
results_GMM <- list(cluster = cluster, cluster_type = cluster_type,
                   acc = acc_list$acc,
                    balanced_acc = acc_list$balanced_acc,
                    runtime = runtime)
Summary of results
### all clustering accuracy
b_acc_all <- rbind(results_SASC$balanced_acc,</pre>
                   results_GMM$balanced_acc,
                   results_seurat$balanced_acc,
                   results_zinb$balanced_acc)
rownames(b_acc_all) <- c("SASC", "GMM", "Seurat", "ZINB-WaVE")</pre>
cat("\n Balanced accuracy: \n")
##
## Balanced accuracy:
round(b_acc_all, 2)
##
                C
                     Α
                         В
## SASC
            0.95 0.98 0.92
## GMM
            0.96 0.83 0.84
## Seurat
            1.00 0.96 0.95
## ZINB-WaVE 0.24 0.54 0.47
ari <- c(
  adjustedRandIndex(celltype_sim, results_SASC$cluster),
  adjustedRandIndex(celltype_sim, results_GMM$cluster),
  adjustedRandIndex(celltype sim, results seurat$cluster),
  adjustedRandIndex(celltype_sim, results_zinb$cluster)
names(ari) <- c("SASC", "GMM", "Seurat", "ZINB-WaVE")</pre>
cat("\n ARI: \n", round(ari, 2))
##
## ARI:
## 0.85 0.5 0.6 0.18
### save running times
runtime_tab <- matrix(c(results_SASC$runtime,</pre>
                        results_GMM$runtime,
                        results_seurat$runtime,
```

results_zinb\$runtime), nrow = 1)