

Differential Expression Analysis for bulk RNA-seq data

CTRL Condition: Vehicle vs BPN

Ximing Ran

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```
library(tibble)
library(tidyr)
library(dplyr)
library(rtracklayer)
```

```
# load function from local files
source(here::here("source", "DEG_functions.R"))
```

1. Read the count data

In this section, we will read the clean count data from the `synaptosomes_bulkRNA` folder. We will read the data and merge them into a single table. The final table will be stored in `../dataresults/bulkRNA_counts_clean.csv`.

```
input_count <- read.csv(here::here("data", "bulkRNA",
                                   "bulkRNA_counts_cleaned.csv"))

counts <- as.data.frame(input_count) %>%
  column_to_rownames(var = "gene")
colnames(counts) <- gsub("_", "-", colnames(counts))

# if the colname is start with X, remove it
colnames(counts) <- gsub("^X", "", colnames(counts))

# raw sample list
sample_list_raw <- read.csv(here::here("data", "bulkRNA",
                                       "sample_info_FXS.csv")) %>%
  mutate(condition = Label,
         sample = gsub("_", "-", Sample_ID))

# Ensure the column names of counts exist in Sample.name
new_colnames <- sample_list_raw$sample[match(colnames(counts), sample_list_raw$sample)]

# Assign new column names
colnames(counts) <- new_colnames

# sort the columns by the colname
condition_list <- data.frame(
  group = sample_list_raw$condition
)

rownames(condition_list) <- sample_list_raw$sample

counts <- counts[, rownames(condition_list)]

gene_name_mapping <- readRDS(here::here("data", "ref", "gene_name_mapping.rds"))
```

2. Differential expression analysis

In this section, we will perform differential expression analysis using DESeq2.

```
# Init the result folder structure for the result
result_folder_all = './results/DEG-CTRL_Veh_vs_BPN'
result_folder = result_folder_all
Result_folder_structure(result_folder)

# load the comparison group information
reference_group <- "CTRL_VEH"
compare_group <- "CTRL_BPN"
reference_group_short <- reference_group
compare_group_short <- compare_group
```

```

filter_sample_info <- condition_list %>%
  filter(group %in% c(reference_group, compare_group))
filter_counts <- counts[, rownames(filter_sample_info)]

# Run the DESeq2 analysis
dds_obj <- DEAnalysis(counts = filter_counts,
                      reference_group = reference_group,
                      compare_group = compare_group,
                      condition_list = filter_sample_info,
                      target_gene = target_gene,
                      result_folder = result_folder)

res <- results(dds_obj)
resOrdered <- res[order(res$padj), ]

# omit the NA values
resOrdered <- resOrdered[!is.na(resOrdered$padj),]
dds_obj <- dds_obj[rownames(resOrdered),]
write.csv(resOrdered, file.path(result_folder, "02-DEG", "01_all_gene_results.csv"))

# DEG with log2fc > 1 and padj < 0.05
deg_1 <- resOrdered %>% as.data.frame() %>% rownames_to_column(var = "gene") %>%
  filter(padj < 0.05 & abs(log2FoldChange) > 1) %>% arrange(padj)
deg_1 <- deg_1[!is.na(deg_1$padj),]
write.csv(deg_1, file.path(result_folder, "02-DEG", "02_DEG_log2fc_1.csv"), row.names = FALSE)

# DEG with log2fc > 1.5 and padj < 0.05
deg_1.5 <- resOrdered %>% as.data.frame() %>% rownames_to_column(var = "gene") %>%
  filter(padj < 0.05 & abs(log2FoldChange) > 1.5) %>% arrange(padj)
deg_1.5 <- deg_1.5 [!is.na(deg_1.5 $padj),]
write.csv(deg_1.5 , file.path(result_folder, "02-DEG", "03_DEG_log2fc_1_5.csv"), row.names = FALSE)
print("DEG analysis is done")

## [1] "DEG analysis is done"

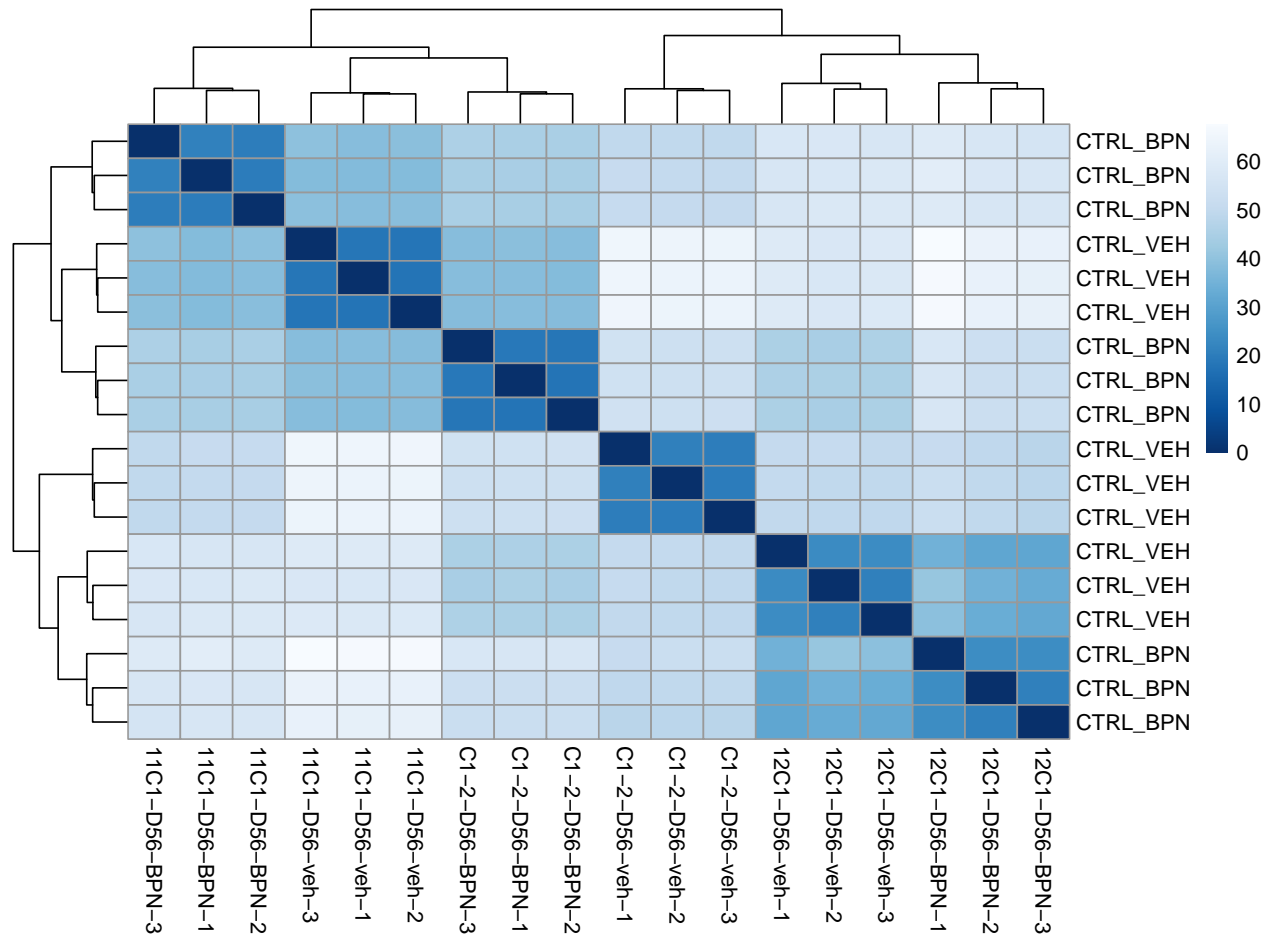
# Save the normalized counts
normalized_counts <- counts(dds_obj, normalized = TRUE)
write.csv(normalized_counts, file.path(result_folder, "02-DEG", "DESeq2_normalized_counts.csv"))

```

3. Visualization for reuslt

(1) Sample information

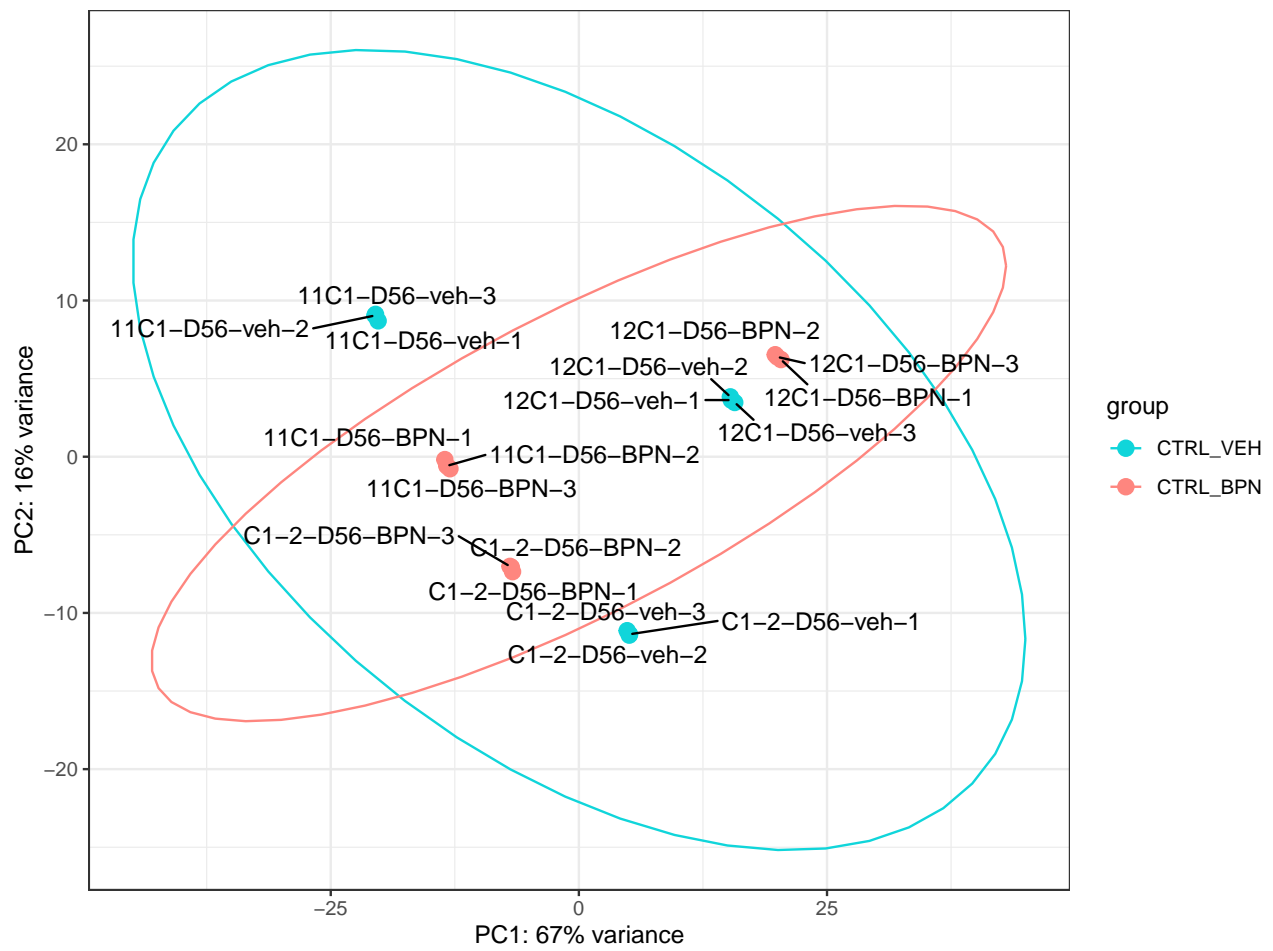
```
# Plot the sample distance
plot_sample_heatmap(dds_obj, figure_folder =
                    file.path(result_folder,"01-Sample_info"),
                    file_name = "01_sample_distance_heatmap")
```



```
## [1] "Sample distance heatmap is done"
```

```
# Plot the PCA plot for the sample
plot_sample_PCA_plot(dds_obj,figure_folder = file.path(result_folder,"01-Sample_info"),
                    file_name = "02_sample_PCA_plot",
                    reference_group, compare_group)
```

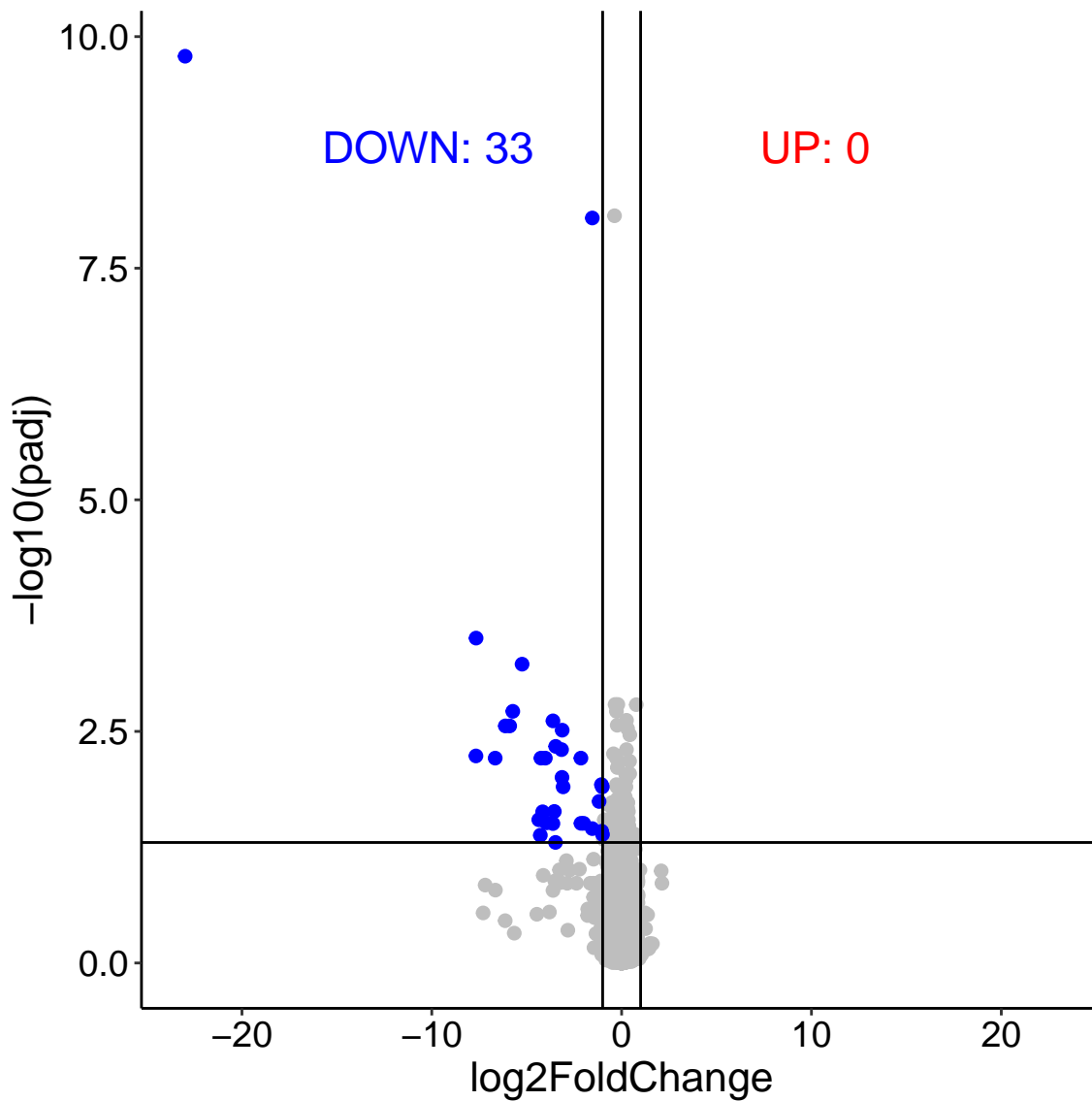
```
## [1] "PCA plot is done"
```



(2) DEG visualization - Volcano plot and Heatmap

```
result_df <- results(dds_obj) %>%  
  as.data.frame() %>%  
  rownames_to_column(var = "GeneName") %>%  
  dplyr::select(GeneName, everything()) %>%  
  filter(!is.na(padj)) %>% # Correct way to filter non-NA values  
  arrange(padj)  
  
# Plot the volcano plot for the DEG  
plot_volcano_plot(result_df=result_df,  
  figure_folder = file.path(result_folder,"02-DEG"),  
  file_name = "02_volcano_plot_log2fc_1",  
  thread = 1 , dot_size =2, label_gene = NULL)
```

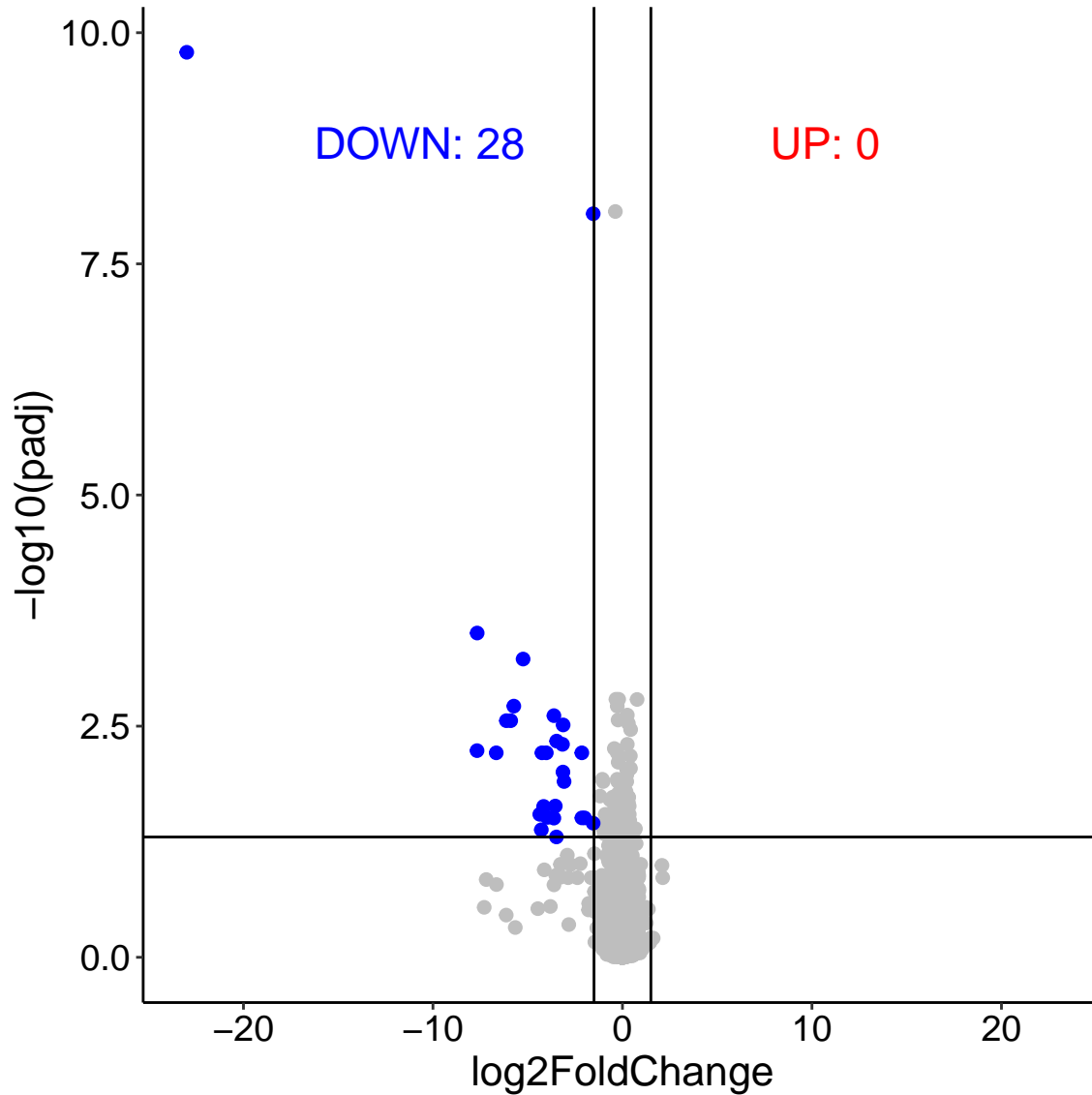
```
## [1] "Volcano plot for 02_volcano_plot_log2fc_1"
```



```
plot_volcano_plot(result_df=result_df,
```

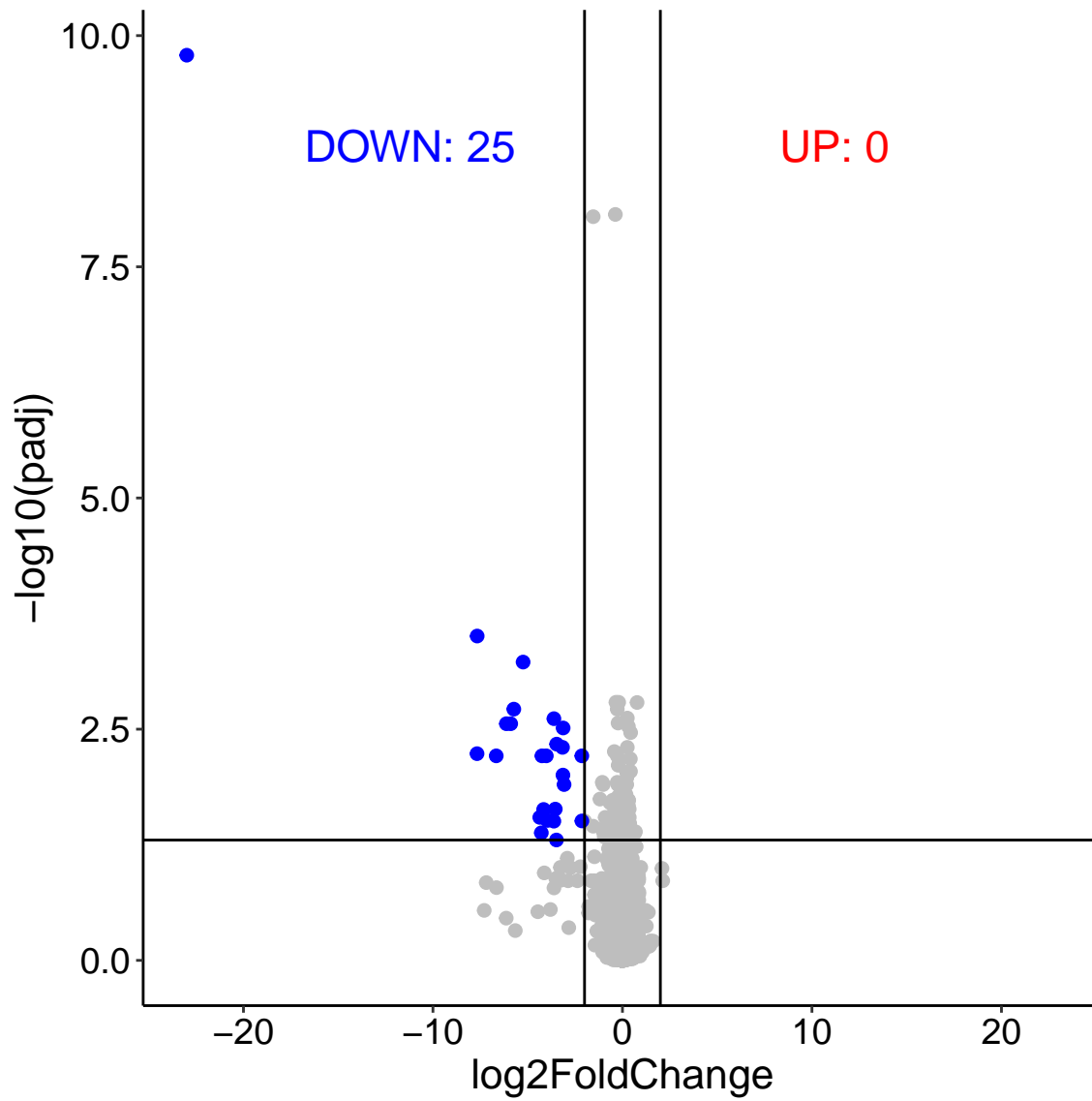
```
figure_folder = file.path(result_folder,"02-DEG"),
file_name = "03_volcano_plot_log2fc_1.5",
thread = 1.5 , dot_size =2,label_gene = NULL)
```

```
## [1] "Volcano plot for 03_volcano_plot_log2fc_1.5"
```



```
plot_volcano_plot(result_df=result_df,
figure_folder = file.path(result_folder,"02-DEG"),
file_name = "03_volcano_plot_log2fc_2",
thread = 2 , dot_size =2,label_gene = NULL)
```

```
## [1] "Volcano plot for 03_volcano_plot_log2fc_2"
```

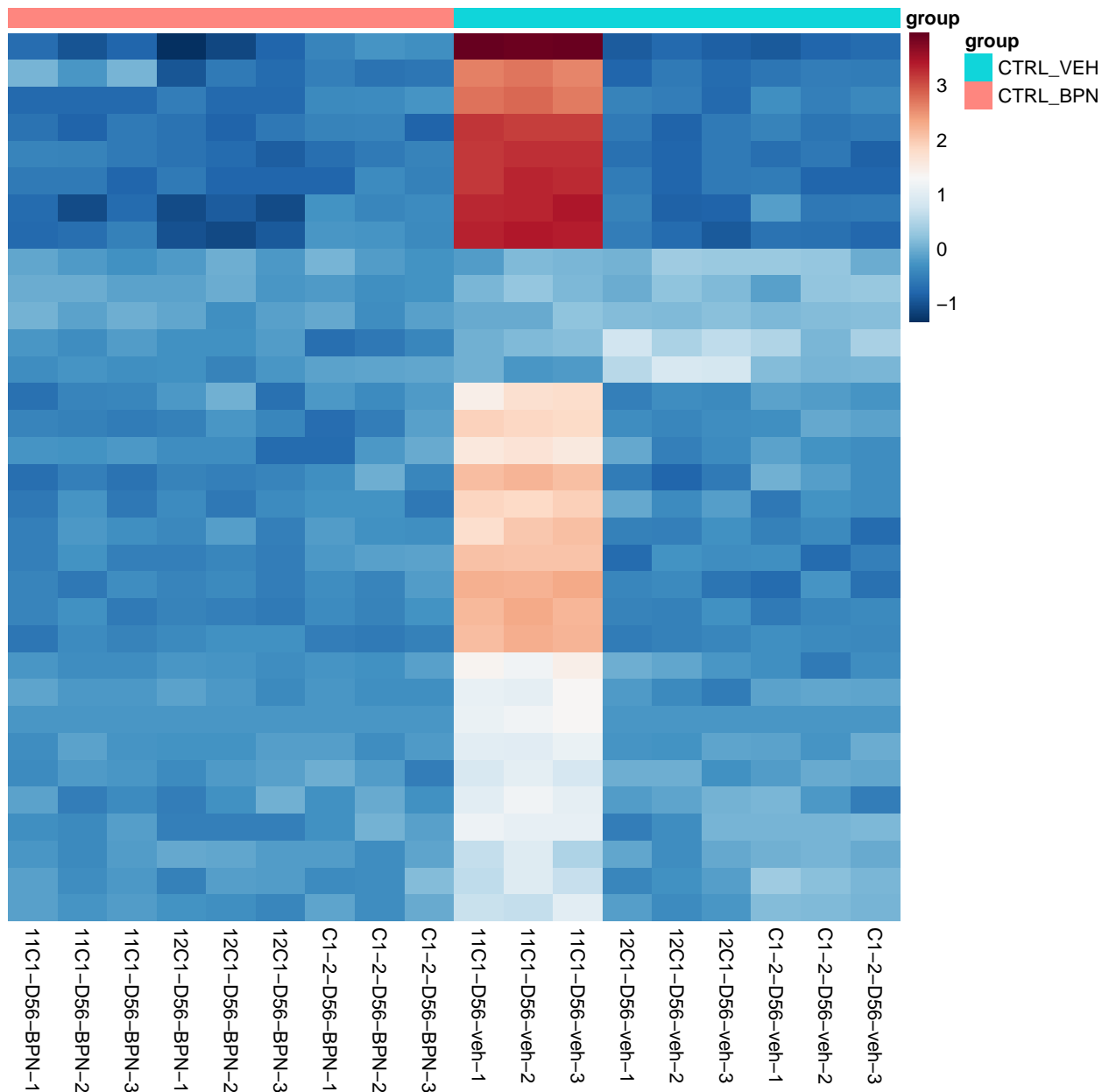


```
# Plot the heatmap for the DEG
vsd_obj <- varianceStabilizingTransformation(dds_obj, blind = TRUE)

DEG_gene_1 <- result_df %>% filter(abs(log2FoldChange) > 1 & padj < 0.05) %>% pull(GeneName)
DEG_gene_1.5 <- result_df %>% filter(abs(log2FoldChange) > 1.5 & padj < 0.05) %>% pull(GeneName)
DEG_gene_2 <- result_df %>% filter(abs(log2FoldChange) > 2 & padj < 0.05) %>% pull(GeneName)

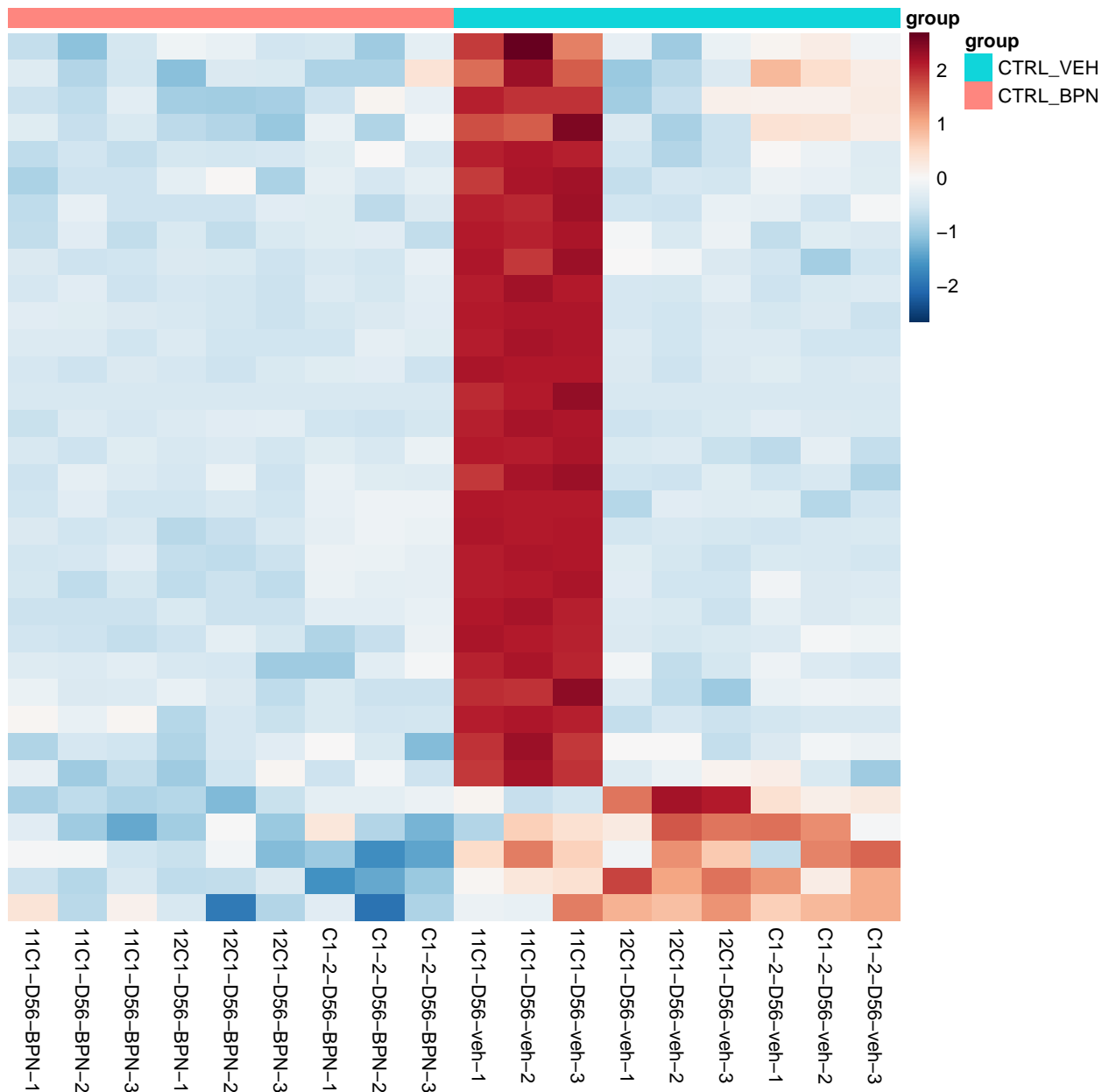
plot_gene_heatmap(vsd_obj, gene_list = DEG_gene_1,
  figure_folder = file.path(result_folder, "O2-DEG"),
  file_name = "O2_heatmap_log2fc_1",
  reference_group, compare_group,
  cluster_rows = TRUE, cluster_cols = FALSE,
  scale = "none")

## [1] "Heatmap for O2_heatmap_log2fc_1 "
```

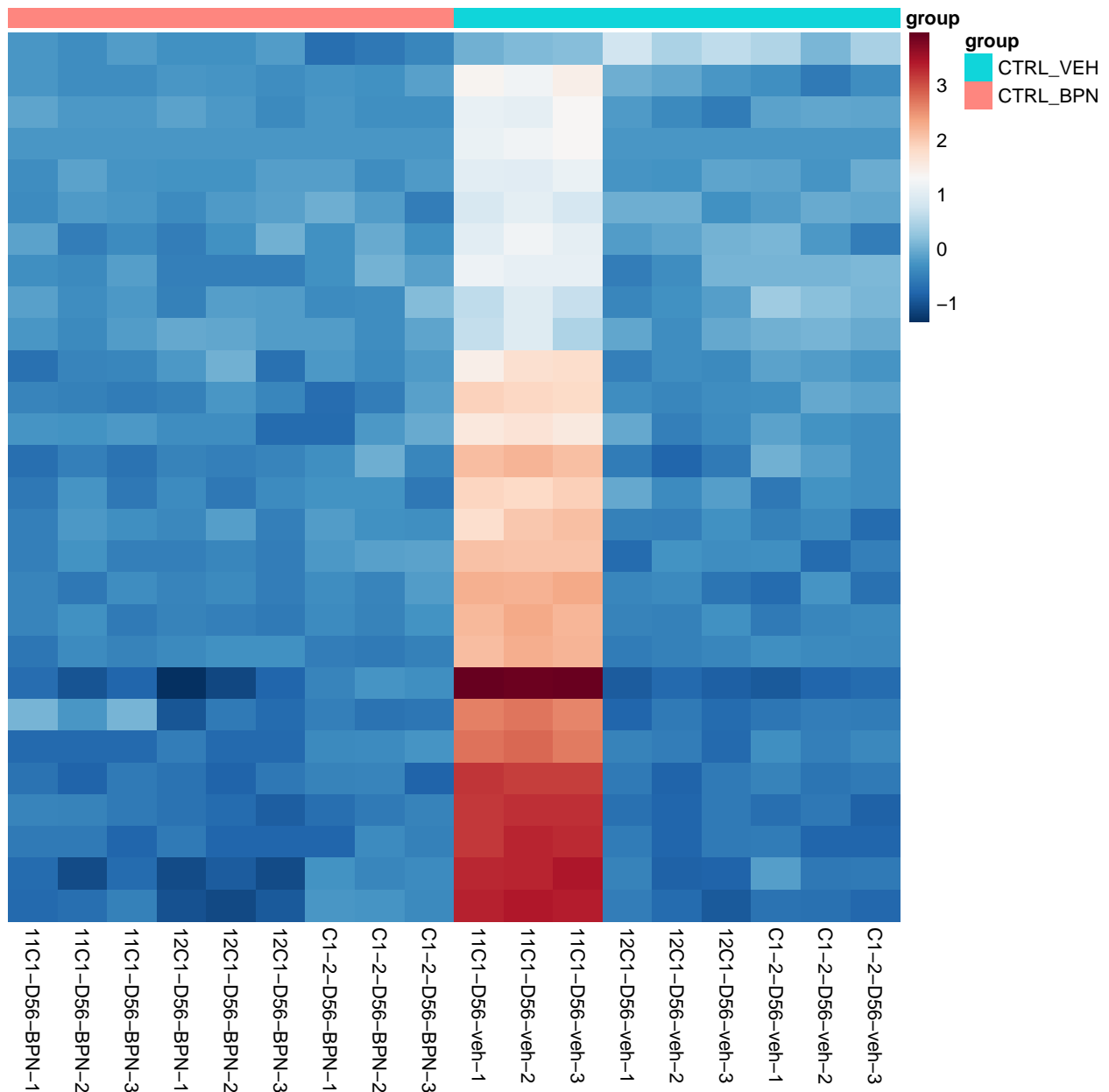
```
plot_gene_heatmap(vsd_obj, gene_list = DEG_gene_1,
                  figure_folder = file.path(result_folder, "02-DEG"),
                  file_name = "02_heatmap_log2fc_1_row",
                  reference_group, compare_group,
                  cluster_rows = TRUE, cluster_cols = FALSE,
                  scale = "row")
```

```
## [1] "Heatmap for 02_heatmap_log2fc_1_row "
```



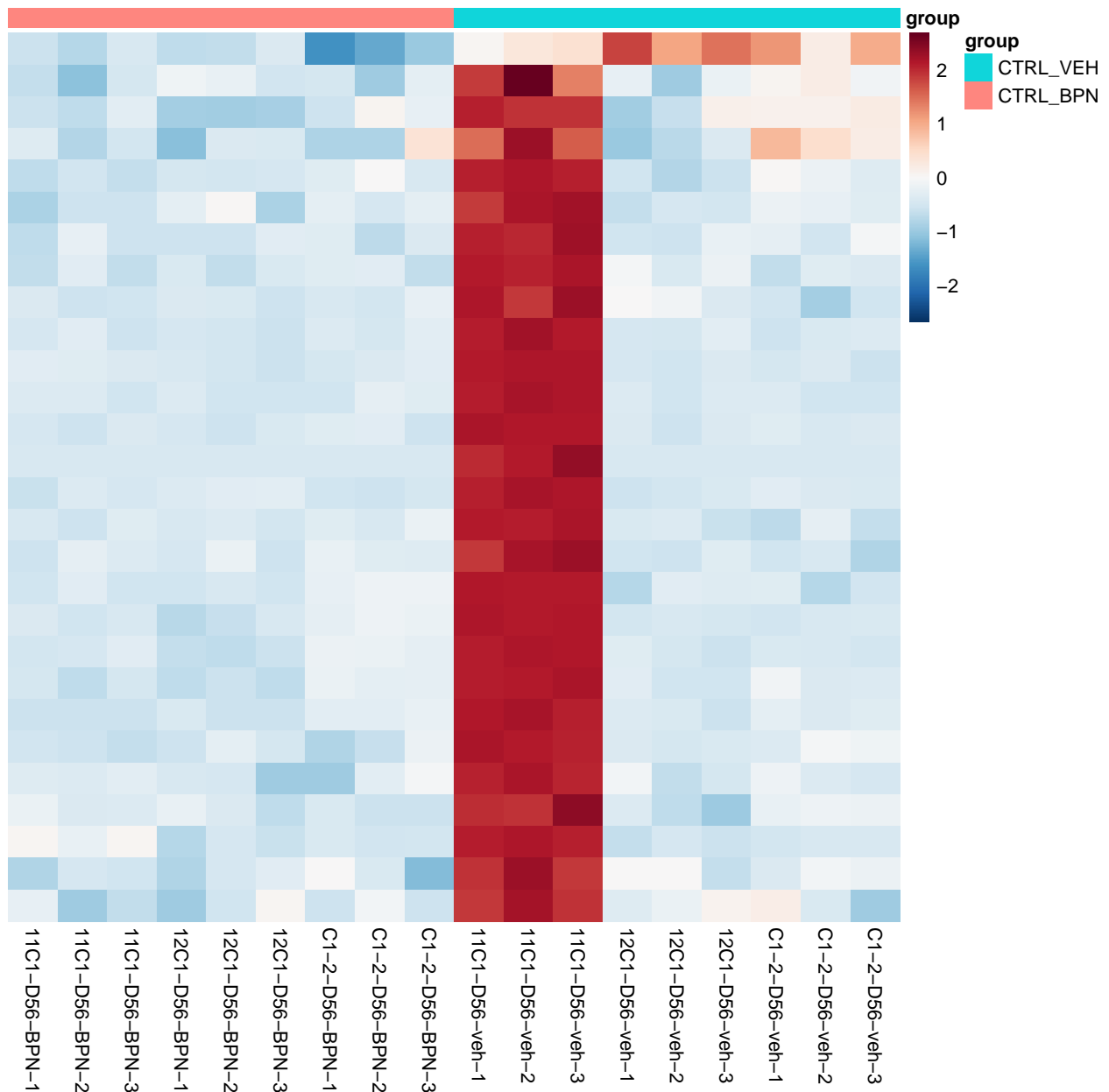
```
plot_gene_heatmap(vsd_obj, gene_list = DEG_gene_1.5,
                  figure_folder = file.path(result_folder, "02-DEG"),
                  file_name = "03_heatmap_log2fc_1.5",
                  reference_group, compare_group,
                  cluster_rows = TRUE, cluster_cols = FALSE,
                  scale = "none")
```

```
## [1] "Heatmap for 03_heatmap_log2fc_1.5 "
```



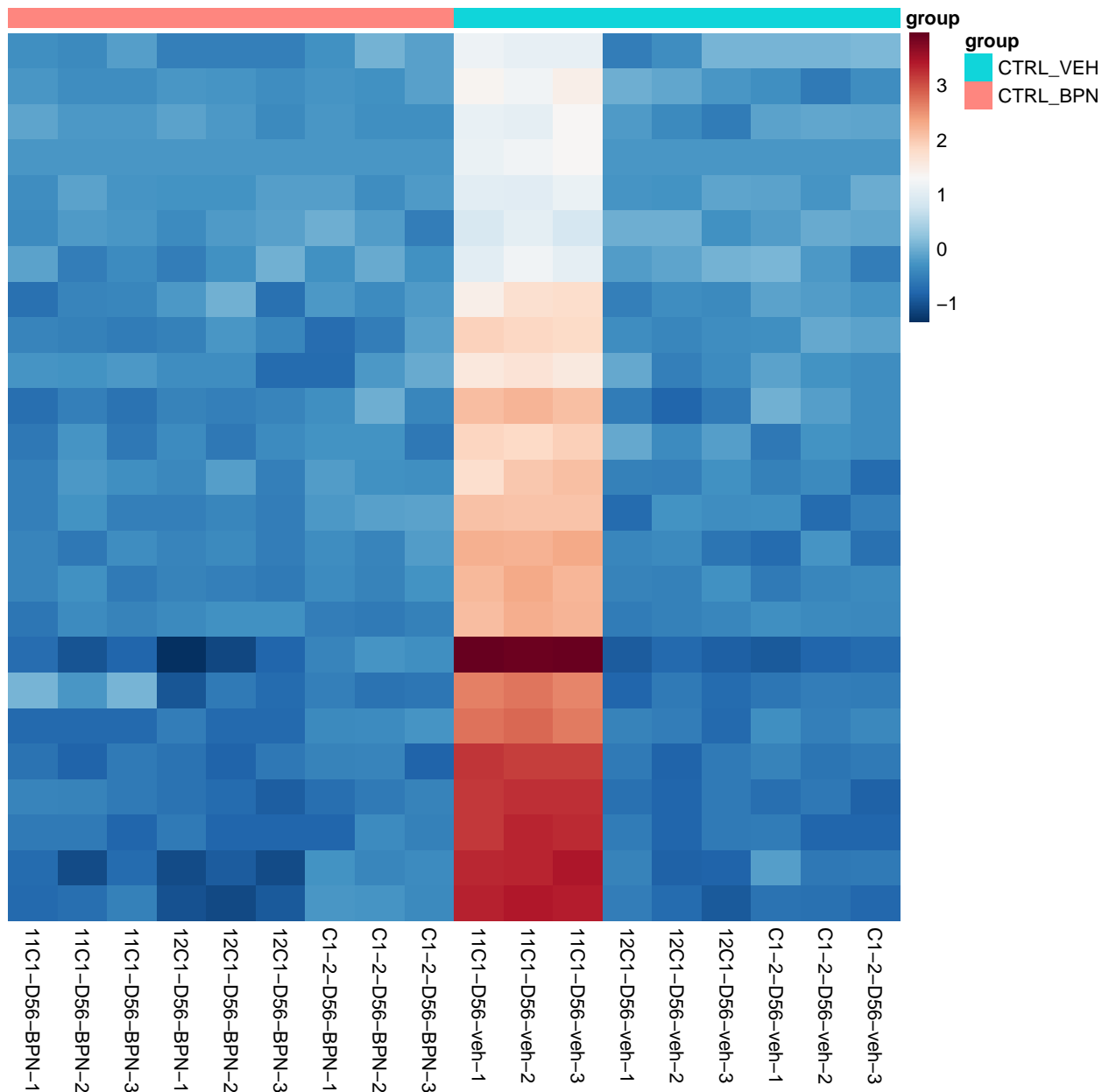
```
plot_gene_heatmap(vsd_obj, gene_list = DEG_gene_1.5,
                  figure_folder = file.path(result_folder, "02-DEG"),
                  file_name = "03_heatmap_log2fc_1.5_row",
                  reference_group, compare_group,
                  cluster_rows = TRUE, cluster_cols = FALSE,
                  scale = "row")
```

```
## [1] "Heatmap for 03_heatmap_log2fc_1.5_row "
```



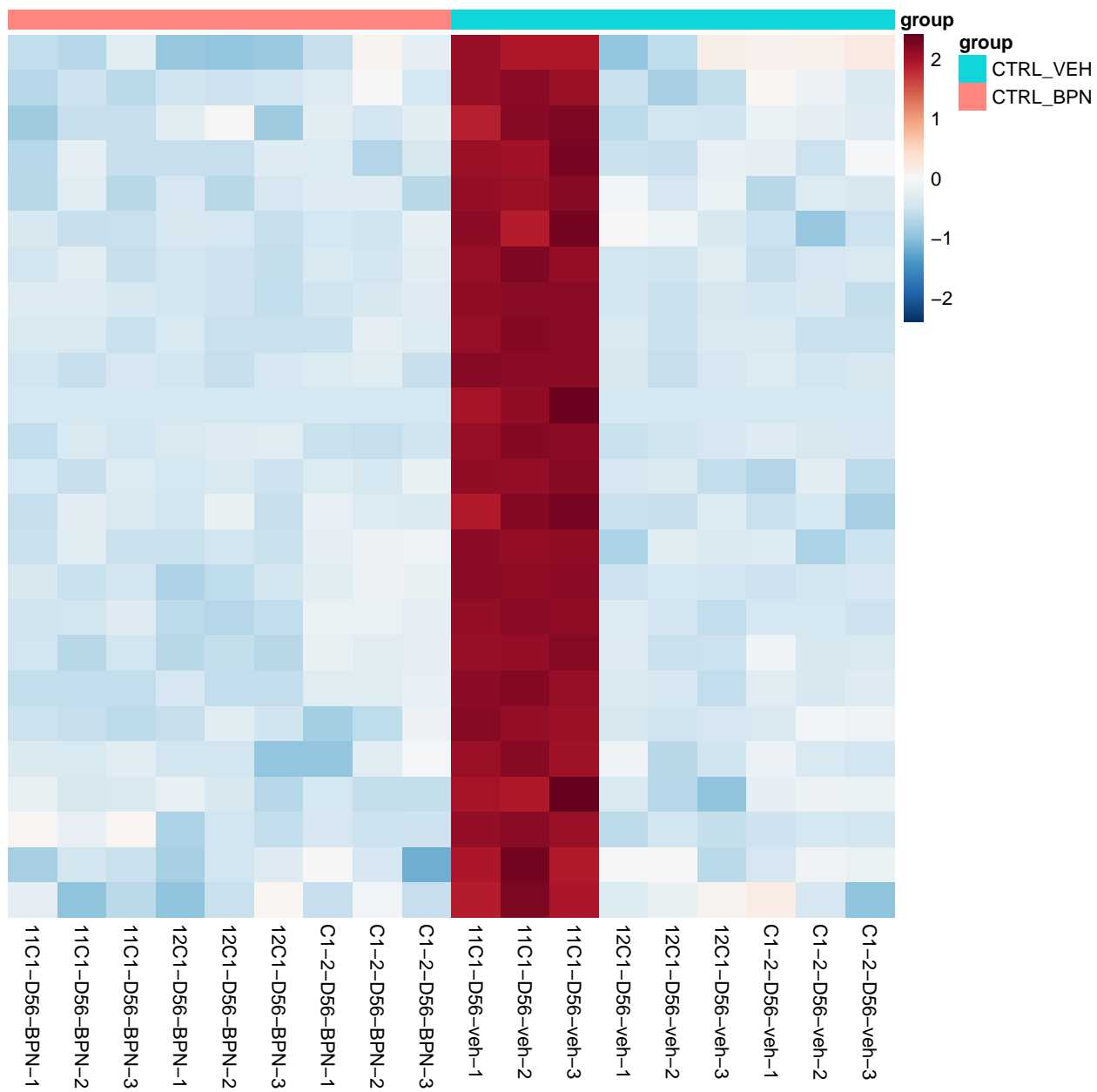
```
plot_gene_heatmap(vsd_obj, gene_list = DEG_gene_2,
                  figure_folder = file.path(result_folder, "02-DEG"),
                  file_name = "03_heatmap_log2fc_2",
                  reference_group, compare_group,
                  cluster_rows = TRUE, cluster_cols = FALSE,
                  scale = "none")
```

```
## [1] "Heatmap for 03_heatmap_log2fc_2 "
```



```
plot_gene_heatmap(vsd_obj, gene_list = DEG_gene_2,
                  figure_folder = file.path(result_folder, "02-DEG"),
                  file_name = "03_heatmap_log2fc_2_row",
                  reference_group, compare_group,
                  cluster_rows = TRUE, cluster_cols = FALSE,
                  scale = "row")
```

```
## [1] "Heatmap for 03_heatmap_log2fc_2_row "
```



4. GSVA analysis

```
# The following code is used to generate the GSVA matrix , only need to run once
gmxFFile <- here::here("data", "ref", "c5.go.v2023.1.Hs.symbols.gmt")
go_list <- getGmt(gmxFFile)

geneset <- go_list
dat <- as.matrix(counts)

gsvaparam <- gsvaParam(dat, geneset, maxDiff=TRUE)
gsva_es <- gsva(gsvaparam)

## No annotation package name available in the input data object.
## Attempting to directly match identifiers in data to gene sets.

## Estimating GSVA scores for 10532 gene sets.
## Estimating ECDFs with Gaussian kernels
## |

gsva_matrix <- as.data.frame(gsva_es)

# save the result
write.csv(gsva_matrix, file.path(result_folder, "04-GSVA", "01_GSVA_matrix.csv"))

gsva_matrix <- read.csv(file.path(result_folder, "04-GSVA", "01_GSVA_matrix.csv"),
                      row.names = 1)

colnames(gsva_matrix) <- sub("^X", "", colnames(gsva_matrix))
condition_list_label <- condition_list %>%
  filter(group %in% c(reference_group, compare_group)) %>%
  mutate(group = case_when(
    group == reference_group ~ reference_group_short,
    group == compare_group ~ compare_group_short,
    TRUE ~ group # fallback just in case
  ))

# plot the heatmap for the GSVA result
pathway_list <- read.csv(here::here("data", "ref", "focus-pathway_2024_10_03.csv"))

# # plot for all pathway
# for (i in 1:nrow(pathway_list)){
#   if (i %% 10 == 0) print(i)
#   pathway_name <- pathway_list$pathway[i]
#   plot_gsva_boxplot(gsva_matrix,
#                     condition_list_label = condition_list_label,
#                     pathway_name = pathway_name,
#                     figure_folder = file.path(result_folder, "04-GSVA", "Boxplot"),
#                     file_name = paste0("GSVA_", pathway_name),
#                     fig.height = 6, fig.width = 4,
#                     reference_group = reference_group_short ,
#                     compare_group = compare_group_short)
```

```

#
# }

box_plot_folder<- file.path(result_folder,"04-GSVA","Boxplot")
# create the folder
dir.create(box_plot_folder, showWarnings = FALSE)

# plot for the focus pathway
for (i in 1:2){
  pathway_name <- pathway_list$pathway[i]
  print(pathway_name)
  p<-plot_gsva_boxplot(gsva_matrix,
                       condition_list_label =condition_list_label,
                       pathway_name = pathway_name,
                       figure_folder = file.path(result_folder,"04-GSVA","Boxplot"),
                       file_name = paste0("GSVA_", pathway_name),
                       fig.height = 6, fig.width = 4,
                       reference_group =reference_group_short ,
                       compare_group = compare_group_short)
  print(p)
}

```

```
## [1] "GOBP_MITOCHONDRIAL_RNA_3_END_PROCESSING"
```

```
## [1] "GOBP_MITOCHONDRIAL_RNA_PROCESSING"
```

Mitochondrial RNA 3 end processing

Mitochondrial RNA processing

GSVA Score

GSVA Score

5. Pathway Enrichment Analysis

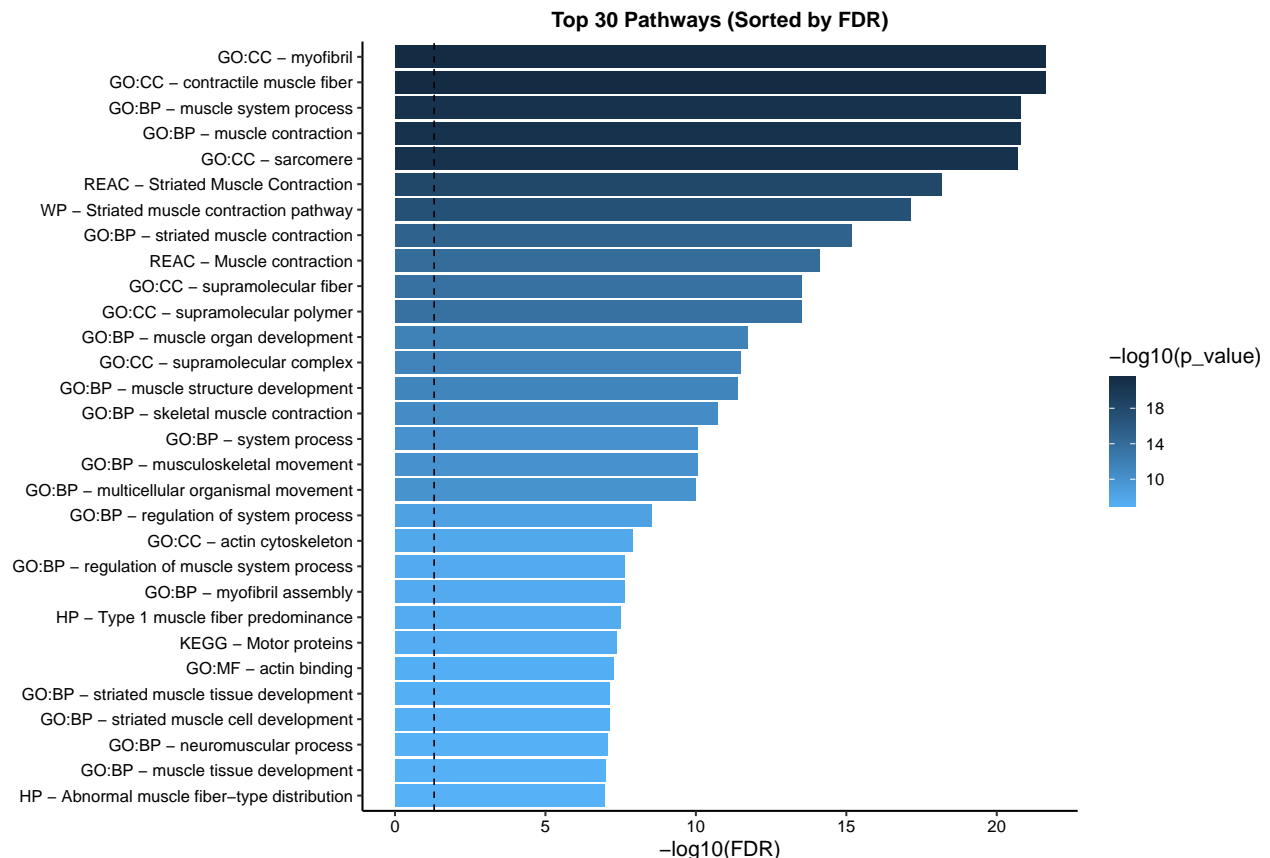
```
deg1 <- result_df %>% filter(padj < 0.05 & abs(log2FoldChange) > 1)
up_gene_1 <- deg1 %>% filter(log2FoldChange > 0) %>% pull(GeneName)
down_gene_1 <- deg1 %>% filter(log2FoldChange < 0) %>% pull(GeneName)

# test for the function
if (length(up_gene_1) == 0) {
  print("The up_gene_1 is empty, skip the analysis")
} else {
  Enrichment_analysis(gene_list = up_gene_1,
    result_folder = file.path(result_folder, "03-Enrichment"),
    file_name = "02-DEG_1_up", gene_name_mapping, flag = "Up")
}
```

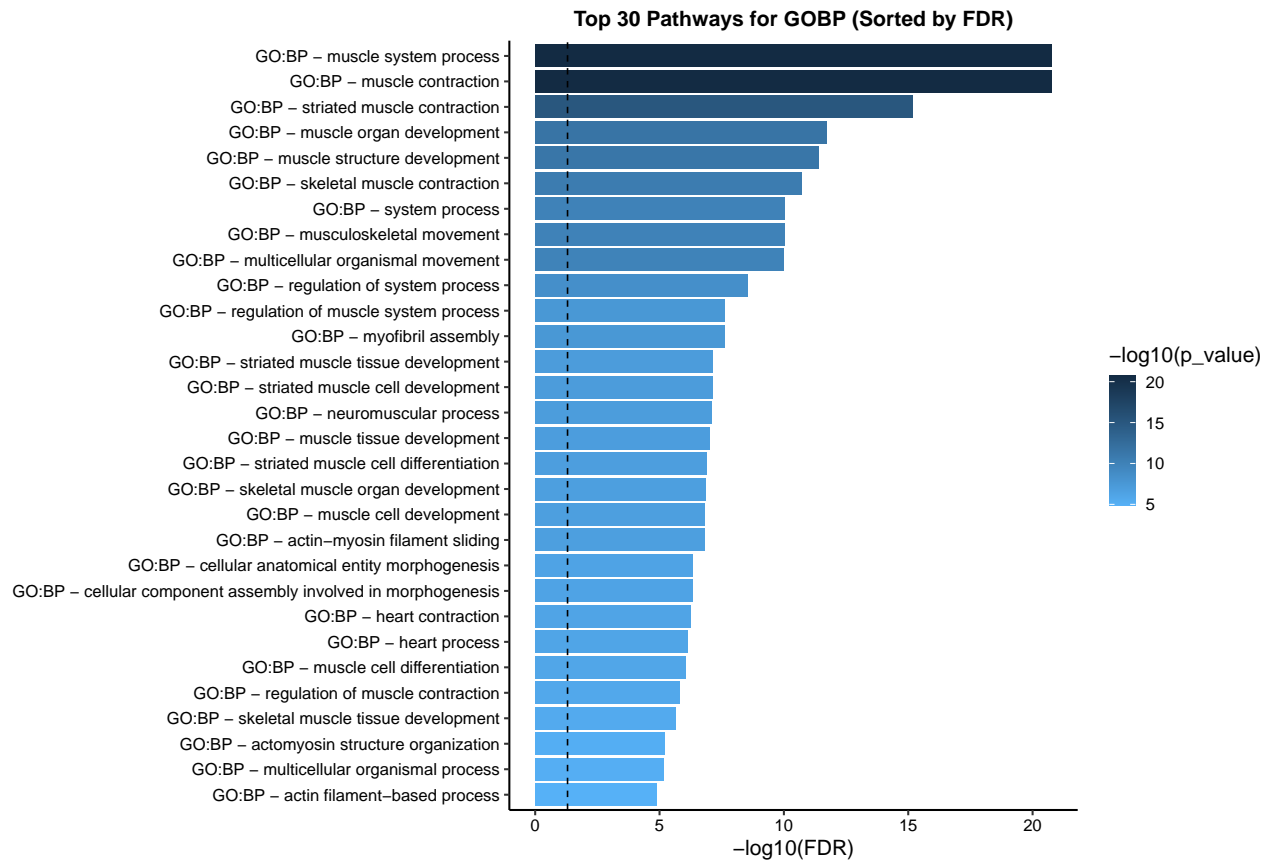
```
## [1] "The up_gene_1 is empty, skip the analysis"
```

```
if (length(down_gene_1) == 0) {
  print("The down_gene_1 is empty, skip the analysis")
} else {
  Enrichment_analysis(gene_list = down_gene_1,
    result_folder = file.path(result_folder, "03-Enrichment"),
    file_name = "02-DEG_1_down", gene_name_mapping, flag = "Down")
}
```

```
## [1] "Enrichment analysis for 02-DEG_1_down "
```



[1] "Enrichment analysis for GOBP 02-DEG_1_down "



```

deg1.5 <- result_df %>% filter(padj < 0.05 & abs(log2FoldChange) > 1.5)
up_gene_1.5 <- deg1.5 %>% filter(log2FoldChange > 0) %>% pull(GeneName)
down_gene_1.5 <- deg1.5 %>% filter(log2FoldChange < 0) %>% pull(GeneName)

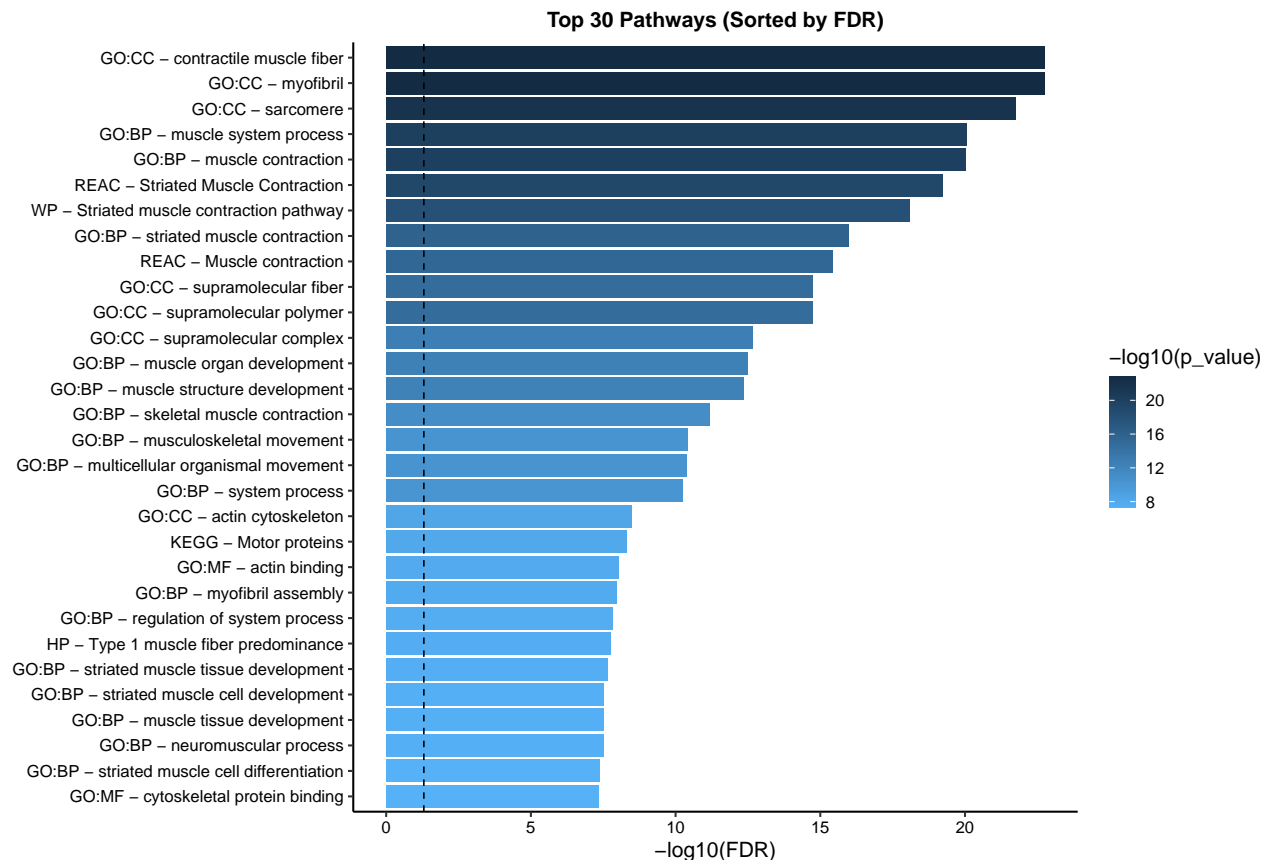
# test for the function
# if the up_gene_1.5 is empty, we will skip the analysis
if (length(up_gene_1.5) == 0) {
  print("The up_gene_1.5 is empty, skip the analysis")
} else {
  Enrichment_analysis(gene_list = up_gene_1.5,
    result_folder = file.path(result_folder, "03-Enrichment"),
    file_name = "02-DEG_1.5_up", gene_name_mapping, flag = "Up")
}

## [1] "The up_gene_1.5 is empty, skip the analysis"

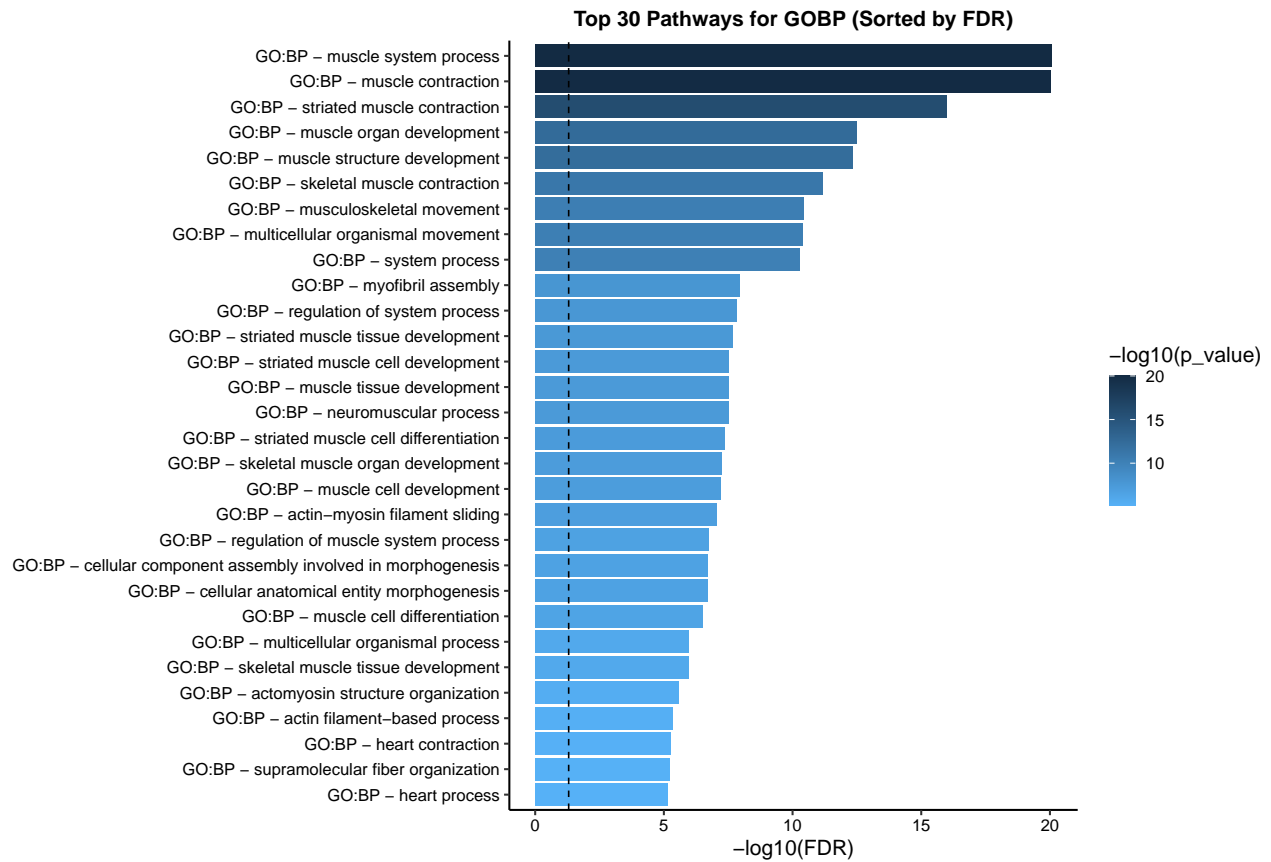
if (length(down_gene_1.5) == 0) {
  print("The down_gene_1.5 is empty, skip the analysis")
} else {
  Enrichment_analysis(gene_list = down_gene_1.5,
    result_folder = file.path(result_folder, "03-Enrichment"),
    file_name = "02-DEG_1.5_down", gene_name_mapping, flag = "Down")
}

## [1] "Enrichment analysis for 02-DEG_1.5_down "

```



```
## [1] "Enrichment analysis for GOBP 02-DEG_1.5_down "
```

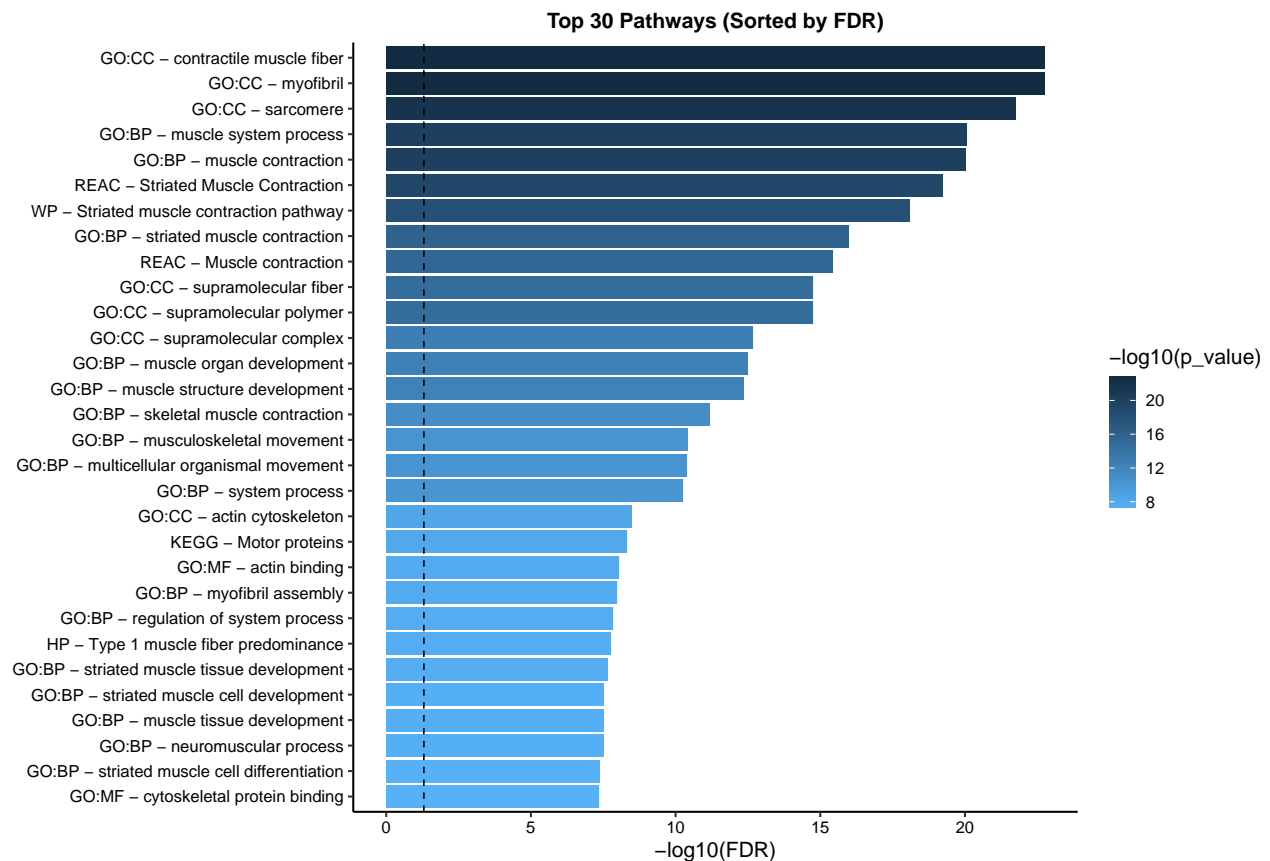


```
deg2 <- result_df %>% filter(padj < 0.05 & abs(log2FoldChange) > 1.5)
up_gene_2 <- deg2 %>% filter(log2FoldChange > 0) %>% pull(GeneName)
down_gene_2 <- deg2 %>% filter(log2FoldChange < 0) %>% pull(GeneName)
result_folder = result_folder_all
# test for the function
if (length(up_gene_2) == 0) {
  print("The up_gene_2 is empty, skip the analysis")
} else {
  Enrichment_analysis(gene_list = up_gene_2,
    result_folder = file.path(result_folder,"03-Enrichment"),
    file_name = "02-DEG_2_up", gene_name_mapping, flag = "Up")
}
```

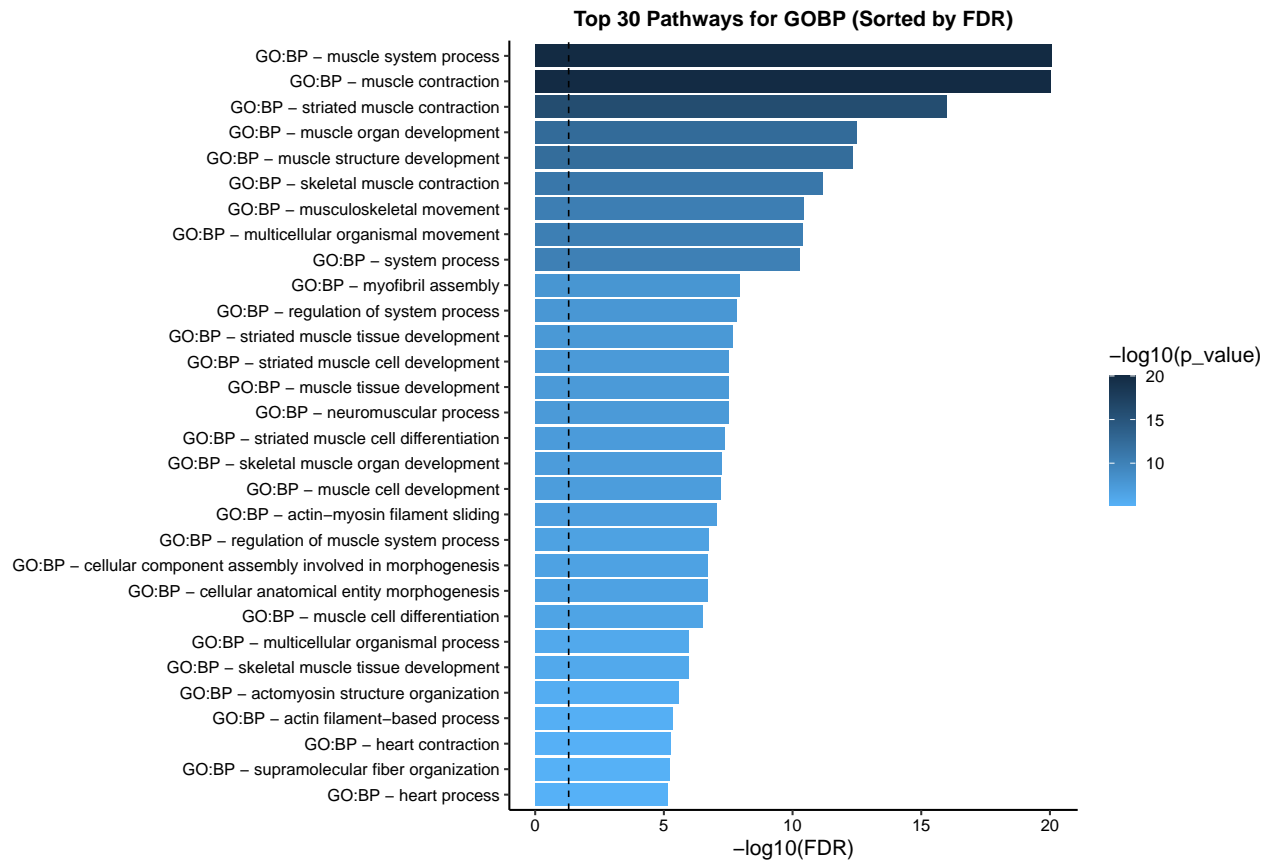
```
## [1] "The up_gene_2 is empty, skip the analysis"
```

```
if (length(down_gene_2) == 0) {
  print("The down_gene_2 is empty, skip the analysis")
} else {
  Enrichment_analysis(gene_list = down_gene_2,
    result_folder = file.path(result_folder,"03-Enrichment"),
    file_name = "02-DEG_2_down", gene_name_mapping, flag = "Down")
}
```

```
## [1] "Enrichment analysis for 02-DEG_2_down "
```



```
## [1] "Enrichment analysis for GOBP 02-DEG_2_down "
```



Session information

```
sessionInfo()
```

```
## R version 4.4.0 (2024-04-24)
## Platform: aarch64-apple-darwin20
## Running under: macOS Sonoma 14.3.1
##
## Matrix products: default
## BLAS:   /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib; LAPACK v
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## time zone: America/New_York
## tzcode source: internal
##
## attached base packages:
## [1] parallel stats4 stats graphics grDevices utils datasets
## [8] methods base
##
## other attached packages:
## [1] GSEABase_1.66.0 graph_1.82.0
## [3] annotate_1.82.0 XML_3.99-0.18
## [5] extrafont_0.19 ggsignif_0.6.4
## [7] patchwork_1.3.0 decoupleR_2.10.0
## [9] GSVA_1.52.3 BiocParallel_1.38.0
## [11] edgeR_4.2.2 limma_3.60.6
## [13] GenomicFeatures_1.56.0 biomaRt_2.60.1
## [15] gprofiler2_0.2.3 RColorBrewer_1.1-3
## [17] data.table_1.16.4 org.Hs.eg.db_3.19.1
## [19] AnnotationDbi_1.66.0 clusterProfiler_4.12.6
## [21] ggfortify_0.4.17 pheatmap_1.0.12
## [23] EnhancedVolcano_1.22.0 ggrepel_0.9.6
## [25] apeglm_1.26.1 DESeq2_1.44.0
## [27] SummarizedExperiment_1.34.0 Biobase_2.64.0
## [29] MatrixGenerics_1.16.0 matrixStats_1.5.0
## [31] reshape2_1.4.4 Matrix_1.7-2
## [33] Signac_1.14.0 Seurat_5.2.1
## [35] SeuratObject_5.0.2 sp_2.2-0
## [37] rtracklayer_1.64.0 GenomicRanges_1.56.2
## [39] GenomeInfoDb_1.40.1 IRanges_2.38.1
## [41] S4Vectors_0.42.1 BiocGenerics_0.50.0
## [43] knitr_1.49 lubridate_1.9.4
## [45] forcats_1.0.0 stringr_1.5.1
## [47] dplyr_1.1.4 purrr_1.0.4
## [49] readr_2.1.5 tidyr_1.3.1
## [51] tibble_3.2.1 ggplot2_3.5.1
## [53] tidyverse_2.0.0
##
## loaded via a namespace (and not attached):
## [1] SpatialExperiment_1.14.0 R.methodsS3_1.8.2
## [3] progress_1.2.3 goftest_1.2-3
```

## [5] HDF5Array_1.32.1	Biostrings_2.72.1
## [7] vctrs_0.6.5	spatstat.random_3.3-2
## [9] digest_0.6.37	png_0.1-8
## [11] deldir_2.0-4	parallelly_1.42.0
## [13] magick_2.8.5	MASS_7.3-64
## [15] httpuv_1.6.15	qvalue_2.36.0
## [17] withr_3.0.2	xfun_0.51
## [19] ggfun_0.1.8	survival_3.8-3
## [21] memoise_2.0.1	gson_0.1.0
## [23] systemfonts_1.2.1	ragg_1.3.3
## [25] tidytree_0.4.6	zoo_1.8-12
## [27] pbapply_1.7-2	R.oo_1.27.0
## [29] prettyunits_1.2.0	KEGGREST_1.44.1
## [31] promises_1.3.2	httr_1.4.7
## [33] restfulr_0.0.15	rhdf5filters_1.16.0
## [35] globals_0.16.3	fitdistrplus_1.2-2
## [37] rhdf5_2.48.0	rstudioapi_0.17.1
## [39] UCSC.utils_1.0.0	miniUI_0.1.1.1
## [41] generics_0.1.3	DOSE_3.30.5
## [43] curl_6.2.1	zlibbioc_1.50.0
## [45] ScaledMatrix_1.12.0	ggraph_2.2.1
## [47] polyclip_1.10-7	GenomeInfoDbData_1.2.12
## [49] SparseArray_1.4.8	xtable_1.8-4
## [51] evaluate_1.0.3	S4Arrays_1.4.1
## [53] BiocFileCache_2.12.0	hms_1.1.3
## [55] irlba_2.3.5.1	colorspace_2.1-1
## [57] filelock_1.0.3	ROCR_1.0-11
## [59] reticulate_1.40.0	spatstat.data_3.1-4
## [61] magrittr_2.0.3	lmtest_0.9-40
## [63] later_1.4.1	viridis_0.6.5
## [65] ggtree_3.12.0	lattice_0.22-6
## [67] spatstat.geom_3.3-5	future.apply_1.11.3
## [69] scattermore_1.2	shadowtext_0.1.4
## [71] cowplot_1.1.3	RcppAnnoy_0.0.22
## [73] pillar_1.10.1	nlme_3.1-167
## [75] compiler_4.4.0	beachmat_2.20.0
## [77] RSpectra_0.16-2	stringi_1.8.4
## [79] tensor_1.5	GenomicAlignments_1.40.0
## [81] plyr_1.8.9	crayon_1.5.3
## [83] abind_1.4-8	BiocIO_1.14.0
## [85] gridGraphics_0.5-1	emdbbook_1.3.13
## [87] locfit_1.5-9.11	graphlayouts_1.2.2
## [89] bit_4.5.0.1	fastmatch_1.1-6
## [91] textshaping_1.0.0	codetools_0.2-20
## [93] BiocSingular_1.20.0	plotly_4.10.4
## [95] mime_0.12	splines_4.4.0
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## [99] sparseMatrixStats_1.16.0	dbplyr_2.5.0
## [101] Rttf2pt1_1.3.12	blob_1.2.4
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## [105] listenv_0.9.1	ggplotify_0.1.2
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## [111] tools_4.4.0	cachem_1.1.0

## [113] RSQLite_2.3.9	viridisLite_0.4.2
## [115] DBI_1.2.3	numDeriv_2016.8-1.1
## [117] fastmap_1.2.0	rmarkdown_2.29
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## [131] lifecycle_1.0.4	uwot_0.2.2
## [133] mvtnorm_1.3-3	timechange_0.3.0
## [135] gtable_0.3.6	rjson_0.2.23
## [137] ggthemes_0.5.6	progressr_0.15.1
## [139] ape_5.8-1	jsonlite_1.9.0
## [141] RcppHNSW_0.6.0	bitops_1.0-9
## [143] bit64_4.6.0-1	Rtsne_0.17
## [145] yulab.utils_0.2.0	spatstat.utils_3.1-2
## [147] bdsmatrix_1.3-7	GOSemSim_2.30.2
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## [151] lazyeval_0.2.2	shiny_1.10.0
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## [155] GO.db_3.19.1	sctransform_0.4.1
## [157] rappdirs_0.3.3	tinytex_0.55
## [159] glue_1.8.0	spam_2.11-1
## [161] httr2_1.1.0	XVector_0.44.0
## [163] RCurl_1.98-1.16	rprojroot_2.0.4
## [165] treeio_1.28.0	gridExtra_2.3
## [167] extrafontdb_1.0	igraph_2.1.4
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## [171] labeling_0.4.3	RcppRoll_0.3.1
## [173] cluster_2.1.8	bbmle_1.0.25.1
## [175] Rhdf5lib_1.26.0	aplot_0.2.4
## [177] DelayedArray_0.30.1	tidyselect_1.2.1
## [179] ggforce_0.4.2	xml2_1.3.6
## [181] future_1.34.0	rsvd_1.0.5
## [183] munsell_0.5.1	KernSmooth_2.23-26
## [185] htmlwidgets_1.6.4	fgsea_1.30.0
## [187] rlang_1.1.5	spatstat.sparse_3.1-0
## [189] spatstat.explore_3.3-4	