

Comparative functional enrichment analysis between proteins annotated as displaying intensity variation in one of the four ciliary locations and annotated as stable

Konstantin Kahnert

2025-05-13

Load libraries

```
library(tidyverse)
library(clusterProfiler)
library(enrichplot)
library(org.Hs.eg.db)
library(Cairo)
library(svglite)
library(VennDiagram)
library(grid)
library(simplifyEnrichment)
library(readxl)
```

Define functions

```
# map_cluster_number:
# - x: enrichResult or compareClusterResult (from clusterProfiler)
# - df: simplifyGO result data.frame with columns ID (term IDs) and Cluster_num
# - comp: if TRUE, use x@compareClusterResult, else x@result
# Returns a data.frame of all enrich columns + term_size + Cluster_num
map_cluster_number <- function(x,
                               df,
                               comp = FALSE) {
  ## 1. Standardize the input DF's column names
  colnames(df) <- c("ID", "Cluster_num")

  ## 2. Select the correct slot from the clusterProfiler object
  if (comp) {
    # from compareClusterResult
    results <- x@compareClusterResult
  } else {
    # from a single enrichResult
    results <- x@result
  }
}
```

```

}

## 3. Get term_size (the numerator of the "BgRatio" string)
## e.g. "12/21273" → 12
results$term_size <- as.numeric(
  sapply(strsplit(results$BgRatio, "/"),
    function(parts) as.numeric(parts[1]))
)

## 4. Merge the enrichment results with the cluster assignments
merged_results <- merge(
  results,
  df,
  by = "ID",
  all.x = TRUE
)

## 5. Return the processed data frame
return(merged_results)
}

# Function to map Ensembl IDs to gene symbols for a single row
map_geneID_to_symbol <- function(geneID_str) {
  geneIDs <- unlist(strsplit(geneID_str, "/"))
  mapped_ids <- bitr(geneIDs,
    fromType = "ENSEMBL",
    toType = "SYMBOL",
    OrgDb = org.Hs.eg.db)

  # drop duplicates in ensembl column and keep first occurrence
  mapped_ids <- mapped_ids[!duplicated(mapped_ids$ENSEMBL),]

  gene_symbols <- mapped_ids$SYMBOL
  return(paste(gene_symbols, collapse = "/"))
}

# Function to map Ensembl IDs to gene symbols for a single row
map_geneID_to_name <- function(geneIDs) {
  mapped_ids <- bitr(geneIDs,
    fromType = "ENSEMBL",
    toType = c("GENENAME", "SYMBOL"),
    OrgDb = org.Hs.eg.db)

  # drop duplicates in ensembl column and keep first occurrence
  mapped_ids <- mapped_ids[!duplicated(mapped_ids$ENSEMBL),]

  return(mapped_ids)
}

```

Data loading and preprocessing

Set input and output paths

```
# set input and output paths
in_path <- "/mnt/Data/Projects/Cilia/revision/NonRestricted/data/"
out_path <- "/mnt/Data/Projects/Cilia/revision/NonRestricted/analysis/GO_BP/"
```

Load data

```
# Load the data
df_all <- read.delim(paste0(in_path, "All_files_combined_as_cytoscape_input.csv"), sep = "\t", header =
head(df_all))
```

```
## BasalBody_ BasalBody_ASC52telo
## ENSG00000001497 True False
## ENSG00000002330 False False
## ENSG00000002549 False False
## ENSG00000003249 False False
## ENSG00000003756 False False
## ENSG00000004766 True False
## BasalBody_hTERT_RPE1_serum_starved BasalBody_RPTEC_TERT1
## ENSG00000001497 True False
## ENSG00000002330 False False
## ENSG00000002549 False False
## ENSG00000003249 False False
## ENSG00000003756 False False
## ENSG00000004766 True True
## PrimaryCilia_ PrimaryCilia_ASC52telo
## ENSG00000001497 True False
## ENSG00000002330 False False
## ENSG00000002549 False False
## ENSG00000003249 False False
## ENSG00000003756 False False
## ENSG00000004766 True True
## PrimaryCilia_hTERT_RPE1_serum_starved PrimaryCilia_RPTEC_TERT1
## ENSG00000001497 True False
## ENSG00000002330 False False
## ENSG00000002549 False False
## ENSG00000003249 False False
## ENSG00000003756 False False
## ENSG00000004766 True True
## PrimaryCiliaTip_ PrimaryCiliaTip_ASC52telo
## ENSG00000001497 False False
## ENSG00000002330 False False
## ENSG00000002549 False False
## ENSG00000003249 False False
## ENSG00000003756 False False
## ENSG00000004766 False False
## PrimaryCiliaTip_hTERT_RPE1_serum_starved
```

##	ENSG00000001497		False	
##	ENSG00000002330		False	
##	ENSG00000002549		False	
##	ENSG00000003249		False	
##	ENSG00000003756		False	
##	ENSG00000004766		False	
##		PrimaryCiliaTip_RPTEC_TERT1	PrimaryCiliaTZ_	
##	ENSG00000001497	False	True	
##	ENSG00000002330	False	False	
##	ENSG00000002549	False	False	
##	ENSG00000003249	False	False	
##	ENSG00000003756	False	False	
##	ENSG00000004766	False	False	
##		PrimaryCiliaTZ_ASC52telo		
##	ENSG00000001497	False		
##	ENSG00000002330	False		
##	ENSG00000002549	False		
##	ENSG00000003249	False		
##	ENSG00000003756	False		
##	ENSG00000004766	False		
##		PrimaryCiliaTZ_hTERT_RPE1_serum_starved		
##	ENSG00000001497	True		
##	ENSG00000002330	False		
##	ENSG00000002549	False		
##	ENSG00000003249	False		
##	ENSG00000003756	False		
##	ENSG00000004766	False		
##		PrimaryCiliaTZ_RPTEC_TERT1	Nucleus	Mitotic Membrane Cytoplasm
##	ENSG00000001497	False	True	False False True
##	ENSG00000002330	False	False	False False True
##	ENSG00000002549	False	False	False False True
##	ENSG00000003249	False	True	False False False
##	ENSG00000003756	False	True	False True True
##	ENSG00000004766	False	False	False True True
##		BasalBody_num	PrimaryCilia_num	PrimaryCiliaTip_num
##	ENSG00000001497	1	1	0
##	ENSG00000002330	0	0	0
##	ENSG00000002549	0	0	0
##	ENSG00000003249	0	0	0
##	ENSG00000003756	0	0	0
##	ENSG00000004766	2	3	0
##		PrimaryCiliaTZ_num	ASC52telo	hTERT_RPE1_serum_starved
##	ENSG00000001497	1	False	True
##	ENSG00000002330	0	False	False
##	ENSG00000002549	0	False	False
##	ENSG00000003249	0	False	False
##	ENSG00000003756	0	False	False
##	ENSG00000004766	0	True	True
##		RPTEC_TERT1		
##	ENSG00000001497	False		
##	ENSG00000002330	False		
##	ENSG00000002549	False		
##	ENSG00000003249	False		
##	ENSG00000003756	False		

```
## ENSG00000004766      True
```

Map gene names and symbols

```
# Identify columns that contain "num" in their names
num_columns <- grep("num", names(df_all), value = TRUE)

# Convert all other columns from string to logical
df_all <- df_all %>%
  mutate(across(-all_of(num_columns), ~ as.logical(.)))

# map gene IDs to gene names
mapped_ids <- map_geneID_to_name(rownames(df_all))

# index by Ensembl
rownames(mapped_ids) <- mapped_ids$ENSEMBL

# pull out exactly one entry per row of df_all
df_all$GeneSymbol <- mapped_ids[ rownames(df_all), "SYMBOL" ]
df_all$GeneName   <- mapped_ids[ rownames(df_all), "GENENAME" ]

# add rownames as column and reset index
df_all$Ensembl_ID <- rownames(df_all)
rownames(df_all) <- NULL

# reorder columns to have Ensembl_ID first, Symbol and Gene name first and then all other columns
df_all <- df_all %>% dplyr::select(Ensembl_ID, GeneSymbol, GeneName, everything())
```

Split data by location

```
# Perform filtering again
df_bb <- df_all %>% filter(BasalBody_ == TRUE)
df_pc <- df_all %>% filter(PrimaryCilia_ == TRUE)
df_tip <- df_all %>% filter(PrimaryCiliaTip_ == TRUE)
df_tz <- df_all %>% filter(PrimaryCiliaTZ_ == TRUE)

# filter gene_id by location
gene_id_all <- df_all$Ensembl_ID
gene_id_bb <- df_bb$Ensembl_ID
gene_id_pc <- df_pc$Ensembl_ID
gene_id_tip <- df_tip$Ensembl_ID
gene_id_tz <- df_tz$Ensembl_ID
```

Combine bb & tz and pc & tip

```
# combine pc and tip and tz
df_pc_tip_tz <- rbind(df_pc, df_tip, df_tz)
```

```
# drop duplicates based on Ensembl_ID column
df_pc_tip_tz <- df_pc_tip_tz[!duplicated(df_pc_tip_tz$Ensembl_ID), ]
gene_id_pc_tip_tz <- df_pc_tip_tz$Ensembl_ID
```

Compare biological themes for different locations with variable proteins

Define variable genes

```
# Load the data from the Excel file
df <- read_excel("/mnt/Data/Projects/Cilia/revision/Filtered_Staining_List_combined_exploded.xlsx")

# Identify Ensembl IDs for each category by searching substrings in the "Annotation (Intensity)" column
tip_variable <- unique(df$`Ensembl id`[grepl("Primary cilium tip", df$`Annotation (Intensity)`)])
pc_variable <- unique(df$`Ensembl id`[grepl("Primary cilium", df$`Annotation (Intensity)`)])
tz_variable <- unique(df$`Ensembl id`[grepl("Primary cilium transition zone", df$`Annotation (Intensity)`)])

# Combine all variable genes and remove duplicates
variable_genes <- unique(c(pc_variable, tip_variable, tz_variable))

# Identify any non-variable genes
non_variable <- setdiff(gene_id_pc_tip_tz, variable_genes)
```

Map as columns to df_all and save as csv file

```
# add "Intensity variation" columns to df_all with true if in variable genes, false if in non_variable
df_all <- df_all %>%
  mutate(
    Intensity_variation = case_when(
      Ensembl_ID %in% variable_genes ~ TRUE,
      Ensembl_ID %in% non_variable ~ FALSE,
      TRUE ~ NA # catch-all case for any other Ensembl IDs
    )
  )

# write to file
write.csv(df_all, file = paste0(out_path, "NonRestricted_all_proteins_enrichment_input_mapping.csv"))
```

GO BP enrichment analysis

```
# Split data by location
input_genes <- list(
  non_variable = non_variable,
  variable = variable_genes
)
```

```
# check length of each list
lapply(input_genes, length)
```

```
## $non_variable
## [1] 109
##
## $variable
## [1] 409
```

```
comp <- compareCluster(geneCluster = input_genes,
                        fun = "enrichGO",
                        OrgDb = org.Hs.eg.db,
                        keyType = 'ENSEMBL',
                        ont = "BP",
                        pAdjustMethod = "BH",
                        pvalueCutoff = 0.01,
                        qvalueCutoff = 0.01)
```

Dot plot of all enriched terms

```
# plot dotplot
dotplot(comp, showCategory = NULL) + ggtitle("GO BP Enrichment Comparison", subtitle = "qvalue < 0.01")
  scale_y_discrete(labels = function(x) str_wrap(x, width = 100))
```




```
# save dotplot as svg file
ggsave(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot.svg"), plot = last_plot(), dev = "svg")
```

Visualize overlap of enriched terms

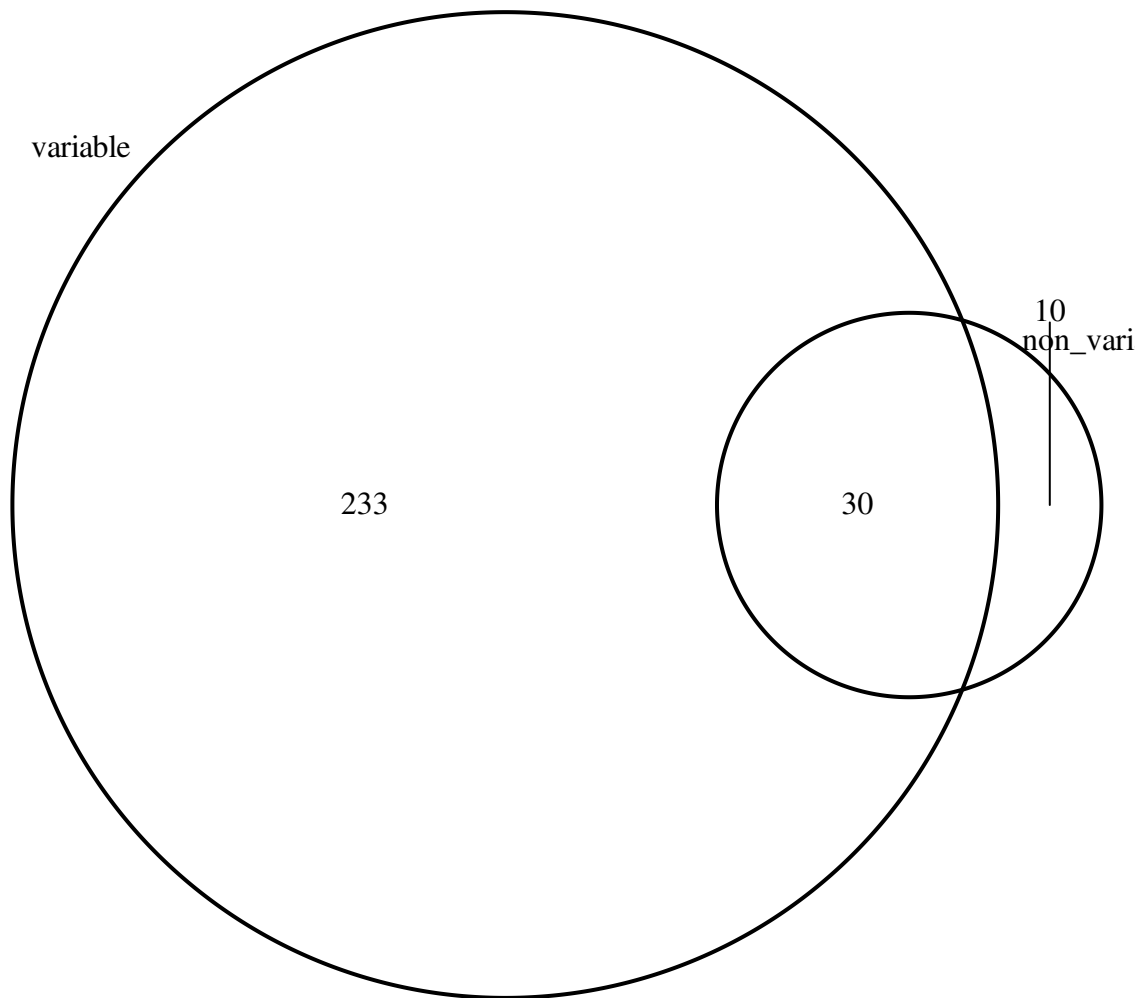
```
# extract results
results <- comp@compareClusterResult

# split by location
variable <- results[results$Cluster == "variable",]
non_variable <- results[results$Cluster == "non_variable",]

# Create a list of sets
go_lists <- list(
  variable = variable$ID,
  non_variable = non_variable$ID
)

# Plot the Venn diagram
venn.plot <- venn.diagram(
  x = go_lists,
  category.names = c("variable", "non_variable"),
  filename = NULL,
  output = TRUE
)

grid.newpage()
grid.draw(venn.plot)
```



```
# Save the captured plot as an SVG file
svglite(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_venn.svg"), width = 6, height = 6)
grid.draw(venn.plot)
dev.off()
```

```
## cairo_pdf
##          2
```

Filter terms only enriched in one condition

```
# calculate intersection of the two
unspecific_terms <- intersect(variable$ID, non_variable$ID)

# remove all unspecific terms
specific_terms <- results %>% filter(!ID %in% unspecific_terms)
```

```
# create a copy of comp
comp_filtered <- comp

# update results in comp
comp_filtered@compareClusterResult <- specific_terms
```

Dot plot of uniquely enriched terms

```
# plot dotplot
dotplot(comp_filtered, showCategory = NULL) + ggtitle("GO BP Enrichment Comparison", subtitle = "qvalue") +
  scale_y_discrete(labels = function(x) str_wrap(x, width = 100))
```



```

# save dotplot as svg file
ggsave(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_specific_terms_only.svg"), plo

# Subset for pc_tip
variable <- comp_filtered@compareClusterResult[
  comp_filtered@compareClusterResult$Cluster == "variable",
]

# Subset for bb_tz
non_variable <- comp_filtered@compareClusterResult[
  comp_filtered@compareClusterResult$Cluster == "non_variable",
]

# Create new compareClusterResult objects for each subset
comp_filtered_variable <- comp_filtered
comp_filtered_non_variable <- comp_filtered

comp_filtered_variable@compareClusterResult <- variable
comp_filtered_non_variable@compareClusterResult <- non_variable

```

Cluster results - variable proteins

```

go_id = comp_filtered_variable@compareClusterResult$ID
mat = GO_similarity(go_id,
  ont = 'BP',
  db = 'org.Hs.eg.db',
  measure = "Sim_Relevance_2006"
)

```

```

# Capture the plot
heatmap_plot <- grid.grabExpr({
  df <- simplifyGO(mat,
    method = 'binary_cut',
    plot = TRUE,
    column_title = "GO BP terms only significant in variable proteins",
    use_raster = FALSE,
    order_by_size = TRUE,
    fontsize_range = c(18,36),
    max_words = 6,
    word_cloud_grob_param = list(col = 'black',
                                  max_width = unit(200, "mm")))
})

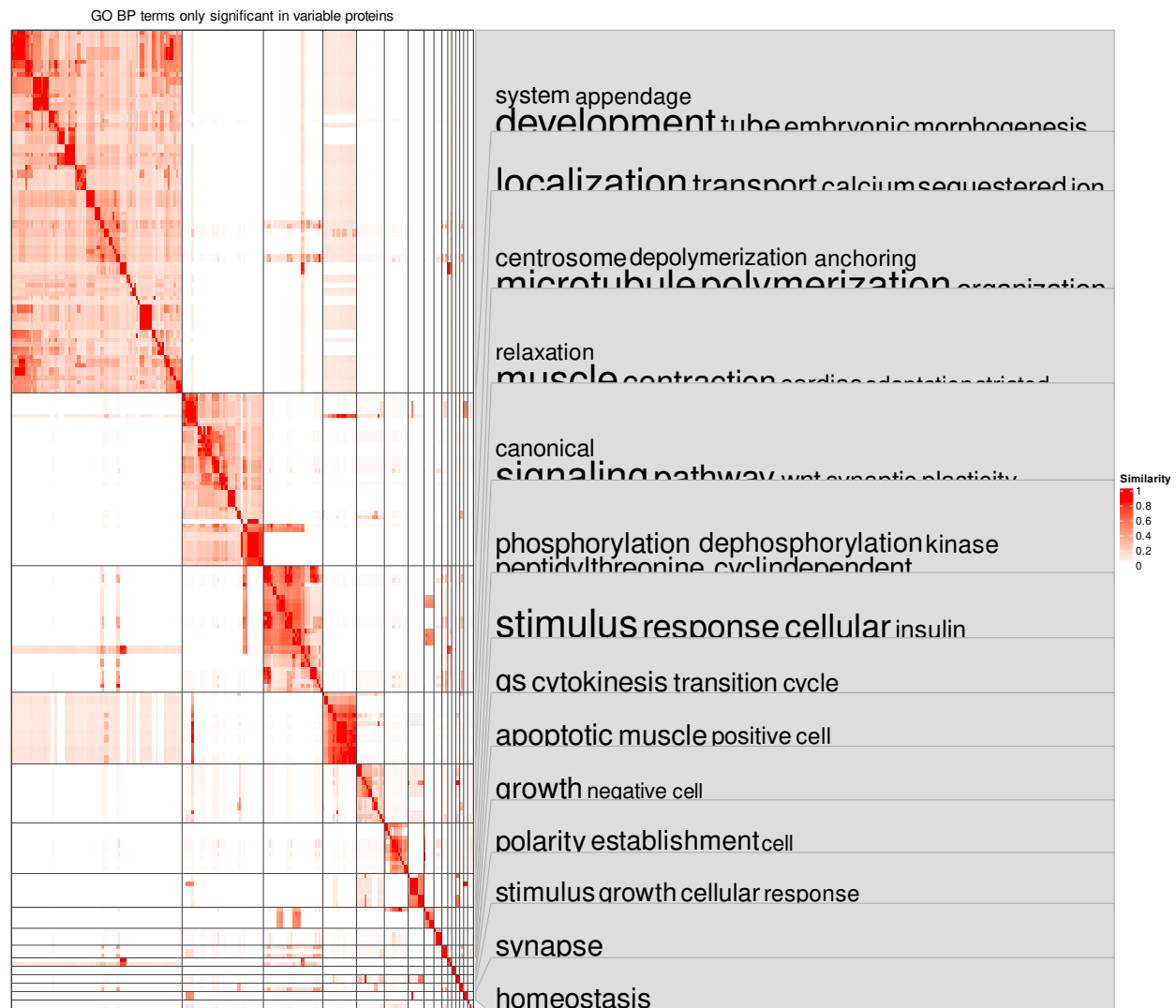
# Save the captured plot as an SVG file
svglite(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_specific_terms_only_variable_1
grid.draw(heatmap_plot)
dev.off()

```

Plot cluster heatmap

```
## cairo_pdf
##      2
```

```
grid.newpage()
grid.draw(heatmap_plot)
```



```
# add cluster number from GO term clustering
results <- map_cluster_number(comp_filtered_variable,
                             df = df,
                             comp = TRUE
)

# Apply the function to each row of the DataFrame
results$GeneSymbol <- sapply(results$geneID, map_geneID_to_symbol)
```

```
# save results as csv file
write.csv(results, file = paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_specific_terms_only_nonvariable_proteins.csv"))
```

Process and save results

Cluster results - non-variable proteins

```
go_id = comp_filtered_non_variable@compareClusterResult$ID
mat = GO_similarity(go_id,
  ont = 'BP',
  db = 'org.Hs.eg.db',
  measure = "Sim_Relevance_2006"
)
```

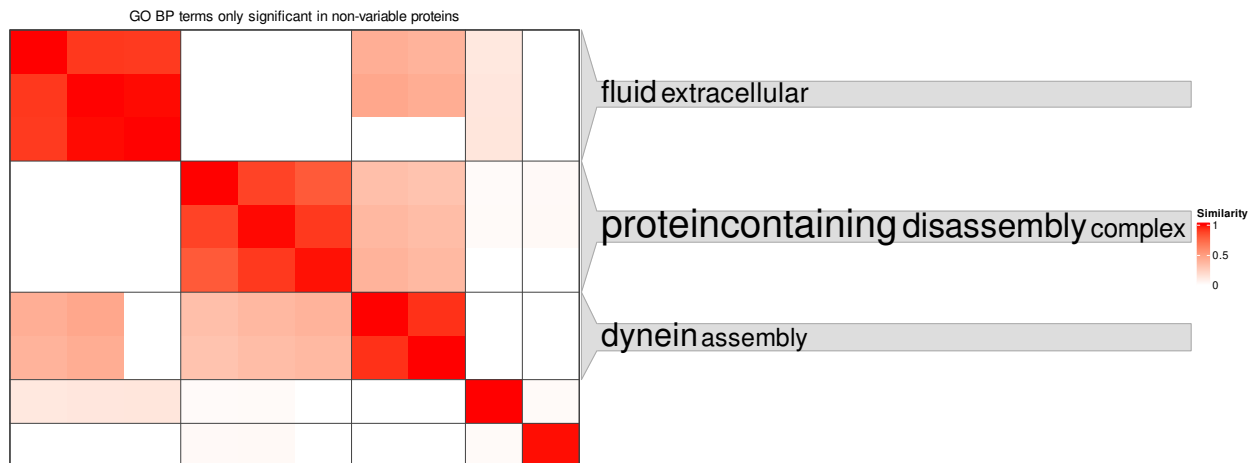
```
# Capture the plot
heatmap_plot <- grid.grabExpr({
  df <- simplifyGO(mat,
    method = 'binary_cut',
    plot = TRUE,
    column_title = "GO BP terms only significant in non-variable proteins",
    use_raster = FALSE,
    order_by_size = TRUE,
    fontsize_range = c(18, 36),
    max_words = 6,
    word_cloud_grob_param = list(col = 'black',
                                  max_width = unit(200, "mm")))
})

# Save the captured plot as an SVG file
svglite(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_specific_terms_only_nonvariable_proteins.svg"))
grid.draw(heatmap_plot)
dev.off()
```

Plot cluster heatmap

```
## cairo_pdf
##      2

grid.newpage()
grid.draw(heatmap_plot)
```



```
# add cluster number from GO term clustering
results <- map_cluster_number(comp_filtered_non_variable,
                             df = df,
                             comp = TRUE
)

# Apply the function to each row of the DataFrame
results$GeneSymbol <- sapply(results$geneID, map_geneID_to_symbol)

# save results as csv file
write.csv(results, file = paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_specific_te
```

Process and save results

Filter for terms enriched in both variable and non-variable

```
# get results as data frame
# extract results
results <- comp@compareClusterResult

# split by location
variable <- results[results$Cluster == "variable",]
non_variable <- results[results$Cluster == "non_variable",]

# calculate intersection of the two
unspecific_terms <- intersect(variable$ID, non_variable$ID)

# remove all unspecific terms
specific_terms <- results %>% filter(!ID %in% unspecific_terms)
unspecific_terms <- results %>% filter(ID %in% unspecific_terms)

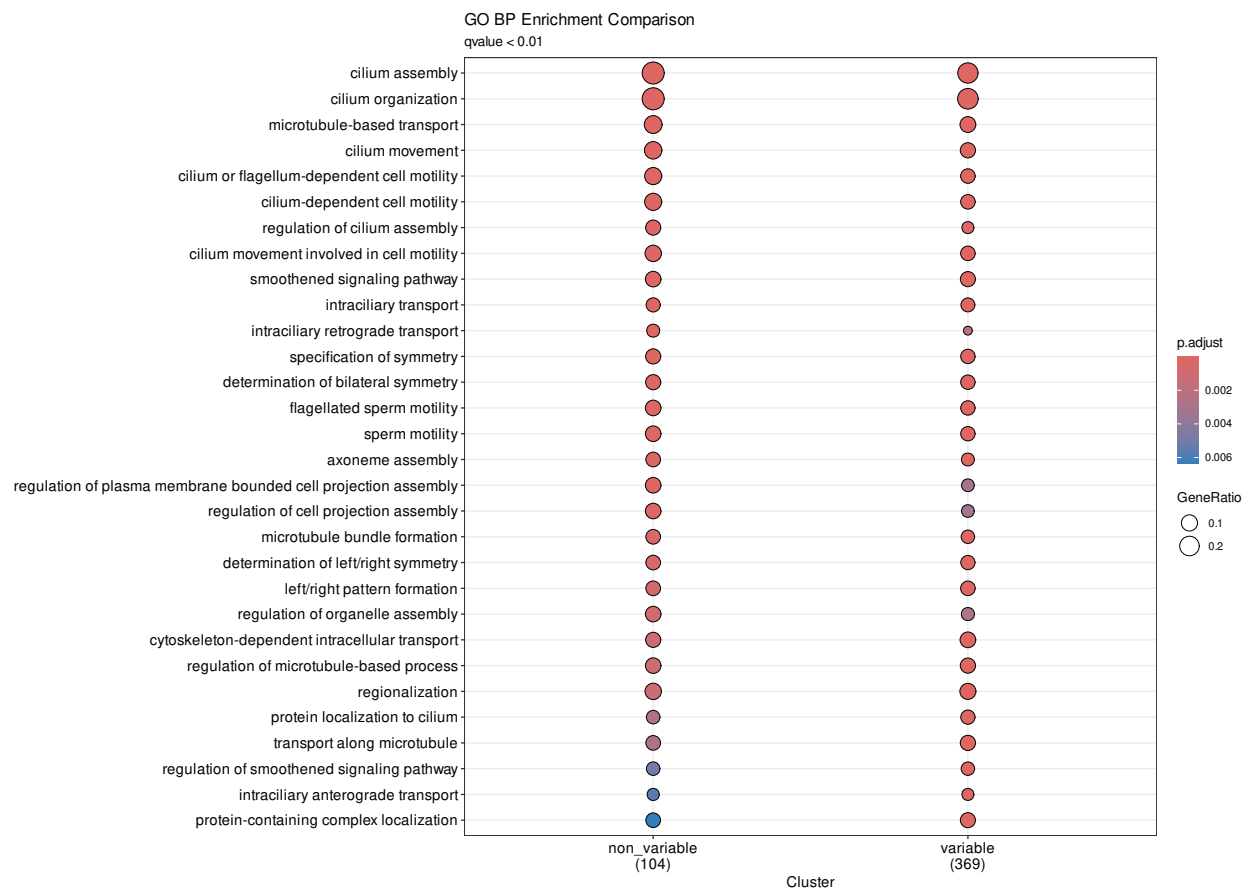
# create a copy of comp
comp_filtered <- comp
```



```
# update results in comp
comp_filtered@compareClusterResult <- unspecific_terms
```

Plot dotplot of shared terms

```
# plot dotplot
dotplot(comp_filtered, showCategory = NULL) + ggtitle("GO BP Enrichment Comparison", subtitle = "qvalue") +
  scale_y_discrete(labels = function(x) str_wrap(x, width = 100))
```



```
# save dotplot as svg file
ggsave(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_shared_terms_only.svg"), plot)
```

Cluster results - Both locations

```
# Split by location
variable <- comp_filtered@compareClusterResult[comp_filtered@compareClusterResult$Cluster == "variable"]
non_variable <- comp_filtered@compareClusterResult[comp_filtered@compareClusterResult$Cluster == "non_variable"]

# Create new compareClusterResult objects for each subset
comp_filtered_variable <- comp_filtered
```

```
comp_filtered_non_variable <- comp_filtered

comp_filtered_variable@compareClusterResult <- variable
comp_filtered_non_variable@compareClusterResult <- non_variable
```

```
go_id = comp_filtered_variable@compareClusterResult$ID
mat = GO_similarity(go_id,
  ont = 'BP',
  db = 'org.Hs.eg.db',
  measure = "Sim_Relevance_2006"
)
```

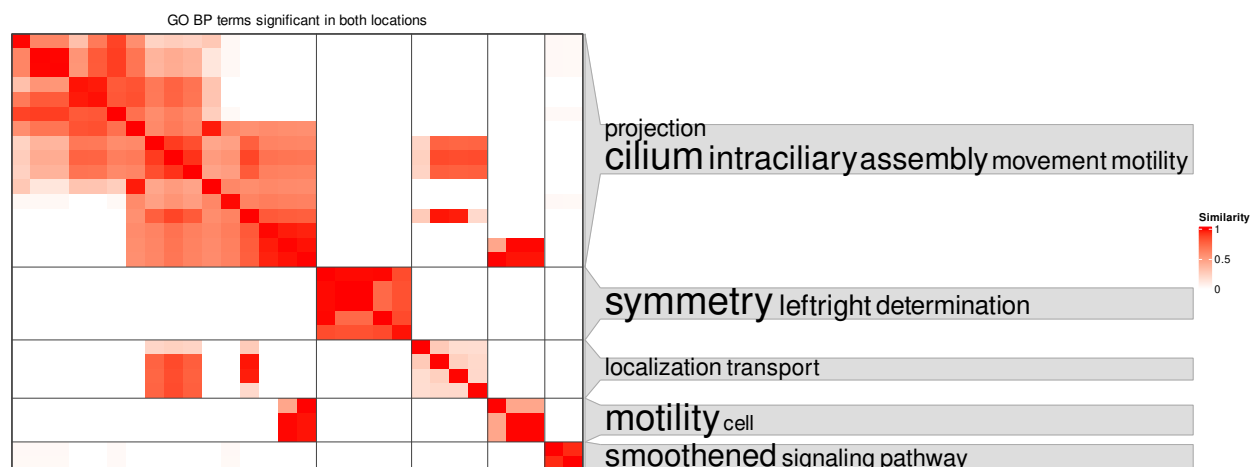
```
# Capture the plot
heatmap_plot <- grid.grabExpr({
  df <- simplifyGO(mat,
    method = 'binary_cut',
    plot = TRUE,
    column_title = "GO BP terms significant in both locations",
    use_raster = FALSE,
    order_by_size = TRUE,
    fontsize_range = c(18, 36),
    max_words = 6,
    word_cloud_grob_param = list(col = 'black',
      max_width = unit(200, "mm")))
})

# Save the captured plot as an SVG file
svglite(paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_shared_terms_heatmap.svg"), w
grid.draw(heatmap_plot)
dev.off()
```

Plot cluster heatmap

```
## cairo_pdf
##      2

grid.newpage()
grid.draw(heatmap_plot)
```



```
# add cluster number from GO term clustering
results <- map_cluster_number(comp_filtered_variable,
                             df = df,
                             comp = TRUE
)

# Apply the function to each row of the DataFrame
results$GeneSymbol <- sapply(results$geneID, map_geneID_to_symbol)

# save results as csv file
write.csv(results, file = paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_shared_terms.csv"))
```

```
# add cluster number from GO term clustering
results <- map_cluster_number(comp_filtered_non_variable,
                             df = df,
                             comp = TRUE
)

# Apply the function to each row of the DataFrame
results$GeneSymbol <- sapply(results$geneID, map_geneID_to_symbol)

# save results as csv file
write.csv(results, file = paste0(out_path, "NonRestricted_comparison_variable_GO_BP_dotplot_shared_terms.csv"))
```

Process and save results

Session info

```
sessionInfo()
```

```

## R version 4.5.0 (2025-04-11)
## Platform: x86_64-pc-linux-gnu
## Running under: Ubuntu 24.04.2 LTS
##
## Matrix products: default
## BLAS: /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.12.0
## LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.12.0 LAPACK version 3.12.0
##
## locale:
## [1] LC_CTYPE=en_US.UTF-8 LC_NUMERIC=C
## [3] LC_TIME=en_US.UTF-8 LC_COLLATE=en_US.UTF-8
## [5] LC_MONETARY=en_US.UTF-8 LC_MESSAGES=en_US.UTF-8
## [7] LC_PAPER=en_US.UTF-8 LC_NAME=C
## [9] LC_ADDRESS=C LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
##
## time zone: America/Los_Angeles
## tzcode source: system (glibc)
##
## attached base packages:
## [1] grid stats4 stats graphics grDevices utils datasets
## [8] methods base
##
## other attached packages:
## [1] readxl_1.4.5 simplifyEnrichment_2.2.0 VennDiagram_1.7.3
## [4] futile.logger_1.4.3 svglite_2.1.3 Cairo_1.6-2
## [7] org.Hs.eg.db_3.21.0 AnnotationDbi_1.70.0 IRanges_2.42.0
## [10] S4Vectors_0.46.0 Biobase_2.68.0 BiocGenerics_0.54.0
## [13] generics_0.1.3 enrichplot_1.28.2 clusterProfiler_4.16.0
## [16] lubridate_1.9.4 forcats_1.0.0 stringr_1.5.1
## [19] dplyr_1.1.4 purrr_1.0.4 readr_2.1.5
## [22] tidyr_1.3.1 tibble_3.2.1 ggplot2_3.5.2
## [25] tidyverse_2.0.0
##
## loaded via a namespace (and not attached):
## [1] RColorBrewer_1.1-3 rstudioapi_0.17.1 jsonlite_2.0.0
## [4] shape_1.4.6.1 magrittr_2.0.3 modeltools_0.2-24
## [7] ggtangle_0.0.6 farver_2.1.2 rmarkdown_2.29
## [10] ragg_1.4.0 GlobalOptions_0.1.2 fs_1.6.6
## [13] vctr_0.6.5 memoise_2.0.1 ggtree_3.16.0
## [16] htmltools_0.5.8.1 lambda.r_1.2.4 cellranger_1.1.0
## [19] gridGraphics_0.5-1 plyr_1.8.9 futile.options_1.0.1
## [22] cachem_1.1.0 igraph_2.1.4 mime_0.13
## [25] lifecycle_1.0.4 iterators_1.0.14 pkgconfig_2.0.3
## [28] Matrix_1.7-3 R6_2.6.1 fastmap_1.2.0
## [31] gson_0.1.0 GenomeInfoDbData_1.2.14 shiny_1.10.0
## [34] clue_0.3-66 digest_0.6.37 aplot_0.2.5
## [37] colorspace_2.1-1 patchwork_1.3.0 textshaping_1.0.1
## [40] RSQLite_2.3.11 labeling_0.4.3 timechange_0.3.0
## [43] httr_1.4.7 compiler_4.5.0 bit64_4.6.0-1
## [46] withr_3.0.2 doParallel_1.0.17 BiocParallel_1.42.0
## [49] DBI_1.2.3 R.utils_2.13.0 scatterplot3d_0.3-44
## [52] rjson_0.2.23 tools_4.5.0 ape_5.8-1
## [55] flexclust_1.5.0 httpuv_1.6.16 R.oo_1.27.1

```

## [58] glue_1.8.0	promises_1.3.2	nlme_3.1-168
## [61] GOSemSim_2.34.0	cluster_2.1.8.1	reshape2_1.4.4
## [64] fgsea_1.34.0	gtable_0.3.6	tzdb_0.5.0
## [67] class_7.3-23	R.methodsS3_1.8.2	data.table_1.17.0
## [70] hms_1.1.3	xml2_1.3.8	XVector_0.48.0
## [73] ggrepel_0.9.6	foreach_1.5.2	pillar_1.10.2
## [76] yulab.utils_0.2.0	later_1.4.2	circlize_0.4.16
## [79] splines_4.5.0	treeio_1.32.0	lattice_0.22-5
## [82] bit_4.6.0	tidyselect_1.2.1	GO.db_3.21.0
## [85] ComplexHeatmap_2.24.0	tm_0.7-16	Biostrings_2.76.0
## [88] knitr_1.50	NLP_0.3-2	xfun_0.52
## [91] matrixStats_1.5.0	stringi_1.8.7	UCSC.utils_1.4.0
## [94] lazyeval_0.2.2	ggfun_0.1.8	yaml_2.3.10
## [97] evaluate_1.0.3	codetools_0.2-20	qvalue_2.40.0
## [100] Polychrome_1.5.4	ggplotify_0.1.2	cli_3.6.5
## [103] xtable_1.8-4	systemfonts_1.2.3	Rcpp_1.0.14
## [106] GenomeInfoDb_1.44.0	png_0.1-8	parallel_4.5.0
## [109] simona_1.6.0	blob_1.2.4	DOSE_4.2.0
## [112] slam_0.1-55	tidytree_0.4.6	scales_1.4.0
## [115] crayon_1.5.3	GetoptLong_1.0.5	rlang_1.1.6
## [118] cowplot_1.1.3	fastmatch_1.1-6	KEGGREST_1.48.0
## [121] formatR_1.14		