Downstream Functional Analysis of Perseus-Processed Proteomics Data

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Data Import and LFQ Matrix Preparation

The protein quantification data generated with Perseus was first imported and processed to extract the LFQ (Label-Free Quantification) matrix required for downstream analyses. Column names were adjusted to comply with the input format required by the AMICA platform.

```
# Load the dataset exported from Perseus
datos <- read.table(file = "perseus.txt", header = TRUE, sep = "\t")</pre>
# Remove default column names to assign headers manually
colnames(datos) <- NULL</pre>
# Extract actual column names from the first row
nombresCol <- datos[1,]</pre>
# Remove the first row containing the original headers and retain only data
datos bien <- datos[-1,]
# Select columns corresponding to LFQ intensities and standardize their names
LFQ names <- nombresCol[45:60]
LFQ_names_bien <- paste("LFQ intensity", LFQ_names)</pre>
nombresCol[45:60] <- LFQ_names_bien
# Apply corrected column names to the dataset
colnames(datos_bien) <- nombresCol</pre>
# Convert appropriate columns to numeric values, replacing commas with periods
# if necessary
cols_a_modificar <- setdiff(7:ncol(datos_bien), c(10, 14))</pre>
datos_bien[, cols_a_modificar] <- lapply(datos_bien[, cols_a_modificar], function(x) {</pre>
  if (is.character(x) | is.factor(x)) {
    as.numeric(gsub(",", ".", as.character(x))) # Replace commas and convert to numeric
  } else {
  }
})
```

LFQ Intensity Normalization

LFQ intensities corresponding to day 1 samples were normalized relative to the average of the remaining days to correct for batch-specific shifts. Additionally, an anti-log2 transformation was applied to revert values to

linear scale, as required by the AMICA tool.

```
# Define columns corresponding to day 1 samples
dia1 cols <- c("LFQ intensity Control-1", "LFQ intensity Cwp2-1",
               "LFQ intensity Emp24-1", "LFQ intensity Gas1-1")
# Identify remaining LFQ columns (other time points)
otros_dias <- setdiff(colnames(datos_bien)[45:60], dia1_cols)
# Compute row-wise mean intensities for day 1 and for other days
media_dia1 <- rowMeans(datos_bien[, dia1_cols])</pre>
media_rest <- rowMeans(datos_bien[, otros_dias])</pre>
# Calculate shift required to center day 1 values with respect to the rest
desplazamiento <- media_dia1 - media_rest</pre>
# Apply normalization by subtracting the shift from day 1 columns
for (col in dia1_cols) {
  datos_bien[[col]] <- datos_bien[[col]] - desplazamiento</pre>
# Perform anti-log2 transformation to obtain linear intensity values (required by AMICA)
datos_bien[, 45:60] <- lapply(datos_bien[, 45:60], function(x) { 2^x })</pre>
# Duplicate LFQ intensity columns with new names ("Intensity") for AMICA compatibility
lfq_cols <- grep("^LFQ intensity ", colnames(datos_bien), value = TRUE)</pre>
for (col in lfq_cols) {
 new_col <- sub("^LFQ intensity", "Intensity", col)</pre>
 datos_bien[[new_col]] <- datos_bien[[col]]</pre>
}
# Export the normalized data matrix to a TSV file
write.table(datos_bien, file = "perseus.tsv", sep = "\t", quote = FALSE,
           row.names = FALSE)
```

Identification of Enriched and Exclusive Interactors

```
# Apply the function to each list of enriched proteins
cwp2_enriched <- separar_proteinas(cwp2_enriched)</pre>
gas1_enriched <- separar_proteinas(gas1_enriched)</pre>
emp24_enriched <- separar_proteinas(emp24_enriched)</pre>
# Identify exclusive interactors for each protein by removing those shared with
# the other two groups
cwp2_exclusivas <- setdiff(cwp2_enriched, union(gas1_enriched, emp24_enriched))</pre>
gas1_exclusivas <- setdiff(gas1_enriched, union(cwp2_enriched, emp24_enriched))</pre>
emp24_exclusivas <- setdiff(emp24_enriched, union(cwp2_enriched, gas1_enriched))</pre>
# Export enriched and exclusive interactors to text files
write(cwp2_enriched, file = "cwp2_TOTALES.txt")
write(gas1_enriched, file = "gas1_TOTALES.txt")
write(emp24_enriched, file = "emp24_TOTALES.txt")
write(cwp2_exclusivas, file = "cwp2_exclusivas.txt")
write(gas1_exclusivas, file = "gas1_exclusivas.txt")
write(emp24_exclusivas, file = "emp24_exclusivas.txt")
# Display the number of unique enriched interactors for each bait
length(unique(cwp2_enriched))
## [1] 555
length(unique(gas1_enriched))
## [1] 373
length(unique(emp24_enriched))
## [1] 510
```

Identification of shared and partially shared interactors among Gas1, Cwp2, and Emp24

```
# 1. Interactors shared between Gas1 and Emp24, but not present in Cwp2
# Overlap between Gas1 and Emp24
gas1_emp24 <- intersect(gas1_enriched, emp24_enriched)
# Remove those also found in Cwp2
gas1_emp24_only <- setdiff(gas1_emp24, cwp2_enriched)</pre>
```

```
# Combine Gas1-exclusive with shared-only-with-Emp24
gas1_total_31 <- union(gas1_exclusivas, gas1_emp24_only)</pre>
# Export shared-only-with-Emp24 interactors
write(gas1_emp24_only, file = "gas1andemp24.txt")
# 2. Interactors shared between Cwp2 and Emp24, but not present in Gas1
# Overlap between Cwp2 and Emp24
cwp2_emp24 <- intersect(cwp2_enriched, emp24_enriched)</pre>
# Remove those also found in Gas1
cwp2_emp24_only <- setdiff(cwp2_emp24, gas1_enriched)</pre>
# Combine Cwp2-exclusive with shared-only-with-Emp24
cwp2_total_213 <- union(cwp2_exclusivas, cwp2_emp24_only)</pre>
# Export shared-only-with-Emp24 interactors
write(cwp2_emp24_only, file = "cwp2andemp24.txt")
# 3. Interactors shared between Gas1 and Cwp2, but not present in Emp24
# Overlap between Gas1 and Cwp2
gas1_cwp2 <- intersect(gas1_enriched, cwp2_enriched)</pre>
# Remove those also found in Emp24
gas1_cwp2_only <- setdiff(gas1_cwp2, emp24_enriched)</pre>
# Export shared-only-with-Cwp2 interactors
write(gas1_cwp2_only, file = "gas1andcwp2.txt")
# 4. Interactors shared across all three bait proteins (Gas1, Cwp2, Emp24)
# Find common interactors across all three
shared_all_three <- Reduce(intersect, list(gas1_enriched, cwp2_enriched, emp24_enriched))</pre>
# Export interactors common to all three
write(shared_all_three, file = "gas1_cwp2_emp24_comunes.txt")
```

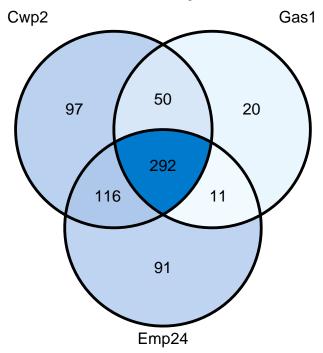
Venn Diagram of Enriched Interactors

```
# Load required packages
library(ggVennDiagram)
library(ggplot2)

# Define a named list of enriched interactors for each bait protein
listas <- list(
   Cwp2 = cwp2_enriched,</pre>
```

```
Gas1 = gas1_enriched,
  Emp24 = emp24_enriched
# Generate Venn diagram showing the overlap of enriched proteins among the three baits
ggVennDiagram(listas, label_alpha = 0, label = "count", set_label_size = 4) +
  # Apply a color gradient for the overlapping areas
  scale_fill_gradient(low = "#f0faff", high = "#007acc") +
  # Remove background and axes for a cleaner appearance
  theme_void() +
  # Customize theme elements: title, margins, background
  theme(
    legend.position = "none",
    plot.title = element_text(size = 14, face = "bold", hjust = 0.5),
    plot.margin = margin(20, 20, 20, 20),
    plot.background = element_rect(fill = "white", color = NA),
   panel.background = element_rect(fill = "white", color = NA)
  ) +
  # Add a title to the plot
  labs(title = "Enriched Interactors of Cwp2, Gas1, and Emp24")
```

Enriched Interactors of Cwp2, Gas1, and Emp24

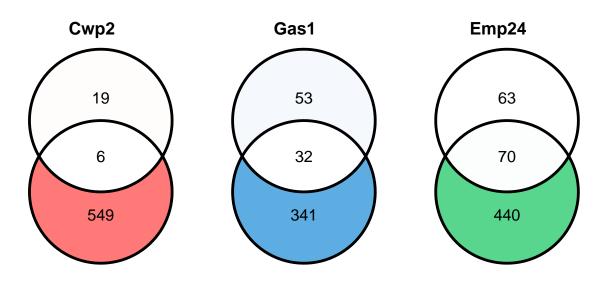


Comparison with SGD Physical Interactome Data

```
# Load physical interaction datasets from Saccharomyces Genome Database (SGD)
CWP2_SGD <- read.delim("CWP2_physical_interactions.txt",</pre>
                       header = TRUE, sep = "\t", quote = "")
CWP2_SGD_PROTS <- unique(CWP2_SGD$Interactor.1)</pre>
GAS1_SGD <- read.delim("GAS1_physical_interactions.txt",</pre>
                       header = TRUE, sep = "\t", quote = "")
GAS1_SGD_PROTS <- unique(GAS1_SGD$Interactor.1)</pre>
EMP24_SGD <- read.delim("EMP24_physical_interactions.txt",</pre>
                        header = TRUE, sep = "\t", quote = "")
EMP24_SGD_PROTS <- unique(EMP24_SGD$Interactor.1)</pre>
# Display number of unique interactors from SGD for each bait
length(CWP2 SGD PROTS)
## [1] 25
length(GAS1_SGD_PROTS)
## [1] 85
length(EMP24_SGD_PROTS)
## [1] 133
# Load necessary libraries for plotting
library(ggplot2)
library(ggVennDiagram)
library(patchwork)
# Create Venn diagram: Cwp2 (experimental vs. SGD interactors)
p1 <- ggVennDiagram(
 list(" " = cwp2_enriched, " " = CWP2_SGD_PROTS),
 label_alpha = 0, label = "count", set_label_size = 4,
  set_name = NULL
) +
  scale_fill_gradient(low = "#ffffff", high = "#ff7979") +
  theme void() +
 labs(subtitle = "Cwp2") +
    plot.subtitle = element_text(size = 13, face = "bold", hjust = 0.5),
    plot.margin = margin(10, 10, 10, 10),
    legend.position = "none"
# Venn diagram: Gas1 (experimental vs. SGD)
p2 <- ggVennDiagram(</pre>
```

```
list(" " = gas1_enriched, " " = GAS1_SGD_PROTS),
  label_alpha = 0, label = "count", set_label_size = 4,
  set_name = NULL
) +
  scale_fill_gradient(low = "#ffffff", high = "#5dade2") +
  theme void() +
  labs(subtitle = "Gas1") +
  theme(
    plot.subtitle = element_text(size = 13, face = "bold", hjust = 0.5),
    plot.margin = margin(10, 10, 10, 10),
    legend.position = "none"
# Venn diagram: Emp24 (experimental vs. SGD)
p3 <- ggVennDiagram(
 list(" " = emp24_enriched, " " = EMP24_SGD_PROTS),
  label_alpha = 0, label = "count", set_label_size = 4,
  set_name = NULL
) +
  scale_fill_gradient(low = "#ffffff", high = "#58d68d") +
  theme_void() +
  labs(subtitle = "Emp24") +
  theme(
    plot.subtitle = element_text(size = 13, face = "bold", hjust = 0.5),
    plot.margin = margin(10, 10, 10, 10),
    legend.position = "none"
# Combine all three Venn diagrams into a single figure
final_plot \leftarrow (p1 + p2 + p3) +
  plot_annotation(
   title = "Overlap between Experimental Enriched Interactors and SGD
    Physical Interactome",
    theme = theme(
      plot.title = element_text(size = 13, face = "bold", hjust = 0.5,
                                margin = margin(b = 18))
    )
  )
# Display and save the final plot
final_plot
```

Overlap between Experimental Enriched Interactors and SGD Physical Interactome



```
ggsave("venn_comparison.png", plot = final_plot, width = 25, height = 5, dpi = 300)
```

Create Cytoscape-Compatible Table for ER-to-Golgi Transport Proteins Enriched in Cwp2

```
# Vector of Cytoscape-compatible IDs (shared name), in the same order as the gene list
shared_names <- c(
   "4932.YLR078C", "4932.YLR080W", "4932.YAR002C-A", "4932.YHR110W",
   "4932.YGL054C", "4932.YPL053C", "4932.YDR189W", "4932.YOR307C",
   "4932.YJL192C", "4932.YGL145W", "4932.YKR044W", "4932.YNL044W"
)

# Corresponding gene names
gene_names <- c(
   "BOS1", "EMP46", "ERP1", "ERP5", "ERV14",
   "KTR6", "SLY1", "SLY41", "SOP4", "TIP20",
   "UIP5", "YIP3"
)

# Match rows from the input data frame (datos_bien) based on gene names
filtered_data <- datos_bien[match(gene_names, datos_bien$"Gene names`), ]

# Create the output data frame with log2 fold changes and identifiers</pre>
```