Final Project

PBX

2025-08-19

Step0:Environment

Step1:Read and Adjust data

```
meta_raw <- read_csv("QBS103_GSE157103_series_matrix-1.csv", show_col_types = FALSE)</pre>
genes_raw <- readr::read_csv("QBS103_GSE157103_genes (1).csv", show_col_types = FALSE)
#cleaning names in dataset
meta <- meta_raw %>% clean_names()
glimpse(meta, width = 80)
## Rows: 126
## Columns: 25
                                             <chr> "COVID_01_39y_male_NonICU", "C~
## $ participant_id
                                             <chr> "GSM4753021", "GSM4753022", "G~
## $ geo_accession
## $ status
                                             <chr> "Public on Aug 29 2020", "Publ~
                                             <chr> "Aug 28 2020", "Aug 28 2020", ~
## $ sample submission date
                                             <chr> "Aug 29 2020", "Aug 29 2020", ~
## $ last_update_date
                                             <chr> "SRA", "SRA", "SRA", "SRA", "S~
## $ type
## $ channel count
                                             <dbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ source name ch1
                                             <chr> "Leukocytes from whole blood",~
## $ organism ch1
                                             <chr> "Homo sapiens", "Homo sapiens"~
## $ disease_status
                                             <chr> "disease state: COVID-19", "di~
                                             <chr> "39", "63", "33", "49", "49", ~
## $ age
                                             <chr> "male", "male", "male", "male"~
## $ sex
                                             <chr> "no", "no", "no", "no", "no", ~
## $ icu_status
                                             <chr> "15", "unknown", "unknown", "u~
## $ apacheii
                                             <dbl> 0, 2, 2, 1, 1, 1, 7, 7, 2, 1, ~
## $ charlson_score
                                             <chr> "yes", "no", "no", "no", "yes"~
## $ mechanical_ventilation
## $ ventilator_free_days
                                              <dbl> 0, 28, 28, 28, 23, 28, 28, 0, ~
## $ hospital_free_days_post_45_day_followup <dbl> 0, 39, 18, 39, 27, 36, 42, 0, ~
## $ ferritin_ng_ml
                                              <chr> "946", "1060", "1335", "583", ~
                                              <chr> "73.1", "unknown", "53.2", "25~
## $ crp mg 1
                                             <chr> "1.3", "1.03", "1.48", "1.32",~
## $ ddimer_mg_l_feu
                                             <chr> "36", "0.37", "0.07", "0.98", ~
## $ procalcitonin ng ml
                                             <chr> "0.9", "unknown", "unknown", "~
## $ lactate_mmol_l
                                             <chr> "513", "unknown", "513", "949"~
## $ fibrinogen
## $ sofa
                                             <chr> "8", "unknown", "unknown", "un~
meta <- meta %>%
```

mutate(across(where(is.character), ~str_trim(.x))) %>%

```
mutate(across(where(is.character),
                ~ifelse(str_to_lower(.x) %in% c("unknown", "na", ""),
                        NA_character_, .x)))
#creating 3 continuous variables + categorical variables
clean num <- function(x) {</pre>
 x %>%
   str trim() %>%
   na_if("") %>% na_if("NA") %>% na_if("na") %>% na_if("unknown") %>% na_if(":") %>%
   readr::parse_number(locale = readr::locale(decimal_mark = ".", grouping_mark = ","))
}
meta <- meta %>%
 mutate(
           = clean_num(age),
   ferritin = clean_num(ferritin_ng_ml),
           = clean_num(crp_mg_1)
    crp
meta <- meta %>%
 mutate(
   sex = case_when(
     str to lower(sex) == "female" ~ "Female",
     str_to_lower(sex) == "male" ~ "Male",
                                    ~ "Unknown"
     TRUE
   ),
   icu status = case when(
      str_to_lower(icu_status) %in% c("yes","icu")
                                                                     ~ "ICU",
     str_to_lower(icu_status) %in% c("no", "nonicu", "non-icu")
                                                                   ~ "Non-ICU".
     TRUE
                                                                     ~ "Unknown"
   ),
   disease_status = case_when(
     !is.na(disease_status) & str_detect(str_to_lower(disease_status), "covid") ~ "COVID-19",
                                                                                 ~ "Non-COVID"
   )
  ) %>%
  mutate(
                  = factor(sex,
                                          levels = c("Female", "Male", "Unknown")),
                 = factor(icu_status, levels = c("ICU", "Non-ICU", "Unknown")),
   icu_status
   disease_status = factor(disease_status, levels = c("COVID-19", "Non-COVID"))
#transpose matrix and combining dataset
genes t <- genes raw %>%
 rename(gene = 1) %>%
 column_to_rownames("gene") %>%
 t() %>% as.data.frame() %>%
 rownames_to_column("participant_id") %>%
 as_tibble()
full_data <- meta %>% left_join(genes_t, by = "participant_id")
#checking
```

```
dim(full_data)
```

```
## [1] 126 127
```

Step3: Generating Latex summary table (stratified by icu_status)

```
# Step3: Generate LaTeX summary table (stratified by icu_status)
#selecting variables
table_dat <- full_data %>%
  dplyr::select(icu_status, sex, disease_status, age, crp, ferritin)
tab1 <- gtsummary::tbl_summary(</pre>
 data = table_dat,
  by = icu_status,
  statistic = list(
   gtsummary::all_continuous() ~ "{mean} ({sd})",
   gtsummary::all_categorical() ~ "{n} ({p}%)"
  ),
  digits = gtsummary::all_continuous() ~ 1,
 missing = "ifany",
 label = list(
                  ~ "Age (years)",
   age
   crp
                 ~ "CRP (mg/L)",
                 ~ "Ferritin (ng/mL)",
   ferritin
   sex
                 ~ "Sex",
   disease_status ~ "Disease status"
) %>%
  gtsummary::add_overall(last = TRUE) %>%
  gtsummary::bold_labels() %>%
  gtsummary::modify_caption("**Summary statistics stratified by ICU status**")
#export Latex
library(kableExtra)
dir.create("tables", showWarnings = FALSE)
invisible(tab1)
latex_tab1 <- gtsummary::as_kable_extra(</pre>
 x = tab1, format = "latex", booktabs = TRUE, escape = FALSE
 kable_styling(latex_options = "HOLD_position")
invisible(save_kable(latex_tab1, "tables/summary_table.tex"))
```

Step4: Making previous graph

```
#select main gene
gene_main <- "AAK1"

#histogram</pre>
```

```
p_hist <- ggplot(full_data, aes(x = .data[[gene_main]])) +</pre>
  geom histogram(binwidth = 0.5, color = "white") +
  labs(
   title = glue("Histogram of {gene_main} Expression"),
   x = glue("{gene_main} Expression (a.u.)"),
   y = "Count"
  ) +
  theme(plot.title = element text(hjust = 0.04))
ggsave("figs/fig1_histogram_AAK1.png", p_hist, width = 6, height = 4, dpi = 300)
#scatter plot
p scatter <- ggplot(full data, aes(x = ferritin, y = .data[[gene main]], color = icu status)) +
  geom_point() +
  labs(
   title = glue("{gene_main} vs Ferritin"),
   x = "Ferritin (ng/mL)",
   y = glue("{gene_main} Expression"),
   color = "ICU"
  ) +
  scale_x_continuous(labels = scales::label_comma()) +
  scale_color_brewer(palette = "Set1") +
  theme(plot.title = element_text(hjust = 0.04))
ggsave("figs/fig2_scatter_AAK1_ferritin.png", p_scatter, width = 6, height = 4, dpi = 300)
#boxplot
p_box <- ggplot(full_data, aes(x = sex, y = .data[[gene_main]], fill = icu_status)) +</pre>
  geom_boxplot() +
  labs(
   title = glue("{gene_main} Expression by Sex and ICU Status"),
   x = "Sex"
   y = glue("{gene_main} Expression"),
   fill = "ICU"
  ) +
  scale_fill_brewer(palette = "Set2") +
  theme(plot.title = element_text(hjust = 0.04))
ggsave("fig3/fig3_box_AAK1_sex_icu.png", p_box, width = 6, height = 4, dpi = 300)
```

Step5:Heatmap

Step6: hexbin plot

Introduction

This report analyzes gene expression profiles from the dataset with clinical covariates. I focus on AAK1 for the main figures and summarize key continuous (age, CRP, ferritin) and categorical variables (sex, ICU status, disease status). Data were cleaned (string trimming, standardization of categories), merged with the gene matrix, and visualized.

Methods

I create a LaTeX-formatted descriptive table of three continuous (age, CRP, ferritin) and three categorical variables (sex, ICU status, disease status), stratified by ICU status. For the main plots, I produced: (i) a histogram of AAK1 expression, (ii) a scatterplot of AAK1 vs ferritin colored by ICU status, and (iii) a boxplot of AAK1 by sex with ICU status as fill. For the heatmap, I select the top 10 genes by variance (ensuring AAK1 is included), applied row-wise z-score normalization, clustered rows and columns, and added two tracking bars (sex, ICU status). Moreover, I included a new plot type (hexbin) to depict dense 2D distributions.

Results

Summary Table

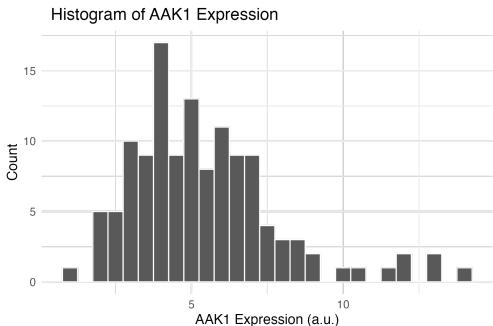
Table 1: Summary statistics stratified by ICU status

Characteristic	ICU N = 66	Non-ICU $ N = 60$	Unknown N = 0	Overall N = 126
Sex				
Female	24 (36%)	27 (45%)	0 (NA%)	51 (40%)
Male	41 (62%)	33 (55%)	0 (NA%)	74 (59%)
Unknown	1(1.5%)	0 (0%)	0 (NA%)	1 (0.8%)
Disease status				
COVID-19	66 (100%)	60 (100%)	0 (NA%)	126 (100%)
Non-COVID	0 (0%)	0 (0%)	0 (NA%)	0 (0%)
Age (years)	63.5(14.0)	59.7 (18.4)	NA (NA)	$61.7\ (16.2)$
Unknown	0	1	0	1
$\mathrm{CRP}\ (\mathrm{mg/L})$	$149.6 \ (105.5)$	109.4 (94.4)	NA (NA)	$131.2\ (102.1)$
Unknown	8	11	0	19
Ferritin (ng/mL)	935.3 (1,019.0)	715.7 (1,067.6)	NA (NA)	833.5 (1,042.8)
Unknown	7	9	0	16

¹ n (%); Mean (SD)

Figures

Figure 1. Histogram of AAK1 expression.



AAK1 expression is right-skewed, with most samples concentrated between 3–7 units, suggesting variability across patients.

AAK1 vs Ferritin

ICU
ICU
Non-ICU

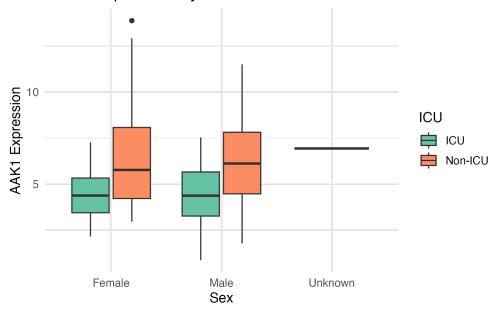
Ferritin (ng/mL)

Figure 2. Scatterplot of AAK1 expression vs ferritin, colored by ICU status.

The scatterplot shows no clear linear relationship between ferritin and AAK1 expression, although most patients cluster at low ferritin values.

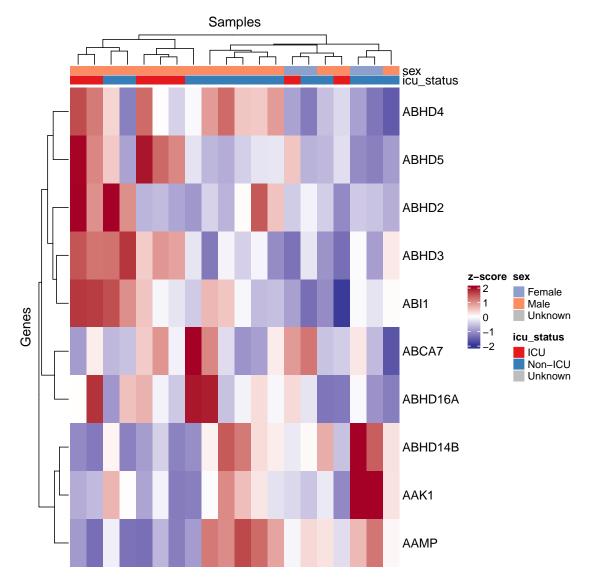
Figure 3. Boxplot of AAK1 expression by sex and ICU status.

AAK1 Expression by Sex and ICU Status



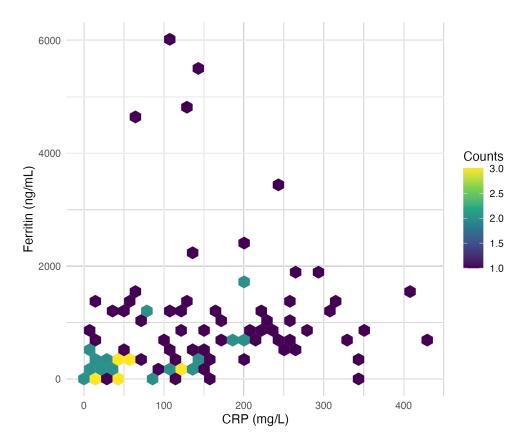
The boxplot suggests that ICU patients may have slightly higher AAK1 expression than non-ICU patients, with no major differences across sex groups.

Figure 4. Heatmap of the top 10 most variable genes.



The heatmap of the first 20 patients shows distinct clustering patterns, indicating that both ICU status and sex are associated with gene expression variation.

Figure 5. Hexbin plot of CRP vs Ferritin.



The hexbin plot shows most patients clustered in the low CRP and low ferritin range, while a few outliers display very high ferritin levels.

References

- Wickham H et al. (2019). tidyverse: Easily Install and Load the 'Tidyverse'. R package version 1.3.1. Available at: https://CRAN.R-project.org/package=tidyverse
- Firke S. (2023). janitor: Simple Tools for Examining and Cleaning Dirty Data. R package version 2.2.0.

Available at: https://CRAN.R-project.org/package=janitor

- Sjoberg DD, Whiting K, Curry M, Lavery JA, Larmarange J. (2021). gtsummary: Presentation-Ready Data Summaries and Analytic Result Tables.
 Available at: https://CRAN.R-project.org/package=gtsummary
- Zhu H. (2021). kableExtra: Construct Complex Table with 'kable' and Pipe Syntax. Available at: https://CRAN.R-project.org/package=kableExtra
- Wickham H. (2022). scales: Scale Functions for Visualization. Available at: https://CRAN.R-project.org/package=scales
- Gu Z, Eils R, Schlesner M. (2016). ComplexHeatmap: Making Complex Heatmaps Simple. *Bioinformatics* 32(18):2847-2849.
 - Available at: https://bioconductor.org/packages/ComplexHeatmap
- Gu Z, Gu L. (2014). circlize: Circular Visualization in R. *Bioinformatics* 30(19):2811-2812. Available at: https://CRAN.R-project.org/package=circlize

- Hester J, Wickham H, Chang W, Bryan J. (2023). glue: Interpreted String Literals. Available at: https://CRAN.R-project.org/package=glue
- Pedersen TL. (2020). patchwork: The Composer of Plots. Available at: https://CRAN.R-project.org/package=patchwork
- Carr DB, Lewin-Koh NJ, Maechler M. (2022). hexbin: Hexagonal Binning Routines. Available at: https://CRAN.R-project.org/package=hexbin