#### R final submission

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2025-08-18

#### Final Project

Before class on 8/21, you should submit (1) your compiled PDF from LaTeX, (2) your corresponding .tex file, (3) a knitted PDF of the R markdown file with your code for the final submission, and (4) a link to your now public facing GitHub repository.

```
#Final subission
#Load gene expression data
gene_data <- read.csv(file = "QBS103_GSE157103_genes.csv",row.names=1)
#Load metadata for participants
series <- read.csv(file = "QBS103_GSE157103_series_matrix-1.csv")</pre>
```

Generate a table formatted in LaTeX of summary statistics for all the covariates you looked at and 2 additional continuous (3 total) and 1 additional categorical variable (3 total). (5 pts) Stratifying by one of your categorical variables Tables should report n (%) for categorical variables Tables should report mean (sd) or median [IQR] for continuous variables

```
only_num <- function(x) {</pre>
  is_num \leftarrow grepl("^[[:space:]]*[0-9]+(?:\.[0-9]+)?[[:space:]]*$", x)
  out <- rep(NA_real_, length(x))</pre>
  out[is_num] <- as.numeric(trimws(x[is_num]))</pre>
  out
}
df <- series %>%
  mutate(
    # Remove any extra spaces from 'sex' variable
    sex = trimws(sex),
    # Recode ICU status to standardized labels
    icu_status = ifelse(tolower(trimws(icu_status)) == "yes", "ICU", "NonICU"),
    # Convert 'age' to numeric
    age = only_num(age),
    # Convert ferritin to numeric (ignore warnings if conversion fails)
    ferritin = only_num(ferritin.ng.ml.),
    # Convert CRP to numeric (ignore warnings)
    crp = only_num(crp.mg.l.),
    # Recode mechanical ventilation variable
    mech_vent = ifelse(trimws(mechanical_ventilation) == "yes", "Yes", "No")
  )
```

```
# Summarize categorical variables (sex and mechanical ventilation) by ICU status
df %>%
 group_by(icu_status) %>%
 summarise(
   n_sex_female = sum(sex == "female", na.rm=TRUE),
                                                     # Count number of females
   n_sex_male = sum(sex == "male", na.rm=TRUE), # Count number of males
   n_mechvent_yes = sum(mech_vent == "Yes", na.rm=TRUE), # Count patients with mechanical ventilation
   n mechanical vent == "No", na.rm=TRUE), # Count patients without mechanical ventilati
    .groups="drop"
## # A tibble: 2 x 5
    icu_status n_sex_female n_sex_male n_mechvent_yes n_mechvent_no
                      <int>
                                 <int>
                                                <int>
## 1 ICU
                                                                 20
                         24
                                    41
                                                   46
## 2 NonICU
                         27
                                    33
                                                                 55
# Summarize continuous variables (age, ferritin, CRP) by ICU status
df %>%
 group_by(icu_status) %>%
 summarise(
   age_mean = mean(age, na.rm=TRUE), # Calculate mean age
   age_sd = sd(age, na.rm=TRUE),
                                     # Calculate standard deviation of age
   ferr_median = median(ferritin, na.rm=TRUE), # Median ferritin level
   ferr_IQR1 = quantile(ferritin, 0.25, na.rm=TRUE), # 25th percentile (Q1) of ferritin
   ferr_IQR3 = quantile(ferritin, 0.75, na.rm=TRUE), # 75th percentile (Q3) of ferritin
   crp mean = mean(crp, na.rm=TRUE), # Mean CRP level
   crp_sd = sd(crp, na.rm=TRUE)
                                      # Standard deviation of CRP
 )
## # A tibble: 2 x 8
   icu_status age_mean age_sd ferr_median ferr_IQR1 ferr_IQR3 crp_mean crp_sd
    <chr>>
                  <dbl> <dbl>
                                     <dbl>
                                               <dbl>
                                                         <dbl>
                                                                  <dbl> <dbl>
## 1 ICU
                   63.5
                          14.0
                                       685
                                                 325
                                                          1212
                                                                   150. 106.
## 2 NonICU
                   58.7
                          17.8
                                       401
                                                 131
                                                           870
                                                                   109.
                                                                         94.4
```

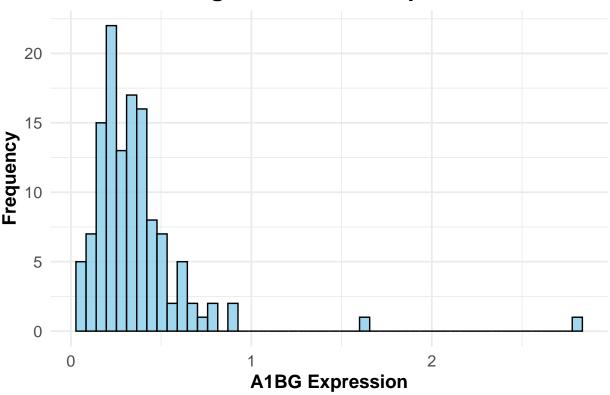
Generate final a publication quality histogram, scatter plot, and boxplot from submission 1 (i.e. only for your first gene of interest) (5 pts)

```
#1.
#Histogram of gene expression
#Select the first gene (A1BG) for analysis and convert to numeric vector
new_gene <- gene_data[1, ]

#Plot histogram for A1BG expression values across all participants
ggplot(data.frame(value = as.numeric(new_gene)), aes(x = value)) +
    geom_histogram(bins = 50, color = "black", fill = "skyblue", alpha = 0.8) +
    labs(
        title = "Histogram of A1BG Expression",
        x = "A1BG Expression",
        y = "Frequency"
    ) +</pre>
```

```
theme_minimal(base_size = 14) +
theme(
  plot.title = element_text(size = 18, face = "bold", hjust = 0.5),
  axis.title = element_text(size = 14, face = "bold"),
  axis.text = element_text(size = 12)
)
```

### **Histogram of A1BG Expression**

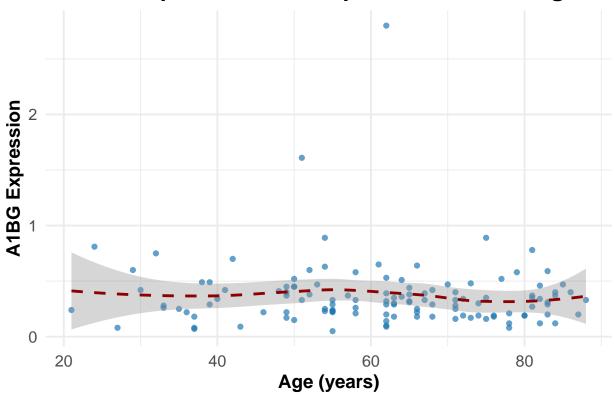


```
#2.
#Scatterplot of gene expression vs age
#Pivot gene expression row into long format for merging
gene_line1 <- new_gene %>%
  pivot_longer(cols = everything(),names_to = "participant_id",values_to = "A1BG_value")
# Merge with metadata by participant_id to obtain age and other covariates
new_df<-merge(series,gene_line1,by="participant_id")</pre>
#Prepare age levels for the x-axis: numeric ages in ascending order, then special categories
num_ages <- sort(as.numeric(unique(new_df$age)[!grepl("[^0-9]", unique(new_df$age))]))</pre>
special_ages <- unique(new_df$age)[grepl("[^0-9]", unique(new_df$age))]</pre>
age_levels <- c(as.character(num_ages), special_ages)</pre>
# Scatterplot: A1BG expression vs age (x-axis shows all ages in specified order)
new_df <- new_df %>%
  mutate(age_num = suppressWarnings(as.numeric(trimws(age)))) %>% # avoid coercion warnings
  filter(is.finite(age_num), is.finite(A1BG_value))
                                                                     # drop NAs/Inf before ggplot
```

```
ggplot(new_df, aes(x = age_num, y = A1BG_value)) +
  geom_point(alpha = 0.7, color = "#1F77B4") +
  geom_smooth(method = "loess", se = TRUE, color = "darkred", linetype = "dashed") +
  labs(
    title = "Scatterplot of A1BG Expression versus Age",
    x = "Age (years)",
    y = "A1BG Expression"
) +
  theme_minimal(base_size = 14) +
  theme(
    plot.title = element_text(size = 18, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 14, face = "bold"),
    axis.text = element_text(size = 12)
)
```

## 'geom\_smooth()' using formula = 'y ~ x'

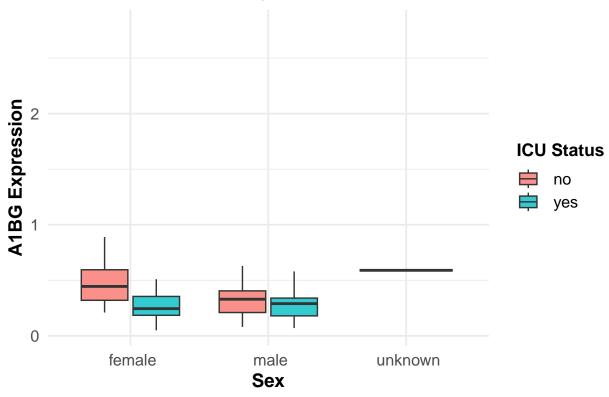
## Scatterplot of A1BG Expression versus Age



```
#3.
#Boxplot of gene expression by sex and ICU status
#sex: categorical variable; icu_status: categorical variable
ggplot(new_df, aes(x = sex, y = A1BG_value, fill = icu_status)) +
    geom_boxplot(alpha = 0.8, outlier.shape = NA) +
    labs(
        title = "A1BG Expression by Sex and ICU Status",
```

```
x = "Sex",
y = "A1BG Expression",
fill = "ICU Status"
) +
theme_minimal(base_size = 14) +
theme(
  plot.title = element_text(size = 18, face = "bold", hjust = 0.5),
  axis.title = element_text(size = 14, face = "bold"),
  axis.text = element_text(size = 12),
  legend.title = element_text(size = 13, face = "bold"),
  legend.text = element_text(size = 12)
)
```

# A1BG Expression by Sex and ICU Status



Generate a heatmap (5 pts) Heatmap should include at least 10 genes Include tracking bars for the 2 categorical covariates in your boxplot Heatmaps should include clustered rows and columns

```
# Convert gene expression data into a numeric matrix
mat <- as.matrix(gene_data)

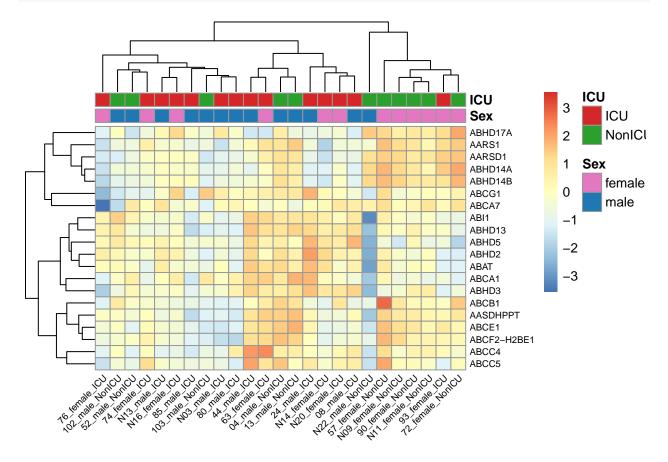
# Find overlapping participant IDs between gene expression and metadata
common <- intersect(colnames(mat), series$participant_id)

# Keep only common participants in the gene expression matrix
mat <- mat[, common, drop = FALSE]

# Subset metadata for the same participants and align the order</pre>
```

```
series_sub <- series %>% filter(participant_id %in% common) %>%
  arrange(match(participant_id, colnames(mat)))
# Log2-transform the expression values to stabilize variance
mat_log <- log2(mat + 1)</pre>
# Select the top 20 most variable genes across samples
topN <- 20
top_idx <- order(apply(mat_log, 1, var, na.rm = TRUE), decreasing = TRUE)[1:topN]</pre>
sub <- mat_log[top_idx, , drop = FALSE]</pre>
# Clean up metadata variables: sex and ICU status
sex_clean <- trimws(series_sub$sex)</pre>
icu_raw <- trimws(series_sub$icu_status)</pre>
# Convert ICU status to binary factor
icu_clean <- ifelse(tolower(icu_raw) == "yes", "ICU", "NonICU")</pre>
# Convert sex into male/female categories
sex_clean <- ifelse(grep1("^f", tolower(sex_clean)), "female", "male")</pre>
# Prepare annotation dataframe for heatmap
annotationData <- data.frame(</pre>
 Sex = factor(sex_clean, levels = c("female", "male")),
 ICU = factor(icu_clean, levels = c("ICU", "NonICU")),
 row.names = series_sub$participant_id
# Define color scheme for annotations
annotationColors <- list(</pre>
 Sex = c(female = "#E377C2", male = "#1F77B4"),
 ICU = c(ICU = "#D62728", NonICU = "#2CA02C")
# Randomly select 25 samples for visualization
set.seed(100)
subset_cols <- sample(colnames(sub), 25)</pre>
sub25 <- sub[, subset_cols, drop = FALSE]</pre>
ann25 <- annotationData[subset_cols, , drop = FALSE]</pre>
# Shorten column labels for readability (keep ID, sex, and ICU status)
short_labels <- ifelse(</pre>
  grepl("^COVID", colnames(sub25)),
  sub("^COVID_(\d+)_\d+y_([^]+)_(ICU|NonICU)$",
      "\\1_\\2_\\3", colnames(sub25)),
 sub("^NONCOVID_(\\d+)_\\d+y_([^_]+)_(ICU|NonICU)$",
      "N\\1_\\2_\\3", colnames(sub25))
# Draw heatmap of top variable genes with sample annotations
pheatmap(
  sub25,
  scale = "row",
                                            # scale genes across samples
```

```
clustering_distance_cols = "euclidean", # distance metric for columns
clustering_distance_rows = "euclidean", # distance metric for rows
clustering_method = "complete",  # hierarchical clustering method
annotation col = ann25,
                                     # add sample metadata as annotation
annotation_colors = annotationColors, # use predefined colors
show_colnames = TRUE,
                                       # display sample labels
labels_col = short_labels,
                                       # use shortened labels
fontsize_col = 7,
                                       # font size for column labels
angle_col = 45,
                                       # rotate labels for clarity
fontsize_row = 7,
                                       # font size for row labels
```



Going through the documentation for ggplot2, generate a plot type that we did not previously discuss in class that describes your data in a new and unique way (5 pts)

```
# Select top N genes with highest variance across samples
topN <- 50
top_idx <- order(apply(mat_log, 1, var, na.rm = TRUE), decreasing = TRUE)[1:topN]
mat_top <- mat_log[top_idx, , drop = FALSE]

# PCA on samples (transpose), centered & scaled
pca <- prcomp(t(mat_top), center = TRUE, scale. = TRUE)

# First two PCs to data frame
pca_df <- as.data.frame(pca$x[, 1:2])
pca_df$participant_id <- rownames(pca_df)</pre>
```

```
# Clean metadata; standardize labels
meta <- series_sub %>%
 mutate(
   sex = tolower(trimws(sex)),
   sex = ifelse(sex %in% c("female", "male"), sex, "unknown"),
   icu_status = ifelse(tolower(trimws(icu_status)) == "yes", "ICU", "NonICU")
  ) %>%
  select(participant_id, sex, icu_status)
# Join and coerce to factors (good for legends/shapes)
pca_df <- dplyr::left_join(pca_df, meta, by = "participant_id") %>%
 mutate(
   sex = factor(sex, levels = c("female", "male", "unknown")),
   icu_status = factor(icu_status, levels = c("ICU", "NonICU"))
  )
# Variance explained for axis labels
var_exp <- round(100 * (pca$sdev^2 / sum(pca$sdev^2)))[1:2]</pre>
# Plot: map color globally, shape only on points (no warning from stat_ellipse)
ggplot(pca df, aes(PC1, PC2, color = icu status)) +
  geom_point(aes(shape = sex), size = 2.5, alpha = 0.9) +
  stat_ellipse(aes(group = icu_status), linetype = 2) +
 labs(
   title = "PCA of Top-Variance Genes",
   x = paste0("PC1 (", var_exp[1], "%)"),
   y = paste0("PC2 (", var_exp[2], "%)"),
   color = "ICU", shape = "Sex"
  theme_minimal(base_size = 14) +
  theme(
   plot.title = element_text(size = 18, face = "bold", hjust = 0.5),
   axis.title = element_text(size = 14, face = "bold"),
   axis.text = element_text(size = 12),
   legend.title = element_text(size = 13, face = "bold"),
   legend.text = element_text(size = 12)
```

