1.27-b4

```
## [1] 66.80082 \hat{\sigma}^2 = s^2 = 66.8
```

2.28-a

```
## [82.83471, 87.05895]
```

We are 95% confident that the mean of muscle mass of age 60 women is within this interval.

2.28-b

```
## [68.45067, 101.443]
```

The prediction interval should be precise because it tell us where we expect the individual values to land. However, the range is large. ??

Appendix

```
knitr::opts_chunk$set(echo = FALSE)
# 1.27-64
file_path <- "CHO1PR27.txt"</pre>
data <- read.table(file_path)</pre>
y <- data$V1
x <- data$V2
n <- nrow(data)</pre>
model \leftarrow lm(y \sim x, data = data)
residual <- residuals(model)</pre>
sum(residual ** 2) / (n - 2)
# 2.28-a
summary_model <- summary(model)</pre>
mse <- summary_model$sigma ** 2</pre>
beta <- coefficients(model)</pre>
beta_0 <- beta[1]</pre>
beta_1 <- beta[2]
x_h <- 60
x_bar <- mean(x)</pre>
y_hat_h \leftarrow beta_0 + beta_1 * x_h
se_y_hat_h \leftarrow sqrt(mse * (1 / n + (x_h - x_bar) ** 2 / sum((x - x_bar) ** 2)))
t_{value} \leftarrow qt(0.975, df = 58)
```

```
lower_bound <- y_hat_h - t_value * se_y_hat_h
upper_bound <- y_hat_h + t_value * se_y_hat_h
cat("[", lower_bound, ", ", upper_bound, "]", sep = "")

# 2.28-b
s_Y_h_new <- sqrt(mse * (1 + 1 / n + (x_h - x_bar) ** 2 / sum((x - x_bar) ** 2)))
lower_bound <- y_hat_h - t_value * s_Y_h_new
upper_bound <- y_hat_h + t_value * s_Y_h_new

cat("[", lower_bound, ", ", upper_bound, "]", sep = "")</pre>
```