homework04

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Question 2:

```
setwd('E:/stat-visp/601')
data = read.csv("boxers.csv", header = TRUE, sep = ",")
y = data$height
y = y / mean(y) #mentioned in discussion
X = select(data,-c(Name, height))
lambda = seq(-23, 7, length.out = 500)
row1 \leftarrow rep(1, 19)
X1 = cbind(row1, X)
X1 = as.matrix(X1)
L = solve(t(X1) %*% X1)
n = nrow(X1)
p = ncol(X1)
\log_1 = \operatorname{rep}(0,500)
for (i in 1:500){
       y_{lambda} = 0
       if (lambda[i] == 0){
              y_{\text{lambda}} = \log(y)
       else{
              y_lambda = (y^(lambda[i]) - 1) / lambda[i]
       beta_lambda = L %*% t(X1) %*% y_lambda
       sigma_lambda = sqrt((t(y_lambda - X1 %*% beta_lambda) %*% (y_lambda - X1 %*% beta_lambda)) / (n - p -
       1 = -1 * n * \log(\text{sigma\_lambda}) + (\text{lambda[i]} - 1) * (\text{sum}(\log(y))) - ((n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * \log(2 * pi)) / (n - p - 1) + n * 
       log_1[i] = 1
}
g_lambda_hat = max(log_l)
alpha = 0.05
chi <- qchisq(1 - alpha, 1)</pre>
minus = rep(0,500)
for (i in 1:500){
      y_{\text{lambda}} = 0
       if (lambda[i] == 0){
      y_{\text{lambda}} = \log(y)
```

```
}
      else{
            y_lambda = (y^(lambda[i]) - 1) / lambda[i]
      beta_lambda = L %*% t(X1) %*% y_lambda
      1 = -1 * n * log(sigma_lambda) + (lambda[i] - 1) * (sum(log(y))) - ((n - p - 1) + n * log(2 * pi)) / (lambda[i] - 1) * (sum(log(y))) - ((n - p - 1) + n * log(2 * pi)) / (lambda[i] - 1) * (lambda[i] - 1) * (sum(log(y))) - ((n - p - 1) + n * log(2 * pi)) / (lambda[i] - 1) * (sum(log(y))) - ((n - p - 1) + n * log(2 * pi)) / (lambda[i] - 1) * (sum(log(y))) - ((n - p - 1) + n * log(2 * pi)) / (lambda[i] - 1) * (sum(log(y))) - ((n - p - 1) + n * log(2 * pi)) / (lambda[i] - 1) * (
     minus[i] = 1 - g_lambda_hat + chi / 2
}
min_index <- which.min(abs(minus))</pre>
minus_without_min <- minus[-min_index]</pre>
second min index <- which.min(abs(minus without min))</pre>
lambda_1 = lambda[min_index]
lambda_2 = lambda[second_min_index]
lambda_hat = lambda[which.max(log_l)]
g_lambda_1 = log_l[min_index]
g_lambda_2 = log_l[second_min_index]
plot(lambda, log_1, main = 'Manual plot for boxer data using all predictors',
                type = '1',
                xlab = expression(lambda),
               ylab = 'log-likelihood')
abline(v = lambda_1, lty = 2)
abline(v = lambda_2, lty = 2)
abline(h = g_lambda_1, lty = 2)
abline(v = lambda_hat, lty = 2)
```

Manual plot for boxer data using all predictors

