Homework2

2023-03-10

Data Input

Likelihood Function

```
# for n(200) and n(20)
like <- function(theta, time, dj){
    lj <- rep(NA, nrow(time))
    for(j in 1:nrow(time)){
        lj[j] <- dj[j]*log(pexp(time[j,2], rate = 1/theta)-pexp(time[j,1], rate = 1/theta))
    }
    return(exp(sum(lj)))
}

# for n(2000)
like.2000 <- function(theta, time, dj){
    lj <- rep(NA, nrow(time))
    for(j in 1:nrow(time)){
        lj[j] <- dj[j]*log(pexp(time[j,2], rate = 1/theta)-pexp(time[j,1], rate = 1/theta))
    }
    return(sum(lj))
}</pre>
```

Ngative Log-Likelihood Function

```
# for n(200) and n(20)
negLike <- function(theta, time, dj){
  logLike <- log(like(theta, time, dj))
  return(-logLike)
}
# for n(2000)
negLike.2000 <- function(theta, time, dj){
  logLike <- like.2000(theta, time, dj)
  return(-logLike)
}</pre>
```

Find Initial

```
theta.x \leftarrow seq(200, 1000, 0.01)
# n(2000)
like.y.2000 \leftarrow c()
for(i in 1:length(theta.x)){
  like.y.2000 \leftarrow c(like.y.2000, like.2000(theta.x[i], time, n2000))
indx.max.2000 \leftarrow which(like.y.2000 == max(like.y.2000))
max.y.2000 <- theta.x[indx.max.2000]
max.y.2000
## [1] 612.77
# n(200)
like.y.200 < - c()
for(i in 1:length(theta.x)){
  like.y.200 \leftarrow c(like.y.200, like(theta.x[i], time, n200))
indx.max.200 \leftarrow which(like.y.200 == max(like.y.200))
max.y.200 <- theta.x[indx.max.200]</pre>
max.y.200
## [1] 572.27
# n(20)
like.y.20 \leftarrow c()
for(i in 1:length(theta.x)){
  like.y.20 \leftarrow c(like.y.20, like(theta.x[i], time, n20))
indx.max.20 \leftarrow which(like.y.20 == max(like.y.20))
max.y.20 <- theta.x[indx.max.20]</pre>
max.y.20
## [1] 440.17
```

Optim

```
# n(2000)
initial.2000 <- 620
theta.hat.2000 <- optim(initial.2000, negLike.2000, time = time, dj = n2000,
                   hessian = TRUE, method = "Brent", lower = 400, upper = 900)
theta.h.2000 <- theta.hat.2000$par
theta.h.2000
## [1] 612.7727
# n(200)
initial.200 <- 550
theta.hat.200 <- optim(initial.200, negLike, time = time, dj = n200,
                   hessian = TRUE, method = "Brent", lower = 300, upper = 800)
theta.h.200 <- theta.hat.200$par
theta.h.200
## [1] 572.2742
```

Confidence Interval by Likelihood Ratio

```
LR <- function(theta, theta.hat, time, dj){
   like(theta, time, dj)/like(theta.hat, time, dj)
}
LR.2000 <- function(theta, theta.hat, time, dj){
   exp(like.2000(theta, time, dj) - like.2000(theta.hat, time, dj))
}
LR.CI <- function(theta, theta.hat, time, dj, alpha = 0.05){
   (LR(theta, theta.hat, time, dj) - exp(-qchisq(1-alpha, 1)/2))^2
}
LR.CI.2000 <- function(theta, theta.hat, time, dj, alpha = 0.05){
   (LR.2000(theta, theta.hat, time, dj) - exp(-qchisq(1-alpha, 1)/2))^2
}</pre>
```

Lower

n(2000)

U.2000\$par

```
# n(2000)
L.2000 <- optim(550, LR.CI.2000, lower = 450, upper = 650, method = "Brent",
                theta.hat = theta.hat.2000$par, time = time, dj = n2000, alpha = 0.05)
L.2000$par
## [1] 585.8676
# n(200)
L.200 <- optim(500, LR.CI, lower = 400, upper = 550, method = "Brent",
               theta.hat = theta.hat.200$par, time = time, dj = n200, alpha = 0.05)
L.200$par
## [1] 497.5785
# n(20)
L.20 <- optim(300, LR.CI, lower = 200, upper = 400, method = "Brent",
              theta.hat = theta.hat.20$par, time = time, dj = n20, alpha = 0.05)
L.20$par
## [1] 288.8806
Upper
```

theta.hat = theta.hat.2000\$par, time = time, dj = n2000, alpha = 0.05)

U.2000 <- optim(700, LR.CI.2000, lower = 600, upper = 800, method = "Brent",

Summary

```
LR.interval.2000 <- paste("[", round(L.2000$par), ",", round(U.2000$par), "]", sep = "")
LR.interval.200 <- paste("[", round(L.200$par), ",", round(U.200$par), "]", sep = "")
LR.interval.20 <- paste("[", round(L.20$par), ",", round(U.20$par), "]", sep = "")</pre>
```

Confidence interval by normal approximation

Standard Error

```
se.theta.2000 <- sqrt(1/theta.hat.2000$hessian)
se.theta.200 <- sqrt(1/theta.hat.200$hessian)
se.theta.20 <- sqrt(1/theta.hat.20$hessian)</pre>
```

Normal-Approximate CI

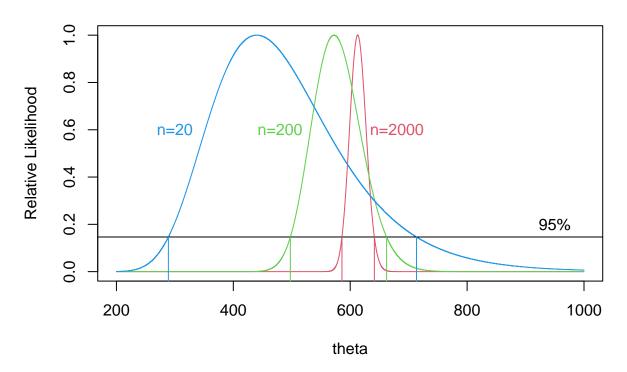
Homework(a)

```
LR.interval.20, N.interval.20))
results
##
            Inference
                         n.2000
                                    n.200
                                                n.20
## 1
          ML Estimate
                            613
                                       572
                                                 440
## 2
       Standard Error
                          14.13
                                     41.72
                                                 101
## 3 CI by Likelihood [586,641] [498,662] [289,713]
         CI by Normal [585,640] [491,654] [242,638]
```

Homework(b)

```
LR.theta.2000 <- c()
for(i in 1:length(theta.x)){
  LR.theta.2000 \leftarrow c(LR.theta.2000,
                     LR.2000(theta.x[i],theta.hat.2000$par, time, n2000))
LR.theta.200 <- c()
for(i in 1:length(theta.x)){
  LR.theta.200 \leftarrow c(LR.theta.200,
                    LR(theta.x[i], theta.hat.200$par, time, n200))
LR.theta.20 <- c()
for(i in 1:length(theta.x)){
  LR.theta.20 <- c(LR.theta.20,
                   LR(theta.x[i], theta.hat.20$par, time, n20))
}
plot(theta.x, LR.theta.2000, type = "1", col = "2",
     xlab = "theta", ylab = "Relative Likelihood", main = "Likelihood Ratio R(theta)")
lines(theta.x, LR.theta.200, col = "3")
lines(theta.x, LR.theta.20, col = "4")
lines(c(0, 1200), c(exp(-qchisq(0.95, 1)/2), exp(-qchisq(0.95, 1)/2)))
lines(c(L.2000$par, L.2000$par),
      c(-1, exp(-qchisq(0.95, 1)/2)), col = 2)
lines(c(U.2000$par, U.2000$par),
      c(-1, exp(-qchisq(0.95, 1)/2)), col = 2)
lines(c(L.200$par, L.200$par),
      c(-1, exp(-qchisq(0.95, 1)/2)), col = 3)
lines(c(U.200$par, U.200$par),
      c(-1, exp(-qchisq(0.95, 1)/2)), col = 3)
lines(c(L.20$par, L.20$par),
      c(-1, exp(-qchisq(0.95, 1)/2)), col = 4)
lines(c(U.20$par, U.20$par),
      c(-1, exp(-qchisq(0.95, 1)/2)), col = 4)
text(950, 0.2, "95%")
text(680, 0.6, "n=2000", col = 2)
text(480, 0.6, "n=200", col = 3)
text(300, 0.6, "n=20", col = 4)
```

Likelihood Ratio R(theta)



Homework(c)

When the confidence level is the same at 95%, as the sample size increases, the standard error becomes smaller and the width of the confidence interval also becomes smaller.