流行病學與生物統計計算 Homework 11

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Homework7

#EX 22-1

```
# n = 30
n < -30
norep <- 1000
beta <- c(-6, 1, 0.005)
y <- c()
mle30 <- matrix(NA, norep, 3)</pre>
for (i in 1 : norep) {
   set.seed(i)
   gpa <- rnorm(n = n, mean = 3.1, sd = 0.3)
   gre <- rnorm(n = n, mean = 580, sd = 80)
   linear <- beta[1] + beta[2] * gpa + beta[3] * gre</pre>
   pii <- exp(linear) / (1 + exp(linear))</pre>
   x <- cbind(rep(1, n), gpa, gre)
   for (k in 1 : n) {
       y[k] \leftarrow sample(c(0, 1), 1, c(1 - pii[k], pii[k]), replace = FALSE)
   }
   ftn <- function(betacoef) {</pre>
       pi1 \leftarrow exp(x \% \% betacoef) / (1 + exp(x \% \% betacoef))
       gradient \leftarrow t(x) % (y - pi1)
       hessian \leftarrow -t(x) \% \% diag(c(pi1 * (1 - pi1)), n) \% \% x
       return(list(gradient, hessian))
   }
   highnew <- function(ftn, x0, tol, maxiter) {</pre>
       x <- x0
       fx \leftarrow ftn(x)
       iter <- 0
       while ((\max(abs(fx[[1]])) > tol) \&\& (iter < maxiter)) {
```

```
x \leftarrow x - (solve(fx[[2]]) % * fx[[1]])
          fx \leftarrow ftn(x)
          iter <- iter + 1
       }
       if (max(abs(fx[[1]])) > tol) {
          cat("algorithm failed to converge\n")
          return(NA)
       } else {
          cat("algorithm converges\n")
          return(x)
      }
   }
   mle30[i, ] <- highnew(ftn, x0 = c(0, 0, 0), tol = 1e-9, maxiter = 100)
mle30
meanmle30 <- colSums(mle30) / norep</pre>
(bias30 <- meanmle30 - beta)</pre>
```

MLE of regression coefficient

Bias of MLE of regression coefficient

```
> (meanmle30 <- colSums(mle30) / norep)
[1] -7.25114905 1.18544476 0.00620335
> (bias30 <- meanmle30 - beta)
[1] -1.25114905 0.18544476 0.00120335</pre>
```

```
\# n = 230
n <- 230
norep <- 1000
beta <- c(-6, 1, 0.005)
y < -c()
mle230 <- matrix(NA, norep, 3)</pre>
for (i in 1 : norep) {
    set.seed(i)
    gpa <- rnorm(n = n, mean = 3.1, sd = 0.3)
   gre <- rnorm(n = n, mean = 580, sd = 80)
    linear \leftarrow beta[1] + beta[2] * gpa + beta[3] * gre
    pii <- exp(linear) / (1 + exp(linear))</pre>
   x <- cbind(rep(1, n), gpa, gre)
   for (k in 1 : n) {
       y[k] \leftarrow sample(c(0, 1), 1, c(1 - pii[k], pii[k]), replace = FALSE)
   }
    ftn <- function(betacoef) {</pre>
       pi1 \leftarrow exp(x \% *\% betacoef) / (1 + exp(x \% *\% betacoef))
       gradient \leftarrow t(x) % * (y - pi1)
       hessian \leftarrow -t(x) \% \% diag(c(pi1 * (1 - pi1)), n) \% \% x
       return(list(gradient, hessian))
   }
    highnew <- function(ftn, x0, tol, maxiter) {</pre>
       x <- x0
       fx \leftarrow ftn(x)
       iter <- 0
       while ((\max(abs(fx[[1]])) > tol) \&\& (iter < maxiter)) {
           x \leftarrow x - (solve(fx[[2]]) % * fx[[1]])
           fx \leftarrow ftn(x)
           iter <- iter + 1
       }
       if (max(abs(fx[[1]])) > tol) {
```

```
cat("algorithm failed to converge\n")
         return(NA)
      } else {
         cat("algorithm converges\n")
         return(x)
     }
   }
  mle230[i, ] <- highnew(ftn, x0 = c(0, 0, 0), tol = 1e-9, maxiter = 100)
mle230
meanmle230 <- colSums(mle230) / norep</pre>
(bias230 <- meanmle230 - beta)</pre>
                > (meanmle230 <- colSums(mle230) / norep)</pre>
 MLE of regression
                 [1] -6.130837632 1.022874614 0.005112127
 coefficient
                 > (bias230 <- meanmle230 - beta)</pre>
 Bias of MLE of
                 [1] -0.1308376319 0.0228746136 0.0001121271
 regression coefficient
```

```
# n = 430
n <- 430
norep <- 1000
beta <- c(-6, 1, 0.005)
y < -c()
mle430 <- matrix(NA, norep, 3)</pre>
for (i in 1 : norep) {
    set.seed(i)
    gpa <- rnorm(n = n, mean = 3.1, sd = 0.3)
   gre <- rnorm(n = n, mean = 580, sd = 80)
    linear \leftarrow beta[1] + beta[2] * gpa + beta[3] * gre
    pii <- exp(linear) / (1 + exp(linear))</pre>
   x <- cbind(rep(1, n), gpa, gre)
   for (k in 1 : n) {
       y[k] \leftarrow sample(c(0, 1), 1, c(1 - pii[k], pii[k]), replace = FALSE)
   }
    ftn <- function(betacoef) {</pre>
       pi1 \leftarrow exp(x \% *\% betacoef) / (1 + exp(x \% *\% betacoef))
       gradient \leftarrow t(x) % * (y - pi1)
       hessian \leftarrow -t(x) \% \% diag(c(pi1 * (1 - pi1)), n) \% \% x
       return(list(gradient, hessian))
   }
    highnew <- function(ftn, x0, tol, maxiter) {</pre>
       x <- x0
       fx \leftarrow ftn(x)
       iter <- 0
       while ((\max(abs(fx[[1]])) > tol) \&\& (iter < maxiter)) {
           x \leftarrow x - (solve(fx[[2]]) % * fx[[1]])
           fx \leftarrow ftn(x)
           iter <- iter + 1
       }
       if (max(abs(fx[[1]])) > tol) {
```

```
cat("algorithm failed to converge\n")
         return(NA)
      } else {
         cat("algorithm converges\n")
         return(x)
      }
   }
  mle430[i, ] <- highnew(ftn, x0 = c(0, 0, 0), tol = 1e-9, maxiter = 100)
mle430
meanmle430 <- colSums(mle430) / norep</pre>
(bias430 <- meanmle430 - beta)</pre>
                  > (meanmle430 <- colSums(mle430) / norep)</pre>
 MLE of regression
                  [1] -6.09723871 1.01552158 0.00509303
 coefficient
                  > (bias430 <- meanmle430 - beta)</pre>
 Bias of MLE of
                  [1] -9.723871e-02 1.552158e-02 9.303015e-05
 regression coefficient
```

```
# n = 630
n <- 630
norep <- 1000
beta <- c(-6, 1, 0.005)
y < -c()
mle630 <- matrix(NA, norep, 3)</pre>
for (i in 1 : norep) {
    set.seed(i)
    gpa <- rnorm(n = n, mean = 3.1, sd = 0.3)
   gre <- rnorm(n = n, mean = 580, sd = 80)
    linear \leftarrow beta[1] + beta[2] * gpa + beta[3] * gre
    pii <- exp(linear) / (1 + exp(linear))</pre>
   x <- cbind(rep(1, n), gpa, gre)
   for (k in 1 : n) {
       y[k] \leftarrow sample(c(0, 1), 1, c(1 - pii[k], pii[k]), replace = FALSE)
   }
    ftn <- function(betacoef) {</pre>
       pi1 \leftarrow exp(x \% *\% betacoef) / (1 + exp(x \% *\% betacoef))
       gradient \leftarrow t(x) % * (y - pi1)
       hessian \leftarrow -t(x) \% \% diag(c(pi1 * (1 - pi1)), n) \% \% x
       return(list(gradient, hessian))
   }
    highnew <- function(ftn, x0, tol, maxiter) {</pre>
       x <- x0
       fx \leftarrow ftn(x)
       iter <- 0
       while ((\max(abs(fx[[1]])) > tol) \&\& (iter < maxiter)) {
           x \leftarrow x - (solve(fx[[2]]) % * fx[[1]])
           fx \leftarrow ftn(x)
           iter <- iter + 1
       }
       if (max(abs(fx[[1]])) > tol) {
```

```
cat("algorithm failed to converge\n")
         return(NA)
      } else {
         cat("algorithm converges\n")
         return(x)
      }
   }
  mle630[i, ] <- highnew(ftn, x0 = c(0, 0, 0), tol = 1e-9, maxiter = 100)
mle630
meanmle630 <- colSums(mle630) / norep</pre>
(bias630 <- meanmle630 - beta)</pre>
                  > (meanmle630 <- colSums(mle630) / norep)</pre>
 MLE of regression
                   [1] -6.038478311 1.007797278 0.005027782
 coefficient
                  > (bias630 <- meanmle630 - beta)</pre>
```

[1] -3.847831e-02 7.797278e-03 2.778161e-05

Bias of MLE of

regression coefficient

plotting

```
setwd("/Users/raymond/Desktop/")
png(filename = "logistic1.png", width = 2000, height = 6000, res = 500)
par(mfrow = c(3, 1))
boxplot(mle30[, 1], mle230[, 1], mle430[, 1], mle630[, 1], main = "beta0")
abline(h = beta[1], col = 2)
boxplot(mle30[, 2], mle230[, 2], mle430[, 2], mle630[, 2], main = "beta1")
abline(h = beta[2], col = 2)
boxplot(mle30[, 3], mle230[, 3], mle430[, 3], mle630[, 3], main = "beta2")
abline(h = beta[3], col = 2)
dev.off()
png(filename = "logistic2.png", width = 6000, height = 6000, res = 500, main =
"beta0")
par(mfrow = c(1, 1))
boxplot(mle30[, 1], mle230[, 1], mle430[, 1], mle630[, 1])
abline(h = beta[1], col = 2)
dev.off()
png(filename = "logistic3.png", width = 6000, height = 6000, res = 500, main =
"beta1")
boxplot(mle30[, 2], mle230[, 2], mle430[, 2], mle630[, 2])
abline(h = beta[2], col = 2)
dev.off()
png(filename = "logistic4.png", width = 6000, height = 6000, res = 500, main =
"beta2")
boxplot(mle30[, 3], mle230[, 3], mle430[, 3], mle630[, 3])
abline(h = beta[3], col = 2)
dev.off()
```











