

STA 250: HW3: OPTIMIZATION MODULE

CHRISTOPHER CONLEY

1. QUESTION 1: BISECTION & NEWTON-RAPHSON

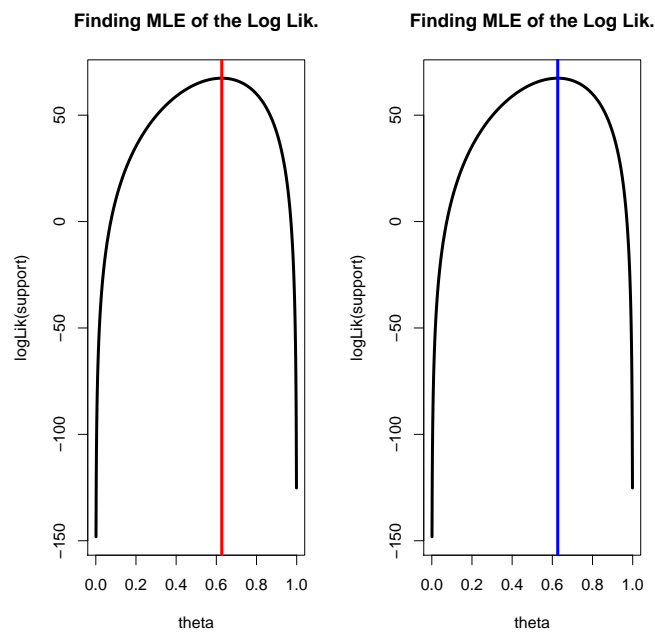


FIGURE 1. The log likelihood (y axis) is maximized with negligible differing error at $\hat{\theta} = 0.6268215$ for both bisection (red) and Newton-Raphson (blue).

1.1. **Bisection.** _____

```
bisection <- function(a, b, g, epsilon, maxIter, verbose = FALSE) {
```

```

#constraints on the paramters
stopifnot(g(a) * g(b) < 0,
          a < b,
          epsilon > 0,
          is.function(g),
          is.integer(maxIter))

#initialize
lower <- a
upper <- b
converged <- FALSE
iterCount <- 0

while (!converged && iterCount < maxIter) {

  if (verbose == TRUE) {
    cat("iteration: ", iterCount, "\n");
    cat("root (current): ", center, "\n");
  }

  iterCount <- iterCount + 1
  center <- (lower + upper) / 2
  if ( abs(g(center)) < epsilon) {
    converged <- TRUE
  } else {
    if ( g(lower) * g(center) < 0) {
      upper <- center
    } else {

```

```

    #g(center) * g(upper) < 0
    lower <- center
  }
}
}
return(center)
}

```

1.2. Newton-Raphson. ---

```

newtonRaphson <- function(g, gprime, xInit, epsilon, maxIter, verbose = FALSE) {

  #constraints on the parameters
  stopifnot(is.function(g),
            is.function(gprime),
            epsilon > 0,
            is.integer(maxIter))

  #small update function
  eta <- function(u, g, gprime) {
    - g(u) / gprime(u)
  }

  #inititalize key variables
  x <- xInit
  converged <- FALSE
  iterCount <- 0

```

```

while(!converged && iterCount < maxIter) {

  if (verbose == TRUE) {
    cat("iteration: ", iterCount, "\n");
    cat("root (current): ", x, "\n");
  }
  iterCount <- iterCount + 1

  if ( abs(g(x))< epsilon ) {
    converged <- TRUE
  }

  x <- x + eta(x, g, gprime)
}
return(x)
}

```

1.3. Linkage Problem. ---

```
setwd("~/myrepos/sta250/Stuff/HW3/")
```

```

#genotype frequencies
countAB <- 125
countAb <- 18
countaB <- 20
countab <- 34

```

```
lambdaFunc <- function(theta) {  
  1 - 2*theta + theta*theta  
}  
  
#log likelihood  
logLik <- function(lambda) {  
  countAB*log( 2 + lambda) + (countAb + countaB)*log (1 - lambda) + countab*log(lambda)  
}  
  
#first derivative of the log likelihood  
logLikPrime <- function(lambda) {  
  (countAB / (2 + lambda)) - ((countAb + countaB) / (1 - lambda)) + (countab / lambda)  
}  
  
#first derivative of the log likelihood  
logLikDoublePrime <- function(lambda) {  
  (-countAB / (2 + lambda)^2) - ((countAb + countaB) / (1 - lambda)^2) - (countab / lambda^2)  
}  
  
#find a good starting value for either algorithm  
#0 <= theta <= 1  
large.lambda <- lambdaFunc(0)  
small.lambda <- lambdaFunc(1)  
support <- seq(from = 0, to = 1, length = 1000)  
plot(support, lambdaFunc(support), type = 'l')  
  
#numerical accuracy
```

```
epsilon <- 1e-10
maxIter <- as.integer(1e5)

source("bisection.r")
lambdaBisection <- bisection(a = small.lambda, b = large.lambda,
                             g = logLikPrime, epsilon = epsilon,
                             maxIter = maxIter)

lambdaBisection

source("newton-raphson.r")
lambdaNR <- newtonRaphson(g = logLikPrime, gprime = logLikDoublePrime,
                          xInit = 0.5, epsilon = epsilon, maxIter)

lambdaNR

pdf("log-lik-mle.pdf")
par(mfrow = c(1,2))
plot(support, logLik(support), type = 'l', lwd = 3,
     main = "Finding MLE of the Log Lik.", xlab = "theta")
abline(v = lambdaBisection, col = "red", lwd = 3)

plot(support, logLik(support), type = 'l', lwd = 3,
     main = "Finding MLE of the Log Lik.", xlab = "theta")
abline(v = lambdaNR, col = "blue", lwd = 3)
dev.off()
```