Optimization HW3

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Abstract

This project is about using various optimization techniques, such as Newton-Raphson, Fisher's Scoring, Fixed point method in trying to maximize likelihood of Cauchy distribution functions. Also needs to compare speed and stability of these techniques.

Question 1

Log-likelihood function and Fisher information

$$L(\theta) = p_1 p_2 \dots p_n = \ln L(\theta) = \ln(p_1) + \dots + \ln(p_n)$$

$$l(\theta) = -\sum_{i=1}^n \ln \pi - \sum_{i=1}^n \ln[1 + (x_i - \theta)^2] = l(\theta) = -n \ln \pi - \sum_{i=1}^n \ln[1 + (x_i - \theta)^2]$$

$$l'(\theta) = -0 - \sum_{i=1}^n \frac{2(\theta - x_i)}{1 + (\theta - x_i)^2} = l'(\theta) = -2\sum_{i=1}^n \frac{(\theta - x_i)}{1 + (\theta - x_i)^2}$$

$$l''(\theta) = -2\sum_{i=1}^n \frac{1 + (\theta - x_i)^2 - 2(\theta - x_i)^2}{[1 + (\theta - x_i)^2]^2} = l''(\theta) = -2\sum_{i=1}^n \frac{1 - (\theta - x_i)^2}{[1 + (\theta - x_i)^2]^2}$$

$$2(x - \theta) \qquad 4n \quad f^{\infty} \qquad (x - \theta)^2$$

$$p'(x) = -\frac{2(x-\theta)}{\pi[1 + (x-\theta)^2]^2} = I(\theta) = \frac{4n}{\pi} \int_{-\infty}^{\infty} \frac{(x-\theta)^2}{[1 + (x-\theta)^2]^3} dx$$
$$I(\theta) = \frac{4n}{\pi} \int_{-\infty}^{\infty} \frac{x^2}{[1 + (x-\theta)^2]^3} dx$$

$$I(\theta) = \frac{4n}{\pi} \int_{-\infty}^{\infty} \frac{x^2}{[1+x^2]^3} dx$$

Let $x = tan\theta$, $1 + x^2 = \frac{1}{cos^2\theta}$, then:

$$I(\theta) = \frac{4n}{\pi} \int_{-\infty}^{\infty} \frac{\sin^2 \theta}{\cos^2 \theta} \cos^6 \theta d(\tan \theta) = I(\theta) = \frac{4n}{\pi} \int_{-\infty}^{\infty} \frac{1}{\cos^2 \theta} \sin^2 \theta \cos^4 \theta d\theta$$
$$I(\theta) = \frac{4n}{\pi} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} \frac{\sin^2 2\theta}{4} d\theta = I(\theta) = \frac{4n}{\pi} \frac{\pi}{8} = n/2$$

Question 2

Loglikelihood function and plot

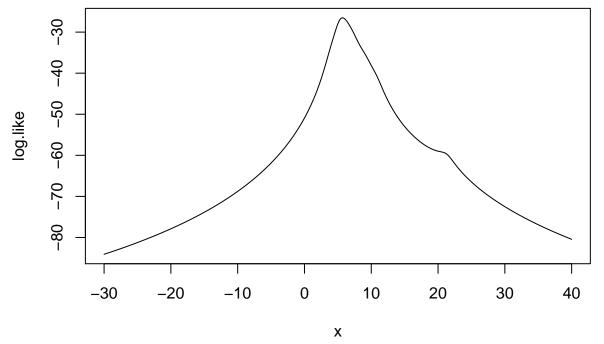
Log-likelihood function are given as follows:

$$l(\theta) = -nln\pi - \sum_{i=1}^{n} ln[1 + (x_i - \theta)^2]$$

```
set.seed(20180909)
y <- rcauchy(10, 5)

loglike <- function(x, y){
    loglike <- 0
    for (i in 1:length(y))
    {
        loglike <- loglike -log(pi) - log(1 + (x - y[i])^2)
    }
    loglike
}

curve(loglike(x,y), from=-30,to=40, n=1000 , xlab="x", ylab="log.like")</pre>
```



Question 3

Newton-Raphson method

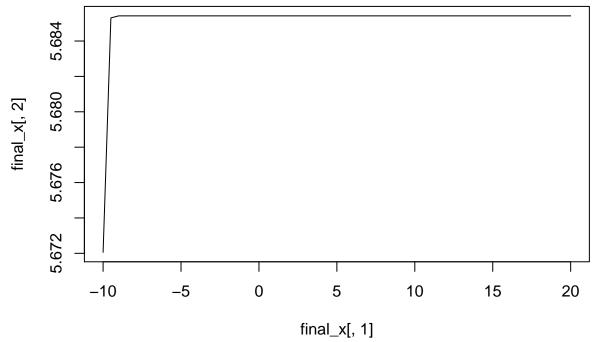
```
set.seed(20180909)

sample <- rcauchy(10, 5)

x <- sample
f <- function(theta)
{
    -2*sum((theta-x)/(1+(theta-x)^2))</pre>
```

```
}
fdash <- function(theta)</pre>
  -2*sum((1-(theta-x)^2)/((1+(theta-x)^2))^2)
g=function(theta){theta-f(theta)/fdash(theta)}
s.p \leftarrow seq(-10,20,by=0.5)
xold=5.5
xnew <- numeric(length(s.p))</pre>
for (i in 1:length(s.p)) {
  xtemp <- g(xold)</pre>
  xnew[i] <- xtemp</pre>
  xold <- xtemp</pre>
}
final_x <- cbind(s.p, xnew)</pre>
final_x
##
           s.p
                   xnew
## [1,] -10.0 5.672057
## [2,] -9.5 5.685308
## [3,] -9.0 5.685422
## [4,] -8.5 5.685422
## [5,] -8.0 5.685422
## [6,] -7.5 5.685422
## [7,] -7.0 5.685422
## [8,] -6.5 5.685422
## [9,] -6.0 5.685422
## [10,] -5.5 5.685422
## [11,] -5.0 5.685422
## [12,] -4.5 5.685422
## [13,] -4.0 5.685422
## [14,] -3.5 5.685422
## [15,] -3.0 5.685422
## [16,] -2.5 5.685422
## [17,] -2.0 5.685422
## [18,] -1.5 5.685422
## [19,]
         -1.0 5.685422
## [20,]
         -0.5 5.685422
## [21,]
          0.0 5.685422
## [22,]
           0.5 5.685422
## [23,]
           1.0 5.685422
## [24,]
           1.5 5.685422
## [25,]
           2.0 5.685422
## [26,]
           2.5 5.685422
## [27,]
           3.0 5.685422
## [28,]
           3.5 5.685422
## [29,]
           4.0 5.685422
## [30,]
           4.5 5.685422
## [31,]
           5.0 5.685422
## [32,]
           5.5 5.685422
```

```
## [33,]
           6.0 5.685422
## [34,]
           6.5 5.685422
## [35,]
           7.0 5.685422
## [36,]
           7.5 5.685422
## [37,]
           8.0 5.685422
## [38,]
           8.5 5.685422
## [39,]
           9.0 5.685422
## [40,]
           9.5 5.685422
## [41,]
          10.0 5.685422
## [42,]
          10.5 5.685422
## [43,]
          11.0 5.685422
## [44,]
          11.5 5.685422
## [45,]
          12.0 5.685422
## [46,]
          12.5 5.685422
## [47,]
          13.0 5.685422
## [48,]
          13.5 5.685422
## [49,]
          14.0 5.685422
   [50,]
          14.5 5.685422
   [51,]
          15.0 5.685422
   [52,]
          15.5 5.685422
## [53,]
          16.0 5.685422
## [54,]
          16.5 5.685422
## [55,]
          17.0 5.685422
## [56,]
          17.5 5.685422
          18.0 5.685422
## [57,]
  [58,]
          18.5 5.685422
  [59,]
          19.0 5.685422
## [60,]
          19.5 5.685422
## [61,]
          20.0 5.685422
plot(final_x[,1], final_x[,2], type = "1")
```



uniroot(f,c(1,30)) ## \$root ## [1] 5.685404 ## ## \$f.root ## [1] 9.522495e-05 ## ## \$iter ## [1] 9 ## ## \$init.it ## [1] NA ## ## \$estim.prec

The choice for inital point of theta has great impact of the function for if it is converging or not. After I tries several theta, when theta_1 is smaller than 3.68, the function tend to be extremely small when x become bigger.

Question 4

[1] 6.103516e-05

Fixed-point iterations method

```
p.fixed <- function(p0,alpha,obs,tol = 1E-6,max.iter = 1000,verbose = F){</pre>
  pold <- p0
  pnew <- pold + alpha * (-2)*sum((pold-obs)/(1+(pold-obs)^2))</pre>
  iter <- 1
  while ((abs(pnew - pold) > tol) && (iter < max.iter)){</pre>
    pold <- pnew
    pnew <- pold + alpha * (-2)*sum((pold-obs)/(1+(pold-obs)^2))
    iter <- iter + 1</pre>
    if(verbose)
      cat("At iteration", iter, "value of p is:", pnew, "\n")
  if (abs(pnew - pold) > tol) {
    cat("Algorithm failed to converge \n")
    return(c("Failed to Converge"))
  }
  else {
    cat("Algorithm converged, in :" ,iter,"iterations \n")
    return(pnew)
  }
}
set.seed(20180909)
sample <- rcauchy(10, 5)</pre>
x \leftarrow sample
s.p <- seq(-10,20,by=0.5)
```

```
p.fixed(p0 = -1,alpha = 1,obs = x)
## Algorithm failed to converge
## [1] "Failed to Converge"
p.fixed(p0 = -1, alpha = 0.64, obs = x)
## Algorithm failed to converge
## [1] "Failed to Converge"
p.fixed(p0 = -1, alpha = 0.25, obs = x)
## Algorithm converged, in : 20 iterations
## [1] 5.685421
#alpha = 1 will not converge.
s.p \leftarrow seq(-10,20,by=0.5)
alpha \leftarrow c(1, 0.64, 0.25)
i <- 1
for (i in 1:length("s.p")){
  j <- 1
 for (j in 1:3){
    result <- p.fixed(p0 = s.p[i],alpha = alpha[j],obs = x)
    print(paste0("For starting point ",s.p[i], ", Alpha ",
                 alpha[j], ". Fix-point result is ", result,"."))
 }
}
## Algorithm failed to converge
## [1] "For starting point -10, Alpha 1. Fix-point result is Failed to Converge."
## Algorithm failed to converge
## [1] "For starting point -10, Alpha 0.64. Fix-point result is Failed to Converge."
## Algorithm converged, in : 43 iterations
## [1] "For starting point -10, Alpha 0.25. Fix-point result is 5.68542149413703."
```

Question 5

Fisher scoring method

```
set.seed(20180909)
sample <- rcauchy(10, 5)
x <- sample

s.p <- seq(-10,20,by=0.5)

i <- 1

for (i in 1:length(s.p)) {
   f <- function(theta){length(x)*log(pi)+sum(log(1+(theta-x)^2))}
   grf <- function(theta){2*sum((theta-x)/(1+(theta-x)^2))}
   fs <- function(theta){matrix(length(x)/2,nrow = 1)}</pre>
```

```
print(paste0("starting point =",s.p[i]))
  z <- nlminb(s.p[i],f,grf,fs)</pre>
  print(paste0("Fisher scoring result is ",z$par))
  f \leftarrow function(theta) \{-2*sum((theta-x)/(1+(theta-x)^2))\}
  fdash \leftarrow function(theta) \{-2*sum((1-(theta-x)^2)/((1+(theta-x)^2))^2)\}
  result <- newton_raphson(z$par, f, fdash, maxiter = 1000)
  print(paste0("Newton-Rasphon result is ", result$root))
## [1] "starting point =-10"
## [1] "Fisher scoring result is 5.68542218758492"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-9.5"
## [1] "Fisher scoring result is 5.68542305603475"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-9"
## [1] "Fisher scoring result is 5.68542323805905"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-8.5"
## [1] "Fisher scoring result is 5.68542315667094"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-8"
## [1] "Fisher scoring result is 5.68542333616358"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-7.5"
## [1] "Fisher scoring result is 5.68542190522392"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-7"
## [1] "Fisher scoring result is 5.68542332287778"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-6.5"
## [1] "Fisher scoring result is 5.68542086853989"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-6"
## [1] "Fisher scoring result is 5.68542188779938"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-5.5"
## [1] "Fisher scoring result is 5.6854229428065"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-5"
## [1] "Fisher scoring result is 5.68541965606818"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-4.5"
## [1] "Fisher scoring result is 5.68542250240315"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-4"
## [1] "Fisher scoring result is 5.68542016871926"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-3.5"
## [1] "Fisher scoring result is 5.68542310453331"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =-3"
## [1] "Fisher scoring result is 5.68541970399257"
## [1] "Newton-Rasphon result is 5.68542165048946"
```

```
## [1] "starting point =-2.5"
```

- ## [1] "Fisher scoring result is 5.68542197810396"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =-2"
- ## [1] "Fisher scoring result is 5.68541967925622"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =-1.5"
- ## [1] "Fisher scoring result is 5.68542021059833"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =-1"
- ## [1] "Fisher scoring result is 5.68542075236584"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =-0.5"
- ## [1] "Fisher scoring result is 5.68542036736388"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =0"
- ## [1] "Fisher scoring result is 5.68541966001209"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =0.5"
- ## [1] "Fisher scoring result is 5.68542125596141"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =1"
- ## [1] "Fisher scoring result is 5.68542219832117"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =1.5"
- ## [1] "Fisher scoring result is 5.68542339333529"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =2"
- ## [1] "Fisher scoring result is 5.68541965603704"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =2.5"
- ## [1] "Fisher scoring result is 5.68542021634363"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =3"
- ## [1] "Fisher scoring result is 5.68541977815309"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =3.5"
- ## [1] "Fisher scoring result is 5.68542330363483"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =4"
- ## [1] "Fisher scoring result is 5.68542130439848"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =4.5"
- ## [1] "Fisher scoring result is 5.68542336158565"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =5"
- ## [1] "Fisher scoring result is 5.68541995229512"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =5.5"
- ## [1] "Fisher scoring result is 5.68542315984845"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =6"
- ## [1] "Fisher scoring result is 5.6854233059152"
- ## [1] "Newton-Rasphon result is 5.68542165048946"

```
## [1] "starting point =6.5"
```

- ## [1] "Fisher scoring result is 5.68542336806929"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =7"
- ## [1] "Fisher scoring result is 5.68542295551322"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =7.5"
- ## [1] "Fisher scoring result is 5.68542210988204"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =8"
- ## [1] "Fisher scoring result is 5.68542124032149"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =8.5"
- ## [1] "Fisher scoring result is 5.68542231100037"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =9"
- ## [1] "Fisher scoring result is 5.68542300231563"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =9.5"
- ## [1] "Fisher scoring result is 5.68542337534368"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =10"
- ## [1] "Fisher scoring result is 5.68542300848121"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =10.5"
- ## [1] "Fisher scoring result is 5.68542239609214"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =11"
- ## [1] "Fisher scoring result is 5.68542102478289"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =11.5"
- ## [1] "Fisher scoring result is 5.68542336200543"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =12"
- ## [1] "Fisher scoring result is 5.68542252180399"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =12.5"
- ## [1] "Fisher scoring result is 5.68542190109544"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =13"
- ## [1] "Fisher scoring result is 5.68542138704335"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =13.5"
- ## [1] "Fisher scoring result is 5.6854219452241"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =14"
- ## [1] "Fisher scoring result is 5.6854211813292"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =14.5"
- ## [1] "Fisher scoring result is 5.68542198606944"
- ## [1] "Newton-Rasphon result is 5.68542165048946"
- ## [1] "starting point =15"
- ## [1] "Fisher scoring result is 5.68542280902432"
- ## [1] "Newton-Rasphon result is 5.68542165048946"

```
## [1] "starting point =15.5"
  [1] "Fisher scoring result is 5.68542317119674"
  [1] "Newton-Rasphon result is 5.68542165048946"
  [1] "starting point =16"
  [1] "Fisher scoring result is 5.68542238668947"
  [1] "Newton-Rasphon result is 5.68542165048946"
  [1] "starting point =16.5"
## [1] "Fisher scoring result is 5.68542206552221"
  [1] "Newton-Rasphon result is 5.68542165048946"
  [1] "starting point =17"
  [1] "Fisher scoring result is 5.68542331806219"
  [1] "Newton-Rasphon result is 5.68542165048946"
  [1] "starting point =17.5"
## [1] "Fisher scoring result is 5.6854233062661"
## [1] "Newton-Rasphon result is 5.68542165048946"
  [1] "starting point =18"
  [1] "Fisher scoring result is 5.68542111894479"
  [1] "Newton-Rasphon result is 5.68542165048946"
  [1] "starting point =18.5"
  [1] "Fisher scoring result is 5.68542332676352"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =19"
## [1] "Fisher scoring result is 5.68542338830802"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =19.5"
## [1] "Fisher scoring result is 5.68542333986286"
## [1] "Newton-Rasphon result is 5.68542165048946"
## [1] "starting point =20"
## [1] "Fisher scoring result is 5.68542376316169"
## [1] "Newton-Rasphon result is 5.68542165048946"
```

Question 6

Comment

The project is comparing the the efficiency and stability of Newton-Raphson method, fixed-point method and scoring-Newton-Raphson method when appling those techniques to find the MLE estimator of Cauchy distribution.

Fixed point iteration is very unstable. It depends on the choice of the alpha vvery much to have the result converage. However, Newton Raphson and Fisher Newton method are both stable. The number of iteration for Fisher Newton method is samller than the one with just Newton method. Thus, Fisher Newton method is the best one with efficiency and stability.