## Quiz 2 Solutions

## March 26, 2019

## 0.1 Q1

Have  $f(x; v) = vx^{-(v+1)}$  for  $x \ge 1$ . Need to solve

$$\sum_{r=1}^{R} 1\{X_r > \gamma_i\} \frac{f(X_r)}{f(X_r; v_{i-1})} \frac{d}{dv} \{ \log[f(X_r; v)] \} = 0$$

start with

$$\frac{\mathrm{d}}{\mathrm{d}v}\log \left(vx^{-v-1}\right) = \frac{\mathrm{d}}{\mathrm{d}v}\log(v) - (v+1)\log(x) = \frac{1}{v} - \log(x).$$

Then

$$\begin{split} \sum_{r=1}^{R} 1\{X_r > \gamma_i\} \frac{f(X_r)}{f(X_r; v_{i-1})} \left(\frac{1}{v_i} - \log(X_r)\right) &= 0 \\ \frac{1}{v_i} \sum_{r=1}^{R} 1\{X_r > \gamma_i\} \frac{f(X_r)}{f(X_r; v_{i-1})} &= \sum_{r=1}^{R} 1\{X_r > \gamma_i\} \frac{f(X_r)}{f(X_r; v_{i-1})} \log(X_r) \\ \Rightarrow v_i &= \frac{\sum_{r=1}^{R} 1\{X_r > \gamma_i\} \frac{f(X_r)}{f(X_r; v_{i-1})}}{\sum_{r=1}^{R} 1\{X_r > \gamma_i\} \frac{f(X_r)}{f(X_r; v_{i-1})} \log(X_r)}. \end{split}$$

Compare with "maximum likelihood estimator" for Pareto on Wikipedia, which is

$$\hat{v} = \frac{R}{\sum_{r=1}^{R} \log(X_r)}.$$

## 0.2 Q2

```
In [1]: library(actuar, warn.conflicts = FALSE)
    set.seed(1337)

    gamma <- 100000
    maxIter <- 100
    R <- 10^6
    rho <- 0.1

    u <- c(11, 12, 13)

    v <- u

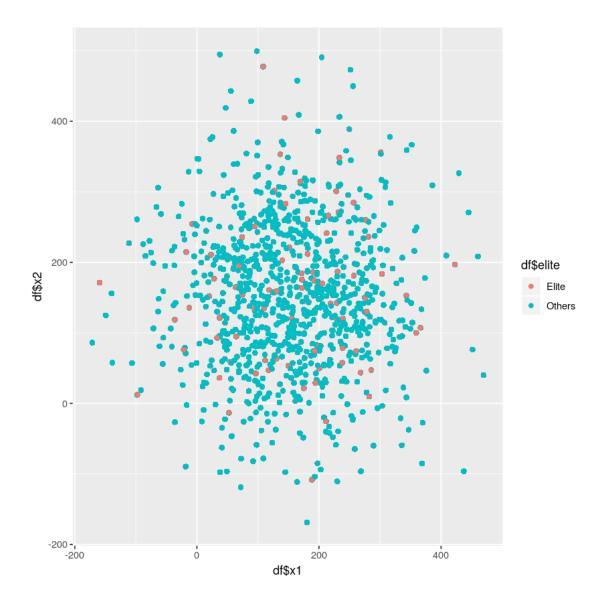
    for (i in 1:maxIter) {
        X1s <- rpareto(R, v[1], 1) + 1
        X2s <- rpareto(R, v[2], 1) + 1</pre>
```

```
X3s \leftarrow rpareto(R, v[3], 1) + 1
            Ss <- X1s + X2s + X3s
            gamma_i <- as.numeric(quantile(Ss, 1-rho))</pre>
            print("Threshold: ")
            print(gamma_i)
            if (gamma_i >= gamma)
                break
            indicators <- (Ss > gamma_i)
            # Stable way.
            LR1s <- dpareto(X1s - 1, u[1], 1) / dpareto(X1s - 1, v[1], 1)
            LR2s <- dpareto(X2s - 1, u[2], 1) / dpareto(X2s - 1, v[2], 1)
            LR3s <- dpareto(X3s - 1, u[3], 1) / dpareto(X3s - 1, v[3], 1)
            LRs <- LR1s * LR2s * LR3s
        #
              # Less numerically stable way!
              LRNumer \leftarrow dpareto(X1s - 1, u[1], 1) *
        #
        #
                          dpareto(X2s - 1, u[2], 1) *
        #
                          dpareto(X3s - 1, u[3], 1)
              LRDenom \leftarrow dpareto(X1s - 1, v[1], 1) *
        #
                          dpareto(X2s - 1, v[2], 1) *
        #
                          dpareto(X3s - 1, v[3], 1)
              LRsUnst <- LRNumer / LRDenom
              print("Stability max error: ")
              print(max(abs(LRs - LRsUnst)))
            v[1] <- sum(indicators * LRs) / sum(indicators * LRs * log(X1s))
            v[2] <- sum(indicators * LRs) / sum(indicators * LRs * log(X2s))</pre>
            v[3] <- sum(indicators * LRs) / sum(indicators * LRs * log(X3s))
            print("Next proposal dist: ")
            print(v)
        }
[1] "Threshold: "
[1] 3.500503
[1] "Next proposal dist: "
[1] 4.627691 5.438179 6.274371
[1] "Threshold: "
[1] 4.301588
[1] "Next proposal dist: "
[1] 2.021405 3.012076 4.216351
[1] "Threshold: "
[1] 6.416932
[1] "Next proposal dist: "
[1] 0.8451012 2.6014956 5.6327097
[1] "Threshold: "
[1] 18.34283
[1] "Next proposal dist: "
[1] 0.3675018 4.1716436 12.2280573
```

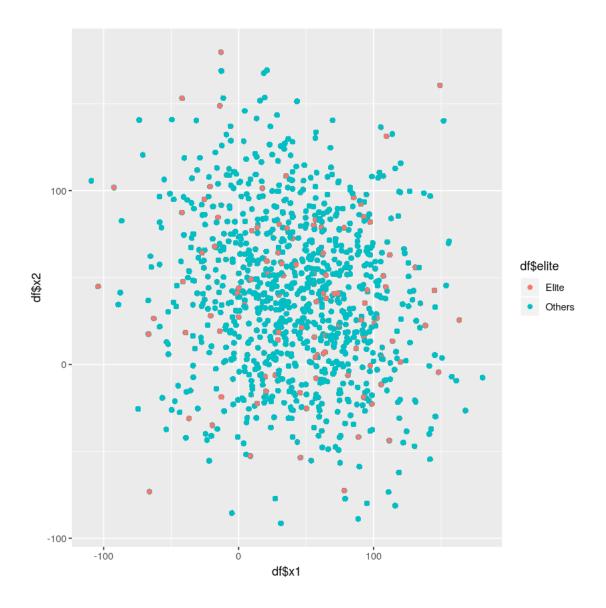
```
[1] "Threshold: "
[1] 524.6714
[1] "Next proposal dist: "
[1] 0.1575161 12.1473515 13.1947135
[1] "Threshold: "
[1] 2248143
In [2]: # Perform the importance sampling
        X1s <- rpareto(R, v[1], 1) + 1</pre>
        X2s \leftarrow rpareto(R, v[2], 1) + 1
        X3s \leftarrow rpareto(R, v[3], 1) + 1
        Ss \leftarrow X1s + X2s + X3s
        indicators <- (Ss > gamma)
        LR1s <- dpareto(X1s - 1, u[1], 1) / dpareto(X1s - 1, v[1], 1)
        LR2s <- dpareto(X2s - 1, u[2], 1) / dpareto(X2s - 1, v[2], 1)
        LR3s <- dpareto(X3s - 1, u[3], 1) / dpareto(X3s - 1, v[3], 1)
        LRs <- LR1s * LR2s * LR3s
        ests <- indicators * LRs
        estMean <- mean(ests)</pre>
        estVar <- var(ests)</pre>
        print("Estimate: ")
        print(estMean)
        q <- abs(qnorm(0.01/2))</pre>
        lowerCI <- estMean - q * sqrt(estVar / R)</pre>
        upperCI <- estMean + q * sqrt(estVar / R)</pre>
        print("99% CI's: ")
        print(c(lowerCI, upperCI))
[1] "Estimate: "
[1] 1.023112e-55
[1] "99% CI's: "
[1] 9.844785e-56 1.061746e-55
0.3 Q3
In [3]: library(ggplot2)
        library(mvtnorm)
        set.seed(1337)
        S \leftarrow function(x) \{ 20 + x[,1]^2 - 10*cos(2*pi*x[,1]) + x[,2]^2 - 10*cos(2*pi*x[,2]) \}
        R <- 10<sup>6</sup>
        rho <- 0.10
        maxIter <- 20
```

```
bestValues <- c()</pre>
        for (iter in 1:maxIter) {
             Xs <- rmvnorm(R, mu, sigma)</pre>
             Ss \leftarrow S(Xs)
             bestValues <- c(bestValues, min(Ss))</pre>
             gamma <- quantile(Ss, rho)</pre>
             elites <- Xs[Ss<gamma,]</pre>
             if (iter < 5 || iter %% 10 == 0) {</pre>
                 df \leftarrow data.frame(x1=Xs[1:1000,1], x2=Xs[1:1000,2], elite=(Ss < gamma))
                  df$elite[Ss<gamma] <- "Elite"</pre>
                 df$elite[Ss>=gamma] <- "Others"</pre>
                 print(qplot(df$x1, df$x2, color=df$elite))
             }
             mu <- c(mean(elites[,1]), mean(elites[,2]))</pre>
             sigma <- cov(elites)</pre>
             cat("Iteration: ", iter, "\tBest value: ", min(Ss),
                  "\t\tMax(mu): ", max(mu), "\tMax(sigma): ", max(sigma), "\n")
        }
        plot(1:maxIter, bestValues, log = "y")
Iteration: 1
                       Best value: 1.900643
                                                                  Max(mu): 39.23646
                                                                                                 Max(sigma): 223
```

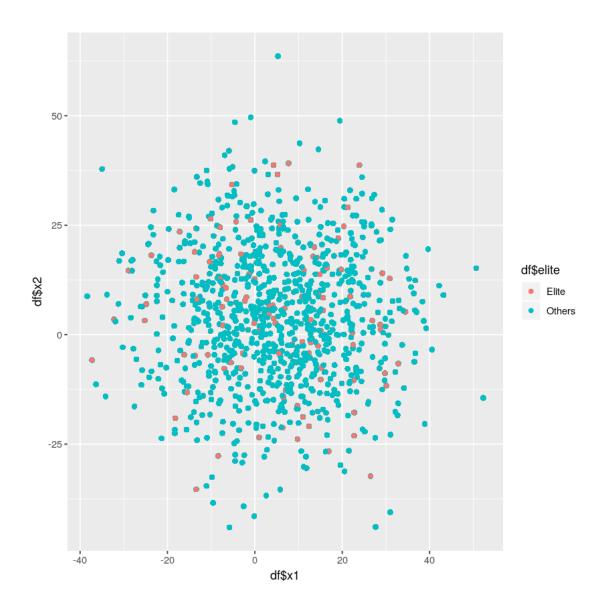
mu <- rep(150, 2)
sigma <- 10000\*diag(2)</pre>



Iteration: 2 Best value: 1.098078 Max(mu): 4.870495 Max(sigma): 227



Iteration: 3 Best value: 0.0534268 Max(mu): 0.2886972 Max(sigma): 1



 Iteration:
 4
 Best value:
 0.002592227

 Iteration:
 5
 Best value:
 0.002936498

 Iteration:
 6
 Best value:
 0.000177803

 Iteration:
 7
 Best value:
 3.184571e-05

 Iteration:
 8
 Best value:
 0.0003860067

 Iteration:
 9
 Best value:
 0.0004497498

 Max(mu):
 0.05956493
 Max(sigma):

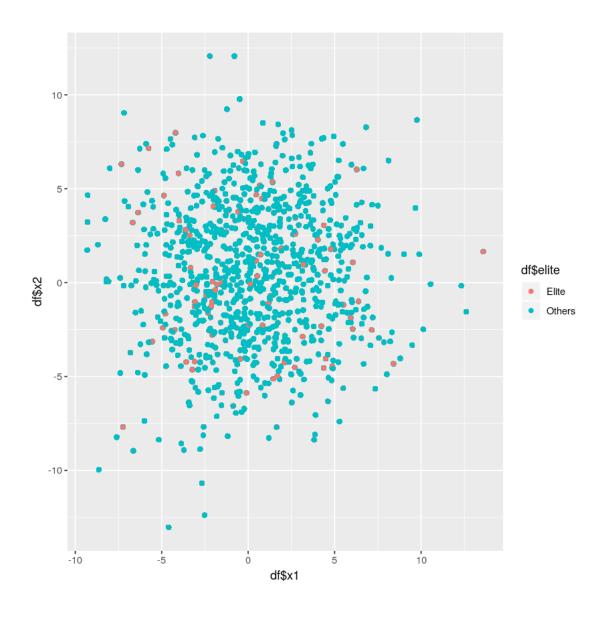
 Max(mu):
 0.02623596
 Max(sigma):

 Max(mu):
 0.01264946
 Max(sigma):

 Max(mu):
 0.006256309
 Max(sigma)

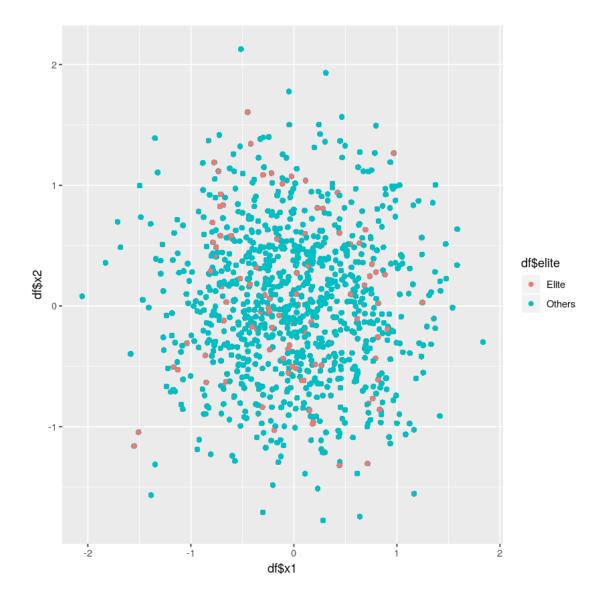
 Max(mu):
 0.005300794
 Max(sigma)

 Max(mu):
 0.0033962
 Max(sigma):

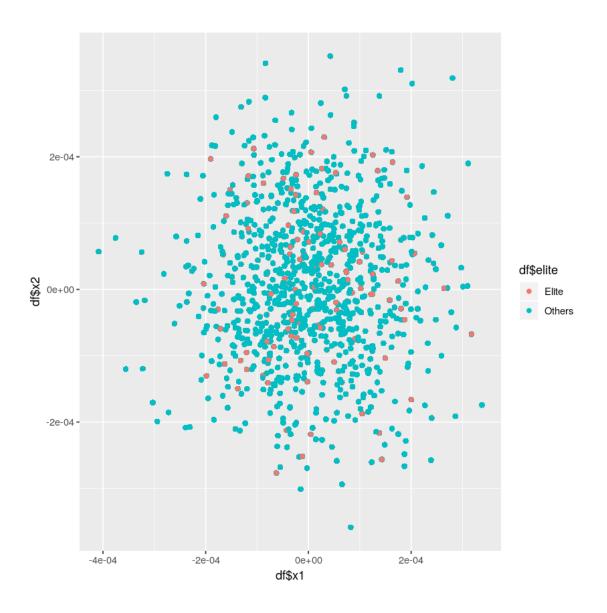


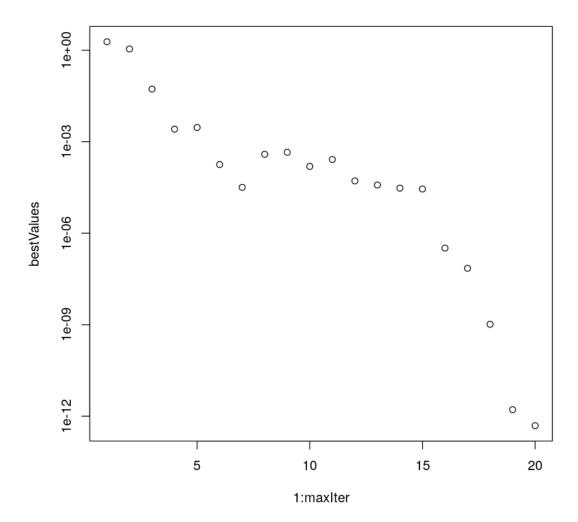
Iteration:	10	Best	value:	0.0001549733
Iteration:	11	Best	value:	0.0002611206
Iteration:	12	Best	value:	5.168094e-05
Iteration:	13	Best	value:	3.796178e-05
Iteration:	14	Best	value:	3.004398e-05
Iteration:	15	Best	value:	2.811228e-05
Iteration:	16	Best	value:	3.238833e-07
Iteration:	17	Best	value:	7.06562e-08
Iteration:	18	Best	value:	1.028054e-09
Iteration:	19	Best	value:	1.630696e-12

Max(mu):	0.004923299
Max(mu):	0.005293373
Max(mu):	0.004298656
Max(mu):	0.00240521
Max(mu):	0.001145529
Max(mu):	0.0004379217
Max(mu):	7.360588e-05
Max(mu):	8.785794e-06
Max(mu):	5.739931e-07
Max(mu):	5.687833e-08



Iteration: 20 Best value: 4.867218e-13 Max(mu): 1.168593e-07 Max(signature)





In [4]: bestValues

 1.
 1.90064338916066
 2.
 1.09807799737755
 3.
 0.0534267970766731
 4.
 0.00259222739365761

 5.
 0.00293649829586329
 6.
 0.000177802981061959
 7.
 3.18457131260885e-05
 8.
 0.000386006712775

 9.
 0.000449749756164408
 10.
 0.000154973292351812
 11.
 0.000261120617020438
 12.
 5.16809429953469e-05

 13.
 3.79617845194247e-05
 14.
 3.00439844309608e-05
 15.
 2.81122816474522e-05
 16.
 3.23883327268959e-07

 17.
 7.06561991137278e-08
 18.
 1.02805408630502e-09
 19.
 1.63069557856943e-12
 20.
 4.86721773995669e-13