

# Quiz 2 Solutions

March 29, 2019

## 0.1 Q1

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

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

start with

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

Then

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

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
```

```

X3s <- rpareto(R, v[3], 1) + 1
Ss <- X1s + X2s + X3s

gamma_i <- as.numeric(quantile(Ss, 1-rho))
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 <- dpareto(X1s - 1, u[1], 1) *
#           dpareto(X2s - 1, u[2], 1) *
#           dpareto(X3s - 1, u[3], 1)
# LRDenum <- dpareto(X1s - 1, v[1], 1) *
#           dpareto(X2s - 1, v[2], 1) *
#           dpareto(X3s - 1, v[3], 1)
# LRsUnst <- LRNumer / LRDenum

# 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))
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
  X2s <- rpareto(R, v[2], 1) + 1
  X3s <- rpareto(R, v[3], 1) + 1
  Ss <- 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)
  estVar <- var(ests)

  print("Estimate: ")
  print(estMean)

  q <- abs(qnorm(0.01/2))

  lowerCI <- estMean - q * sqrt(estVar / R)
  upperCI <- estMean + q * sqrt(estVar / R)

  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 <- function(x) { 20 + x[,1]^2 - 10*cos(2*pi*x[,1]) + x[,2]^2 - 10*cos(2*pi*x[,2])}

        R <- 10^6
        rho <- 0.10
        maxIter <- 20

```

```

mu <- rep(150, 2)
sigma <- 10000*diag(2)

bestValues <- c()

for (iter in 1:maxIter) {
  Xs <- rmvnorm(R, mu, sigma)
  Ss <- S(Xs)
  bestValues <- c(bestValues, min(Ss))

  gamma <- quantile(Ss, rho)
  elites <- Xs[Ss<gamma,]

  if (iter < 5 || iter %% 10 == 0) {
    df <- data.frame(x1=Xs[,1], x2=Xs[,2], elite=(Ss<gamma))
    df$Category[Ss<gamma] <- "Elite"
    df$Category[Ss>=gamma] <- "Others"

    print(qplot(x1, x2, color=Category, data = df[1:10000,]))
  }

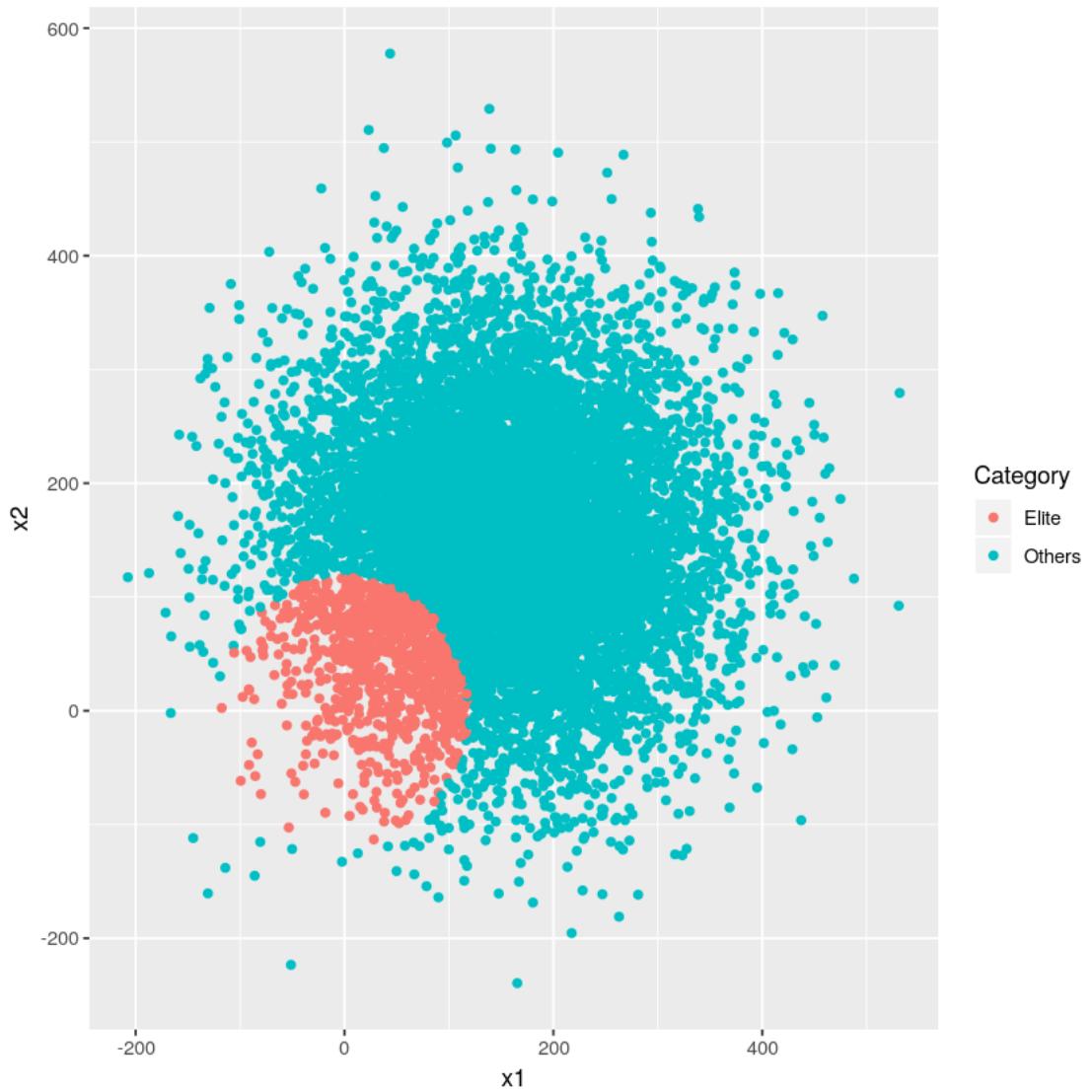
  mu <- c(mean(elites[,1]), mean(elites[,2]))
  sigma <- cov(elites)

  cat("Iteration: ", iter, "\tBest value: ", min(Ss),
      "\t\tMax(mu): ", max(mu), "\t\tMax(sigma): ", max(sigma), "\n")
}

plot(1:maxIter, bestValues, log = "y")

```

Iteration: 1              Best value: 1.900643              Max(mu): 39.23646              Max(sigma): 223

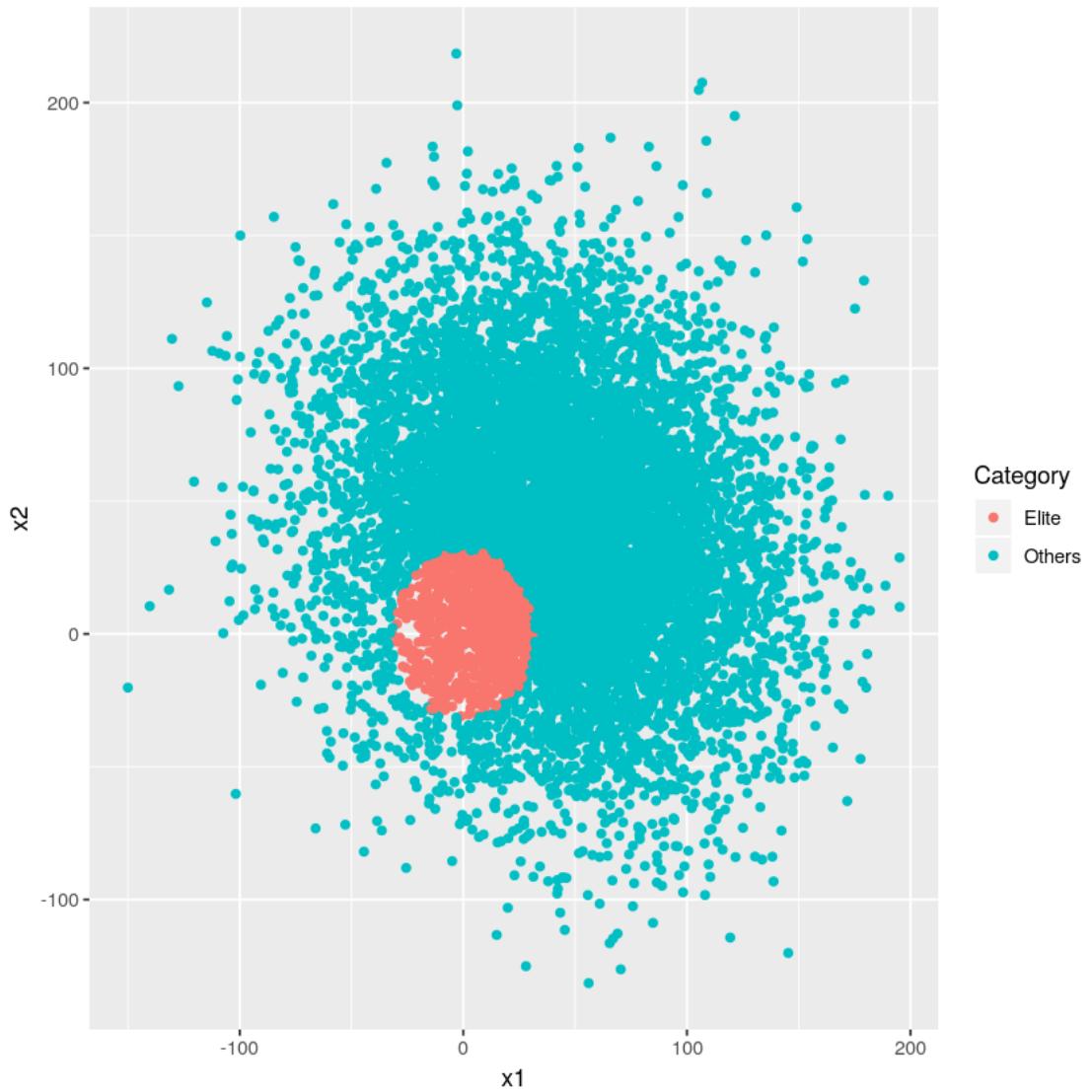


Iteration: 2

Best value: 1.098078

Max(mu): 4.870495

Max(sigma): 227

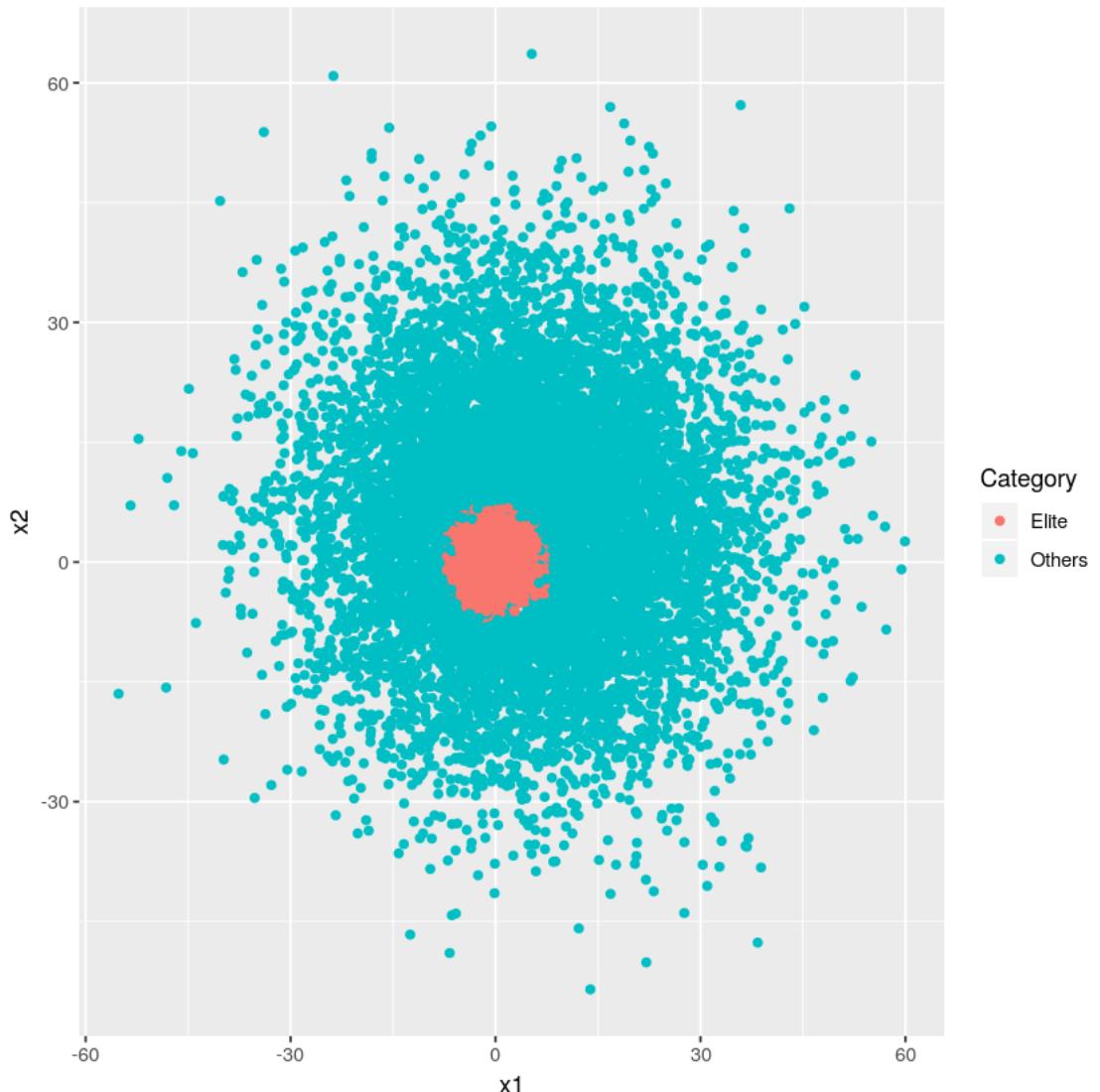


Iteration: 3

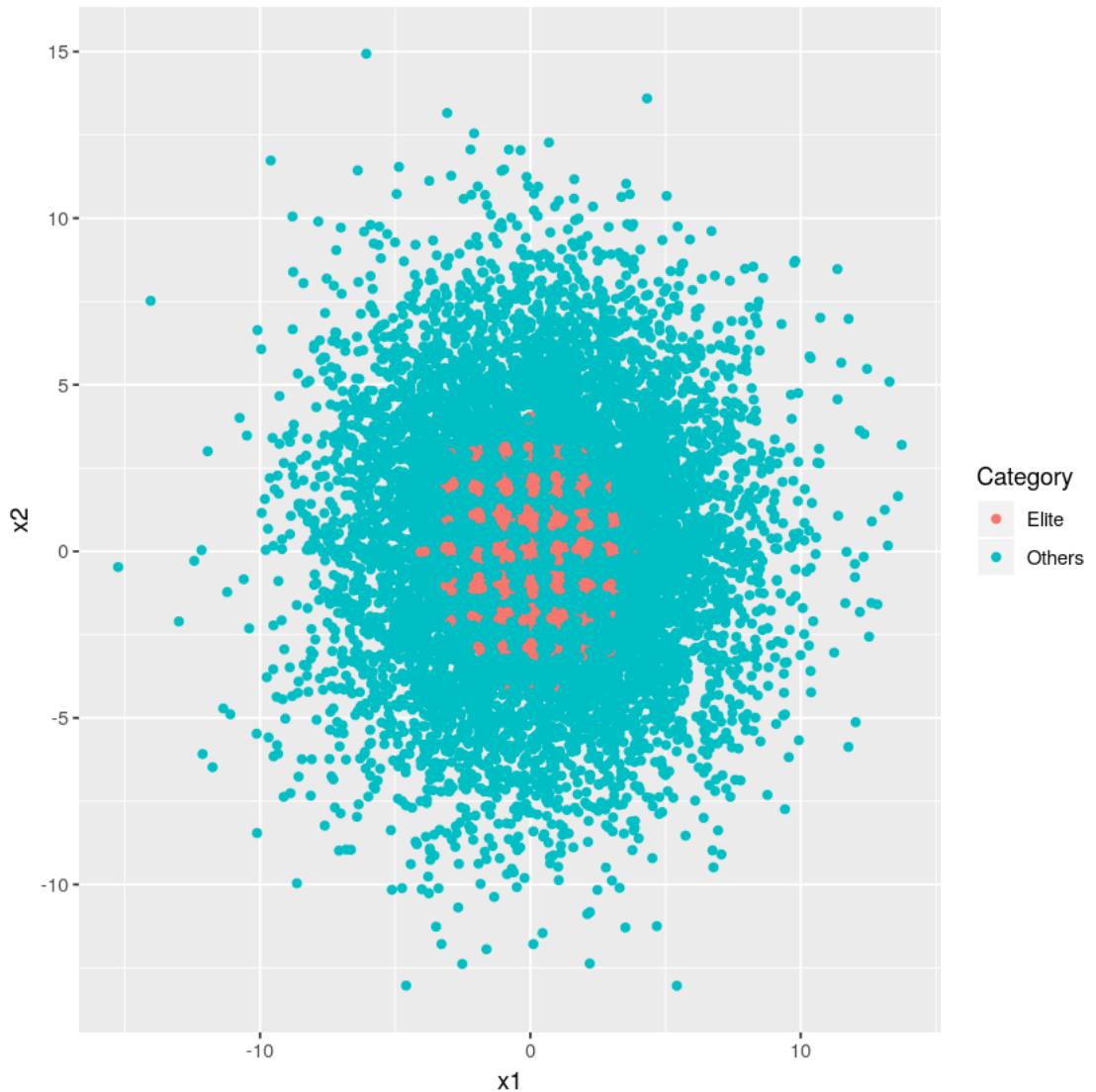
Best value: 0.0534268

Max(mu): 0.2886972

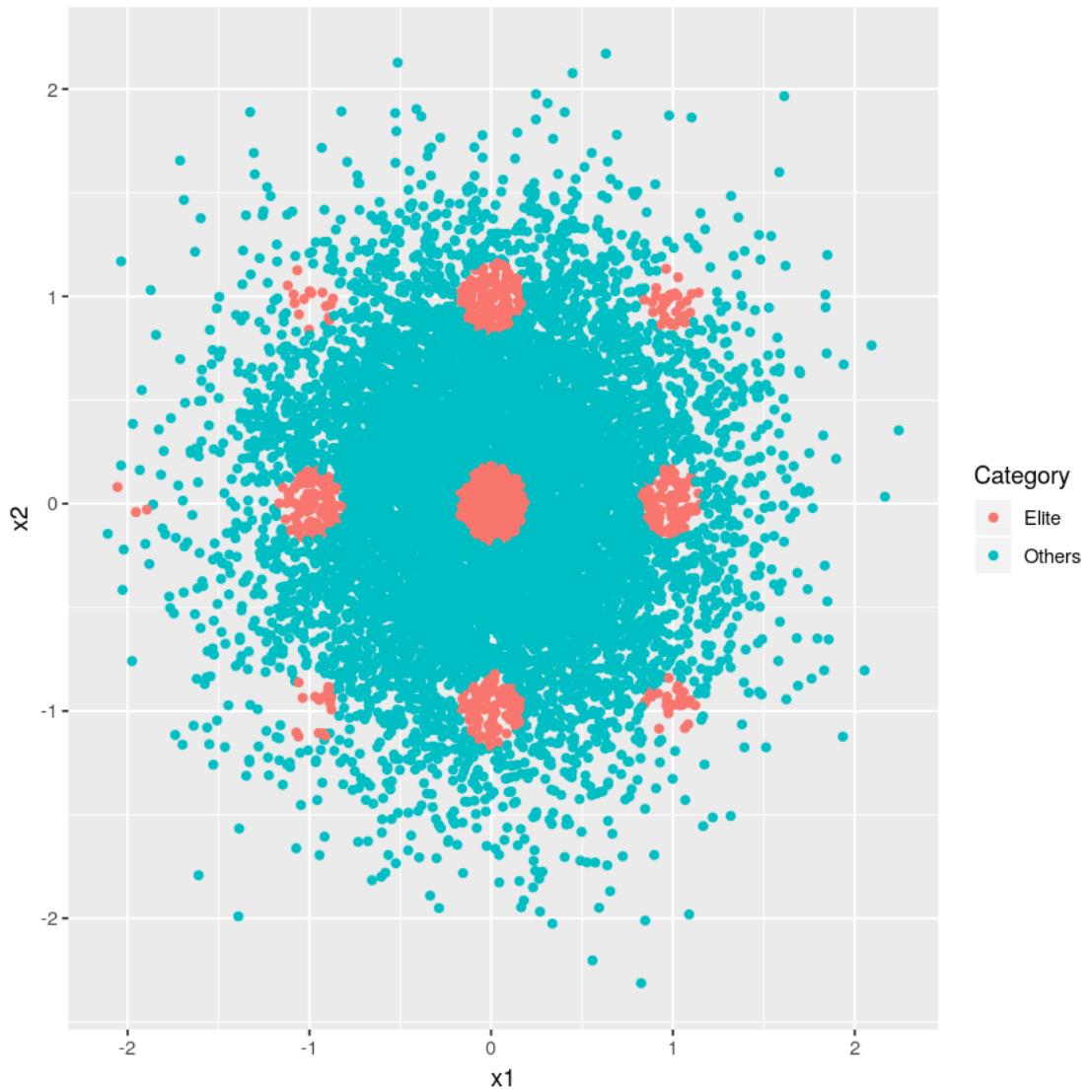
Max(sigma): 1



Iteration: 4	Best value: 0.002592227	Max(mu): 0.05956493	Max(sigma):
Iteration: 5	Best value: 0.002936498	Max(mu): 0.02623596	Max(sigma):
Iteration: 6	Best value: 0.000177803	Max(mu): 0.01264946	Max(sigma):
Iteration: 7	Best value: 3.184571e-05	Max(mu): 0.006256309	Max(sigma):
Iteration: 8	Best value: 0.0003860067	Max(mu): 0.005300794	Max(sigma):
Iteration: 9	Best value: 0.0004497498	Max(mu): 0.0033962	Max(sigma):



Iteration: 10	Best value: 0.0001549733	Max(mu): 0.004923299	Max(sigma):
Iteration: 11	Best value: 0.0002611206	Max(mu): 0.005293373	Max(sigma):
Iteration: 12	Best value: 5.168094e-05	Max(mu): 0.004298656	Max(sigma):
Iteration: 13	Best value: 3.796178e-05	Max(mu): 0.00240521	Max(sigma):
Iteration: 14	Best value: 3.004398e-05	Max(mu): 0.001145529	Max(sigma):
Iteration: 15	Best value: 2.811228e-05	Max(mu): 0.0004379217	Max(sigma):
Iteration: 16	Best value: 3.238833e-07	Max(mu): 7.360588e-05	Max(sigma):
Iteration: 17	Best value: 7.06562e-08	Max(mu): 8.785794e-06	Max(sigma):
Iteration: 18	Best value: 1.028054e-09	Max(mu): 5.739931e-07	Max(sigma):
Iteration: 19	Best value: 1.630696e-12	Max(mu): 5.687833e-08	Max(sigma):

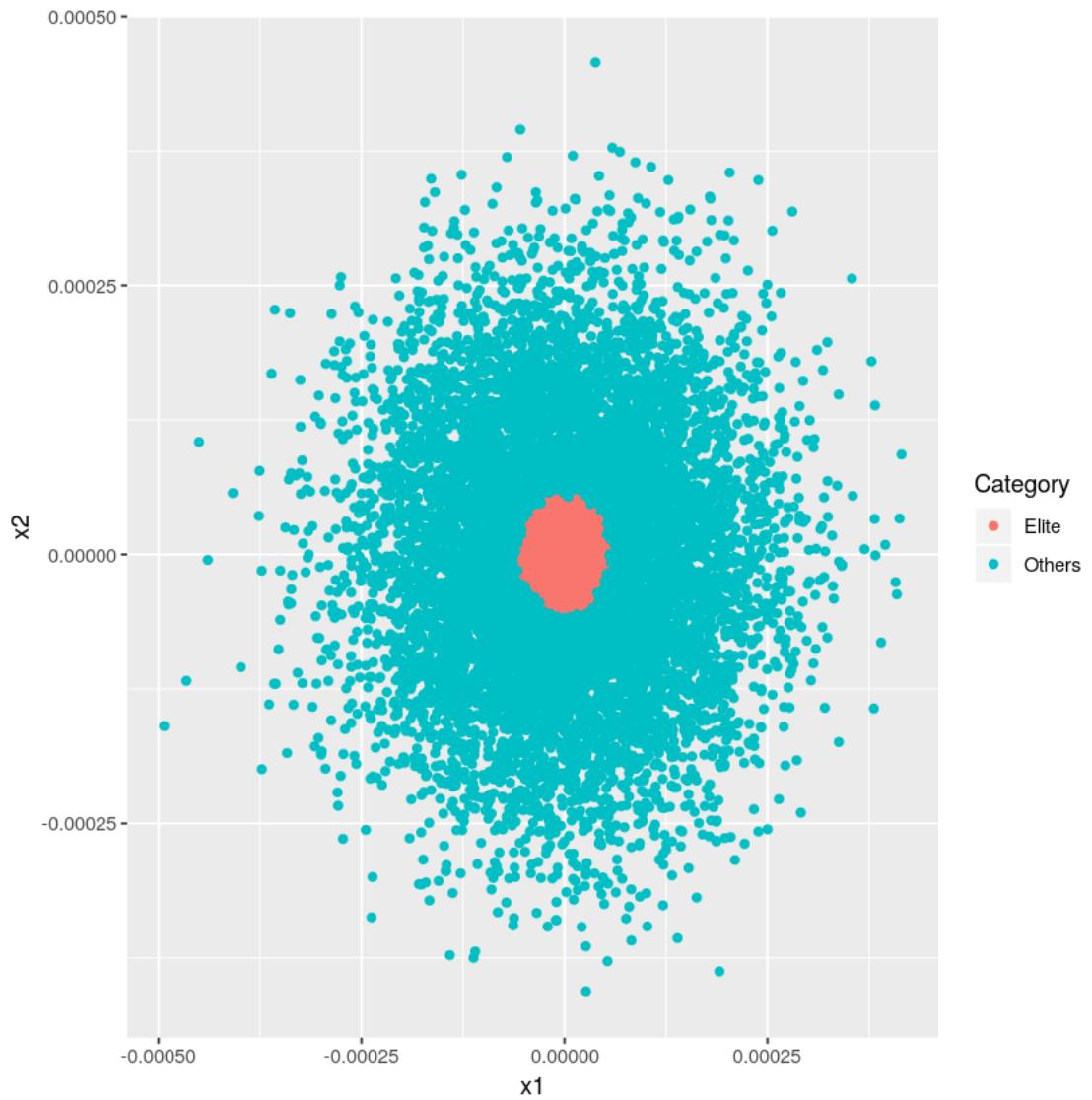


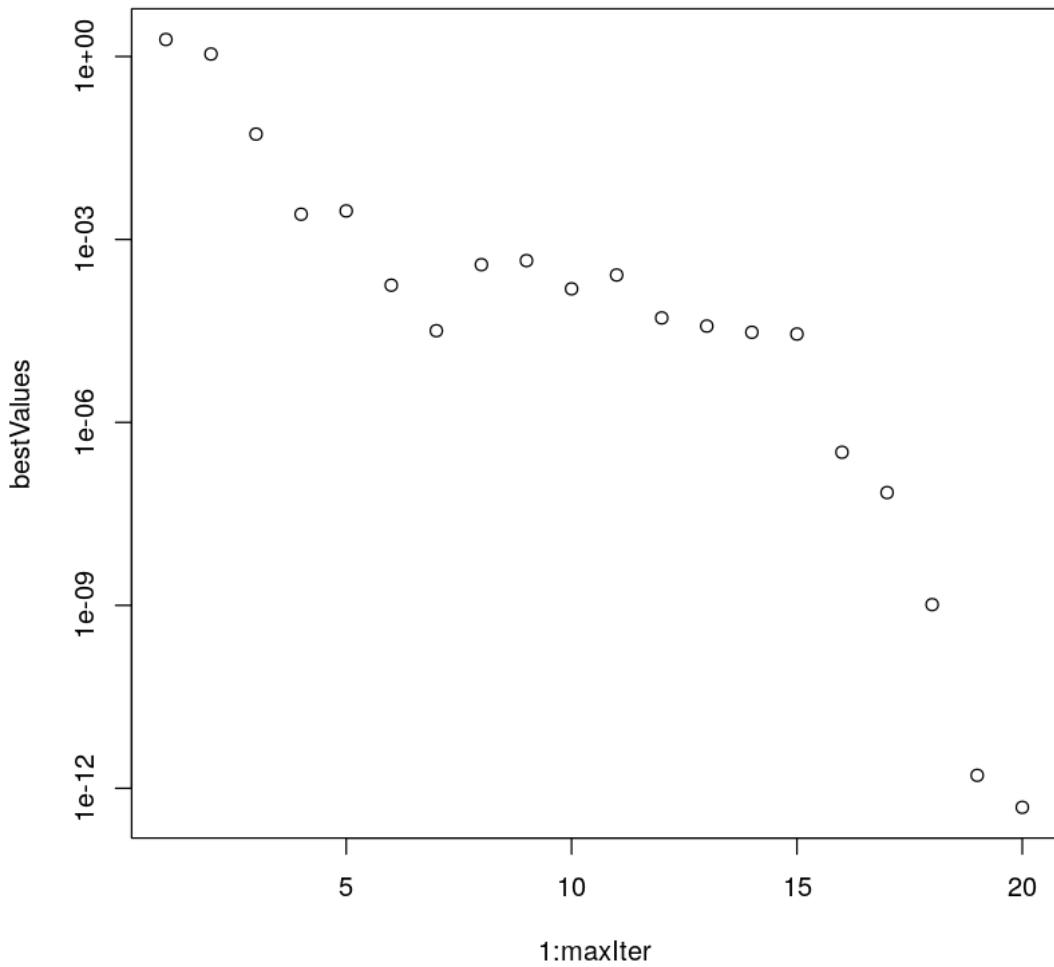
Iteration: 20

Best value: 4.867218e-13

Max(mu): 1.168593e-07

Max(sigma)





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
```