## STAT 206 Lab 6 Lihua Xu

## Due Monday, November 13, 5:00 PM

General instructions for labs: You are encouraged to work in pairs to complete the lab. Labs must be completed as an R Markdown file. Be sure to include your lab partner (if you have one) and your own name in the file. Give the commands to answer each question in its own code block, which will also produce plots that will be automatically embedded in the output file. Each answer must be supported by written statements as well as any code used.

## Agenda: Accept-Reject algorithm

1. Write a function to simulate n N(0,1) random variates using the Accept-Reject algorithm with a Cauchy candidate.

Here, our target density is the standard Normal distribution with PDF  $f(x) = \frac{1}{\sqrt{2\pi}}e^{-x^2/2}$  The Cauchy(0,1) random variable has the instrumental PDF

$$g(x) = \frac{1}{\pi \gamma} \left[ \frac{\gamma^2}{(x - x_0)^2 + \gamma^2} \right]$$

standard PDF:

$$g(x) = \frac{1}{\pi} \frac{1}{1 + x^2}, \quad -\infty < x < \infty,$$

and CDF

$$F(x) = \frac{1}{\pi} arctan(\frac{x-\mu}{\sigma}) + \frac{1}{2}$$

Standard CDF

$$F(x) = \frac{1}{\pi} arctan(x) + \frac{1}{2}$$

It is now easy to calculate the inverse transformation to simulate:

$$F^{-1}(y) = x = \sigma tan(??(y - 1/2)) + \mu$$

We know already that Cauchy has heavier tails than the Normal distribution, suggesting that there is an M such that  $Normal \leq M * Cauchy$ . That is, we should find M that satisfies the following experssion:

$$\frac{1}{\sqrt{2\pi}}e^{-x^2/2} \le M \, \frac{1}{\pi(1+x^2)}$$

We can approximate M numerically using the following function.

$$M = \sup \frac{\frac{1}{\sqrt{2\pi}} e^{-x^2/2}}{\frac{1}{\pi} \frac{1}{1+x^2}}, \quad -\infty < x < \infty,$$

And THE maximum value is M=1.5203469. This means that on the average we need to generate approximately 1.52 Cauchys before we will get a Normal.

```
Accept_Reject <- function(n){
    M <- 1.5203469
    i <-1
    Y <-c()
    while(i<(n+1)){
        u <- runif(1)
        X <- rcauchy(1,location = 0,scale = 1)
        i <- 1+i
        if (u < ((1/(sqrt(2*pi))*exp(-X^2/2))/(M*1/pi*1/(1+X^2))))
        {Y <- c(Y,X)}}
    return(Y)}</pre>
```

2. Simulate 1000 N(0,1) random variates using your function to estimate  $E[Y^3 \log(1+|Y|)]$  and  $Pr(Y \in [-1,2])$ . Be sure to include a Monte Carlo standard error with your estimates.

```
data <- Accept_Reject(1000)</pre>
Z \leftarrow data^3*log10(1+abs(data))
Expect <- mean(Z)</pre>
mcse <- sd(data) / sqrt(length(data))</pre>
#The expected value should be:
Expect
## [1] -0.1675298
#The Monte Carlo standard error will be:
mcse
## [1] 0.0379133
num <- length(data)</pre>
N <- 0
for (i in 1:num){
  if (-1<=data[i]&data[i]<=2){N=N+1}}
Probability <- N/num
#The probability should be:
Probability
```

## [1] 0.797546

3. What was the acceptance rate of the Accept-Reject algorithm? Is this close to the theoretical acceptance rate?

The acceptance rate resulted from my code is:

```
num/1000
```

```
## [1] 0.652
```

And the theoretical acceptance rate is:

```
M <- 1.5203469
1/M
```

## ## [1] 0.6577446

They are very close to each other.

4. Write a function that continues simulation until the sample size is large enough so that your Monte Carlo error is less than  $\epsilon = 0.01$  for estimating a general statistic stat (which will be a function). You function should also return the observed acceptance rate of the Accept-Reject algorithm and a Monte Carlo standard error.

```
stat <- function(n){
    M <- 1.5203469
    i <-1
    Y <-c()
    value <- c()
    for(m in 1:n){
        u <- runif(1)
        X <- rcauchy(1,location = 0,scale = 1)
        i <- 1+i
        if (u < ((1/(sqrt(2*pi))*exp(-X^2/2))/(M*1/pi*1/(1+X^2))))
        {Y <- c(Y,X)}</pre>
```

```
if ((length(Y)>=10)&(sd(Y)/(sqrt(length(Y)))<=0.01))</pre>
      {break}}
  List <- list(accpt_rate=length(Y)/i,data=Y,mcse=sd(Y)/sqrt(length(Y)),total_times=i)</pre>
  return(List)}
  5. Use your function to estimate E[Y^3 \log(1+|Y|)] and Pr(Y \in [-1,2]) with \epsilon = 0.01. Report your
     estimates along with confidence intervals based on the Monte Carlo standard error. What was the
     acceptance rate?
#n (# of times) is a very big value. Once the tolerance 0.01 was achieved, it will stop.
#The acceptance rate and the MCSE would be as the following values respectively:
OUTPUT <- stat(1000000)
OUTPUT$accpt_rate
## [1] 0.6604726
OUTPUT$mcse
## [1] 0.009999308
OUTPUT$total_times #The total time they simulate is less than n=1000000.
## [1] 15445
#Estimation of the expected value and the probability
data new <- OUTPUT$data
Z_new <- data_new^3*log10(1+abs(data_new))</pre>
Expect <- mean(Z new)</pre>
#The expected value would be:
Expect
## [1] -0.003746866
C_{interval} \leftarrow Expect + c(-1,1)*1.96*OUTPUT$mcse
#The confidence intervals should be:
C interval
## [1] -0.02334551 0.01585178
num_new <- length(data_new)</pre>
N new <- 0
for (i in 1:num_new){
  if (-1<=data_new[i]&data_new[i]<=2){N_new=N_new+1}}</pre>
Probability <- N_new/num_new
```

## [1] 0.8154103

Probability

#The probability would be: