# **Adaptive Squeezed Rejection Sampling**

Amy Chen October 23, 2018

# Overview

Adaptive squeezed rejection sampling is a method of drawing points from a target distribution, and goes a step further than rejection sampling by utilizing an automatic envelope generation strategy for squeezed rejection sampling.

Suppose we are interested in drawing points from f(x), a concave function. Let g denote another density from which we know how to sample and for which we can easily calculate g(x). Let e denote an envelope such that  $e(x) = \frac{g(x)}{\alpha} \ge f(x) \forall x$  for which f(x) > 0 for a given constant  $\alpha \le 1$ . It is simpler to generate this envelope function in the log space. Take n points on log(f(x)) and connect their tangent lines to determine log(e(x)). This ensures that when exponentiated, the envelope function encompasses f(x).

We will also define a squeeze function s(x) such that  $s(x) \le f(x) \forall x$  for which f(x) > 0. Using the selected points from generating the envelope function, connect the points to determine log(s(x)). This ensures that when exponentiated, the squeeze function is below f(x).

Then adaptive rejection sampling can be completed in the following steps:

- 1. Sample  $Y \sim g$ 2. Sample  $U \sim Unif(0, 1)$ 3. If  $U \leq \frac{s(Y)}{e(Y)}$ , keep Y4. If  $U > \frac{s(Y)}{e(Y)}$  and  $U \leq \frac{f(Y)}{e(Y)}$ , keep Y5. Otherwise, reject Y
- 6. Repeat for desired sample size

# Demonstration

Suppose we would like to estimate  $S = E[x^2]$  where *X* has density proportional to  $q(x) = e^{\frac{-|x|^3}{3}}$ 

#### **Target Function**

Let the target function be  $f(x) = e^{\frac{-|x|^3}{3}}$ Then  $log(f(x)) = \frac{-|x|^3}{3}$ 

#### **Envelope Function**

Select points  $(-1, -\frac{1}{3})$ , (0, 0), and  $(1, -\frac{1}{3})$  from log(f(x))

By computing the tangent lines at each point, finding the points of intersection, and merging the functions, we get the log of the envelope function,

Adaptive Squeezed Rejection Sampling

$$log(e(x)) = \begin{cases} 0 & \text{if } -\frac{2}{3} < x < \frac{2}{3} \\ \frac{2}{3} - |x| & otherwise \end{cases}$$

Exponentiate to get the envelope function,

$$e(x) = \begin{cases} 1 & \text{if } -\frac{2}{3} < x < \frac{2}{3} \\ e^{\frac{2}{3}} - |x| & \text{otherwise} \end{cases}$$

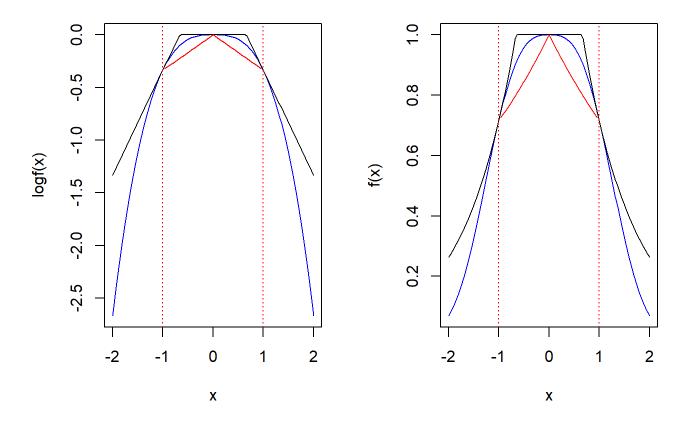
#### **Squeezing Function**

Let the log of the squeezing function be  $log(s(x)) = -\frac{|x|}{3}$  if -1 < x < 1Exponentiate to get the squeezing function  $s(x) = e^{-\frac{|x|}{3}}$  if -1 < x < 1

```
logf <- function(x) {
   -abs(x^3)/3
}
loge <- function(x) {
   ifelse( (x>-2/3)&(x<2/3), 0, 2/3-abs(x) );
}
logs <- function(x) {
   ifelse( (x>-1)&(x<1), -abs(x)/3, NA );
}</pre>
```

```
f <- function(x) {
    exp(logf(x))
}
e <- function(x) {
    exp(loge(x));
}
s <- function(x) {
    exp(logs(x));
}</pre>
```

```
par(mfrow=c(1,2))
curve(logf(x), from = -2, to = 2, col = "blue")
curve(loge(x), add = T)
curve(logs(x), add = T, col = "red")
abline(v = -1, lty = 3, col = "red")
abline(v = 1, lty = 3, col = "red")
curve(f(x), from = -2, to = 2, col = "blue")
curve(e(x), add = T)
curve(s(x), add = T, col = "red")
abline(v = -1, lty = 3, col = "red")
abline(v = 1, lty = 3, col = "red")
```



## Finding Inverse CDF $G^{-1}$

 $\int_{-\infty}^{\infty} e(x) \, dx = \int_{-\infty}^{-\frac{2}{3}} e^{\frac{2}{3}+x} \, dx + \int_{-\frac{2}{3}}^{\frac{2}{3}} e^{0} \, dx + \int_{\frac{2}{3}}^{\infty} e^{\frac{2}{3}-x} \, dx = \frac{10}{3}$ So the normalizing constant is  $\frac{3}{10}$ 

Then the CDF of g is

$$G(x) = \begin{cases} \frac{3}{10}e^{\frac{2}{3}+x} & \text{if } x < -\frac{2}{3}\\ \frac{3}{10}x + \frac{1}{2} & \text{if } -\frac{2}{3} \le x \le \frac{2}{3}\\ 1 - \frac{3}{10}e^{\frac{2}{3}-x} & \text{if } x > \frac{2}{3} \end{cases}$$

Take the inverse of G(x),

$$G^{-1}(u) = \begin{cases} log(\frac{10}{3}u) - \frac{2}{3} & \text{if } 0 < u < \frac{3}{10} \\ \frac{10}{3}(u - \frac{1}{2}) & -\frac{3}{10} \le u \le \frac{7}{10} \\ \frac{2}{3} - log(\frac{10}{3}(1 - u)) & \frac{7}{10} < u < 1 \end{cases}$$

Ginv <- function(u) {
 ifelse(u<3/10, log(u\*10/3)-2/3, ifelse(u>7/10, 2/3-log((1-u)\*10/3),(u-1/2)\*10/3));
}

#### Adaptive Rejection Sampling

Below is a function for performing rejection sampling with a sample size of *n* points.

```
# adaptive rejection sampling function
ars <- function(n) {</pre>
  x <- rep(NA, n);</pre>
 # number of points accepted
  ct <- 0;
  # number of points sampled
  total <- 0;
  # number of points caught by squeeze
  squeeze <-0;
  while(ct < n) {</pre>
    y <- Ginv(runif(1));</pre>
    u <- runif(1);
    # check squeeze range
    if(y > -1 \& k y < 1) {
      # under squeeze
      if(u < s(y)/e(y)) {
          ct <- ct + 1;
          x[ct] <- y;
           squeeze <- squeeze + 1;</pre>
      }
      # above squeeze
      else {
        # under f
        if(u < f(y)/e(y)) {
          ct <- ct + 1;
          x[ct]<-y;
        }
      }
    }
    # outside squeeze but under f
    else if(u < f(y)/e(y)) {
      ct <- ct + 1;
      x[ct] <- y;
    }
    total <- total + 1;</pre>
  }
  list(x = x, acratio_sx = squeeze/total, acratio = ct/total);
}
```

Choose a sample size of 100,000. Below are a few points drawn using this method.

samp\_size = 100000
set.seed(920)
ars\_points <- ars(samp\_size)
head(ars\_points\$x)</pre>

**##** [1] 0.2950836 1.3434853 1.1435056 -1.3004348 -0.3099414 0.1288534

The theoretical evelope ratio is  $\frac{\int_{-\infty}^{\infty} f(x) dx}{\int_{-\infty}^{\infty} e(x) dx}$ , the proportion of points in f that are in e.

integrate(f, lower = -Inf, upper = Inf)\$value / integrate(e, lower = -Inf, upper = Inf)
\$value

## [1] 0.7727395

For this simulation, the envelope ratio is

ars\_points\$acratio

## [1] 0.7754643

```
The theoretical squeeze ratio is \frac{\int_{-1}^{1} s(x) dx}{\int_{-\infty}^{\infty} e(x) dx}, the proportion of points in s that are in e.
```

```
integrate(s, lower = -1, upper = 1)$value / integrate(e, lower = -Inf, upper = Inf)$valu
e
```

```
## [1] 0.5102436
```

For this simulation, the squeeze ratio is

ars\_points\$acratio\_sx

## [1] 0.512706

### Calculate $E[x^2]$

Now that we have our points from the sample, square each accepted x, and take the mean to get  $E[x^2]$ .

mean(ars\_points\$x^2)

## [1] 0.7762001