# Statistical Computing Homework 6, Chapter 5

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19 October, 2018

#### Contents

Rejection Sampling	1
Mixture Proposal	3

#### Rejection Sampling

(a) It's obvious that first component is  $Gamma(\theta, 1)$ , second component is  $Gamma(\theta + 0.5, 1)$ . In order to make sure the integral of pdf is 1,  $C = (2\Gamma(\theta) + \Gamma(\theta + 0.5))^{-1}$ . Weights for two components are  $2\Gamma(\theta)C$  and  $\Gamma(\theta + 0.5)C$ 

```
(b)
```

```
theta <- 3
C <- function(theta) {
    ( 2 * gamma(theta) + gamma(theta+0.5) )^(-1) }
cmp1 <- function(theta) {
    2 * gamma(theta) * C(theta) }

#cmp2 <- function(theta) {
    # gamma(theta+0.5) * C(theta)
#}

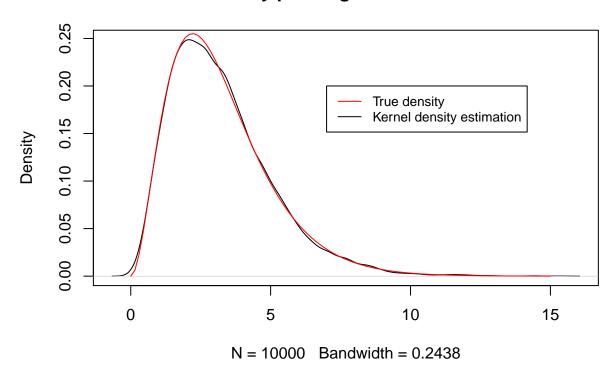
G <- function(x) {
    cmp1(theta) * dgamma(x, shape = theta, scale = 1) + (1-cmp1(theta)) * dgamma(x, shape = theta + 0.5, }
#curve(G, 0, 10)</pre>
```

#### Draw 10,000 samples from g: mixture gamma

```
p1 <- cmp1(theta)
#p2 <- cmp2(theta)
n <- 10000; mixture_g <- numeric(n)

for (i in 1:n) {
    u <- runif(1, min = 0, max = 1)
    if (u < p1) {
        mixture_g[i] <- rgamma(1, shape = theta, scale = 1)
    } else {
        mixture_g[i] <- rgamma(1, shape = theta + 0.5, scale = 1)
    }
}</pre>
```

### Density plot of g and estimation



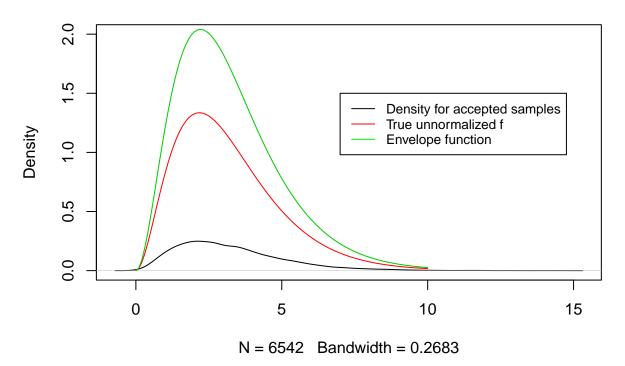
(c) Choose  $\alpha=8$ , this can be proved that envelope function is greater than unnormalized f on positve real line.

```
f <- function(x) {
    sqrt(4 + x) * x^(theta - 1) * exp(-x)
}
#curve(8*G(x), 0, 10, add = T, col = 2)

u <- runif(10000, min = 0, max = 1)
accept <- mixture_g[u < f(mixture_g)/(8*G(mixture_g))]

plot(density(accept), ylim = c(0, 2))
curve(f, 0, 10, add = T, col = 2)
curve(8*G(x), 0, 10, add = T, col = 3)
legend(7, 1.5, legend=c("Density for accepted samples", "True unnormalized f", "Envelope function"),
    col=c("black", "red", "green"), lty = 1, cex=0.8)</pre>
```

## density.default(x = accept)

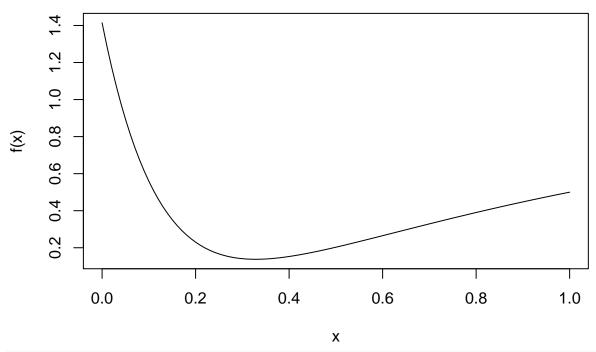


## Mixture Proposal

(a) Mixture beta as envelope function: set  $\theta = 3$ ,  $\beta = 10$ 

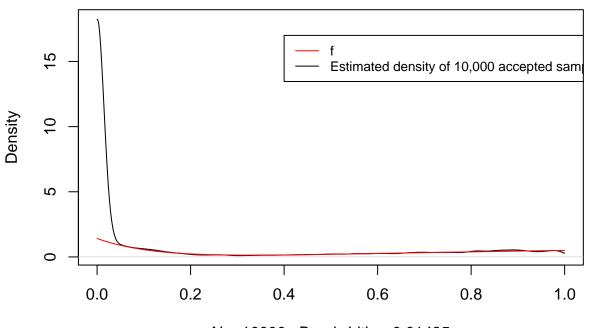
```
theta <- 3
bet <- 10

f <- function(x) {
    x^(theta - 1)/(1+x^2) + sqrt(x^2+2)*(1-x)^(bet-1)
}
curve(f, 0, 1)</pre>
```



```
### mixture beta with component probability 0.5 and 0.5
mixture_beta_fun <- function(x) {</pre>
  0.5*dbeta(x, shape1 = theta, shape2 = 1) + 0.5*dbeta(x, shape1 = 1, shape2 = bet)
}
accepted_f_1 <- numeric(10000)</pre>
n <- 0
### Draw the sample based on mixture beta envelope:
while(n <= 10000) {
  # First sample from mixture beta with component probability 0.5 and 0.5
  u <- runif(1, min = 0, max = 1)
  if (u < 0.5) {
    mixture_b <- rbeta(1, shape1 = theta, shape2 = 1)
    mixture_b <- rbeta(1, shape1 = 1, shape2 = bet)</pre>
  u \leftarrow runif(1, min = 0, max = 1)
  if (u <= f(mixture_b)/mixture_beta_fun(mixture_b)) {accepted_f_1[n] <- mixture_b}
  n \leftarrow n + 1
}
plot(density(accepted_f_1, from = 0, to = 1), main = "Sampling using mixture of beta")
curve(f, 0, 1, col = 2, add = T)
legend(0.4, 17, legend=c("f", "Estimated density of 10,000 accepted samples"),
       col=c("red", "black"), lty = 1, cex=0.8)
```

### Sampling using mixture of beta

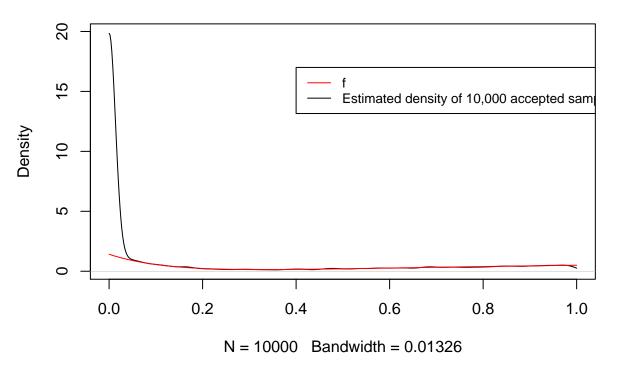


N = 10000 Bandwidth = 0.01435

(b)

```
\#curve(x^{(theta-1)/(1+x^2)}, 0, 1)
\#curve(theta*x^(theta-1)*0.6, 0, 1, add = T, col = 2)
\#curve(sqrt(2+x^2)*(1-x)^(bet-1), 0, 1)
\#curve(bet*(1-x)^(bet-1)*0.4, 0, 1, add = T, col = 2)
alpha1 <- 0.6
alpha2 <- 0.4
q1 <- function(x) {
  x^{(theta-1)/(1+x^2)}
q2 <- function(x) {
  sqrt(2+x^2)*(1-x)^(bet-1)
accepted_f_2 <- numeric(10000)</pre>
n <- 0
while (n \le 10000) {
  k \leftarrow sample(c(1, 2), size = 1, prob = c(alpha1, alpha2))
  u \leftarrow runif(1, min = 0, max = 1)
  if (k == 1) {
    propose <- rbeta(1, shape1 = theta, shape2 = 1)</pre>
    if (u <= q1(propose)/(alpha1*dbeta(propose, shape1 = theta, shape2 = 1))) {accepted_f_2[n] <- propo
  } else {
```

## Sampling using two components



 $\alpha_1$  should be greater than  $\theta^{-1}=0.333$ ,  $\alpha_2$ should be greater than  $\sqrt{3}\beta^{-1}=0.173$ , because  $p_1,\ p_2$  proportional to  $\alpha_1$  and  $\alpha_2$ , so we choose  $\alpha_1=0.6$ ,  $\alpha_2=0.4$