

# Homework 6

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## 5.2.1) Rejection Sampling

Let  $f$  and  $g$  be two probability densities on  $(0, \infty)$  such that

$$f(x) \propto \sqrt{4+x} x^{\theta-1} e^{-x}$$

$$g(x) \propto (2x^{\theta-1} + x^{\theta-\frac{1}{2}}) e^{-x}$$

Find the value of normalizing constant,  $C$ , such that

$$C \int_0^{\infty} (2x^{\theta-1} + x^{\theta-\frac{1}{2}}) e^{-x} dx = 1$$

.

First we show  $g$  is a mixture of gamma distributions, by separating  $g$  we get:

$$\begin{aligned} 2 \int_0^{\infty} (x^{\theta-1})(e^{-x}) dx + \int_0^{\infty} (x^{\theta-\frac{1}{2}})(e^{-x}) dx \\ = 2\Gamma(\theta) + \Gamma(\theta + 1/2) \end{aligned}$$

Substituting these terms into our equation

$$\begin{aligned} \Rightarrow C[2\Gamma(\theta) + \Gamma(\theta + 1/2)] &= 1 \\ \Rightarrow C &= \frac{1}{2\Gamma(\theta) + \Gamma(\theta + 1/2)} \end{aligned}$$

Therefore

$$g(x) = \frac{1}{2\Gamma(\theta) + \Gamma(\theta + 1/2)} (2x^{\theta-1})(e^{-x}) + \frac{1}{2\Gamma(\theta) + \Gamma(\theta + 1/2)} (x^{\theta-1/2})(e^{-x})$$

By Multiplying each term by  $\frac{\Gamma(\theta)}{\Gamma(\theta)}$  and  $\frac{\Gamma(\theta+1/2)}{\Gamma(\theta+1/2)}$  respectively, we get the mixture of distributions:

$$g(x) = \frac{\Gamma(\theta)}{2\Gamma(\theta) + \Gamma(\theta + 1/2)} (2x^{\theta-1})(e^{-x}) \frac{1}{\Gamma(\theta)} + \frac{\Gamma(\theta + 1/2)}{2\Gamma(\theta) + \Gamma(\theta + 1/2)} (x^{\theta-1/2})(e^{-x}) \frac{1}{\Gamma(\theta + 1/2)}$$

Therefore the component distributions are  $\Gamma(\theta, 1)$  and  $\Gamma(\theta + 1/2, 1)$  And corresponding weights:  $\frac{2\Gamma(\theta)}{2\Gamma(\theta) + \Gamma(\theta + 1/2)}$  and  $\frac{\Gamma(\theta + 1/2)}{2\Gamma(\theta) + \Gamma(\theta + 1/2)}$

To sample from this code, we need to sample from a uniform distribution, comparing whether  $U$  is less than the weight of each Gamma distribution component

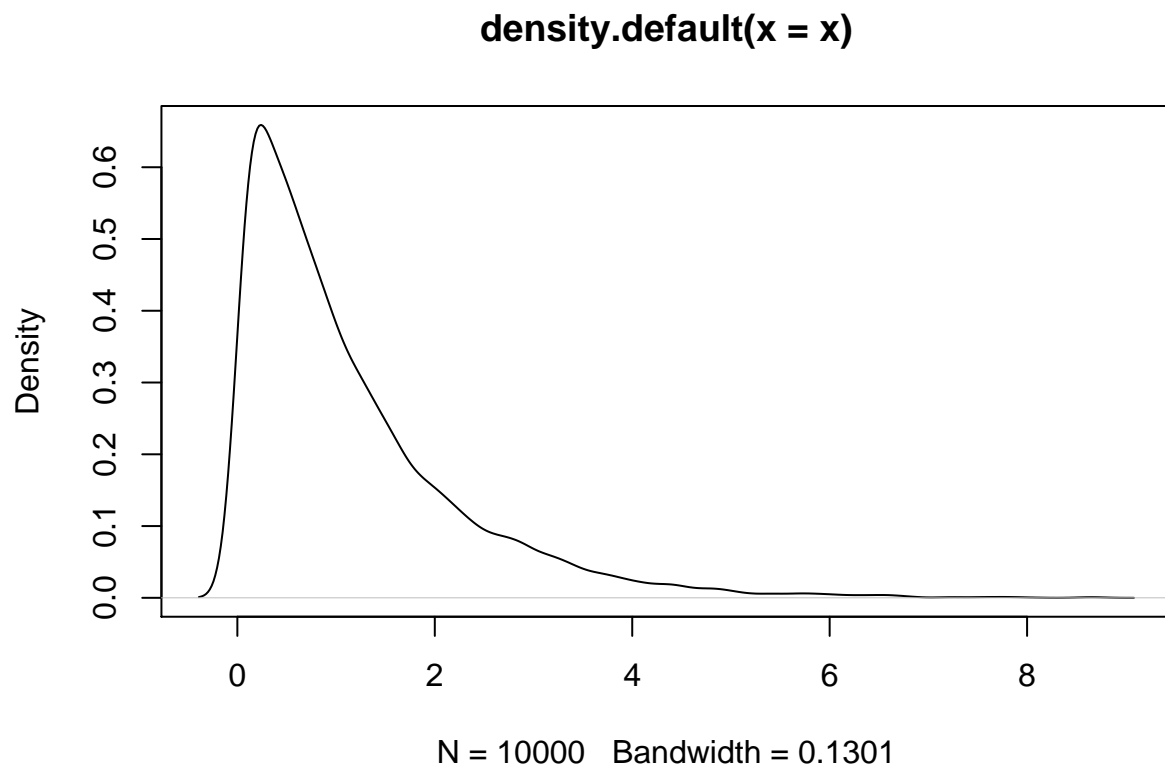
```
gamma.sample <- function(n){
  n <- 10000
  U <- runif(n)
  x <- rep(NA,n)
  theta <- 1 #arbitrary theta value
  C <- 1 / (2 * gamma(theta) + gamma(theta + 0.5)) # Normalizing Constant
  weight.g1 <- (2 * gamma(theta)) * C
  weight.g2 <- (gamma(theta + 0.5)) * C
```

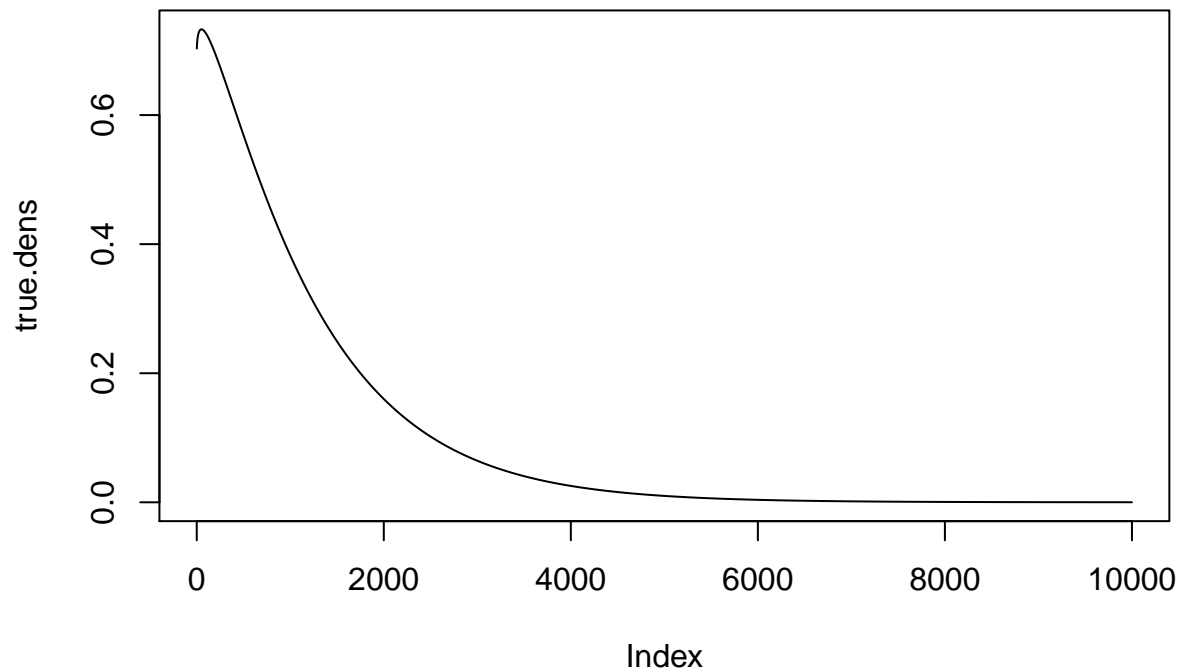
```

hh <- seq(.001, 10, .001)
true.dens <- weight.g1*dgamma(hh,1,1) + weight.g2*dgamma(hh,1.5,1)

for (i in 1:n){
  if (U[i] < weight.g1) {
    x[i] <- rgamma(1, theta, 1)
  }
  else {
    x[i] <- rgamma(1, theta + 0.5, 1)
  }
}
aa <- plot(density(x))
z <- plot(true.dens, type = "l")
lines(aa,z)
}
gamma.sample(10000)

```

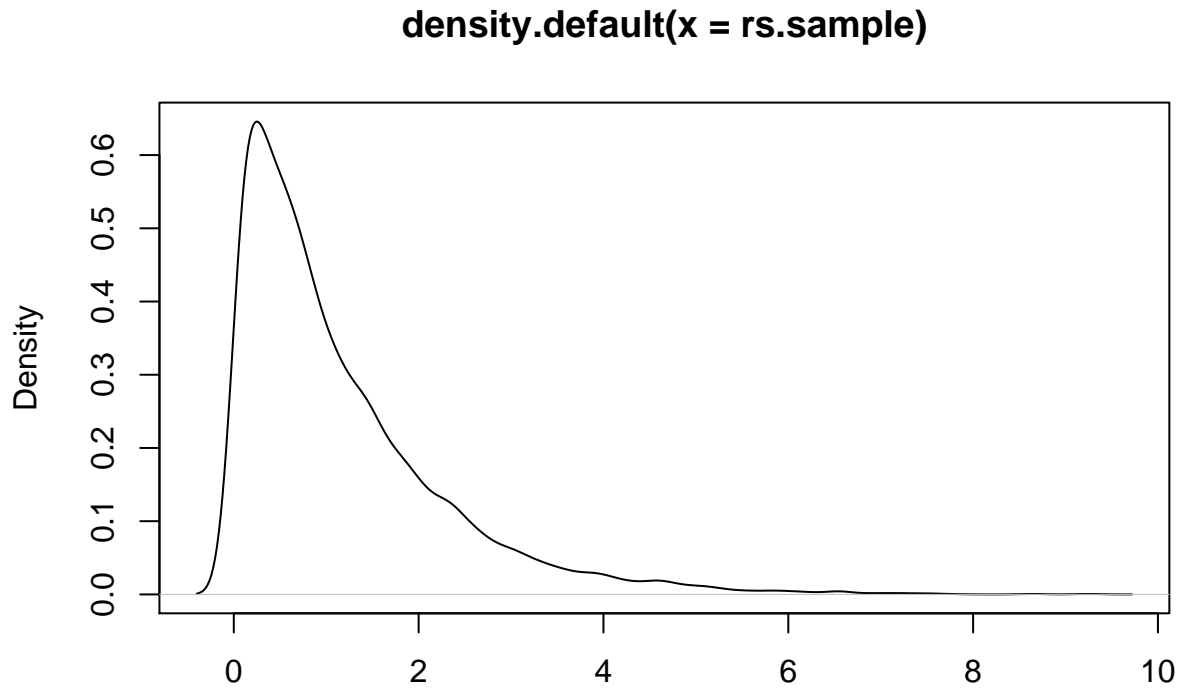




#### Rejection Sampling Method

```
n <- 10000
theta <- 1
rs.sample <- rep(NA, n)
C <- 1 / (2 * gamma(theta) + gamma(theta + 0.5))
weight.f1 <- C * (2 * gamma(theta))
alpha <- sqrt(2) / (2*C)

for (i in 1:n){
  U <- runif(1)
  if (U < weight.f1){
    x <- rgamma(1, theta, 1)
  }
  else{
    x <- rgamma(1, theta + 0.5, 1)
  }
  UU <- runif(1)
  f.x <- sqrt(4 + x) * x^(theta - 1) * exp(-x)
  g.x <- C * (2 * x ^ (theta - 1) + x^(theta - 0.5)) * exp(-x)
  z <- f.x / (alpha * g.x)
  if (U > z) {
    i <- i + 1; next
  }
  else rs.sample[i] <- x
}
plot(density(rs.sample))
```



N = 10000    Bandwidth = 0.1334

### 5.2.2) Mixture Proposal

Given probability density f:

$$f(x) \propto \frac{x^{\theta-1}}{1+x^2} + \sqrt{2+x^2}(1-x)^{\beta-1}$$

Sample from f as a mixture of beta distributions Let  $0 < x < 1$ , since

$$\frac{x^{\theta-1}}{1+x^2} \leq x^{\theta-1}$$

and

$$\sqrt{2+x^2}(1-x)^{\beta-1} \leq (\sqrt{2}+x)(1-x)^{\beta-1}$$

we can say

$$Cf(x) \leq q(x) = x^{\theta-1} + (\sqrt{3})(1-x)^{\beta-1}$$

where  $\sqrt{3}$  is derived from the max of  $\sqrt{2+x^2}$  given x's domain.

Since beta distributions are of the form,  $p(y|a, b) = \frac{y^{a-1}(1-y)^{b-1}}{B(a, b)}$ , we get distributions  $\text{beta}(\theta, 1)$  and  $\text{beta}(1, \beta)$  with weights  $w_1 = (1/\theta) * C$  and  $w_2 = C * (\sqrt{3}/\beta)$

```
beta.samp <- function(n, theta, beta0){
  bsamp <- rep(NA, n)
  w1 <- beta(theta, 1) / (beta(theta, 1) + sqrt(3)* beta(1, beta0))
  for (i in 1:n){
    U <- runif(1)
    if (U < w1){bsamp[i] <- rbeta(1, theta, 1)}
```

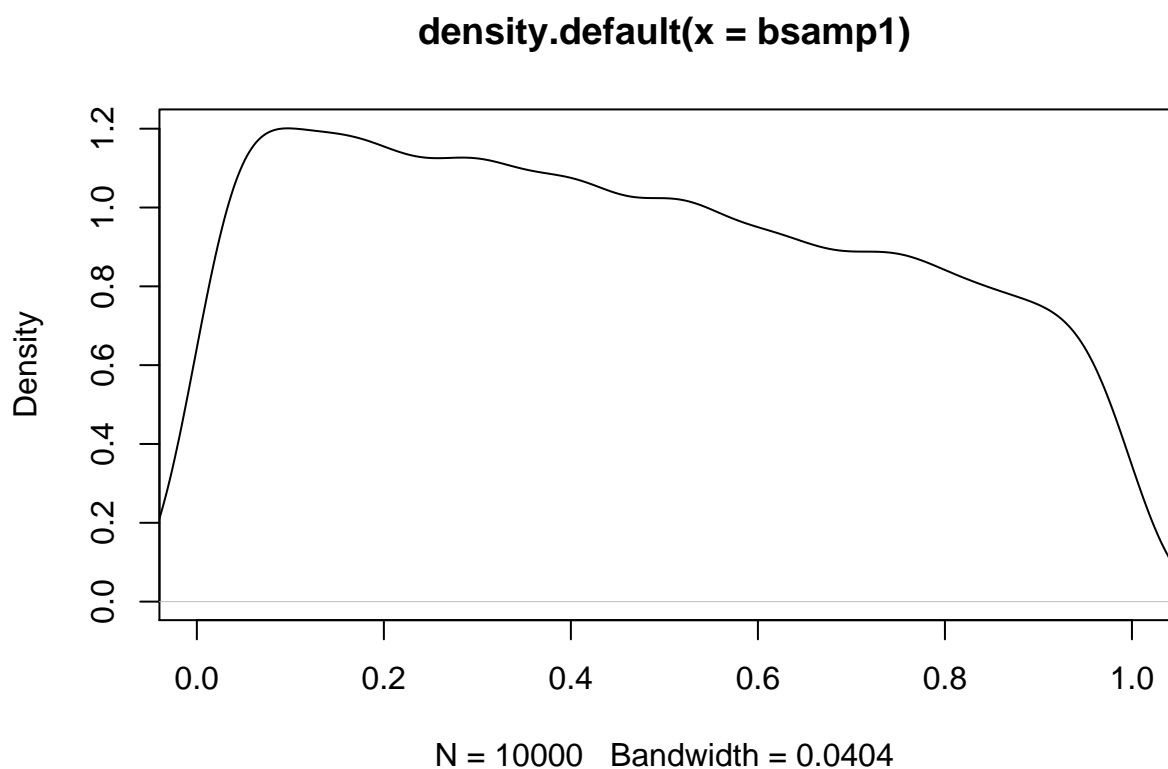
```

    else {bsamp[i] <- rbeta(1,1, beta0)}
  }
  return(list(samp = bsamp,w1= w1, w2 =1-w1))
}

n <- 10000
theta <- 2
beta0 <- 2
bsamp1 <- rep(0, n)
C <- 1 / (theta) + sqrt(3) / beta0
for (i in 1:n){
  U <- 10000
  upper <- 1
  lower <- 0.5
  while (U > C){
    U <- runif(1)
    b <- beta.samp(1, theta, beta0)
    x <- b$samp
    bsamp1[i] <- x
    upper <- (x^(theta - 1) / (1 + x^2) + sqrt(2+x^2)*(1-x)^(beta0-1))
    lower <- b$w1*dbeta(x, theta, 1) + b$w2*dbeta(x,1,beta0)
  }
}

gg <- function(x) {x ^ (theta-1) / (1+x^2) + sqrt(2+x^2) * (1-x)^(beta0-1)}
dens.beta <- integrate(gg, lower=0, upper=1)$value
plot(density(bsamp1), xlim=c(0,1))

```



Sampling Separately

```
beta.sep <- function(n, beta, theta){  
  b <- rep(NA, n)  
  r <- rep(NA, n)  
  U <- runif(n)  
  wt <- 1/theta + sqrt(3) / beta  
  while (U > wt)  
}
```