

# HW7 - Exercise7

JooChul Lee

26 October 2018

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## 1 Normal mixture revisited

Under the assumption,  $\delta$  is an uninformative prior and we use a simple random walk. The prior for  $\mu_1$  and  $\mu_2$  are  $N(0, 10^2)$ , that the prior for  $\sigma_1^2$  and  $\sigma_2^2$  are  $IV\Gamma(a, b)$  with shape  $a = .5$  and scale  $b = 10, \pi(\delta), \pi(\mu_1), \pi(\mu_2), \pi(\sigma_1^2)$ , and  $\pi(\sigma_2^2)$ , respectively.

Suppose that we consider finite mixture normal distribution,  $\mathbf{X}$

$$f(x) = \delta N(\mu_1, \sigma_1^2) + (1 - \delta) N(\mu_2, \sigma_2^2)$$

$$f(x, \delta, \mu_1, \mu_2, \sigma_1^2, \sigma_2^2) \propto (\delta \frac{1}{\sigma_1} \exp\{-\frac{(x-\mu_1)^2}{2\sigma_1^2}\} + (1 - \delta) \frac{1}{\sigma_2} \exp\{-\frac{(x-\mu_2)^2}{2\sigma_2^2}\}) \times \pi(\delta) \times \pi(\mu_1) \times \pi(\mu_2) \times \pi(\sigma_1^2) \times \pi(\sigma_2^2)$$

Then, MCMC using Gibbs sampling is the following;

Step 1 : Draw  $\delta^t$  from  $\pi(\delta^t | \mu_1^{t-1}, \mu_2^{t-1}, \sigma_1^{t-1}, \sigma_2^{t-1}, x^t)$

Step 2 : Draw  $\mu_1^t$  from  $\pi(\mu_1 | \delta^t, \mu_2^{t-1}, \sigma_1^{t-1}, \sigma_2^{t-1}, x^t)$

Step 3 : Draw  $\mu_2^t$  from  $\pi(\mu_2 | \delta^t, \mu_1^t, \sigma_1^{t-1}, \sigma_2^{t-1}, x^t)$

Step 4 : Draw  $\sigma_1^t$  from  $\pi(\sigma_1 | \delta^t, \mu_1^t, \mu_2^t, \sigma_2^{t-1}, x^t)$

Step 5 : Draw  $\sigma_2^t$  from  $\pi(\sigma_2 | \delta^t, \mu_1^t, \mu_2^t, \sigma_1^t, x^t)$

Repeat Step 1 - 5

```
library(MCMCpack)
```

```
## Loading required package: coda
```

```
## Loading required package: MASS
```

```
## ##
```

```
## ## Markov Chain Monte Carlo Package (MCMCpack)
```

```
## ## Copyright (C) 2003-2018 Andrew D. Martin, Kevin M. Quinn, and Jong Hee Park
```

```
## ##
```

```
## ## Support provided by the U.S. National Science Foundation
```

```
## ## (Grants SES-0350646 and SES-0350613)
## ##
```

```
library(HI)
delta <- 0.7
n <- 100
set.seed(254)
u <- rbinom(n, prob = delta, size = 1)
x <- rnorm(n, ifelse(u == 1, 7, 10), 0.5)

loglik = function(x,mu1,mu2,sigma1,sigma2,delta)
{
  Fisrt = log( ( delta * dnorm(x, mu1,sigma1) ) + ( (1-delta) * dnorm(x,mu2,sigma2) ) )
  return(sum(Fisrt))
}

HW7 = function(x,ini.D,ini.mu1,ini.mu2, ini.sigma1,ini.sigma2, iter)
{
  inix=x
  current.D <- ini.D; current.mu1 <- ini.mu1; current.mu2 <- ini.mu2; current.sigma1 <- ini.s
  current.sigma2 <- ini.sigma2
  new.D <- new.mu1 <- new.mu2 <- new.sigma1 <- new.sigma2 <- rep(0,iter)
  for(i in 1:iter)
  {
    D_v =function(delta) loglik(x =inix, mu1 = current.mu1, mu2 = current.mu2, sigma1 = curr
      sigma2 = current.sigma2, delta = delta)
    new.D[i] = arms(current.D, D_v, function(delta) {(delta>0)*(delta<1)}), 1)

    mu1.v = function(mu1) loglik(x =inix, mu1 = mu1, mu2 = current.mu2,
      sigma1 = current.sigma1, sigma2 = current.sigma2, delta = new.D[i]
      log( dnorm(mu1,0,10^2) )
    new.mu1[i] = arms(current.mu1, mu1.v, function(mu1) (mu1 > -30) * (mu1 < 30)), 1)

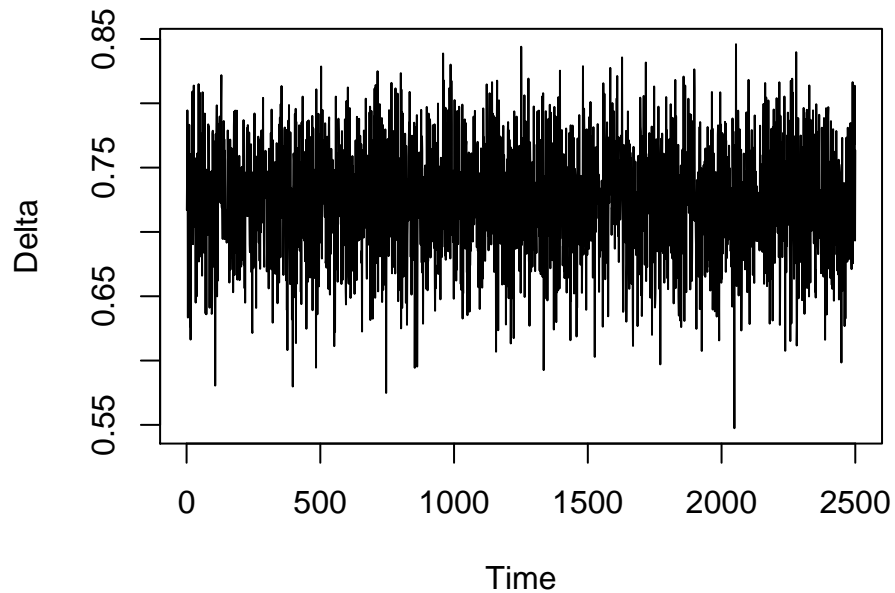
    mu2.v = function(mu2) loglik(x =inix, mu1 = new.mu1[i], mu2 = mu2,
      sigma1 = current.sigma1, sigma2 = current.sigma2, delta = new.D[i]
      log( dnorm(mu2,0,10^2) )
    new.mu2[i] = arms(current.mu2, mu2.v, function(mu2) (mu2 > -30) * (mu2 < 30)), 1)

    sigma1.v = function(sigma1) loglik(x =inix, mu1 = new.mu1[i], mu2 = new.mu2[i],
      sigma1 = sigma1, sigma2 = current.sigma2, delta = new.D[i]) +
      log(dinvgamma(sigma1, 0.5 , 0.1))
    new.sigma1[i] = arms(current.sigma1, sigma1.v, function(sigma1) (sigma1> 0) * (sigma1 <
    sigma2.v = function(sigma2) loglik(x =inix, mu1 = new.mu1[i], mu2 = new.mu2[i],
      sigma1 = new.sigma1[i], sigma2 = sigma2, delta = new.D[i])
      log(dinvgamma(sigma2, 0.5 , 0.1))
    new.sigma2[i] = arms(current.sigma2, sigma2.v, function(sigma2) (sigma2 > 0) * (sigma2 <
```

```

    current.D = new.D[i]; current.mu1 = new.mu1[i]; current.mu2 = new.mu2[i];
    current.sigma1 = new.sigma1[i]; current.sigma2 = new.sigma2[i]
  }
  list(new.D = new.D, new.mu1= new.mu1, new.mu2 = new.mu2, new.sigma1= new.sigma1, new.sigma2 =
}
joo = HW7(x,0.5,1,1,1,1, iter = 3000)
plot(ts(joo$new.D[-(1:500)]), ylab = 'Delta')

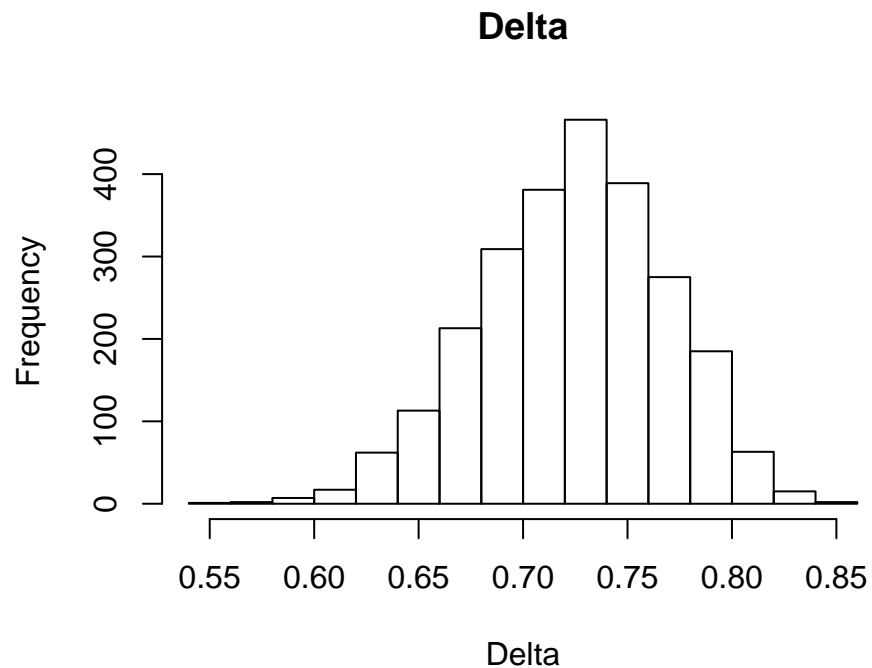
```



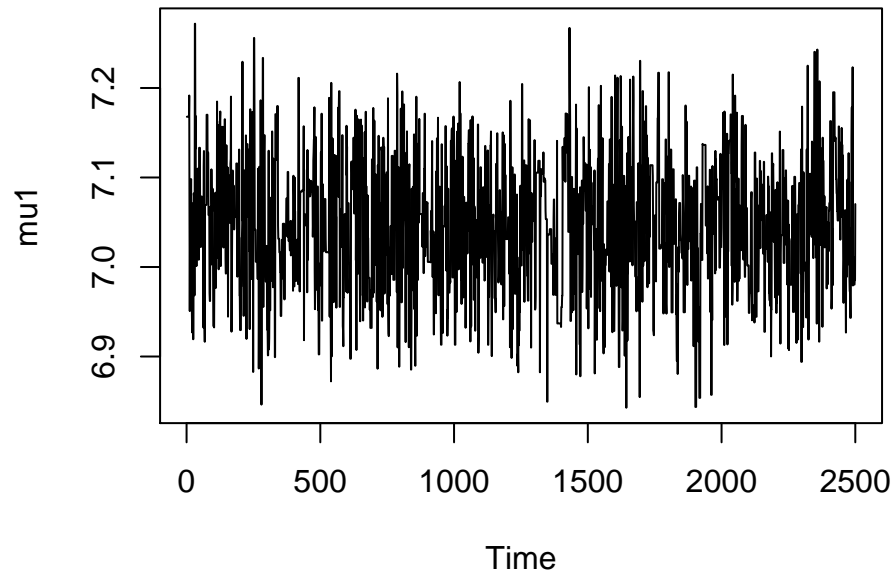
```

hist(ts(joo$new.D[-(1:500)]), main = 'Delta', xlab = 'Delta')

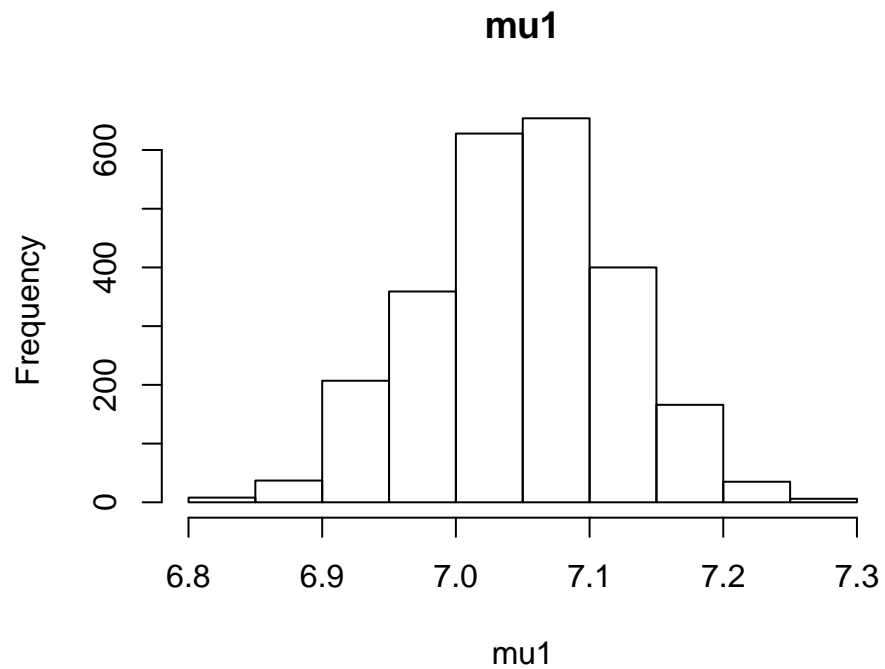
```



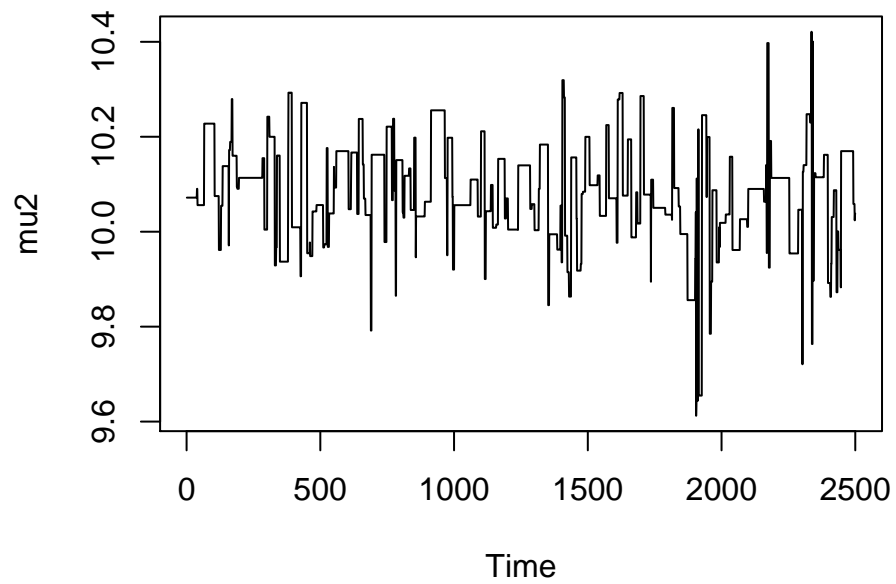
```
plot(ts(joo$new.mu1[-(1:500)]), ylab = 'mu1')
```



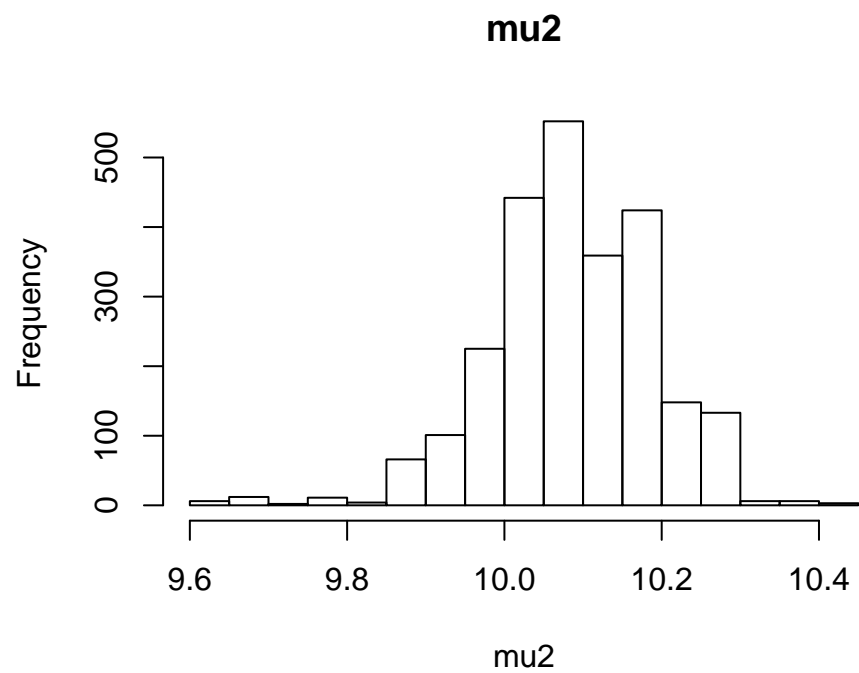
```
hist(ts(joo$new.mu1[-(1:500)]), main = 'mu1', xlab = 'mu1')
```



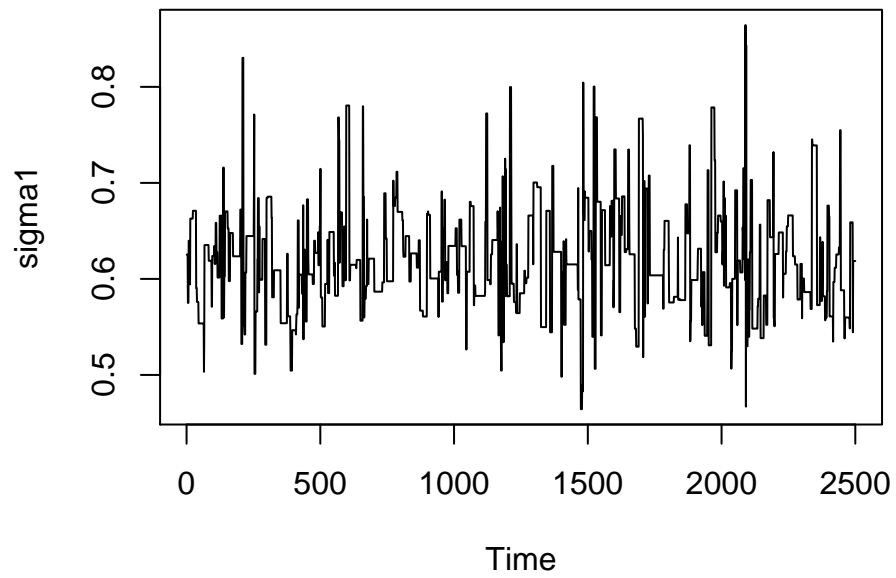
```
plot(ts(joo$new.mu2[-(1:500)]), ylab = 'mu2')
```



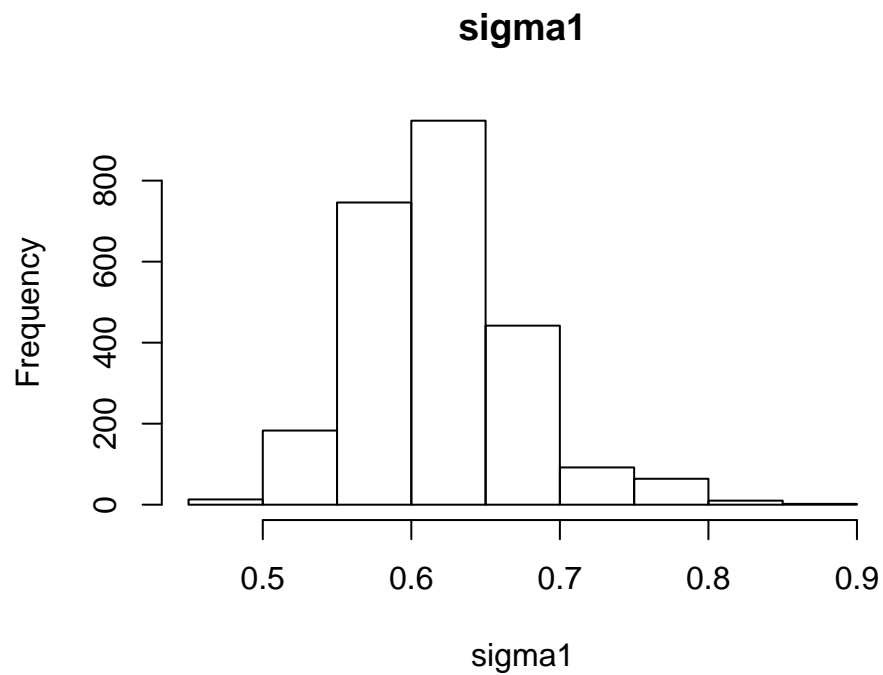
```
hist(ts(joo$new.mu2[-(1:500)]), main = 'mu2', xlab = 'mu2')
```



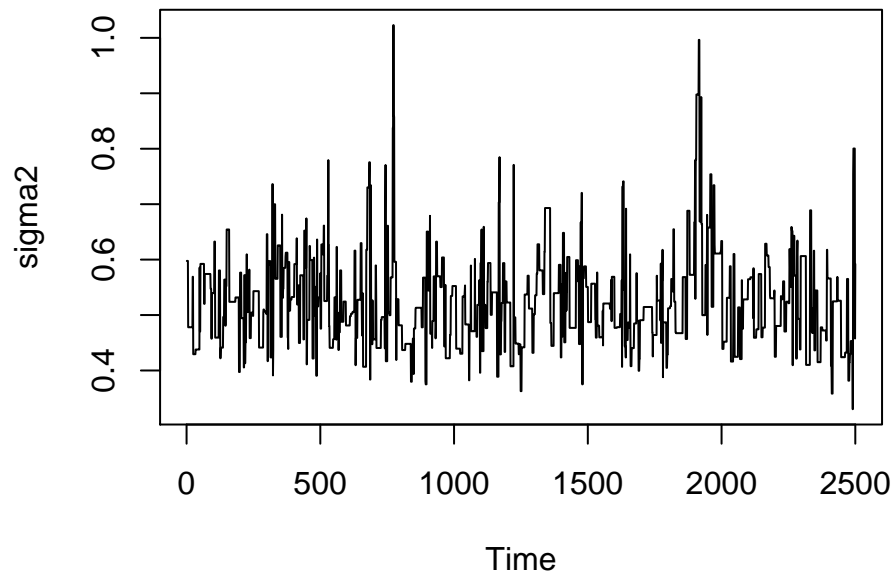
```
plot(ts(joo$new.sigma1[-(1:500)]), ylab = 'sigma1')
```



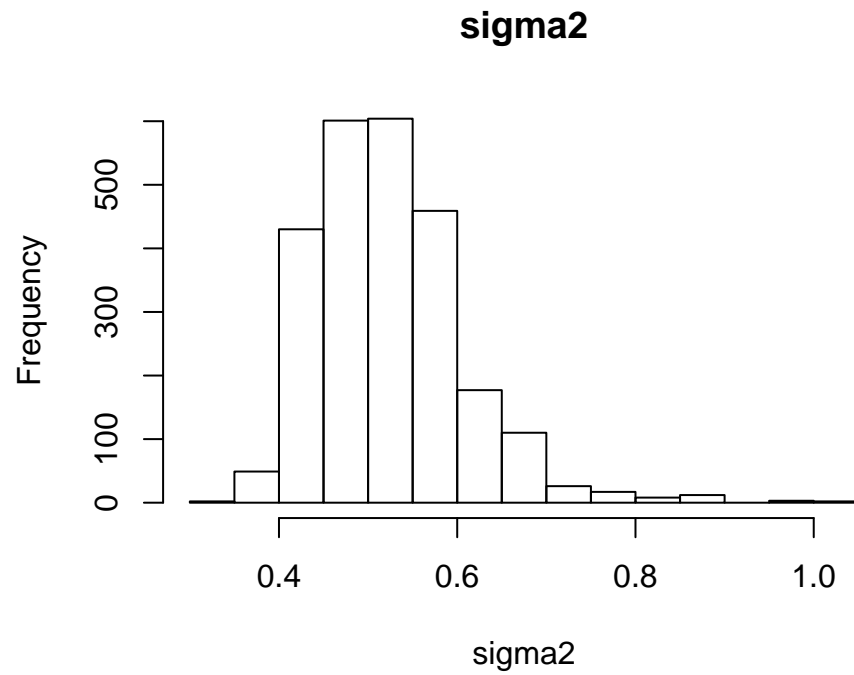
```
hist(ts(joo$new.sigma1[-(1:500)]), main = 'sigma1', xlab = 'sigma1')
```



```
plot(ts(joo$new.sigma2[-(1:500)]), ylab = 'sigma2')
```



```
hist(ts(joo$new.sigma2[-(1:500)]), main = 'sigma2', xlab = 'sigma2')
```



500 runs are dropped for the burn-in period.