Computational Statistics

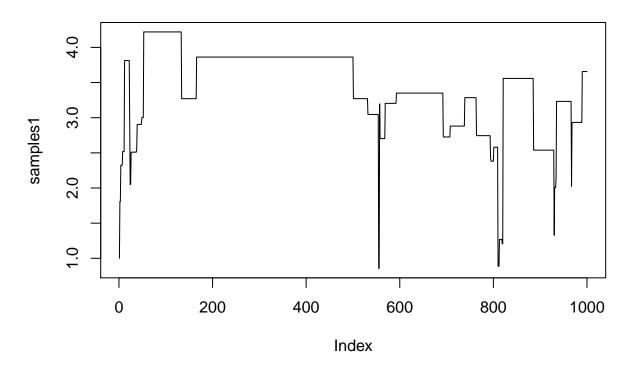
Lab 4

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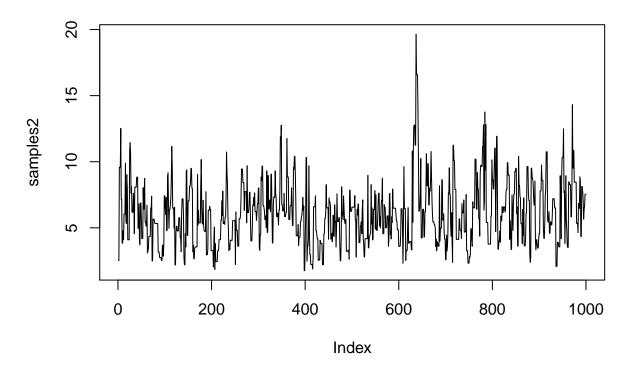
```
targetdensity <- function(x) {</pre>
    x^5 * exp(-x)
}
lognormalfuncs <- list(propsample=function(x) { rlnorm(1, meanlog=x, sdlog=1) },</pre>
                         propdensity=function(x, y) { dlnorm(x, meanlog=y, sdlog=1) },
                         targdensity=targetdensity)
chisquarefuncs <- list(propsample=function(x) { rchisq(1, df=floor(x + 1)) },</pre>
                         propdensity=function(x, y) { dchisq(x, df=floor(y + 1)) },
                         targdensity=targetdensity)
metropolis_hastings <- function(XO, iters, funcs) {</pre>
    x <- X0
    values <- rep(0, iters)</pre>
    alpha <- function(x, y) {</pre>
        numerator <- funcs$targdensity(y) * funcs$propdensity(x, y)</pre>
        denominator <- funcs$targdensity(x) * funcs$propdensity(y, x)</pre>
        numerator / denominator
    }
    for (i in 1:iters) {
        y <- funcs$propsample(x)</pre>
        u <- runif(1)
        if (u < alpha(x, y)) {
             x = y
        values[i] <- x</pre>
    }
    values
}
iters <- 1000
XO <- 1
actual <- rgamma(iters, shape=6, rate=1)</pre>
set.seed(123456)
samples1 <- metropolis_hastings(X0=X0, iters=iters, funcs=lognormalfuncs)</pre>
mean(samples1)
```

[1] 3.42

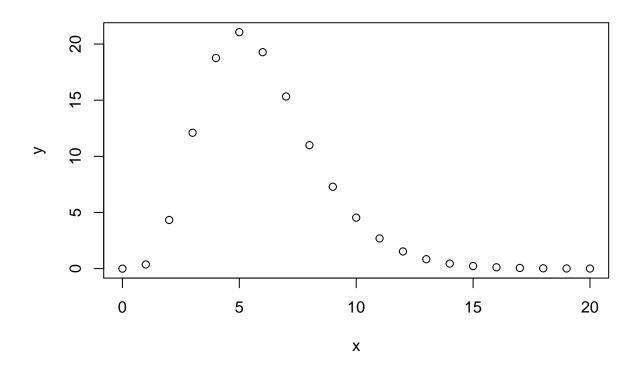
```
plot(samples1, type="l")
```



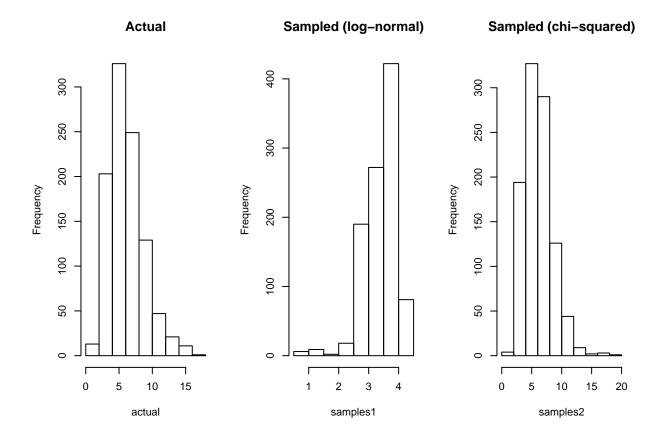
```
set.seed(123456)
samples2 <- metropolis_hastings(X0=X0, iters=iters, funcs=chisquarefuncs)
mean(samples2)
## [1] 6.05
plot(samples2, type="l")</pre>
```



```
x <- 0:20
y <- sapply(x, targetdensity)
plot(x, y)</pre>
```



```
oldpar <- par(mfrow = c(1, 3))
hist(actual, main="Actual")
hist(samples1, main="Sampled (log-normal)")
hist(samples2, main="Sampled (chi-squared)")</pre>
```



Geldman rubin

```
Geldman<-function(x){
    k <- nrow(x)
    n <- ncol(x)

B <- (n / (k - 1)) * sum((rowMeans(x) - mean(x))^2)

W <- sum((x - rowMeans(x))^2) / (k * (n - 1))

VarV <- ((n - 1) / n) * W + B / n

sqrtR <- sqrt(VarV / W)
    sqrtR
}

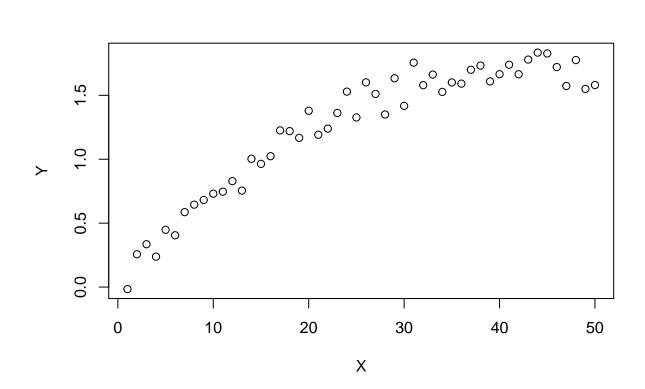
resultMatrix <- do.call(rbind, lapply(1:10, FUN = function(x)
    metropolis_hastings(XO = x, iters = iters, funcs = chisquarefuncs)))

GeldmanRes <- Geldman(resultMatrix)
GeldmanRes</pre>
```

[1] 1

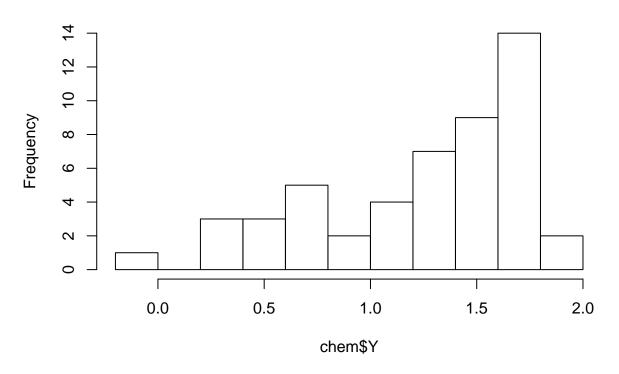
Question 2

```
load("../data/chemical.RData")
chem <- data.frame(X = X , Y = Y)
plot(chem)</pre>
```



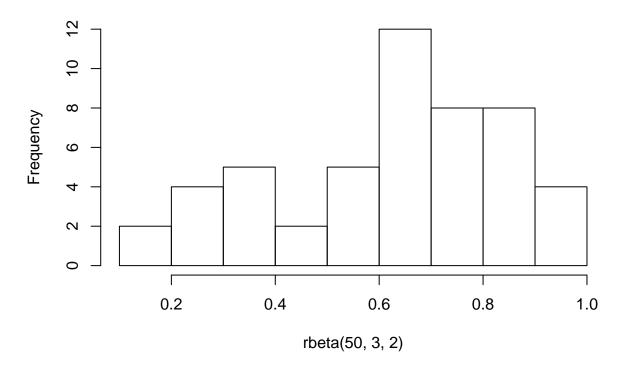
hist(chem\$Y)

Histogram of chem\$Y



hist(rbeta(50,3,2))

Histogram of rbeta(50, 3, 2)



$$p(\mu) = p(\mu_1)p(\mu_2|\mu_1)\cdots p(\mu_n|\mu_{n-1})$$

$$= \frac{1}{\sqrt{(2\pi\sigma^2)^{n-1}}} \exp\left(-\frac{1}{2\sigma^2} \sum_{i=2}^n (\mu_i - \mu_{i-1})^2\right)$$

$$p(y|\mu) = p(y_1|\mu_1)p(y_2|\mu_2)\cdots p(y_n|\mu_n)$$

$$= \frac{1}{\sqrt{(2\pi\sigma^2)^n}} \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mu_i)^2\right)$$

$$p(\mu|y) \propto p(y|\mu)p(\mu)$$

 $\propto \exp\left(-\frac{1}{2\sigma^2} \left(\sum_{i=1}^n (y_i - \mu_i)^2 + \sum_{i=2}^n (\mu_i - \mu_{i-1})^2\right)\right)$

This

$$p(\mu_1|\mu_{-1}, y) \propto \exp\left(-\frac{1}{\sigma^2}(\mu_1 - (y_1 + \mu_2)/2)^2\right)$$

$$p(\mu_i|\mu_{-i}, y) \propto p(\mu_1|\mu_{-1}, y) \exp\left(-\frac{3}{2\sigma^2} \sum_{j=2}^i (\mu_j - (y_j + \mu_{j-1} + \mu_{j+1})/3)^2\right)$$

$$p(\mu_n|\mu_{-n}, y) \propto p(\mu_{n-1}|\mu_{-n-1}, y) \exp\left(-\frac{1}{\sigma^2}(\mu_n - (y_n + \mu_{n-1})/2)^2\right)$$

Alternatively this

$$p(\mu_1|\mu_{-1}, y) \propto \exp\left(-\frac{1}{\sigma^2}(\mu_1 - (y_1 + \mu_2)/2)^2\right)$$
$$p(\mu_i|\mu_{-i}, y) \propto \exp\left(-\frac{3}{2\sigma^2}(\mu_i - (y_i + \mu_{i-1} + \mu_{i+1})/3)^2\right)$$
$$p(\mu_n|\mu_{-n}, y) \propto \exp\left(-\frac{1}{\sigma^2}(\mu_n - (y_n + \mu_{n-1})/2)^2\right)$$