Bayesian Learning

Lab 3

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Question 1

rainfall <- read.table("../data/rainfall.dat", header=F)</pre>

Prior

$$\mu \sim N(\mu_0, \tau_0^2)$$

$$\sigma^2 \sim \text{Inv} - \chi^2(\nu_0, \sigma_0^2)$$

Likelihood

$$\mathbf{y}|\mu,\sigma^2 \sim N(\mu,\sigma^2)$$

Posterior

$$\mu|\sigma^2, \mathbf{x} \sim N(\mu_n, \tau_n^2)$$

$$\sigma^2|\mu, \mathbf{x} \sim \text{Inv}-\chi^2\left(\nu_n, \frac{\nu_0 \sigma_0^2 + \sum_{i=1}^n (x_i - \mu)^2}{n + \nu_0}\right)$$

where

$$\mu_n = \frac{\frac{1}{\tau_0^2}}{\frac{1}{\tau_0^2} + \frac{n}{\sigma^2}} \mu_0 + \frac{\frac{n}{\sigma^2}}{\frac{1}{\tau_0^2} + \frac{n}{\sigma^2}} \bar{x}$$

$$\frac{1}{\tau_n^2} = \frac{1}{\tau_0^2} + \frac{n}{\sigma^2}$$

$$\nu_n = \nu_0 + n$$

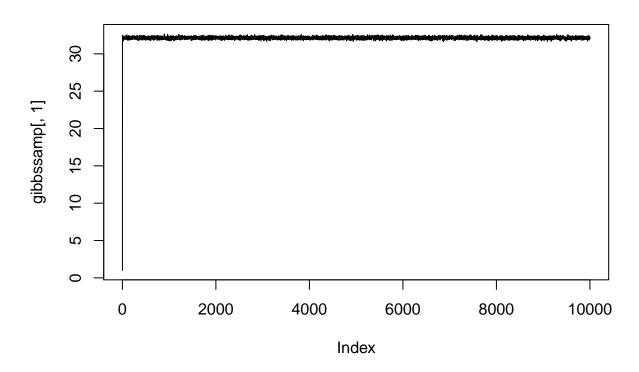
a)

```
library(geoR)

mun <- function(x, sigmasq, hyperparams){
    n <- length(x)
    denom <- ((1 / hyperparams$tausq0) + (n / sigmasq))</pre>
```

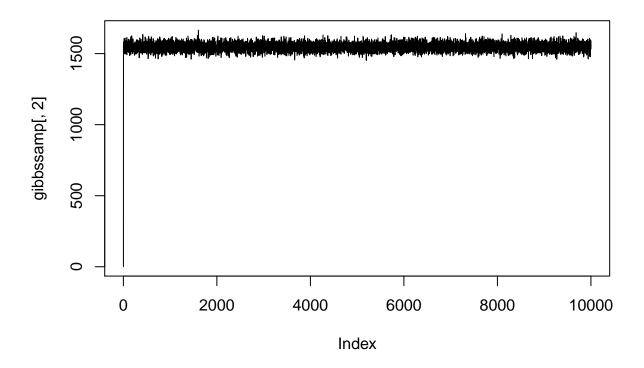
```
pt1 <- ((1 / hyperparams$tausq0) / denom) * hyperparams$mu0</pre>
    pt2 <- ((n / sigmasq) / denom) * mean(x)
    pt1 + pt2
}
taun <- function(x, sigmasq, hyperparams){</pre>
    hyperparams$tausq0 + (sigmasq / length(x))
    1/ hyperparams$tausq0
}
nun <- function(x, hyperparams){</pre>
    hyperparams$nu0 + length(x)
sigmasqn <- function(x, mu, hyperparams){</pre>
    (hyperparamsnu0 * hyperparams<math>sigmasq0 + sum((x - mu)^2)) / (length(x) + hyperparams<math>nu0)
}
musampler <- function(x,sigmasq, hyperparams){</pre>
    mu <- mun(x, sigmasq, hyperparams)</pre>
    sigma <- sqrt(taun(x,sigmasq, hyperparams))</pre>
    rnorm(1, mu, sigma)
}
sigmasampler <- function(x, mu, hyperparams){</pre>
    scale <- sigmasqn(x, mu, hyperparams)</pre>
    df <- nun(x, hyperparams)</pre>
    rinvchisq(1, df, scale)
}
Now we start start sampling
gibbs <- function(x, iter, init, hyperparams){</pre>
    samples <- matrix(NA, ncol = 2, nrow = iter + 1)</pre>
    samples[1,] <- init</pre>
    for (i in 2:(iter+1)){
        mu <- musampler(x, samples[i-1, 2], hyperparams)</pre>
         sigma <- sigmasampler(x, mu, hyperparams)</pre>
         samples[i,] <- c(mu, sigma)</pre>
    }
    samples
}
set.seed(123456)
hyperparams <- list(mu0=0, tausq0=50, nu0=1, sigmasq0=50)
gibbssamp <- gibbs(x=rainfall$V1, iter = 10000, init = c(1,1), hyperparams)
```

Mean



plot(gibbssamp[,2], type = "1", main="Variance")

Variance

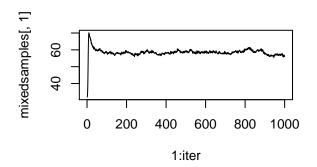


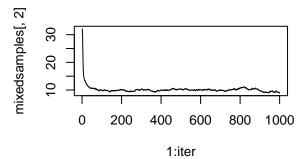
b)

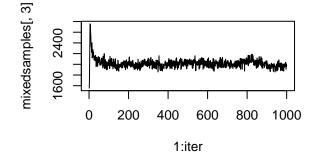
```
nn <- function(I){</pre>
    n1 \leftarrow sum(I == 1)
    n2 \leftarrow sum(I == 2)
    list(n1 = n1, n2 = n2)
}
dd <- function(x, I) {</pre>
    d1 \leftarrow x[I == 1]
    d2 <- x[I == 2]
    list(d1=d1, d2=d2)
}
pisampler <- function(x, I, hyperparams){</pre>
    n \leftarrow nn(I)
    rbeta(1, shape1 = n$n1 + hyperparams$a1 , shape2 = n$n2 + hyperparams$a2)
}
sigma2sampler <- function(x, I, hyperparams){</pre>
    n \leftarrow nn(I)
    d \leftarrow dd(x, I)
    vnsn1 <- hyperparams$nu0 * hyperparams$sigmasq0 +</pre>
```

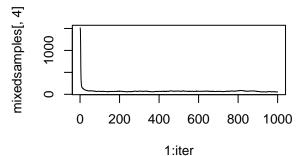
```
(n$n1 - 1) * var(d$d1) +
         (((1 / hyperparams$tausq0) * n$n1) / ((1 / hyperparams$tausq0) + n$n1)) *
         ((mean(d$d1) - hyperparams$mu0)^2)
    vnsn2 <- hyperparams$nu0 * hyperparams$sigmasq0 +</pre>
         (n$n2 - 1) * var(d$d2) +
         (((1 / hyperparams$tausq0) * n$n2) / ((1 / hyperparams$tausq0) + n$n2)) *
         ((mean(d$d2) - hyperparams$mu0)^2)
    vn1
          <- hyperparams$nu0 + n$n1
    vn2
          <- hyperparams$nu0 + n$n2
          <- vnsn1 / vn1
    sn1
         <- vnsn2 / vn2
    sigma1 <- rinvchisq(1, vn1, sn1)</pre>
    sigma2 <- rinvchisq(1, vn2, sn2)</pre>
    c(sigma1, sigma2)
}
mu2sampler <- function(x, I, sigmasq, hyperparams){</pre>
    d \leftarrow dd(x, I)
    tau1 <- taun(d$d1, sigmasq[1], hyperparams)</pre>
    mun1 <- mun(d$d1, sigmasq[1], hyperparams)</pre>
    tau2 <- taun(d$d2, sigmasq[2], hyperparams)</pre>
    mun2 <- mun(d$d2, sigmasq[2], hyperparams)</pre>
    mu1 <- rnorm(1, mun1, tau1)</pre>
    mu2 <- rnorm(1, mun2, tau2)</pre>
    c(mu1, mu2)
}
Isampler <- function(x, pi, mu, sigmasq){</pre>
    nom \leftarrow (1 - pi) * dnorm(x, mu[2], sqrt(sigmasq[2]))
    denom <- pi * dnorm(x, mu[1], sqrt(sigmasq[1])) + nom</pre>
    theta <- nom / denom
    rbinom(length(x), prob = theta, size = 1) + 1 # to get them into 1 and 2
}
ysampler <- function(pi, mu, sigmasq){</pre>
    pi * rnorm(1, mu[1], sqrt(sigmasq[1])) +
         (1 - pi) * rnorm(1, mu[2], sqrt(sigmasq[2]))
}
And now we sample.
mixedgibbs <- function(x, iter, init, hyperparams){</pre>
    samples <- matrix(NA, ncol = 4, nrow = iter)</pre>
    I <- init
```

```
for (i in 1:iter){
        pi <- pisampler(x, I, hyperparams)</pre>
        sigmasq <- sigma2sampler(x, I, hyperparams)</pre>
        mu <- mu2sampler(x, I, sigmasq, hyperparams)</pre>
        I <- Isampler(x, pi, mu, sigmasq)</pre>
        samples[i,] <- c(mu, sigmasq)</pre>
    }
    samples
}
set.seed(123456)
hyperparams <- list(a1=1, a2=1, mu0=1, tausq0=50, nu0=1, sigmasq0=50)
init <- sample(c(1, 2), size=length(rainfall$V1), replace=TRUE)</pre>
iter <- 1000
mixedsamples <- mixedgibbs(rainfall$V1, iter, init, hyperparams)</pre>
par(mfrow = c(2,2))
plot(x = 1:iter, mixedsamples[,1], type = "1")
plot(x = 1:iter, mixedsamples[,2], type = "1")
plot(x = 1:iter, mixedsamples[,3], type = "1")
plot(x = 1:iter, mixedsamples[,4], type = "1")
```









par(mfrow = c(1,1))

c)

Question 2

- **a**)
- b)
- **c**)