Lasso

John Tipton May 4, 2015

First, let's load the packages and some helper functions

```
##
## libraries and functions
##
library(statmod)
source('~/Linear-Model/dinvgamma.R')
source('~/Linear-Model/rMVN.R')

source("~/Linear-Model/bayesianLassoRegression/fixedEffectModel/mcmc.lm.lasso.R")

make.model.plot <- function(out){
    n.burn <- floor(n.mcmc / 5) + 1
    layout(matrix(1:9, 3))
    matplot(t(out$beta.save[, n.burn:n.mcmc]), type = 'l')
    hist(out$beta.save[2,][n.burn:n.mcmc], main = 'Posterior of Beta2')
    abline(v = beta[2], col = 'red')
    plot(out$sigma.squared.epsilon.save[n.burn:n.mcmc], type = 'l')
    abline(h = sigma.squared.epsilon, col = 'red')
    plot(out$lambda.squared.save, type = 'l')
}</pre>
```

Then we simulate some data

```
##
## Simulate some data
##

N <- 1000
n <- 100
beta <- -3:3
sigma.squared.epsilon <- 0.25
tau <- length(beta)

make.lm.data <- function(N, n, beta, sigma.sqaured.epsilon){
   tau <- length(beta)
   X <- matrix(nrow = N, ncol = tau)
   for(i in 1:tau){
        X[, i] <- rnorm(N, 0, 1)
   }
        Y <- X %*% beta + rnorm(N, 0, sigma.squared.epsilon)</pre>
```

```
data.frame(Y, X)
}
data <- make.lm.data(N, n, beta, sigma.squared.epsilon)</pre>
```

Subsample the data

```
samp <- sample(1:N, n)
data.samp <- data[samp, ]</pre>
```

Examine a linear regression model

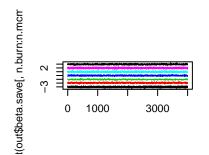
```
lm(Y \sim . ,data = data)
##
## Call:
## lm(formula = Y ~ ., data = data)
##
## Coefficients:
## (Intercept)
                        X1
                                    X2
                                                 ХЗ
                                                              Х4
    -0.009582 -2.998608
                              -1.996605
                                           -1.001350
                                                        0.001352
##
           X5
                        Х6
                                    Х7
     1.002206 2.004336
                               2.999200
```

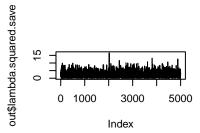
Specify priors for a Bayesian model

```
##
## Setup priors
##
# hyperparameters for mu.beta and sigma.squared.beta
alpha.epsilon <- 1
beta.epsilon <- 1
alpha.lambda <- 1
beta.lambda <- 1
n.mcmc < -5000
##
## Fit mcmc
##
Y <- data.samp[, 1]
X <- as.matrix(data.samp[, 2:(tau + 1)], ncol = tau)</pre>
out <- mcmc.lm.lasso(Y, X, n.mcmc, alpha.epsilon, beta.epsilon, alpha.lambda, beta.lambda)
     100 200 300 400 500 600 700 800 900 1000 1100 1200 1300 1400 1500 1600 1700 1800
##
```

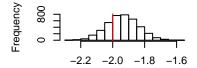
Examine model output

make.model.plot(out)

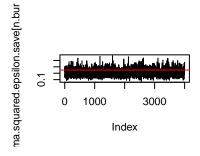




Posterior of Beta2



out\$beta.save[2,][n.burn:n.mcmc]



Examine estimates $\hat{\beta}$

##

```
library(pander)
results=data.frame(rbind(c(beta), c(rowMeans(out$beta))), row.names=c("Truth", "Estimate"))
names(results)=c("Beta1", "Beta2", "Beta3", "Beta4", "Beta5", "Beta6", "Beta7")
pandoc.table(results, style="rmarkdown")
```

```
##
##
##
         
                     Beta1 | Beta2 |
                                         Beta3 | Beta4
      **Truth**
                      -3
                                -2
                                          -1
                                                    0
##
                  | -3.002 | -1.939 | -1.04 | -0.01437 | 0.9778
##
##
## Table: Table continues below
##
##
```

##		-	Beta6		Beta7		
## ::							
##	**Truth**	-	2		3		
##	**Estimate**	-	1.971		2.967		