

Regression with Slice Sampling

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```
library(bisemSliceSampler)
library(latex2exp)
library(patchwork)
library(ggplot2)

data(BostonHousing, package = "mlbench")
keep_cols <- c("crim", "lstat", "age", "medv")
housing_sub <- subset(BostonHousing, select = keep_cols)

#summary(BostonHousing)

# Add column with all 1s for intercept
X <- cbind("intercept" = 1, as.matrix(subset(housing_sub, select = -medv)))
y <- housing_sub$medv
```

Model:

$y_{\text{medv},i} = \beta_0 + \beta_1 x_{\text{crim},i} + \beta_2 x_{\text{lstat},i} + \beta_3 x_{\text{age},i} + \beta_4 x_{\text{medv},i} + \epsilon_i, \quad i = 1, \dots, n$
with $\epsilon_i \stackrel{\text{iid}}{\sim} \text{Normal}(0, \sigma^2)$

$$\implies y_i \mid \beta, \sigma^2 \stackrel{\text{i.i.d}}{\sim} \text{N}(\mathbf{x}_i^T \beta, \sigma^2)$$

Likelihood:

$$L(\mu, \sigma^2) \propto (\sigma^2)^{-n/2} \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mathbf{x}_i^T \beta)^2\right)$$

Prior

In this example we use the reference prior for multiple linear regression:

$$p(\beta, \sigma^2) \propto \frac{1}{\sigma^2}$$

Posterior

$$p(\beta, \sigma^2 \mid \mathbf{y}) \propto (\sigma^2)^{-n/2} \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mathbf{x}_i^T \beta)^2\right) * \frac{1}{\sigma^2}$$

Log-Posterior

$$\log(p(\beta, \sigma^2 \mid \mathbf{y})) \propto -\frac{n+2}{2} \log(\sigma^2) - \frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mathbf{x}_i^T \beta)^2$$

```
logpost <- function(theta) {
  beta <- theta[-length(theta)]
  sigma_sq <- theta[length(theta)]
  n <- length(y)
  -((n + 2)/2 * log(sigma_sq)) - (1/(2 * sigma_sq)) * crossprod(y - (X %*% beta))
}
```

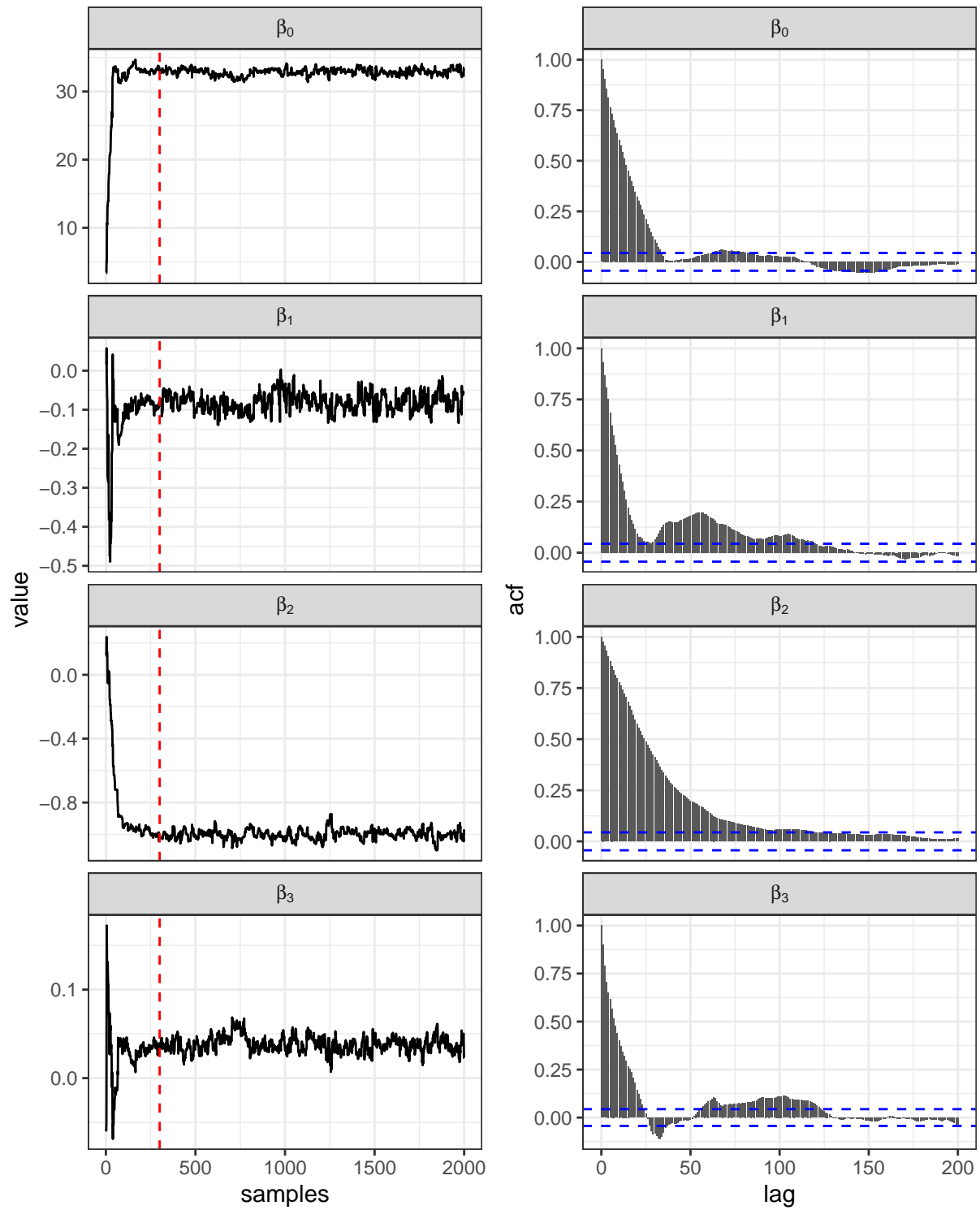
Note: $\theta = (\beta, \sigma^2) = (\beta_0, \beta_1, \beta_2, \beta_3, \sigma^2)$

Sampling from Log-Posterior

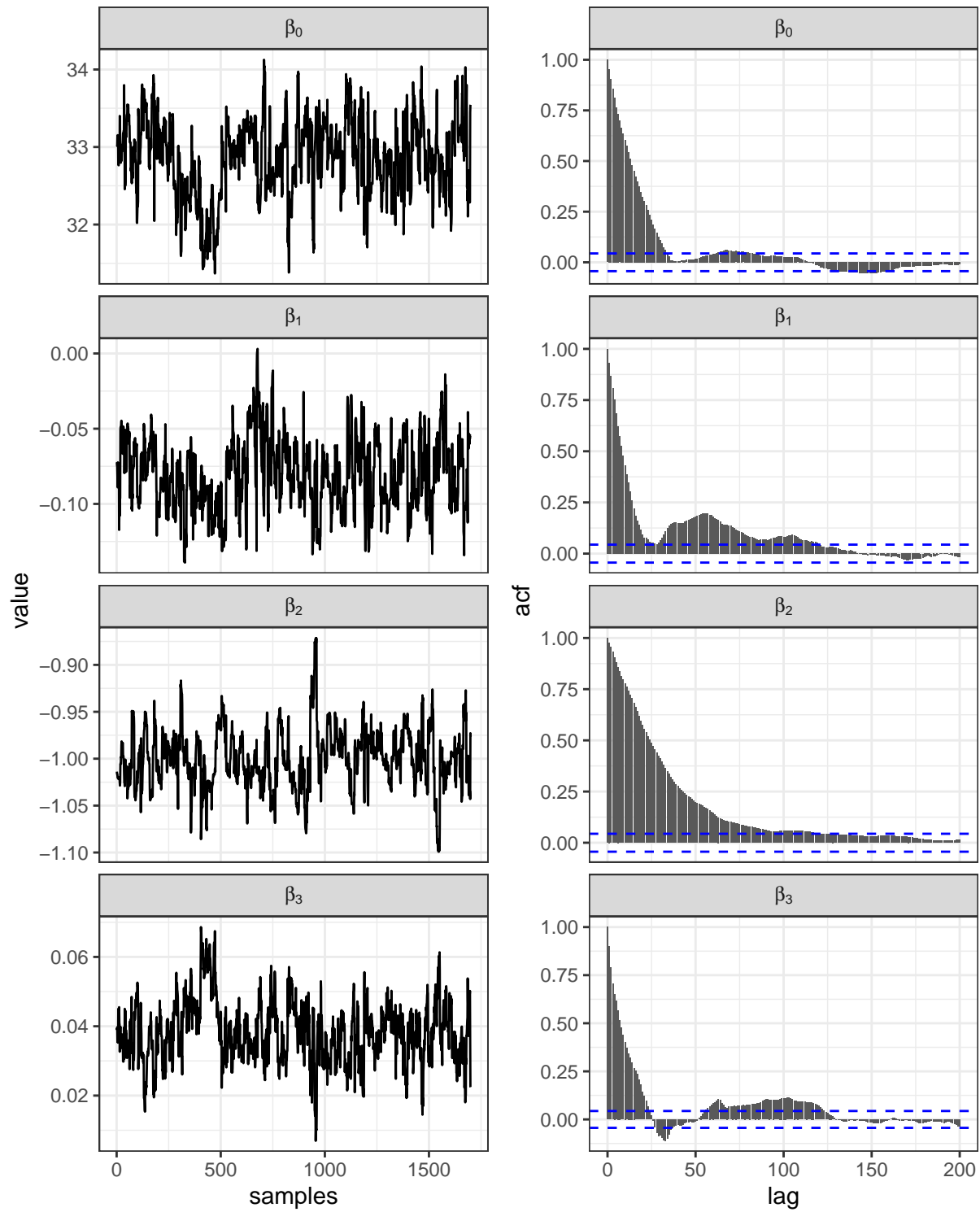
```
w_intercept <- 10
w_rest <- 0.4
w <- c(w_intercept, rep(w_rest, times = ncol(X)))
n_samples <- 2000
theta_init <- c(rep(0, ncol(X)), 10)
```

```
post_samples <- slice_sampler(
  logpost, x_init = theta_init, w = w, n_samples = n_samples
)
```

Convergence Diagnostic



Convergence Diagnostic - 300 burn-in samples removed



Coefficients and Credibility Intervals

	Mean	2.5%	97.5%
intercept	32.858	31.727	33.787
crim	-0.080	-0.124	-0.033
lstat	-1.000	-1.059	-0.939
age	0.038	0.021	0.058

Comparison with frequentist calculation

Expected to be similar as non-informative prior was chosen.

	Mean	2.5%	97.5%
intercept	32.828	31.359	34.297
crim	-0.083	-0.153	-0.012
lstat	-0.994	-1.094	-0.894
age	0.038	0.014	0.062