

STA3105-01 Bayesian Statistics
Homework 6
DUE Friday, December 2

Copying homework solutions from others lead to a 0 score. No late submission is allowed. Your solution should contain both code and a corresponding explanation for the answer. Submit your HW through LearnUs. **You should submit (1) a report file (pdf) and (2) relevant code files.** We will implement the MCMC algorithm for a Bayesian regression as follows:

$$\begin{aligned}\mathbf{Y} &\sim \text{Bernoulli}(\mu(\mathbf{X})) \\ \log\left(\frac{\mu(\mathbf{X})}{1 - \mu(\mathbf{X})}\right) &= \mathbf{X}\beta \\ \beta_j &\stackrel{\text{ind}}{\sim} \lambda_j N(0, \sigma_1^2) + (1 - \lambda_j) N(0, \sigma_2^2) \text{ for } j = 1, \dots, 10 \\ \lambda_j &\stackrel{\text{ind}}{\sim} \text{Bernoulli}(1/2) \\ \sigma_1^2 &\sim \text{IG}(1, 20), \quad \sigma_2^2 \sim \text{G}(1, 20)\end{aligned}$$

The notations follow our lecture notes.

1. (10 points) Simulate the dataset as follows.
 - (a) Let $\mathbf{X}_i \in \mathbb{R}^{10}$ be the predictors for i th observation. For $i = 1, \dots, 10,000$, simulate $\mathbf{X}_i \sim N(0, \mathbf{I})$ independently. Here \mathbf{I} is an identity matrix.
 - (b) For $i = 1, \dots, 10,000$, simulate $Y_i \sim \text{Bernoulli}(\mu(\mathbf{X}_i))$ independently. Set the true regression coefficient value as $\beta = (0.5, -0.5, 1, -1, 0.7, 0, 0, 0, 0, 0)$.
2. (45 points) Implement the MCMC algorithm using the simulated dataset in Problem 1. Here, you should write down the code without using any packages (e.g., `nimble`, `adaptMCMC`). Report followings:
 - Trace plots, density plots, 95% HPD intervals, posterior mean, acceptance probability, and effective sample size for $\beta_1, \dots, \beta_{10}, \sigma_1^2, \sigma_2^2$.
 - Posterior inclusion probability for all β_j . Based on the posterior inclusion probabilities, discuss whether we have to include j th predictor or not. Compare these results with the simulated truth $\beta = (0.5, -0.5, 1, -1, 0.7, 0, 0, 0, 0, 0)$.
3. (45 points) Implement the MCMC algorithm using the simulated dataset in Problem 1. Here, you should use `nimble` package to implement the algorithm. Report followings:
 - Trace plots, density plots, 95% HPD intervals, posterior mean, acceptance probability, and effective sample size for $\beta_1, \dots, \beta_{10}, \sigma_1^2, \sigma_2^2$.
 - Posterior inclusion probability for all β_j . Based on the posterior inclusion probabilities, discuss whether we have to include j th predictor or not. Compare these results with the simulated truth $\beta = (0.5, -0.5, 1, -1, 0.7, 0, 0, 0, 0, 0)$.