

Homework 8

Setup

$$y_i \sim N(\beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_{100} x_{i(100)}, \sigma^2)$$

$$\sigma^2 = 1$$

$$\beta_1, \dots, \beta_5 \sim N(0, 2)$$

$$\beta_6, \dots, \beta_{100} = 0$$

$$x_i \sim N_{100}(0, I)$$

$$n = 100$$

For this simulation, I set 5 non-zero betas β_1, \dots, β_5 [each drawn from $N(0, 2)$] and 95 betas $\beta_6, \dots, \beta_{100}$ equal to zero. I then simulated 100 x_i 's each with its predictors drawn independently from a standard normal $N(0, 1)$. Finally I calculated the values of y_i using the beta and x_i values plus noise distributed $N(0, 1)$.

Implementation of different methods

Fully Bayes ridge: I set the priors for τ and σ^{-2} to be $\tau \sim Ga(0.5, 0.5)$ and $\sigma^{-2} \sim Ga(0.001, 0.001)$. Using the full conditionals (derived in the Appendix), I performed 1000 iterations of Gibbs sampling with the initial value of the betas set to the true values of the betas. After examining the trace plots, no burn-ins were needed, perhaps because my initial values of the betas were the true values.

t-prior with $\nu = 1$: Similarly, I set the priors for τ and σ^{-2} to be $\tau \sim Ga(0.5, 0.5)$ and $\sigma^{-2} \sim Ga(0.001, 0.001)$. Using the full conditionals (derived in the Appendix), I performed 1000 iterations of Gibbs sampling with the initial value of the betas set to the true values of the betas. After examining the trace plots, no burn-ins were needed, perhaps because my initial values of the betas were the true values.

t-prior with $\nu = 0.001$: Similarly, I set the priors for τ and σ^{-2} to be $\tau \sim Ga(0.5, 0.5)$ and $\sigma^{-2} \sim Ga(0.001, 0.001)$. Using the full conditionals (derived in the Appendix), I performed 1000 iterations of Gibbs sampling with the initial value of the betas set to the true values of the betas. After examining the trace plots, no burn-ins were needed, perhaps because my initial values of the betas were the true values.

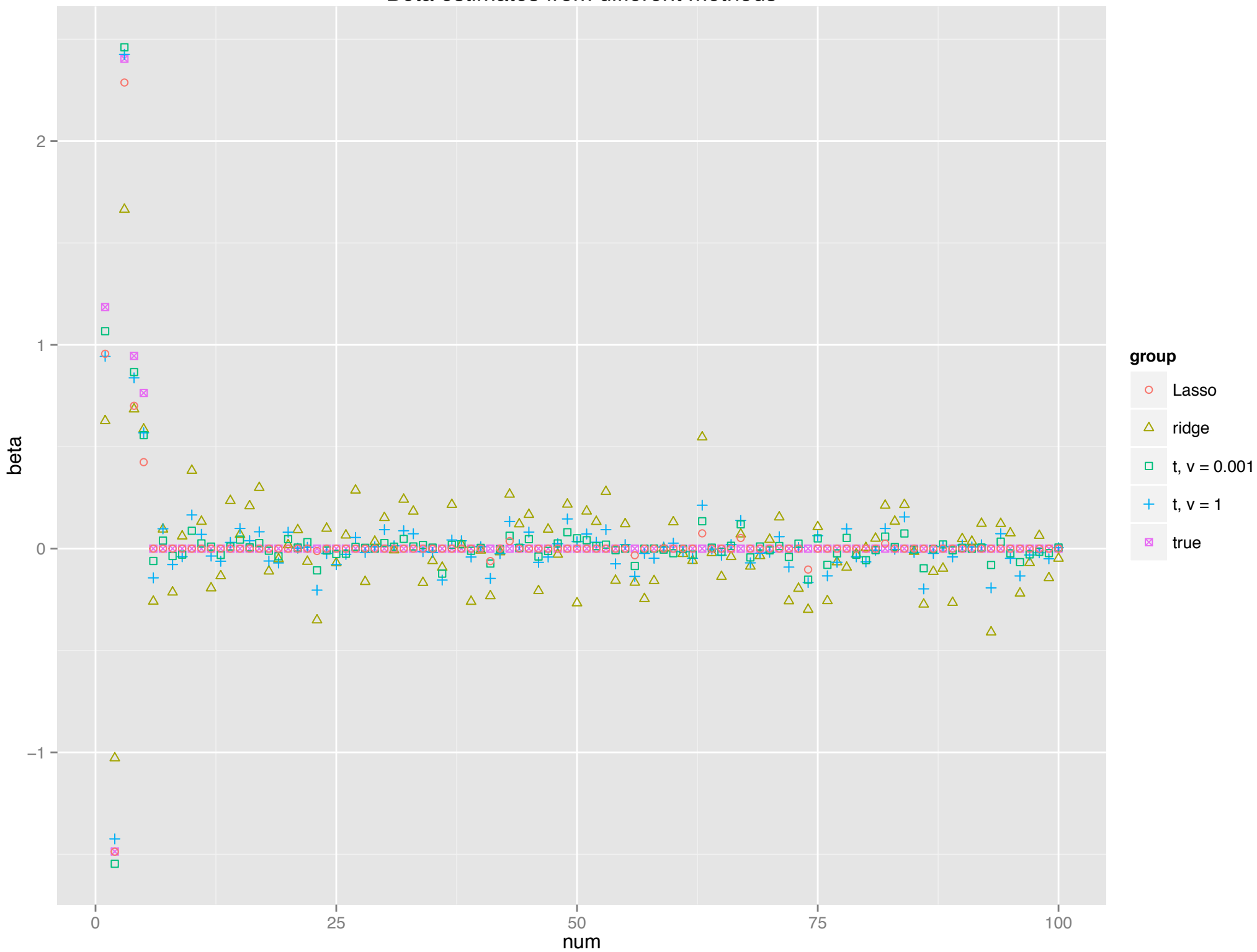
Lasso: I utilized the glmnet package in R to get Lasso estimates of the betas.

Comparison among different methods

The following is a plot comparing the beta estimates among the different methods.

- The fully Bayes ridge method appears to pull the estimates of the true (non-zero) betas too close to 0 while pulling the estimates of the fake betas too much out from 0
- The t-prior methods appears to estimate the true (non-zero) betas slightly better than Lasso, but Lasso estimates the fake betas better

Beta estimates from different methods



For each method, I calculated the MSE on the β 's and the predictive MSE on a newly generated test dataset of 100 observations. The results are summarized in the table below.

Table: Comparison of MSE's among different methods for estimating β 's

	MSE on β 's	Predictive MSE
Fully Bayes ridge	0.0411	4.4731
t -prior with $\nu = 1$	0.0074	1.7688
t -prior with $\nu = 0.001$	0.0027	1.2153
Lasso	0.0026	1.2270

- The fully Bayes ridge regression performs the worst in terms of both estimating β 's and predicting new observations
- The t -prior with the smaller ν value performs better than the t -prior with the larger ν value in terms of both MSE on β 's and predictive MSE
- The Bayesian method of using a t -prior with $\nu = 0.001$ performs about the same as the frequentist Lasso method in terms of both estimating β 's and predicting new observations. Both of these methods do very well compared to the others.

Fully Bayes ridge

$$y \sim N_{100}(X\beta, \sigma^2 I)$$

$$\beta \sim N_{100}(0, \tau^{-1} \sigma^2 I)$$

$$\tau \sim \text{Ga}(a, b)$$

$$\sigma^{-2} \sim \text{Ga}(c, d)$$

$$\pi(\beta, \tau, \sigma^{-2} | x, y) = L(y | x, \beta, \sigma^{-2}) \pi(\beta | \tau, \sigma^{-2}) \pi(\tau) \pi(\sigma^{-2})$$

$$\propto |\sigma^2 I|^{-\frac{1}{2}} \exp\left[-\frac{1}{2}(y - X\beta)^T (\sigma^2 I)^{-1} (y - X\beta)\right] |\tau^{-1} \sigma^2 I|^{-\frac{1}{2}} \exp\left[-\frac{1}{2}\beta^T (\tau^{-1} \sigma^2 I)^{-1} \beta\right] \tau^{a-1} e^{-b\tau} (\sigma^{-2})^{c-1} e^{-d\sigma^{-2}}$$

$$\propto (\sigma^{-2})^{-50} \exp\left[-\frac{1}{2\sigma^2}(y^T y - y^T X\beta - \beta^T X^T y + \beta^T X^T X \beta) - \frac{\tau}{2\sigma^2} \beta^T \beta\right] (\tau^{-1} \sigma^2)^{-50} \tau^{a-1} e^{-b\tau} (\sigma^{-2})^{c-1} e^{-d\sigma^{-2}}$$

$$\propto (\sigma^{-2})^{c+99} e^{-d\sigma^{-2}} \tau^{a+99} e^{-b\tau} \exp\left[-\frac{1}{2\sigma^2}(y^T y - 2\beta^T X^T y + \beta^T X^T X \beta + \tau \beta^T \beta)\right]$$

$$\pi(\beta | \tau, \sigma^{-2}, x, y) \propto \exp\left[-\frac{1}{2\sigma^2}(-2\beta^T X^T y + \beta^T X^T X \beta + \tau \beta^T \beta)\right]$$

$$\propto \exp\left[-\frac{1}{2\sigma^2}[\beta^T (X^T X + \tau I) \beta - 2\beta^T (X^T y)]\right]$$

$$\boxed{\beta | \tau, \sigma^{-2}, x, y \sim N_{100}((X^T X + \tau I)^{-1} (X^T y), (X^T X + \tau I)^{-1} \sigma^2)}$$

$$\pi(\tau | \beta, \sigma^{-2}, x, y) \propto \tau^{a+99} \exp\left[-\tau(b + \frac{1}{2}\sigma^{-2} \beta^T \beta)\right]$$

$$\boxed{\tau | \beta, \sigma^{-2}, x, y \sim \text{Ga}(a+50, b + \frac{1}{2}\sigma^{-2} \beta^T \beta)}$$

$$\pi(\sigma^{-2} | \beta, \tau, x, y) \propto (\sigma^{-2})^{c+99} \exp\left[-\sigma^{-2}(d + \frac{1}{2}(y - X\beta)^T (y - X\beta) + \frac{1}{2}\tau \beta^T \beta)\right]$$

$$\boxed{\sigma^{-2} | \beta, \tau, x, y \sim \text{Ga}(c+100, d + \frac{1}{2}(y - X\beta)^T (y - X\beta) + \frac{1}{2}\tau \beta^T \beta)}$$

t-prior

$$y \sim N_{100}(X\beta, \sigma^2 I)$$

$$\beta \sim N_{100}(0, \tau^{-1} L^{-1})$$

$$\text{where } L = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_{100} \end{bmatrix}$$

$$\lambda_j \sim \text{Ga}(\frac{\nu}{2}, \frac{\nu}{2})$$

$$\tau \sim \text{Ga}(a, b)$$

$$\sigma^{-2} \sim \text{Ga}(c, d)$$

$$\pi(\beta, \lambda, \tau, \sigma^{-2} | x, y) = L(y | x, \beta, \sigma^{-2}) \pi(\beta | \lambda, \tau) \pi(\lambda) \pi(\tau) \pi(\sigma^{-2})$$

$$\propto |\sigma^2 I|^{-\frac{1}{2}} \exp[-\frac{1}{2}(y - X\beta)^T (\sigma^2 I)^{-1} (y - X\beta)] |\tau^{-1} L^{-1}|^{-\frac{1}{2}} \exp[-\frac{1}{2} \beta^T (\tau^{-1} L^{-1})^{-1} \beta] \left[\prod_{j=1}^{100} \lambda_j^{\frac{\nu}{2}-1} e^{-\frac{\nu}{2} \lambda_j} \right] \tau^{a-1} e^{-b\tau} (\sigma^{-2})^{c-1} e^{-d\sigma^{-2}}$$

$$\propto (\sigma^2)^{-50} \exp[-\frac{1}{2\sigma^2} (y^T y - y^T X\beta - \beta^T X^T y + \beta^T X^T X \beta) - \frac{\tau}{2} \beta^T L \beta] \tau^{50} \left[\prod_{j=1}^{100} \lambda_j^{\frac{1}{2}} \right] \left[\prod_{j=1}^{100} \lambda_j^{\frac{\nu}{2}-1} \right] \exp[-\frac{\nu}{2} \sum_{j=1}^{100} \lambda_j] \tau^{a-1} e^{-b\tau} (\sigma^{-2})^{c-1} e^{-d\sigma^{-2}}$$

$$\propto (\sigma^{-2})^{c+49} e^{-d\sigma^{-2}} \tau^{a+49} e^{-b\tau} \left[\prod_{j=1}^{100} \lambda_j^{\frac{\nu}{2}-\frac{1}{2}} \right] \exp[-\frac{\nu}{2} \sum_{j=1}^{100} \lambda_j] \exp[-\frac{1}{2\sigma^2} (y^T y - 2\beta^T X^T y + \beta^T X^T X \beta) - \frac{\tau}{2} \beta^T L \beta]$$

$$\pi(\beta | \lambda, \tau, \sigma^{-2}, x, y) \propto \exp[-\frac{1}{2} (-2\sigma^{-2} \beta^T X^T y + \sigma^{-2} \beta^T X^T X \beta + \tau \beta^T L \beta)]$$

$$\propto \exp[-\frac{1}{2} [\beta^T (\sigma^{-2} X^T X + \tau L) \beta - 2\beta^T (\sigma^{-2} X^T y)]]$$

$$\beta | \lambda, \tau, \sigma^{-2}, x, y \sim N_{100}((\sigma^{-2} X^T X + \tau L)^{-1} (\sigma^{-2} X^T y), (\sigma^{-2} X^T X + \tau L)^{-1})$$

$$\text{where } L = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_{100} \end{bmatrix}$$

$$\pi(\lambda_j | \beta, \lambda_{(-j)}, \tau, \sigma^{-2}, x, y) \propto \lambda_j^{\frac{\nu}{2}-\frac{1}{2}} \exp[-\lambda_j \frac{1}{2} (\tau \beta_j^2 + \nu)]$$

$$\lambda_j | \beta, \lambda_{(-j)}, \tau, \sigma^{-2}, x, y \sim \text{Ga}(\frac{\nu+1}{2}, \frac{1}{2} (\tau \beta_j^2 + \nu))$$

$$\pi(\tau | \beta, \lambda, \sigma^{-2}, x, y) \propto \tau^{a+49} \exp[-\tau (b + \frac{1}{2} \beta^T L \beta)]$$

$$\tau | \beta, \lambda, \sigma^{-2}, x, y \sim \text{Ga}(a+50, b + \frac{1}{2} \beta^T L \beta)$$

$$\pi(\sigma^{-2} | \beta, \lambda, \tau, x, y) \propto (\sigma^{-2})^{c+49} \exp[-\sigma^{-2} (d + \frac{1}{2} (y - X\beta)^T (y - X\beta))]$$

$$\sigma^{-2} | \beta, \lambda, \tau, x, y \sim \text{Ga}(c+50, d + \frac{1}{2} (y - X\beta)^T (y - X\beta))$$