

## Assignment 5

File `stockprices.txt` contains monthly opening prices of fifteen randomly-selected stocks listed on the NASDAQ stock market for each month of 2017. You will build and compare two different varying-coefficient hierarchical normal regression models for the *log*-prices, using JAGS and `rjags`.

Let  $y_{ij}$  be the *natural logarithm* of the price of stock  $j$  at the beginning of month  $i$  ( $i = 1, \dots, 12$ ,  $j = 1, \dots, 15$ ), where the months are numbered in the order in which they occurred. For each stock, model the log-price as a simple linear regression on the *centered* month index:

$$y_{ij} \mid \beta^{(j)}, \sigma_y^2, X \sim \text{indep. } N(\beta_1^{(j)} + \beta_2^{(j)}(x_i - \bar{x}), \sigma_y^2)$$

where

$$\beta^{(j)} = \begin{pmatrix} \beta_1^{(j)} \\ \beta_2^{(j)} \end{pmatrix} \quad j = 1, \dots, 15 \quad x_i = i \quad i = 1, \dots, 12$$

Note that the coefficients are allowed to depend on the stock, but the variance is not.

- (a) [2 pts] Let  $\hat{\beta}_1^{(j)}$  and  $\hat{\beta}_2^{(j)}$  be the *ordinary least squares* estimates of  $\beta_1^{(j)}$  and  $\beta_2^{(j)}$ , estimated for each stock separately. Produce a scatterplot of the pairs  $(\hat{\beta}_1^{(j)}, \hat{\beta}_2^{(j)})$ , and also compute the sample average of  $\hat{\beta}_1^{(j)}$  and of  $\hat{\beta}_2^{(j)}$ .

- (b) Consider the bivariate prior

$$\begin{aligned} \beta^{(j)} \mid \mu_\beta, \Sigma_\beta &\sim \text{iid } N(\mu_\beta, \Sigma_\beta) \\ \mu_\beta &= \begin{pmatrix} \mu_{\beta_1} \\ \mu_{\beta_2} \end{pmatrix} \quad \Sigma_\beta = \begin{pmatrix} \sigma_{\beta_1}^2 & \rho \sigma_{\beta_1} \sigma_{\beta_2} \\ \rho \sigma_{\beta_1} \sigma_{\beta_2} & \sigma_{\beta_2}^2 \end{pmatrix} \end{aligned}$$

with hyperpriors

$$\begin{aligned} \mu_\beta &\sim N(0, 1000^2 I) \\ \Sigma_\beta^{-1} &\sim \text{Wishart}_2(\Sigma_0^{-1}/2) \end{aligned}$$

in the notation used in the lecture videos. For your analysis, use

$$\Sigma_0 = \begin{pmatrix} 1 & 0 \\ 0 & 0.01 \end{pmatrix}$$

based on preliminary analyses. Let the prior on  $\sigma_y^2$  be

$$\sigma_y^2 \sim \text{Inv-gamma}(0.0001, 0.0001)$$

- (i) [2 pts] List an appropriate JAGS model. Make sure to create nodes for  $\Sigma_\beta$ ,  $\rho$ , and  $\sigma_y^2$ .

Remember that the stock prices are to be analyzed on the *log scale*.

Now run your model using `rjags`. Make sure to use multiple chains with overdispersed starting points, check convergence, and monitor  $\mu_\beta$ ,  $\Sigma_\beta$ ,  $\sigma_y^2$ , and  $\rho$  (after convergence) long enough to obtain *effective sample sizes of at least 4000* for each parameter.

- (ii) [2 pts] Display the `coda` summary of the results for the monitored parameters.
- (iii) [2 pts] Give an approximate 95% central posterior credible interval for the correlation parameter  $\rho$ , and also produce a graph of its (estimated) posterior density.
- (iv) [2 pts] Approximate the posterior probability that  $\rho > 0$ . Also, compute the Bayes factor favoring  $\rho > 0$  versus  $\rho < 0$ . (You may use the fact that  $\rho > 0$  and  $\rho < 0$  have equal prior probability.) Describe the level of data evidence that  $\rho > 0$ . ? ? ? ?
- (v) [2 pts] Over the period of the data, the NASDAQ composite index rose from 5,425.62 to 6,844.04, a gain of 26.1%.

Your model implies that, over the 11 months from the first time point to the last, the (population) median stock price should have changed by a factor of

$$e^{11\mu_{\beta_2}}$$

Form an approximate 95% central posterior credible interval for this quantity, and compare it with the change in the NASDAQ composite.

- (vi) [2 pts] Use the `rjags` function `dic.samples` to compute the effective number of parameters (“penalty”) and Plummer’s DIC (“Penalized deviance”). Use at least 100,000 iterations.
- (c) Now consider a different model with “univariate” hyperpriors for the model coefficients, which do not allow for a coefficient correlation parameter:

$$\begin{aligned}\beta_1^{(j)} \mid \mu_{\beta_1}, \sigma_{\beta_1}^2 &\sim \text{iid } N(\mu_{\beta_1}, \sigma_{\beta_1}^2) \\ \beta_2^{(j)} \mid \mu_{\beta_2}, \sigma_{\beta_2}^2 &\sim \text{iid } N(\mu_{\beta_2}, \sigma_{\beta_2}^2)\end{aligned}$$

with hyperpriors

$$\begin{aligned}\mu_{\beta_1}, \mu_{\beta_2} &\sim \text{iid } N(0, 1000^2) \\ \sigma_{\beta_1}, \sigma_{\beta_2} &\sim \text{iid } U(0, 1000)\end{aligned}$$

Let the prior on  $\sigma_y^2$  be the same as in the previous model.

- (i) [2 pts] Draw a complete DAG for this new model.
- (ii) [2 pts] List an appropriate JAGS model. Make sure that there are nodes for  $\sigma_{\beta_1}^2$ ,  $\sigma_{\beta_2}^2$ , and  $\sigma_y^2$ .

Remember that the stock prices are to be analyzed on the *log* scale.

Now run your model using `rjags`. Make sure to use multiple chains with overdispersed starting points, check convergence, and monitor  $\mu_{\beta_1}$ ,  $\mu_{\beta_2}$ ,  $\sigma_{\beta_1}^2$ ,  $\sigma_{\beta_2}^2$ ,  $\sigma_y^2$  (after convergence) long enough to obtain effective sample sizes of at least 4000 for each parameter.

- (iii) [2 pts] Display the `coda` summary of the results for the monitored parameters.

- (iv) [2 pts] Recall the (population) median stock price change factor

$$e^{11\mu_{\beta_2}}$$

from the previous analysis. Form an approximate 95% central posterior credible interval for this quantity, and compare it with the previous results.

- (v) [2 pts] Use the `rjags` function `dic.samples` to compute the effective number of parameters (“penalty”) and Plummer’s DIC (“Penalized deviance”). Use at least 100,000 iterations.
- (vi) [1 pt] Compare the (Plummer’s) DIC values for this model and the previous one. Which is preferred?
- (d) (i) [2 pts] It is possible that the variability in log-price (volatility) depends on the stock. How might you modify your model to account for this? Would your solution need more hyperparameters?
- (ii) [1 pt] It is possible that there are time-series correlations that are not captured by the simple linear regression model. What specific model assumption would this violate?

Total: 28 pts