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Week 2 Homework

1. Using the results in Chapter 2, explore Bayes regression:
  - a. simulate from a regression model with  $k=3$ .  $n=30$ . To calibrate your simulations, pool the cheese data and regress  $\log(\text{VOLUME})$  on  $\text{DISP}$  and  $\log(\text{PRICE})$  to set beta and sigma-squared. Sample from the rows of  $X$  for this pooled regression. see the R command `sample` to do this.
  - b. Write R code to sample from the prior. Note you will need to sample from a multivariate normal and a Chi-squared. Chi-squared is available as `rchisq` in R; for multivariate normal see Chapter 2 notes.
  - c. Plot the prior distribution of beta elements and sigma for various prior settings. Consider the “standard diffuse” prior

$$\sigma^2 \sim \frac{\nu_0 s_0^2}{\chi_{\nu_0}^2} \quad \nu_0 = 3; \quad s_0^2 = 1$$

$$\beta \mid \sigma^2 \sim N(0, \sigma^2 A^{-1}) \quad A = .01 I_{\dim(\beta)}$$

as well as a much tighter prior (change  $A$  by a factor of 100!)

- d. What is a possible objection to the prior on  $\sigma^2$ ? Hint: look at your simulated data for the standard diffuse prior settings.
  - e. Compare the prior and posterior distributions of the model parameters. (here you might find the `boxplot` command useful. If you have data in a matrix  $X$ , `boxplot(data.frame(X))` will give you a set of boxplots on the same plotting frame of the columns of  $X$ . Use `runireg` to draw from the posterior.
2. The SUR model is discussed in section 2.8.6. Given  $\Sigma$ , there is a normal conjugate prior for this model. Write down the posterior using this prior.
3. In the homework directory, there is a file `rmultireg_wrong.R`. This is R code to simulate from the posterior of the multivariate regression model. There is a conceptual error in this code. Locate the error and explain why it might be hard to detect by pure simulation methods alone. Hints: 1). the error results the wrong value of  $S$  being computed. 2). Read section 2.12 carefully.
4. Gelman et al p. 581 suggest that you can simulate from a Wishart distribution with df parameter  $\nu$  and location  $S$ ,  $W(\nu, S)$ , by simulating  $x_i \sim N(0, S); i = 1, \dots, \nu$  and computing  $W = \sum_{i=1}^{\nu} x_i x_i'$ . Code this up in R. `rwishart` in *bayesm* does this by a much more efficient method. Use `rwishart` to check your code for accuracy. Why is the method in `rwishart` and discussed on p. 44 more efficient?