## CPMA 573 — Homework #5

Exercise 1: Normal model inference. Let  $Y_1, \ldots, Y_{43}$  be iid normal random variables from the density

$$f(y|\mu,\sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{1}{2\sigma^2}(y-\mu)^2\right\}$$
.

Realizations from this data model are located at

www.mathcs.duq.edu/~kern/hw5.dat

It is your goal to make inference on the values of  $\mu$  and  $\sigma^2$  used to generate these data.

**a.** Assuming  $\mu$  and  $\sigma^2$  are a priori independent, with prior densities  $\pi(\mu) \propto 1$  and  $\pi(\sigma^2) \propto \sigma^{-2}$ , obtain 25000 draws from the marginal posteriors  $\pi(\mu|\vec{y})$  and  $\pi(\sigma^2|\vec{y})$ using

Method 1: The Gibbs sampler. (Be sure to check for zero autocorrelation using 'acf' plots.)

Method 2: Independent draws directly from the (theoretical) marginal distribution of  $\sigma^2$  in conjunction with the full conditional distribution for  $\mu$ .

For both methods, provide trace plots ('ts.plot') and histograms of your  $\mu$  and  $\sigma^2$ realizations. Just for kicks, superpose the theoretical marginal density of  $\sigma^2$  on both histograms of  $\sigma^2$  realizations. The four plots for Method 1 can be produced in R as follows:

par(mfrow=c(2,2)) #Splits the plotting window into two rows and two columns ts.plot(mu1,xlab=''Iteration'') #mu1 represents Method 1 realizations of mu ts.plot(sigsq1,xlab="'Iteration") #sigsq1 as with mu1

hist(mu1,probability=T)

hist(sigsq1,probability=T)

lines(yy, IGdens(yy)) #Here yy is a vector, and IGdens is your own #inverse gamma density function

b. Use the quantile function in R in conjunction with your posterior draws to find the values  $(b_1, b_2)$  and  $(c_1, c_2)$  that satisfy the following posterior probabilities:

• 
$$\Pr(b_1 < \mu < b_2) = k$$
 [with  $\Pr(\mu < b_1) = (1 - k)/2$ ]

• 
$$\Pr(c_1 < \sigma^2 < c_2) = k$$
 [with  $\Pr(\sigma^2 < c_1) = (1 - k)/2$ ]

for  $k = \{0.95, 0.99\}.$ 

c. Provide your estimates of  $\mu$  and  $\sigma^2$ , along with corresponding 95% credible intervals. (Credible intervals are the Bayesian analog to confidence intervals. They are named differently because their interpretation is different.)

Exercise 2: Posterior predictive distribution. Use the 25000  $(\mu, \sigma^2)$  pairs generated in the previous problem—from either method—to generate 25000 predicted y-values. Based on these predicted y-values, answer the following:

- a. What is the chance that the next (44th) observation is greater than 10?
- b. What is the shortest interval that has a 95% chance of containing the next

Values of yours vse quantill totaly 2.5%, 97,5%

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