Lab 4: Hierarchical models for Kdrama rating

Author:	_(Insert	your	name l	here)	
Total Grade for	Lab 4:	/18			

Comments (optional)

Template for lab report

Instructions: This is the template you will use to type up your responses to the exercises. To produce a document that you can print out and turn in just click on Knit PDF above. All you need to do to complete the lab is to type up your BRIEF answers and the R code (when necessary) in the spaces provided below.

It is strongly recommended that you knit your document regularly (minimally after answering each exercise) for two reasons.

- 1. Ensure that there are no errors in your code that would prevent the document from knitting.
- 2. View the instructions and your answers in a more legible, attractive format.

```
# Any text BOTH preceded by a hashtag AND within the ```{r}``` code chunk is a comment.
# R indicates a comment by turning the text green in the editor, and brown in the knitted
# document.
# Comments are not treated as a command to be interpreted by the computer.
# They normally (briefly!) describe the purpose of your command or chunk in plain English.
# However, for this class, they will have a different goal, as the text above and below
# each chunk should sufficiently describe the chunk's contents.
# For this class, comments will be used to indicate where your code should go, or to give
# hints for what the code should look like.
```

Overview

We have explored the Kdrama rating dataset with a hierarchical model. The sampling density, the two-stage prior distribution for μ_i 's and the prior distribution for σ are presented below.

• The sampling density for group j, and $j = 1, \dots, J$:

$$Y_{ij} \overset{i.i.d.}{\sim} \text{Normal}(\mu_i, \sigma),$$
 (1)

where $i = 1, \dots, n_j$ and n_j is the number of observations in group j.

• The stage 1 prior distribution for μ_i :

$$\mu_j \sim \text{Normal}(\mu, \tau).$$
 (2)

• The stage 2 prior distribution for μ_j :

$$\mu, \tau \sim g(\mu, \tau).$$
 (3)

Hyperpriors:

$$\mu \mid \mu_0, \gamma_0 \sim \text{Normal}(\mu_0, \gamma_0),$$
 (4)

$$1/\tau^2 \mid \alpha_{\tau}, \beta_{\tau} \sim \operatorname{Gamma}(\alpha_{\tau}, \beta_{\tau}).$$
 (5)

• The prior distribution for σ :

$$1/\sigma^2 \sim \text{Gamma}(\alpha_{\sigma}, \beta_{\sigma}).$$
 (6)

We have set $\mu_0 = 0.1$, $\gamma_0 = 0.5$, $\alpha_\tau = \beta_\tau = \alpha_\sigma = \beta_\sigma = 1$, and obtained posterior summaries below:

	Lower95	Median	Upper95	Mean	SD	Mode	MCerr	MC%ofSD	SSeff	AC.10	psrf
mu	-0.4905	0.1080	0.668	0.1047	0.2884	NA	0.004181	1.4	4758	-0.01052	NA
tau	0.3527	0.6585	1.250	0.7206	0.2749	NA	0.004198	1.5	4288	-0.01511	NA
mu_j[1]	-0.0720	0.0713	0.217	0.0727	0.0724	NA	0.001024	1.4	5000	0.00223	NA
mu_j[2]	-0.0569	0.0994	0.253	0.0993	0.0800	NA	0.001131	1.4	5000	0.01881	NA
mu_j[3]	-0.2415	0.0448	0.349	0.0427	0.1511	NA	0.002073	1.4	5310	-0.00797	NA
mu_j[4]	-0.0248	0.1914	0.399	0.1924	0.1075	NA	0.001520	1.4	5000	-0.01727	NA
sigma	0.2011	0.2616	0.333	0.2650	0.0346	NA	0.000554	1.6	3908	0.00988	NA

We have noticed the issues of negative draws of some parameter which should have been strictly non-negative, including mu, $mu_j[1]$ through $mu_j[4]$, corresponding to μ and μ_1 through μ_4 in the model. Since these 5 parameters indicate the mean of the mean rating, and the means of ratings, they should be non-negative.

This lab is to explore different prior specifications that would prevent this from happening.

Truncated Normal distributions for the hyperprior for μ and priors for μ_j 's

We know that μ and μ_j 's should be non-negative. If we still want to use a hyperprior/prior distribution related to the Normal distribution, we can consider the truncated normal distribution.

The truncated Normal distribution

From Wikipedia: The truncated normal distribution is the probability distribution derived from that of a normally distributed random variable by bounding the random variable from either below or above (or both).

Suppose $Y \sim \text{Normal}(\mu, \sigma)$ has a Normal distribution and lies within the interval $Y \in (a, b), -\infty \leq a < b \leq \infty$. Then Y conditional on a < Y < b has a truncated Normal distribution, with pdf:

$$f(y \mid \mu, \sigma, a, b) = \frac{\phi(\frac{y-\mu}{\sigma})}{\sigma\left(\Phi(\frac{b-\mu}{\sigma}) - \Phi(\frac{a-\mu}{\sigma})\right)},\tag{7}$$

where $\phi(.)$ is the pdf of the standard Normal distribution (i.e. Normal(0,1)) and $\Phi(.)$ is the cdf (cumulative distribution function) of the standard Normal distribution.

Specifying a truncated Normal hyperprior/prior in JAGS

In the previous hierarchical model, where regular Normal prior distribution is assigned to μ_i , the syntax is:

```
for (j in 1:J){
mu_j[j] ~ dnorm(mu, invtau2)
}
```

If we want to use a truncated Normal prior distribution with only non-negative values of μ_j 's, one can use the following syntax:

```
for (j in 1:J){
mu_j[j] ~ dnorm(mu, invtau2)T(0,)
}
```

Exercise 1: Give appropriate truncated Normal prior distribution for μ_j 's and truncated Normal hyperprior distribution for μ . Run the new hierarchical model, and obtain the posterior summaries for all 7 parameters. Verify that the posterior draws of mu and mu_j[1] through mu_j[4] are all non-negative. Include the 2-by-2 traceplot + cdf + historgram + ACF plot for mu_j[1] (Hint: use the plot(posterior, vars = "mu_j[1]" command). Comment on the MCMC diagnostics for mu_j[1].

Grade for Exercise 1: /6

Comments:

Log-normal distributions for the hyperprior for μ and priors for μ_i 's

In addition to truncated Normal distribution, we can also consider the log-normal distribution.

The log-normal distribution

From Wikipedia: a log-normal (or lognormal) distribution is a continuous probability distribution of a random variable whose logarithm is Normally distributed. Thus, if the random variable Y is log-normally distributed, then $Y' = \ln(Y)$ has a Normal distribution.

A random variable which is log-normally distributed takes only positive real values, an appealing feature for μ and μ_j 's in the Kdrama rating application.

If $Y \sim Normal(\mu, \sigma)$, its pdf is:

$$f(y \mid \mu, \sigma) = \frac{1}{y} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(\ln(y) - \mu)^2}{2\sigma^2}\right). \tag{8}$$

Specifying a log-normal hyperprior/prior in JAGS

In the previous hierarchical model, where regular Normal prior distribution is assigned to μ_i , the syntax is:

```
for (j in 1:J){
mu_j[j] ~ dnorm(mu, invtau2)
}
```

If we want to use a log-normal prior distribution with only non-negative values of μ_j 's, one can use the following syntax:

```
for (j in 1:J){
mu_j[j] ~ dlnorm(mu, invtau2)
}
```

Exercise 2: Give appropriate log-normal prior distribution for μ_j 's and log-normal hyperprior distribution for μ . Run the new hierarchical model, and obtain the posterior summaries for all 7 parameters. Verify that the posterior draws of mu and mu_j[1] through mu_j[4] are all non-negative. Include the 2-by-2 traceplot + cdf + historgram + ACF plot for mu_j[1] (Hint: use the plot(posterior, vars = "mu_j[1]" command). Comment on the MCMC diagnostics for mu_j[1].

Grade for Exercise 2: /6

Comments:

Your choice of distribution for the hyperprior for μ and priors for μ_j 's

Exercise 3: Give appropriate prior distribution for μ_j 's and hyperprior distribution for μ of your own choosing. Run the new hierarchical model, and obtain the posterior summaries for all 7 parameters. Verify that the posterior draws of mu and mu_j[1] through mu_j[4] are all non-negative. Include the 2-by-2 traceplot + cdf + historgram + ACF plot for mu_j[1] (Hint: use the plot(posterior, vars = "mu_j[1]" command). Comment on the MCMC diagnostics for mu_j[1].

Grade for Exercise 3: /6

Comments: