Full conditional distributions for a Bayesian change point model

Our goal is to draw samples from the 5-dimensional **posterior** distribution $f(k, \theta, \lambda, b_1, b_2 | \mathbf{Y})$ The posterior distribution is

$$f(k, \theta, \lambda, b_1, b_2 | \mathbf{Y}) \propto \prod_{i=1}^{k} \frac{\theta^{Y_i} e^{-\theta}}{Y_i!} \prod_{i=k+1}^{n} \frac{\lambda^{Y_i} e^{-\lambda}}{Y_i!} \times \frac{1}{\Gamma(0.5)b_1^{0.5}} \theta^{-0.5} e^{-\theta/b_1} \times \frac{1}{\Gamma(0.5)b_2^{0.5}} \lambda^{-0.5} e^{-\lambda/b_2}$$

$$\times \frac{e^{-1/b_1}}{b_1} \frac{e^{-1/b_2}}{b_2} \times \frac{1}{n}$$
(1)

From 1 we can obtain full conditional distributions for each parameter by ignoring all terms that are constant with respect to the parameter. For θ :

$$f(\theta|k,\lambda,b_{1},b_{2},\mathbf{Y}) \propto \prod_{i=1}^{k} \frac{\theta^{Y_{i}}e^{-\theta}}{Y_{i}!} \times \frac{1}{\Gamma(0.5)b_{1}^{0.5}} \theta^{-0.5}e^{-\theta/b_{1}}$$

$$\propto \theta^{\sum_{i=1}^{k} Y_{i}-0.5}e^{-\theta(k+1/b_{1})}$$

$$\propto \operatorname{Gamma}\left(\sum_{i=1}^{k} Y_{i}+0.5, \frac{b_{1}}{kb_{1}+1}\right)$$
(2)

For λ :

$$f(\lambda|k, \theta, b_1, b_2, \mathbf{Y}) \propto \prod_{i=k+1}^{n} \frac{\lambda^{Y_i} e^{-\lambda}}{Y_i!} \times \frac{1}{\Gamma(0.5) b_2^{0.5}} \lambda^{-0.5} e^{-\lambda/b_2}$$

$$\propto \text{Gamma}\left(\sum_{i=k+1}^{n} Y_i + 0.5, \frac{b_2}{(n-k)b_2 + 1}\right)$$
(3)

For k:

$$f(k|\theta, \lambda, b_1, b_2, \mathbf{Y}) \propto \prod_{i=1}^{k} \frac{\theta^{Y_i} e^{-\theta}}{Y_i!} \prod_{i=k+1}^{n} \frac{\lambda^{Y_i} e^{-\lambda}}{Y_i!}$$
 (4)

For b_1 :

$$f(b_1|k,\theta,\lambda,b_2,\mathbf{Y}) \propto \frac{1}{b_1^{0.5}} e^{-\theta/b_1} \times \frac{e^{-1/b_1}}{b_1} \propto b_1^{-1.5} e^{-(1+\theta)/b_1} \propto IG(0.5,1/(\theta+1))$$
(5)

For b_2 :

$$f(b_2|k,\theta,\lambda,b_1|\mathbf{Y}) \propto \times \frac{1}{b_2^{0.5}} e^{-\lambda/b_2} \times \frac{e^{-1/b_2}}{b_2} \propto b_2^{-1.5} e^{-(1+\lambda)/b_2} \propto IG(0.5,1/(\lambda+1))$$
(6)

Note: The Inverse Gamma density is said to be a **conjugate** prior in this case since it results in a posterior that is also Inverse Gamma and therefore trivial to sample. As such, this density is mathematically convenient (due to its conjugacy property) but has poorly behaved moments; it may be better to adopt another prior density (such as a Gamma) instead.