## Penn State Astrostatistics MCMC tutorial

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## A Bayesian change point model

Consider the following hierarchical changepoint model for the number of occurrences  $Y_i$  of some event during time interval i with change point k.

$$\begin{aligned} Y_i | k, \theta, \lambda &\sim \text{Poisson}(\theta) \text{ for } i = 1, \dots, k \\ Y_i | k, \theta, \lambda &\sim \text{Poisson}(\lambda) \text{ for } i = k + 1, \dots, n \end{aligned}$$

Assume the following prior distributions:

$$\begin{aligned} \theta | b_1 \sim & \text{Gamma}(0.5, b_1) & (\text{pdf}=g_1(\theta | b_1)) \\ \lambda | b_2 \sim & \text{Gamma}(0.5, b_2) & (\text{pdf}=g_2(\lambda | b_2)) \\ b_1 \sim & \text{IG}(0, 1) & (\text{pdf}=h_1(b_1)) \\ b_2 \sim & \text{IG}(0, 1) & (\text{pdf}=h_2(b_2)) \\ k \sim & \text{Uniform}(1, \dots, n) & (\text{pmf}=u(k)) \end{aligned}$$

 $k, \theta, \lambda$  are conditionally independent and  $b_1, b_2$  are independent. Assume the Gamma density parameterization  $\operatorname{Gamma}(\alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^{\alpha}}x^{\alpha-1}e^{-x/\beta}$ and IG (Inverse Gamma) density parameterization  $\operatorname{IG}(\alpha, \beta) = \frac{e^{-1/\beta x}}{\Gamma(\alpha)\beta^{\alpha}x^{\alpha+1}}$ 

Inference for this model is therefore based on the 5-dimensional **posterior** distribution  $f(k, \theta, \lambda, b_1, b_2 | \mathbf{Y})$  where  $\mathbf{Y} = (Y_1, \ldots, Y_n)$ . The posterior distribution is obtained *up to a constant* (that is, the normalizing constant is unknown) by taking the product of all the conditional distributions. Thus we have

$$\begin{split} f(k,\theta,\lambda,b_{1},b_{2}|\mathbf{Y}) &\propto \prod_{i=1}^{k} f_{1}(Y_{i}|\theta,\lambda,k) \prod_{i=k+1}^{n} f_{2}(Y_{i}|\theta,\lambda,k) \\ &\times g_{1}(\theta|b_{1})g_{2}(\lambda|b_{2})h_{1}(b_{1})h_{2}(b_{2})u(k) \\ &= \prod_{i=1}^{k} \frac{\theta^{Y_{i}}e^{-\theta}}{Y_{i}!} \prod_{i=k+1}^{n} \frac{\lambda^{Y_{i}}e^{-\lambda}}{Y_{i}!} \\ &\times \frac{1}{\Gamma(0.5)b_{1}^{0.5}}\theta^{-0.5}e^{-\theta/b_{1}} \times \frac{1}{\Gamma(0.5)b_{2}^{0.5}}\lambda^{-0.5}e^{-\lambda/b_{2}} \\ &\times \frac{e^{-1/b_{1}}}{b_{1}}\frac{e^{-1/b_{2}}}{b_{2}}\frac{1}{n} \end{split}$$