

**Regression model with an unknown changepoint.** Suppose that the density for  $y_t$ ,  $t = 1, 2, \dots, T$ , can be expressed as

$$(y_t|x_t, \theta_1, \theta_2, \beta_1, \beta_2, \sigma^2, \tau^2, \lambda) \sim \begin{cases} N(\theta_1 + \theta_2 x_t, \sigma^2) & t \leq \lambda \\ N(\beta_1 + \beta_2 x_t, \tau^2) & t > \lambda \end{cases},$$

and that you employ priors of the form  $(\theta_1, \theta_2)' \sim N(\mu_\theta, V_\theta)$ ,  $(\beta_1, \beta_2)' \sim N(\mu_\beta, V_\beta)$ ,  $\sigma^2 \sim IG(a_1, b_1)$ ,  $\tau^2 \sim IG(a_2, b_2)$ , and  $\lambda \sim U\{1, 2, \dots, T - 1\}$ .

**Simulation.** Generate  $T = 500$  observations from the changepoint model  $y_t|x_t \sim N(2 + x_t, 0.2)$  for  $t \leq 85$  and  $N(1.5 + 0.8x_t, 0.5)$ , for  $t > 85$ , where  $x_t \sim N(0, 1)$ . Select prior hyperparameters as follows:  $\mu_\theta = \mu_\beta = 0_2$ ,  $V_\theta = V_\beta = 100I_2$ ,  $a_1 = b_1 = 3$ , and  $a_2 = b_2 = 0.5$ . sample from the joint posterior,  $p(\theta_1, \theta_2, \beta_1, \beta_2, \sigma^2, \tau^2, \lambda|y, x)$ , via a Gibbs sampler.

**Real data.** The data below represents yearly homicide rates,  $y_t$ , of young men between 15 and 29 years of age (in 100,000 inhabitants) in the city of São Paulo from 1980 to 2009 ( $T = 30$ ).

67.01 85.89 82.77 119.73 160.40 161.66 166.90 177.10 161.52 197.68  
 205.04 210.76 195.83 180.30 198.05 223.14 223.86 220.06 238.41 273.12  
 267.07 257.76 214.29 227.34 168.58 110.06 83.79 63.47 51.43 56.26

MLE of the two regressions, when  $\hat{\lambda} = 1998$  is assumed to be the change point, are given by  $y_t = 90 + 8x_t$  ( $\hat{\sigma} = 20$ ), for  $t \leq 1998$  and  $y_t = 822 - 26x_t$  ( $\hat{\tau} = 21$ ), for  $t > 1998$ , where  $x_t = t - 1979$ . Assuming the same priors as above, compute the posterior means, standard deviations, medians and 95% posterior credibility intervals of  $\theta_1, \theta_2, \sigma, \beta_1, \beta_2, \tau$  and  $\lambda$ .