
A Fully Bayesian Model for RNA-seq Data

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1 The Model

Let $y_{g,n}$ be the expression level of gene g ($g = 1, \dots, G$) in library n ($n = 1, \dots, N$). Let $\mu(n, \phi_g, \alpha_g, \delta_g)$ be the function given by:

$$\mu(n, \phi_g, \alpha_g, \delta_g) = \begin{cases} \phi_g - \alpha_g & \text{library } n \text{ is in treatment group 1} \\ \phi_g + \delta_g & \text{library } n \text{ is in treatment group 2} \\ \phi_g + \alpha_g & \text{library } n \text{ is in treatment group 3} \end{cases}$$

Then:

$$\begin{aligned} y_{g,n} &\sim \text{Poisson}(\exp(c_n + \varepsilon_{g,n} + \mu(n, \phi_g, \alpha_g, \delta_g))) \\ c_n &\sim \text{N}(0, \sigma_c^2) \\ \sigma_c &\sim \text{U}(0, \sigma_{c0}) \\ \varepsilon_{g,n} &\sim \text{N}(0, \sigma_g^2) \\ \sigma_g^2 &\sim \text{Inv-Gamma}\left(\text{shape} = \frac{d \cdot \tau^2}{2}, \text{rate} = \frac{d}{2}\right) \\ d &\sim \text{U}(0, d_0) \\ \tau^2 &\sim \text{Gamma}(\text{shape} = a_\tau, \text{rate} = b_\tau) \\ \phi_g &\sim \text{N}(\theta_\phi, \sigma_\phi^2) \\ \theta_\phi &\sim \text{N}(0, \gamma_\phi^2) \\ \sigma_\phi &\sim \text{U}(0, \sigma_{\phi 0}) \\ \alpha_g &\sim \pi_\alpha \text{I}(\alpha_g = 0) + (1 - \pi_\alpha) \text{I}(\alpha_g \neq 0) \text{N}(\theta_\alpha, \sigma_\alpha^2) \\ \theta_\alpha &\sim \text{N}(0, \gamma_\alpha^2) \\ \sigma_\alpha &\sim \text{U}(0, \sigma_{\alpha 0}) \\ \pi_\alpha &\sim \text{Beta}(a_\alpha, b_\alpha) \\ \delta_g &\sim \pi_\delta \text{I}(\delta_g = 0) + (1 - \pi_\delta) \text{I}(\delta_g \neq 0) \text{N}(\theta_\delta, \sigma_\delta^2) \\ \theta_\delta &\sim \text{N}(0, \gamma_\delta^2) \\ \sigma_\delta &\sim \text{U}(0, \sigma_{\delta 0}) \\ \pi_\delta &\sim \text{Beta}(a_\delta, b_\delta) \end{aligned}$$

where:

- Independence is implied unless otherwise specified.
- The parameters to the left of the “ \sim ” are implicitly conditioned on the parameters to the right.

This gives us the (unsimplified) full conditional posteriors below, where $k(n)$ is the treatment group of library n .

$$\begin{aligned}
p(c_n \mid \cdots) &= \prod_{g=1}^G \text{Poisson}(y_{g,n} \mid \exp(c_n + \varepsilon_{g,n} + \mu(n, \phi_g, \alpha_g, \delta_g))) \cdot \text{N}(c_n \mid 0, \sigma_c^2) \\
p(\sigma_c \mid \cdots) &= \prod_{n=1}^N \prod_{g=1}^G \text{N}(c_n \mid 0, \sigma_c^2) \cdot \text{U}(\sigma_c \mid 0, \sigma_{c0}) \\
p(\varepsilon_{g,n} \mid \cdots) &= \text{Poisson}(y_{g,n} \mid \exp(c_n + \varepsilon_{g,n} + \mu(n, \phi_g, \alpha_g, \delta_g))) \cdot \text{N}(\varepsilon_{g,n} \mid 0, \sigma_g^2) \\
p(\sigma_g^2 \mid \cdots) &= \prod_{n=1}^N \text{N}(\varepsilon_{g,n} \mid 0, \sigma_g^2) \cdot \text{Inv-Gamma}\left(\sigma_g^2 \mid \frac{d \cdot \tau^2}{2}, \frac{d}{2}\right) \\
p(d \mid \cdots) &= \prod_{n=1}^N \prod_{g=1}^G \text{Inv-Gamma}\left(\sigma_g^2 \mid \frac{d \cdot \tau^2}{2}, \frac{d}{2}\right) \cdot \text{U}(d \mid 0, d_0) \\
p(\tau^2 \mid \cdots) &= \prod_{n=1}^N \prod_{g=1}^G \text{Inv-Gamma}\left(\sigma_g^2 \mid \frac{d \cdot \tau^2}{2}, \frac{d}{2}\right) \cdot \text{Gamma}(\text{shape} = a_\tau, \text{rate} = b_\tau) \\
p(\phi_g \mid \cdots) &= \prod_{n=1}^N \text{Poisson}(y_{g,n} \mid \exp(c_n + \varepsilon_{g,n} + \mu(n, \phi_g, \alpha_g, \delta_g))) \cdot \text{N}(\phi_g \mid \theta_\phi, \sigma_\phi^2) \\
p(\theta_\phi \mid \cdots) &= \prod_{n=1}^N \prod_{g=1}^G \text{N}(\phi_g \mid \theta_\phi, \sigma_\phi^2) \cdot \text{N}(\theta_\phi \mid 0, \gamma_\phi^2) \\
p(\sigma_\phi \mid \cdots) &= \prod_{n=1}^N \prod_{g=1}^G \text{N}(\phi_g \mid \theta_\phi, \sigma_\phi^2) \cdot \text{U}(\sigma_\phi \mid 0, \sigma_{\phi 0}) \\
p(\alpha_g \mid \cdots) &= \prod_{k(n) \neq 2} \text{Poisson}(y_{g,n} \mid \exp(c_n + \varepsilon_{g,n} + \mu(n, \phi_g, \alpha_g, \delta_g))) \\
&\quad \times (\pi_\alpha \text{I}(\alpha_g = 0) + (1 - \pi_\alpha) \text{I}(\alpha_g \neq 0) \text{N}(\alpha_g \mid \theta_\alpha, \sigma_\alpha^2)) \\
p(\theta_\alpha \mid \cdots) &= \prod_{k(n) \neq 2} \prod_{g=1}^G (\pi_\alpha \text{I}(\alpha_g = 0) + (1 - \pi_\alpha) \text{I}(\alpha_g \neq 0) \text{N}(\alpha_g \mid \theta_\alpha, \sigma_\alpha^2)) \cdot \text{N}(\theta_\alpha \mid 0, \gamma_\alpha^2) \\
p(\sigma_\alpha \mid \cdots) &= \prod_{k(n) \neq 2} \prod_{g=1}^G (\pi_\alpha \text{I}(\alpha_g = 0) + (1 - \pi_\alpha) \text{I}(\alpha_g \neq 0) \text{N}(\alpha_g \mid \theta_\alpha, \sigma_\alpha^2)) \cdot \text{U}(\sigma_\alpha \mid 0, \sigma_{\alpha 0}) \\
p(\pi_\alpha \mid \cdots) &= \prod_{k(n) \neq 2} \prod_{g=1}^G (\pi_\alpha \text{I}(\alpha_g = 0) + (1 - \pi_\alpha) \text{I}(\alpha_g \neq 0) \text{N}(\alpha_g \mid \theta_\alpha, \sigma_\alpha^2)) \cdot \text{Beta}(\pi_\alpha \mid a_\alpha, b_\alpha) \\
p(\delta_g \mid \cdots) &= \prod_{k(n)=2} \text{Poisson}(y_{g,n} \mid \exp(c_n + \varepsilon_{g,n} + \mu(n, \phi_g, \delta_g, \delta_g))) \\
&\quad \times (\pi_\delta \text{I}(\delta_g = 0) + (1 - \pi_\delta) \text{I}(\delta_g \neq 0) \text{N}(\delta_g \mid \theta_\delta, \sigma_\delta^2)) \\
p(\theta_\delta \mid \cdots) &= \prod_{k(n)=2} \prod_{g=1}^G (\pi_\delta \text{I}(\delta_g = 0) + (1 - \pi_\delta) \text{I}(\delta_g \neq 0) \text{N}(\delta_g \mid \theta_\delta, \sigma_\delta^2)) \cdot \text{N}(\theta_\delta \mid 0, \gamma_\delta^2) \\
p(\sigma_\delta \mid \cdots) &= \prod_{k(n)=2} \prod_{g=1}^G (\pi_\delta \text{I}(\delta_g = 0) + (1 - \pi_\delta) \text{I}(\delta_g \neq 0) \text{N}(\delta_g \mid \theta_\delta, \sigma_\delta^2)) \cdot \text{U}(\sigma_\delta \mid 0, \sigma_{\delta 0}) \\
p(\pi_\delta \mid \cdots) &= \prod_{k(n)=2} \prod_{g=1}^G (\pi_\delta \text{I}(\delta_g = 0) + (1 - \pi_\delta) \text{I}(\delta_g \neq 0) \text{N}(\delta_g \mid \theta_\delta, \sigma_\delta^2)) \cdot \text{Beta}(\pi_\delta \mid a_\delta, b_\delta)
\end{aligned}$$