

Example Model:

$$y_i \sim N(\mu, \sigma^2), i = 1, \dots, n$$

$$\mu \sim N(\mu_0, \sigma_0^2) = (\frac{1}{2\pi\sigma_0^2})^{\frac{1}{2}} \exp\left\{-\frac{1}{2}(\mu - \mu_0)^2\right\}$$

$$\sigma^2 \sim \text{IG}(\nu, \tau) = \frac{1}{2\Gamma(\nu)} (\sigma^2)^{-(\nu+1)} e^{-\frac{\tau}{\sigma^2}}$$

Posterior:

$$[\mu, \sigma^2 | y] \propto \left(\prod_{i=1}^n [y_i | \mu, \sigma^2] \right) [\mu] [\sigma^2]$$

Full-conditionals:

$$[\mu | \cdot] \propto \left(\prod_{i=1}^n [y_i | \mu, \sigma^2] \right) [\mu]$$

$$\propto \left(\prod_{i=1}^n \exp\left\{-\frac{1}{2\sigma^2}(y_i - \mu)^2\right\} \right) \exp\left\{-\frac{1}{2\sigma_0^2}(\mu - \mu_0)^2\right\}$$

$$\propto \exp\left\{-\frac{1}{2\sigma^2}(\sum_{i=1}^n y_i^2 - 2\mu \sum_{i=1}^n y_i + n\mu^2)\right\} \exp\left\{-\frac{1}{2\sigma_0^2}(\mu^2 - 2\mu\mu_0 + \mu_0^2)\right\}$$

$$\propto \exp\left\{-\frac{1}{2} \underbrace{\left(-2\left(\frac{\sum y_i}{\sigma^2} + \frac{\mu_0}{\sigma_0^2}\right)\mu + \mu^2\left(\frac{n}{\sigma^2} + \frac{1}{\sigma_0^2}\right)\right)}_a\right\}$$

$$= N(a^{-1}b, a^{-1})$$

Why? Complete the square! Note:

$$\frac{(\mu - a^{-1}b)^2}{a^{-1}} = \frac{\mu^2}{a^{-1}} - \frac{2\mu a^{-1}b}{a^{-1}} + \frac{(a^{-1}b)^2}{a^{-1}} = a\mu^2 - 2\mu b + \underbrace{a^{-1}b^2}_{\text{constant}}$$

$$a^{-1}b^2 = \left(\frac{n}{\sigma^2} + \frac{1}{\sigma_0^2}\right)^{-1} \left(\frac{\sum y_i}{\sigma^2} + \frac{\mu_0}{\sigma_0^2}\right)^2$$

doesn't involve μ , so we can multiply $e^{a^{-1}b^2}$ by $[\mu | \cdot]$ proportionality and it will have a Gaussian form (i.e., conjugate)

$$[\sigma^2 | \cdot] \propto \left(\prod_{i=1}^n C_{y_i} | \mu, \sigma^2 \right) [\sigma^2]$$

$$\propto \left(\prod_{i=1}^n (\cancel{2\sigma^2})^{\frac{1}{2}} \exp\left\{-\frac{1}{2\sigma^2}(y_i - \mu)^2\right\} \right) \frac{1}{\tilde{r} \Gamma(\tilde{r})} (\sigma^2)^{-(\tilde{r}+1)} \exp\left\{-\frac{1}{\tilde{r}\sigma^2}\right\}$$

$$\propto (\sigma^2)^{\frac{\tilde{r}}{2}} \exp\left\{-\frac{1}{\sigma^2} \frac{\sum_i (y_i - \mu)^2}{2}\right\} (\sigma^2)^{-(\tilde{r}+1)} \exp\left\{-\frac{1}{\sigma^2} \cdot \frac{1}{\tilde{r}}\right\}$$

$$\propto (\sigma^2)^{-\left(\frac{\tilde{r}}{2} + \tilde{r} + 1\right)} \exp\left\{-\frac{1}{\sigma^2} \left(\underbrace{\frac{\sum_i (y_i - \mu)^2}{2}}_{\tilde{b}} + \underbrace{\frac{1}{\tilde{r}}}_{\tilde{r}^{-1}}\right)\right\}$$

$$= \mathcal{IG}(\tilde{b}, \tilde{r})$$

why? Because we can multiply by $\frac{1}{\tilde{r} \tilde{b} \Gamma(\tilde{r})}$

and $[\sigma^2 | \cdot]$ will take the form of another inverse gamma (i.e., conjugate)

MCMC algorithm:

1.) set $\mu^{(0)} = y$

2.) $k=1$

3.) Sample $\sigma^{2(k)} \sim [\sigma^{2(k)} | y, \mu^{(k-1)}]$

4.) Sample $\mu^{(k)} \sim [\mu^{(k)} | y, \sigma^{2(k)}]$

5.) $k = k+1$

6.) Goto 3, repeat for $k=1, \dots, K$

This is called a Gibbs sampler.