#### **Appendix**

The model joint probability

$$p(\mathbf{Y}, \mathbf{x}, \beta_i, \sigma, \Delta, s, w) = p(\mathbf{Y} \mid \mathbf{x}\beta_i, \sigma)p(\beta_i \mid \Delta, s)p(\Delta \mid 0, w)p(\sigma)p(w)p(s)$$

where the individual factors are:

$$p(\mathbf{Y} \mid \mathbf{x}\beta_i, \sigma) = \prod_{i=1}^{I} \prod_{c=1}^{C} \mathcal{N}(y_c \mid x_c\beta_i, \sigma^{-1})$$

$$p(\beta_i \mid \Delta, s) = \prod_{i=1}^{I} \mathcal{N}\left(\beta_i \mid \Delta, s^{-1}\right)$$

$$p(\Delta \mid 0, w) = \mathcal{N}(\Delta \mid 0, w^{-1})$$

$$p(\sigma) = \operatorname{Gamma}(\sigma \mid a_0, b_0)$$

$$p(s) = \operatorname{Gamma}(s \mid c_0, d_0)$$

$$p(w) = \operatorname{Gamma}(w \mid e_0, f_0)$$

We want to infer the posterior distibution:

$$q(\beta_i, \Delta, \sigma, s, w) = p(\beta_i, \Delta, \sigma, s, w \mid \mathbf{D})$$

we assume that the posterior distribution factorizes into independent factors:

$$q(\beta_i, \Delta, \sigma, s, w) = q(\beta_i)q(\Delta)q(\sigma)q(s)q(w)$$

The true posterior distribution does not in fact factor this way and we will obtain an approximation.

## Derivation of $q(\beta_i)$

$$\begin{split} & \ln q_{\beta_i}^*(\beta_i) = \mathbf{E}_{\Delta,\sigma,s} \left[ \ln p(\mathbf{Y} \mid \mathbf{x}\beta_i, \sigma) + \ln p(\beta_i \mid \Delta, s) \right] + C \\ & = \mathbf{E}_{\sigma} \left[ \sum_{c=1}^{C} \frac{1}{2} \ln \sigma - \frac{\sigma}{2} (y_c - x_c \beta_i)^2 \right] + \mathbf{E}_{\Delta,s} \left[ \frac{1}{2} \ln s - \frac{s}{2} (\Delta - \beta_i)^2 \right] + C \\ & = \mathbf{E}_{\sigma} \left[ \sum_{c=1}^{C} -\frac{\sigma}{2} (y_c - x_c)^2 \right] + \mathbf{E}_{\Delta,s} \left[ -\frac{s}{2} (\Delta - \beta_i)^2 \right] + C \\ & = \mathbf{E}_{\sigma} \left[ -\frac{\sigma}{2} \sum_{c=1}^{C} y_c^2 + \sigma \beta_i \sum_{c=1}^{C} y_c x_c - \frac{\sigma \beta_i^2 \sum_{c=1}^{C} x_c^2}{2} \right] + \mathbf{E}_{\Delta,s} \left[ -\frac{s\Delta^2}{2} + s\Delta \beta_i - \frac{s\beta_i^2}{2} \right] + C \\ & = \mathbf{E}_{\sigma} \left[ \sigma \beta_i \sum_{c=1}^{C} y_c x_c - \frac{\sigma \beta_i^2 \sum_{c=1}^{C} x_c^2}{2} \right] + \mathbf{E}_{\Delta,s} \left[ s\Delta \beta_i - \frac{s\beta_i^2}{2} \right] + C \\ & = (\mathbf{E}_{\sigma} \left[ \sigma \right] \sum_{c=1}^{C} x_c y_c + \mathbf{E}_s \left[ s \right] \mathbf{E}_{\Delta} \left[ \Delta \right]) \beta_i - \frac{1}{2} (\mathbf{E}_{\sigma} \left[ \sigma \right] \sum_{c=1}^{C} x_c^2 + \mathbf{E}_s \left[ s \right]) \beta_i^2 + C \end{split}$$

$$\lambda_{\beta,i} = \frac{a_n}{b_n} \sum_{c=1}^C x_c^2 + \frac{c_n}{d_n}$$
$$\beta_i = \frac{\frac{a_n}{b_n} \sum_{c}^C x_c y_c + \frac{c_n}{d_n} \Delta}{\lambda_{\beta,i}}$$

#### Derivation of $q(\Delta)$

$$\begin{split} & \ln q_{\Delta}^{*}(\Delta) = \mathbf{E}_{\beta_{i},s} \left[ \ln p(\beta_{i} \mid \Delta, s) + \ln p(\Delta \mid \mu_{0}, w) \right] + C \\ & = \mathbf{E}_{\beta_{i},s} \left[ \sum_{i=1}^{I} \frac{1}{2} \ln s - \frac{s}{2} (\Delta - \beta_{i})^{2} \right] + \mathbf{E}_{w} \left[ \frac{1}{2} \ln w - \frac{w}{2} (\Delta - \mu_{0})^{2} \right] + C \\ & = \mathbf{E}_{\beta_{i},s} \left[ \sum_{i=1}^{I} -\frac{s}{2} (\Delta - \beta_{i})^{2} \right] + \mathbf{E}_{w} \left[ -\frac{w}{2} (\Delta - \mu_{0})^{2} \right] + C \\ & = \mathbf{E}_{\beta_{i},s} \left[ -\frac{Is\Delta^{2}}{2} + s\Delta \sum_{i=1}^{I} \beta_{i} \right] + \mathbf{E}_{w} \left[ w\Delta - \frac{w\Delta^{2}}{2} \right] + C \\ & = \mathbf{E}_{\beta_{i},s} \left[ -\frac{1}{2} (Is + w)\Delta^{2} \right] + \mathbf{E}_{w,s,\beta_{i}} \left[ \Delta (s\sum_{i}^{I} \beta_{i} + w\mu_{0}) \right] + C \end{split}$$

$$\lambda_{\Delta} = I \frac{c_n}{d_n} + \frac{e_n}{f_n}$$
$$\Delta = \frac{\frac{c_n}{d_n} \sum_{i}^{I} \beta_i + \frac{e_n}{f_n} \mu_0}{\lambda_{\Delta}}$$

## Derivation of $q(\sigma)$

$$\begin{split} & \ln q_{\sigma}^{*}(\sigma) = \mathbf{E}_{\beta_{i}} \left[ \ln p(\mathbf{Y} \mid \mathbf{x}\beta_{i}, \sigma) + \ln p(\sigma \mid a, b) \right] + C \\ & = \mathbf{E}_{\beta_{i}} \left[ \sum_{i=1}^{I} \sum_{c=1}^{C} \frac{1}{2} \ln \sigma - \frac{\sigma}{2} (y_{i,c} - \mathbf{x}\beta_{i})^{2} \right] + \left[ (a-1) \ln \sigma - b\sigma \right] + C \\ & = \mathbf{E}_{\beta_{i}} \left[ -\frac{\sigma}{2} \sum_{i=1}^{I} \sum_{c=1}^{C} (y_{i,c} - \mathbf{x}\beta_{i})^{2} \right] + (a-1) \ln \sigma - b\sigma + \frac{\sum_{i=1}^{I} \sum_{c=1}^{C} 1}{2} \ln \sigma + C \\ & = \mathbf{E}_{\beta_{i}} \left[ -\frac{\sigma}{2} \sum_{i=1}^{I} \sum_{c=1}^{C} y_{i,c}^{2} - 2y_{i,c} \mathbf{x}\beta_{i} + \mathbf{x}\beta_{i}^{2} \right] + \left( (a-1) + \frac{\sum_{i=1}^{I} \sum_{c=1}^{C} 1}{2} \right) \ln \sigma - b\sigma + C \\ & = \mathbf{E}_{\beta_{i}} \left[ -\left( (\sum_{i=1}^{I} \mathbf{x}\beta_{i}^{2} - 2\sum_{i=1}^{I} \sum_{c=1}^{C} y_{i,c} \mathbf{x}\beta_{i} + \sum_{i=1}^{I} \sum_{c=1}^{C} y_{i,c}^{2} \right) \frac{1}{2} + b)\sigma \right] + \left( (a-1) + \frac{\sum_{i=1}^{I} \sum_{c=1}^{C} 1}{2} \right) \ln \sigma + C \end{split}$$

$$N = \sum_{i=1}^{I} \sum_{c=1}^{C} 1$$

$$a_n = a + \frac{N}{2}$$

$$b_n = b + \frac{1}{2} \left( \sum_{i=1}^{I} \sum_{c=1}^{C} (y_{i,c} - \mathbf{x}\beta_i)^2 + \sum_{i=1}^{I} trace(\lambda_{\beta,i} X_i^T X_i) \right)$$

### Derivation of q(s)

$$\begin{split} & \ln q_s^*(s) = \mathbf{E}_{\beta_i,\Delta} \left[ \ln p(\beta_i \mid \Delta, s) + \ln p(s \mid c, d) \right] + C \\ & = \mathbf{E}_{\beta_i,\Delta} \left[ \sum_{i=1}^{I} \sum_{d=1}^{D} \frac{1}{2} \ln s - \frac{s}{2} (\beta_{i,d} - \Delta_d)^2 \right] + \left[ (c-1) \ln s - ds \right] + C \\ & = \mathbf{E}_{\beta_i,\Delta} \left[ \sum_{i=1}^{I} \sum_{d=1}^{D} -\frac{s}{2} (\beta_{i,d} - \Delta_d)^2 \right] + \left( (c-1) + \frac{\sum_{i=1}^{I} \sum_{d=1}^{D} 1}{2} \right) \ln s - ds + C \\ & = \mathbf{E}_{\beta_i,\Delta} \left[ - \left( (\sum_{i=1}^{I} \sum_{d=1}^{D} \beta_{i,d}^2 - 2 \sum_{i=1}^{I} \sum_{d=1}^{D} \beta_{i,d} \Delta_d + \sum_{d=1}^{D} \Delta_d^2 \right) \frac{1}{2} + d \right) s \right] + \left( (c-1) + \frac{ID}{2} \right) \ln s + C \end{split}$$

$$\begin{aligned} c_n &= c + \frac{DI}{2} \\ d_n &= d + \frac{1}{2} (\sum_{i=1}^{I} (\beta_i - \delta)^2 + trace(\lambda_{\Delta}) + \sum_{i=1}^{I} trace(\lambda_{\beta,i})) \end{aligned}$$

# Derivation of q(w)

$$\begin{split} \ln q_w^*(w) &= \mathcal{E}_\Delta \left[ \ln p(\Delta \mid 0, w) + \ln p(w \mid e, f) \right] + C \\ &= \mathcal{E}_\Delta \left[ \sum_{d=1}^D \frac{1}{2} \ln w - \frac{w}{2} (\delta_d)^2 \right] + \left[ (e-1) \ln w - f w \right] + C \\ &= \mathcal{E}_\Delta \left[ \sum_{d=1}^D -\frac{w}{2} (\delta_d)^2 \right] + \left( (e-1) + \frac{D}{2} \right) \ln w - f w + C \\ &= \mathcal{E}_\Delta \left[ -w (\frac{1}{2} \sum_{d=1}^D (-\delta_d)^2 + f) \right] + \left( (e-1) + \frac{D}{2} \right) \ln w + C \end{split}$$

$$e_n = e + \frac{D}{2}$$
  
$$f_n = f + \frac{1}{2}(\Delta^T \Delta + trace(\lambda_{\Delta}))$$