

# Derivative of the posterior of the gestalt model

## 1 Rules of differentiation

Assuming that  $y$  and  $a$  are vectors and  $M$  is a symmetric matrix of appropriate dimension, and  $f$  is a scalar function, and  $s$  is a scalar variable.

$$\frac{\partial}{\partial y} y^T M y = 2M y \quad (1)$$

$$\frac{\partial}{\partial y} a^T y = a \quad (2)$$

$$\frac{\partial}{\partial M} y^T M^{-1} y = -M^{-1} y y^T M^{-1} \quad (3)$$

$$\frac{\partial}{\partial M} \log \det M = M^{-1} \quad (4)$$

$$\frac{\partial}{\partial s} f(M(s)) = \text{Tr} \left[ \frac{\partial f}{\partial M} \frac{\partial M}{\partial s} \right] \quad (5)$$

$$\frac{\partial}{\partial s} s M = M \quad (6)$$

## 2 Form of the full posterior

$$p(v, g, z | x) = p(x | v, g, z) p(v | g) p(g) p(z) \frac{1}{p(x)} \sim p(x | v, z) p(v | g) p(g) p(z) \quad (7)$$

so the log-posterior will be the following, up to an additive constant, using Gamma priors over  $g$  and  $z$  defined by shape and scale parameters:

$$\begin{aligned} \log p(v, g, z | x) &\sim \log p(x | v, z) + \log p(v | g) + \log p(g) + \log p(z) = \\ &= \log \mathcal{N}(x; z A v, \sigma_x I) + \log \mathcal{N}(v; 0, C_v) + \log \text{Gam}(g; sh_g, sc_g) + \log \text{Gam}(z; sh_z, sc_z) \end{aligned} \quad (8)$$

logarithms of the used pdfs look as follows:

$$\log \mathcal{N}(y; \mu, C) = -\frac{1}{2} [\log(2\pi) + \log \det(C) + (y - \mu)^T C^{-1} (y - \mu)] \quad (9)$$

$$\log \text{Gam}(y; sh, sc) = \log(1) - \log(\Gamma(sh)) - sh \log(sc) + (sh - 1) \log(y) - \frac{y}{sc} \quad (10)$$

so discarding all terms that are constant w.r.t. all three variables, the log-posterior is composed as follows:

$$\begin{aligned} \log p(v, g, z | x) \sim & -\frac{1}{2\sigma_x} (x - zAv)^T (x - zAv) - \\ & -\frac{1}{2} [\log(\det(C_v)) + v^T C_v^{-1} v] + \\ & + \sum_{j=1}^K \left[ (sh_g - 1) \log(g_j) - \frac{g_j}{sc_g} \right] + (sh_z - 1) \log(z) - \frac{z}{sc_z} \end{aligned} \quad (11)$$

expanding the quadratic form in the first term

$$\begin{aligned} (x - zAv)^T (x - zAv) &= x^T x - zv^T A^T x + z^2 v^T A^T Av - zx^T Av = \\ &= x^T x - 2zx^T Av + z^2 v^T A^T Av \end{aligned} \quad (12)$$

as  $zx^T Av$  is a scalar, thus equal to its transpose. Discarding the term not dependent on any variables of the posterior we get

$$\begin{aligned} \log p(v, g, z | x) \sim & -\frac{z}{2\sigma_x} (zv^T A^T Av - 2x^T Av) - \\ & -\frac{1}{2} [\log(\det(C_v)) + v^T C_v^{-1} v] + \\ & + \sum_{j=1}^K \left[ (sh_g - 1) \log(g_j) - \frac{g_j}{sc_g} \right] + \\ & + (sh_z - 1) \log(z) - \frac{z}{sc_z} \end{aligned} \quad (13)$$

### 3 Derivative in $v$

$$\log p(v, g, z | x) \sim -\frac{1}{2} \left[ \frac{z^2}{\sigma_x} v^T A^T Av - \frac{2z}{\sigma_x} x^T Av + v^T C_v^{-1} v \right] + f_1(g, z) \quad (14)$$

lumping the two quadratic forms together

$$\log p(v, g, z | x) \sim \frac{z}{\sigma_x} x^T Av - \frac{1}{2} v^T \left[ \frac{z^2}{\sigma_x} A^T A + C_v^{-1} \right] v + f_1(g, z) \quad (15)$$

Taking the derivative using Eq. 1 and 2 we get

$$\frac{\partial}{\partial v} \log p(v, g, z | x) = \frac{z}{\sigma_x} A^T x - \left[ \frac{z^2}{\sigma_x} A^T A + C_v^{-1} \right] v \quad (16)$$

#### 4 Derivative in $g$

$$\begin{aligned} \log p(v, g, z | x) \sim & -\frac{1}{2} [\log \det(C_v) + v^T C_v^{-1} v] + \\ & + \sum_{j=1}^K \left[ (sh_g - 1) \log(g_j) - \frac{g_j}{sc_g} \right] + f_2(v, z) \end{aligned} \quad (17)$$

Taking the derivative w.r.t. a single  $g_i$  using Eq. 5 we get

$$\begin{aligned} \frac{\partial}{\partial g_i} \log p(v, g, z | x) = & -\frac{1}{2} \text{Tr} \left[ \frac{\partial}{\partial C_v} [\log \det(C_v) + v^T C_v^{-1} v] \frac{\partial C_v}{\partial g_i} \right] + \\ & + \frac{\partial}{\partial g_i} \left[ (sh_g - 1) \log(g_i) - \frac{g_i}{sc_g} \right] \end{aligned} \quad (18)$$

using Eq. 4, 3 and 6 we arrive to

$$\frac{\partial}{\partial g_i} \log p(v, g, z | x) = -\frac{1}{2} \text{Tr} \left[ [C_v^{-1} - C_v^{-1} v v^T C_v^{-1}] C_i \right] + \frac{sh_g - 1}{g_i} - \frac{1}{sc_g} \quad (19)$$

#### 5 Derivative in $z$

$$\log p(v, g, z | x) \sim -\frac{z}{2\sigma_x} (z v^T A^T A v - 2x^T A v) + (sh_z - 1) \log(z) - \frac{z}{sc_z} + f_3(g, v) \quad (20)$$

$$\frac{\partial}{\partial z} \log p(v, g, z | x) = \frac{1}{\sigma_x} [x^T A v - z v^T A^T A v] + \frac{sh_z - 1}{z} - \frac{1}{sc_z} \quad (21)$$