Model selection for the gestalt model

1 The likelihood function

In order to evaluate the log-likelihood for cross validation steps, we have to use an approximation.

$$p(X \mid C_{1..k}) = \prod_{n=1}^{N} \iiint_{-\infty}^{\infty} p(x_n \mid v_n, z_n) p(v_n \mid g_n) p(g_n) p(z_n) dv_n dg_n dz_n =$$

$$= \prod_{n=1}^{N} \iint_{-\infty}^{\infty} p(g_n) p(z_n) \int_{-\infty}^{\infty} p(x_n \mid v_n, z_n) p(v_n \mid g_n) dv_n dg_n dz_n$$
(1)

the product of the two conditionals can be rewritten as a Gaussian over v times another Gaussian

$$p(x \mid v, z)p(v \mid g) = \frac{1}{\sqrt{\det(A^T A)}} \frac{1}{z^{D_v}} \mathcal{N}(\frac{1}{z} A^+ x; 0, \frac{\sigma_x}{z^2} (A^T A)^{-1} + C_v) \mathcal{N}(v; \mu_c, C_c)$$
(2)

where v only appears in the second Gaussian, so integrating this formula will set that term to 1 and leave everything else as it is.

$$p(X \mid C_{1..k}) = \det(A^T A)^{-\frac{N}{2}} \prod_{n=1}^{N} \int_0^\infty p(z) \frac{1}{z^{D_v}} \int_0^\infty p(g) \mathcal{N}(\frac{1}{z} A^+ x; 0, \frac{\sigma_x}{z^2} (A^T A)^{-1} + C_v) \mathrm{d}g \mathrm{d}z$$
(3)

2 Approximation with sampling

The integrals may be approximated by averaging over samples from p(g) and p(z), leaving out terms constant in g, z and x

$$p(X \mid C_{1..k}) \sim \prod_{n=1}^{N} \frac{1}{L^2} \sum_{l_1=1}^{L} \frac{1}{z_{l_1}^{D_v}} \sum_{l_2=1}^{L} \mathcal{N}(\frac{1}{z_{l_1}} A^+ x_n; 0, \frac{\sigma_x}{z_{l_1}^2} (A^T A)^{-1} + C_v^{l_2})$$
 (4)

as z and g are independent, we can choose a large enough L and merge the sums, and bring the division by L outside of the product

$$p(X \mid C_{1..k}) \sim \frac{1}{L^N} \prod_{n=1}^N \sum_{l=1}^L \frac{1}{z_{l_1}^{D_v}} \mathcal{N}(\frac{1}{z_l} A^+ x_n; 0, \frac{\sigma_x}{z_l^2} (A^T A)^{-1} + C_v^l)$$
 (5)

taking the logarithm we get

$$\log p(X \mid C_{1..k}) \sim -N \log(L) \sum_{n=1}^{N} \log \left[\sum_{l=1}^{L} \frac{1}{z_{l_1}^{D_v}} \mathcal{N}(\frac{1}{z_l} A^+ x_n; 0, \frac{\sigma_x}{z_l^2} (A^T A)^{-1} + C_v^l) \right]$$
(6)