current model $f_k(x)$, k = 1, ..., K, with $\sum_{k=1}^{K} f_k(x) = 0$ (see (10.21) in Section 10.6). We wish to update the model for observations in a region h in predictor space, by adding constants $f_{\nu}(x) + \gamma_{\nu}$, with $\gamma_{\nu} = 0$. (a) Write down the multinomial log-likelihood for this problem, and its first and second derivatives.

(b) Using only the diagonal of the Hessian matrix in (1), and starting
from γ_k = 0 ∀k, show that a one-step approximate Newton update
for γ_k is
$$\gamma_k^1 = \sum_{x_k \in R} (y_{ik} - p_{ik}), \quad k = 1, ..., K - 1, \quad (10.57)$$

Ex. 10.8 Consider a K-class problem where the targets up are coded as 1 if observation i is in class k and zero otherwise. Suppose we have a

(10.57)where $p_{ik} = \exp(f_k(x_i)) / (\sum_{\ell=1}^{K} f_{\ell}(x_i))$.

(c) We prefer our update to sum to zero, as the current model does. Using symmetry arguments, show that

 $\hat{\gamma}_k = \frac{K-1}{K} (\gamma_k^1 - \frac{1}{K} \sum_{\ell}^K \gamma_{\ell}^1), k = 1, ..., K$ (10.58)is an appropriate update, where γ_L^1 is defined as in (10.57) for all

k = 1, ..., K.