

1. Let

$$x(n)=s(n)+v(n); \quad y(n)=\sum_{k=0}^{N-1} h_n(k)x(n-k)=\vec{h}_n^T \vec{x}_n$$

where

$$\vec{h}_n = [h_n(0) \quad \cdots \quad h_n(N-1)]^T, \quad \vec{x}_n = [x(n) \quad \cdots \quad x(n-N+1)]^T$$

prove that

$$f(\vec{h}_n) = E[(y(n) - s(n))^2] = \vec{h}_n^T R_x \vec{h}_n - 2\vec{p}^T \vec{h}_n + E[s^2(n)]$$

where

$$R_x = \begin{bmatrix} R_x(0) & \cdots & R_x(N-1) \\ \vdots & \ddots & \vdots \\ R_x(N-1) & \cdots & R_x(0) \end{bmatrix}, \quad \vec{p} = \begin{bmatrix} R_{xs}(0) \\ \vdots \\ R_{xs}(N-1) \end{bmatrix}$$

2. Suppose the random discrete signal $x(n)$ is modeled by ARMA, i.e.,

$$x(n) = -\sum_{i=1}^N \alpha_i x(n-i) + \sum_{j=0}^M \beta_j u(n-j)$$

prove that

$$R_x(m) = E[x(n)x(n+m)] = -\sum_{i=1}^N \alpha_i R_x(m-i) + \sigma^2 \sum_{j=0}^M \beta_j h(j-m)$$

3. Let

$$E = \sum_{i=1}^{N_{out}} (y_i - d_i)^2$$

where

$$y_i = f_{neu}(\hat{y}_i), \quad \hat{y}_i = \sum_{j=1}^{N_K} \tilde{w}_{ij}^{out} h_j^K + \tilde{b}_i^{out}$$

$$h_j^K = f_{neu}(\hat{h}_j^K), \quad \hat{h}_j^K = \sum_{t=1}^{N_{K-1}} w_{jt}^K h_t^{K-1} + b_j^K$$

prove that

$$\frac{\partial E}{\partial w_{pq}^K} = 2 \sum_{i=1}^{N_{out}} (y_i - d_i) f'_{neu}(\hat{y}_i) \tilde{w}_{ip}^{out} f'_{neu}(\hat{h}_p^K) h_q^{K-1}$$

4. Let

$$N(x|\mu_1, \Sigma_1) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma_1|^{\frac{1}{2}}} e^{-\frac{(x-\mu_1)^T \Sigma_1^{-1} (x-\mu_1)}{2}}, \quad N(x|\mu_2, \Sigma_2) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma_2|^{\frac{1}{2}}} e^{-\frac{(x-\mu_2)^T \Sigma_2^{-1} (x-\mu_2)}{2}}$$

Prove that

$$N(x|\mu_1, \Sigma_1) N(x|\mu_2, \Sigma_2) = C \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma_0|^{\frac{1}{2}}} e^{-\frac{(x-\mu_0)^T \Sigma_0^{-1} (x-\mu_0)}{2}}$$

where

$$\Sigma_0 = (\Sigma_1^{-1} + \Sigma_2^{-1})^{-1}; \quad \mu_0 = \Sigma_0 (\Sigma_1^{-1} \mu_1 + \Sigma_2^{-1} \mu_2)$$

or further

$$\Sigma_0 = \Sigma_1 - \Sigma_1 (\Sigma_1 + \Sigma_2)^{-1} \Sigma_1; \quad \mu_0 = \mu_1 + \Sigma_1 (\Sigma_1 + \Sigma_2)^{-1} (\mu_2 - \mu_1)$$