

$$(7.81) \quad p(w|t, X, \alpha, \beta) = N(w|m, \Sigma)$$

$$(7.82) \quad m = \beta \Sigma \bar{X}^T t$$

$$(7.83) \quad \Sigma = (A + \beta \bar{X}^T \bar{X})^{-1}$$

$$(7.79) \quad p(t|X, w, \beta) = \prod_{n=1}^N p(t_n|x_n, w, \beta)$$

$$(7.80) \quad p(w|\alpha) = \prod_{i=1}^M N(w_i|0, \alpha_i)$$

$$(7.76) \quad p(t|x, w, \beta) = N(t|y(x), \beta^{-1})$$

$$(7.77) \quad y(x) = \sum_{i=1}^M w_i \phi(x)_i$$

7.97

$$\begin{aligned} E_w[\ln p(t|x, w, \beta) p(w|\alpha)] &= \int p(w|t) \ln p(t|x, w, \beta) p(w|\alpha) dw \\ &= \int p(w|t) \ln \left( \prod_{n=1}^N N(t_n|y_n, \beta^{-1}) \prod_{i=1}^M N(w_i|0, \alpha_i) \right) dw \\ &= \int p(w|t) \left\{ \sum_n \ln N(t_n|y_n, \beta^{-1}) + \sum_i \ln N(w_i|0, \alpha_i) \right\} dw \\ &= \int p(w|t) \left[ \sum_n \left\{ \frac{1}{2} \ln \frac{\beta}{2\pi} - \frac{1}{2} \beta (t_n - y_n)^2 \right\} + \sum_i \left\{ \frac{1}{2} \ln \frac{\alpha_i}{2\pi} - \frac{1}{2} \alpha_i w_i^2 \right\} \right] dw \\ &= \frac{N}{2} \ln \frac{\beta}{2\pi} - \frac{\beta}{2} \sum_n E_w[(t_n - y_n)^2] + \sum_i \frac{1}{2} \ln \frac{\alpha_i}{2\pi} - \frac{1}{2} \sum_i \alpha_i E_w[w_i^2] \quad \dots \textcircled{1} \end{aligned}$$

7.98

$$E_w[w_i] = (E_w[w])_i = \left( \int p(w|t) w dw \right)_i = \left( \int N(w|m, \Sigma) w dw \right)_i = m_i$$

$$E_w[w_i w_j] = (E_w[w w^T])_{ij} = \left( \int p(w|t) w w^T dw \right)_{ij} = \left( \int N(w|m, \Sigma) w w^T dw \right)_{ij} = (m m^T + \Sigma)_{ij}$$

$$E_w[y_n] = E_w\left[\sum_i w_i \phi_n(x)_i\right] = \sum_i E_w[w_i] \phi_n(x)_i = \sum_i m_i \phi_n(x)_i = m^T \phi_n$$

$$E_w[y_n^2] = E_w\left[\left(\sum_i w_i \phi_n(x)_i\right)^2\right] = E_w\left[\sum_i \sum_j w_i w_j \phi_n(x)_i \phi_n(x)_j\right] = \sum_i \sum_j E_w[w_i w_j] \phi_n(x)_i \phi_n(x)_j = \sum_i \sum_j (m m^T + \Sigma)_{ij} \phi_n(x)_i \phi_n(x)_j$$

$$\begin{aligned} &= \sum_i \sum_j (m m^T)_{ij} \phi_n(x)_i \phi_n(x)_j + \sum_i \sum_j \Sigma_{ij} \phi_n(x)_i \phi_n(x)_j \\ &= \phi_n^T m m^T \phi_n + \phi_n^T \Sigma \phi_n \quad \left\{ \begin{array}{l} \phi_n^T m m^T \phi_n = (\phi_n, \phi_n) \begin{pmatrix} m_1 \\ \vdots \\ m_M \end{pmatrix} \begin{pmatrix} m_1 \\ \vdots \\ m_M \end{pmatrix} (\phi_n) \\ = (\phi_n, m_1 + \phi_n m_2) (m_1 \phi_n + m_2 \phi_n) \\ = m_1 m_1 \phi_n \phi_n + m_1 m_2 \phi_n \phi_n + m_2 m_1 \phi_n \phi_n + m_2 m_2 \phi_n \phi_n \\ \phi_n^T \Sigma \phi_n = (\phi_n, \phi_n) \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix} \begin{pmatrix} \phi_n \\ \phi_n \end{pmatrix} = (\phi_n, \phi_n) \begin{pmatrix} \Sigma_{11} \phi_n + \Sigma_{12} \phi_n \\ \Sigma_{21} \phi_n + \Sigma_{22} \phi_n \end{pmatrix} \\ = \Sigma_{11} \phi_n \phi_n + \Sigma_{12} \phi_n \phi_n + \Sigma_{21} \phi_n \phi_n + \Sigma_{22} \phi_n \phi_n \end{array} \right. \end{aligned}$$

$$E_w[(t_n - y_n)^2] = E_w[t_n^2 - 2 t_n y_n + y_n^2] = t_n^2 - 2 t_n m^T \phi_n + \phi_n^T m m^T \phi_n + \phi_n^T \Sigma \phi_n$$

$$= (t_n - m^T \phi_n)^2 + \phi_n^T \Sigma \phi_n$$

7.99

①  $E$   $\alpha$  で微分して 0 とおく

$$\frac{\partial E}{\partial \alpha_i} = \frac{1}{2} \frac{1}{\alpha_i} - \frac{1}{2} E w_i^2 = 0$$

$$\therefore \frac{1}{\alpha_i} - (m m^T + \Sigma)_{ii} = 0$$

$$\therefore \alpha_i = \frac{1}{m_i^2 + \Sigma_{ii}}$$

とすれば、(9.67) を得る。

②  $E$   $\beta$  で微分して 0 とおく

$$\frac{\partial E}{\partial \beta} = \frac{N}{2} \frac{1}{\beta} - \frac{1}{2} \sum_n E w_n^2 = 0$$

$$\therefore \frac{1}{\beta} = \frac{1}{N} \sum_n \{ (t_n - \phi_n^T m)^2 + \phi_n^T \Sigma \phi_n \}$$

$$= \frac{1}{N} \{ \|t - \Phi m\|^2 + \text{Tr}(\Phi \Sigma \Phi^T) \}$$

$$= \frac{1}{N} \{ \|t - \Phi m\|^2 + \frac{1}{\beta} \sum_i (1 - \alpha_i \Sigma_{ii}) \}$$

とすれば、(9.68) を得る。

$$\begin{aligned} \Phi &= \begin{pmatrix} \phi_1^T \\ \vdots \\ \phi_n^T \end{pmatrix} \text{ (1.16) } \Rightarrow \Phi m = \begin{pmatrix} \phi_1^T m \\ \vdots \\ \phi_n^T m \end{pmatrix} \\ \therefore (t - \Phi m)^T (t - \Phi m) &= \sum_n (t_n - \phi_n^T m)^2 \\ \Phi \Sigma \Phi^T &= \begin{pmatrix} \phi_1^T \\ \vdots \\ \phi_n^T \end{pmatrix} \Sigma \begin{pmatrix} \phi_1 \\ \vdots \\ \phi_n \end{pmatrix} = \begin{pmatrix} \phi_1^T \Sigma \phi_1 & \phi_1^T \Sigma \phi_2 \\ \vdots & \vdots \\ \phi_n^T \Sigma \phi_1 & \phi_n^T \Sigma \phi_n \end{pmatrix} \\ \therefore \text{Tr}(\Phi \Sigma \Phi^T) &= \phi_1^T \Sigma \phi_1 + \phi_2^T \Sigma \phi_2 + \dots = \sum_n \phi_n^T \Sigma \phi_n \end{aligned}$$

$$(7.83) \quad \Sigma = (A + \beta \Phi^T \Phi)^{-1}$$

より

$$(A + \beta \Phi^T \Phi) \Sigma = I$$

$$\therefore \Phi^T \Phi \Sigma = \frac{1}{\beta} (I - A \Sigma)$$

$$\therefore \text{Tr}(\Phi^T \Phi \Sigma) = \frac{1}{\beta} \text{Tr}(I - A \Sigma) = \frac{1}{\beta} \sum_i (1 - \alpha_i \Sigma_{ii})$$

より

$$\text{Tr}(\Phi \Sigma \Phi^T) \stackrel{(c.1)}{=} \text{Tr}(\Phi^T \Phi \Sigma) = \frac{1}{\beta} \sum_i (1 - \alpha_i \Sigma_{ii})$$