

3.19

$$\int \exp \{-E(\mathbf{w})\} d\mathbf{w}$$

Applying the result from 3.80, this becomes:

$$\begin{aligned} &= \int \exp \left\{ - \left(E(\mathbf{m}_N) + \frac{1}{2}(\mathbf{w} - \mathbf{m}_N)^T \mathbf{A}(\mathbf{w} - \mathbf{m}_N) \right) \right\} d\mathbf{w} \\ &= \int \exp \{-E(\mathbf{m}_N)\} \exp \left\{ -\frac{1}{2}(\mathbf{w} - \mathbf{m}_N)^T \mathbf{A}(\mathbf{w} - \mathbf{m}_N) \right\} d\mathbf{w} \\ &= \exp \{-E(\mathbf{m}_N)\} \int \exp \left\{ -\frac{1}{2}(\mathbf{w} - \mathbf{m}_N)^T \mathbf{A}(\mathbf{w} - \mathbf{m}_N) \right\} d\mathbf{w} \\ &= \exp \{-E(\mathbf{m}_N)\} \int \frac{(2\pi)^{M/2} |\mathbf{A}^{-1}|^{1/2}}{(2\pi)^{M/2} |\mathbf{A}^{-1}|^{1/2}} \exp \left\{ -\frac{1}{2}(\mathbf{w} - \mathbf{m}_N)^T \mathbf{A}(\mathbf{w} - \mathbf{m}_N) \right\} d\mathbf{w} \\ &= \exp \{-E(\mathbf{m}_N)\} (2\pi)^{M/2} |\mathbf{A}^{-1}|^{1/2} \int \frac{1}{(2\pi)^{M/2} |\mathbf{A}^{-1}|^{1/2}} \exp \left\{ -\frac{1}{2}(\mathbf{w} - \mathbf{m}_N)^T \mathbf{A}(\mathbf{w} - \mathbf{m}_N) \right\} d\mathbf{w} \\ &= \exp \{-E(\mathbf{m}_N)\} (2\pi)^{M/2} |\mathbf{A}^{-1}|^{1/2} \int \mathcal{N}(\mathbf{w}|\mathbf{m}_N, \mathbf{A}^{-1}) d\mathbf{w} \\ &= \exp \{-E(\mathbf{m}_N)\} (2\pi)^{M/2} |\mathbf{A}^{-1}|^{1/2} (1) \\ &= \exp \{-E(\mathbf{m}_N)\} (2\pi)^{M/2} |\mathbf{A}|^{-1/2} \end{aligned}$$

which is the same 3.85.

Now, using 3.78 and this result, the marginal likelihood becomes:

$$\begin{aligned} p(\mathbf{t}|\alpha, \beta) &= \left(\frac{\beta}{2\pi} \right)^{N/2} \left(\frac{\alpha}{2\pi} \right)^{M/2} \exp \{-E(\mathbf{m}_N)\} (2\pi)^{M/2} |\mathbf{A}|^{-1/2} \\ &= \left(\frac{\beta}{2\pi} \right)^{N/2} \alpha^{M/2} \exp \{-E(\mathbf{m}_N)\} |\mathbf{A}|^{-1/2} \end{aligned}$$

The log-likelihood becomes:

$$\ln p(\mathbf{t}|\alpha, \beta) = \frac{N}{2} \ln \beta - \frac{N}{2} \ln(2\pi) + \frac{M}{2} \ln \alpha - E(\mathbf{m}_N) - \frac{1}{2} \ln |\mathbf{A}|$$

which is the same as 3.86.