

Q, we have from (2.11.5) and (2.11.6)

$$p(y) = N(y | A\mu + b, L^{-1} + A\Lambda^{-1}A^T)$$
$$p(x|y) = N(x | \Sigma [A^T L(y-b) + A\mu], \Sigma)$$

$$\Sigma = (\Lambda + A^T L A)^{-1}$$

The predictive distribution is given by:-

$$p(t|t, \alpha, \beta) = \int p(t|w, \beta) p(w|t, \alpha, \beta) dw$$

Substituting value of $p(y)$ in $p(t|t, \alpha, \beta)$:-

$$p(t|t, \alpha, \beta) = \int N(t | A\mu + b, L^{-1} + A\Lambda^{-1}A^T) \\ N(t | \Sigma [A^T L(y-b) + A\mu], \Sigma)$$

$$= \int N(t | \Sigma [A^T L(y-b) + A\mu], \Sigma) dw$$

We see $m_N = \sum \{A^T L(y, b) + A\mu\}$

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

$$p(t|x, w, \beta) = \prod_{n=1}^N \mathcal{N}(t_n | w^T \phi(x_n), \beta^{-1})$$

$$p(w) \sim \mathcal{N}(w|m_0, \Sigma_0)$$

$$p(t|x, w, \beta) = \prod_{n=1}^N \mathcal{N}(t_n | w^T \phi(x_n), \beta^{-1})$$

Let $p(w) = \mathcal{N}(w|\mu, \Lambda^{-1})$

$$p(t|w) = \mathcal{N}(t|Aw+b, L^{-1})$$

$$p(t) = \mathcal{N}(t|A\mu+b, L^{-1}+A\Lambda^{-1}A^T)$$

$$p(w|t) = \mathcal{N}(w|\sum \{A^T L(t-b) + \Lambda\mu\}, \Sigma)$$

$$p(w) = \mathcal{N}(w|m_0, \Sigma_0)$$

$$p(t|x, w, \beta) = \prod_{n=1}^N \mathcal{N}(t_n | w^T \phi(x_n), \beta^{-1})$$

$$p(w|t) = p(w|m_N, \Sigma_N)$$

$$m_N = \Sigma_N (\Sigma_0^{-1} m_0 + \beta \phi^T t)$$

given by $\Sigma_N^{-1} = \Sigma_0^{-1} + \beta \phi^T \phi$

$$p(t|x, t, \alpha, \beta) = \mathcal{N}(t | m_N^T \phi(x), \sigma_N^2(x))$$

$$\sigma_N^2(x) = \frac{1}{\beta} + \phi(x)^T \Sigma_N \phi(x)$$