

1.

$$E_D(\mathbf{w}) = \frac{1}{2} \sum_{n=1}^N r_n \{t_n - \mathbf{w}^T \phi(\mathbf{x}_n)\}^2$$

$$\begin{aligned}\nabla_{\mathbf{w}} E_D(\mathbf{w}) &= \frac{\partial}{\partial \mathbf{w}} \frac{1}{2} \sum_{n=1}^N r_n \{t_n - \mathbf{w}^T \phi(\mathbf{x}_n)\}^2 \\ &= \frac{1}{2} \sum_{n=1}^N r_n \cdot 2 \{t_n - \mathbf{w}^T \phi(\mathbf{x}_n)\} (-\phi(\mathbf{x}_n)^T) \\ &= - \sum_{n=1}^N r_n \{\mathbf{w}^T \phi(\mathbf{x}_n) - t_n\} \phi(\mathbf{x}_n)^T \\ &= - \sum_{n=1}^N r_n \{\mathbf{w}^T \phi(\mathbf{x}_n) \phi(\mathbf{x}_n)^T - t_n \phi(\mathbf{x}_n)^T\} \\ &= \sum_{n=1}^N r_n t_n \phi(\mathbf{x}_n)^T - \mathbf{w}^T \sum_{n=1}^N r_n \phi(\mathbf{x}_n) \phi(\mathbf{x}_n)^T\end{aligned}$$

$$\begin{aligned}\nabla_{\mathbf{w}} E_D(\mathbf{w}) &= 0 \\ &\Rightarrow \sum_{n=1}^N r_n t_n \phi(\mathbf{x}_n)^T - \mathbf{w}^T \sum_{n=1}^N r_n \phi(\mathbf{x}_n) \phi(\mathbf{x}_n)^T = 0 \\ &\Rightarrow \mathbf{w}^T \sum_{n=1}^N r_n \phi(\mathbf{x}_n) \phi(\mathbf{x}_n)^T = \sum_{n=1}^N r_n t_n \phi(\mathbf{x}_n)^T \\ &\Rightarrow \sum_{n=1}^N r_n \phi(\mathbf{x}_n) \phi(\mathbf{x}_n)^T \mathbf{w} = \sum_{n=1}^N r_n t_n \phi(\mathbf{x}_n) \\ &\Rightarrow \Phi^T \mathbf{R} \Phi \mathbf{w} = \Phi^T \mathbf{R} \mathbf{t} \\ &\Rightarrow \mathbf{w}^* = (\Phi^T \mathbf{R} \Phi)^{-1} \Phi^T \mathbf{R} \mathbf{t}\end{aligned}$$

For weighted sum-of-squares for replicated data points, if original data point (x_n, t_n) is replicated r_n times or the original data point is observed r_n times, then there are essentially $\sum_n r_n$ data points with the same (x_n, t_n) resulting in the same function as the original data dependent variance version.

$$\begin{aligned}&\frac{1}{2} \sum_{n=1}^N \underbrace{[(t_n - \mathbf{w}^T \phi(x_n))^2 + \dots + (t_n - \mathbf{w}^T \phi(x_n))^2]}_{r_n \text{ times}} \\ &= \frac{1}{2} \sum_{n=1}^N r_n (t_n - \mathbf{w}^T \phi(x_n))^2 \\ E_D(\mathbf{w}) &= \frac{1}{2} \sum_{n=1}^N r_n (t_n - \mathbf{w}^T \phi(x_n))^2.\end{aligned}$$

Thus, minimizing the error function on the replicated data points data set (with points repeated r_n times) gives the identical solution \mathbf{w}^* as the weighted sum-of-squares error function minimization on the original data.

2.

$$p(\mathbf{w} \mid \mathbf{t}) = \mathcal{N}(\mathbf{w} \mid \mathbf{m}_N, \mathbf{S}_N)$$

$$p(\mathbf{w} \mid \mathbf{t}) \propto p(\mathbf{t} \mid \mathbf{w}) p(\mathbf{w})$$

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

$$p(\mathbf{w}) = \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \mathbf{S}_0)$$

$$\mathcal{N}(\mathbf{w} \mid \mathbf{m}_N, \mathbf{S}_N) \propto \prod_{n=1}^N \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \mathbf{S}_0)$$

$$\begin{aligned} \ln \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \mathbf{S}_0) &= \ln \left(\frac{1}{(2\pi)^{D/2} |\mathbf{S}_0|^{1/2}} \exp \left\{ -\frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \right\} \right) \\ &= \ln \frac{1}{(2\pi)^{D/2} |\mathbf{S}_0|^{1/2}} + \ln \exp \left\{ -\frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \right\} \\ &= -\frac{D}{2} \ln(2\pi) - \frac{1}{2} \ln |\mathbf{S}_0| - \frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \\ &= -\frac{1}{2} (\mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{w} - 2 \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{m}_0 + \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0) \\ &= -\frac{1}{2} \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{w} + \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{m}_0 - \frac{1}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \end{aligned}$$

$$\begin{aligned} \ln \prod_{n=1}^N \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) &= \sum_{n=1}^N \ln \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) \\ &= \sum_{n=1}^N \ln \left(\frac{1}{(2\pi)^{1/2} (\beta^{-1})^{1/2}} \exp \left\{ -\frac{1}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \beta \right\} \right) \\ &= \sum_{n=1}^N \left[\ln \frac{1}{(2\pi)^{1/2} (\beta^{-1})^{1/2}} + \ln \exp \left\{ -\frac{1}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \beta \right\} \right] \\ &= \sum_{n=1}^N \left[-\frac{1}{2} \ln(2\pi) - \frac{1}{2} \ln(\beta^{-1}) - \frac{1}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \beta \right] \\ &= -\frac{N}{2} \ln(2\pi) - \frac{N}{2} \ln(\beta^{-1}) - \frac{1}{2} \beta \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \\ &= -\frac{N}{2} \ln(2\pi) + \frac{N}{2} \ln \beta - \frac{1}{2} \beta \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \\ &= -\frac{\beta}{2} \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \\ &= -\frac{\beta}{2} \sum_{n=1}^N (t_n^2 - 2 t_n (\mathbf{w}^T \phi(x_n)) + (\mathbf{w}^T \phi(x_n))^2) \\ &= -\frac{\beta}{2} \sum_{n=1}^N (t_n^2 - 2 t_n (\mathbf{w}^T \phi(x_n)) + \mathbf{w}^T \phi(x_n) \phi(x_n)^T \mathbf{w}) \\ &= -\frac{\beta}{2} \sum_{n=1}^N t_n^2 + \beta \sum_{n=1}^N t_n (\mathbf{w}^T \phi(x_n)) - \frac{\beta}{2} \sum_{n=1}^N \mathbf{w}^T \phi(x_n) \phi(x_n)^T \mathbf{w} \end{aligned}$$

$$\begin{aligned}
\ln p(\mathbf{w} \mid \mathbf{t}) &= \ln p(\mathbf{t} \mid \mathbf{w}) p(\mathbf{w}) \\
&= \ln \left(\prod_{n=1}^N \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \mathbf{S}_0) \right) \\
&= \ln \prod_{n=1}^N \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) + \ln \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \mathbf{S}_0) \\
&= -\frac{\beta}{2} \sum_{n=1}^N t_n^2 + \beta \sum_{n=1}^N t_n (\mathbf{w}^T \phi(x_n)) - \frac{\beta}{2} \sum_{n=1}^N \mathbf{w}^T \phi(x_n) \phi(x_n)^T \mathbf{w} - \frac{1}{2} \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{w} + \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{m}_0 - \frac{1}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \\
&= -\frac{\beta}{2} \sum_{n=1}^N \mathbf{w}^T \phi(x_n) \phi(x_n)^T \mathbf{w} - \frac{1}{2} \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{w} + \beta \sum_{n=1}^N t_n (\mathbf{w}^T \phi(x_n)) + \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{m}_0 - \frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{1}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \\
&= -\frac{1}{2} \mathbf{w}^T \left(\sum_{n=1}^N \beta \phi(x_n) \phi(x_n)^T + \mathbf{S}_0^{-1} \right) \mathbf{w} + \mathbf{w}^T \left(\sum_{n=1}^N \beta \phi(x_n) t_n + \mathbf{S}_0^{-1} \mathbf{m}_0 \right) + \left(-\frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{1}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \right) \\
&= -\frac{1}{2} \mathbf{w}^T \left(\beta \Phi^T \Phi + \mathbf{S}_0^{-1} \right) \mathbf{w} + \mathbf{w}^T \left(\beta \Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0 \right) + \left(-\frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{1}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \right) \longrightarrow (1)
\end{aligned}$$

$$\ln \mathcal{N}(\mathbf{w} \mid \mathbf{m}_N, \mathbf{S}_N) = -\frac{1}{2} \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{w} + \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{m}_N - \frac{1}{2} \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N \longrightarrow (2)$$

From (1) and (2), comparing the quadratic and linear terms:

$$\begin{aligned}
-\frac{1}{2} \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{w} &= -\frac{1}{2} \mathbf{w}^T \left(\beta \Phi^T \Phi + \mathbf{S}_0^{-1} \right) \mathbf{w} \\
\Rightarrow \mathbf{S}_N^{-1} &= \beta \Phi^T \Phi + \mathbf{S}_0^{-1}
\end{aligned}$$

$$\begin{aligned}
\mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{m}_N &= \mathbf{w}^T \left(\beta \Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0 \right) \\
\Rightarrow \mathbf{S}_N^{-1} \mathbf{m}_N &= \left(\beta \Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0 \right) \\
\Rightarrow \mathbf{m}_N &= \mathbf{S}_N \left(\beta \Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0 \right)
\end{aligned}$$

3.

$$\begin{aligned}
p(\mathbf{w}, \beta) &= \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \beta^{-1} \mathbf{S}_0) \Gamma(\beta \mid a_0, b_0) \\
p(\mathbf{w}, \beta \mid \mathbf{t}) &= \mathcal{N}(\mathbf{w} \mid \mathbf{m}_N, \beta^{-1} \mathbf{S}_N) \Gamma(\beta \mid a_N, b_N) \\
p(\mathbf{w}, \beta \mid \mathbf{t}) &\propto p(\mathbf{t} \mid \mathbf{w}, \beta) p(\mathbf{w}, \beta) \\
p(\mathbf{t} \mid \mathbf{w}, \beta) &= \prod_{n=1}^N \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) \\
\mathcal{N}(\mathbf{x} \mid \boldsymbol{\mu}, \Sigma) &= \frac{1}{(2\pi)^{D/2} |\Sigma|^{1/2}} \exp \left\{ -\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu})^\top \Sigma^{-1} (\mathbf{x} - \boldsymbol{\mu}) \right\}.
\end{aligned}$$

$$\begin{aligned}
\ln \mathcal{N}(\mathbf{w} \mid \mathbf{m}_0, \beta^{-1} \mathbf{S}_0) &= \ln \frac{1}{(2\pi)^{D/2} |\beta^{-1} \mathbf{S}_0|^{1/2}} \exp \left\{ -\frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \beta^{-1} \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \right\} \\
&= \ln \frac{1}{(2\pi)^{D/2} |\beta^{-1} \mathbf{S}_0|^{1/2}} \exp \left\{ -\frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \beta \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \right\} \\
&= \ln \frac{1}{(2\pi)^{D/2} |\beta^{-1} \mathbf{S}_0|^{1/2}} + \ln \exp \left\{ -\frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \beta \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \right\} \\
&= -\frac{D}{2} \ln(2\pi) - \frac{1}{2} \ln |\beta \mathbf{S}_0| - \frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \beta \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \\
&= -\frac{1}{2} (\mathbf{w} - \mathbf{m}_0)^T \beta \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) \\
&= -\frac{1}{2} \mathbf{w}^T \beta \mathbf{S}_0^{-1} \mathbf{w} + \mathbf{w}^T \beta \mathbf{S}_0^{-1} \mathbf{m}_0 - \frac{1}{2} \mathbf{m}_0^T \beta \mathbf{S}_0^{-1} \mathbf{m}_0
\end{aligned}$$

$$\begin{aligned}
\ln \Gamma(\beta \mid a_0, b_0) &= \ln \left(\frac{b_0^{a_0}}{\Gamma(a_0)} \beta^{a_0-1} \exp(-b_0 \beta) \right) \\
&= \ln \frac{b_0^{a_0}}{\Gamma(a_0)} + \ln \beta^{a_0-1} + \ln \exp(-b_0 \beta) \\
&= \ln b_0^{a_0} - \ln \Gamma(a_0) + (a_0 - 1) \ln \beta - b_0 \beta \\
&= a_0 \ln b_0 - \ln \Gamma(a_0) + (a_0 - 1) \ln \beta - b_0 \beta
\end{aligned}$$

$$\begin{aligned}
\ln \prod_{n=1}^N \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) &= \sum_{n=1}^N \ln \mathcal{N}(t_n \mid \mathbf{w}^T \phi(x_n), \beta^{-1}) \\
&= \sum_{n=1}^N \ln \left(\frac{1}{(2\pi)^{1/2} (\beta^{-1})^{1/2}} \exp \left\{ -\frac{1}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \beta \right\} \right) \\
&= \sum_{n=1}^N \left[\ln \frac{1}{(2\pi)^{1/2} (\beta^{-1})^{1/2}} + \ln \exp \left\{ -\frac{1}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \beta \right\} \right] \\
&= \sum_{n=1}^N \left[-\frac{1}{2} \ln(2\pi) - \frac{1}{2} \ln(\beta^{-1}) - \frac{1}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \beta \right] \\
&= -\frac{N}{2} \ln(2\pi) - \frac{N}{2} \ln(\beta^{-1}) - \frac{1}{2} \beta \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \\
&= -\frac{N}{2} \ln(2\pi) + \frac{N}{2} \ln \beta - \frac{1}{2} \beta \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \\
&= -\frac{\beta}{2} \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \\
&= -\frac{\beta}{2} \sum_{n=1}^N (t_n^2 - 2 t_n (\mathbf{w}^T \phi(x_n)) + (\mathbf{w}^T \phi(x_n))^2) \\
&= -\frac{\beta}{2} \sum_{n=1}^N (t_n^2 - 2 t_n (\mathbf{w}^T \phi(x_n)) + \mathbf{w}^T \phi(x_n) \phi(x_n)^T \mathbf{w}) \\
&= -\frac{\beta}{2} \sum_{n=1}^N t_n^2 + \beta \sum_{n=1}^N t_n (\mathbf{w}^T \phi(x_n)) - \frac{\beta}{2} \sum_{n=1}^N \mathbf{w}^T \phi(x_n) \phi(x_n)^T \mathbf{w}
\end{aligned}$$

$$\begin{aligned}
\ln p(\mathbf{w}, \beta | \mathbf{t}) &= \ln [p(\mathbf{t} | \mathbf{w}, \beta) p(\mathbf{w}, \beta)] \\
&= \ln p(\mathbf{t} | \mathbf{w}, \beta) + \ln p(\mathbf{w}, \beta) \\
&= \ln \prod_{n=1}^N \mathcal{N}(t_n | \mathbf{w}^T \phi(x_n), \beta^{-1}) + \ln \mathcal{N}(\mathbf{w} | \mathbf{m}_0, \beta^{-1} \mathbf{S}_0) + \ln \Gamma(\beta | a_0, b_0) \\
&\propto -\frac{\beta}{2} \sum_{n=1}^N \left[t_n^2 - 2t_n (\mathbf{w}^T \phi(x_n)) + (\mathbf{w}^T \phi(x_n))^2 \right] \\
&\quad - \frac{\beta}{2} (\mathbf{w} - \mathbf{m}_0)^T \mathbf{S}_0^{-1} (\mathbf{w} - \mathbf{m}_0) + (a_0 - 1) \ln \beta - b_0 \beta + a_0 \ln b_0 - \ln \Gamma(a_0) \\
&= -\frac{\beta}{2} \sum_{n=1}^N t_n^2 + \beta \sum_{n=1}^N t_n (\mathbf{w}^T \phi(x_n)) - \frac{\beta}{2} \sum_{n=1}^N (\mathbf{w}^T \phi(x_n))^2 \\
&\quad - \frac{\beta}{2} \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{w} + \beta \mathbf{w}^T \mathbf{S}_0^{-1} \mathbf{m}_0 - \frac{\beta}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \\
&\quad + (a_0 - 1) \ln \beta - b_0 \beta + a_0 \ln b_0 - \ln \Gamma(a_0) \\
&= -\frac{\beta}{2} \mathbf{w}^T \left(\sum_{n=1}^N \phi(x_n) \phi(x_n)^T + \mathbf{S}_0^{-1} \right) \mathbf{w} \\
&\quad + \beta \mathbf{w}^T \left(\sum_{n=1}^N t_n \phi(x_n) + \mathbf{S}_0^{-1} \mathbf{m}_0 \right) - \frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{\beta}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 \\
&\quad + (a_0 - 1) \ln \beta - b_0 \beta + a_0 \ln b_0 - \ln \Gamma(a_0) \\
&= -\frac{\beta}{2} \mathbf{w}^T (\Phi^T \Phi + \mathbf{S}_0^{-1}) \mathbf{w} + \beta \mathbf{w}^T (\Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0) \\
&\quad - \frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{\beta}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 + (a_0 - 1) \ln \beta - b_0 \beta + a_0 \ln b_0 - \ln \Gamma(a_0) \longrightarrow (1)
\end{aligned}$$

$$\begin{aligned}
\ln p(\mathbf{w}, \beta | \mathbf{t}) &= \ln [\mathcal{N}(\mathbf{w} | \mathbf{m}_N, \beta^{-1} \mathbf{S}_N) \Gamma(\beta | a_N, b_N)] \\
&= \ln \mathcal{N}(\mathbf{w} | \mathbf{m}_N, \beta^{-1} \mathbf{S}_N) + \ln \Gamma(\beta | a_N, b_N) \\
&= -\frac{1}{2} (\mathbf{w} - \mathbf{m}_N)^T \beta \mathbf{S}_N^{-1} (\mathbf{w} - \mathbf{m}_N) + a_N \ln b_N - \ln \Gamma(a_N) + (a_N - 1) \ln \beta - b_N \beta \\
&= -\frac{\beta}{2} \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{w} + \beta \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{m}_N \\
&\quad - \frac{\beta}{2} \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N + a_N \ln b_N - \ln \Gamma(a_N) + (a_N - 1) \ln \beta - b_N \beta \longrightarrow (2)
\end{aligned}$$

From (1) and (2), comparing the quadratic, linear and constant terms:

$$\begin{aligned}
-\frac{\beta}{2} \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{w} &= -\frac{\beta}{2} \mathbf{w}^T (\Phi^T \Phi + \mathbf{S}_0^{-1}) \mathbf{w} \\
\Rightarrow \mathbf{S}_N^{-1} &= (\Phi^T \Phi + \mathbf{S}_0^{-1}) \\
\Rightarrow \mathbf{S}_N &= (\Phi^T \Phi + \mathbf{S}_0^{-1})^{-1}
\end{aligned}$$

$$\beta \mathbf{w}^T \mathbf{S}_N^{-1} \mathbf{m}_N = \beta \mathbf{w}^T (\Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0)$$

$$\Rightarrow \mathbf{S}_N^{-1} \mathbf{m}_N = (\Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0)$$

$$\Rightarrow \mathbf{m}_N = \mathbf{S}_N (\Phi^T \mathbf{t} + \mathbf{S}_0^{-1} \mathbf{m}_0)$$

$$-\frac{\beta}{2} \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N - b_N \beta = -\frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{\beta}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 - b_0 \beta$$

$$\Rightarrow -b_N \beta = -\frac{\beta}{2} \sum_{n=1}^N t_n^2 - \frac{\beta}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 - b_0 \beta + \frac{\beta}{2} \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N$$

$$\Rightarrow b_N \beta = \frac{\beta}{2} \sum_{n=1}^N t_n^2 + \frac{\beta}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 + b_0 \beta - \frac{\beta}{2} \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N$$

$$\Rightarrow b_N = \frac{1}{2} \sum_{n=1}^N t_n^2 + \frac{1}{2} \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 + b_0 - \frac{1}{2} \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N$$

$$\Rightarrow b_N = b_0 + \frac{1}{2} \left(\sum_{n=1}^N t_n^2 + \mathbf{m}_0^T \mathbf{S}_0^{-1} \mathbf{m}_0 - \mathbf{m}_N^T \mathbf{S}_N^{-1} \mathbf{m}_N \right)$$

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

$$\Rightarrow p(\mathbf{t} \mid \mathbf{w}, \beta) = \prod_{n=1}^N \sqrt{\frac{\beta}{2\pi}} \exp \left(-\frac{\beta}{2} (t_n - \mathbf{w}^T \phi(x_n))^2 \right)$$

$$\Rightarrow p(\mathbf{t} \mid \mathbf{w}, \beta) = \beta^{\frac{N}{2}} \cdot (2\pi)^{-\frac{N}{2}} \exp \left(-\frac{\beta}{2} \sum_{n=1}^N (t_n - \mathbf{w}^T \phi(x_n))^2 \right)$$

$$\Rightarrow p(\mathbf{t} \mid \mathbf{w}, \beta) = \beta^{\frac{N}{2}} \exp \left(-\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{w}\|^2 \right)$$

$$p(\mathbf{t} \mid \mathbf{w}, \beta) \propto \beta^{\frac{N}{2}} \exp \left(-\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{w}\|^2 \right) \rightarrow (3)$$

$$\Gamma(\beta \mid a_0, b_0) \propto \beta^{a_0-1} \exp(-b_0 \beta) \rightarrow (4)$$

From (3) and (4)

$$\beta^{a_0-1} \times \beta^{\frac{N}{2}} = \beta^{a_0-1+\frac{N}{2}}$$

For the case where the likelihood consists of N univariate Gaussian terms where each term contributes a factor of $\beta^{\frac{1}{2}}$ when multiplying with N results in factor $\beta^{\frac{N}{2}}$, which increases the parameter a_0 by $\frac{N}{2}$. Therefore,

$$(a_N - 1) \ln \beta = \left(a_0 - 1 + \frac{N}{2} \right) \ln \beta$$

$$\Rightarrow a_N = a_0 + \frac{N}{2}$$