3.14 ψ_j is a linear combination of the basis functions $\phi_0 \dots \phi_{M-1}$, giving us:

$$\psi_j(\mathbf{x}) = \sum_{i=0}^{M-1} \alpha_{i,j} \, \phi_i(\mathbf{x})$$

Let us define a matrix **A** such that:

$$\mathbf{A} = \begin{bmatrix} \alpha_{0,0} & \alpha_{1,0} & \dots & \alpha_{(M-1),0} \\ \alpha_{0,1} & \alpha_{1,1} & \dots & \alpha_{(M-1),1} \\ \vdots & \vdots & \vdots & \vdots \\ \alpha_{0,j} & \alpha_{1,j} & \dots & \alpha_{(M-1),j} \\ \vdots & \vdots & \vdots & \vdots \\ \alpha_{0,(M-1)} & \alpha_{1,(M-1)} & \dots & \alpha_{(M-1),(M-1)} \end{bmatrix}$$

This implies that:

$$\mathbf{A}\mathbf{\Phi}^T = \begin{bmatrix} \alpha_{0,0} & \alpha_{1,0} & \dots & \alpha_{(M-1),0} \\ \alpha_{0,1} & \alpha_{1,1} & \dots & \alpha_{(M-1),1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \alpha_{0,j} & \alpha_{1,j} & \dots & \alpha_{(M-1),j} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \alpha_{0,(M-1)} & \alpha_{1,(M-1)} & \dots & \alpha_{(M-1),(M-1)} \end{bmatrix} \begin{bmatrix} \vdots & \vdots & \vdots & \vdots \\ \phi(\mathbf{x}_1) & \phi(\mathbf{x}_2) & \dots & \phi(\mathbf{x}_N) \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

$$= \left[egin{array}{ccccc} dots & dots & dots & dots \ oldsymbol{\psi}(\mathbf{x}_1) & oldsymbol{\psi}(\mathbf{x}_2) & \ldots & oldsymbol{\psi}(\mathbf{x}_N) \ dots & dots & dots & dots \end{array}
ight] = oldsymbol{\Psi}^T$$

and $\mathbf{A}\phi(\mathbf{x}_i) = \psi(\mathbf{x}_i)$

$$\Longrightarrow \Psi = (\mathbf{A}\mathbf{\Phi}^T)^T = \mathbf{\Phi}\mathbf{A}^T$$

$$\Longrightarrow \mathbf{\Phi} = \mathbf{\Psi}\mathbf{A}^{-T}$$

Note: A^{-1} exists because change of basis matrices are invertible. Source.

Using 3.54,

$$\mathbf{S}_N = (\alpha \mathbf{I} + \beta \mathbf{\Phi}^T \mathbf{\Phi})^{-1}$$

$$= (\beta \mathbf{\Phi}^T \mathbf{\Phi})^{-1} \quad \text{Since } \alpha = 0$$

$$= (\beta \mathbf{\Phi}^T \mathbf{\Phi})^{-1}$$

$$= (\beta (\mathbf{\Psi} \mathbf{A}^{-T})^T (\mathbf{\Psi} \mathbf{A}^{-T}))^{-1}$$

$$= (\beta \mathbf{A}^{-1} \mathbf{\Psi}^T \mathbf{\Psi} \mathbf{A}^{-T})^{-1}$$

Since Ψ is orthonormal basis, $\Psi^T \Psi = \mathbf{I}$, giving us:

$$\mathbf{S}_N = (\beta \mathbf{A}^{-1} \mathbf{I} \mathbf{A}^{-T})^{-1}$$
$$= \frac{1}{\beta} (\mathbf{A}^{-1} \mathbf{A}^{-T})^{-1}$$
$$= \frac{1}{\beta} (\mathbf{A}^T \mathbf{A})$$

Applying this to 3.62, we get

$$k(\mathbf{x}, \mathbf{x}') = \beta \phi(\mathbf{x})^T \mathbf{S}_N \phi(\mathbf{x}')$$

$$= \beta \phi(\mathbf{x})^T \left(\frac{1}{\beta} (\mathbf{A}^T \mathbf{A})\right) \phi(\mathbf{x}')$$

$$= \phi(\mathbf{x})^T \mathbf{A}^T \mathbf{A} \phi(\mathbf{x}')$$

$$= (\mathbf{A} \phi(\mathbf{x}))^T (\mathbf{A} \phi(\mathbf{x}'))$$

$$= \psi(\mathbf{x})^T \psi(\mathbf{x}')$$

Now to prove that the kernel satisfies the summation constraint:

$$\sum_{n=1}^{N} k(\mathbf{x}, \mathbf{x}_n) = \sum_{n=1}^{N} \psi(\mathbf{x})^T \psi(\mathbf{x}_n)$$

$$= \sum_{n=1}^{N} \sum_{m=0}^{M-1} \psi_m(\mathbf{x}) \psi_m(\mathbf{x}_n)$$

$$= \sum_{m=0}^{M-1} \sum_{n=1}^{N} \psi_m(\mathbf{x}) \psi_m(\mathbf{x}_n)$$

$$= \sum_{m=0}^{M-1} \psi_m(\mathbf{x}) \sum_{n=1}^{N} \psi_m(\mathbf{x}_n)$$

$$= \sum_{m=0}^{M-1} \psi_m(\mathbf{x}) \sum_{n=1}^{N} \psi_m(\mathbf{x}_n) \psi_0(\mathbf{x}_n)$$

$$= \psi_0(\mathbf{x}) \sum_{n=1}^{N} \psi_0(\mathbf{x}_n) \psi_0(\mathbf{x}_n) + \sum_{m=1}^{M-1} \psi_m(\mathbf{x}) \sum_{n=1}^{N} \psi_m(\mathbf{x}_n) \psi_0(\mathbf{x}_n)$$

$$= 1.1 + \sum_{m=1}^{M-1} \psi_m(\mathbf{x}).0$$

$$= 1.$$