**6.10** First we check that the kernel  $k(\mathbf{x}, \mathbf{x'}) = f(\mathbf{x})f(\mathbf{x'})$  is a valid kernel.

Based on the solution to exercise 6.5, the Gram matrix corresponding to this kernel is given by :

$$\begin{bmatrix} f(\mathbf{x}_1)^2 & f(\mathbf{x}_1)f(\mathbf{x}_2) & \dots & f(\mathbf{x}_1)f(\mathbf{x}_N) \\ f(\mathbf{x}_2)f(\mathbf{x}_1) & f(\mathbf{x}_2)^2 & \dots & f(\mathbf{x}_2)f(\mathbf{x}_N) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_N)f(\mathbf{x}_1) & f(\mathbf{x}_N)f(\mathbf{x}_2) & \dots & f(\mathbf{x}_N)^2 \end{bmatrix}$$

$$= \begin{bmatrix} f(\mathbf{x}_1) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_2) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_N) \end{bmatrix} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} f(\mathbf{x}_1) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_2) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_N) \end{bmatrix}$$

The matrix of all ones can be represented as an outer product of a vector of ones with itself, a.k.a,  $\mathbf{11}^T$ .

This matrix is positive semidefinite as can be seen:

$$\forall \mathbf{x}, \mathbf{x}^T (\mathbf{1}\mathbf{1}^T) \mathbf{x} = (\mathbf{x}^T \mathbf{1}) (\mathbf{1}^T \mathbf{x}) = (\mathbf{x}^T \mathbf{1}) (\mathbf{x}^T \mathbf{1}) = (\mathbf{x}^T \mathbf{1})^2 \ge 0$$

The Gram matrix will also be PSD (Positive Semidefinite), based on the discussion in exercise 6.5.

The matrix of ones can be decomposed into the following matrix product:

$$\begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} = \left( \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \right) \left( \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \right)$$

Then, the Gram matrix can be written as:

$$\begin{bmatrix} f(\mathbf{x}_1) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_2) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_N) \end{bmatrix} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} f(\mathbf{x}_1) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_2) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_N) \end{bmatrix}$$

$$= \begin{bmatrix} f(\mathbf{x}_{1}) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_{2}) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \begin{pmatrix} 1 \\ \sqrt{N} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \end{pmatrix} \begin{pmatrix} 1 \\ \sqrt{N} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \end{pmatrix} \begin{bmatrix} f(\mathbf{x}_{1}) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_{2}) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_{N}) \end{bmatrix}$$

$$= \begin{pmatrix} 1 \\ \sqrt{N} \begin{bmatrix} f(\mathbf{x}_{1}) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_{2}) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \end{pmatrix} \begin{pmatrix} 1 \\ \sqrt{N} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} f(\mathbf{x}_{1}) & 0 & \dots & 0 \\ 0 & f(\mathbf{x}_{2}) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ \sqrt{N} \begin{bmatrix} f(\mathbf{x}_{1}) & f(\mathbf{x}_{1}) & \dots & f(\mathbf{x}_{1}) \\ f(\mathbf{x}_{2}) & f(\mathbf{x}_{2}) & \dots & f(\mathbf{x}_{N}) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_{N}) & f(\mathbf{x}_{N}) & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \end{pmatrix} \begin{pmatrix} 1 \\ \sqrt{N} \begin{bmatrix} f(\mathbf{x}_{1}) & f(\mathbf{x}_{2}) & \dots & f(\mathbf{x}_{N}) \\ f(\mathbf{x}_{1}) & f(\mathbf{x}_{2}) & \dots & f(\mathbf{x}_{N}) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_{N}) & f(\mathbf{x}_{N}) & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ \frac{1}{\sqrt{N}} \begin{bmatrix} f(\mathbf{x}_{1}) & f(\mathbf{x}_{1}) & \dots & f(\mathbf{x}_{1}) \\ f(\mathbf{x}_{2}) & f(\mathbf{x}_{2}) & \dots & f(\mathbf{x}_{N}) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_{N}) & f(\mathbf{x}_{N}) & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \end{pmatrix} \begin{pmatrix} 1 \\ \frac{1}{\sqrt{N}} \begin{bmatrix} f(\mathbf{x}_{1}) & f(\mathbf{x}_{1}) & \dots & f(\mathbf{x}_{1}) \\ f(\mathbf{x}_{2}) & f(\mathbf{x}_{2}) & \dots & f(\mathbf{x}_{N}) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_{N}) & f(\mathbf{x}_{N}) & \dots & f(\mathbf{x}_{N}) \end{bmatrix} \end{pmatrix}^{T}$$

Let

$$\mathbf{\Phi} = \begin{pmatrix} \frac{1}{\sqrt{N}} \begin{bmatrix} f(\mathbf{x}_1) & f(\mathbf{x}_1) & \dots & f(\mathbf{x}_1) \\ f(\mathbf{x}_2) & f(\mathbf{x}_2) & \dots & f(\mathbf{x}_2) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_N) & f(\mathbf{x}_N) & \dots & f(\mathbf{x}_N) \end{bmatrix} \end{pmatrix}$$

Then, using equation 6.9, the linear model based on this kernel will be:

$$y(\mathbf{x}) = \mathbf{k}(\mathbf{x})^T (\mathbf{K} + \lambda \mathbf{I}_N)^{-1} \mathbf{t}$$
$$= f(\mathbf{x}) [f(\mathbf{x}_1) \quad f(\mathbf{x}_2) \quad \dots \quad f(\mathbf{x}_N)] (\mathbf{\Phi} \mathbf{\Phi}^T + \lambda \mathbf{I}_N)^{-1} \mathbf{t}$$

$$f(\mathbf{x})[f(\mathbf{x}_1) \quad f(\mathbf{x}_2) \quad \dots \quad f(\mathbf{x}_N)]$$
 can be written as:

$$f(\mathbf{x}) [f(\mathbf{x}_1) \quad f(\mathbf{x}_2) \quad \dots \quad f(\mathbf{x}_N)] = \begin{bmatrix} f(\mathbf{x}) f(\mathbf{x}_1) \\ f(\mathbf{x}) f(\mathbf{x}_2) \\ \vdots \\ f(\mathbf{x}) f(\mathbf{x}_N) \end{bmatrix}^T$$

$$= \left( \left( \frac{1}{\sqrt{N}} \begin{bmatrix} f(\mathbf{x}_1) & f(\mathbf{x}_1) & \dots & f(\mathbf{x}_1) \\ f(\mathbf{x}_2) & f(\mathbf{x}_2) & \dots & f(\mathbf{x}_2) \\ \vdots & \vdots & \ddots & \vdots \\ f(\mathbf{x}_N) & f(\mathbf{x}_N) & \dots & f(\mathbf{x}_N) \end{bmatrix} \right) \left( \frac{1}{\sqrt{N}} \begin{bmatrix} f(\mathbf{x}) \\ f(\mathbf{x}) \\ \vdots \\ f(\mathbf{x}) \end{bmatrix} \right) \right)^T$$

$$= (\mathbf{\Phi} \boldsymbol{\phi}(\mathbf{x}))^T$$

where 
$$\phi(\mathbf{x}) = \frac{1}{\sqrt{N}} \begin{bmatrix} f(\mathbf{x}) \\ f(\mathbf{x}) \\ \vdots \\ f(\mathbf{x}) \end{bmatrix}$$

Thus, the simplified model becomes:

$$y(\mathbf{x}) = (\mathbf{\Phi}\phi(\mathbf{x}))^T (\mathbf{K} + \lambda \mathbf{I}_N)^{-1} \mathbf{t}$$
$$= (\mathbf{\Phi}\phi(\mathbf{x}))^T \mathbf{a}$$
$$= \mathbf{a}^T \mathbf{\Phi}\phi(\mathbf{x})$$
$$= \mathbf{w}^T \phi(\mathbf{x})$$
$$= \frac{1}{\sqrt{N}} f(\mathbf{x}) (\mathbf{w}^T \mathbf{1})$$

proving that the solution is proportional to  $f(\mathbf{x})$ .