

Ex. 13.6

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A set of N basis functions $\varphi_n(x)$, $n = 1, 2, \dots, N$, is defined on the interval $0 \leq x \leq L$. Given the velocity $U(x)$, the filtered velocity $\bar{U}(x)$ is defined as the least-square projection of $U(x)$ onto the basis functions. That is, $\bar{U}(x)$ is given by

$$\bar{U}(x) = \sum_{n=1}^N a_n \varphi_n(x) \quad (1)$$

where the basis-function coefficients a_n are determined by the condition that they minimize the mean-square residual

$$\chi \equiv \frac{1}{L} \int_0^L [\bar{U}(x) - U(x)]^2 dx \quad (2)$$

By substituting Eq. (1) into Eq. (2) and differentiating with respect to a_m , show that the coefficients satisfy the matrix equation

$$\sum_{n=1}^N B_{mn} a_n = v_m \quad (3)$$

with

$$B_{mn} \equiv \frac{1}{L} \int_0^L \varphi_m(x) \varphi_n(x) dx \quad (4)$$

$$v_m \equiv \frac{1}{L} \int_0^L \varphi_m(x) U(x) dx \quad (5)$$

Show that the matrix \mathbf{B} is symmetric positive definite. Show that $\bar{U}(x)$ can be expressed as the result of a filtering operation

$$\bar{U}(x) = \int_{x-L}^x G(r, x) U(x-r) dr \quad (6)$$

where the filter is

$$G(r, x) = \frac{1}{L} \sum_{n=1}^N \sum_{m=1}^N B_{nm}^{-1} \varphi_n(x) \varphi_m(x-r) \quad (7)$$

and B_{nm}^{-1} denotes the m - n element of the inverse of the matrix \mathbf{B} . Argue, both from Eq. (6) and from the equation for $\partial\chi/\partial a_m$, that the filtered residual $\bar{u}(x)$ is zero.

Solution

By substituting Eq. (1) into Eq. (2) and differentiating with respect to a_m , we get

$$\begin{aligned} \frac{\partial\chi}{\partial a_m} &= \frac{\partial}{\partial a_m} \left[\frac{1}{L} \int_0^L \left[\sum_{n=1}^N a_n \varphi_n(x) - U(x) \right]^2 dx \right] \\ &= 2 \frac{1}{L} \int_0^L \left[\sum_{n=1}^N a_n \varphi_n(x) - U(x) \right] \varphi_m(x) dx \end{aligned} \quad (8)$$

Since a_n is a set of coefficients which minimize Eq. (2), then Eq. (8) should equal zero. This is

$$2 \frac{1}{L} \int_0^L \left[\sum_{n=1}^N a_n \varphi_n(x) - U(x) \right] \varphi_m(x) dx = 0 \quad (9)$$

After dropping the constant 2 and rearranging the remaining terms we have

$$\frac{1}{L} \int_0^L \left[\varphi_m(x) \sum_{n=1}^N a_n \varphi_n(x) \right] dx = \underbrace{\frac{1}{L} \int_0^L \varphi_m(x) U(x) dx}_{v_m} \quad (10)$$

The right hand side term is v_m as shown in Eq. (10). As for the left hand side of Eq. (10), we could write

$$\frac{1}{L} \int_0^L \left[\varphi_m(x) \sum_{n=1}^N a_n \varphi_n(x) \right] dx = \sum_{n=1}^N a_n \underbrace{\left[\frac{1}{L} \int_0^L [\varphi_m(x) \varphi_n(x)] dx \right]}_{B_{nm}} \quad (11)$$

Here Eq. (3) holds.

Based on Eq. (4) we can rewrite matrix \mathbf{B} into the following form.

$$\mathbf{B} = \frac{1}{L} \int_0^L \boldsymbol{\varphi} \boldsymbol{\varphi}^T dx \quad (12)$$

where $\boldsymbol{\varphi}$ is a column vector

$$\boldsymbol{\varphi} = \begin{Bmatrix} \varphi_1 \\ \varphi_2 \\ \vdots \\ \varphi_N \end{Bmatrix} \quad (13)$$

Let \mathbf{y} be an arbitrary non-zero vector, do the following matrix multiplication

$$\begin{aligned} & \mathbf{y}^T \mathbf{B} \mathbf{y} \\ &= \mathbf{y}^T \left(\frac{1}{L} \int_0^L \boldsymbol{\varphi} \boldsymbol{\varphi}^T dx \right) \mathbf{y} \\ &= \frac{1}{L} \int_0^L \mathbf{y}^T \boldsymbol{\varphi} \boldsymbol{\varphi}^T \mathbf{y} dx \\ &= \frac{1}{L} \int_0^L (\mathbf{y}^T \boldsymbol{\varphi}) (\boldsymbol{\varphi}^T \mathbf{y}) dx \\ &= \frac{1}{L} \int_0^L \left(\sum_{j=1}^N y_j \varphi_j \right) \left(\sum_{j=1}^N y_j \varphi_j \right) dx \\ &= \frac{1}{L} \int_0^L \left(\sum_{j=1}^N y_j \varphi_j \right)^2 dx \end{aligned} \quad (14)$$

Since $\varphi^2 \geq 0$ and φ is a basis function, the integral of φ^2 in $[0, L]$ should not be zero. Therefore for arbitrary non-zero vector \mathbf{y} , Eq. (14) satisfies

$$\mathbf{y}^T \mathbf{B} \mathbf{y} > 0 \quad (15)$$

This is equivalent to that Eq. (12) is a positive definite matrix. In fact, Eq. (7) can be expressed in matrix notations

$$G(r, x) = \frac{1}{L} \sum_{n=1}^N \sum_{m=1}^N B_{nm}^{-1} \varphi_n(x) \varphi_m(x-r) = \frac{1}{L} (\boldsymbol{\varphi}(x))^T \mathbf{B}^{-1} \boldsymbol{\varphi}(x-r) \quad (16)$$

where \mathbf{B}^{-1} is the inverse matrix of \mathbf{B} . Noting that \mathbf{B} and \mathbf{B}^{-1} only depend on x , substitute Eq. (16) into the right hand side of Eq. (6) we obtain

$$\begin{aligned}
& \int_{x-L}^x \left[\frac{1}{L} (\boldsymbol{\varphi}(x))^T \mathbf{B}^{-1} \boldsymbol{\varphi}(x-r) \right] U(x-r) dr \\
&= (\boldsymbol{\varphi}(x))^T \mathbf{B}^{-1} \frac{1}{L} \int_{x-L}^x \boldsymbol{\varphi}(x-r) U(x-r) dr
\end{aligned} \tag{17}$$

Let's take out the integral term in Eq. (17) and perform a variable substitution

$$\frac{1}{L} \int_{x-L}^x \boldsymbol{\varphi}(x-r) U(x-r) dr \stackrel{t=x-r}{=} -\frac{1}{L} \int_L^0 \boldsymbol{\varphi}(t) U(t) dt = \frac{1}{L} \int_0^L \boldsymbol{\varphi}(t) U(t) dt \tag{18}$$

Recalling Eq. (3) and Eq. (5) we can turn Eq. (18) into

$$\frac{1}{L} \int_0^L \boldsymbol{\varphi}(t) U(t) dt = \mathbf{B} \mathbf{a} \tag{19}$$

where \mathbf{a} is a column vector expressed as Eq. (20)

$$\mathbf{a} = \begin{Bmatrix} a_1 \\ a_2 \\ \vdots \\ a_N \end{Bmatrix} \tag{20}$$

Substitute Eq. (19) back into Eq. (17)

$$\begin{aligned}
& (\boldsymbol{\varphi}(x))^T \mathbf{B}^{-1} \frac{1}{L} \int_{x-L}^x \boldsymbol{\varphi}(x-r) U(x-r) dr \\
&= (\boldsymbol{\varphi}(x))^T \mathbf{B}^{-1} \mathbf{B} \mathbf{a} \\
&= (\boldsymbol{\varphi}(x))^T \mathbf{a} \\
&= \sum_{j=1}^N a_j \varphi_j \\
&= \bar{U}(x)
\end{aligned} \tag{21}$$

It means that Eq. (6) holds true.