

PROBLEM 2. 20 points (Paper and Pencil)

1.  $C$  is an  $(N + L - 1) \times L$  matrix:

$$C = \begin{pmatrix} x_0 & 0 & \dots & 0 \\ x_1 & x_0 & \dots & 0 \\ \vdots & \vdots & & \vdots \\ x_{N-1} & x_{N-2} & \dots & x_{N-L} \\ 0 & x_{N-1} & \dots & x_{N-L+1} \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & x_{N-1} \end{pmatrix}.$$

2. We are seeking for the  $\hat{\mathbf{h}}$  which minimizes

$$\|\mathbf{y} - C\mathbf{h}\|^2$$

over all  $\mathbf{h}$ .

3.  $C\hat{\mathbf{h}}$  is an element of the inner-product space spanned by the columns of  $C$ . So  $\mathcal{V} = \text{span}\{C_1, \dots, C_L\}$  where  $C_i$  is the  $i$ th column of  $C$ ,  $i = 1, 2, \dots, L$ .

4. We are seeking the vector  $C\hat{\mathbf{h}} = \hat{\mathbf{y}} \in \mathcal{V}$  which minimizes

$$\|\mathbf{y} - \hat{\mathbf{y}}\|^2.$$

The projection theorem tells us that  $\hat{\mathbf{y}}$  is the projection of  $\mathbf{y}$  into  $\mathcal{V}$ . It has the property that the error vector  $\mathbf{y} - \hat{\mathbf{y}}$  is orthogonal to every element of  $\mathcal{V}$ . In particular, it is orthogonal to the columns of  $C$ . Hence

$$\langle \mathbf{y} - C\hat{\mathbf{h}}, C_i \rangle = 0, \quad i = 1, 2, \dots, L.$$

Hence

$$\langle \mathbf{y}, C_i \rangle = \langle C\hat{\mathbf{h}}, C_i \rangle, \quad i = 1, 2, \dots, L.$$

Equivalently,

$$C_i^\dagger \mathbf{y} = C_i^\dagger C\hat{\mathbf{h}} \quad i = 1, 2, \dots, L.$$

In matrix form

$$C^\dagger \mathbf{y} = C^\dagger C\hat{\mathbf{h}}.$$

5. Solving for  $\hat{\mathbf{h}}$  yields the least-squares approximation of  $\mathbf{h}$ :

$$\hat{\mathbf{h}} = (C^\dagger C)^{-1} C^\dagger \mathbf{y}.$$

6. The MMSE estimator is  $\hat{\mathbf{h}}_{\text{MMSE}}(\mathbf{y}) = K_{\mathbf{H}\mathbf{Y}} K_{\mathbf{Y}}^{-1} \mathbf{y}$ , where  $K_{\mathbf{H}\mathbf{Y}} = E[\mathbf{H}\mathbf{Y}^\dagger] = K_{\mathbf{H}} C^\dagger$  and  $K_{\mathbf{Y}} = E[\mathbf{Y}\mathbf{Y}^\dagger] = CK_{\mathbf{H}} C^\dagger + K_{\mathbf{Z}}$ .