

PROBLEM 1. (Paper and Pencil)

Note: the message in this problem is that we can do the up-conversion from baseband to passband by an appropriate choice of the interpolation filter if the target center frequency  $f_c$  is an integer multiple of  $\frac{1}{T_s}$ . If the center frequency  $f_d$  is an arbitrary one, then an additional pre-processing is needed.

1.  $T_s < \frac{1}{B}$ .

2. The sampling theorem tells us that we can reconstruct  $b(t)$  as follows:

$$b(t) = \sum_i b(iT_s) \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right).$$

Hence,

$$b(t) = \sum_i b(iT_s) h_0(t - iT_s).$$

3.  $p_{\mathcal{F}}(f) = b_{\mathcal{F}}(f - f_c)$ .

4.  $p(nT_s) = b(nT_s)e^{j2\pi f_c nT_s}$ . Hence, if we wish to have  $p(nT_s) = b(nT_s)$  for every integer  $n$ , we should have that  $f_c T_s = k$  for some integer  $k$ . So we should have  $f_c = \frac{k}{T_s}$ .

5.

$$\begin{aligned} \sum_i b(iT_s) h_c(t - iT_s) &= \sum_i b(iT_s) e^{j2\pi f_c(t - iT_s)} \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right) \\ &= e^{j2\pi f_c t} \sum_i b(iT_s) \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right) \\ &= b(t) e^{j2\pi f_c t} \\ &= p(t). \end{aligned}$$

6.

$$\begin{aligned} \sum_i b(iT_s) q_i h_d(t - iT_s) &= \sum_i b(iT_s) q_i e^{j2\pi f_d(t - iT_s)} \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right) \\ &= e^{j2\pi f_d t} \sum_i b(iT_s) q_i e^{-j2\pi(f_c + \Delta)iT_s} \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right) \\ &= e^{j2\pi f_d t} \sum_i b(iT_s) q_i e^{-j2\pi \Delta iT_s} \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right). \end{aligned}$$

We immediately see that if  $q_i = e^{j2\pi \Delta iT_s}$  the above becomes

$$e^{j2\pi f_d t} \sum_i b(iT_s) \operatorname{sinc}\left(\frac{t - iT_s}{T_s}\right) = b(t) e^{j2\pi f_d t} = q(t).$$

PROBLEM 2. (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  $\hat{\mathbf{y}} \in \mathcal{V}$  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  $\mathbf{h}$ :

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