## EE5040: Adaptive Signal Processing Problem Set 7: Recursive least-squares

1. (Sayed VII.12, QR method) Consider the normal equations  $H^H H \hat{\mathbf{w}} = H^H \mathbf{x}$ , where H is  $N \times M$ . Assume H has full column rank (i.e., the rank of H is M) so that  $H^H H$  is positive definite. A method to reduce the effects of ill-conditioning of H on the solution of the normal equations is to avoid forming the product  $H^H H$  and to determine the Cholesky factor L by working directly with H. This can be achieved by appealing to the so-called QR decomposition of H namely,

$$H = Q \left[ \begin{array}{c} R \\ 0 \end{array} \right],$$

where Q is  $N \times N$  unitary and R is  $M \times M$  upper-triangular with positive diagonal entries.

- (a) Show that  $L = R^{H}$ . (Remark: With the L so determined, we can solve the normal equations by using the two-step procedure in problem set 6. Alternatively, we can proceed as below, which is nowadays the preferred way of solving the normal equations due to its numerical reliability.)
- (b) Let  $\begin{bmatrix} \mathbf{z}_1 \\ \mathbf{z}_2 \end{bmatrix} = Q^H \mathbf{x}$ , where  $\mathbf{z}_1$  is  $M \times 1$ . Verify that  $||\mathbf{x} H\mathbf{w}||^2 = ||\mathbf{z}_1 R\mathbf{w}||^2 + ||\mathbf{z}_2||^2$ . Conclude that the solution  $\hat{\mathbf{w}}$  can be obtained by solving the triangular linear system of equations  $R\hat{\mathbf{w}} = \mathbf{z}_1$ . Conclude further that the resulting minimum cost is  $||\mathbf{z}_2||^2$ .
- 2. (Sayed, Chapter 34, Lemma 34.2, Complex Givens rotation) Consider a  $1 \times 2$  vector  $[a \ b]$  with possibly complex entries. Then choose  $\Theta$  as

$$\Theta = \frac{1}{\sqrt{1+|\rho|^2}} \left[ \begin{array}{cc} 1 & -\rho \\ \rho^* & 1 \end{array} \right] \ \, {\rm where} \ \ \rho = \frac{b}{a}, \ a \neq 0$$

to get

$$[a \ b]\Theta = \pm e^{j\phi_a} \sqrt{|a|^2 + |b|^2} [1 \ 0]$$

where  $\phi_a$  denotes the phase of a. If a = 0, then choose  $\Theta$  as

$$\Theta = \left[ \begin{array}{cc} 0 & -1 \\ 1 & 0 \end{array} \right]$$

to get  $[0 \ b]\Theta = [b \ 0].$ 

3. (Sayed, Example 34.1, Using Givens rotations) Assume we are given a  $2 \times 3$  matrix A

$$A = \left[ \begin{array}{rrr} 1 & 0.75 & 0.75 \\ 0.4 & 0.2 & 0.2 \end{array} \right]$$

and that we wish to reduce it to the form

$$A\Theta = \left[\begin{array}{ccc} \times & 0 & 0 \\ \times & \times & 0 \end{array}\right]$$

via a sequence of Givens rotations. (Hint: First make the (1, 3) entry 0, then the (1, 2) entry, and finally the (2, 3) entry.)