


 1 How would you get an orthonormal basis of  $X$ ?


**Solution:** Just take any basis and orthonormalize it with Gram-Schmidt or Householder QR.

 2 Show how you can get a decomposition in which  $C$  is lower (or upper) triangular, from the above factorization.

**Solution:** You first get any factorization in the form shown in Page 9-7 – Then to get an upper triangular  $C$  you use the QR factorization  $C = QR$ . Then  $U$  is replaced by

$$U_{new} = U \times \begin{pmatrix} Q & 0 \\ 0 & I \end{pmatrix}$$

and  $C$  is replaced by  $R$ . To get a lower triangular  $C$  you can use the same trick applied to  $A^T$  and transpose the final result.


 3 How can you get the ULV decomposition by using only the Householder QR factorization

(possibly with pivoting)?

**Solution:** You first get the Householder QR factorization  $A = Q_1 R_1$  of the matrix  $A$ . The second step is to perform a Householder QR factorization of the matrix  $R_1^T$ , so you will get:  $R_1^T = Q_2 R_2$ . The final step is to write:

$$A = Q_1 * R_2^T * Q_2^T \equiv URV^T$$

where  $U = Q_1 \in \mathbb{R}^{m \times m}$ ;  $V = Q_2 \in \mathbb{R}^{n \times n}$ ;  $R = R_2 \in \mathbb{R}^{m \times n}$   $\square$

 4 In the proof of the SVD decomposition, define  $U, V$  as single Householder reflectors.

**Solution:** We deal with  $U$  only [proceed similarly with  $V$ ]. We need a matrix  $P = I - 2ww^T$  such that the first column of  $A$  is  $u_1$  and all columns are orthonormal. The second requirement is satisfied by default since  $P$  is unitary. Note that if  $P$  is available we will have  $Pu_1 = e_1$  because  $P^2 = I$ . Therefore, the wanted  $w$  is simply the vector that transforms the vector  $u_1$  into  $\alpha e_1$ .  
...  $\square$

 5 How can you obtain the thin SVD from the QR factorization of  $A$  and the SVD of an  $n \times n$

matrix?

**Solution:** We first get the thin QR factorization of  $\mathbf{A}$ , namely  $\mathbf{A} = \mathbf{Q}\mathbf{R}$  where  $\mathbf{Q} \in \mathbb{R}^{m \times n}$  and  $\mathbf{R} \in \mathbb{R}^{n \times n}$ ; Then we can get the SVD  $\mathbf{R} = \mathbf{U}_R \mathbf{\Sigma} \mathbf{V}_R^T$  of  $\mathbf{R}$  and this yields:

$$\mathbf{A} = \mathbf{Q} \times \mathbf{U}_R \mathbf{\Sigma} \mathbf{V}_R^T \rightarrow \mathbf{A} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}^T, \quad \text{with } \mathbf{U} = \mathbf{Q} \times \mathbf{U}_R; \quad \mathbf{\Sigma} = \mathbf{\Sigma}_R; \quad \mathbf{V} = \mathbf{V}_R.$$