

### Least squares and QR

Problem:

Given  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^m$ ,  $m \geq n$ , find  $x$  such that  $\|Ax - b\|$  is minimal. This  $x$  is called the *least squares* solution of  $Ax = b$ .

Solution:

Let  $x, h \in \mathbb{R}^n$ , and let  $f(t) = \|A(x+th) - b\|^2 = (A(x+th) - b)^T(A(x+th) - b)$ . By a direct computation (done in class),  $f'(0) = h^T(A^T Ax - A^T b)$ . The number  $f'(0)$  is the derivative of  $\|Ax - b\|^2$  in the direction  $h$ . It is known that if  $\|Ax - b\|^2$  is minimal then all directional derivatives are zero, hence  $h^T(A^T Ax - A^T b) = 0$  for all  $h$ . This implies that the solution  $x$  of the minimization problem satisfies the system of equations

$$A^T Ax = A^T b,$$

called the *normal equations*.

The QR decomposition of  $A$  gives  $A = QR$ , where  $Q \in \mathbb{R}^{m \times m}$  is unitary and  $R \in \mathbb{R}^{m \times n}$  is upper triangular, that is

$$R = \begin{bmatrix} L^T \\ 0 \end{bmatrix},$$

where  $L^T \in \mathbb{R}^{n \times n}$  (that is,  $L^T$  is the same size as  $A^T A$ ). Then

$$A^T A = R^T Q^T QR = R^T R = LL^T,$$

that is,  $L$  is the Choleski factor of  $A^T A$ . If the columns of  $A$  are linearly independent, then  $A^T A$  is nonsingular, and the normal equations can be solved by solving in turn the two nonsingular triangular systems  $Ly = A^T b$ ,  $L^T x = y$ .

Homework:

Let

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 1 \\ 0 & 0 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

1. Compute by hand the QR decomposition of  $A$ . Use the algorithm from the book. Compute both  $Q$  and  $R$ .
2. Find the least squares solution of  $Ax = b$ .
3. Verify the results in Matlab, that is, compute  $A - QR$ ,  $Q^T Q - I$ , and  $A^T Ax - A^T b$  for your hand calculated results. You may but do not need to print the Matlab log, you can just copy the results by hand.