

# QR Decomposition Method

**Case 1:**  $A$  is full rank, i.e.,  $\text{rank } A = m$ .

Let  $A = Q \begin{bmatrix} R_1 \\ 0 \end{bmatrix}$  be a QR decomposition of  $A$  and  $Q_1$  be the isometry formed by the first  $m$  columns of  $Q$ .

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Then  $Q_1 Q_1^T$  is the orthogonal projection onto  $R(A)$  or  $\text{Col}(A)$ .  
Hence,  $b_1 = Q_1 Q_1^T b$  and the solution  $x_0$  of the LSP associated with  $Ax = b$  satisfies

$$\begin{aligned} Ax_0 &= Q_1 Q_1^T b \\ \Rightarrow Q \begin{bmatrix} R_1 \\ 0 \end{bmatrix} x_0 &= Q[e_1 \cdots e_m][e_1 \cdots e_m]^T \underbrace{Q^T b}_{:=c} \\ \Rightarrow \begin{bmatrix} R_1 \\ 0 \end{bmatrix} x_0 &= \begin{bmatrix} I_m & 0 \\ 0 & 0 \end{bmatrix}_{n \times n} c \\ \Rightarrow R_1 x_0 &= c_1 \end{aligned}$$

where  $c_1$  is the vector of first  $m$  rows of  $c$ .

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$$Q = \underbrace{Q_1}_m \underbrace{Q_2}_{n-m}$$

$$A e_i \uparrow \text{the } i\text{th row of } A$$

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$$Ax_0 = Q_1 Q_1^T b$$

$$\Rightarrow Q \begin{bmatrix} R_1 \\ 0 \end{bmatrix} x_0 = \underbrace{Q[e_1 \cdots e_m]}_{=Q_1} \underbrace{[e_1 \cdots e_m]^T}_{=Q_1^T} \underbrace{Q^T b}_{:=c}$$

$$\begin{bmatrix} R_1 x_0 \\ 0 \end{bmatrix}$$

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$$\begin{bmatrix} I_m c_1 + 0 c_2 \\ 0 c_1 + 0 c_2 \end{bmatrix}$$

$$\text{where } c = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} \begin{matrix} \vdots m \\ \vdots n-m \end{matrix}$$

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**Exercise:** Let  $AP = Q \begin{bmatrix} R_1 & R_2 \\ 0 & 0 \end{bmatrix}$  be a rank revealing QR

decomposition of  $A$  where  $R_1 \in \mathbb{R}^{r \times r}$ ,  $R_2 \in \mathbb{R}^{r \times m-r}$ ,  $R_1$  being upper triangular and nonsingular. Let  $Q_1 = [q_1 \cdots q_r]$  be the isometry formed by the first  $r$  columns of  $Q$ . Then prove that

$\text{Col}(A)(= R(A)) = \text{span} \{APe_1, \dots, APe_r\} = \text{span} \{q_1, \dots, q_r\} = \text{Col}(Q_1)$ .

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$$\text{Col}(A) (= R(A)) = \text{span} \{ A P e_1, \dots, A P e_r \} = \text{span} \{ q_1, \dots, q_r \} = \text{Col}(Q_1).$$

$$[A P e_1 \cdots A P e_k \cdots A P e_{k+1} \cdots A P e_m] = \begin{bmatrix} Q_1 & Q_2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} R_1 & R_2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} Q_1 R_1 & Q_1 R_2 \\ 0 & 0 \end{bmatrix} \text{ so } [A P e_1 \cdots A P e_k] = Q_1 R_1 \text{ and thus } \mu \text{ or condensed QR}$$

Therefore,  $Q_1 Q_1^T$  is the orthogonal projection onto  $\text{Col}(A)$  and any solution  $x_0$  of the LSP associated with  $Ax = b$  satisfies

$$\begin{aligned} A x_0 = Q_1 Q_1^T b &\Rightarrow Q \begin{bmatrix} R_1 & R_2 \\ 0 & 0 \end{bmatrix} \underbrace{P^T x_0}_{:=y} = Q [e_1 \cdots e_r] [e_1 \cdots e_r]^T \underbrace{Q^T b}_{:=c} \\ &\Rightarrow \begin{bmatrix} R_1 & R_2 \\ 0 & 0 \end{bmatrix}_{n \times m} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}_{m \times 1} = \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}_{n \times n} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}_{n \times 1} \end{aligned}$$

$c_1, y_1$  being the vectors of first  $r$  rows of  $c$  and  $y$  respectively and  $c_2, y_2$  the vector formed by the last  $m - r$  rows of  $y$  and last  $n - r$  rows of  $c$  respectively

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For ease of computation choose  $y_2 = 0$ . Then one solution  $x_0 \in \mathbb{R}^m$  satisfying

$$\|b - Ax_0\|_2^2 = \min_{x \in \mathbb{R}^m} \|b - Ax\|_2^2$$

is given by

$$x_0 = P \begin{bmatrix} y_1 \\ 0 \end{bmatrix}$$

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**Exercise:** If  $r = b - Ax_0$  prove that  $\|r\|_2 = \|c_2\|_2$ .

# QR Decomposition Method

**Pseudocode for solving the LSP associated with  $Ax = b$  via QR decomposition method:**

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1. Find  $R = \begin{bmatrix} R_1 & R_2 \\ 0 & 0 \end{bmatrix}$  of a rank revealing QR decomposition of  $A$  and suppose  $Q_1, \dots, Q_r$  are the reflectors and  $P_1, \dots, P_r$  are the permutations required in the process. Here  $r \leq m$ .  
( costs  $2nm^2 - \frac{2}{3}m^3 + O(nm) + O(m^2)$  flops if  $r = m$ .)

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2. Compute  $c = Q_r \cdots Q_1 b$  and extract vectors  $c_1, c_2$  from its ~~100~~ entries.  
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4. Solve  $R_1 y = c_1$ .  
(costs  $O(m^2)$  flops if  $r = m$ .)
5. Set  $x_0 = P_1 \cdots P_r \begin{bmatrix} y \\ 0 \end{bmatrix}_{m \times 1}$ .

(Here the zeroes are necessary only if  $r < m$ .)