

## Least Squares Problem (Reminders from Lecture 13)

# Problem Statement

Given  $A \in \mathbb{C}^{n \times p}$  with  $n > p$ , and  $b \in \mathbb{C}^n$ ,

- ▶ solve

$$\min_{x \in \mathbb{C}^p} \|b - Ax\|_2,$$

- ▶ find also minimizing  $x$ .

# Optimal Solution

## Theorem

*The following are equivalent:*

- (1)  $\|b - Ax_*\|_2 \leq \|b - Ax\|_2 \quad \forall x \in \mathbb{C}^p.$
- (2)  $Ax_*$  is the orthogonal projection of  $b$  onto  $\text{Col}(A)$ .

# Numerical Solution

Let  $A = \widehat{Q}\widehat{R}$  be a reduced QR factorization.  
The solution  $x_*$  of LSP satisfies

$$\widehat{R}x_* = \widehat{Q}^*b.$$

# Numerical Procedure

1. Compute a QR factorization

$$A = QR = \begin{bmatrix} \underbrace{\hat{Q}}_{n \times p} & \underbrace{\tilde{Q}}_{n \times n-p} \end{bmatrix} \begin{bmatrix} \underbrace{\hat{R}}_{p \times p} \\ \underbrace{0}_{n-p \times p} \end{bmatrix}$$

by Householder reflectors.  $\sim 2np^2 - 2p^3/3$  flops

2. Form  $\hat{b} = \hat{Q}^* b$ .  $\sim 4np - 2p^2$  flops

3. Solve

$$\hat{R}x_* = \hat{b}$$

by back substitution.  $\sim p^2$  flops

## Forming $\widehat{b} = \widehat{Q}^* b$

$$\widehat{b} = (Q^* b)(1 : p),$$

QR factorization algorithm returns vectors  $q_1, \dots, q_p$  s.t.

$$Q = Q_1 Q_2 \dots Q_p, \quad Q_j = \begin{bmatrix} I_{j-1} & 0 \\ 0 & I_{n-j+1} - 2q_j q_j^* \end{bmatrix}$$

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Pseudocode to form  $Q^* b$

**for**  $j = 1, 2, \dots, p$  **do**

$b(j : n) \leftarrow b(j : n) - 2q_j(q_j^* b(j : n))$

**end for**

# Back Substitution

$$\begin{bmatrix} \hat{r}_{11} & \dots & \hat{r}_{1(p-1)} & \hat{r}_{1p} \\ 0 & \ddots & & \vdots \\ & & \hat{r}_{(p-1)(p-1)} & \hat{r}_{(p-1)p} \\ 0 & & 0 & \hat{r}_{pp} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_{p-1} \\ x_p \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_{p-1} \\ b_p \end{bmatrix}$$

$$x_p = b_p / \hat{r}_{pp}$$

$$x_{p-1} = \{b_{p-1} - \hat{r}_{(p-1)p}x_p\} / \hat{r}_{(p-1)(p-1)}$$

$$x_j = \{b_j - \hat{r}_{j(j+1)}x_{j+1} - \dots - \hat{r}_{jp}x_p\} / \hat{r}_{jj}$$

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**for**  $j = p$  down to 1 **do**

$$x_j \leftarrow b_j$$

**for**  $k = j + 1, \dots, p$  **do**

$$x_j \leftarrow x_j - \hat{r}_{jk}x_k$$

**end for**

$$x_j \leftarrow x_j / \hat{r}_{jj}$$

**end for**