

## Discussion 6

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# 1 Definition reviews

## 1.1 Gram-Schmidt Process

**Definition 1.1** (Orthonormal vectors). *A set of vectors  $q_1, q_2, \dots, q_k \in \mathbb{R}^n$  is said to be orthonormal if the vectors are pairwise orthogonal, and each vector has Euclidean norm 1; that is,*

$$\langle q_i, q_j \rangle = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

**Theorem 1.2.** *Let  $Q \in \mathbb{R}^{n \times n}$ . Then  $Q$  is an orthogonal matrix if and only if its columns (rows) form an orthonormal set.*

**Definition 1.3** (Isometric matrix). *The matrix  $Q \in \mathbb{R}^{m \times n}$ ,  $m \geq n$  will be called isometric if its columns are orthonormal, i.e.,  $Q$  is an isometric if and only if  $Q^T Q = I$ .*

**Theorem 1.4** (Condensed QR Decomposition). *Let  $A \in \mathbb{R}^{m \times n}$ ,  $m \geq n$ . Then there exist matrices  $\hat{Q}$  and  $\hat{R}$  such that  $\hat{Q} \in \mathbb{R}^{m \times n}$  is an isometric matrix,  $\hat{R} \in \mathbb{R}^{n \times n}$  is upper triangular, and*

$$A = \hat{Q} \hat{R}$$

**Definition 1.5** (The procedure of Modified Gram-Schmidt Process).

**Exercise 1.6.** Find the  $QR$  decomposition of  $A$

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} = QR$$

## 1.2 Solution of least square problem

Consider an overdetermined system

$$Ax = b, \quad A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m, m > n$$

Usually,  $Ax = b$  does not have an exact solution. Thus, we try to find a solution  $\hat{x}$  that minimizes  $\|A\hat{x} - b\|_2$ .

**Theorem 1.7.** Let  $A \in \mathbb{R}^{m \times n}, m > n$ . Then there exist  $Q \in \mathbb{R}^{m \times m}$  and  $R \in \mathbb{R}^{m \times n}$ , such that  $Q$  is orthogonal and  $R = \begin{bmatrix} \hat{R} \\ 0 \end{bmatrix}$ , where  $\hat{R} \in \mathbb{R}^{n \times n}$  is upper triangular, and  $A = QR$ .

**Theorem 1.8.** Let  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^m, m > n$ , and suppose  $A$  has full rank. Then the least squares problem for the overdetermined system  $Ax = b$  has a unique solution, which can be found by solving the nonsingular system  $\hat{R}x = \hat{b}_1$ , where  $\begin{bmatrix} \hat{b}_1 \\ \hat{b}_2 \end{bmatrix} = Q^T b$ ,  $\hat{R} \in \mathbb{R}^{n \times n}$  and  $Q \in \mathbb{R}^{m \times m}$  are as in Theorem 1.6.

**Remark 1.9.** If  $\text{rank}(A) < n, \exists P \in \mathbb{R}^{n \times n}$  such that

$$A \cdot P = Q \begin{bmatrix} R_{11} & R_{12} \\ 0 & 0 \end{bmatrix}$$

$$\hat{x} = P \cdot \hat{y} = P \cdot \begin{bmatrix} R_{11}^{-1} \cdot \hat{b}_1 \\ 0 \end{bmatrix}$$

**Exercise 1.10.** Find the least square solution to the overdetermined system

$$\begin{bmatrix} 2 & 1 \\ 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ -3 \end{bmatrix}$$