

MATH50003

Numerical Analysis

III.5 QR Factorisation

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Chapter III

Numerical Linear Algebra

1. **Structured matrices** such as banded
2. **LU and PLU factorisations** for solving linear systems
3. **Cholesky factorisation** for symmetric positive definite
4. **Orthogonal matrices** such as Householder reflections
5. **QR factorisation** for solving least squares

Definition 23 (QR factorisation). The *QR factorisation* is

$$A = QR = \underbrace{\left[\mathbf{q}_1 | \cdots | \mathbf{q}_m \right]}_{Q \in U(m)} \underbrace{\begin{bmatrix} \times & \cdots & \times \\ & \ddots & \vdots \\ & & \times \\ & & 0 \\ & & \vdots \\ & & 0 \end{bmatrix}}_{R \in \mathbb{C}^{m \times n}}$$

Definition 24 (Reduced QR factorisation). The *reduced QR factorisation*

$$A = \hat{Q}\hat{R} = \underbrace{\left[\mathbf{q}_1 | \cdots | \mathbf{q}_n \right]}_{\hat{Q} \in \mathbb{C}^{m \times n}} \underbrace{\begin{bmatrix} \times & \cdots & \times \\ & \ddots & \vdots \\ & & \times \end{bmatrix}}_{\hat{R} \in \mathbb{C}^{n \times n}}$$

QR gives reduced QR

Embedded in a QR factorisation is the reduced QR

III.5.1 Reduced QR and Gram–Schmidt

Gram–Schmidt is a way of computing the reduced QR

Define

$$\mathbf{v}_j := \mathbf{a}_j - \sum_{k=1}^{j-1} \underbrace{\mathbf{q}_k^* \mathbf{a}_j}_{r_{kj}} \mathbf{q}_k$$

$$r_{jj} := \|\mathbf{v}_j\|$$

$$\mathbf{q}_j := \frac{\mathbf{v}_j}{r_{jj}}$$

Theorem (Gram–Schmidt and reduced QR) Define \mathbf{q}_j and r_{kj} as above (with $r_{kj} = 0$ if $k > j$). Then a reduced QR factorisation is given by:

$$A = \underbrace{[\mathbf{q}_1 | \cdots | \mathbf{q}_n]}_{\hat{Q} \in \mathbb{C}^{m \times n}} \underbrace{\begin{bmatrix} r_{11} & \cdots & r_{1n} \\ & \ddots & \vdots \\ & & r_{nn} \end{bmatrix}}_{\hat{R} \in \mathbb{C}^{n \times n}}$$

III.5.2 Householder reflections and QR

Householder is a more stable way to compute QR

Theorem 7 (QR). *Every matrix $A \in \mathbb{C}^{m \times n}$ has a QR factorisation:*

$$A = QR$$

where $Q \in U(m)$ and $R \in \mathbb{C}^{m \times n}$ is right triangular.

Example 17 (QR by hand).

III.5.3 QR and least squares

Use QR to solve least squares problems

Theorem 8 (least squares via QR). *Suppose $A \in \mathbb{C}^{m \times n}$ with $m \geq n$ has full rank. Given a QR factorisation $A = QR$ then*

$$\mathbf{x} = \hat{R}^{-1} \hat{Q}^* \mathbf{b}$$

minimises $\|A\mathbf{x} - \mathbf{b}\|$.

