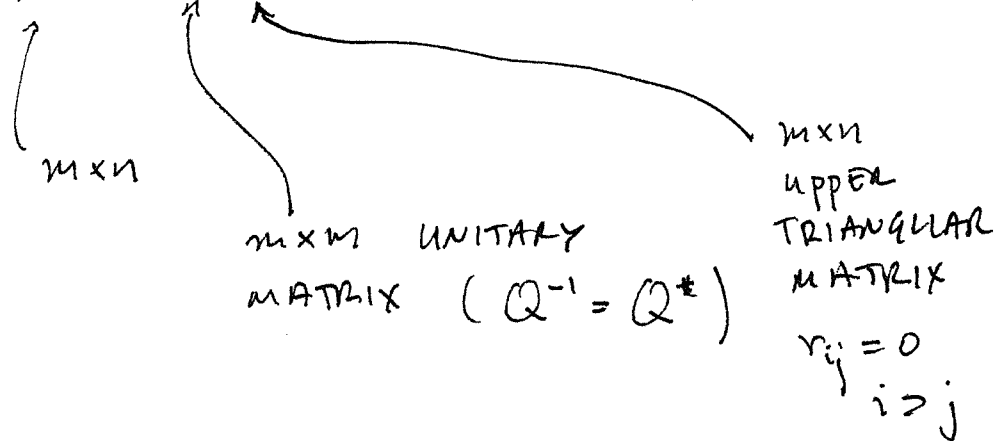


# QR FACTORIZATION

①

$$A = QR \quad (\text{FULL QR})$$



$$R = \begin{bmatrix} \text{shaded triangle} \\ 0 \end{bmatrix}$$

$$A = \hat{Q} \hat{R} \quad (\text{REDUCED QR})$$

$m \times n$  MATRIX USE COLUMNS FROM AN ORTHONORMAL SET OF VECTORS.  
 $n \times n$  UPPER TRIANGULAR

WHY DO THIS?

②

(1) SOLVE LINEAR SYSTEMS

$$Ax = b$$

$m \times m$ , FULL RANK.

(a)  $QRx = b$  (PERFORM QR)

(b) MULTIPLY BY  $Q^*$

$$\underbrace{Q^* Q}_I R x = \underbrace{Q^* b}_y$$

(c) USE BACKWARDS SUBSTITUTION TO SOLVE  $Rx = y$

LAST ROW

$$r_{nn} x_n = y_n$$

$$x_n = y_n / r_{nn}$$

PENULTIMATE ROW

$$r_{n-1,n-1} x_{n-1} + r_{n-1,n} x_n = y_{n-1}$$

(2) LEAST SQUARE PROBLEM. (3)

$$A \in \mathbb{C}^{m \times n}$$

FIND  $x$  s.t.

$\|Ax - b\|_2$  is MINIMIZED.

THE SOLUTION IS GIVEN  
BY

$$Ax = \hat{Q} \hat{Q}^* b$$

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

MULTIPLY BY  $\hat{Q}^*$

$$\Rightarrow \hat{R} x = \hat{Q}^* b$$

SOLVE FOR  $x$  USING  
BACKWARDS SUBSTITUTION.

COMPUTING QR FACTORIZATION. (4)

• CLASSICAL GRAM-SCHMIDT.

• MODIFIED GRAM-SCHMIDT.

• HOUSEHOLDER TRIANGULARIZATION.

USING THE MATRIX  $A$  ( $m \times n$ )  
WE CAN CONSTRUCT SUCCESSIVE  
SPACES SPANNED BY ITS  
COLUMNS.

$$\langle a_1 \rangle \subseteq \langle a_1, a_2 \rangle \subseteq \langle a_1, a_2, a_3 \rangle$$

...

$\langle \dots \rangle$  SPACE SPANNED BY  
THE VECTORS IN THE BRACKETS.

IDEA BEHIND QR:

CONSTRUCT SEQUENCE OF  
ORTHONORMAL VECTORS  $q_1, q_2, \dots$   
THAT SPAN THE SUCCESSIVE  
SPACES.

$$\underbrace{\begin{bmatrix} | & | & \dots & | \\ a_1 & a_2 & \dots & a_n \\ | & | & \dots & | \end{bmatrix}}_{A \ (m \times n)} = \underbrace{\begin{bmatrix} | & | & \dots & | \\ q_1 & q_2 & \dots & q_n \\ | & | & \dots & | \end{bmatrix}}_{Q \ (m \times n)} \times \underbrace{\begin{bmatrix} r_{11} & r_{12} & \dots & r_{1n} \\ & r_{22} & & \\ & & \ddots & \\ 0 & & & r_{nn} \end{bmatrix}}_{\hat{R} \ (n \times n)}$$

THE REDUCED QR FACTORIZATION  
For  $m \geq n$

$$\begin{matrix} n \\ m \end{matrix} \begin{bmatrix} | \\ A \\ | \end{bmatrix} = \begin{matrix} n \\ m \end{matrix} \begin{bmatrix} | \\ Q \\ | \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} | \\ \hat{R} \\ | \end{bmatrix}$$

⑤

Full QR ( $m \geq n$ )

$$\begin{matrix} n \\ m \end{matrix} \begin{bmatrix} | \\ A \\ | \end{bmatrix} = \begin{matrix} m \\ m \end{matrix} \begin{bmatrix} | & | \\ Q & \\ | & \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} | \\ \hat{R} \\ | \\ 0 \end{bmatrix}$$

$Q$  is ORTHONORMAL  
ADDITIONAL VECTORS THAT IF A IS FULL RANK SPAN  $\text{RANGE}(A)^\perp$   
 $\{ \text{on } \text{NULL}(A^*) \}$   
DONE TO MAKE Q UNITARY.

⑥

GRAM-SCHMIDT ORTHOGONALIZATION

GIVEN  $a_1, a_2, \dots, a_n$  FIND  
 $q_1, q_2, \dots, q_n$  AND  $r_{ij}$   
BY SUCCESSIVE ORTHOGONALIZATION

FIND  $q_j \in \langle a_1, a_2, \dots, a_j \rangle$  ⑦

THAT IS ORTHOGONAL TO

$q_1, q_2, \dots, q_{j-1}$

$$1. \quad v_j = a_j - (q_1^* a_j) q_1 - (q_2^* a_j) q_2 \\ - \dots - (q_{j-1}^* a_j) q_{j-1}$$

$$2. \quad q_j = v_j / \|v_j\|_2$$

BY THE FACT THAT  $A = \overset{1 \times n}{Q} \overset{n \times n}{R}$

WE KNOW

$$q_1 = a_1 / r_{11}$$

$$q_2 = \frac{a_2 - r_{12} q_1}{r_{22}}$$

$\vdots$

$$q_n = \frac{a_n - \sum_{i=1}^{n-1} r_{in} q_i}{r_{nn}} \quad ⑧$$

BY COMPARING THE TWO APPROACHES WE SEE THAT

$$r_{ij} = q_i^* a_j$$

$$\text{AND } \|r_{jj}\| = \|a_j - \sum_{i=1}^{j-1} r_{ij} q_i\|_2$$

CLASSICAL GRAM-SCHMIDT (CGS)  
ALGORITHM

for  $j=1$  to  $n$

$$v_j = a_j$$

for  $i=1$  to  $j-1$

$$r_{ij} = q_i^* a_j$$

$$v_j = v_j - r_{ij} q_i$$

END FOR

$$r_{jj} = \|v_j\|_2$$

$$q_j = v_j / r_{jj}$$

END FOR

CAN USE CGS TO PROVE

EXISTENCE & UNIQUENESS (FULL RANK <sub>A</sub>)

(TREFETHEN & LAU p. 51)

BUT TURNS OUT TO BE

UNSTABLE FOR NUMERICAL

PURPOSES.

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