

$u$  is stored over  $n$ .

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p ← max1 ≤ i ≤ n |xi|
if p = 0
  then r ← 0
else
  x ← x/p (vector divided by p)
  r ← √(x12 + x22 + ... + xn2)
  if (x1 < 0) r ← -r
  x1 ← r + x1
  r ← x1/r
  x(1 to n-1) ← x(1 to n-1) / x1
  x1 ← 1
  r ← r/p
  
```

Cancellation cannot occur

$x_1^2 + x_2^2 + \dots + x_n^2$  involves only positive numbers.

Cancellation does not occur sum  $r + x_1$ .

Therefore,  $r$ ,  $\gamma$ , and  $u$  are accurate.

Many algorithms, that use QR decomposition, do not calculate  $Q$  explicitly.

It suffices to save  $r$  and  $u$ .

When reflectors are used to compute the QR-decomposition,

we need to multiply each reflector  $Q$  by a matrix that is a submatrix of the matrix going under the transformation,

Suppose  $Q \in \mathbb{R}^{n \times n}$   
and  $B \in \mathbb{R}^{n \times m}$

$$\text{then } QB = (I - \gamma uu^T) B = B - \gamma uu^T B.$$

So our main focus  $\gamma uu^T B$ .

There are "good" ways and "bad" ways

to do this.

The first good thing we can do is absorb the scalar  $\gamma$  into one of the vectors,

$$\text{let } v^T = \gamma u^T$$

$$\text{then } QB = B - uv^T B$$

$$\text{Let } v' = ru' \\ \text{so that } QB = B - uv^T B$$

\* See exercise online  
(Exercise 3.2.37)

$$B \leftarrow B - u(v^T B)$$

Algorithm to calculate  $QB$  and store it over  $B$ , where  $B \in \mathbb{R}^{n \times m}$  and  $Q = I - ruu^T$ . An auxiliary  $v \in \mathbb{R}^n$

$$v^T \leftarrow ru^T \\ v^T \leftarrow v^T B$$

$$B \leftarrow B - uv^T$$

The total flop count is about  $4nm$ , which should be significantly less than multiplying  $QB$  with "normal" matrix multiplication.

This saving of flop count, because  $Q$  is rank-one update of the identity.

Theorem (QR-decomposition with reflectors)  
(Proof by induction on  $n$ )

Let  $n=1$  take  $Q = [1]$  and  $R = [a_{11}]$   
then  $A = QR$ .

Now take an arbitrary  $n \geq 2$

Show theorem holds for an  $n$  by  $n$  matrix if it holds for  $(n-1)$  by  $(n-1)$  matrices.

Let  $Q_1 \in \mathbb{R}^{n \times n}$  be a reflector that creates zeros in the first column of  $A$ .

$$\begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{bmatrix} \rightarrow \begin{bmatrix} -r_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

(creates zeros in the first column of  $A$ .)

$$Q_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{bmatrix} = \begin{bmatrix} -\tau_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Recall  $Q_1$  is symmetric

$$Q_1^T A = Q_1 A = \begin{bmatrix} -\tau_1 & \hat{a}_{12} & \dots & \hat{a}_{1n} \\ 0 & \hat{A}_2 & & \\ \vdots & & & \\ 0 & & & \end{bmatrix}$$

By the induction hypothesis  $\hat{A}_2$  has a QR-decomposition  
 $\hat{A}_2 = \hat{Q}_2 \hat{R}_2$  where  $\hat{Q}_2$  is orthogonal

and  $\hat{R}_2$  is upper triangular.

Define  $\tilde{Q}_2 \in \mathbb{R}^{n \times n}$

$$\tilde{Q}_2 = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & \hat{Q}_2 & & \\ \vdots & & & \\ 0 & & & \end{bmatrix}$$

$\tilde{Q}_2$  is orthogonal

and  $R = \tilde{Q}_2^T Q_1^T A = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & \hat{Q}_2 & & \\ \vdots & & & \\ 0 & & & \end{bmatrix} \begin{bmatrix} -\tau_1 & \hat{a}_{12} & \dots & \hat{a}_{1n} \\ 0 & \hat{A}_2 & & \\ \vdots & & & \\ 0 & & & \end{bmatrix}$

$$= \begin{bmatrix} -\tau_1 & \hat{a}_{12} & \dots & \hat{a}_{1n} \\ 0 & \hat{R}_2 & & \\ \vdots & & & \\ 0 & & & \end{bmatrix}$$

This matrix is upper triangular

Let  $Q = Q_1 \tilde{Q}_2$  then  $Q$  is orthogonal and  $QA = R$ . Therefore  $A = QR$ .

Step 1

$$Q_1 = I - \tau_1 u^{(1)} u^{(1)T}$$

Step 1  $Q_1 = I - \sigma_1 u^{(1)} u^{(1)T}$

then  $Q_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{n1} \end{bmatrix} = \begin{bmatrix} \tau_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$

as noted it suffices to only store  $\sigma_1, -\tau_1, u^{(1)}$

$$A \rightarrow R_1 = \left[ \begin{array}{c|ccc} \tau_1 & \hat{a}_{21} & \dots & \hat{a}_{n1} \\ \hline 0 & \hat{A}_2 & & \end{array} \right]$$

We could store  $-\tau_1$  in the first position  
 $u^{(1)}$  stored in first column because  $u^{(1)}$

The rest of  $u^{(1)}$  is first column of  $A$  divided  
 by  $\tau_1 + a_{11}$ .

The rest of  $A$  can be transformed as indicated  
 above since  $m=n-1$  columns are involved  
 the total flop count of step 1 is about  $4n^2$ .

need  $\hat{A}_2$  upper triangular.

follow outline of step 1

$$\left[ \begin{array}{c|ccc} -\tau_2 & \hat{a}_{23} & \dots & \hat{a}_{2n} \\ \hline 0 & \hat{A}_3 & & \\ \vdots & & & \end{array} \right]$$

Once again store  $u^{(2)}$  in first column.

cost for this step is  $4(n-1)^2$  flops.

(process identical to step 1 except it performed  
 on  $(n-1) \times (n-1)$  matrix)

Step 3  $4(n-2)^2$  flops

After  $n-1$  steps

$A$  has been transformed to upper triangular

$$u^{(1)}, \dots, u^{(n-1)}$$

The array that held  $A$  now holds  $R$ .

Another array holds  $\gamma_1, \dots, \gamma_{n-1}$

$$R = Q_{n-1} Q_{n-2} \dots Q_1 A \quad \text{where } Q_1 = I - \gamma_1 u^{(1)} u^{(1)T}$$

$$Q_2 = \left[ \begin{array}{c|ccc} 1 & 0 & \dots & 0 \\ \hline 0 & & & \\ \vdots & & & \\ 0 & & & \end{array} \right] \begin{array}{c} \\ \hline I - \gamma_2 u^{(2)} u^{(2)T} \end{array}$$

In general

$$Q_i = \left[ \begin{array}{c|ccc} I_{i-1} & & & 0 \\ \hline 0 & & & \\ \vdots & & & \\ 0 & & & \end{array} \right] \begin{array}{c} \\ \hline I - \gamma_i u^{(i)} u^{(i)T} \end{array}$$

$$Q = Q_1 Q_2 Q_3 \dots Q_{n-1}$$

$$Q^T = Q_{n-1}^T Q_{n-2}^T \dots Q_1^T$$

$$R = Q^T A$$

$$\text{and thus } A = QR$$

Algorithm to calculate QR decomposition of  $A \in \mathbb{R}^{n \times n}$  using reflectors

for  $k=1 \dots n-1$

Determine a reflector

$$Q_k = I - \gamma_k u^{(k)} u^{(k)T} \quad \text{such that}$$

$$Q_k \begin{bmatrix} a_{kk} \\ \vdots \\ a_{nk} \end{bmatrix} = \begin{bmatrix} -\gamma_k \\ 0 \\ \vdots \end{bmatrix}$$

store  $u^{(k)}$  over  $a_{k:n,k}$

$$a_{k:n,k+1:n} \leftarrow Q_k a_{k:n,k+1:n} \quad (\text{transforming values of } A \text{ to set } R)$$

$$r_{kk} \leftarrow \gamma_k$$

$$a_{k:n, k:n} \leftarrow Q_k a_{k:n, k:n} \quad (\text{transforming values of } A \text{ to set } R)$$

$$a_{kk} \leftarrow r_k$$

$$r_n \leftarrow a_{nn}$$

## Uniqueness of the QR-decomposition

Theorem Let  $A \in \mathbb{R}^{n \times n}$  be non singular,  
 Then there exists unique  $Q, R \in \mathbb{R}^{n \times n}$  such that  
 $Q$  is orthogonal,  $R$  is upper triangular with positive  
 main diagonal entries, and  $A = QR$

By previous theorems  
 $A = \hat{Q} \hat{R}$  where  $\hat{Q}$  is orthogonal and  $\hat{R}$   
 is upper triangular but does not necessarily have positive  
 main diagonal entries

Since  $A$  is nonsingular,  
 $\hat{R}$  must also be nonsingular

So the main-diagonal entries are non-zero  
 Let  $D$  be the diagonal matrix.

given by

$$d_{ii} = \begin{cases} 1 & \text{if } \hat{r}_{ii} > 0 \\ -1 & \text{if } \hat{r}_{ii} < 0 \end{cases}$$

Then  $D = D^T = D^{-1}$  is orthogonal.

Let  $Q = \hat{Q} D^{-1}$  and  $R = D \hat{R}$

Then  $Q$  is orthogonal,  $R$  is upper triangular.  
 with  $r_{ii} = d_{ii} \hat{r}_{ii} > 0$  and  $A = QR$ .

This establishes existence,

With this we have existence,

This establishes existence,

There are various to show uniqueness

Suppose  $A = Q_1 R_1 = Q_2 R_2$  where  $Q_1, Q_2$  orthogonal  
 $R_1, R_2$  upper triangular with positive main diagonal entries  
 $A^T A$  is a positive definite matrix

$$\text{and } A^T A = (Q_1 R_1)^T Q_1 R_1 = R_1^T \underbrace{Q_1^T Q_1}_{=I} R_1 = R_1^T R_1$$

(since  $Q_1^T Q_1 = I$ )

Thus  $R_1$  is a Cholesky factor of  $A^T A$

Same applies to  $R_2$

But we argued previously the uniqueness of Cholesky decomposition.

$$\text{So } R_1 = R_2 \quad \text{and} \quad Q_1 = A R_1^{-1} = A R_2^{-1} = Q_2.$$

## The Complex Case

Inner product on  $\mathbb{C}^n$  is defined by

$$\langle x, y \rangle = \sum_{i=1}^n x_i \overline{y_i} \quad x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \text{ and } y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

Exercise Show that the inner product of  $\mathbb{C}^n$  satisfies the following properties

$$(a) \quad \langle x, y \rangle = \overline{\langle y, x \rangle}$$

$$(b) \quad \langle \alpha_1 x_1 + \alpha_2 x_2, y \rangle = \alpha_1 \langle x_1, y \rangle + \alpha_2 \langle x_2, y \rangle$$

$$(c) \quad \langle x, \alpha_1 y_1 + \alpha_2 y_2 \rangle = \overline{\alpha_1} \langle x, y_1 \rangle + \overline{\alpha_2} \langle x, y_2 \rangle$$

$$(d) \quad \langle x, x \rangle \text{ is real. } \langle x, x \rangle \geq 0 \text{ and}$$

(d)  $\langle x, x \rangle$  is real,  $\langle x, x \rangle \geq 0$  and  
 $\langle x, x \rangle = 0$  if and only if  $x = 0$ .

(e)  $\sqrt{\langle x, x \rangle} = \|x\|_2$

Complex analogues of orthogonal matrices are  
Unitary matrices.

A matrix  $U \in \mathbb{C}^{n \times n}$  is unitary if  
 $U U^* = I$ , where  $U^*$  is the conjugate transpose  
of  $U$ .

Equivalent statements to above are:

$$U^* U = I$$

$$U^* = U^{-1}$$

Exercise (complex rotators)  $\begin{bmatrix} a \\ b \end{bmatrix} \in \mathbb{C}^2$

define  $U \in \mathbb{C}^{2 \times 2}$  by

$$U = \frac{1}{r} \begin{bmatrix} a & -\bar{b} \\ b & \bar{a} \end{bmatrix} \quad \text{where } r = \sqrt{|a|^2 + |b|^2}$$

Verify that

(a)  $U$  is unitary

(b)  $\det(U) = 1$

(c)  $U^* \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} r \\ 0 \end{bmatrix}$ .



Exercise (complex reflector) Let  $u \in \mathbb{C}^n$  with  $\|u\|_2 = 1$  and define  $Q \in \mathbb{C}^{n \times n}$  by  $Q = I - 2uu^*$  verify that

(a)  $Qu = -u$

(b)  $Qv = v$  if  $\langle u, v \rangle = 0$

(c)  $Q = Q^*$  ( $Q$  is Hermitian)

(d)  $Q^* = Q^{-1}$  ( $Q$  is unitary)

(e)  $Q^{-1} = Q$  ( $Q$  is an involution)

### Solution of least squares problem

Consider an overdetermined system

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