

# Simpler Problems: Triangular


$$\|Ax - b\|_2 \quad A = QR \quad \begin{matrix} \nearrow \text{upper triangular} \\ \searrow \text{orthogonal} \end{matrix}$$

What do we win from transforming a least-squares system to upper triangular form?

$$\|Ax - b\|_2 = \|Q^T(Ax - b)\|_2 = \|Q^T(Qx - b)\|_2 = \|Rx - Q^T b\|_2$$

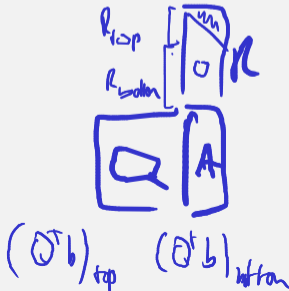
→ min

How would we minimize the residual norm?



A diagram showing a vertical vector with a horizontal line through it. The top portion of the vector is enclosed in a box with a diagonal slash, and an arrow points to it from the text below. The bottom portion of the vector is also enclosed in a box.

$$= \underbrace{\|(Q^T b)_{\text{top}} - R_{\text{top}} x\|_2^2}_0 + \|(Q^T b)_{\text{bottom}}\|_2^2$$



# Computing QR

- ▶ Gram-Schmidt
- ▶ Householder Reflectors
- ▶ Givens Rotations

[Demo: Gram-Schmidt–The Movie \[cleared\]](#)

[Demo: Gram-Schmidt and Modified Gram-Schmidt \[cleared\]](#)

[Demo: Keeping track of coefficients in Gram-Schmidt \[cleared\]](#)

Seen: Even modified Gram-Schmidt still unsatisfactory in finite precision arithmetic because of roundoff.

**NOTE:** Textbook makes further modification to ‘modified’ Gram-Schmidt:

- ▶ Orthogonalize *subsequent* rather than *preceding* vectors.
- ▶ Numerically: no difference, but sometimes algorithmically helpful.

## Economical/Reduced QR

Is QR with square  $Q$  for  $A \in \mathbb{R}^{m \times n}$  with  $m > n$  efficient?

$$\|Ax - b\|_2 = \|Q(Rx - Q^T b)\|_2$$

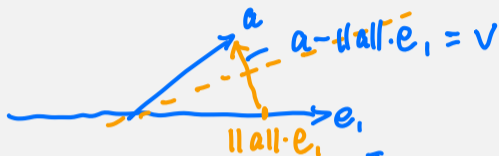
$$A = \begin{matrix} m \\ \left[ \begin{array}{c|c} n & m \\ \hline & \end{array} \right] \end{matrix} \begin{matrix} n \\ \left[ \begin{array}{c} \\ \\ \\ \end{array} \right] \end{matrix} \Rightarrow \begin{matrix} n \\ \left[ \begin{array}{c} \\ \\ \\ \end{array} \right] \end{matrix} \begin{matrix} n \\ \left[ \begin{array}{c} \\ \\ \\ \end{array} \right] \end{matrix}$$

In-Class Activity: QR

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## Householder Transformations

Find an *orthogonal* matrix  $Q$  to zero out the lower part of a vector  $\mathbf{a}$ .



$$I - 2 \frac{\mathbf{v}}{\|\mathbf{v}\|} \frac{\mathbf{v}^T}{\|\mathbf{v}\|} = I - 2 \frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}} = H$$

Householder reflector.

## Householder Reflectors: Properties

Seen from picture (and easy to see with algebra):

$$H\mathbf{a} = \pm \|\mathbf{a}\|_2 \mathbf{e}_1.$$

Remarks:

- ▶ **Q:** What if we want to zero out only the  $i + 1$ th through  $n$ th entry?  
**A:** Use  $\mathbf{e}_i$  above.
- ▶ A product  $H_n \cdots H_1 A = R$  of Householders makes it easy (and quite efficient!) to build a QR factorization.
- ▶ It turns out  $\mathbf{v}' = \mathbf{a} + \|\mathbf{a}\|_2 \mathbf{e}_1$  works out, too—just pick whichever one causes less cancellation.
- ▶  $H$  is symmetric
- ▶  $H$  is orthogonal

Demo: 3x3 Householder demo [cleared]

## Givens Rotations

If reflections work, can we make rotations work, too?

$$\begin{bmatrix} c & s \\ -s & c \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} \sqrt{a_1^2 + a_2^2} \\ 0 \end{bmatrix}$$

( produces 1 zero at a time )

Demo: 3x3 Givens demo [cleared]

## Rank-Deficient Matrices and QR

What happens with QR for rank-deficient matrices?

