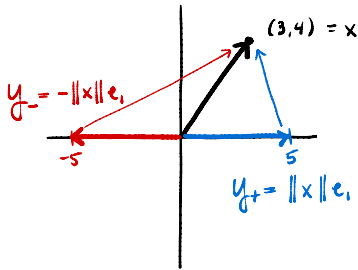


LECTURE 18

Q. Write down the Householder reflection matrix that takes the vector $x = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$ to a vector y in the direction of $e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.

Find y . * Choose a vector y ! *

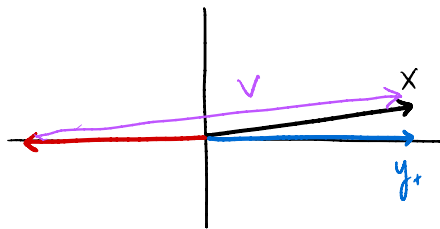
$\|x\| = \sqrt{3^2 + 4^2} = 5$. There are two vectors: $y = \begin{bmatrix} 5 \\ 0 \end{bmatrix}$ or $\begin{bmatrix} -5 \\ 0 \end{bmatrix}$



Recall: we want $v = x - y$, $u = \frac{v}{\|v\|}$.

Then $H = I - 2uu^T$.

In general, we want a vector that is "most different" from x , because of situations like:



here $x \approx +\|x\|e_1 = y_+$

so $v = x - y_+ \approx 0$ (Error!)

sign of e_1 component of x

Best to always choose $v = x + \text{sgn}(x_1) \|x\| e_1$:

$$v = x - y_- = x - (-\|x\|e_1) = \begin{bmatrix} 3 \\ 4 \end{bmatrix} + \begin{bmatrix} 5 \\ 0 \end{bmatrix} = \begin{bmatrix} 8 \\ 4 \end{bmatrix} = 4 \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\|v\| = 4\sqrt{4+1} = 4\sqrt{5}$$

$$u = \frac{v}{\|v\|} = \frac{1}{4\sqrt{5}} \cdot 4 \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{bmatrix}$$

$$H = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - 2 \begin{bmatrix} \frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} \frac{8}{5} & \frac{4}{5} \\ \frac{4}{5} & \frac{2}{5} \end{bmatrix} = \begin{bmatrix} -\frac{3}{5} & -\frac{4}{5} \\ -\frac{4}{5} & \frac{3}{5} \end{bmatrix}$$

#

H is an orthogonal matrix. $y = Hx$

MGS as triangular orthogonalization:

Let's write the algorithm as matrix operations:

Step $i=1$

$$\underbrace{[v_1^{(1)} \ v_2^{(1)} \ \dots \ v_n^{(1)}]}_{=A} \underbrace{\begin{bmatrix} \frac{1}{r_{11}} & -\frac{r_{12}}{r_{11}} & \dots & -\frac{r_{1n}}{r_{11}} \\ 0 & & & \\ & & I_{n-1} & \end{bmatrix}}_{\text{triangular operation } "R_1"} = \underbrace{[q_1 \ v_2^{(1)} \ \dots \ v_n^{(1)}]}_{=AR_1}$$

Step $i=2$

$$\underbrace{[q_1 \ v_2^{(1)} \ \dots \ v_n^{(1)}]}_{AR_1} \underbrace{\begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & \frac{1}{r_{22}} & -\frac{r_{23}}{r_{22}} & \dots & -\frac{r_{2n}}{r_{22}} \\ & 0 & & & \\ & & & I_{n-2} & \\ & 0 & & & \end{bmatrix}}_{R_2} = \underbrace{[q_1 \ q_2 \ v_3^{(1)} \ \dots \ v_n^{(1)}]}_{AR_1 R_2}$$

$$\dots \ AR_1 R_2 \dots R_n = \hat{Q} = [q_1 \ \dots \ q_n]$$

$$R_n = \begin{bmatrix} 1 & & & 0 \\ & \ddots & & \\ & & 1 & \\ 0 & & & \frac{1}{r_{nn}} \end{bmatrix}$$

* Note that each R_i is invertible!
(full rank, square matrix)

$$\text{let } \hat{R} = (R_1 R_2 \dots R_n)^{-1}. \text{ Then } A = \hat{Q} \hat{R} \text{ indeed.}$$

~

Householder Triangularization is orthogonal triangularization:

$$\begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{bmatrix} \xrightarrow{Q_1} \begin{bmatrix} \Delta & \Delta & \Delta \\ 0 & \Delta & \Delta \\ 0 & \Delta & \Delta \\ 0 & \Delta & \Delta \end{bmatrix} \xrightarrow{Q_2} \begin{bmatrix} \Delta & \Delta & \Delta \\ 0 & \square & \square \\ 0 & 0 & \square \\ 0 & 0 & \square \end{bmatrix} \xrightarrow{Q_3} \begin{bmatrix} \Delta & \Delta & \Delta \\ 0 & \square & \square \\ 0 & 0 & * \\ 0 & 0 & 0 \end{bmatrix} = R \quad (\text{full } R!) \\ A \qquad \qquad Q_1 A \qquad \qquad Q_2 Q_1 A \qquad \qquad Q_3 Q_2 Q_1 A = R$$

$$\text{let } Q^T = Q_3 Q_2 Q_1. \text{ Then } Q^T A = R \Rightarrow A = QR \quad (\text{full QR decomp})$$

Algorithm (Householder QR decomposition)

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

for $k=1:n$

$$x = A(k:m, k)$$

eg. $n=3, m=4, k=2$:

$$v_k = \text{sgn}(x_i) \|x\| e_i + x$$

in standard basis for \mathbb{R}^{m-k+1}
 $= \mathbb{R}_k \times \mathbb{R}_{k+1} \times \dots \times \mathbb{R}_m$

$$\begin{bmatrix} \Delta & \Delta & \Delta \\ 0 & \Delta & \Delta \\ 0 & \Delta & \Delta \\ 0 & \Delta & \Delta \end{bmatrix}$$

\uparrow
 x

$$v_k = \frac{v_k}{\|v_k\|} \quad \text{normalized ("}u_k\text{")}$$

$$A(k:m, k:n) = \underbrace{A(k:m, k:n) - 2v_k v_k^T A(k:m, k:n)}_{(I - 2u_k u_k^T)(A(k:m, k:n))}$$

Output R , whose top square \hat{R} is upper triangular

#

You might not wish to store the Q_i if not needed:

eg. $Ax=b$ (determined or overdetermined)

If you have $Q^T A = R$ as linear transformation,

then $Rx = Q^T Ax$ ($= Q^T b$ if $Ax=b$ is determined)

$$\text{So if we want } \min_x \|b - Ax\|^2 = \min_x \|Q^T b - Q^T Ax\|^2$$

$$= \min_x \|Rx - Q^T b\|$$

so we didn't need to store Q^T .

Note You could compute $Q^T b$ by running the algorithm

for $k=1:n$ on $[a_1, a_2, \dots, a_n | b]$

\uparrow not $n+1$!

& not store Q^T .

Numerical stability example

$$A = \begin{bmatrix} 1 & 1 & 1 \\ \varepsilon & & \\ & \varepsilon & \\ & & \varepsilon \end{bmatrix}$$

$$a_1 = [1 \ \varepsilon \ 0 \ 0]^T$$

$$a_2 = [1 \ 0 \ \varepsilon \ 0]^T$$

$$a_3 = [1 \ 0 \ 0 \ \varepsilon]^T$$

Using Householder reflections, we get

$$\underbrace{\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & -\frac{1}{\sqrt{5}} & \frac{\sqrt{5}}{\sqrt{3}} \\ & & \frac{\sqrt{5}}{\sqrt{5}} & \frac{1}{\sqrt{3}} \end{bmatrix}}_{Q_3} \underbrace{\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ & & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ & & & 1 \end{bmatrix}}_{Q_2} \underbrace{\begin{bmatrix} -1 & -\varepsilon & & \\ -\varepsilon & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}}_{Q_1} \underbrace{\begin{bmatrix} 1 & 1 & 1 \\ \varepsilon & & \\ & \varepsilon & \\ & & \varepsilon \end{bmatrix}}_A = \underbrace{\begin{bmatrix} -1 & -1 & -1 \\ & 2\varepsilon/\sqrt{2} & \varepsilon/\sqrt{2} \\ & & (\frac{1}{\sqrt{5}} + \frac{\sqrt{5}}{\sqrt{3}})\varepsilon \end{bmatrix}}_R$$

$\Rightarrow A = QR$

So $Q^T = Q_3 Q_2 Q_1$ and $Q = (Q^T)^T = Q_1^T Q_2^T Q_3^T$

and $Q^T Q = Q_3 Q_2 Q_1 Q_1^T Q_2^T Q_3^T$

there are
some epsilons
here:

$$Q_1^T Q_1 = \begin{bmatrix} -1 & -\varepsilon \\ -\varepsilon & 1 \end{bmatrix} \begin{bmatrix} -1 & -\varepsilon \\ -\varepsilon & 1 \end{bmatrix} = \begin{bmatrix} 1+\varepsilon^2 & 0 \\ 0 & 1+\varepsilon^2 \end{bmatrix} \approx I_2!$$

Turns out $Q^T Q \approx I_4$; only multiples of ε^2 appear!