

Welcome to Ma 221! Lecture 18, Oct 4

Algorithms for $A = QR$

Classical + Modified Gram-Schmidt
CGS + MGS

Equate columns of A and QR
 $A(:, i) = \sum_{j=1}^i Q(:, j) \cdot R(j, i)$

since columns of Q orthogonal

$$(Q(:, j))^T A(:, i) = R(j, i)$$

for $i = 1:n$

$$\text{tmp} = A(:, i)$$

for $j = 1:i-1$

2m flops $R(j, i) = (Q(:, j))^T \cdot A(:, i) \dots$ CGS

" $R(j, i) = (Q(:, j))^T \cdot \text{tmp} \dots$ MGS

" $\text{tmp} = \text{tmp} - R(j, i) \cdot Q(:, j)$

end for ... $\text{tmp} = R(i, i) \cdot Q(:, i)$

" $R(i, i) = \|\text{tmp}\|_2$

m flops $Q(:, i) = \text{tmp} / R(i, i)$

end for

$$\# \text{ flops} = 2mn^2 + O(mn)$$

$\sim 2 \times$ cost of NE if $m \gg n$

Householder - stable

MGS - less stable

CGS - even less stable

) still used, because
Q returned
explicitly
but for Householder
Q implicit

2 Metrics of Backward Stability

want accurate decomposition $A = QR$

$$A + E = QR, \quad \|E\| = O(\epsilon) \cdot \|A\|$$

$$\frac{\|A - QR\|}{\|A\|} = O(\epsilon)$$

want Q to be close to orthogonal

$$\|Q^T Q - I\| = O(\epsilon)$$

How to "fix" CGS and MGS: do them twice

i.e. run it again on output Q, better,
not perfect

Perturbation Theory for LS

How much can $x = \operatorname{argmin}_x \|Ax - b\|_2$
change if A and b change?

When $m = n$, same as solving $Ax = b$

so expect $\kappa(A) = \frac{\sigma_{\max}}{\sigma_{\min}}$ to appear

Another source of ill-conditioning

$$A = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad b = \begin{bmatrix} 0 \\ b_2 \end{bmatrix}, \quad x = [1, 0] \cdot \begin{bmatrix} 0 \\ b_2 \end{bmatrix} = 0$$

$$A = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}, \quad x = [1, 0] \cdot \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = b_1$$

\Rightarrow large relative change in x

In general, ill conditioned if b (nearly) orthogonal to $\text{span}(A)$

$$x + e = \underset{x}{\text{argmin}} \| (A + \delta A)x - (b + \delta b) \|_2$$

$$x = \underset{x}{\text{argmin}} \| Ax - b \|_2$$

$$e = (x + e) - x = ((A + \delta A)^T (A + \delta A))^{-1} (A + \delta A)^T (b + \delta b) - (A^T A)^{-1} A^T b$$

Do Taylor expansion, keep first term

$$(I - X)^{-1} = I + X + O(X^2)$$

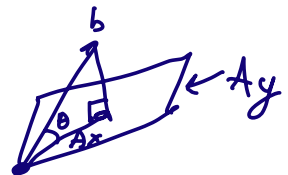
only keep terms with one "small" factor δA or δb

$$\text{Def: } \varepsilon = \max \left(\frac{\|\delta A\|_2}{\|A\|_2}, \frac{\|\delta b\|_2}{\|b\|_2} \right)$$

$$\frac{\|e\|_2}{\|x\|_2} \leq \varepsilon \left(2 \cdot \kappa(A) \cdot \frac{1}{\cos \theta} + \tan \theta \cdot \kappa^2(A) \right) + O(\varepsilon^2)$$

$$\theta = \text{angle}(b, Ax)$$

$$\sin(\theta) = \frac{\|Ax - b\|_2}{\|b\|_2}$$



$$\theta = 0 \Rightarrow \frac{\|e\|_2}{\|x\|_2} \leq \varepsilon (2 \cdot \kappa(A)) \dots \text{best case}$$

condition number large when:

① $\kappa(A)$ large

② θ near $\frac{\pi}{2} \Rightarrow \frac{1}{\cos \theta} \tan \theta$ large

③ error like $\kappa^2(A)$ when θ not near 0

Householder: Stable Alg for QR

Represent Q as product $Q = Q_1 \cdot Q_2 \cdots Q_n$
of "simple" orthogonal Q_i , chosen to
make progress zeroing out entries
of A to turn it into R

2 kinds of "simple" Q_i :

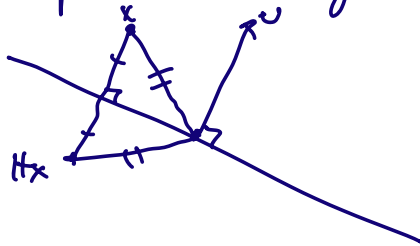
Householder Transformation (reflections)

Givens rotations $\begin{bmatrix} c & s \\ -s & c \end{bmatrix}$ $\begin{matrix} c = \cos \theta \\ s = \sin \theta \end{matrix}$

Householder: $H = I - 2uu^T$, $\|u\|_2 = 1$

$$\begin{aligned} HH^T &= (I - 2uu^T)(I - 2uu^T) \\ &= I - 2uu^T - 2uu^T + 4u \underbrace{u^T u}_{=1} = I \end{aligned}$$

Reflection Hx is reflection of x
in plane orthogonal to u



Given x , choose u so Hx has zeros
in certain locations

$$Hx = \begin{bmatrix} c \\ 0 \\ \vdots \\ 0 \end{bmatrix} = c \cdot e_1 \text{ for some } c$$

$$\|Hx\|_2 = \|x\|_2 = |c| \Rightarrow c = \pm \|x\|_2$$

$$Hx = (I - 2uu^T)x = x - 2u \underbrace{u^T x}_c = c \cdot e_1$$

$$\Rightarrow u = \frac{x - ce_1}{2u^T x}$$

denominator scalar, choose it so $\|u\|_2 = 1$

$$y = x - ce_1 = x \pm \|x\|_2 e_1$$

$$u = \frac{y}{\|y\|_2}$$

$$u = \text{House}(x)$$

How to pick sign of $\pm \|x\|_2$: choose to avoid cancellation in

$$y = \begin{bmatrix} x_1 \pm \|x\|_2 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} x_1 + \text{sign}(x_1) \cdot \|x\|_2 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

(may get R_{ii} positive or negative)
 $Q^T \cdot A$

$$\left[\begin{array}{c|c} I^3 & 0 \\ \hline 0 & H^{2 \times 2} \end{array} \right] \left[\begin{array}{c|c} 1 & 0 \\ \hline 0 & H^{3 \times 3} \end{array} \right] \left[\begin{array}{c|c} 1 & 0 \\ \hline 0 & H^{4 \times 4} \end{array} \right] H^{5 \times 5}$$

$$\begin{matrix} A^{5 \times 5} \\ \begin{bmatrix} x & x & x & x \\ x & x & x & x \\ x & x & x & x \\ x & x & x & x \\ x & x & x & x \end{bmatrix} \end{matrix} = R$$

product of 4 orthogonal matrices, so $= Q^T$

Q implicit

$$A = QR$$

GER rank 1 update

$$\text{need } (I - 2uu^T)A = A - 2u \underbrace{(u^T A)}_{\substack{\text{matrix} \\ \text{GEMV}}}$$

QR for $A^{m \times n}$ $m \geq n$

for $i = 1$ to $\min(m-1, n)$.. only need to do last col if $m > n$

$$u(i) = \text{House}(A(i:m, i))$$

$$\begin{aligned} A(i:m, i:n) &= \dots (I - 2u(i)u(i)^T) A(i:m, i:n) \\ &= A(i:m, i:n) - 2u(i) \cdot \underbrace{(u(i)^T A(i:m, i:n))}_{\text{GEHV}} \end{aligned}$$

$\underbrace{\hspace{10em}}_{\text{GER}}$

Where store $u(i)$ s? same trick as storing L in GE: in zero entries below diagonal of A

$$\text{Cost} = \sum_{i=1}^n 4(m-i+1)(n-i+1) = 2n^2m - \frac{2}{3}n^3$$

$m \gg n$: dominated by $2n^2m$
 $m = n$: $\frac{4}{3}n^3$, twice GE

Solve LS using implicit Q.

$$Q_n \dots Q_2 Q_1 A = R$$

$$A = Q_1^T Q_2^T \dots Q_n^T R$$

$$x = R^{-1} Q^T b = \underset{x}{\text{argmin}} \|Ax - b\|_2$$

for $i = 1$ to n

$$\begin{aligned} b &= Q_i b = (I - 2u_i u_i^T) b \\ &= b - 2u_i \underbrace{(u_i^T b)}_{\text{dot product}} \end{aligned}$$

endfor

axpy

$$\begin{aligned} & \hat{b} = R^{-1}b \\ \Rightarrow \text{cost} &= O(m \cdot n) \text{ much less than QR} \end{aligned}$$