

Three Hints:

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Block Jacobi: try not to invert anything

$N \times N$

try to think of individual blocks

Neumann boundary conditions:

approx to 2nd deriv

$$\frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} + O(h^2)$$

recommended to use a second order approx also for first deriv

$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix}$ A SPD, pos + neg: Think Schur complements $BA^{-1}B^T$

Reform Block G.E. $\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} = LDL^T$

$\uparrow \uparrow \uparrow$

2×2 block matrices.

ID Projection:

MR, SD

$$x \in x_0 + V\gamma$$

V is a basis for K - search space

L constant space

if $L = K$, A SPD \rightarrow CG $\|e_k\|_A$

$L = AK$, A nonsingular \rightarrow MINRES, GMRES $\|r_k\|_2$

W is a basis for L

W^TAV needs to be inverted

$$x = x_0 + V(W^TAV)^{-1}W^Tr_0$$

Krylov Subspace Methods:

$$K^k(A, r_0) = \text{Span} \{r_0, Ar_0, \dots, A^{k-1}r_0\}$$

① optimality condition

② Forming a basis for $K^n(A, r_0)$

③ preconditioning

focus on ②

↳ suppose $V = [r_0, Ar_0, \dots, A^{K-1}r_0]$

↑ bad basis b/c Ar_0 large K is similar to say, $A^{K-1}r_0$
(Precisely why power method is good - later)

• Want a good basis
↳ orthogonal

QR decomposition

w.l.o.g.: assume $x_0 = 0$ (if not, then solve $Ay = b - Ax_0$ w/ $y_0 = 0$)

so: $r_0 = b$, $x_0 + K^n(A, r_0) \rightarrow K^n(A, b)$

$$u_j = A^{j-1}r_0 = A^{j-1}b$$

$$U_k = [u_1, u_2, \dots, u_k]$$

$$AU_k = [u_2, \dots, u_{k+1}] = U_k \cdot B_k + u_{k+1} e_k^T$$

$$Au_j = u_{j+1}$$

↑ $A \cdot A^{j-1}r_0$ $A^j r_0$

$$\begin{pmatrix} 0 & \dots & 0 \\ 1 & & \\ & \ddots & \\ 0 & & 1 & 0 \end{pmatrix}$$

QR Factorization:

$$\begin{array}{c} B \\ \begin{array}{|c|} \hline n \\ \hline \end{array} \\ m \end{array} = \begin{array}{c} m \\ \begin{array}{|c|} \hline m \\ \hline \end{array} \\ Q \\ \text{orthogonal} \end{array} \begin{array}{c} \begin{array}{|c|} \hline \text{shaded triangle} \\ \hline \end{array} \\ \begin{array}{|c|} \hline 0 \\ \hline \end{array} \\ \begin{array}{|c|} \hline 0 \\ \hline \end{array} \\ \begin{array}{|c|} \hline 0 \\ \hline \end{array} \\ R \\ \text{square } (n \times n) \end{array} \quad \text{or} \quad = \begin{array}{c} \begin{array}{|c|} \hline \text{shaded triangle} \\ \hline \end{array} \\ \begin{array}{|c|} \hline \\ \hline \end{array} \\ Q \\ \text{economy size} \end{array} \quad R \leftarrow \text{square } (n \times n)$$

Three ways to find QR

Graham-Schmidt

Householder

Givens Rotations

$$U_k = Q_k R_k$$

$$A Q_k R_k = Q_k R_k B_k + u_k e_k^T$$

mult. by R_k^{-1} on the right.

$$A Q_k = Q_k C_k + u_k e_k^T R_k^{-1}$$

$$R_k B_k R_k^{-1} = \begin{pmatrix} \text{upper triangular} \\ \text{upper Hessenberg} \end{pmatrix} = \begin{pmatrix} \text{upper triangular} \\ \text{upper Hessenberg} \end{pmatrix} = H_k$$

$$A Q_k = Q_k H_k + \text{correction} \leftarrow 1 \text{ column @ } k\text{-th position}$$

$$= Q_{k+1} H_{k+1,k}$$

if A is symmetric

$$Q_k^T A Q_k = H_k$$

$$H_k^T = (Q_k^T A Q_k)^T = Q_k^T A^T Q_k = Q_k^T A Q_k = H_k$$

$$\begin{pmatrix} H_k \\ \text{tridiagonal} \end{pmatrix}$$

JDI:

$$A Q_k = \dots$$

column-by-column: Arnoldi-Process

$$A q_1 = (q_1, q_2) \begin{pmatrix} h_{11} \\ h_{21} \end{pmatrix} = h_{11} q_1 + h_{21} q_2 \leftarrow \begin{matrix} 4 \text{ unknowns } q_1, q_2, h_{11}, h_{21} \\ \text{orthogonality: } q_1^T q_1 = 1 \end{matrix}$$

$$A(q_1, q_2) = (q_1, q_2, q_3) \begin{pmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \\ 0 & h_{32} \end{pmatrix}$$

$$\begin{matrix} q_1^T q_1 = 1 \\ q_2^T q_1 = 0 \\ q_2^T q_2 = 0 \end{matrix}$$

start taking inner products

$$q_1^T A q_1 = h_{11} q_1^T q_1 + h_{21} q_1^T q_2 = h_{11}$$

what is q_1 ?

q_1 spans r_0

$\{q_1, q_2\}$ span $\{r_0, A r_0\}$

$$\Rightarrow q_1 = \frac{r_0}{\|r_0\|}$$

$$h_{21} q_2 = A q_1 - h_{11} q_1$$

$$h_{21} = \|A q_1 - h_{11} q_1\|$$