

Iterative Methods

Part 3: Arnoldi for Linear Sys. (GMRES)

Recap

Arnoldi generates ONB Q_n s.t.

ONB
for
Krylov
Subspace

$$\begin{bmatrix} | & | & \dots & | \\ x & Ax & \dots & A^{n-1}x \\ | & | & \dots & | \end{bmatrix} = \begin{bmatrix} | & | & \dots & | \\ q_1 & q_2 & \dots & q_n \\ | & | & \dots & | \end{bmatrix} \begin{bmatrix} \text{upper triangular} \\ \text{matrix} \\ \text{with zeros below diagonal} \end{bmatrix}$$

K_n Q_n R_n

Arnoldi
Decomp.

$$A \begin{bmatrix} | & | & \dots & | \\ q_1 & \dots & q_n \\ | & | & \dots & | \end{bmatrix} = \begin{bmatrix} | & | & \dots & | \\ q_1 & \dots & q_n \\ | & | & \dots & | \end{bmatrix} \begin{bmatrix} \text{upper triangular} \\ \text{matrix} \\ \text{with zeros below diagonal} \end{bmatrix} + h_{n+1} q_{n+1} e_n^T$$

Q_n Q_n H_n

$$\Rightarrow H_n = Q_n^* A Q_n$$

Approximating Eigenvalues

$\theta_1, \dots, \theta_m$ eigenvalues of H_n are roots of the minimizing polynomial

$$\|p_n(A)x\|_2 = \min_{\substack{p_n \in P_n \\ \text{deg. } n \text{ poly's}}} \text{over}$$

\Rightarrow As n increases, p_n tends to place roots near extremal eigenvalues of A .

\Rightarrow Can be much more efficient than power iterations / simultaneous iterations.

Generalized Minimal Residuals (GMRES)

The Arnoldi iteration also produces powerful approximations to the solns of linear systems.

$$\text{Solve } Ax = b \quad \Rightarrow \quad x_{\text{ex}} = A^{-1}b \quad \begin{matrix} \text{Exact} \\ \text{soln.} \end{matrix}$$

$$\begin{matrix} \text{Krylov} \\ \text{Subspace} \end{matrix} \quad K_n(A, b) = \text{span} \{b, Ab, A^2b, \dots, A^{n-1}b\}$$

Idea: choose $x_n \in K_n(A, b)$ to minimize the residual norm

$$x_n = \underset{x \in K_n(A, b)}{\operatorname{argmin}} \|Ax - b\|_2$$

\Rightarrow This is just a Least-Squares problem!

Arnoldi ! Least-Squares

Approach 1: write $\underbrace{AK_n c}_{x \in K_n(A,b)}$ to get

$$x_n = \argmin_{c \in \mathbb{R}^n} \|AK_n c - b\|_2$$

Solve w/QR factorization of tall skinny mat

$$\begin{bmatrix} A_1 & A_2 & \dots & A_n \\ 1 & 1 & \dots & 1 \end{bmatrix}$$

\Rightarrow ill-conditioned Least-Squares problem

Approach 2: write $\underbrace{AQ_n y}_{x \in K_n(A,b)}$ to get

$$x_n = \argmin_{Q_n y \in \mathbb{R}^n} \|AQ_n y - b\|_2$$

Use Arnoldi decomp $AQ_n = Q_{n+1} \tilde{H}_n$

$$\|AQ_n y - b\|_2 = \|Q_{n+1} \tilde{H}_n y - b\|_2$$

$$\text{b/c } Q_{n+1} \tilde{H}_n y - b \in K_n(A,b) \rightarrow = \|\tilde{H}_n y - Q_n^* b\|_2$$

Since $q_1 = b/\|b\|$, we need to solve

$$1) y_n = \underset{y \in \mathbb{R}^n}{\operatorname{argmin}} \| \tilde{H}_n y - \|b\|e \|_2$$

$$2) x_n = Q_n y_n$$

\Rightarrow Avoids ill-conditioning of K_n .

\Rightarrow Special structure of \tilde{H}_n allows us to update Least-Squares soln. y_n fast as we build up \tilde{H}_n column-by-column with Arnoldi iterations.

GMRES (Pseudocode)

$$q_1 = b/\|b\|$$

for $k=1, 2, 3, \dots$

- Step n of Arnoldi adds k^{th} column to \tilde{H}_k, Q_k
- solve LS problem $y = \underset{y}{\operatorname{argmin}} \| \tilde{H}_k y - \|b\|e \|_2$ using fast QR update.
- Compute $x_k = Q_k y_k$

GMRES : Polynomial Approx

Since $x_n \in K_n(A, b) = \operatorname{span} \{b, Ab, A^2b, \dots, A^{n-1}b\}$,

$$x_n = q_{n-1}(A)b = c_1 b + c_2 Ab + \dots + c_{n-1} A^{n-1}b$$

We minimize the residual

$$r_n = b - Ax_n = b - A(c_0b + c_1Ab + \dots + c_{n-1}A^{n-1}b)$$

so $r_n = p_n(A)b$ where $p_n(z) = 1 - zq_{n-1}(z)$

is a degree n polynomial w/ $p_n(0) = 1$.

Then Over all degree n polynomials with $p(0) = 1$, GMRES constructs $p_n(z)$

s.t. $\|p_n(A)b\|_2 = \text{minimum.}$

Convergence

First, note that $\|r_{n+1}\| \leq \|r_n\|$ for all $n \geq 1$
b/c $K_n(A, b) \subset K_{n+1}(A, b)$.

Second, note that $\|r_n\| = 0$ since
 $K_n(A, b) = \mathbb{R}^m$ and A is invertible: $A^{-1}b \in K_n(A, b)$

so $\|r_n\| \rightarrow 0$ monotonically (non-increasing)
in no more than m steps.

We need $n < m$, so key question is how fast does $\|r_n\|$ decrease?

Extremal problems in polynomial approx

Denote $P_n^0 = \{\text{set of deg. } n \text{ poly with } p(0)=1\}$

$$\|r_n\| = \min_{p \in P_n^0} \|p(A)b\| \leq \|p(A)\| \|b\| \quad \text{for any } p \in P_n^0$$

$$\frac{\|r_n\|}{\|b\|} \leq \inf_{p \in P_n^0} \|p(A)\|$$

$A = V\Lambda V^{-1}$ diagonalizable, then

$$p(A) = V p(\Lambda) V^{-1} \Rightarrow \|p(A)\| \leq \underbrace{\|V\| \|V^{-1}\|}_{K(V)} \underbrace{\|p(\Lambda)\|}_{\max_{1 \leq i \leq n} p(\lambda_i)}$$

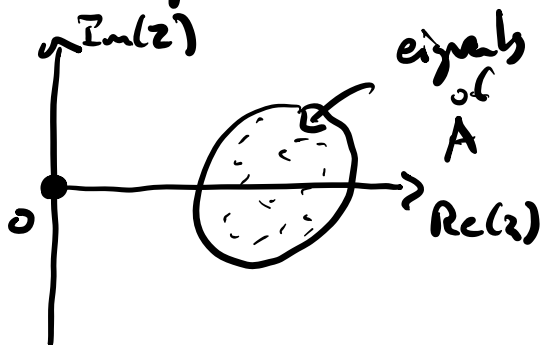
$$\therefore \frac{\|r_n\|}{\|b\|} \leq K(V) \min_{p \in P_n^0} \max_{1 \leq i \leq n} p(\lambda_i)$$

Min-max problem on spectrum of A .

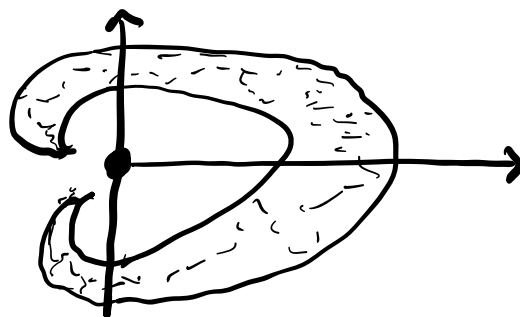
Note: Non-normality (large $K(V)$) can degrade convergence of iterative methods like

Arnoldi : GMRES.

Example 1



Example 2



$P(0) = 1$, how small
can $p(z)$ be on
spectrum of A ?