

# The worst-case GMRES for normal matrices

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A system of linear algebraic equations

$$\mathbf{A}x = b,$$

$\mathbf{A} \in \mathbb{C}^{n \times n}$  is **nonsingular and normal**,  $b \in \mathbb{C}^n$ .

Eigendecomposition of  $\mathbf{A}$ :

$$\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^H, \quad \mathbf{Q}^H\mathbf{Q} = \mathbf{I}, \quad \mathbf{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_n).$$

We will assume that all eigenvalues of  $\mathbf{A}$  are **distinct**.

Given  $x_0 \in \mathbb{C}^n$ ,  $r_0 = b - \mathbf{A}x_0$ .

GMRES computes a sequence of iterates  $x_i$ ,

$$x_i \in x_0 + \mathcal{K}_i(\mathbf{A}, r_0)$$

so that  $r_i = b - \mathbf{A}x_i$  satisfies

$$\|r_i\| = \min_{p \in \pi_i} \|p(\mathbf{A}) r_0\|,$$

where

$$\mathcal{K}_i(\mathbf{A}, r_0) \equiv \text{span} \{r_0, \dots, \mathbf{A}^{i-1} r_0\},$$

$$\pi_i \equiv \{p \text{ is a polynomial; } \deg(p) \leq i; p(0) = 1\}.$$

# Worst-case GMRES residual norm

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Let  $\|r_0\| = 1$ ,  $L \equiv \{\lambda_1, \dots, \lambda_n\}$ . Then

$$\begin{aligned}\|r_i\| &= \min_{p \in \pi_i} \|p(\mathbf{A}) r_0\| \\ &\leq \min_{p \in \pi_i} \|p(\mathbf{A})\| = \min_{p \in \pi_i} \max_{\lambda_j \in L} |p(\lambda_j)|,\end{aligned}$$

This “standard” bound **is sharp**. [Greenbaum, Gurvits-94, Joubert-94]

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**Definition:** An  $i$ th worst-case GMRES residual  $r_i^w$  for  $\mathbf{A} \in \mathbb{C}^{n \times n}$  is a GMRES residual that satisfies

$$\|r_i^w\| = \max_{\|r_0\|=1} \min_{p \in \pi_i} \|p(\mathbf{A}) r_0\|,$$

where  $i = 1, \dots, n - 1$ .

For every GMRES residual  $r_i$  it holds

$$\|r_i\| \leq \|r_i^w\| = \min_{p \in \pi_i} \max_{\lambda_j \in L} |p(\lambda_j)|.$$

## Evaluation of the bound

- Can we describe the standard bound in terms of eigenvalues?

## Relevance of the bound

- How good describes the standard bound the GMRES convergence?

## Other questions

- Which  $r_0$  yields the worst-case residual norm?
- How does the worst-case polynomial look like?

1. Worst-case GMRES residual norm
2. Numerical experiments
3. Relevance of the standard bound
4. A model problem with known eigenvalues
5. Conclusions

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# 1. Worst-case GMRES residual norm



# Factorization of Krylov matrix

**Krylov matrix:**

$$\mathbf{K}_{i+1} \equiv [r_0, \mathbf{A}r_0, \dots, \mathbf{A}^i r_0].$$

We consider  $\mathbf{A}$  and  $r_0$  in the form

$$\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^H, \quad r_0 = \mathbf{Q}[\varrho_1, \dots, \varrho_n]^T.$$

**Factorization:**

$$\mathbf{K}_{i+1} = \mathbf{Q}\mathbf{D}\mathbf{V}_{i+1}$$

where

$$\mathbf{D} \equiv \begin{bmatrix} \varrho_1 & & & \\ & \ddots & & \\ & & \ddots & \\ & & & \varrho_n \end{bmatrix}, \quad \mathbf{V}_{i+1} \equiv \begin{bmatrix} 1 & \lambda_1 & \cdots & \lambda_1^i \\ \vdots & \vdots & & \vdots \\ 1 & \lambda_n & \cdots & \lambda_n^i \end{bmatrix}.$$

Residual  $r_i$  can be written as

[LiRoSt-02, Ipsen-00]

$$\begin{aligned} r_i &= \|r_i\|^2 (\mathbf{K}_{i+1}^+)^H e_1 \\ &= \|r_i\|^2 \mathbf{Q} [(\mathbf{D}\mathbf{V}_{i+1})^+]^H e_1. \end{aligned}$$

and

$$\|r_i\| = \frac{1}{\|[(\mathbf{D}\mathbf{V}_{i+1})^+]^H e_1\|}.$$

( Assumption:  $\mathbf{K}_{i+1}$  has full column rank )

# GMRES residual norm (next-to-last step)

Let  $q_j \neq 0$  for all  $j$ . Then

[Liesen, T.-04, Ipsen-00]

$$\|r_{n-1}\| = \frac{1}{\|\mathbf{D}^{-H} \mathbf{V}_n^{-H} e_1\|} = \frac{1}{\left( \sum_{j=1}^n \left| \frac{l_j(0)}{q_j} \right|^2 \right)^{1/2}},$$

where

$$l_j(\lambda) \equiv \prod_{\substack{k=1 \\ k \neq j}}^n \frac{\lambda_k - \lambda}{\lambda_k - \lambda_j}.$$

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Let  $\|r_0\| = 1$ . Using Cauchy's inequality,

[Liesen, T.-04]

$$\|r_{n-1}^w\| = \frac{1}{\sum_{j=1}^n |l_j(0)|}.$$

For each  $S \subseteq L = \{\lambda_1, \dots, \lambda_n\}$  we denote

$$M_i^S \equiv \min_{p \in \pi_i} \max_{\lambda_j \in S} |p(\lambda_j)|.$$

- **We want** to determine the value  $M_i^L = \|r_i^w\|$ .
- **We are able** to determine

$$M_i^S = \left( \sum_{k=1}^{i+1} |l_k^S(0)| \right)^{-1}, \quad S \subseteq L, \quad |S| = i + 1.$$

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For each subset  $S \subseteq L$  it holds  $M_i^L \geq M_i^S$ , i.e.

$$M_i^L \geq \max_{\substack{S \subseteq L \\ |S|=i+1}} M_i^S \equiv B_i^L.$$

**lower bound**

# Tightness of the bound

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All eigenvalues are real:

[Liesen, T.-04, Greenbaum-79]

$$\|r_i^w\| = M_i^L = B_i^L .$$

At least one non-real eigenvalue:

[Liesen, T.-04]

$$B_i^L \leq \|r_i^w\| \leq \sqrt{(i+1)(n-i)} B_i^L .$$

Numerical Experiments show that  $B_i^L$  is very close to  $\|r_i^w\|$ .

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**We conjecture:**

$$B_i^L \leq \|r_i^w\| \leq C B_i^L ,$$

$C > 1$  is a constant independent on  $i$ ,  $n$  and on eigenvalue distribution



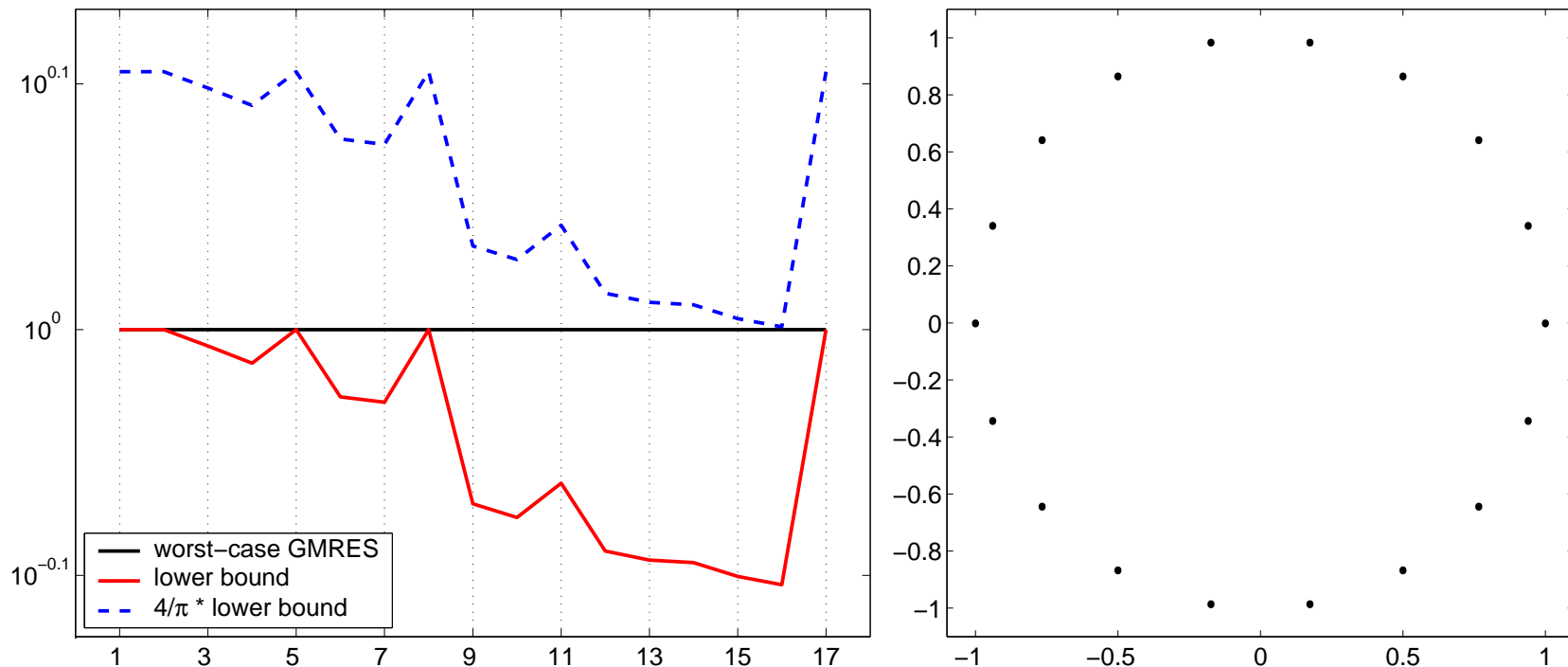
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## 2. Numerical Experiments in MATLAB

# Experiment 1: roots of unity

In this case the worst-case GMRES completely stagnates, i.e.

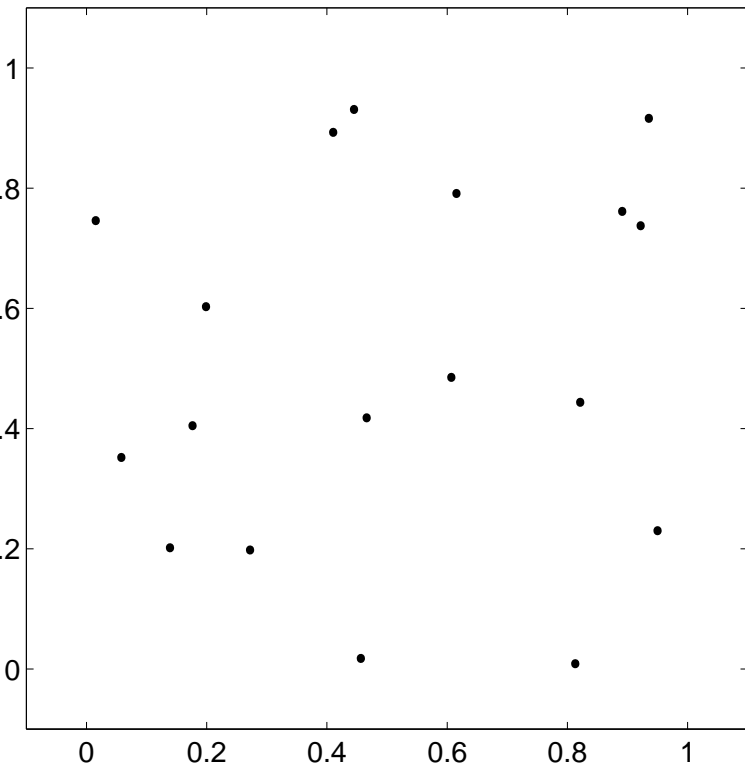
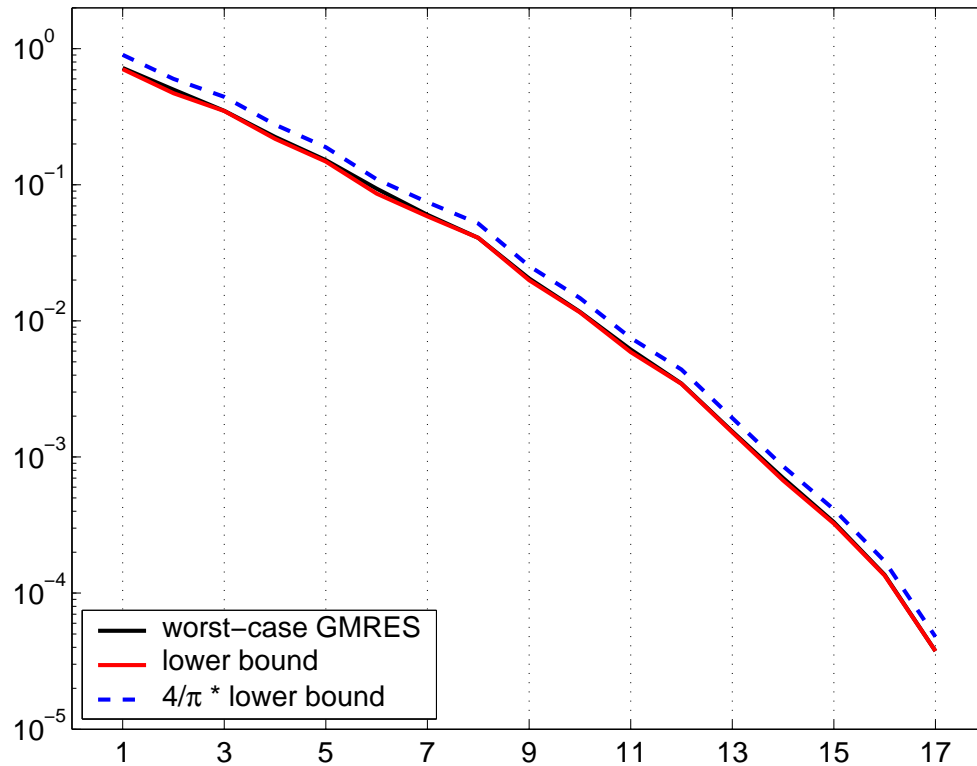
$$1 = \|r_i^w\|, \quad i = 0, \dots, n-1.$$



We proved:  $\|r_{n-2}^w\| < \frac{4}{\pi} B_{n-2}^L, \quad \lim_{n \rightarrow \infty} \left[ \frac{4}{\pi} B_{n-2}^L \right] = \|r_{n-2}^w\|,$

# Experiment 2: random eigenvalues

Random eigenvalues in the region  $[0, 1] \times i [0, 1]$



Numerical experiments predict that

$$B_i^L \leq \|r_i^w\| \leq \frac{4}{\pi} B_i^L$$

holds for all sets  $L$  containing  $n$  distinct complex numbers, where

$$B_i^L \equiv \max_{\substack{S \subseteq L \\ |S|=i+1}} \frac{1}{\sum_{j=1}^{i+1} \prod_{\substack{k=1 \\ k \neq j}}^n \frac{|\lambda_k^S|}{|\lambda_k^S - \lambda_j^S|}} .$$

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### 3. Relevance of the standard bound

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How good describes the standard bound the convergence of GMRES?

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How good describes the standard bound the convergence of GMRES?

It depends on  $r_0$ , in case of  $x_0 = 0$  on the right-hand side  $b$ :

- $b$  is **unbiased**  $\rightarrow b$  has components in the matrix eigenvectors of approximately equal size. Example:  $\rho_i \equiv n^{-1/2}$ ,

$$\|r_i^w\| \geq \|r_i^u\| \geq n^{-1/2} \|r_i^w\|.$$

The standard bound describes the convergence (up to factor  $n^{-1/2}$ ).

- $b$  is **biased**  $\rightarrow$  the standard (worst-case) bound need not describe the convergence (often is an overestimation)

## 4. Model problem



# Model problem: Poisson equation

$$-u''(z) = f(z), \quad z \in (0, 1), \quad u(0) = u_0, \quad u(1) = u_1.$$

The central finite difference approximation on the uniform grid  $kh$ ,  $k = 1, \dots, n$ ,  $h = 1/(n+1)$ , leads to a linear algebraic system  $\mathbf{A}x = b$

$$\overbrace{\begin{bmatrix} 2 & -1 & & & \\ -1 & \ddots & \ddots & & \\ & \ddots & \ddots & \ddots & \\ & & \ddots & \ddots & -1 \\ & & & -1 & 2 \end{bmatrix}}^{\mathbf{A}} x = h^2 \overbrace{\begin{bmatrix} f(h) \\ \vdots \\ \vdots \\ f(nh) \end{bmatrix}}^b + \begin{bmatrix} u_0 \\ \vdots \\ \vdots \\ u_1 \end{bmatrix}.$$

The eigenvalues  $\lambda_k$  and the eigenvectors  $q_k$  of  $\mathbf{A}$  are known.

We can apply MINRES or CG to this system.

## Worst-case × unbiased case (MINRES)

$$\|r_{n-1}^w\| = \frac{1}{n}, \quad \|r_{n-1}^u\| > \sqrt{\frac{2}{3}} \frac{1}{n}.$$

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## Worst data for linear solver (CG)

worst  $\varrho$

worst  $r_0$

data for diff. eq.

$$\begin{bmatrix} \sin(\pi h) \\ \vdots \\ \sin(n\pi h) \end{bmatrix} \leftrightarrow \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \leftrightarrow \begin{aligned} f(z) &= 0, \\ u(0) &= 1, \\ u(1) &= 0. \end{aligned}$$

CG started with  $x_0 = 0$  and  $b = e_1$  attains the worst-case relative  $A$ -norm of the error in the  $(n - 1)$ st iteration step ( $= n^{-1}$ ).

- We completely characterized (for normal matrices):
  - the GMRES residual in the next-to-last step,
  - the worst-case GMRES residual in the next-to-last step.
- The worst-case GMRES norm  $\|r_i^w\|$  can be estimated by  $B_i^L$  :
  - $\|r_i^w\| = B_i^L$  for real eigenvalues,
  - $\|r_i^w\| \geq B_i^L$  in general case.
- Numerical experiments predict that

$$B_i^L \leq \|r_i^w\| \leq \frac{4}{\pi} B_i^L .$$

- The standard bound seems to be reasonable for unbiased  $b$ .
- Our results allow to study model problems with known eigenvalues.
- If  $A$  is SPD, all results transfer to the  $A$ -norm of the error in CG.

## More details can be found in

**Liesen, J. and Tichý, P., The worst-case GMRES for normal matrices**, BIT Numerical Mathematics, 44 (2004), pp. 79–98.

**Liesen, J. and Tichý, P., Behavior of CG and MINRES for symmetric tridiagonal Toeplitz matrices**, in preparation, (2004).

or at

<http://www.math.tu-berlin.de/~tichy>

<http://www.math.tu-berlin.de/~liesen>