

On prescribing Ritz values and GMRES residual norms generated by Arnoldi processes

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joint work with

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Outline

1. Motivation
2. Prescribing convergence behavior for Arnoldi's method
3. Prescribing convergence behavior for *both* Arnoldi's method and GMRES



1. Motivation

Given a nonsingular matrix and nonzero vector

$$A \in \mathbb{C}^{n \times n}, \quad b \in \mathbb{C}^n,$$

the k th iteration of the **Arnoldi orthogonalization process** [Arnoldi - 1951] (without breakdown) computes the decomposition

$$AV_k = V_{k+1} \tilde{H}_k,$$

where the columns of $V_k = [v_1, \dots, v_k]$ (the **Arnoldi vectors**) contain an orthogonal basis for the k th **Krylov subspace**

$$\mathcal{K}_k(A, b) \equiv \text{span}\{b, Ab, \dots, A^{k-1}b\}$$

and $\tilde{H}_k \in \mathbb{C}^{(k+1) \times k}$ is **rectangular upper Hessenberg**; by deleting its last row we get the square matrix

$$H_k = V_k^* AV_k \in \mathbb{C}^{k \times k}.$$



1. Motivation

Essentially,

- for eigenpair approximations of A , the **Arnoldi method** [Arnoldi - 1951], [Saad - 1980] uses the eigenvalues (so-called Ritz values) and eigenvectors of H_k and the first k Arnoldi vectors,
- for approximate solutions to linear systems $Ax = b$, the **GMRES method** [Saad, Schultz - 1986] solves least squares problems with \tilde{H}_k and $\|b\|_{e_1}$ and the first k Arnoldi vectors.
- Both the GMRES and the Arnoldi method are **very popular methods** that are successful for a large variety of problem classes.
- Nevertheless, **convergence behavior** of the two methods is **not fully understood**, analysis is particularly challenging with highly **non-normal** input matrices.



1. Motivation

- Often one tries to **use the tools** that are successful for analysis of hermitian counterparts of GMRES and Arnoldi like the Conjugate Gradient and the Lanczos method.
- For example, the basic tool for explaining Krylov subspace methods for hermitian **linear systems** is the **eigenvalue distribution**.
- However, for GMRES it is known for some time that if GMRES generates a certain residual norm history, the same history can be generated with any nonzero spectrum [Greenbaum, Strakoš - 1994].
- Complemented with the fact that GMRES can generate arbitrary non-increasing residual norms, this gives the result that **any non-increasing convergence curve is possible with any nonzero spectrum** [Greenbaum , Pták, Strakoš - 1996].
- A **complete description of the class** of matrices and right hand sides with prescribed convergence and eigenvalues was given in [Arioli , Pták, Strakoš - 1998].



1. Motivation

An important tool for **hermitian eigenproblems** solved with Krylov subspace methods is the following **interlacing property**:

Consider a tridiagonal Jacobi matrix T_m and its leading principal submatrix T_k for some $k < m$. If the ordered eigenvalues of T_k are

$$\rho_1^{(k)} < \rho_2^{(k)} < \dots < \rho_k^{(k)},$$

then in **every open interval** between two subsequent eigenvalues

$$(\rho_{i-1}^{(k)}, \rho_i^{(k)}), \quad i = 2, \dots, k,$$

there lies **at least one eigenvalue of T_m** .

This interlacing property enables, among others, to prove the **persistence theorem** (see [Paige - 1971, 1976, 1980] or [Meurant, Strakoš - 2006]) which is crucial for controlling the convergence of Ritz values in the Lanczos method.



1. Motivation

- There are **generalizations of the interlacing property to the non-hermitian but normal case** [Fan, Pall - 1957], [Thompson- 1966], [Ericsson - 1990], [Malamud - 2005], though a geometric interpretation is difficult.
- There is **no interlacing property for the eigenvalues of principal submatrices of general non-normal matrices** [de Oliveira - 1969], [Shomron, Parlett - 2009].
- This makes convergence analysis of the Arnoldi method for non-normal input matrices rather delicate, just as it is for the GMRES method.
- The GMRES and Arnoldi methods being closely related through the Arnoldi process, **can we show that arbitrary convergence behavior of Arnoldi is possible?**
- By arbitrary behavior we mean **arbitrary Ritz values for all iterations** (we do not consider eigenvectors). Note that this involves many more conditions than prescribing one residual norm per GMRES iteration.



2. Prescribed convergence for Arnoldi's method

Notation: Let the k th Hessenberg matrix H_k generated in Arnoldi's method have the eigenvalue ρ and normalized eigenvector y ,

$$H_k y = \rho y.$$

With the Arnoldi decomposition $AV_k = V_{k+1}\tilde{H}_k$, the **Ritz pair**

$$\{\rho, V_k y\},$$

i.e. the **Ritz value - Ritz vector** pair has residual norm

$$\|A(V_k y) - \rho(V_k y)\| = \|A(V_k y) - V_k H_k y\| = \|V_{k+1} \tilde{H}_k y - V_k H_k y\| = h_{k+1,k} |e_k^T y|.$$

Often for small $h_{k+1,k} |e_k^T y|$, the Arnoldi method takes $\{\rho, V_k y\}$ as an approximate eigenvalue-eigenvector pair of A . **Note** that a small value $h_{k+1,k} |e_k^T y|$ needs not imply that ρ is close to a true eigenvalue of A , see e.g. [Chatelin - 1993], [Godet-Thobie - 1993]; convergence analysis cannot be based on this value but focusses instead on the quality of approximate invariant subspaces [Beattie, Embree, Sorensen - 2005].



2. Prescribed convergence for Arnoldi's method

Theorem 1 [DT, Meurant - 2012]. Let the set

$$\mathcal{R} = \left\{ \begin{array}{l} \rho_1^{(1)}, \\ (\rho_1^{(2)}, \rho_2^{(2)}), \\ \vdots \\ (\rho_1^{(n-1)}, \dots, \rho_{n-1}^{(n-1)}), \\ (\lambda_1, \dots, \lambda_n) \end{array} \right\},$$

represent any **choice of $n(n+1)/2$ complex Ritz values** and denote by $C^{(k)}$ the **companion matrix** of the polynomial with roots $\rho_1^{(k)}, \dots, \rho_k^{(k)}$, i.e.

$$C^{(k)} = \begin{pmatrix} 0 & \dots & 0 & -\alpha_0 \\ 1 & 0 & \dots & 0 & -\alpha_1 \\ & \ddots & \vdots & \vdots & \\ & & 1 & -\alpha_{k-1} \end{pmatrix}, \quad \prod_{j=1}^k (z - \rho_j^{(k)}) = z^k + \sum_{j=0}^{k-1} \alpha_j z^j.$$



2. Prescribed convergence for Arnoldi's method

If we define the **unit upper triangular matrix** $U(\mathcal{S})$ through

$$U(\mathcal{S}) = I_n - \begin{bmatrix} 0 & C^{(1)}e_1 & \vdots & \vdots \\ & 0 & C^{(2)}e_2 & \vdots \\ & & 0 & \vdots \\ & & & C^{(n-1)}e_{n-1} \\ & & & & 0 \end{bmatrix},$$

then the upper Hessenberg matrix

$$H(\mathcal{R}) = U(\mathcal{S})^{-1}C^{(n)}U(\mathcal{S})$$

has the spectrum $\lambda_1, \dots, \lambda_n$ and its k th leading principal submatrix has spectrum

$$\rho_1^{(k)}, \dots, \rho_k^{(k)}, \quad k = 1, \dots, n - 1.$$

It has **unit subdiagonal**.



2. Prescribed convergence for Arnoldi's method

Proof: The $k \times k$ leading principal submatrix of $H(\mathcal{R})$ is

$$\begin{aligned} [I_k, 0] H(\mathcal{R}) \begin{bmatrix} I_k \\ 0 \end{bmatrix} &= [I_k, 0] U(\mathcal{S})^{-1} C^{(n)} U(\mathcal{S}) \begin{bmatrix} I_k \\ 0 \end{bmatrix} \\ &= [U_k^{-1}, \tilde{u}_{k+1}, \dots, \tilde{u}_n] \begin{bmatrix} 0 \\ U_k \\ 0 \end{bmatrix} = [U_k^{-1}, \tilde{u}_{k+1}] \begin{bmatrix} 0 \\ U_k \end{bmatrix}, \end{aligned}$$

where U_k denotes the $k \times k$ leading principal submatrix of $U(\mathcal{S})$ and \tilde{u}_j denotes the vector of the first k entries of the j th column of $U(\mathcal{S})^{-1}$ for $j > k$. Its spectrum is also the spectrum of the matrix

$$U_k [U_k^{-1}, \tilde{u}_{k+1}] \begin{bmatrix} 0 \\ U_k \end{bmatrix} U_k^{-1} = [I_k, U_k \tilde{u}_{k+1}] \begin{bmatrix} 0 \\ I_k \end{bmatrix},$$

which is a companion matrix with last column $U_k \tilde{u}_{k+1}$.



2. Prescribed convergence for Arnoldi's method

From

$$e_{k+1} = U_{k+1}U_{k+1}^{-1}e_{k+1} = \begin{bmatrix} U_k & -C^{(k)}e_k \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \tilde{u}_{k+1} \\ 1 \end{bmatrix} = \begin{bmatrix} U_k\tilde{u}_{k+1} - C^{(k)}e_k \\ 1 \end{bmatrix}$$

we obtain $U_k\tilde{u}_{k+1} = C^{(k)}e_k$. \square

Remark: The matrix

$$H(\mathcal{R}) = U(\mathcal{S})^{-1}C^{(n)}U(\mathcal{S}).$$

is the *unique* upper Hessenberg matrix $H(\mathcal{R})$ with the prescribed spectrum and Ritz values and **the entry one along the subdiagonal** (see also [Parlett, Strang - 2008] where $H(\mathcal{R})$ is constructed in a different way).

Note that $U(\mathcal{S})$ transforms the matrix $C^{(n)}$ with all Ritz values zero to the matrix $H(\mathcal{R})$ with prescribed Ritz values. It is composed of (columns of) companion matrices and we will call $U(\mathcal{S})$ the **Ritz value companion transform**.



2. Prescribed convergence for Arnoldi's method

Thus the Ritz values generated in **the Arnoldi method can exhibit any convergence behavior**: Apply the Arnoldi method to the pair

$$\{H(\mathcal{R}), e_1\}$$

where $H(\mathcal{R})$ has the desired Ritz values. Then the method generates the Hessenberg matrix $H(\mathcal{R})$ itself.

Question: Can the same prescribed Ritz values be generated with positive entries other than one on the subdiagonal?

For $\sigma_1, \sigma_2, \dots, \sigma_{n-1} > 0$ consider the **diagonal similarity transformation**

$$H \equiv \text{diag} (1, \sigma_1, \sigma_1\sigma_2, \dots, \prod_{j=1}^{n-1} \sigma_j) H(\mathcal{R}) \left(\text{diag} (1, \sigma_1, \sigma_1\sigma_2, \dots, \prod_{j=1}^{n-1} \sigma_j) \right)^{-1}.$$

Then the subdiagonal of H is $\sigma_1, \dots, \sigma_{n-1}$ and all leading principal submatrices of H are similar to corresponding leading principal submatrices of $H(\mathcal{R})$.



2. Prescribed convergence for Arnoldi's method

This immediately leads to a **parametrization** of the matrices and initial Arnoldi vectors that generate a given set of Ritz values \mathcal{R} :

Theorem 2 [DT, Meurant - 2012]. Assume we are given a set of tuples

$$\mathcal{R} = \left\{ \begin{array}{l} \rho_1^{(1)}, \\ (\rho_1^{(2)}, \rho_2^{(2)}), \\ \vdots \\ (\rho_1^{(n-1)}, \dots, \rho_{n-1}^{(n-1)}), \\ (\lambda_1, \dots, \lambda_n) \end{array} \right\},$$

of complex numbers and $n - 1$ positive real numbers

$$\sigma_1, \dots, \sigma_{n-1}.$$

If A is a matrix of order n and b a nonzero n -dimensional vector, then the following assertions are equivalent:



2. Prescribed convergence for Arnoldi's method

1. The Hessenberg matrix generated by the Arnoldi method applied to A and initial Arnoldi vector b has eigenvalues $\lambda_1, \dots, \lambda_n$, subdiagonal entries $\sigma_1, \dots, \sigma_{n-1}$ and $\rho_1^{(k)}, \dots, \rho_k^{(k)}$ are the eigenvalues of its k th leading principal submatrix for all $k = 1, \dots, n - 1$.
2. The matrix A and initial vector b are of the form

$$A = VD_\sigma U(\mathcal{S})^{-1}C^{(n)}U(\mathcal{S})D_\sigma^{-1}V^*, \quad b = \|b\|Ve_1,$$

where V is unitary, $U(\mathcal{S})$ is the Ritz value companion transform,

$$D_\sigma = \text{diag}(1, \sigma_1, \sigma_1\sigma_2, \dots, \prod_{j=1}^{n-1} \sigma_j),$$

and $C^{(n)}$ is the companion matrix of the polynomial with roots $\lambda_1, \dots, \lambda_n$.

This also shows how little on the quality of the Ritz value ρ needs be said by

$$\|A(V_k y) - \rho(V_k y)\| = h_{k+1,k} |e_k^T y|.$$

Any distance from ρ to the spectrum of A is possible with **any** value of $h_{k+1,k}$!



2. Prescribed convergence for Arnoldi's method

Counterintuitive example 1: Convergence of interior Ritz values only:

$$\mathcal{R} = \{ \begin{array}{l} 3, \\ (3, 3), \\ (2, 3, 4), \\ (3, 3, 3, 3), \\ (1, 2, 3, 4, 5) \end{array} \}.$$

This gives the unit upper Hessenberg matrix

$$H(\mathcal{R}) = U(\mathcal{S})^{-1}C^{(5)}U(\mathcal{S}) = \begin{bmatrix} 3 & 0 & 0 & 0 & 0 \\ 1 & 3 & 1 & 0 & 1 \\ & 1 & 3 & -1 & 0 \\ & & 1 & 3 & 5 \\ & & & 1 & 3 \end{bmatrix}.$$



2. Prescribed convergence for Arnoldi's method

Thus these Ritz values are generated by the Arnoldi method applied to

$$A = V \operatorname{diag} (1, \sigma_1, \sigma_1 \sigma_2, \dots, \prod_{j=1}^{n-1} \sigma_j) \begin{bmatrix} 3 & 0 & 0 & 0 & 0 \\ 1 & 3 & 1 & 0 & 1 \\ & 1 & 3 & -1 & 0 \\ & & 1 & 3 & 5 \\ & & & 1 & 3 \end{bmatrix} \operatorname{diag} (1, \sigma_1, \sigma_1 \sigma_2, \dots, \prod_{j=1}^{n-1} \sigma_j)^{-1} V^*,$$

with initial vector $b = V e_1$ and for any unitary V and positive values $\sigma_1, \dots, \sigma_{n-1}$.

This is [not a highly non-normal example](#), for instance with $\sigma_i \equiv 1$:

$$\|A\| \|A^{-1}\| = 9.7137,$$

and the eigenvector basis W of A has condition number

$$\|W\| \|W^{-1}\| = 4.8003.$$



2. Prescribed convergence for Arnoldi's method

Counterintuitive example 2: We can prescribe the “diverging” Ritz values

$$\mathcal{R} = \{ \begin{array}{l} 1, \\ (0, 2), \\ (-1, 1, 3), \\ (-2, 0, 2, 4), \\ (1, 1, 1, 1, 1) \end{array} \},$$

with $H(\mathcal{R}) = U(\mathcal{S})^{-1} C^{(5)} U(\mathcal{S}) = \begin{bmatrix} 1 & 1 & 0 & -3 & 0 \\ 1 & 1 & 3 & 0 & -31 \\ & 1 & 1 & 6 & 0 \\ & & 1 & 1 & -10 \\ & & & 1 & 1 \end{bmatrix}.$

They are generated by Arnoldi applied to $A = VH(\mathcal{R})V^*$, $b = Ve_1$ for unitary V .



2. Prescribed convergence for Arnoldi's method

The same “diverging” Ritz values are generated with the **exponentially decreasing values** 2^{-1} , 2^{-2} , 2^{-3} and 2^{-4} on the subdiagonal of the Hessenberg matrix:

$$A = V \begin{bmatrix} 1 & 2 & 0 & -192 & 0 \\ 0.5 & 1 & 12 & 0 & -15872 \\ & 0.25 & 1 & 48 & 0 \\ & & 0.125 & 1 & -160 \\ & & & 0.0625 & 1 \end{bmatrix} V^*, \quad b = \|b\| V e_1.$$

Then the rounded residual norms $\|A(V_k y) - \rho(V_k y)\| = h_{k+1,k} |e_k^T y|$ **seem to indicate convergence:**

$$\left\{ \begin{array}{l} \frac{1}{2}, \\ (0.1118, 0.1118), \\ (0.011, 0.0052, 0.011), \\ (0.0006, 0.0001, 0.0001, 0.0006) \end{array} \right\}.$$



3. Prescribed convergence for Arnoldi *and* GMRES

Starting with initial guess $x_0 = 0$, GMRES iterates x_k minimize the residual norm,

$$\|b - Ax_k\| = \min \|b - As\| \quad \text{over all } s \in \mathcal{K}_k(A, b).$$

Writing x_k in the Arnoldi basis,

$$x_k = V_k y_k \in \mathcal{K}_k(A, r_0),$$

and using the Arnoldi decomposition $AV_k = V_{k+1} \tilde{H}_k$, we see that the residual norm is

$$\begin{aligned} \|b - Ax_k\| &= \|b - AV_k y_k\| = \|V_{k+1} \|b\| e_1 - AV_k y_k\| \\ &= \|V_{k+1} (\|b\| e_1 - \tilde{H}_k y_k)\| = \min_{y \in \mathbb{C}^k} \|\|b\| e_1 - \tilde{H}_k y\|. \end{aligned}$$

Thus the residual norms generated by the GMRES method are fully determined by the Hessenberg matrix \tilde{H}_k and $\|b\|$.



3. Prescribed convergence for Arnoldi *and* GMRES

- We have seen that the subdiagonal entries of \tilde{H}_k can be chosen arbitrarily, for any prescribed Ritz values in the k th iteration.
- Hence there is a chance we can modify the behavior of GMRES while maintaining the prescribed Ritz values.

Example from earlier: Consider the prescribed 'diverging' Ritz values

$$\mathcal{R} = \left\{ \begin{array}{l} 1, \\ (0, 2), \\ (-1, 1, 3), \\ (-2, 0, 2, 4), \\ (1, 1, 1, 1, 1) \end{array} \right\},$$

and the prescribed subdiagonal entries of the generated Hessenberg matrix

$$\sigma_1 = 2^{-1}, \quad \sigma_2 = 2^{-2}, \quad \sigma_3 = 2^{-3}, \quad \sigma_4 = 2^{-4}.$$



3. Prescribed convergence for Arnoldi *and* GMRES

The corresponding GMRES convergence curve is

$$\|r^{(0)}\| = 1, \quad \|r^{(1)}\| = \sqrt{\frac{1}{5}}, \quad \|r^{(2)}\| = \sqrt{\frac{1}{5}}, \quad \|r^{(3)}\| = 0.0052, \quad \|r^{(4)}\| = 0.0052.$$

Question: Can we force any GMRES convergence speed with arbitrary Ritz values by modifying the subdiagonal entries?

Not **any**, because there is a relation between GMRES stagnation and zero Ritz values: A **singular Hessenberg matrix corresponds to stagnation** in the parallel GMRES process, see [Brown - 1991]. In our example we have

$$\begin{aligned} \rho_1^{(1)} &= 1, & \|r^{(1)}\| &= \frac{1}{\sqrt{5}} \\ (\rho_1^{(2)}, \rho_2^{(2)}) &= (0, 2), & \|r^{(2)}\| &= \frac{1}{\sqrt{5}} \\ (\rho_1^{(3)}, \rho_2^{(3)}, \rho_3^{(3)}) &= (-1, 1, 3), & \|r^{(3)}\| &= 0.0052 \\ (\rho_1^{(4)}, \rho_2^{(4)}, \rho_3^{(4)}, \rho_4^{(4)}) &= (-2, 0, 2, 4), & \|r^{(4)}\| &= 0.0052. \end{aligned}$$



3. Prescribed convergence for Arnoldi *and* GMRES

However, this is **the *only* restriction Ritz values put on GMRES** residual norms:

Theorem 3 [DT, Meurant - 2012]. Consider a set of tuples of complex numbers

$$\mathcal{R} = \left\{ \begin{array}{l} \rho_1^{(1)}, \\ (\rho_1^{(2)}, \rho_2^{(2)}), \\ \vdots \\ (\rho_1^{(n-1)}, \dots, \rho_{n-1}^{(n-1)}), \\ (\lambda_1, \dots, \lambda_n) \end{array} \right\},$$

such that $(\lambda_1, \dots, \lambda_n)$ contains no zero number and n positive numbers

$$f(0) \geq f(1) \geq \dots \geq f(n-1) > 0,$$

such that the k -tuple $(\rho_1^{(k)}, \dots, \rho_k^{(k)})$ **contains a zero number if and only if**

$$f(k-1) = f(k).$$



3. Prescribed convergence for Arnoldi *and* GMRES

Let A be a square matrix of size n and let b be a nonzero n -dimensional vector. The following assertions are equivalent:

1. The GMRES method applied to A and right-hand side b with zero initial guess yields **residuals** $r^{(k)}$, $k = 0, \dots, n - 1$ such that

$$\|r^{(k)}\| = f(k), \quad k = 0, \dots, n - 1,$$

A has **eigenvalues**

$$\lambda_1, \dots, \lambda_n,$$

and

$$\rho_1^{(k)}, \dots, \rho_k^{(k)}$$

are **the Ritz values generated at the k th iteration** for $k = 1, \dots, n - 1$.



3. Prescribed convergence for Arnoldi *and* GMRES

2. The matrix A and right hand side b are of the form

$$A = V \text{diag}(f(0), D_c^{-*}) U(\mathcal{S})^{-1} C^{(n)} U(\mathcal{S}) \text{diag}(f(0)^{-1}, D_c^*) V^*, \quad b = \|b\| V e_1,$$

where V is a unitary matrix, $U(\mathcal{S})$ is the **Ritz value companion transform** for \mathcal{R} and $C^{(n)}$ is the companion matrix of the polynomial with roots $\lambda_1, \dots, \lambda_n$. D_c is a nonsingular **diagonal matrix such that**

$$R_h^{-T} \hat{h} = -f(0)^2 D_c c,$$

$$\hat{h} = [\eta_1, \dots, \eta_{n-1}]^T, \quad \eta_k = (f(k-1)^2 - f(k)^2)^{1/2},$$

R_h being the upper triangular factor of the Cholesky decomposition

$$R_h^T R_h = I - \frac{\hat{h} \hat{h}^T}{f(0)^2},$$

and c is the first row of $U(\mathcal{S})$ without its diagonal entry.

Note we exhausted all freedom modulo unitary transformation.



3. Prescribed convergence for Arnoldi *and* GMRES

Example: Standardly **converging** Ritz values and '**nearly stagnating**' GMRES:

$$\mathcal{R} = \{ \quad 5, \\ \quad \quad (1, 5), \\ \quad \quad (1, 4, 5), \\ \quad \quad (1, 3, 4, 5), \\ \quad \quad (1, 2, 3, 4, 5) \},$$

$\|r^{(0)}\| = 1, \quad \|r^{(1)}\| = 0.9, \quad \|r^{(2)}\| = 0.8, \quad \|r^{(3)}\| = 0.7, \quad \|r^{(4)}\| = 0.6, \quad \|r^{(5)}\| = 0$ gives

$$A = V \begin{bmatrix} 5 & 0 & 0 & 0 & 0 \\ 10.3237 & 1 & 0 & 0 & 0 \\ & 0.8458 & 4 & 0 & 0 \\ & & 3.312 & 3 & 0 \\ & & & 2.4169 & 2 \end{bmatrix} V^*, \quad b = \|b\| V e_1.$$



3. Prescribed convergence for Arnoldi *and* GMRES

Again, this is **not a highly non-normal example**:

$$\|A\| \|A^{-1}\| = 28.9498,$$

and the eigenvector basis W of A has condition number

$$\|W\| \|W^{-1}\| = 57.735.$$

The **residual norms** $\|A(V_k y) - \rho(V_k y)\| = h_{k+1,k} |e_k^T y|$ for the Ritz pairs are

$$10.3237,$$

$$(0.8458, 0.7886),$$

$$(0.8987, 3.312, 2.0509),$$

$$(0.9906, 2.4169, 2.3137, 1.7303) .$$

respectively, i.e. they **give misleading information**.



Conclusions and future work:

- There is **no interlacing property** for the Hessenberg matrices in the Arnoldi method.
- The Ritz values generated in the **Arnoldi method can behave arbitrarily badly**.
- Convergence of **Ritz values need not say anything about the behavior of GMRES** residual norms (zero Ritz values excepted). For close to normal matrices, the opposite has been suggested [van der Vorst, Vuik - 1993].
- Extension to **harmonic Ritz values** which determine the GMRES polynomials?
- It is desirable to have similar results for **popular restarted versions** of GMRES and the Arnoldi method (see e.g. the failure of restarted Arnoldi with exact shifts explained in [Embree - 2009]).



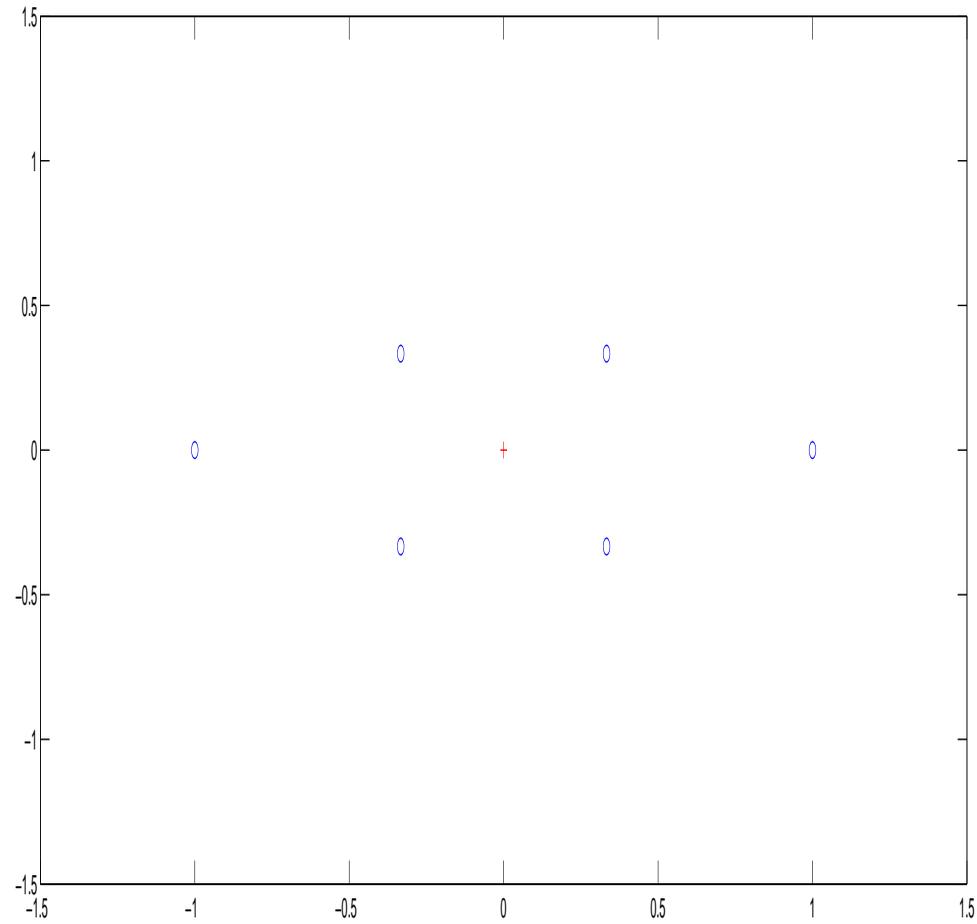
For **more details** see: DUINTJER TEBBENS J, MEURANT G: *Any Ritz value behavior is possible for Arnoldi and for GMRES*, SIMAX vol. 33 no. 3, pp 958–978, 2012, available at www.cs.cas.cz/duintjertebbens/duintjertebbens_pub.html

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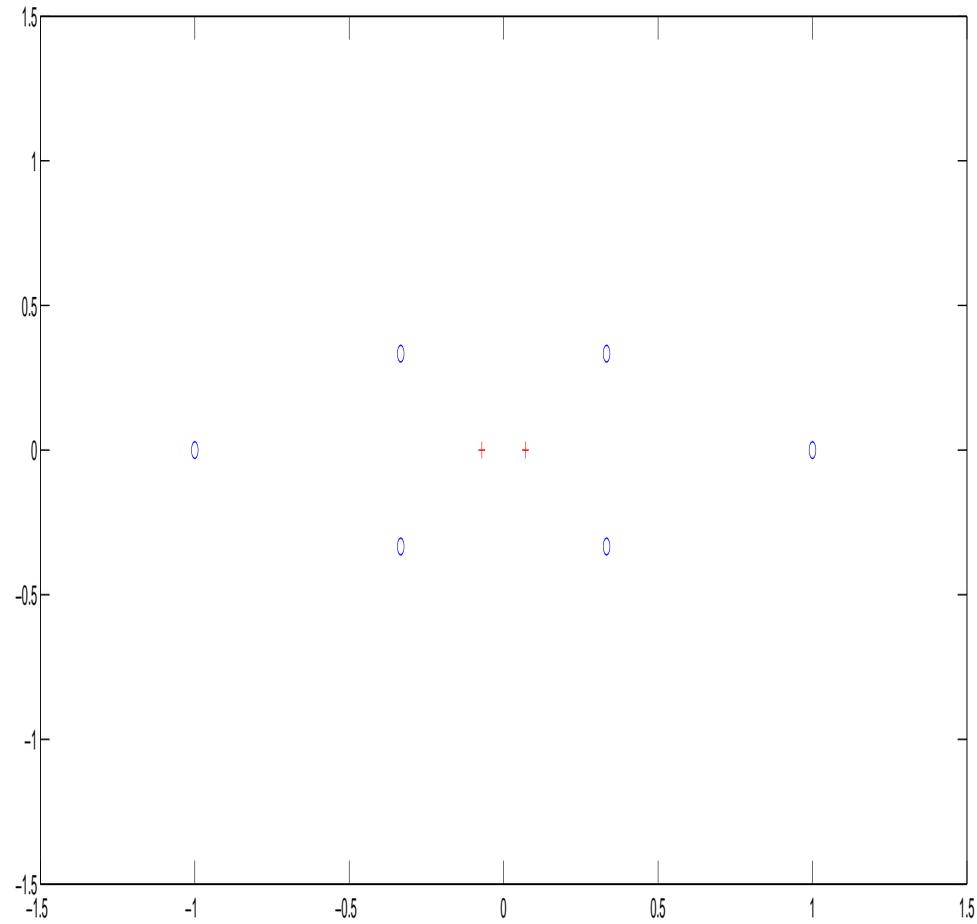


Counterintuitive normal example



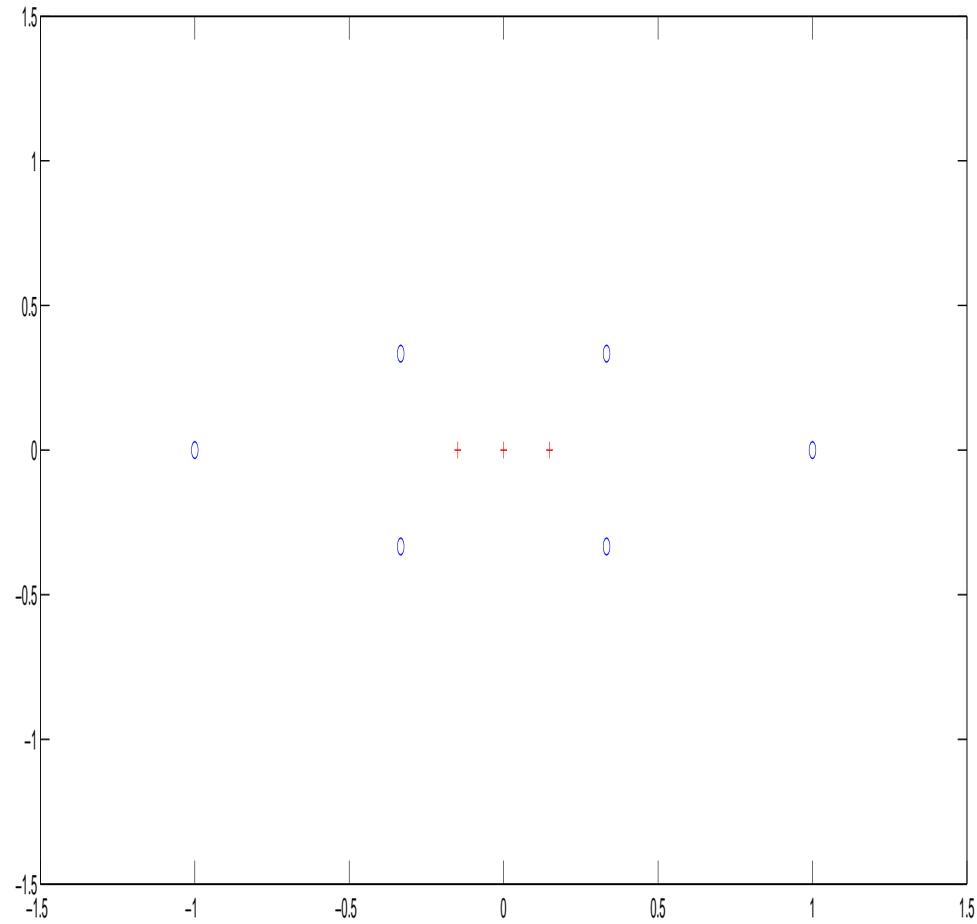


Counterintuitive normal example



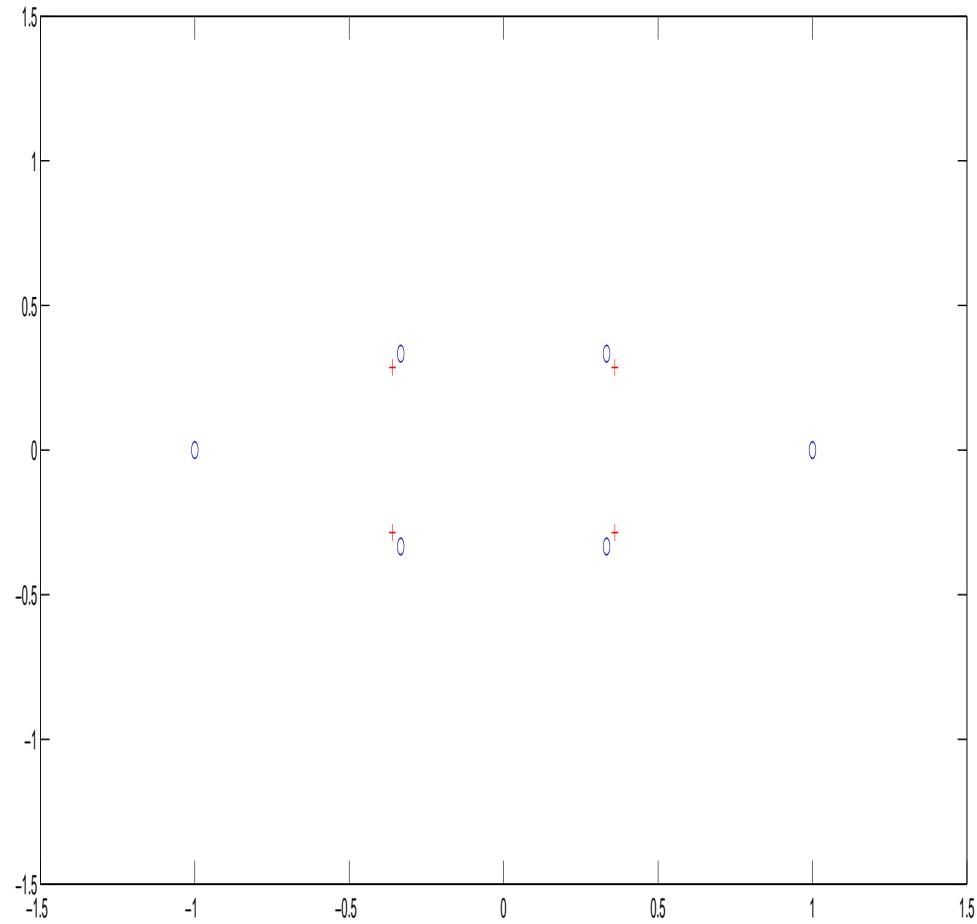


Counterintuitive normal example





Counterintuitive normal example





Counterintuitive normal example

