# On the convergence of Q-OR and Q-MR Krylov methods for solving linear systems

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Many Krylov methods have been proposed over the years for solving linear systems

Most of them can be classified as quasi-orthogonal (Q-OR) or quasi-minimum residual (Q-MR)

Q-OR: FOM, BiCG, Hessenberg, ...

Q-MR: GMRES, QMR, CMRH, ...

Whatever their definition, these methods share many fundamental properties

See the nice paper by M. Eiermann and O.G. Ernst, Geometric aspects in the theory of Krylov subspace methods, Acta Numerica, v 10 n 10 (2001), pp. 251–312

The methods differ by the basis of the Krylov space that is constructed:

- orthogonal for FOM/GMRES (true OR/MR methods)
- bi-orthogonal for BiCG/QMR
- based on an LU factorization for Hessenberg/CMRH

Our aim is to show that some results about GMRES convergence can be extended to other Q-Q-MR methods

#### **GMRES**

GMRES uses the Arnoldi process to construct an orthonormal basis of the Krylov susbpsace

$$\mathcal{K}_n(A,b) = \{b \mid Ab \mid \cdots \mid A^{n-1}b\}$$

Assume the basis vectors are linearly independent. Then,

$$AV = VH, \quad V^*V = I,$$

and H is (unreduced) upper Hessenberg

With  $x_0 = 0$ , the GMRES iterates  $x_k = V_k y_k$  are computed by solving

$$\min_{x_k \in \mathcal{K}_k(A,b)} \|b - Ax_k\|$$

with  $V_k$   $n \times k$ , k first columns of V

### What do we know about GMRES?

Let

$$K = \begin{pmatrix} b & Ab & A^2b & \cdots & A^{n-1}b \end{pmatrix}$$

be the Krylov matrix that we assume of full rank. Then

$$K = VU$$

with V orthogonal (or unitary) and U upper triangular with positive real diagonal entries

As we know, the matrix  $H = V^*AV$  is upper Hessenberg

We have

$$H = UCU^{-1}$$

where C is the companion matrix for the eigenvalues of A

[This is a consequence of AK = KC]

Let  $x_k^G$  (resp.  $x_k^F$ ) be the iterates for GMRES (resp. FOM) and the residual vectors  $r_k^G = b - Ax_k^G$  (resp.  $r_k^F = b - Ax_k^F$ ) We assume  $x_0 = 0$  and ||b|| = 1

#### We know that

- every non-increasing residual norm convergence curve is possible for GMRES
- one can construct matrices *A* with a prescribed spectrum and right-hand sides *b* such that GMRES yields a prescribed decreasing residual norm convergence curve. In addition one can prescribe the Ritz values for all iterations
- a first parametrization of this class of matrices and right-hand sides was introduced by Arioli, Pták and Strakoš

#### For these properties see

- A. Greenbaum and Z. Strakoš, *Matrices that generate the same Krylov residual spaces*, in Recent advances in iterative methods, G.H. Golub, A. Greenbaum and M. Luskin, eds., Springer, (1994), pp. 95–118
- A. Greenbaum, V. Pták and Z. Strakoš, Any nonincreasing convergence curve is possible for GMRES, SIAM J. Matrix Anal. Appl., v 17 (1996), pp. 465–469
- M. Arioli, V. Pták and Z. Strakoš, *Krylov sequences of maximal length and convergence of GMRES*, BIT Numerical Mathematics, v 38 n 4 (1998), pp. 636–643
- J. Duintjer Tebbens and G. Meurant, Any Ritz value behavior is possible for Arnoldi and for GMRES, SIAM J. Matrix Anal. Appl., v 33 n 3 (2012), pp. 958–978

## Another parametrization, JDT-GM

Assume we are given n positive numbers

$$1 = f_0 \ge f_1 \ge \cdots \ge f_{n-1} > 0$$

and n complex numbers  $\lambda_1, \ldots, \lambda_n$  all different from 0. Let A be a matrix of order n and b an n-dimensional vector of unit norm. The following assertions are equivalent:

1- The spectrum of A is  $\{\lambda_1, \ldots, \lambda_n\}$  and GMRES applied to A and b yields residuals  $r_j^G$ ,  $j=0,\ldots,n-1$  such that

$$||r_j^G|| = f_j, \quad j = 0, \dots, n-1$$



2- The matrix A is of the form  $A = VUCU^{-1}V^*$  and  $b = Ve_1$ , where V is any unitary matrix, U is nonsingular upper triangular such that

$$U_{1,1}^{-1} = 1$$
,  $U_{1,j}^{-1} = \left(\frac{1}{f_{j-1}^2} - \frac{1}{f_{j-2}^2}\right)^{\frac{1}{2}}$ ,  $j = 2, \dots, n$ 

and *C* is the companion matrix corresponding to the prescribed eigenvalues

This type of parametrization can also be used to prescribe all the Ritz values at every iteration

From the relation between FOM and GMRES residual norms we have

$$-|(U^{-1})_{1,k}|=1/||r_{k-1}^F||$$

Moreover

- 
$$\|r_k^G\|^2 = 1/(M_{k+1}^{-1})_{1,1}$$
 with  $M_{k+1} = U_{k+1}^* U_{k+1}$ 

This last result has been proved by several people: Stewart, Zitko, Ipsen, Liesen, Rozložník and Strakoš, and Sadok

To compute  $(M_{k+1}^{-1})_{1,1}$ , following ideas from H. Sadok, we use two simple tools:

- ► Cramer's rule (1750 but known before that)
- ► The Cauchy-Binet formula (1812) for det(AB) with A and B rectangular

## Diagonalizable matrices

Let 
$$A = X\Lambda X^{-1}$$
 and  $c = X^{-1}b$ . Then

$$K = X (c \land c \lor \land \land^{n-1}c)$$

Therefore

$$M = K^*K = (c \land c \lor \land \land^{n-1}c)^*X^*X(c \land c \lor \land \land^{n-1}c)$$

and

$$M_{k+1} = \mathcal{V}_{k+1}^* D_{\bar{c}} X^* X D_c \mathcal{V}_{k+1}$$

with  $D_c$  diagonal with  $c_i$  as diagonal entries and ...

$$\mathcal{V}_{k+1} = \begin{pmatrix} 1 & \lambda_1 & \cdots & \lambda_1^k \\ 1 & \lambda_2 & \cdots & \lambda_2^k \\ \vdots & \vdots & & \vdots \\ 1 & \lambda_n & \cdots & \lambda_n^k \end{pmatrix}$$

an  $n \times (k+1)$  Vandermonde matrix

Then using Cramer's rule for the first column of the inverse and twice the Cauchy-Binet formula we obtain the following (complicated but exact) result for the GMRES residual norms

#### The residual norms

Let A be a diagonalizable matrix with  $A = X \Lambda X^{-1}$ ,  $c = X^{-1}b$ . Then

$$||r_k^G||^2 = \sigma_{k+1}^N / \sigma_k^D$$

with

$$\sigma_{k+1}^{N} = \sum_{I_{k+1}} \left| \sum_{J_{k+1}} \det(X_{I_{k+1}, J_{k+1}}) c_{j_{1}} \cdots c_{j_{k+1}} \prod_{j_{1} \le j_{l} < j_{p} \le j_{k+1}} (\lambda_{j_{p}} - \lambda_{j_{l}}) \right|^{2}$$

$$\sigma_{1}^{D} = \sum_{i=1}^{n} \left| \sum_{j=1}^{n} X_{i, j} c_{j} \lambda_{j} \right|^{2}$$

and

$$\sigma_k^D = \sum_{l_k} \left| \sum_{J_k} \det(\mathsf{X}_{l_k,J_k}) c_{j_1} \cdots c_{j_k} \, \lambda_{j_1} \cdots \lambda_{j_k} \prod_{j_1 \leq j_l < j_p \leq j_k} (\lambda_{j_p} - \lambda_{j_l}) \right|^2, \ k > 1$$

where the summations are over all sets of indices  $I_{k+1}, J_{k+1}, I_k, J_k$  defined as  $I_\ell$  to be a set of  $\ell$  indices  $(i_1, i_2, \ldots, i_\ell)$  such that  $1 \leq i_1 < \cdots < i_\ell \leq n$ ,  $X_{I_\ell, J_\ell}$  is the submatrix of X whose row and column indices are defined by  $I_\ell$  and  $J_\ell$ 

If the matrix A is normal, we have  $X^*X = I$  and simpler formulas

$$\sigma_{k+1}^{N} = \sum_{I_{k+1}} |c_{j_1}|^2 \cdots |c_{j_{k+1}}|^2 \prod_{j_1 \leq j_l < j_p \leq j_{k+1}} |(\lambda_{j_p} - \lambda_{j_l})|^2$$

$$\sigma_1^D = \sum_{i=1}^n |c_i|^2 |\lambda_j|^2$$

and

$$\sigma_k^D = \sum_{I_k} |c_{j_1}|^2 \cdots |c_{j_k}|^2 \, |\lambda_{j_1}|^2 \cdots |\lambda_{j_k}|^2 \prod_{j_1 \leq j_l < j_p \leq j_k} \left| (\lambda_{j_p} - \lambda_{j_l}) 
ight|^2, \,\, k > 1$$

See the paper by J. Duintjer Tebbens, GM, H. Sadok and Z. Strakoš, LAA v 450 (2014)

The results for diagonalizable matrices can somehow be extended to the case of non-diagonalizable matrices using the Jordan canonical form

In particular we can obtain nice expressions of the residual norms for one Jordan block



### Q-OR and Q-MR methods

Can we extend some of these results to Q-OR and Q-MR methods?

We assume that we have an ascending basis V of the Krylov space (with columns of unit norm) such that K = VU with V nonsingular and U upper triangular

We define  $H = UCU^{-1}$ . As a consequence AV = VH. The iterates are

$$x_k = V_k y_k$$

where  $V_k$  is the matrix of the k first columns of V. The residual  $r_k$  is

$$V_k e_1 - AV_k y_k = V_k (e_1 - H_k y_k) - h_{k+1,k} (y_k)_k v_{k+1} = V_{k+1} (e_1 - \underline{H}_k y_k)$$



The Q-OR method is defined (provided that  $H_k$  is nonsingular) by

$$H_k y_k^O = e_1$$

where  $H_k$  is the principal submatrix of order k. This annihilates the first term in the residual

In the Q-MR method  $y_k^M$  is computed as the solution of the least squares problem

$$\min_{y} \|e_1 - \underline{H}_k y\|$$

where  $\underline{H}_k$  is  $(k+1) \times k$ 

The vector  $\mathbf{z}_k^M = \mathbf{e}_1 - \underline{H}_k \mathbf{y}_k^M$  is referred as the quasi-residual

The residual vector is  $r_k^M = V_{k+1} z_k^M$ 

Generally, the two problems are solved using Givens rotations with sines  $s_i$  and cosines  $c_i$ . It is known that

$$||z_k^M|| = |s_1 s_2 \cdots s_k|$$

Moreover we have a relation between the Q-OR residual norms and the Q-MR quasi-residual norms

$$\frac{1}{\|r_k^O\|^2} = \frac{1}{\|z_k^M\|^2} - \frac{1}{\|z_{k-1}^M\|^2}$$

See Eiermann and Ernst (2001), Freund and Nachtigal (1991)

## Properties of Q-OR and Q-MR methods

From these results we can show that

$$|(U^{-1})_{1,k}| = \frac{1}{\|r_{k-1}^O\|}$$

This is proved by using the rotation matrices  $G_i$  such that

$$G_{n-1}\cdots G_1H=\mathcal{R}$$

From  $H = UCU^{-1}$  we have

$$U^{-1}G_1^{-1}\cdots G_{n-1}^{-1}=CU^{-1}\mathcal{R}^{-1}$$

The first row of the matrix on the right-hand side is zero except from the last entry

Looking at the product of the inverses of the rotation matrices, we can show by induction that

$$U_{1,j+1}^{-1} = \frac{c_j}{s_1 \cdots s_j} = \pm \left( \frac{1}{(s_1 \cdots s_j)^2} - \frac{1}{(s_1 \cdots s_{j-1})^2} \right)^{1/2}, \ j = 1, \dots, n-1$$

Let  $M_{k+1} = U_{k+1}^* U_{k+1}$ . Since  $M_{k+1}^{-1} = U_{k+1}^{-1} U_{k+1}^{-*}$  and from the first row of  $U^{-1}$ , a consequence of the previous result is the following

$$||z_k^M||^2 = \frac{1}{(M_{k+1}^{-1})_{1,1}}$$

The difference with GMRES is that we only have the norm of the quasi-residual

Then, we obtain expressions of the quasi-residual norms

## The quasi-residual norms

Let A be a diagonalizable matrix with  $A = X\Lambda X^{-1}$ ,  $Z = V^{-1}X$  and  $c = X^{-1}b$ . Then

$$\|z_k^M\|^2 = \sigma_{k+1}^N / \sigma_k^D$$

with

$$\sigma_{k+1}^{N} = \sum_{l_{k+1}} \left| \sum_{J_{k+1}} \det(Z_{l_{k+1},J_{k+1}}) c_{j_{1}} \cdots c_{j_{k+1}} \prod_{j_{1} \leq j_{l} < j_{p} \leq j_{k+1}} (\lambda_{j_{p}} - \lambda_{j_{l}}) \right|^{2}$$

$$\sigma_{1}^{D} = \sum_{i=1}^{n} \left| \sum_{j=1}^{n} Z_{i,j} c_{j} \lambda_{j} \right|^{2}$$

and

$$\sigma_k^D = \sum_{l_k} \left| \sum_{J_k} \det(Z_{l_k,J_k}) c_{j_1} \cdots c_{j_k} \, \lambda_{j_1} \cdots \lambda_{j_k} \prod_{j_1 \leq j_l < j_p \leq j_k} (\lambda_{j_p} - \lambda_{j_l}) \right|^2, \ k > 1$$

where the summations are over all sets of indices  $I_{k+1}, J_{k+1}, I_k, J_k$  where  $I_\ell$  is a set of  $\ell$  indices  $(i_1, i_2, \ldots, i_\ell)$  such that  $1 \leq i_1 < \cdots < i_\ell \leq n$ ,  $Z_{I_\ell, J_\ell}$  is the submatrix of Z whose row and column indices are defined by  $I_\ell$  and  $J_\ell$ 



This result arises from  $\|z_k^M\|^2 = 1/(M_{k+1}^{-1})_{1,1}$  and

$$M = U^*U = K^*V^{-*}V^{-1}K$$
  
=  $(c \land c \lor \cdots \land^{n-1}c)^*Z^*Z(c \land c \lor \cdots \land^{n-1}c)$ 

It yields

$$M_{k+1} = \mathcal{V}_{k+1}^* D_{\bar{c}} Z^* Z D_c \mathcal{V}_{k+1}$$

where  $D_c$  is diagonal and  $V_{k+1}$  is an  $n \times (k+1)$  Vandermonde matrix

As for GMRES, we compute the (1,1) entry of the inverse using Cramer's rule and the Cauchy-Binet determinant formula

Note that there is no simplification when A is normal

## Construction of linear systems with a prescribed convergence curve

Can we construct linear systems with a prescribed convergence curve and a prescribed spectrum for Q-OR and Q-MR methods?

For FOM/GMRES this is easy since we just have to construct an upper triangular matrix  $U^{-1}$  with the inverses of the FOM residual norms (obtained from the GMRES norms) on the first row. Then we take

$$A = VUCU^{-1}V^*, \qquad b = Ve_1$$

where C is the companion matrix of the given eigenvalues and V is any unitary matrix

Things are more difficult for some Q-Q-MR methods because we may ask for some non-zero structure in H



#### **BiCG**

We would like to find matrices H (with a given spectrum) and U such that

$$H = \begin{pmatrix} \gamma_1 & \beta_2 & 0 & 0 & 0\\ \rho_2 & \gamma_2 & \beta_3 & 0 & 0\\ 0 & \ddots & \ddots & \ddots & 0\\ 0 & 0 & \rho_{n-1} & \gamma_{n-1} & \beta_n\\ 0 & 0 & 0 & \rho_n & \gamma_n \end{pmatrix} = UCU^{-1}$$

and the first row of  $U^{-1}$  is prescribed as  $\begin{pmatrix} 1 & g_1 & \cdots & g_{n-1} \end{pmatrix}$  with  $g_j \neq 0$ 

Let  $\omega_2, \ldots, \omega_n$  be arbitrary chosen entries of the last column of  $U^{-1}$  with  $\omega_n \neq 0$ ,  $\omega_1 = g_{n-1}$  and  $-\alpha_0, \ldots, -\alpha_{n-1}$  be the entries of the last column of C that we know from the prescribed spectrum

We compute  $U^{-1}$  and H recursively column-wise The last column of  $U^{-1}H = CU^{-1}$  yields

$$\begin{pmatrix} g_{n-2}\beta_n + g_{n-1}\gamma_n \\ \vdots \\ \omega_n\gamma_n \end{pmatrix} = \begin{pmatrix} -\alpha_0\omega_n \\ \vdots \\ \omega_{n-1} - \alpha_{n-1}\omega_n \end{pmatrix}$$

We use the first and last equations

$$\begin{pmatrix} g_{n-2} & g_{n-1} \\ 0 & \omega_n \end{pmatrix} \begin{pmatrix} \beta_n \\ \gamma_n \end{pmatrix} = \begin{pmatrix} -\alpha_0 \omega_n \\ \omega_{n-1} - \alpha_{n-1} \omega_n \end{pmatrix}$$

The solution of this  $2 \times 2$  non singular linear system yields  $\gamma_n, \beta_n$ 

From the other equations that we discarded we can compute the unknown entries  $\nu_{j,n-1}$  of column n-1 of  $U^{-1}$ 

Then we go on backwards with column n-1

We have three unknowns  $\beta_{n-1}, \gamma_{n-1}$  and  $\rho_n$ We first take the first and the last two equations This gives us a linear system with an upper triangular matrix

$$\begin{pmatrix} g_{n-3} & g_{n-2} & g_{n-1} \\ 0 & \nu_{n-1,n-1} & \omega_{n-1} \\ 0 & 0 & \omega_n \end{pmatrix} \begin{pmatrix} \beta_{n-1} \\ \gamma_{n-1} \\ \rho_n \end{pmatrix} = \begin{pmatrix} 0 \\ \nu_{n-2,n-1} \\ \nu_{n-1,n-1} \end{pmatrix}$$

And so on... Then  $A = VHV^{-1}$  and  $b = Ve_1$  for an appropriately chosen matrix V

So far we don't know how to completely handle the case with zero entries on the first row of  $U^{-1}$ 

This algorithm can be extended to a larger upper bandwidth but what about stability?

This allows to prescribe BiCG residual norm convergence (or QMR quasi-residual norms)



## Summary

Many known properties of FOM/GMRES are also valid for general Q-OR/Q-MR methods

We express the Q-MR quasi-residual norms as functions of the eigenvalues, the eigenvectors, the right-hand side and the basis of the Krylov space

We (almost) have a parametrization of the class of matrices with a prescribed spectrum and a prescribed Q-OR/Q-MR convergence curve

In particular we can construct examples with a BiCG (finite) convergence curve