

IDR: A new generation of Krylov subspace methods?

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Outline

Krylov subspace methods

- Hessenberg decompositions

- Polynomial representations

- Perturbations

IDR

- IDR and IDR(s)

- IDREIG

- IDR(s)STAB(ℓ) and IDRSTABEIG

- (Flexible and multi-shift) QMRIDR

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Introduction

Krylov subspace methods: approximations

$$\left. \begin{array}{l} \mathbf{x}_k, \underline{\mathbf{x}}_k, \\ \mathbf{y}_k, \underline{\mathbf{y}}_k \end{array} \right\} \in \mathcal{K}_k(\mathbf{A}, \mathbf{q}) := \text{span} \{ \mathbf{q}, \mathbf{A}\mathbf{q}, \dots, \mathbf{A}^{k-1}\mathbf{q} \} = \{ p(\mathbf{A})\mathbf{q} \mid p \in \mathbb{P}_{k-1} \},$$

where

$$\mathbb{P}_{k-1} := \left\{ \sum_{j=0}^{k-1} \alpha_j z^j \mid \alpha_j \in \mathbb{C}, 0 \leq j < k \right\},$$

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to solutions of linear systems

$$\mathbf{Ax} = \mathbf{r}_0 \quad (= \mathbf{b} - \mathbf{Ax}_0), \quad \mathbf{A} \in \mathbb{C}^{n \times n}, \quad \mathbf{b}, \mathbf{x}_0 \in \mathbb{C}^n,$$

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and (partial) eigenproblems

$$\mathbf{Av} = \mathbf{v}\lambda, \quad \mathbf{A} \in \mathbb{C}^{n \times n}.$$

Hessenberg decompositions

Construction of basis vectors resembled in structure of arising **Hessenberg decomposition**

$$\mathbf{AQ}_k = \mathbf{Q}_{k+1} \underline{\mathbf{H}}_k,$$

where

- ▶ $\mathbf{Q}_{k+1} = (\mathbf{Q}_k, \mathbf{q}_{k+1}) \in \mathbb{C}^{n \times (k+1)}$ collects basis vectors,
- ▶ $\underline{\mathbf{H}}_k \in \mathbb{C}^{(k+1) \times k}$ is unreduced extended Hessenberg.

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Aspects of perturbed Krylov subspace methods: captured with perturbed Hessenberg decompositions

$$\mathbf{AQ}_k + \mathbf{F}_k = \mathbf{Q}_{k+1} \underline{\mathbf{H}}_k,$$

$\mathbf{F}_k \in \mathbb{C}^{n \times k}$ accounts for perturbations (finite precision & inexact methods).

Karl Hessenberg & “his” matrix + decomposition



„Behandlung linearer Eigenwertaufgaben mit Hilfe der Hamilton-Cayleyschen Gleichung“, Karl Hessenberg, 1. Bericht der Reihe „Numerische Verfahren“, July, 23rd 1940, page 23:

Man kann nun die Vektoren $\mathbf{z}_v^{(n-v)}$ ($v = 1, 2, \dots, n$) ebenfalls in einer Matrix zusammenfassen, und zwar ist nach Gleichung (55) und (56)

$$(57) \quad (\mathbf{z}_1 \ \mathbf{z}_2 \ \mathbf{z}_3 \ \cdots \ \mathbf{z}_n^{(n-n)}) = \mathbf{A} \cdot \mathbf{z}' = \mathbf{z}' \cdot \mathbf{P},$$

worin die Matrix \mathbf{P} zur Abkürzung gesetzt ist für

$$(58) \quad \mathbf{P} = \begin{pmatrix} \alpha_{1,0} & \alpha_{2,0} & \cdots & \alpha_{n-1,0} & \alpha_{n,0} \\ 1 & \alpha_{2,1} & \cdots & \alpha_{n-1,1} & \alpha_{n,1} \\ 0 & 1 & \cdots & \alpha_{n-1,2} & \alpha_{n,2} \\ 0 & 0 & \cdots & 1 & \alpha_{n,n} \end{pmatrix}.$$

- ▶ Hessenberg decomposition, Eqn. (57),
- ▶ Hessenberg matrix, Eqn. (58).

Karl Hessenberg (* September 8th, 1904, † February 22nd, 1959)

Important Polynomials

Residuals of OR and MR approximation

$$\mathbf{x}_k := \mathbf{Q}_k \mathbf{z}_k \quad \text{and} \quad \underline{\mathbf{x}}_k := \mathbf{Q}_k \underline{\mathbf{z}}_k$$

with coefficient vectors

$$\mathbf{z}_k := \mathbf{H}_k^{-1} \mathbf{e}_1 \|\mathbf{r}_0\| \quad \text{and} \quad \underline{\mathbf{z}}_k := \underline{\mathbf{H}}_k^\dagger \mathbf{e}_1 \|\mathbf{r}_0\|$$

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$$\mathcal{R}_k(z) := \det(\mathbf{I}_k - z \mathbf{H}_k^{-1}) \quad \text{and} \quad \underline{\mathcal{R}}_k(z) := \det(\mathbf{I}_k - z \underline{\mathbf{H}}_k^\dagger \mathbf{I}_k).$$

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Convergence of OR and MR depends on (harmonic) Ritz values.

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In perturbed case

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polynomial representation

$$\mathbf{r}_k = \mathcal{R}_k(\mathbf{A})\mathbf{r}_0 - \sum_{\ell=1}^k z_{\ell k} \mathcal{R}_{\ell+1:k}(\mathbf{A})\mathbf{f}_\ell + \mathbf{F}_k \mathbf{z}_k$$

(all trailing square Hessenberg matrices are assumed to be regular).

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Convergence: $\mathbf{F}_k \mathbf{z}_k$ bounded (inexact methods) & $\mathcal{R}_{\ell+1:k}(\mathbf{A})$ “small”.

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- 1979 First talk on IDR
- 1980 Proceedings
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- 1992 IDR \rightsquigarrow BICGSTAB
- 1993 BICGSTAB2, BICGSTAB(ℓ)
- later “acronym explosion” . . .

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IDR(s) is a Krylov subspace method \rightsquigarrow all techniques from 90’s applicable!

IDR(s)

IDR spaces:

$\mathcal{G}_0 := \mathcal{K}(\mathbf{A}, \mathbf{q})$, (full Krylov subspace)

$\mathcal{G}_j := (\alpha_j \mathbf{A} + \beta_j \mathbf{I})(\mathcal{G}_{j-1} \cap \mathcal{S}), \quad j \geq 1, \quad \alpha_j, \beta_j \in \mathbb{C}, \quad \alpha_j \neq 0,$

where

$\text{codim}(\mathcal{S}) = s, \quad \text{e.g.,} \quad \mathcal{S} = \text{span } \{\tilde{\mathbf{R}}_0\}^\perp, \quad \tilde{\mathbf{R}}_0 \in \mathbb{C}^{n \times s}.$

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Interpreted as **Sonneveld spaces** (Sleijpen, Sonneveld, van Gijzen 2010):

$$\mathcal{G}_j = \mathcal{S}_j(P_j, \mathbf{A}, \tilde{\mathbf{R}}_0) := \left\{ P_j(\mathbf{A})v \mid v \perp \mathcal{K}_j(\mathbf{A}^H, \tilde{\mathbf{R}}_0) \right\},$$

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Image of shrinking space: **Induced Dimension Reduction**.

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Generalized Hessenberg decomposition:

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where $\mathbf{U}_k \in \mathbb{C}^{k \times k}$ upper triangular.

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Structure of Sonneveld pencils:

$$\mathbf{H}_k = \begin{pmatrix} \times \times \times \times \circ \circ \circ \circ \circ \circ \\ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \circ \circ + \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \circ \circ \circ + \times \times \circ \circ \circ \circ \circ \circ \circ \\ \circ + \times \circ \circ \circ \circ \circ \circ \circ \circ \end{pmatrix}, \quad \begin{pmatrix} \times \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \circ \circ \times \times \times \circ \circ \circ \circ \circ \circ \\ \circ \circ \circ \circ \circ \circ \circ \circ \circ \times \times \circ \circ \circ \circ \circ \circ \circ \end{pmatrix} = \mathbf{U}_k$$

IDREig

Eigenvalues of **Sonneveld pencil** ($\mathbf{H}_k, \mathbf{U}_k$) are roots of residual polynomials.
Those distinct from roots of

$$P_j(z) = \prod_{i=1}^j (\alpha_i z + \beta_i), \quad \text{i.e.,} \quad z_i = -\frac{\beta_i}{\alpha_i}, \quad 1 \leq i \leq j$$

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First: IDR(s)ORES, Olaf Rendel: IDR(s)BIO, Anisa Rizvanollie: IDR(s)STAB(ℓ).

IDRSTAB

IDR(s)STAB(ℓ) (Tani & Sugihara; Sleijpen & van Gijzen): combine ideas of IDR(s) and BICGSTAB(ℓ).

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IDR^{STAB} (Sleijpen's implementation) recursively computes “(extended) Hessenberg matrices of basis matrices and residuals” ($k \geq 1$):

$$\begin{matrix} \mathbf{G}_{11}^{(k)}, \mathbf{r}_{11}^{(k)} & \mathbf{G}_{12}^{(k)}, \mathbf{r}_{12}^{(k)} & \cdots & \mathbf{G}_{1,\ell+1}^{(k)}, & \mathbf{r}_{1,\ell+1}^{(k)} \\ \mathbf{G}_{21}^{(k)}, \mathbf{r}_{21}^{(k)} & \mathbf{G}_{22}^{(k)}, \mathbf{r}_{22}^{(k)} & \cdots & \mathbf{G}_{2,\ell+1}^{(k)}, & \mathbf{r}_{2,\ell+1}^{(k)} \\ & \mathbf{G}_{32}^{(k)}, \mathbf{r}_{32}^{(k)} & \ddots & & \vdots \\ & & \ddots & \mathbf{G}_{\ell+1,\ell+1}^{(k)}, & \mathbf{r}_{\ell+1,\ell+1}^{(k)} \\ & & & \mathbf{G}_{\ell+2,\ell+1}^{(k)} & \end{matrix}$$

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Initialization using Arnoldi's method:

$$\mathbf{G}_{21}^{(1)} = \mathbf{A}\mathbf{G}_{11}^{(1)} = (\mathbf{G}_{11}^{(1)}, \mathbf{g}_{\text{tmp}})\underline{\mathbf{H}}_s^{(0)},$$

$$\mathbf{r}_{11}^{(1)} = \mathbf{r}_0 - \mathbf{G}_{21}^{(1)} \boldsymbol{\alpha}^{(1)} = (\mathbf{I} - \mathbf{G}_{21}^{(1)}(\tilde{\mathbf{R}}_0^\top \mathbf{G}_{21}^{(1)})^{-1} \tilde{\mathbf{R}}_0^\top) \mathbf{r}_0, \quad \mathbf{r}_{21}^{(1)} = \mathbf{A}\mathbf{r}_{11}^{(1)}.$$

IDRSTAB

Columnwise update (**IDR part**) such that diagonal blocks

- ▶ form basis of $\mathcal{G}_j \setminus \mathcal{G}_{j+1}$ with expansion $\mathcal{G}_j = \mathbf{A}(\mathcal{G}_{j-1} \cap \mathcal{S}) \rightsquigarrow \boldsymbol{\beta}^{(j)} \in \mathbb{C}^{s \times s}$,
- ▶ are orthonormalized $\rightsquigarrow \underline{\mathbf{H}}_{s-1}^{(j)} \in \mathbb{C}^{s \times s-1}$

All other blocks in column treated in same manner.

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New cycle (**STAB part**, $\mathbf{r}_{21}^{(k+1)} = \mathbf{A}\mathbf{r}_{11}^{(k+1)}$, $\boldsymbol{\gamma}^{(\ell)} \in \mathbb{C}^s$ such that $\|\mathbf{r}_{11}^{(k+1)}\| = \min$):

$$\mathbf{r}_{11}^{(k+1)} = \mathbf{r}_{1,\ell+1}^{(k)} - \sum_{i=1}^{\ell} \mathbf{r}_{i+1,\ell+1}^{(k)} \gamma_i^{(\ell)}, \quad \begin{cases} \mathbf{G}_{11}^{(k+1)} = \mathbf{G}_{1,\ell+1}^{(k)} - \sum_{i=1}^{\ell} \mathbf{G}_{i+1,\ell+1}^{(k)} \gamma_i^{(\ell)}, \\ \mathbf{G}_{21}^{(k+1)} = \mathbf{G}_{2,\ell+1}^{(k)} - \sum_{i=1}^{\ell} \mathbf{G}_{i+2,\ell+1}^{(k)} \gamma_i^{(\ell)}. \end{cases}$$

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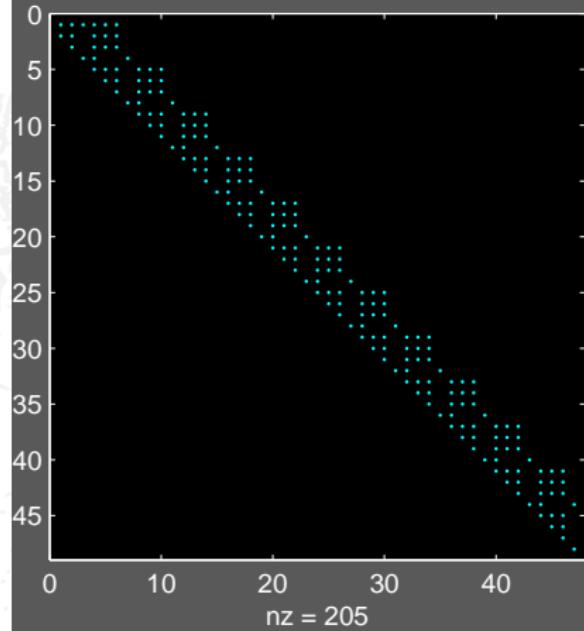
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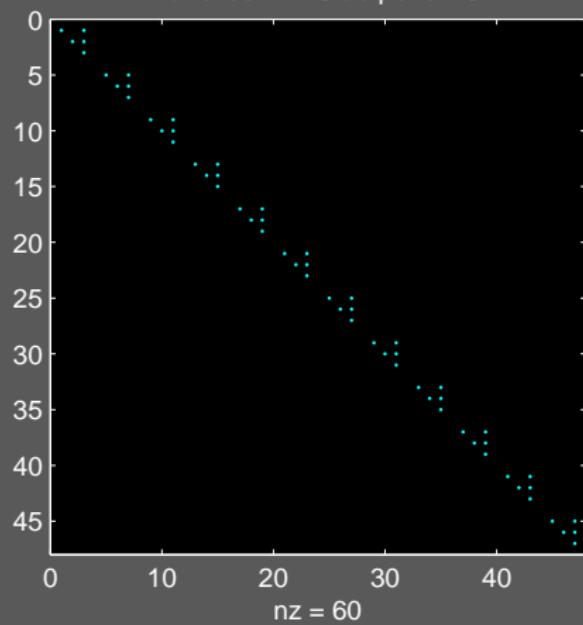
Anisa Rizvanolli: \rightsquigarrow Lanczos-IDRSTAB pencil for eigenvalues, IDRSTABEIG.

Structure of (undeflated) Lanczos-IDRStab pencil

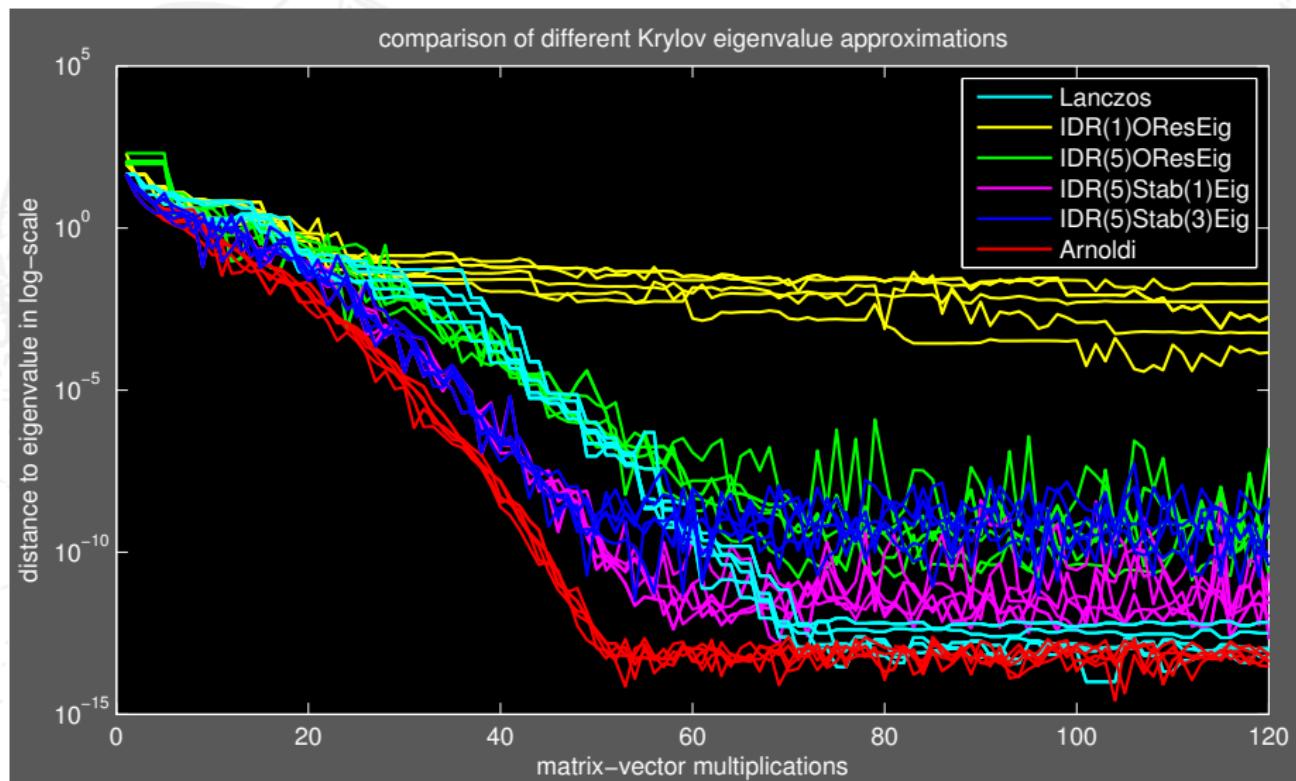
Lanczos-IDRStab pencil: uH



Lanczos-IDRStab pencil: U



A comparison: IDR based eigenvalue solvers



QMRIDR

MR methods: use extended Hessenberg matrix

$$\underline{\mathbf{x}}_k := \mathbf{Q}_k \mathbf{z}_k, \quad \underline{\mathbf{z}}_k := \underline{\mathbf{H}}_k^\dagger \underline{\mathbf{e}}_1 \|\mathbf{r}_0\|.$$

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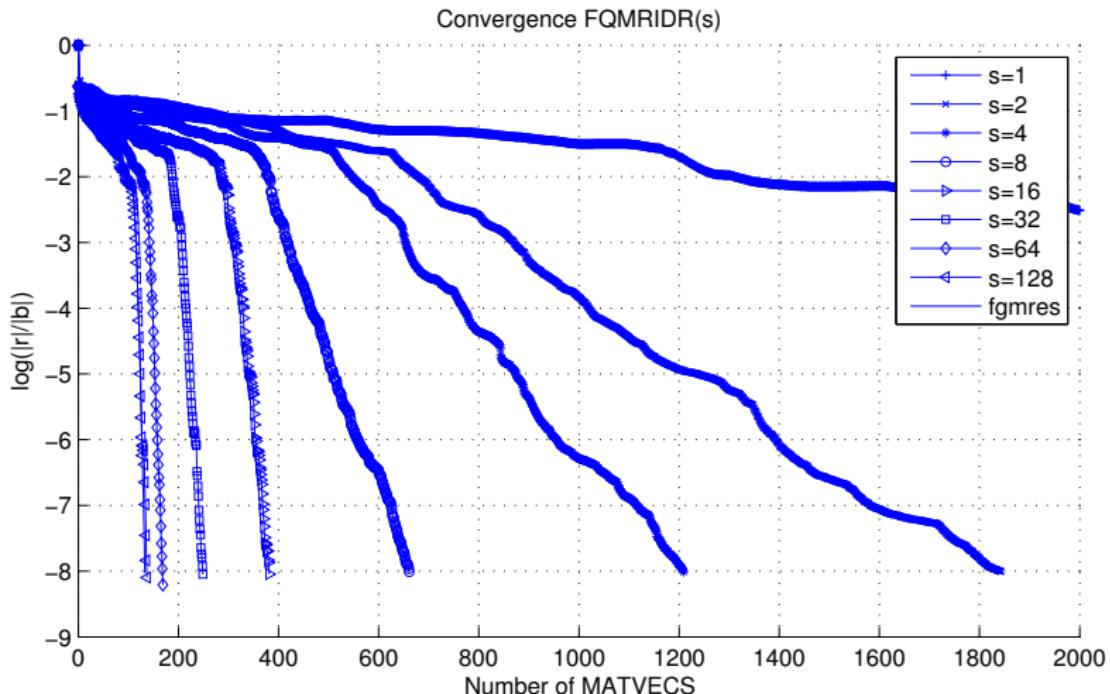
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Olaf Rendel, Gerard Sleijpen, Martin van Gijzen: \rightsquigarrow QMRIDRStab.

Flexible QMRIDR(s)



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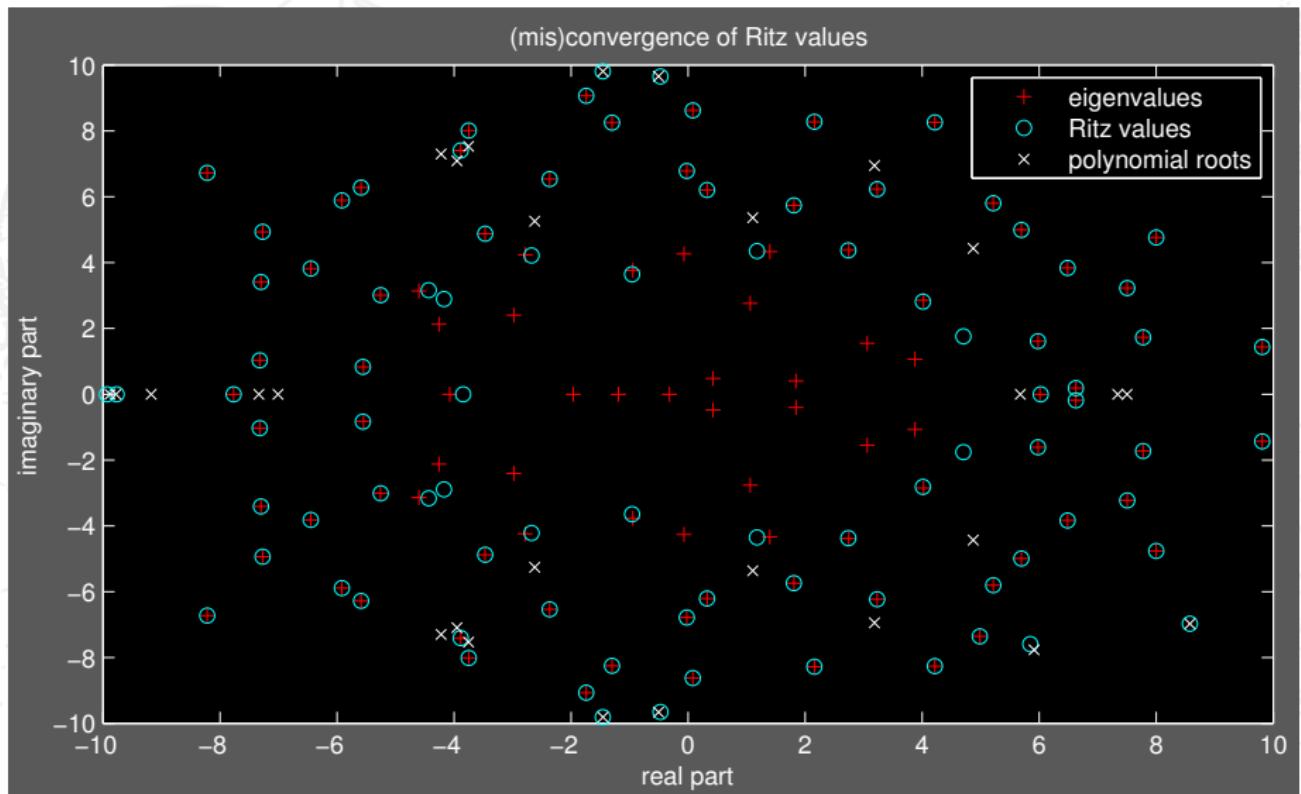
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But:

- ▶ IDR transpose-free,
- ▶ easy to implement,
- ▶ more stable (for large values of s),
- ▶ often close to “optimal” methods (for large values of s).

IDR(3)STAB(3): “Ghost polynomial roots”



Conclusion and Outview

- ▶ IDR is both **old** (original IDR, CGS, BICGSTAB, BICGSTAB2, BICGSTAB(ℓ), ...) and **new** (IDR(s), IDRSTAB, QMRIDR, ...).

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ILAS related:

- ▶ The analysis & development of IDR based methods is a **new branch of Krylov subspace methods**.
- ▶ The pencils of IDR based methods are **specially structured pencils** (adapted backward stable algorithms; perturbation theory, ...).

Thank you for your attention!

In case of questions feel free to ask Anisa, Olaf & myself at any time.

This talk is partially based on the following technical reports:

Eigenvalue computations based on IDR, Martin H. Gutknecht and Z., Bericht 145, Institut für Numerische Simulation, TUHH, 2010,

Flexible and multi-shift induced dimension reduction algorithms for solving large sparse linear systems, Martin B. van Gijzen, Gerard L.G. Sleijpen, and Z., Bericht 156, Institut für Numerische Simulation, TUHH, 2011.