

# Deflation techniques for two classes of structured linear systems

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joint work with **Jia-Jun Fan** and **Fang Wang**

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# Outline

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- ① Linear systems, Krylov subspace methods, and deflation
- ② Nonsymmetric positive definite linear systems
- ③ Symmetric quasi-definite linear systems
- ④ Summary

# Linear systems, Krylov subspace methods, and deflation

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- Linear systems of equations

$$\mathbf{Ax} = \mathbf{b}, \quad \mathbf{A} \in \mathbb{R}^{m \times m}, \quad \mathbf{b} \in \mathbb{R}^m$$

- Krylov subspace methods

CG, MINRES; GMRES, CMRH, Bi-CG, QMR, Bi-CGSTAB ...

- Acceleration techniques

preconditioning, randomization, inexact or mixed-precision, inner-product free (orthogonalization-free), **deflation** ...

- When solving linear systems, deflation refers to reducing the influence of some eigenvalues that tend to slow convergence. Deflation can be implemented
  - (1) by **adding approximate eigenvectors to a subspace**, or
  - (2) by building a preconditioner from eigenvectors.

## Some solvers incorporating augmentation-based deflation

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- Nonsymmetric linear systems: Arnoldi + augmentation + restart  
FOM-IR, GMRES-IR (Morgan, SIMAX, 2000), FOM-DR, GMRES-DR  
(Morgan, SISC, 2002)

$$\mathbf{A}\mathbf{V}_m^{(i)} = \mathbf{V}_{m+1}^{(i)}\mathbf{H}_{m+1,m}^{(i)}, \quad i = 1, 2, \dots$$

- Symmetric linear systems: Lanczos + augmentation + restart  
Lanczos-DR, MINRES-DR (Abdel-Rehim et al., SISC, 2010)

$$\mathbf{A}\mathbf{V}_m^{(i)} = \mathbf{V}_{m+1}^{(i)}\mathbf{T}_{m+1,m}^{(i)}, \quad i = 1, 2, \dots$$

- Symmetric saddle point linear systems: Golub–Kahan + augmentation + restart  
Augmented LSQR (Baglama, Reichel, and Richmond, NA, 2013)  
Augmented CRAIG (Dumitrasc, Kruse, and Rde, SIMAX, 2024)

$$\mathbf{A}\mathbf{V}_m^{(i)} = \mathbf{U}_{m+1}^{(i)}\mathbf{B}_{m+1,m}^{(i)}, \quad \mathbf{A}^\top\mathbf{U}_{m+1}^{(i)} = \mathbf{V}_{m+1}^{(i)}(\mathbf{B}_{m+1}^{(i)})^\top, \quad i = 1, 2, \dots$$

## Nonsymmetric positive definite linear systems

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- The symmetric and skew-symmetric splitting

$$\mathbf{A} = \mathbf{H} + \mathbf{S}, \quad \mathbf{H} := \frac{1}{2}(\mathbf{A} + \mathbf{A}^\top), \quad \mathbf{S} := \frac{1}{2}(\mathbf{A} - \mathbf{A}^\top).$$

Assume that the symmetric part  $\mathbf{H}$  of  $\mathbf{A}$  is SPD.

- Stationary iterative methods  
HSS ([Bai, Golub, and Ng, SIMAX, 2003](#)) ...
- Krylov subspace methods based on the skew-symmetric Lanczos process  
CGW ([Concus and Golub, 1976](#), [Widlund, 1978](#)) : Galerkin condition  
Rapoport's method ([Rapoport, 1978](#)): MR condition

# Nonsymmetric positive definite linear systems

- Skew-symmetric Lanczos

$$\mathbf{S}\mathbf{U}_m = \mathbf{H}\mathbf{U}_{m+1}\mathbf{T}_{m+1,m}, \quad \mathbf{T}_{m+1,m} = \begin{bmatrix} 0 & -\gamma_2 & & & \\ \gamma_2 & 0 & \ddots & & \\ & \ddots & \ddots & -\gamma_m & \\ & & \gamma_m & 0 & \\ & & & \gamma_{m+1} & \end{bmatrix} = \begin{bmatrix} \mathbf{T}_m \\ \gamma_{m+1}\mathbf{e}_m^\top \end{bmatrix}.$$

- Note that  $\lambda(\mathbf{H}^{-1}\mathbf{S}) = \{\pm\sigma_1\mathbf{i}, \pm\sigma_2\mathbf{i}, \dots, \pm\sigma_{r/2}\mathbf{i}, 0\}$  with  $r = \text{rank}(\mathbf{H}^{-1}\mathbf{S})$  and  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_{r/2} > 0$ .

$$\text{CGW convergence: } \frac{\|\mathbf{x}_{2j} - \mathbf{A}^{-1}\mathbf{b}\|_{\mathbf{H}}}{\|\mathbf{x}_0 - \mathbf{A}^{-1}\mathbf{b}\|_{\mathbf{H}}} \leq 2 \left( \frac{\sqrt{1 + \sigma_1^2} - 1}{\sqrt{1 + \sigma_1^2} + 1} \right)^j$$

$$\text{Rapoport convergence: } \frac{\|\mathbf{b} - \mathbf{A}\mathbf{x}_j\|_{\mathbf{H}^{-1}}}{\|\mathbf{b} - \mathbf{A}\mathbf{x}_0\|_{\mathbf{H}^{-1}}} \leq 2 \left( \frac{\sigma_1}{\sqrt{1 + \sigma_1^2} + 1} \right)^j$$

# Skew-symmetric Lanczos with deflated restarting

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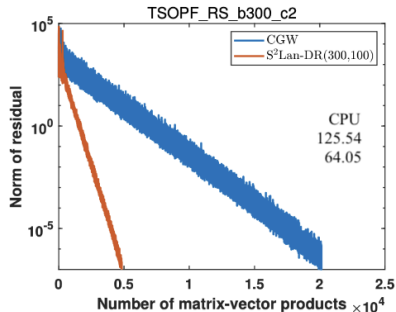
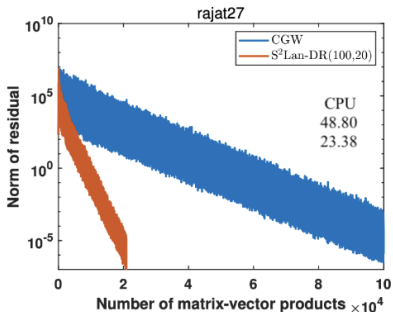
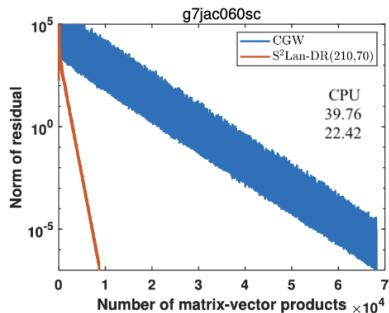
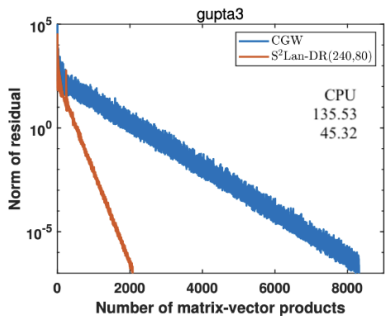
- $S^2\text{Lan-DR}(m, 2k)$

$$\mathbf{S}\mathbf{U}_m^{(i)} = \mathbf{H}\mathbf{U}_{m+1}^{(i)} \mathbf{T}_{m+1,m}^{(i)}$$

For example,  $m = 8$ ,  $k = 2$ ,  $i = 2, 3, \dots$ ,

$$\mathbf{T}_{m+1,m}^{(i)} = \begin{bmatrix} 0 & \times & & & \times & & & \\ \times & 0 & & & \times & & & \\ & & 0 & \times & \times & & & \\ & & \times & 0 & \times & & & \\ \times & \times & \times & \times & 0 & \times & & \\ & & & \times & 0 & \times & & \\ & & & & \times & 0 & \times & \\ & & & & & \times & 0 & \times \\ & & & & & & \times & 0 \end{bmatrix}, \quad \mathbf{T}_m^{(i)} = -(\mathbf{T}_m^{(i)})^\top.$$

# CGW vs. $S^2$ Lan-DR( $m, 2k$ )





# Rapoport with deflated restarting

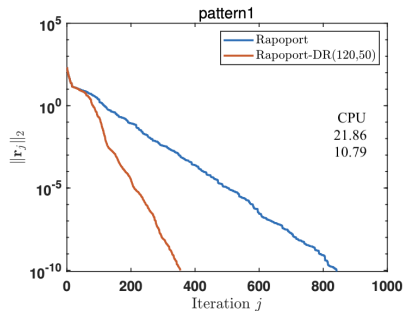
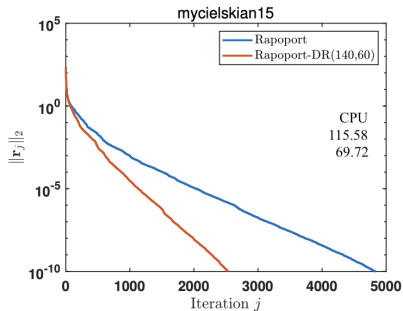
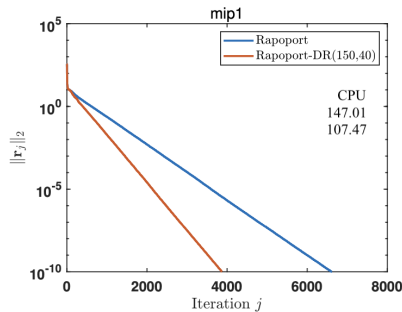
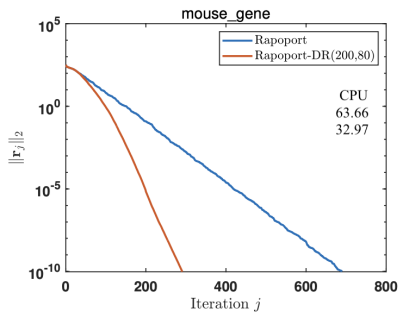
- Rapoport-DR( $m, 2k$ )

$$\mathbf{S}\mathbf{U}_m^{(i)} = \mathbf{H}\mathbf{U}_{m+1}^{(i)} \mathbf{T}_{m+1,m}^{(i)}$$

For example,  $m = 8$ ,  $k = 2$ ,  $i = 2, 3, \dots$ ,

$$\mathbf{T}_{m+1,m}^{(i)} = \begin{bmatrix} 0 & \times & \times & \times & \times & & & \\ \times & 0 & \times & \times & \times & & & \\ \times & \times & 0 & \times & \times & & & \\ \times & \times & \times & 0 & \times & & & \\ \times & \times & \times & \times & 0 & \times & & \\ & & & & \times & 0 & \times & \\ & & & & & \times & 0 & \times \\ & & & & & & \times & 0 \\ & & & & & & & \times \end{bmatrix}, \quad \mathbf{T}_m^{(i)} = -(\mathbf{T}_m^{(i)})^\top.$$

# Rapoport vs. Rapoport-DR( $m, 2k$ )



# Symmetric quasi-definite linear systems

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- $\mathbf{M} \in \mathbb{R}^{m \times m}$  and  $\mathbf{N} \in \mathbb{R}^{n \times n}$  are SPD,  $\mathbf{A} \in \mathbb{R}^{m \times n}$  is nonzero,  $\mathbf{b} \in \mathbb{R}^m$ , and  $\mathbf{c} \in \mathbb{R}^n$ :

$$\begin{bmatrix} \mathbf{M} & \mathbf{A} \\ \mathbf{A}^\top & -\mathbf{N} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ \mathbf{c} \end{bmatrix}, \quad \mathbf{H} = \begin{bmatrix} \mathbf{M} & \\ & \mathbf{N} \end{bmatrix}, \quad \mathbf{K} = \begin{bmatrix} \mathbf{M} & \mathbf{A} \\ \mathbf{A}^\top & -\mathbf{N} \end{bmatrix}.$$

- Symmetric, indefinite, nonsingular
- **Monolithic** methods: solving the system as a whole, for example, SYMMLQ, MINRES

**Segregated** methods: tailored specifically to the block structure, for example, TriCG: **generalized Saunders–Simon–Yip tridiagonalization + Galerkin condition**, mathematically equivalent to preconditioned block-CG

TriMR: **generalized Saunders–Simon–Yip tridiagonalization + MR condition**, mathematically equivalent to preconditioned block-MINRES

(Montoison and Orban, SISC, 2021)

# The generalized SSY tridiagonalization

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- Let  $\beta_1 \mathbf{M} \mathbf{u}_1 = \mathbf{b}$  and  $\gamma_1 \mathbf{N} \mathbf{v}_1 = \mathbf{c}$ . After  $j$  steps of gSSY, we have

$$\mathbf{A} \mathbf{V}_j = \mathbf{M} \mathbf{U}_{j+1} \mathbf{T}_{j+1,j}, \quad \mathbf{A}^\top \mathbf{U}_j = \mathbf{N} \mathbf{V}_{j+1} \mathbf{T}_{j,j+1}^\top,$$

$$\mathbf{U}_{j+1}^\top \mathbf{M} \mathbf{U}_{j+1} = \mathbf{V}_{j+1}^\top \mathbf{N} \mathbf{V}_{j+1} = \mathbf{I}_{j+1}.$$

with

$$\mathbf{T}_{j+1,j} = \begin{bmatrix} \alpha_1 & \gamma_2 & & \\ \beta_2 & \alpha_2 & \ddots & \\ & \ddots & \ddots & \gamma_j \\ & & \beta_j & \alpha_j \\ & & & \beta_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{T}_j \\ \beta_{j+1} \mathbf{e}_j^\top \end{bmatrix}.$$

- Assume that no breakdowns occur for the first  $j$  steps, i.e.,  $\mathbf{U}_j$ ,  $\mathbf{V}_j$ , and  $\mathbf{T}_j$  are well defined. The  $j$ th TriCG iterate is

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{U}_j & \mathbf{0} \\ \mathbf{0} & \mathbf{V}_j \end{bmatrix} \begin{bmatrix} \mathbf{I}_j & \mathbf{T}_j \\ \mathbf{T}_j^\top & -\mathbf{I}_j \end{bmatrix}^{-1} \begin{bmatrix} \beta_1 \mathbf{e}_1 \\ \gamma_1 \mathbf{e}_1 \end{bmatrix},$$

which satisfies the Galerkin condition

$$\begin{bmatrix} \mathbf{U}_j & \mathbf{0} \\ \mathbf{0} & \mathbf{V}_j \end{bmatrix}^\top \left( \begin{bmatrix} \mathbf{b} \\ \mathbf{c} \end{bmatrix} - \begin{bmatrix} \mathbf{M} & \mathbf{A} \\ \mathbf{A}^\top & -\mathbf{N} \end{bmatrix} \begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} \right) = \mathbf{0}.$$

- Equivalent to preconditioned block-CG:

$$\begin{bmatrix} \mathbf{M} & \mathbf{A} \\ \mathbf{A}^\top & -\mathbf{N} \end{bmatrix} \begin{bmatrix} \mathbf{x}^1 & \mathbf{x}^2 \\ \mathbf{y}^1 & \mathbf{y}^2 \end{bmatrix} = \begin{bmatrix} \mathbf{b} & \mathbf{0} \\ \mathbf{0} & \mathbf{c} \end{bmatrix}.$$

## Elliptic singular value decomposition (ESVD)

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- Given SPD  $\mathbf{M}$  and  $\mathbf{N}$ , ESVD of  $\mathbf{A}$  is

$$\mathbf{A} = \mathbf{M}\mathbf{P}\mathbf{\Sigma}\mathbf{Q}^{\top}\mathbf{N},$$

where  $\mathbf{\Sigma} = \text{diag}(\sigma_1, \dots, \sigma_p)$ ,  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_p \geq 0$ ,  $p = \min(m, n)$ , and  $\mathbf{P}$  and  $\mathbf{Q}$  satisfy

$$\mathbf{P}^{\top}\mathbf{M}\mathbf{P} = \mathbf{I}_m, \quad \mathbf{Q}^{\top}\mathbf{N}\mathbf{Q} = \mathbf{I}_n.$$

- Eigenvalues of a two-sided preconditioned matrix (let  $r = \text{rank}(\mathbf{A})$ ):

$$\lambda\left(\mathbf{H}^{-\frac{1}{2}}\mathbf{K}\mathbf{H}^{-\frac{1}{2}}\right) = \begin{cases} \pm\sqrt{\sigma_i^2 + 1}, & i = 1, \dots, r, \\ 1, & (m - r) \text{ times}, \\ -1, & (n - r) \text{ times}. \end{cases}$$

## A gSSY process with deflated restarting

- gSSY-DR( $p, k$ ):

$$\begin{aligned}\mathbf{A}\mathbf{V}_p^{(i)} &= \mathbf{M}\mathbf{U}_p^{(i)}\mathbf{T}_p^{(i)} + \beta_{p+1}^{(i)}\mathbf{M}\mathbf{u}_{p+1}^{(i)}\mathbf{e}_p^\top, \\ \mathbf{A}^\top\mathbf{U}_p^{(i)} &= \mathbf{N}\mathbf{V}_p^{(i)}(\mathbf{T}_p^{(i)})^\top + \gamma_{p+1}^{(i)}\mathbf{N}\mathbf{v}_{p+1}^{(i)}\mathbf{e}_p^\top.\end{aligned}$$

For  $i = 2, 3, \dots$ ,

$$\mathbf{T}_p^{(i)} = \begin{bmatrix} \alpha_1^{(i)} & & & \gamma_2^{(i)} & & & & \\ & \ddots & & \vdots & & & & \\ & & \ddots & \gamma_{k+1}^{(i)} & & & & \\ \beta_2^{(i)} & \dots & \beta_{k+1}^{(i)} & \alpha_{k+1}^{(i)} & \gamma_{k+2}^{(i)} & & & \\ & & & \beta_{k+2}^{(i)} & \alpha_{k+2}^{(i)} & \ddots & & \\ & & & & \ddots & \ddots & \gamma_p^{(i)} & \\ & & & & & \beta_p^{(i)} & \alpha_p^{(i)} & \end{bmatrix}.$$

## TriCG with deflated restarting

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- The recurrences in the first cycle are the same as that of TriCG. Now consider cycle  $i \geq 2$ . The  $j$ th ( $k+1 \leq j \leq p$ ) TriCG-DR( $p, k$ ) iterate is

$$\begin{bmatrix} \mathbf{x}_j^{(i)} \\ \mathbf{y}_j^{(i)} \end{bmatrix} = \begin{bmatrix} \mathbf{U}_j^{(i)} & \mathbf{0} \\ \mathbf{0} & \mathbf{V}_j^{(i)} \end{bmatrix} \begin{bmatrix} \mathbf{I}_j & \mathbf{T}_j^{(i)} \\ (\mathbf{T}_j^{(i)})^\top & -\mathbf{I}_j \end{bmatrix}^{-1} \begin{bmatrix} \beta_1^{(i)} \mathbf{e}_{k+1} \\ \gamma_1^{(i)} \mathbf{e}_{k+1} \end{bmatrix},$$

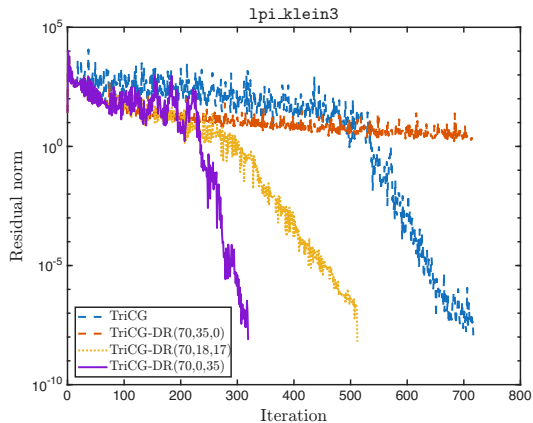
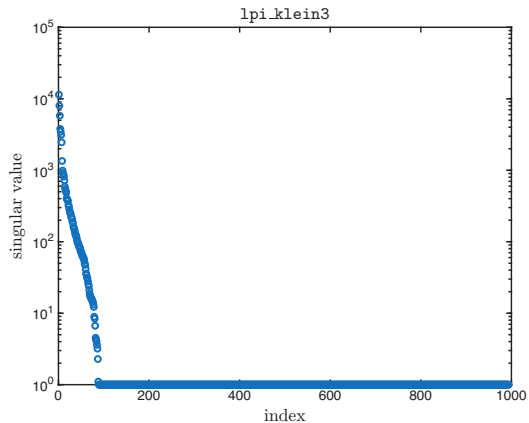
which satisfies the Galerkin condition

$$\begin{bmatrix} \mathbf{U}_j^{(i)} & \mathbf{0} \\ \mathbf{0} & \mathbf{V}_j^{(i)} \end{bmatrix}^\top \left( \begin{bmatrix} \mathbf{b} \\ \mathbf{c} \end{bmatrix} - \begin{bmatrix} \mathbf{M} & \mathbf{A} \\ \mathbf{A}^\top & -\mathbf{N} \end{bmatrix} \begin{bmatrix} \mathbf{x}_j^{(i)} \\ \mathbf{y}_j^{(i)} \end{bmatrix} \right) = \mathbf{0}.$$

- Using an  $\text{LDL}^\top$  decomposition and the same strategy in TriCG, short recurrences can be obtained to compute  $\mathbf{x}_j^{(i)}$  and  $\mathbf{y}_j^{(i)}$  for  $k+1 \leq j \leq p$ .



# TriCG vs. TriCG-DR



## Summary

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- We have developed deflation techniques for nonsymmetric positive definite linear systems and symmetric quasi-definite linear systems.
- We have proposed  $S^2$ Lan-DR and Rapopart-DR for solving nonsymmetric positive definite linear systems, and TriCG-DR for solving symmetric quasi-definite linear systems. The new methods have faster convergence and demonstrate a significant advantage in computational time.
- $S^2$ Lan-DR and gSSY-DR can be used to compute partial spectral information. And the computed spectral information can be used to help solve linear systems with sequential multiple right-hand sides.

## Our recent related work

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- Kui Du, Jia-Jun Fan, Xiao-Hui Sun, Fang Wang, and Ya-Lan Zhang  
On Krylov subspace methods for skew-symmetric and shifted skew-symmetric linear systems, Advances in Computational Mathematics (2024) 50:78
- Kui Du, Jia-Jun Fan, and Fang Wang  
On deflated CGW methods for solving nonsymmetric positive definite linear systems, Calcolo (2025) 62:22.
- Kui Du, Jia-Jun Fan, and Ya-Lan Zhang  
Improved TriCG and TriMR methods for symmetric quasi-definite linear systems  
Numerical Linear Algebra with Applications, 2025, 32:e70026.
- Kui Du and Jia-Jun Fan  
TriCG with deflated restarting for symmetric quasi-definite linear systems.  
In preparation, 2025.

**Thanks!**