



# Can you use MINRES with an indefinite preconditioner?

**Tyrone Rees**

STFC Rutherford Appleton Laboratory

Nick Gould (RAL), Dominique Orban (École Polytechnique de Montréal)

# The problem...

Solve

$$\mathcal{A}x = b$$

where  $A$  is symmetric, but **indefinite**.

# MINRES (Paige & Saunders, 1975)

MINimal RESidual

Finds

$$x_k \in x_0 + \text{span}\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$$

which minimizes

$$\|r_k\|_2 = \|b - \mathcal{A}x_k\|_2$$

# MINRES (Paige & Saunders, 1975)

$$r_k := b - \mathcal{A}x_k$$

MINimal RESidual



Finds

$$x_k \in x_0 + \text{span}\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$$

which minimizes

$$\|r_k\|_2 = \|b - \mathcal{A}x_k\|_2$$

## Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

## Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

$$v_1 = r_0 / \|r_0\|$$

Given an orthonormal set  $\{v_1, \dots, v_m\}$

- ▶ set  $w = \mathcal{A}v_m$
- ▶  $h_{i,m} = v_i^T w$
- ▶  $w \leftarrow w - \sum h_{i,m} v_i$
- ▶  $h_{m+1,m} = \|w\|_2, \quad v_{m+1} = w / h_{m+1,m}$

## Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

$$v_1 = r_0 / \|r_0\|$$

Given an orthonormal set  $\{v_1, \dots, v_m\}$

- ▶ set  $w = \mathcal{A}v_m$
- ▶  $h_{i,m} = v_i^T w$
- ▶  $w \leftarrow w - \sum h_{i,m} v_i$
- ▶  $h_{m+1,m} = \|w\|_2, \quad v_{m+1} = w / h_{m+1,m}$



# Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

$$\mathcal{A} \underbrace{\begin{bmatrix} v_1 & \dots & v_m \\ \vdots & & \vdots \end{bmatrix}}_{V_m} = \underbrace{\begin{bmatrix} v_1 & \dots & v_{m+1} \\ \vdots & & \vdots \end{bmatrix}}_{V_{m+1}} \underbrace{\begin{bmatrix} h_{1,1} & h_{2,1} & \dots & h_{1,m} \\ h_{2,1} & h_{2,2} & \dots & h_{2,m} \\ 0 & \ddots & \ddots & \vdots \\ \vdots & & h_{m,m-1} & h_{m,m} \\ 0 & \dots & 0 & h_{m+1,m} \end{bmatrix}}_{\hat{H}_m}$$



## Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

$$V_m^T \mathcal{A} V_m = \underbrace{\begin{bmatrix} h_{1,1} & h_{1,2} & \cdots & h_{1,m} \\ h_{2,1} & h_{2,2} & & h_{2,m} \\ \vdots & \ddots & \ddots & \vdots \\ 0 & & h_{m,m-1} & h_{m,m} \end{bmatrix}}_{H_m}$$



## Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

$$V_m^T \mathcal{A} V_m = \underbrace{\begin{bmatrix} h_{1,1} & h_{2,1} & 0 & \dots & 0 \\ h_{2,1} & h_{2,2} & \ddots & \ddots & \vdots \\ 0 & \ddots & \ddots & & 0 \\ \vdots & \ddots & & & h_{m,m-1} \\ 0 & \dots & 0 & h_{m,m-1} & h_{m,m} \end{bmatrix}}_{H_m}$$



## Lanczos orthogonalization

Want an **orthonormal basis** for the set  $\{r_0, \mathcal{A}r_0, \dots, \mathcal{A}^{k-1}r_0\}$

$$V_m^T \mathcal{A} V_m = \underbrace{\begin{bmatrix} \alpha_1 & \beta_1 & 0 & \dots & 0 \\ \beta_1 & \alpha_2 & \ddots & \ddots & \vdots \\ 0 & \ddots & \ddots & & 0 \\ \vdots & \ddots & & & \beta_{m-1} \\ 0 & \dots & 0 & \beta_{m-1} & \alpha_m \end{bmatrix}}_{H_m}$$



# From Lanczos to MINRES

$$\|b - \mathcal{A}x_k\|_2 = \|b - \mathcal{A}(x_0 + V_k z_k)\|_2$$

# From Lanczos to MINRES

$$\begin{aligned}\|b - \mathcal{A}x_k\|_2 &= \|b - \mathcal{A}(x_0 + V_k z_k)\|_2 \\ &= \|(b - \mathcal{A}x_0) - \mathcal{A}V_k z_k\|_2\end{aligned}$$

# From Lanczos to MINRES

$$\begin{aligned}\|b - \mathcal{A}x_k\|_2 &= \|b - \mathcal{A}(x_0 + V_k z_k)\|_2 \\ &= \|(b - \mathcal{A}x_0) - \mathcal{A}V_k z_k\|_2 \\ &= \| \|r_0\| v_1 - \mathcal{A}V_k z_k \|_2\end{aligned}$$

$v_1 = r_0 / \|r_0\|$

# From Lanczos to MINRES

$$\begin{aligned}\|b - \mathcal{A}x_k\|_2 &= \|b - \mathcal{A}(x_0 + V_k z_k)\|_2 \\ &= \|(b - \mathcal{A}x_0) - \mathcal{A}V_k z_k\|_2 \\ &= \|\|r_0\|v_1 - \mathcal{A}V_k z_k\|_2 \\ &= \|\|r_0\|v_1 - V_{k+1} \hat{H}_k z_k\|_2\end{aligned}$$

$$\mathcal{A}V_k = V_{k+1} \hat{H}_k$$

## From Lanczos to MINRES

$$\begin{aligned}\|b - \mathcal{A}x_k\|_2 &= \|b - \mathcal{A}(x_0 + V_k z_k)\|_2 \\&= \|(b - \mathcal{A}x_0) - \mathcal{A}V_k z_k\|_2 \\&= \|\|r_0\|v_1 - \mathcal{A}V_k z_k\|_2 \\&= \|\|r_0\|v_1 - V_{k+1} \hat{H}_k z_k\|_2 \\&= \|V_{k+1}(\|r_0\|e_1 - \hat{H}_k z_k)\|_2\end{aligned}$$

$$e_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

# From Lanczos to MINRES

$$\begin{aligned}\|b - \mathcal{A}x_k\|_2 &= \|b - \mathcal{A}(x_0 + V_k z_k)\|_2 \\&= \|(b - \mathcal{A}x_0) - \mathcal{A}V_k z_k\|_2 \\&= \|\|r_0\|v_1 - \mathcal{A}V_k z_k\|_2 \\&= \|\|r_0\|v_1 - V_{k+1} \hat{H}_k z_k\|_2 \\&= \|V_{k+1}(\|r_0\|e_1 - \hat{H}_k z_k)\|_2 \\&= \|\|r_0\|e_1 - \hat{H}_k z_k\|_2\end{aligned}$$

# From Lanczos to MINRES

$$\begin{aligned}\|b - \mathcal{A}x_k\|_2 &= \|b - \mathcal{A}(x_0 + V_k z_k)\|_2 \\&= \|(b - \mathcal{A}x_0) - \mathcal{A}V_k z_k\|_2 \\&= \|\|r_0\|v_1 - \mathcal{A}V_k z_k\|_2 \\&= \|\|r_0\|v_1 - V_{k+1} \hat{H}_k z_k\|_2 \\&= \|V_{k+1}(\|r_0\|e_1 - \hat{H}_k z_k)\|_2 \\&= \|\|r_0\|e_1 - \hat{H}_k z_k\|_2\end{aligned}$$

Least squares problem: solved by Givens rotations

## What about preconditioning?

Convergence of MINRES depends on the clustering of the eigenvalues.

Suppose  $P = MM^T$  is such that the spectrum of

$$M^{-1} \mathcal{A} M^{-T}$$

is ‘nice’.

Apply MINRES to

$$M^{-1} \mathcal{A} M^{-T} y = M^{-1} b,$$

where  $y = M^T x$ .

Expect better convergence.

# Algorithm

$$\mathbf{v}_1 = \mathbf{b} - \mathcal{A}\mathbf{x}_0$$

$$P\mathbf{z}_1 = \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T \mathcal{A} \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = \mathcal{A}\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$P\mathbf{z}_{j+1} = \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

# Algorithm

$$\mathbf{v}_1 = \mathbf{b} - \mathcal{A}\mathbf{x}_0$$

$$P\mathbf{z}_1 = \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T \mathcal{A} \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = \mathcal{A}\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$P\mathbf{z}_{j+1} = \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

**Preconditioner.** Note  $M$  factors not needed.



# Algorithm

$$\mathbf{v}_1 = \mathbf{b} - \mathcal{A}\mathbf{x}_0$$

$$P\mathbf{z}_1 = \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T \mathcal{A} \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = \mathcal{A}\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$P\mathbf{z}_{j+1} = \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

**Preconditioner.** Note  $M$  factors  
not needed.

diagonal entries of  $H_k$   
off-diagonal entries of  $H_k$



# Algorithm

$$\mathbf{v}_1 = \mathbf{b} - \mathcal{A}\mathbf{x}_0$$

$$P\mathbf{z}_1 = \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T \mathcal{A} \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = \mathcal{A}\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$P\mathbf{z}_{j+1} = \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

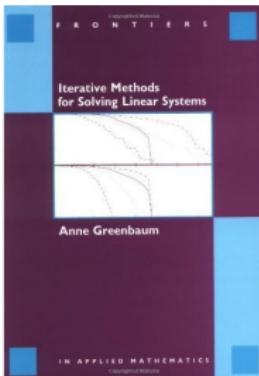
$$\eta = -s_{j+1} \eta$$

**Preconditioner.** Note  $M$  factors  
not needed.

diagonal entries of  $H_k$   
off-diagonal entries of  $H_k$   
Givens' rotations



# Greenbaum



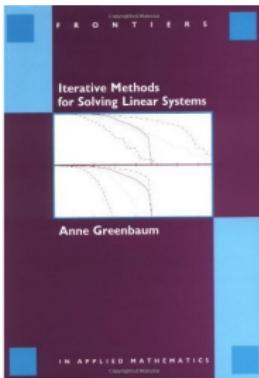
Chapter 8

## Overview and Preconditioned Algorithms

The same modifications can be made to any of the MINRES implementations, provided that the preconditioner  $M$  is positive definite. To obtain a preconditioned version of Algorithm 4, first consider the Lanczos algorithm applied directly to the matrix  $L^{-1}AL^{-H}$  with initial vector  $\hat{q}_1$ . Successive vectors satisfy

$$\begin{aligned}\hat{v}_j &= L^{-1}AL^{-H}\hat{q}_j - \alpha_j\hat{q}_j - \beta_{j-1}\hat{q}_{j-1}, \\ \alpha_j &= \langle L^{-1}AL^{-H}\hat{q}_j, \hat{q}_j \rangle - \beta_{j-1}\langle \hat{q}_{j-1}, \hat{q}_j \rangle, \\ \hat{q}_{j+1} &= \hat{v}_j/\beta_j, \quad \beta_j = \|\hat{v}_j\|.\end{aligned}$$

# Greenbaum



## Chapter 8

### Overview and Preconditioned Algorithms

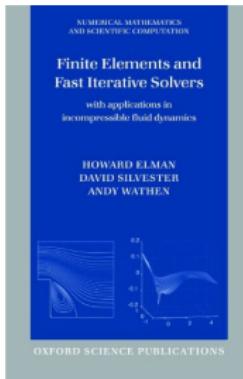
•  
•

The same modifications can be made to any of the MINRES implementations, provided that the preconditioner  $M$  is positive definite. To obtain a preconditioned version of Algorithm 4, first consider the Lanczos algorithm applied directly to the matrix  $L^{-1}AL^{-H}$  with initial vector  $\hat{q}_1$ . Successive vectors satisfy

$$\begin{aligned}\hat{v}_j &= L^{-1}AL^{-H}\hat{q}_j - \alpha_j\hat{q}_j - \beta_{j-1}\hat{q}_{j-1}, \\ \alpha_j &= \langle L^{-1}AL^{-H}\hat{q}_j, \hat{q}_j \rangle - \beta_{j-1}\langle \hat{q}_{j-1}, \hat{q}_j \rangle, \\ \hat{q}_{j+1} &= \hat{v}_j/\beta_j, \quad \beta_j = \|\hat{v}_j\|.\end{aligned}$$



# Elman, Silvester, Wathen



## 6.1 The preconditioned MINRES method

In the generic context of solving a symmetric and indefinite matrix system

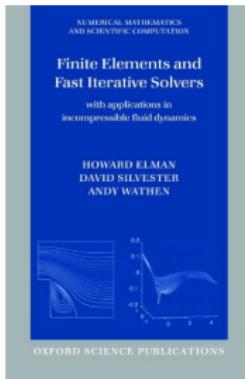
$$K\mathbf{x} = \mathbf{b}, \quad (6.3)$$

•  
•  
•

It is desirable to ensure that any preconditioner does not destroy the symmetry of the discrete Stokes problem; otherwise iterative methods for nonsymmetric systems would have to be employed as for discrete convection-diffusion problems. To preserve symmetry in the preconditioned system, a symmetric and positive-definite preconditioner  $M = HH^T$  is required.



# Elman, Silvester, Wathen



## 6.1 The preconditioned MINRES method

In the generic context of solving a symmetric and indefinite matrix system

$$K\mathbf{x} = \mathbf{b}, \quad (6.3)$$

•  
•  
•

It is desirable to ensure that any preconditioner does not destroy the symmetry of the discrete Stokes problem; otherwise iterative methods for nonsymmetric systems would have to be employed as for discrete convection-diffusion problems. To preserve symmetry in the preconditioned system, a symmetric and positive-definite preconditioner  $M = HH^T$  is required.



# Saunders' implementation

The screenshot shows the homepage of the SOL (Stanford Optimization Library) website. The header includes the SOL logo, the Stanford University Department of Management Science and Engineering (MS&E), and the Huang Engineering Center address. A navigation menu on the left lists Home, Software, Personnel, Students, Alumni, Visitors, Links & Fun Stuff, Contact Us, Research & Applications (with sub-links for Constrained Optimization, Stochastic Programming, and Systems Using SOL), and Publications (with sub-links for Books, Dissertations, Journal Papers, Classics, Technical Reports, User Guides, Talks, and Dantzig Memoriam). The main content area features a section titled "MINRES: Sparse Symmetric Equations" with bullet points for AUTHORS, CONTRIBUTORS, and CONTENTS. It also contains mathematical equations and text explaining the method's properties and usage.

**SOL**

Stanford University  
Dept of Management Science and Engineering (MS&E)  
Huang Engineering Center  
Stanford, CA 94305-4121 USA

---

**MINRES: Sparse Symmetric Equations**

- **AUTHORS:** C. C. Paige, M. A. Saunders.
- **CONTRIBUTORS:** Sou-Cheng Choi, Dominique Orban, Umberto Emanuele Villa.
- **CONTENTS:** Implementation of a conjugate-gradient type method for solving sparse linear equations: Solve

$Ax = b$  or  $(A - sI)x = b$ .

The matrix  $A - sI$  must be symmetric but it may be definite or indefinite or singular. The scalar  $s$  is a shifting parameter -- it may be any number. The method is based on Lanczos tridiagonalization. You may provide a preconditioner, but it must be positive definite.

MINRES is really solving one of the least-squares problems

$$\text{minimize } \|Ax - b\| \text{ or } \|(A - sI)x - b\|.$$

If  $A$  is singular (and  $s = 0$ ), MINRES returns a least-squares solution with small  $\|Ar\|$  (where  $r = b - Ax$ ), but in general it is not the minimum-length solution. To get the min-length solution, use **MINRES-QLP**.

Similarly if  $A - sI$  is singular.

If  $A$  is symmetric (and  $A - sI$  is nonsingular), **SYMMQLQ** may be slightly more reliable.

If  $A$  is unsymmetric, use **LSCP**.



# Saunders' implementation

The screenshot shows the SOL (Stanford Optimization Library) website. The header includes the SOL logo, the text "Stanford University Dept of Management Science and Engineering (MS&E)", and the address "Huang Engineering Center Stanford, CA 94305-4121 USA". The left sidebar has links for Home, Software, Personnel, Students, Alumni, Visitors, Links & Fun Stuff, Contact Us, Research & Applications (Constrained Optimization, Stochastic Programming, Systems Using SOL), and Publications (Books, Dissertations, Journal Papers, Classics, Technical Reports, User Guides, Talks, Dantzig Memoriam). The main content area is titled "MINRES: Sparse Symmetric Equations". It lists authors (C. C. Paige, M. A. Saunders), contributors (Sou-Cheng Choi, Dominique Orban, Umberto Emanuele Villa), and contents (implementation of a conjugate-gradient type method for solving sparse linear equations: Solve  $Ax = b$  or  $(A - sI)x = b$ ). It explains that  $A - sI$  must be symmetric but may be definite or indefinite or singular. The scalar  $s$  is a shifting parameter (it may be any number). The method is based on Lanczos tridiagonalization. You may provide a preconditioner, but it must be positive definite. A red oval highlights this sentence. It also states that MINRES is really solving one of the least-squares problems:

$$\text{minimize } \|Ax - b\| \text{ or } \|(A - sI)x - b\|.$$

If  $A$  is singular (and  $s = 0$ ), MINRES returns a least-squares solution with small  $\|Ar\|$  (where  $r = b - Ax$ ), but in general it is not the minimum-length solution. To get the min-length solution, use MINRES-QLP. Similarly if  $A - sI$  is singular. If  $A$  is symmetric (and  $A - sI$  is nonsingular), SYMMLQ may be slightly more reliable. A small note at the bottom right says "If A is unsymmetric, use LSQR".





## Can you use MINRES with an indefinite preconditioner?

**Tyrone Rees**

STFC Rutherford Appleton Laboratory

Nick Gould (RAL), Dominique Orban (École Polytechnique de Montréal)



Can you use MINRES with an indefinite  
preconditioner?

No

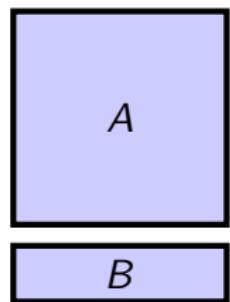
**Tyrone Rees**

STFC Rutherford Appleton Laboratory

Nick Gould (RAL), Dominique Orban (École Polytechnique de Montréal)

## A class of indefinite problems

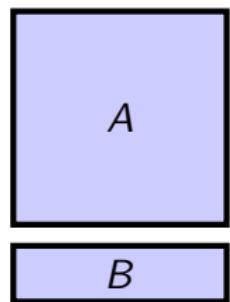
$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \hat{x} \\ y \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix}$$



## A class of indefinite problems

$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} x + x_0 \\ y \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix}$$

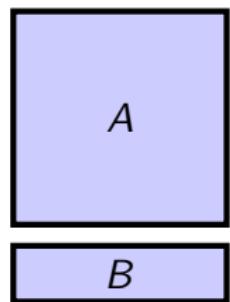
$$\begin{aligned}\hat{x} &= x + x_0 \\ Bx_0 &= b\end{aligned}$$



## A class of indefinite problems

$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} c \\ 0 \end{bmatrix}$$

$$c = a - Ax_0$$

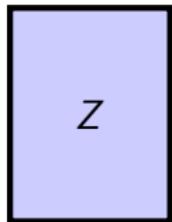
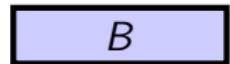
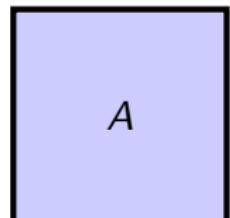


## A class of indefinite problems

$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} c \\ 0 \end{bmatrix}$$

Let  $Z$  span the nullspace of  $B$  (i.e.  $BZ = 0$ )

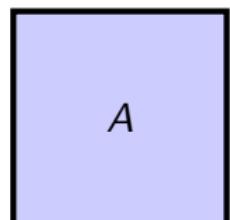
$$BZ = 0 \Rightarrow x = Z\bar{x}$$



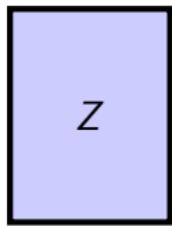
## A class of indefinite problems

$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} c \\ 0 \end{bmatrix}$$

Let  $Z$  span the nullspace of  $B$  (i.e.  $BZ = 0$ )



$$BZ = 0 \Rightarrow x = Z\bar{x}$$



$$AZ\bar{x} + B^T y = c$$

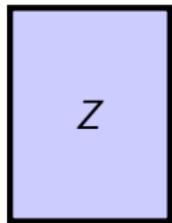
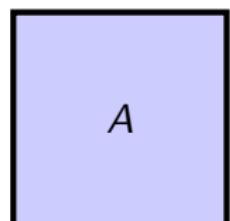
## A class of indefinite problems

$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} c \\ 0 \end{bmatrix}$$

Let  $Z$  span the nullspace of  $B$  (i.e.  $BZ = 0$ )

$$BZ = 0 \Rightarrow x = Z\bar{x}$$

$$Z^T A Z \bar{x} + Z^T B^T y = Z^T c$$



## A class of indefinite problems

$$\begin{bmatrix} A & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} c \\ 0 \end{bmatrix}$$

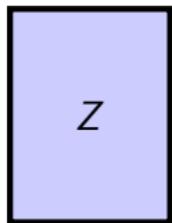
Let  $Z$  span the nullspace of  $B$  (i.e.  $BZ = 0$ )

$$BZ = 0 \Rightarrow x = Z\bar{x}$$

$$Z^T A Z \bar{x} = Z^T c$$

If  $A$  symmetric,  $Z^T A Z$  symmetric, but (possibly) indefinite - apply MINRES!

Use as a preconditioner  $Z^T G Z$ , where  $G$  is positive definite on the null space of  $B$ .



## MINRES applied to the reduced system

$$\bar{\mathbf{v}}_1 = Z^T \mathbf{c} - Z^T A Z \bar{\mathbf{x}}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} \bar{\mathbf{v}}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T \bar{\mathbf{v}}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$\bar{\mathbf{v}}_{j+1} = Z^T A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} \bar{\mathbf{v}}_j - \frac{\beta_j}{\beta_{j-1}} \bar{\mathbf{v}}_{j-1}$$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} \bar{\mathbf{v}}_{j+1}$$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T \bar{\mathbf{v}}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\bar{\mathbf{x}}_j = \bar{\mathbf{x}}_{j-1} + c_{j+1} \eta \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$

## MINRES applied to the reduced system

$$\bar{\mathbf{v}}_1 = Z^T \mathbf{c} - Z^T A \bar{\mathbf{x}}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} \bar{\mathbf{v}}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T \bar{\mathbf{v}}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$\bar{\mathbf{v}}_{j+1} = Z^T A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} \bar{\mathbf{v}}_j - \frac{\beta_j}{\beta_{j-1}} \bar{\mathbf{v}}_{j-1}$$

►  $\mathbf{x}_k = Z \bar{\mathbf{x}}_k$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} \bar{\mathbf{v}}_{j+1}$$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T \bar{\mathbf{v}}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\bar{\mathbf{x}}_j = \bar{\mathbf{x}}_{j-1} + c_{j+1} \eta \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$



## MINRES applied to the reduced system

$$\bar{\mathbf{v}}_1 = Z^T \mathbf{c} - Z^T A \mathbf{x}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} \bar{\mathbf{v}}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T \bar{\mathbf{v}}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$\bar{\mathbf{v}}_{j+1} = Z^T A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} \bar{\mathbf{v}}_j - \frac{\beta_j}{\beta_{j-1}} \bar{\mathbf{v}}_{j-1}$$

►  $\mathbf{x}_k = Z \bar{\mathbf{x}}_k$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} \bar{\mathbf{v}}_{j+1}$$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T \bar{\mathbf{v}}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$



# MINRES applied to the reduced system

$$\bar{\mathbf{v}}_1 = Z^T \mathbf{c} - Z^T A \mathbf{x}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} \bar{\mathbf{v}}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T \bar{\mathbf{v}}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$\bar{\mathbf{v}}_{j+1} = Z^T A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} \bar{\mathbf{v}}_j - \frac{\beta_j}{\beta_{j-1}} \bar{\mathbf{v}}_{j-1}$$

►  $\mathbf{x}_k = Z \bar{\mathbf{x}}_k$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} \bar{\mathbf{v}}_{j+1}$$

►  $Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T \bar{\mathbf{v}}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$



# MINRES applied to the reduced system

$$Z^T \mathbf{v}_1 = Z^T \mathbf{c} - Z^T A \mathbf{x}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T Z^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$Z^T \mathbf{v}_{j+1} = Z^T A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} Z^T \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} Z^T \mathbf{v}_{j-1}$$

$$\triangleright \mathbf{x}_k = Z \bar{\mathbf{x}}_k$$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\triangleright Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T Z^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$



# MINRES applied to the reduced system

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T Z^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$\mathbf{v}_{j+1} = A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T Z^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

►  $\mathbf{x}_k = Z \bar{\mathbf{x}}_k$

►  $Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$



# MINRES applied to the reduced system

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\bar{\mathbf{z}}_1 = (Z^T G Z)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\bar{\mathbf{z}}_1^T Z^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\bar{\mathbf{z}}_j = \bar{\mathbf{z}}_j / \beta_j$$

$$\alpha_j = \bar{\mathbf{z}}_j^T Z^T A Z \bar{\mathbf{z}}_j$$

$$\mathbf{v}_{j+1} = A Z \bar{\mathbf{z}}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\blacktriangleright \mathbf{x}_k = Z \bar{\mathbf{x}}_k$$

$$\bar{\mathbf{z}}_{j+1} = (Z^T G Z)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\blacktriangleright Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$$

$$\beta_{j+1} = \sqrt{\bar{\mathbf{z}}_{j+1}^T Z^T \mathbf{v}_{j+1}}$$

$$\blacktriangleright \mathbf{z}_k = Z \bar{\mathbf{z}}_k$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\bar{\mathbf{w}}_{j+1} = (\bar{\mathbf{z}}_j - \gamma_3 \bar{\mathbf{w}}_{j-1} - \gamma_2 \bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z \bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \eta$$



# MINRES applied to the reduced system

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\mathbf{z}_1 = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\mathbf{z}_{j+1} = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$Z\bar{\mathbf{w}}_{j+1} = (\mathbf{z}_j - \gamma_3 Z\bar{\mathbf{w}}_{j-1} - \gamma_2 Z\bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z\bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$

$$\mathbf{x}_k = Z\bar{\mathbf{x}}_k$$

$$Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$$

$$\mathbf{z}_k = Z\bar{\mathbf{z}}_k$$



# MINRES applied to the reduced system

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\mathbf{z}_1 = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\mathbf{z}_{j+1} = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$Z\bar{\mathbf{w}}_{j+1} = (\mathbf{z}_j - \gamma_3 Z\bar{\mathbf{w}}_{j-1} - \gamma_2 Z\bar{\mathbf{w}}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta Z\bar{\mathbf{w}}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

$$\mathbf{x}_k = Z\bar{\mathbf{x}}_k$$

$$Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$$

$$\mathbf{z}_k = Z\bar{\mathbf{z}}_k$$

$$\mathbf{w}_k = Z\bar{\mathbf{w}}_k$$



# MINRES applied to the reduced system

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\mathbf{z}_1 = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\mathbf{z}_{j+1} = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

$$\mathbf{x}_k = Z \bar{\mathbf{x}}_k$$

$$Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$$

$$\mathbf{z}_k = Z \bar{\mathbf{z}}_k$$

$$\mathbf{w}_k = Z \bar{\mathbf{w}}_k$$



# MINRES applied to the reduced system

$$\begin{aligned}\mathbf{v}_1 &= \mathbf{c} - A\mathbf{x}_0 \\ \mathbf{z}_1 &= Z(Z^T GZ)^{-1} Z^T \mathbf{v}_1 \\ \beta_1 &= \sqrt{\mathbf{z}_1^T \mathbf{v}_1}\end{aligned}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\mathbf{z}_{j+1} = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \eta$$

- ▶  $\mathbf{x}_k = Z\bar{\mathbf{x}}_k$
- ▶  $Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$
- ▶  $\mathbf{z}_k = Z\bar{\mathbf{z}}_k$
- ▶  $\mathbf{w}_k = Z\bar{\mathbf{w}}_k$



# MINRES applied to the reduced system

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\mathbf{z}_1 = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_1$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$\eta = \beta_1$ ,  $s_0 = s_1 = 0$ ,  $c_0 = c_1 = 1$   
 for  $j = 1, 2, \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\mathbf{z}_{j+1} = Z(Z^T GZ)^{-1} Z^T \mathbf{v}_{j+1}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \gamma_1$$

$$\begin{aligned} \mathbf{z} &= Z(Z^T GZ)^{-1} Z^T \mathbf{v} \\ \Rightarrow \begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z} \\ \mathbf{g} \end{bmatrix} &= \begin{bmatrix} \mathbf{v} \\ \mathbf{0} \end{bmatrix} \end{aligned}$$

- $\mathbf{x}_k = Z\bar{\mathbf{x}}_k$
- $Z^T \mathbf{v}_k = \bar{\mathbf{v}}_k$
- $\mathbf{z}_k = Z\bar{\mathbf{z}}_k$
- $\mathbf{w}_k = Z\bar{\mathbf{w}}_k$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ \mathbf{0} \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\mathbf{v}_{j+1} = A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{0} \end{bmatrix}$$

## Projected MINRES

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\mathbf{w}_{j+1} = (\mathbf{z}_j - \gamma_3 \mathbf{w}_{j-1} - \gamma_2 \mathbf{w}_j) / \gamma_1$$

$$\mathbf{x}_j = \mathbf{x}_{j-1} + c_{j+1} \eta \mathbf{w}_{j+1}$$

$$\eta = -s_{j+1} \eta$$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = -B\mathbf{x}_0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ \mathbf{u}_1 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1 + \mathbf{g}_1^T \mathbf{u}_1}$$

$$\eta = \beta_1, \quad s_0 = s_1 = 0, \quad c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

MINRES with (indefinite)  
constraint preconditioner



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = -B\mathbf{x}_0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ \mathbf{u}_1 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1 + \mathbf{g}_1^T \mathbf{u}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Claim:

If  $x_0$  chosen so that  
 $Bx_0 = 0$ , then both  
algorithms are identical

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = -B\mathbf{x}_0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ \mathbf{u}_1 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1 + \mathbf{g}_1^T \mathbf{u}_1}$$

$$\eta = \beta_1, \quad s_0 = s_1 = 0, \quad c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Claim:

If  $x_0$  chosen so that  
 $Bx_0 = 0$ , then both

algorithms are identical

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ \mathbf{u}_1 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1 + \mathbf{g}_1^T \mathbf{u}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Claim:

If  $x_0$  chosen so that  
 $Bx_0 = 0$ , then both

algorithms are identical

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1 + 0}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Claim:

If  $x_0$  chosen so that  
 $Bx_0 = 0$ , then both

algorithms are identical

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Claim:

If  $x_0$  chosen so that  
 $Bx_0 = 0$ , then both

algorithms are identical

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

$$\blacktriangleright \mathbf{u}_k = 0$$

$$\blacktriangleright B\mathbf{z}_k = 0$$

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

$$\mathbf{z}_j^T A \mathbf{z}_j + \mathbf{z}_j^T B^T \mathbf{g}_j + \mathbf{y}_j^T B \mathbf{z}_j$$

Assume, ( $k \leq j$ ) :

- ▶  $\mathbf{u}_k = 0$
- ▶  $B\mathbf{z}_k = 0$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T (A\mathbf{z}_j + B^T \mathbf{g}_j) + \mathbf{y}_j^T B\mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

$$\begin{aligned} & \mathbf{z}_j^T A \mathbf{z}_j + \mathbf{z}_j^T B^T \mathbf{g}_j + \mathbf{y}_j^T B \mathbf{z}_j \\ &= \mathbf{z}_j^T A \mathbf{z}_j \end{aligned}$$

Assume, ( $k \leq j$ ) :

$$\mathbf{u}_k = 0$$

$$B\mathbf{z}_k = 0$$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

$$\blacktriangleright \mathbf{u}_k = 0$$

$$\blacktriangleright B\mathbf{z}_k = 0$$

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ B\mathbf{z}_j \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ \mathbf{u}_j \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ \mathbf{u}_{j-1} \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

$$\blacktriangleright \mathbf{u}_k = 0$$

$$\blacktriangleright B\mathbf{z}_k = 0$$

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

$$\blacktriangleright \mathbf{u}_k = 0$$

$$\blacktriangleright B\mathbf{z}_k = 0$$

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + \mathbf{g}_{j+1}^T \mathbf{u}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

$$\mathbf{u}_k = 0$$

$$B\mathbf{z}_k = 0$$

$$\mathbf{u}_{j+1} = 0$$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, \quad s_0 = s_1 = 0, \quad c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ 0 \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1} + 0}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

►  $\mathbf{u}_k = 0$

►  $B\mathbf{z}_k = 0$

$\mathbf{u}_{j+1} = 0$

$B\mathbf{z}_{j+1} = 0$

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \quad \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, \quad s_0 = s_1 = 0, \quad c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \quad \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ 0 \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; \quad s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

Assume, ( $k \leq j$ ) :

►  $\mathbf{u}_k = 0$

►  $B\mathbf{z}_k = 0$

$\mathbf{u}_{j+1} = 0$

$B\mathbf{z}_{j+1} = 0$

$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ 0 \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

problem?

Assume, ( $k \leq j$ ) :

►  $\mathbf{u}_k = 0$

►  $B\mathbf{z}_k = 0$

$\mathbf{u}_{j+1} = 0$

$B\mathbf{z}_{j+1} = 0$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ 0 \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

problem?

unchanged

Assume, ( $k \leq j$ ) :

$$\mathbf{u}_k = 0$$

$$B\mathbf{z}_k = 0$$

$$\mathbf{u}_{j+1} = 0$$

$$B\mathbf{z}_{j+1} = 0$$



$$\mathbf{v}_1 = \mathbf{c} - A\mathbf{x}_0, \mathbf{u}_1 = 0$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_1 \\ \mathbf{g}_1 \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1 \\ 0 \end{bmatrix}$$

$$\beta_1 = \sqrt{\mathbf{z}_1^T \mathbf{v}_1}$$

$$\eta = \beta_1, s_0 = s_1 = 0, c_0 = c_1 = 1$$

for  $j = 1, 2 \dots$  until convergence

$$\mathbf{z}_j = \mathbf{z}_j / \beta_j, \mathbf{g}_j = \mathbf{g}_j / \beta_j$$

$$\alpha_j = \mathbf{z}_j^T A \mathbf{z}_j$$

$$\begin{bmatrix} \mathbf{v}_{j+1} \\ \mathbf{u}_{j+1} \end{bmatrix} = \begin{bmatrix} A\mathbf{z}_j + B^T \mathbf{g}_j \\ 0 \end{bmatrix} - \frac{\alpha_j}{\beta_j} \begin{bmatrix} \mathbf{v}_j \\ 0 \end{bmatrix} - \frac{\beta_j}{\beta_{j-1}} \begin{bmatrix} \mathbf{v}_{j-1} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} G & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z}_{j+1} \\ \mathbf{g}_{j+1} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_{j+1} \\ 0 \end{bmatrix}$$

$$\beta_{j+1} = \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}}$$

$$\gamma_0 = c_j \alpha_j - c_{j-1} s_j \beta_j$$

$$\gamma_1 = \sqrt{\gamma_0^2 + \beta_{j+1}^2}$$

$$\gamma_2 = s_j \alpha_j + c_{j-1} c_j \beta_j$$

$$\gamma_3 = s_{j-1} \beta_j$$

$$c_{j+1} = \gamma_0 / \gamma_1; s_{j+1} = \beta_{j+1} / \gamma_1$$

$$\begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix} = \frac{1}{\gamma_1} \left( \begin{bmatrix} \mathbf{z}_j \\ \mathbf{g}_{j+1} \end{bmatrix} - \gamma_3 \begin{bmatrix} \mathbf{w}_{j-1} \\ \mathbf{p}_{j-1} \end{bmatrix} - \gamma_2 \begin{bmatrix} \mathbf{w}_j \\ \mathbf{p}_j \end{bmatrix} \right)$$

$$\begin{bmatrix} \mathbf{x}_j \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_{j-1} \\ \mathbf{y}_{j-1} \end{bmatrix} + c_{j+1} \eta \begin{bmatrix} \mathbf{w}_{j+1} \\ \mathbf{p}_{j+1} \end{bmatrix}$$

$$\eta = -s_{j+1} \eta$$

$$\begin{aligned} \beta_{j+1} &= \sqrt{\mathbf{z}_{j+1}^T \left( A\mathbf{z}_j + B^T \mathbf{g}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1} \right)} \\ &= \sqrt{\mathbf{z}_{j+1}^T \left( A\mathbf{z}_j - \frac{\alpha_j}{\beta_j} \mathbf{v}_j - \frac{\beta_j}{\beta_{j-1}} \mathbf{v}_{j-1} \right)} \\ &= \sqrt{\mathbf{z}_{j+1}^T \mathbf{v}_{j+1}^{\text{PPMINRES}}} \end{aligned}$$

Assume, ( $k \leq j$ ) :

►  $\mathbf{u}_k = 0$

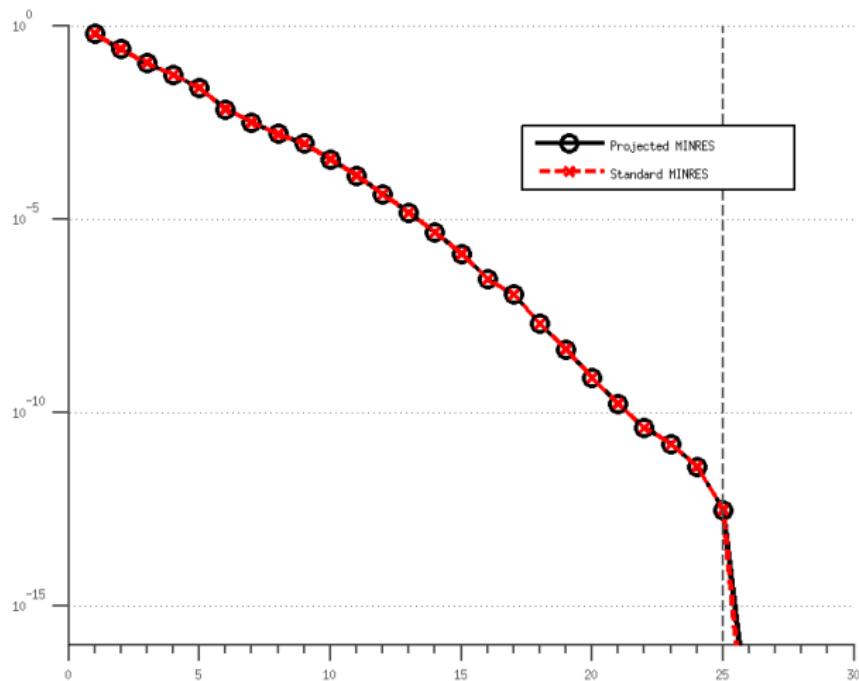
►  $B\mathbf{z}_k = 0$

$\mathbf{u}_{j+1} = 0$

$B\mathbf{z}_{j+1} = 0$



## Numerical Comparison





# Can you use MINRES with an indefinite preconditioner?

~~No~~  
~~Sometimes~~<sup>†</sup>

Tyrone Rees

STFC Rutherford Appleton Laboratory

Nick Gould (RAL), Dominique Orban (École Polytechnique de Montréal)

<sup>†</sup> Gould, N.I.M., Orban, D. and Rees, T.,

*Projected Krylov Methods for Saddle Point Systems* (in preparation)