On some recent algorithms for solving nonsymmetric algebraic Riccati equations

Beatrice Meini Joint work with D. Bini and F. Poloni

Due giorni di algebra lineare numerica Bologna, 6-7 Marzo, 2008



Nonsymmetric Algebraic Riccati Equations

Preliminaries Outline of SDA Outline of Cyclic Reduction

$NARE \rightarrow UQME$

Ramaswami's transform UL based transform "Small size" transform

Eigenvalues transform

Shrink and shift Cayley transform

Numerical results and conclusions



Algorithms	for	NAREs
NARE		

Preliminaries

Nonsymmetric Algebraic Riccati Equations

Given $D \in \mathbb{R}^{n \times n}$, $A \in \mathbb{R}^{m \times m}$, $C \in \mathbb{R}^{n \times m}$, $B \in \mathbb{R}^{m \times n}$, find $X \in \mathbb{R}^{m \times n}$ such that

NARE

$$XCX - AX - XD + B = 0 \tag{1}$$



Algorithms	for	NAREs
NARE		

Preliminaries

Nonsymmetric Algebraic Riccati Equations

Given $D \in \mathbb{R}^{n \times n}$, $A \in \mathbb{R}^{m \times m}$, $C \in \mathbb{R}^{n \times m}$, $B \in \mathbb{R}^{m \times n}$, find $X \in \mathbb{R}^{m \times n}$ such that

NARE

$$XCX - AX - XD + B = 0 \tag{1}$$

Remark: Any solution X of (1) is such that

$$\begin{bmatrix} D & -C \\ B & -A \end{bmatrix} \begin{bmatrix} I \\ X \end{bmatrix} = \begin{bmatrix} I \\ X \end{bmatrix} (D - CX)$$

The eigenvalues of D - CX are eigenvalues of $H = \begin{bmatrix} D & -C \\ B & -A \end{bmatrix}^{-\frac{1}{5}M^2}$

Important case

Assumption : assume that

$$M = \left[\begin{array}{cc} D & -C \\ -B & A \end{array} \right]$$

is either a nonsingular M-matrix or a singular irreducible M-matrix.



Preliminaries

Important case

Assumption : assume that

$$M = \left[\begin{array}{cc} D & -C \\ -B & A \end{array} \right]$$

is either a nonsingular M-matrix or a singular irreducible M-matrix. Spectral properties: let $\sigma(H) = \{\lambda_1, \lambda_2, \dots, \lambda_{m+n}\}$, with $\operatorname{Re}(\lambda_{m+n}) \leq \dots \leq \operatorname{Re}(\lambda_2) \leq \operatorname{Re}(\lambda_1)$.

- ▶ If *M* is nonsingular then $\operatorname{Re}(\lambda_{n+1}) < 0 < \operatorname{Re}(\lambda_n)$
- If *M* is singular, then Re(λ_{n+1}) ≤ 0 ≤ Re(λ_n). Moreover, only one of the following conditions is satisfied:
 - $\lambda_n = 0$ and $\lambda_{n+1} \in \mathbb{R}^-$ (positive recurrent case);
 - $\lambda_n \in \mathbb{R}^+$ and $\lambda_{n+1} = 0$ (transient case);
 - $\lambda_n = \lambda_{n+1} = 0$ (null recurrent case).

Università di Pisa



- Preliminaries

Location of the eigenvalues: singular case





Э

Algorithms for NAREs	
Preliminaries	

Interest

Compute the minimal entrywise nonnegative solution ${\it S}$ of the NARE (1)



◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへぐ

Interest

Compute the minimal entrywise nonnegative solution ${\it S}$ of the NARE (1)

Invariant subspace property:

The seeked solution S is the unique matrix such that

$$H\begin{bmatrix} I\\S\end{bmatrix} = \begin{bmatrix} I\\S\end{bmatrix}R, \quad R = D - CS,$$

and $\sigma(R) = \{\lambda_1, \dots, \lambda_n\}$. The solution *S* is called the extremal solution.

There are many algorithms for solving AREs based on the invariant subspace property. One of the most efficient is the Structure-preserving Doubling UNIVERSITAD (PISA Algorithm (SDA) by [Guo, Lin, Wei, 2006]

Algorithms for NAREs	
NARE	
Outline of SDA	

Outline of SDA

- Assume for simplicity that *M* is a nonsingular M-matrix. Therefore σ(*R*) = {λ₁,...,λ_n} ∈ C⁺.
- ► Apply the Cayley transform $z \to (z \gamma)/(z + \gamma)$ with $\gamma > 0$ to *R* and obtain

$$(H - \gamma I) \begin{bmatrix} I \\ S \end{bmatrix} = (H + \gamma I) \begin{bmatrix} I \\ S \end{bmatrix} R_{\gamma},$$

where $R_{\gamma} = (R + \gamma I)^{-1}(R - \gamma I)$.

Key property: $ho(R_{\gamma}) < 1$



Algorithms for NAREs	
NARE	
Uutline of SDA	

Outline of SDA

SDA generates the matrix sequences

$$L_k = \begin{bmatrix} D_k & 0 \\ -H_k & I \end{bmatrix}, \quad U_k = \begin{bmatrix} I & -G_k \\ 0 & F_k \end{bmatrix}$$

such that

$$L_k \begin{bmatrix} I \\ S \end{bmatrix} = U_k \begin{bmatrix} I \\ S \end{bmatrix} R_{\gamma}^{2^k}, \quad k = 0, 1, \dots$$

Since $\rho(R_{\gamma}) < 1$ then H_k quadratically converges to S



◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 三臣 - のへで

Algorithms for NAREs	
NARE	
Outline of SDA	

Outline of SDA

SDA generates the matrix sequences

$$L_k = \begin{bmatrix} D_k & 0 \\ -H_k & I \end{bmatrix}, \quad U_k = \begin{bmatrix} I & -G_k \\ 0 & F_k \end{bmatrix}$$

such that

$$L_k \begin{bmatrix} I \\ S \end{bmatrix} = U_k \begin{bmatrix} I \\ S \end{bmatrix} R_{\gamma}^{2^k}, \quad k = 0, 1, \dots$$

Since $\rho(R_{\gamma}) < 1$ then H_k quadratically converges to SCost: $\frac{64}{3}n^3$ ops per step (where we assume m = n). Remark: The convergence is still quadratic if M is singular irreducible and $\lambda_n \neq \lambda_{n+1}$. If $\lambda_n = \lambda_{n+1} = 0$ the convergence is linear with rate 1/2. The convergence turns to quadratic by applying a shift technique to the null eigenvalues of H [Guo, $\geq \infty$]

Algorithms for NAREs	
NARE	
Outline of CR	

Cyclic Reduction (CR)

CR is a versatile algorithm invented by G. Golub [Buzbee, Golub, Nielson 1970] for the f.d. Poisson equation.

- Rediscovered by Latouche and Ramaswami (1993) for QBDs
- ► Revisited by Bini and Meini (1996ff), applied to UQMEs and extended to equations of the kind $X = \sum_{i=0}^{+\infty} A_i X^i$

 Applied to the following matrix equations: X = A ± BX⁻¹C [Meini 2002]; matrix square and pth root (Bini, Higham, Meini 2005); NARE [Ramaswami 1999].

Details on this algorithm can be found in the book D. Bini, G. Latouche, B. Meini, "Numerical Solution of Structured Markov Chains", Oxford Univ. Press 2005.

-NARE

└─Outline of CR

Few words about CR for UQME

Given: $A_0, A_1, A_2 \in \mathbb{R}^{N \times N}$ such that the roots of $\varphi(\lambda) = \det(A_0 + A_1\lambda + A_2\lambda^2)$ are

 $|\xi_1| \leq \cdots \leq |\xi_N| \leq 1 < |\xi_{N+1}| \leq \cdots \leq |\xi_{2N}|$

(including zeros at ∞ if deg $\varphi(\lambda) < 2N$)





A	gorithms	for	NA	REs
~	gonums	5 101		1123

-NARE

-Outline of CR

Few words about CR for UQME

Given: $A_0, A_1, A_2 \in \mathbb{R}^{N \times N}$ such that the roots of $\varphi(\lambda) = \det(A_0 + A_1\lambda + A_2\lambda^2)$ are

 $|\xi_1| \leq \cdots \leq |\xi_N| \leq 1 < |\xi_{N+1}| \leq \cdots \leq |\xi_{2N}|$

(including zeros at ∞ if deg $arphi(\lambda) < 2N$)



Goal: compute the solution G of the Unilateral Quadratic Matrix Equation (UQME)

$$A_0 + A_1 X + A_2 X^2 = 0,$$

such that $\rho(G) = |\xi_N|$, provided it exists.



Algorithms for NAREs
NARE
Outline of CR

Few words about CR for UQME

CR generates the matrix sequences

$$\begin{aligned} A_0^{(k+1)} &= -A_0^{(k)} S^{(k)} A_0^{(k)}, \quad S^{(k)} = (A_1^{(k)})^{-1} \\ A_2^{(k+1)} &= -A_2^{(k)} S^{(k)} A_2^{(k)}, \\ A_1^{(k+1)} &= A_1^{(k)} - A_0^{(k)} S^{(k)} A_2^{(k)} - A_2^{(k)} S^{(k)} A_0^{(k)}, \\ \widehat{A}^{(k+1)} &= \widehat{A}^{(k)} - A_0^{(k)} S^{(k)} A_2^{(k)}, \quad k \ge 0 \end{aligned}$$

starting from $A_i^{(0)} = A_i$, i = 1, 2, 3, $\widehat{A}^{(0)} = A_1$, such that

$$A_0 + \widehat{A}^{(k)}G + A_2^{(k)}G^{2^k+1} = 0$$

Convergence property: the convergence is quadratic, more specifically:



$$||(\widehat{A}^{(k)})^{-1}A_0 - G|| = O(|\xi_N / \xi_{N+1}|^{2^k})$$

Algorithms for NAREs	
NARE	
Outline of CR	

Few words about CR for UQME

Cost: 6 matrix products, one PLU factorization: $\frac{38}{3}N^3$ ops Applicability: under mild conditions the matrices $A_0^{(k)}$ are invertible Critical case: If $|\xi_N| = |\xi_{N+1}| = 1$ convergence turns to linear with rate 1/2. Quadratic convergence can be recovered by means of the shift technique [He, Meini, Rhee, 01].



New class of algorithms

Idea: To transform the NARE into a UQME of the kind

$$A_0 + A_1Y + A_2Y^2 = 0, \quad A_0, A_1, A_2 \in \mathbb{R}^{N imes N}$$

with $N \leq m + n$, such that $\det(A_0 + A_1\lambda + A_2\lambda^2)$ has roots

 $|\xi_1| \leq \cdots \leq |\xi_N| \leq 1 \leq |\xi_{N+1}| \leq \cdots \leq |\xi_{2N}|$

and apply cyclic reduction.



New class of algorithms

Idea: To transform the NARE into a UQME of the kind

$$A_0 + A_1Y + A_2Y^2 = 0, \quad A_0, A_1, A_2 \in \mathbb{R}^{N imes N}$$

with $N \leq m + n$, such that $det(A_0 + A_1\lambda + A_2\lambda^2)$ has roots

 $|\xi_1| \leq \cdots \leq |\xi_N| \leq 1 \leq |\xi_{N+1}| \leq \cdots \leq |\xi_{2N}|$

and apply cyclic reduction.

H.-G. Xu and L.-Z. Lu (1995) reduced an ARE to an equation $Y^2 - M^2 = 0$ but with no splitting property.



Ramaswami's transform

The linear matrix pencil

$$H - \lambda I = \begin{bmatrix} D & -C \\ B & -A \end{bmatrix} - \lambda I$$

can be transformed into a quadratic matrix polynomial by multiplying the second block column by λ

$$A(\lambda) = \begin{bmatrix} D & 0 \\ B & 0 \end{bmatrix} + \begin{bmatrix} -I & -C \\ 0 & -A \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & -I \end{bmatrix} \lambda^2$$

This matrix polynomial defines a UQME

$$\begin{bmatrix} D & 0 \\ B & 0 \end{bmatrix} + \begin{bmatrix} -I & -C \\ 0 & -A \end{bmatrix} Y + \begin{bmatrix} 0 & 0 \\ 0 & -I \end{bmatrix} Y^2 = 0 \qquad (2)$$

16.01

Ramaswami's transform

Theorem

The roots of the matrix polynomial $A(\lambda)$ are:

- ▶ *m* equal to 0
- the m + n eigenvalues $\lambda_1, \ldots, \lambda_{m+n}$ of H
- n at infinity.

Moreover

$$V = \left[\begin{array}{cc} D - CS & 0 \\ S & 0 \end{array} \right],$$

where *S* is the extremal solution of (1), is the unique solution of the UQME (2) with *m* eigenvalues equal to $\lambda_1, \ldots, \lambda_n$.

UL based transform

Consider the block UL factorization

$$H = U^{-1}L, \quad U = \begin{bmatrix} I & -U_1 \\ 0 & U_2 \end{bmatrix}, \quad L = \begin{bmatrix} L_1 & 0 \\ -L_2 & I \end{bmatrix},$$

and transform the pencil $H - \lambda I$ into the new pencil

 $L - \lambda U$.



◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 三臣 - のへで

UL based transform

Consider the block UL factorization

$$H = U^{-1}L, \quad U = \begin{bmatrix} I & -U_1 \\ 0 & U_2 \end{bmatrix}, \quad L = \begin{bmatrix} L_1 & 0 \\ -L_2 & I \end{bmatrix},$$

and transform the pencil $H - \lambda I$ into the new pencil

 $L - \lambda U$.

Now multiply the second block row by $-\lambda$ and get

$$A(\lambda) = \begin{bmatrix} L_1 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} -I & U_1 \\ L_2 & -I \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & U_2 \end{bmatrix} \lambda^2,$$

which defines the UQME

$$\begin{bmatrix} L_1 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} -I & U_1 \\ L_2 & -I \end{bmatrix} Y + \begin{bmatrix} 0 & 0 \\ 0 & U_2 \end{bmatrix} Y^2 = 0$$



UL based transform

Theorem

The roots of the matrix polynomial $A(\lambda)$ are:

- m equal to 0
- the m + n eigenvalues $\lambda_1, \ldots, \lambda_{m+n}$ of H
- n at infinity.

Moreover

$$V = \left[\begin{array}{cc} D - CS & 0 \\ S(D - CS) & 0 \end{array} \right],$$

where *S* is the extremal solution of (1), is the unique solution of the UQME (3) with *m* eigenvalues equal to $\lambda_1, \ldots, \lambda_n$.

"Small size" transform

The matrix pencil $H - \lambda I$ is transformed into

$$\begin{bmatrix} I & 0 \\ -U & I \end{bmatrix} H \begin{bmatrix} I & 0 \\ -U & I \end{bmatrix}^{-1} - \lambda I.$$
 (4)

If det $C \neq 0$, by choosing $U = C^{-1}D$, (4) becomes

$$\begin{bmatrix} 0 & I \\ R(C^{-1}D) & A - C^{-1}DC \end{bmatrix} - \lambda I,$$

where R(U) = UCU - AU - UD + B, which defines the UQME

$$(B - AC^{-1}D)C + (C^{-1}DC - A)Y + Y^{2} = 0$$



"Small size" transform

Theorem The roots of

$$A(\lambda) = (B - AC^{-1}D)C + (C^{-1}DC - A)\lambda + I\lambda^{2}$$

are the eigenvalues of *H*. Moreover, $Y = C^{-1}(D - CS)C$ is the unique solution of the UQME

$$(B - AC^{-1}D)C + (C^{-1}DC - A)Y + Y^{2} = 0$$

with eigenvalues $\lambda_1, \ldots, \lambda_n$.



"Small size" transform

Remark: The condition det $C \neq 0$ is not restrictive. Indeed, X solves (1) if and only if $\tilde{X} = X(I - MX)^{-1}$ solves

$$Y\widetilde{C}Y-\widetilde{A}Y-Y\widetilde{D}+\widetilde{B}=0,$$

where M is any matrix such that $det(I - MX) \neq 0$, and

$$\begin{split} \widetilde{A} &= A - BM, \quad \widetilde{B} = B, \\ \widetilde{C} &= \widetilde{R}(M), \quad \widetilde{D} = D - MB, \\ \widetilde{R}(M) &= MBM - DM - MA + C \end{split}$$

Open issue: Find M such that $\widetilde{R}(M)$ is well-conditioned.



Algorithms for NAREs
└─NARE → UQME
-3rd transform

The UQMEs of size m + n are associated with matrix polynomials of the kind

$$A(\lambda) = \begin{cases} \begin{bmatrix} * & 0 \\ * & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ 0 & * \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & -I \end{bmatrix} \lambda^2 \\ \begin{bmatrix} * & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ * & I \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & * \end{bmatrix} \lambda^2$$



◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 三臣 - のへで

Algorithms for NAREs
└─NARE → UQME
-3rd transform

The UQMEs of size m + n are associated with matrix polynomials of the kind

$$A(\lambda) = \begin{cases} \begin{bmatrix} * & 0 \\ * & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ 0 & * \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & -I \end{bmatrix} \lambda^2 \\ \begin{bmatrix} * & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ * & I \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & * \end{bmatrix} \lambda^2$$

• The eigenvalues of *H* are roots of det $A(\lambda)$.



Algorithms for NAREs
└─NARE → UQME
-3rd transform

The UQMEs of size m + n are associated with matrix polynomials of the kind

$$A(\lambda) = \begin{cases} \begin{bmatrix} * & 0 \\ * & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ 0 & * \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & -I \end{bmatrix} \lambda^2 \\ \begin{bmatrix} * & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ * & I \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & * \end{bmatrix} \lambda^2$$

- The eigenvalues of *H* are roots of det $A(\lambda)$.
- The nonzero roots of det A(λ) have a splitting w.r.t. the imaginary axis



The UQMEs of size m + n are associated with matrix polynomials of the kind

$$A(\lambda) = \begin{cases} \begin{bmatrix} * & 0 \\ * & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ 0 & * \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & -I \end{bmatrix} \lambda^2 \\ \begin{bmatrix} * & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} -I & * \\ * & I \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0 \\ 0 & * \end{bmatrix} \lambda^2$$

- The eigenvalues of *H* are roots of $det A(\lambda)$.
- The nonzero roots of det A(λ) have a splitting w.r.t. the imaginary axis
- The solution of the UQME associated with the eigenvalues with the largest real part is the one to be computed



Algorithms for UQME reach the highest efficiency for eigenvalues split w.r.t. the unit circle where the solution with eigenvalues of modulus less than 1 is seeked.

Three approaches to transform a splitting w.r.t the imaginary axis into a splitting w.r.t. the unit circle:

- shrink and shift (Ramaswami 1999)
- Cayley transform applied to the pencil (Guo, Lin, Wei, 2006)
- Cayley transform applied to the UQME (Bini, Latouche, Meini, 2006)



Shrink and shift

Shrink and shift

Multiply the Riccati equation by t,

tXCX - tAX - tXD + tB = 0,



・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・
・

Shrink and shift

Shrink and shift

Multiply the Riccati equation by t,

$$tXCX - tAX - tXD + tB = 0,$$

add / to -tA and subtract / from -tD and get:

$$tXCX - (tA - I)X - X(tD + I) + tB = 0$$
 (5)

The associated matrix is

$$H_t = \left[\begin{array}{cc} I + tD & -tC \\ tB & I - tA \end{array} \right]$$



Shrink and shift

Shrink and shift

Multiply the Riccati equation by t,

$$tXCX - tAX - tXD + tB = 0,$$

add I to -tA and subtract I from -tD and get:

$$tXCX - (tA - I)X - X(tD + I) + tB = 0$$
 (5)

UNIVERSITÀ DI PISA

The associated matrix is

$$H_t = \left[\begin{array}{cc} I + tD & -tC \\ tB & I - tA \end{array} \right]$$

If $0 < t < 1/\max(a_{i,i}, d_{i,i})$ the eigenvalues of H_t have a splitting w.r.t. the unit circle

Shrink and shift

Transformation of the eigenvalues



Original eigenvalues





◆□▶ ◆□▶ ◆ □▶ ◆ □▶ □ のへぐ

Shrink by t

Shift by 1



Cayley transform applied to the pencil

The Cayley transform z → (z − γ)/(z + γ) applied to the pencil H − λI yields the pencil

$$H_{\gamma} - \lambda I$$
, $H_{\gamma} = (H + \gamma I)^{-1}(H - \gamma I)$.



Cayley transform applied to the pencil

• The Cayley transform $z \to (z - \gamma)/(z + \gamma)$ applied to the pencil $H - \lambda I$ yields the pencil

$$H_{\gamma} - \lambda I$$
, $H_{\gamma} = (H + \gamma I)^{-1}(H - \gamma I)$.

Since $\mu = \frac{\gamma - \lambda}{\gamma + \lambda}$ is eigenvalue of H_{γ} iff λ is eigenvalue of H, the eigenvalues of H_{γ} are split w.r.t. the unit circle.



Cayley transform applied to the pencil

• The Cayley transform $z \to (z - \gamma)/(z + \gamma)$ applied to the pencil $H - \lambda I$ yields the pencil

$$H_{\gamma} - \lambda I$$
, $H_{\gamma} = (H + \gamma I)^{-1}(H - \gamma I)$.

- Since $\mu = \frac{\gamma \lambda}{\gamma + \lambda}$ is eigenvalue of H_{γ} iff λ is eigenvalue of H, the eigenvalues of H_{γ} are split w.r.t. the unit circle.
- ► Three UQMEs can be obtained from the pencil $H_{\gamma} \lambda I$. The *UL*-based transform yields

$$A(\lambda) = \begin{bmatrix} -D_{\gamma} & 0\\ 0 & 0 \end{bmatrix} + \begin{bmatrix} I & -G_{\gamma}\\ -H_{\gamma} & I \end{bmatrix} \lambda + \begin{bmatrix} 0 & 0\\ 0 & -F_{\gamma} \end{bmatrix} \lambda_{\text{INVERSITÀ DI PIS/}}^{2}$$

SDA is CR!

Theorem Cyclic Reduction applied to

$$\begin{bmatrix} -D_{\gamma} & 0\\ 0 & 0 \end{bmatrix} + \begin{bmatrix} I & -G_{\gamma}\\ -H_{\gamma} & I \end{bmatrix} Y + \begin{bmatrix} 0 & 0\\ 0 & -F_{\gamma} \end{bmatrix} Y^{2} = 0 \quad (6)$$

coincides with SDA. Moreover, the spectral minimal solution of (6) is $\begin{bmatrix} R_{\gamma} & 0\\ SR_{\gamma} & 0 \end{bmatrix}$, where $R_{\gamma} = (R + \gamma I)^{-1}(R - \gamma I)$.



Cayley transform

Some theoretical results

Theorem

Assume that M is nonsingular and let $Q(\lambda) = \lambda^{-1}A(\lambda)$. Then:

• The matrix function $Q(\lambda)$ is analytic for $|\xi| < |z| < |\eta|$, where $\xi = (\lambda_n - \gamma)/(\lambda_n + \gamma)$, $\eta = (\lambda_{n+1} - \gamma)/(\lambda_{n+1} + \gamma)$.



- Cayley transform

Some theoretical results

Theorem

Assume that M is nonsingular and let $Q(\lambda) = \lambda^{-1}A(\lambda)$. Then:

- The matrix function $Q(\lambda)$ is analytic for $|\xi| < |z| < |\eta|$, where $\xi = (\lambda_n \gamma)/(\lambda_n + \gamma)$, $\eta = (\lambda_{n+1} \gamma)/(\lambda_{n+1} + \gamma)$.
- $\mathcal{Q}(\lambda)$ has the canonical factorization

$$\mathcal{Q}(\lambda) = \left(I - \lambda \begin{bmatrix} 0 & 0 \\ W & WS \end{bmatrix}\right) \left[\begin{array}{cc} I & -G_{\gamma} \\ -S & I \end{array}\right] \left(I - \lambda^{-1} \left[\begin{array}{cc} R_{\gamma} & 0 \\ SR_{\gamma} & 0 \end{bmatrix}\right)$$



Cayley transform

Some theoretical results

Theorem

Assume that M is nonsingular and let $Q(\lambda) = \lambda^{-1}A(\lambda)$. Then:

- ► The matrix function $Q(\lambda)$ is analytic for $|\xi| < |z| < |\eta|$, where $\xi = (\lambda_n \gamma)/(\lambda_n + \gamma)$, $\eta = (\lambda_{n+1} \gamma)/(\lambda_{n+1} + \gamma)$.
- $\mathcal{Q}(\lambda)$ has the canonical factorization

$$\mathcal{Q}(\lambda) = \left(I - \lambda \begin{bmatrix} 0 & 0 \\ W & WS \end{bmatrix}\right) \left[\begin{array}{cc} I & -G_{\gamma} \\ -S & I \end{array}\right] \left(I - \lambda^{-1} \left[\begin{array}{cc} R_{\gamma} & 0 \\ SR_{\gamma} & 0 \end{bmatrix}\right)$$

• The series $\psi(\lambda) = \mathcal{Q}(\lambda)^{-1}$, $\psi(\lambda) = \sum_{k=-\infty}^{+\infty} \lambda^k \psi_k$ is such that

$$\psi_0^{-1} = \left[\begin{array}{cc} I & -T \\ -S & I \end{array} \right]$$

UNIVERSITÀ DI PISA

where T is the solution of the dual NARE of (1).

Different combinations \rightarrow different algorithms

We may combine the different strategies, for instance:

"Shrink and shift" + "Ramaswami transform" lead to an algorithm similar to that of Ramaswami (1999) of cost (68/3)n³ ops per step (ss-ram).



Different combinations \rightarrow different algorithms

We may combine the different strategies, for instance:

- "Shrink and shift" + "Ramaswami transform" lead to an algorithm similar to that of Ramaswami (1999) of cost (68/3)n³ ops per step (ss-ram).
- "Cayley transform" + "UL-based reduction" lead to SDA, having a cost (64/3)n³ per step (sda).



Different combinations \rightarrow different algorithms

We may combine the different strategies, for instance:

- "Shrink and shift" + "Ramaswami transform" lead to an algorithm similar to that of Ramaswami (1999) of cost (68/3)n³ ops per step (ss-ram).
- "Cayley transform" + "UL-based reduction" lead to SDA, having a cost (64/3)n³ per step (sda).
- "Shrink and shift" + "UL-based reduction" lead to a new algorithm with the same cost of SDA. Formally, this algorithm differs from SDA only for the initial values, which are simpler (ss-ul).



Different combinations \rightarrow different algorithms

We may combine the different strategies, for instance:

- "Shrink and shift" + "Ramaswami transform" lead to an algorithm similar to that of Ramaswami (1999) of cost $(68/3)n^3$ ops per step (ss-ram).
- "Cayley transform" + "UL-based reduction" lead to SDA, having a cost $(64/3)n^3$ per step (sda).
- "Shrink and shift" + "UL-based reduction" lead to a new algorithm with the same cost of SDA. Formally, this algorithm differs from SDA only for the initial values, which are simpler (ss-ul).
- "Cayley transform" + "Small-size transform" lead to a new algorithm, having a cost $(38/3)n^3$ (nodoub). UNIVERSITÀ DI PISA



NARE deriving from a problem in neutron transport theory

$$A = \widehat{\Delta} - eq^T, \hspace{0.2cm} B = ee^T, \hspace{0.2cm} C = qq^T, \hspace{0.2cm} D = \Delta - qe^T$$

with

$$\begin{split} \Delta &= \operatorname{diag}(\delta_1, \dots, \delta_n), \qquad \widehat{\Delta} &= \operatorname{diag}(\widehat{\delta}_1, \dots, \widehat{\delta}_n), \\ \delta_i &= \frac{1}{cx_i(1-\alpha)}, \qquad \widehat{\delta}_i &= \frac{1}{cx_i(1+\alpha)}, \quad i = 1, \dots, n, \\ e &= \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix}^T, \qquad q_i &= \frac{w_i}{2x_i}, \quad i = 1, \dots, n, \end{split}$$

 $(x_i)_{i=1}^n$ and $(w_i)_{i=1}^n$ being the nodes and weights of a Gaussian discretization. Here we have chosen $\alpha = 10^{-8}$, $c = 1 - 10^{-6}$, which yields a close-to-null-recurrent Riccati equation.



Running time in seconds

n	sda	ss-ul	ss-ram	nodoub
8	0.045209	0.02735	0.030078	0.027061
16	0.039896	0.041282	0.046027	0.03845
32	0.14559	0.14666	0.18047	0.13432
64	0.92806	0.93415	1.1707	0.8448
128	7.2632	7.3491	9.1974	6.6499
256	60.841	61.926	76.835	55.03
512	499.95	504.37	625.06	448.46



Residual errors

п	sda	ss-ul	ss-ram	nodoub
8	1.654e-13	5.8367e-14	6.6482e-14	1.4294e-11
16	1.328e-12	2.4418e-13	2.7769e-13	1.6405e-10
32	3.4631e-12	1.964e-12	1.7786e-12	7.8717e-10
64	2.2679e-11	1.3598e-11	8.2769e-12	7.8282e-09
128	1.3316e-10	8.1521e-11	6.4269e-11	5.4047e-08
256	1.0096e-09	5.6852e-10	3.7115e-10	4.5315e-07
512	6.7923e-09	4.2861e-09	1.7767e-09	5.4083e-06



◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへぐ

Conclusions and open issues

- The interpretation provided in this talk casts new light on the SDA algorithm and on the relationship between UQMEs and NAREs.
- Several other approaches to the solution of the NARE can be developed with this new setting. Among the possible ideas:
 - using numerical integration and the Cauchy integral theorem for computing the matrix ψ_0 ;
 - using functional iterations borrowed from stochastic processes (QBD) for solving the UQME;
 - using Newton's iteration applied to the UQME trying to exploit the specific matrix structure.
- ► It would be important to find for more general transformations, which map a Hamiltonian matrix *H* to a new one *H* where the block *H*_{1,2} is not only nonsingular but numerically well conditioned.