

# Fast inclusion for the matrix inverse square root

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## Inverse square root of nonsingular $A \in \mathbb{C}^{n \times n}$

is  $X^* \in \mathbb{C}^{n \times n}$  s.t.  $X^{*2} A = I_n$ ,

always exists for  $A$  being nonsingular,

is not unique.

If  $\lambda_i(A) \notin \mathbb{R}_-$ ,  $\forall i$ , the **principal** inverse square root can be **uniquely** defined by requiring  $\operatorname{Re} \lambda_i(X^*) > 0$ ,  $\forall i$ .

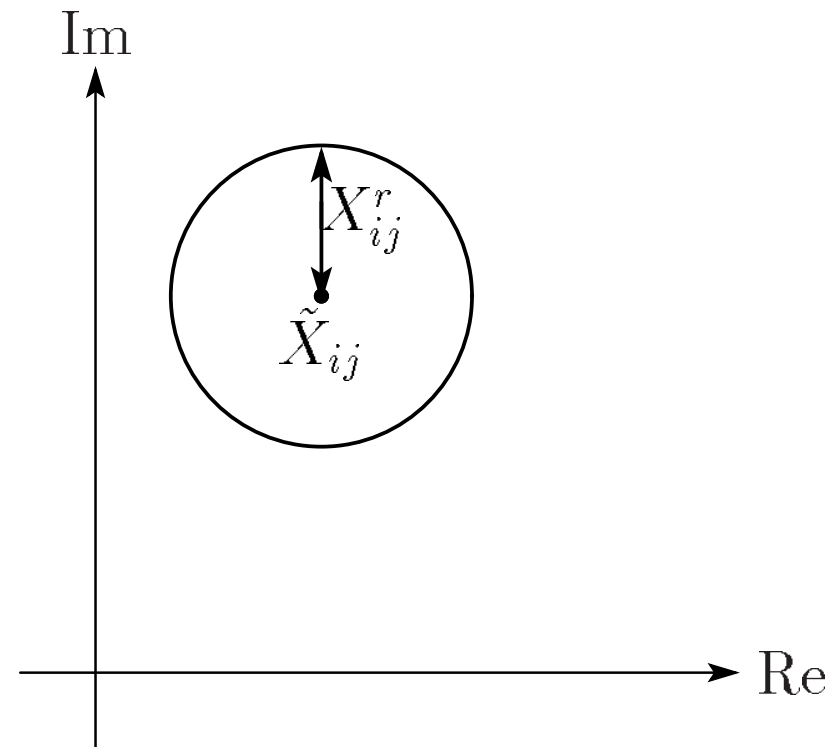
The principal inverse square root is of particular interest.

# Purpose

Numerically computing  $\tilde{X}$  and  $X^r$   
s.t.  $X^* \in \langle \tilde{X}, X^r \rangle$

Preferable

- Smaller  $X_{ij}^r$
- Fast algorithm



## Previous work

2 algorithms by Frommer-Hashemi-Sablik (2014)

- numerical spectral decomposition is utilized  $\Rightarrow \mathcal{O}(n^3)$  operations
- Krawczyk operator is applied for nonlinear matrix equations

**1st algorithm** ( $F(X) = 0$ , where  $F(X) := XAX - I_n$ )

The uniqueness of the contained inverse square root is verified

**2nd algorithm** (an affine transformation of  $F(X) = 0$ )

Although the uniqueness is not verified, more effective than the 1st

## Our contribution

Algorithm for computing  $\tilde{X}$  and  $X^r$  s.t.  $X^* \in \langle \tilde{X}, X^r \rangle$

utilizes the spectral decomposition  $\Rightarrow \mathcal{O}(n^3)$  operations

adopts the affine transformation and Newton operator

verifies the **principal property** and uniqueness

## Affine transformation (1/2)

Assume we have diagonal  $\tilde{D} \in \mathbb{C}^{n \times n}$  and  $\tilde{V}, W \in \mathbb{C}^{n \times n}$  s.t.  $A\tilde{V} \approx \tilde{V}\tilde{D}$ ,  $W \approx \tilde{V}^{-1}$ , and  $\tilde{V}$  and  $W$  are nonsingular (verifiable).

$Q := \tilde{D} - WA\tilde{V}$ ,  $S := I_n - W\tilde{V}$  ( $Q \approx 0$ ,  $S \approx 0$ ),  $Y := \tilde{V}^{-1}XW^{-1}$

$$\begin{aligned}
 F(X) = 0 &\Leftrightarrow \tilde{V}^{-1}XW^{-1}WA\tilde{V}\tilde{V}^{-1}XW^{-1} - \tilde{V}^{-1}W^{-1} = 0 \\
 &\Leftrightarrow Y(\tilde{D} - (\tilde{D} - WA\tilde{V}))Y - (W\tilde{V})^{-1} = 0 \\
 &\Leftrightarrow Y(\tilde{D} - Q)Y - (I_n - S)^{-1} = 0 \\
 &\Leftrightarrow Y(\tilde{D} - Q)Y - I_n - (I_n - S)^{-1}S = 0.
 \end{aligned}$$

## Affine transformation (2/2)

$$F(X) = 0 \Leftrightarrow \hat{F}(Y) = 0, \hat{F}(Y) := Y(\tilde{D} - Q)Y - I_n - (I_n - S)^{-1}S.$$

We compute an interval matrix  $\mathbf{Y}$  s.t.  $Y^* \in \mathbf{Y}$  and  $\hat{F}(Y^*) = 0$

and enclose  $X^*$  by computing the superset of  $\{\tilde{V}YW : Y \in \mathbf{Y}\}$ .

Since  $X^* = \tilde{V}Y^*W$ , the superset contains  $X^*$ .

The superset can be obtained by interval arithmetic.

## How to obtain $Y$ ?

$\hat{F}'_Y(H) = Y(\tilde{D} - Q)H + H(\tilde{D} - Q)Y$ . Let  $\tilde{Y} \approx Y^*$  be diagonal.

If  $\hat{F}'_{\tilde{Y}}(H)$  is invertible,  $\hat{N}(Y) := Y - (\hat{F}'_{\tilde{Y}})^{-1}(\hat{F}(Y))$ .

$\hat{N}(Y) = Y$  is a fixed point equation for  $Y$ .

We thus verify the invertibility and

$\{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\} \subseteq \text{int}(\langle \tilde{Y}, Y^r \rangle)$  for given  $Y^r \in \mathbb{R}^{n \times n}$ .

The Brouwer theorem then implies  $Y^* \in \{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\}$ .

$\mathbf{Y}$  can be obtained s.t.  $\{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\} \subseteq \mathbf{Y}$ .



## How to verify the invertibility?

$\hat{F}'_{\tilde{Y}}(H)$  can be represented in terms of matrix vector products as

$$\text{vec}(\hat{F}'_{\tilde{Y}}(H)) = \hat{P} \text{vec}(H), \quad \hat{P} := I_n \otimes \tilde{Y}(\tilde{D} - Q) + \tilde{Y}^T(\tilde{D} - Q)^T \otimes I_n$$

If  $\hat{P}$  is nonsingular,  $\hat{F}'_{\tilde{Y}}(H)$  is invertible. We verify the nonsingularity.

$\hat{P}$  is a  $n^2 \times n^2$  matrix.

By exploiting the sparsity of  $\tilde{D}$  and  $\tilde{Y}$ , however, this can be achieved with  $\mathcal{O}(n^3)$  operations.

$$\{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\} \subseteq \text{int}(\langle \tilde{Y}, Y^r \rangle)? \quad (1/2)$$

We compute  $Y^\varepsilon \in \mathbb{R}^{n \times n}$  s.t.  $\{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\} \subseteq \langle \tilde{Y}, Y^\varepsilon \rangle$

and verify  $Y^\varepsilon < Y^r$ .  $Y^\varepsilon$  can be obtained by the following idea:

$$\hat{N}(Y) = Y - (\hat{F}'_{\tilde{Y}})^{-1}(\hat{F}(Y)) \Leftrightarrow \hat{F}'_{\tilde{Y}}(\hat{N}(Y)) = \hat{F}'_{\tilde{Y}}(Y) - \hat{F}(Y)$$

$$\text{LHS} = \tilde{Y}(\tilde{D} - Q)\hat{N}(Y) + \hat{N}(Y)(\tilde{D} - Q)\tilde{Y}$$

$$\text{RHS} = \tilde{Y}(\tilde{D} - Q)Y + Y(\tilde{D} - Q)\tilde{Y} - Y(\tilde{D} - Q)Y + I_n + (I_n - S)^{-1}S$$

$$\{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\} \subseteq \text{int}(\langle \tilde{Y}, Y^r \rangle)? \quad (2/2)$$

$\{\hat{N}(Y) : Y \in \langle \tilde{Y}, Y^r \rangle\}$  is the set of all solutions of an equation

$$\tilde{Y}(\tilde{D} - Q)\hat{N}_Y + \hat{N}_Y(\tilde{D} - Q)\tilde{Y}$$

$$= \tilde{Y}(\tilde{D} - Q)Y + Y(\tilde{D} - Q)\tilde{Y} - Y(\tilde{D} - Q)Y + I_n + (I_n - S)^{-1}S,$$

where  $\hat{N}_Y \in \mathbb{C}^{n \times n}$  is unknown and  $Y \in \langle \tilde{Y}, Y^r \rangle$  is the parameter.

We hence enclose the solution set, which can be achieved with  $\mathcal{O}(n^3)$  operations.

## How to determine $Y^r$ ?

In the computation of  $Y^\varepsilon$ ,  $Y^r$  is required.

Let  $Y_0^\varepsilon$  be  $Y^\varepsilon$  when  $Y^r = 0$ . For  $\eta \in \mathbb{R}$ , we put  $Y^r = \eta Y_0^\varepsilon$

and determine  $\eta$  s.t.  $Y^\varepsilon < \eta Y_0^\varepsilon = Y^r$ .

**Note** The determination of  $\eta$  includes some assumptions. If one of the assumptions cannot be verified, this determination fails.

## How to verify the principal property?

Let  $X^* \in \langle \tilde{X}, X^r \rangle$  and  $X \in \langle \tilde{X}, X^r \rangle$  be arbitrary.

We show  $\operatorname{Re}\lambda_i(X) > 0, \forall i$ . Then  $\operatorname{Re}\lambda_i(X^*) > 0, \forall i$ .

Since the principal square root is unique, the **uniqueness** can also be verified.

By reusing the previously computed matrices, this can be achieved with  $\mathcal{O}(n^2)$  operations.

## The uniqueness when $\lambda_i(A) \in \mathbb{R}_-$ ? (1/2)

Let  $X^*$  and  $X^{**}$ , and  $X_1$  and  $X_2$  be the inverse square roots and arbitrarily matrices contained in  $\langle \tilde{X}, X^r \rangle$ , respectively. We show the invertibility of  $F'_{\tilde{V}\tilde{Y}W}(H)$ , use  $N(X) := X - (F'_{\tilde{V}\tilde{Y}W})^{-1}(F(X))$  (Note  $N(X^*) = X^*$ ,  $N(X^{**}) = X^{**}$ ), derive  $S(X_1, X_2)$  s.t.

$$\text{vec}(N(X_1) - N(X_2)) = S(X_1, X_2)\text{vec}(X_1 - X_2),$$

and prove  $\|S(X_1, X_2)\|_\infty < 1$ , which gives  $\|S(X^*, X^{**})\|_\infty < 1$ .

## The uniqueness when $\lambda_i(A) \in \mathbb{R}_-$ ? (2/2)

We obtain

$$\begin{aligned}\|\text{vec}(X^* - X^{**})\|_\infty &= \|\text{vec}(N(X^*) - N(X^{**}))\|_\infty \\ &= \|\mathbf{S}(X^*, X^{**})\text{vec}(X^* - X^{**})\|_\infty \\ &\leq \|\mathbf{S}(X^*, X^{**})\|_\infty \|\text{vec}(X^* - X^{**})\|_\infty,\end{aligned}$$

so that  $(1 - \|\mathbf{S}(X^*, X^{**})\|_\infty) \|\text{vec}(X^* - X^{**})\|_\infty \leq 0$ .

This and  $\|\mathbf{S}(X^*, X^{**})\|_\infty < 1$  implies  $X^* = X^{**}$ .

These can be achieved with  $\mathcal{O}(n^3)$  operations.

## Numerical results (1/2)

Intel Xeon 2.66GHz Dual CPU, 4.00GB RAM, MATLAB 7.5 with Intel MKL, and IEEE 754 double precision

FHS1: 1st algorithm by Frommer-Hashemi-Sablik (2014)  
(the uniqueness is verified)

FHS2: 2nd algorithm by Frommer-Hashemi-Sablik (2014)  
(the uniqueness is not verified)

M: the proposed algorithm  
(the principal property and uniqueness are verified)

–: FHS1 or FHS2 did not succeed after 30 iterations



## Numerical results (2/2)

Let  $X^* \in \langle \tilde{X}, X^r \rangle$ ,  $\text{mrp} := \max_{i,j} X_{ij}^p$  and  $\text{arp} := (\sum_{i,j=1}^n X_{ij}^p) / n^2$ ,

$$\text{where } X_{ij}^p := \begin{cases} \min \left( 1, \frac{X_{ij}^r}{|\tilde{X}_{ij}|} \right) & (\tilde{X}_{ij} \neq 0) \\ 1 & (\tilde{X}_{ij} = 0) \end{cases} .$$

mrp and arp by  $\underline{M}$  with underline: both of the principal property and uniqueness could not be verified

## Results for “frank” matrix

$n$	FHS1	FHS2	M	FHS1	FHS2	M
	mrp	mrp	mrp	time	time	time
	arp	arp	arp			
7	2.6e-10	2.6e-10	6.2e-11	4.3e-1	1.3e-2	3.4e-3
	4.4e-11	4.4e-11	1.6e-11			
8	–	5.5e-9	1.0e-9	1.7e-1	1.1e-2	3.4e-3
	–	1.0e-9	2.1e-10			
9	–	–	1.5e-8	1.4e-1	1.3e-1	3.3e-3
	–	–	2.9e-9			
12	–	–	<u>1.4e-1</u>	1.4e-1	1.3e-1	4.1e-3
	–	–	<u>2.8e-3</u>			
13	–	–	fail	1.4e-1	1.3e-1	3.5e-3
	–	–	fail			

## Results for “poisson” matrix

$n$	FHS1	FHS2	M	FHS1	FHS2	M
	mrp	mrp	mrp	time	time	time
	arp	arp	arp			
400	5.7e-7	5.5e-7	4.3e-8	3.7e+0	3.6e+0	8.2e-1
	3.0e-9	2.8e-9	2.2e-10			
625	–	3.2e-6	1.9e-7	1.4e+2	1.3e+1	2.8e+0
	–	9.6e-9	5.6e-10			
900	–	1.4e-5	6.6e-7	3.8e+2	3.5e+1	7.9e+0
	–	2.7e-8	1.2e-9			
1225	–	5.0e-5	2.0e-6	9.2e+2	8.6e+1	1.9e+1
	–	6.0e-8	2.4e-9			
1600	–	1.5e-4	4.7e-6	1.9e+3	1.8e+2	4.2e+1
	–	1.3e-7	4.1e-9			

## Results for “prolate” matrix

$n$	FHS1	FHS2	M	FHS1	FHS2	M
	mrp	mrp	mrp	time	time	time
	arp	arp	arp			
10	4.6e-9	4.6e-9	6.2e-10	6.7e-2	1.2e-2	3.6e-3
	3.5e-10	3.5e-10	1.8e-10			
12	–	2.2e-7	1.9e-8	1.4e-1	3.6e-2	3.4e-3
	–	1.5e-8	5.7e-9			
14	–	–	1.6e-6	1.4e-1	1.3e-1	3.5e-3
	–	–	4.0e-7			
20	–	–	<u>2.3e-1</u>	1.5e-1	1.4e-1	4.4e-3
	–	–	<u>4.8e-2</u>			
22	–	–	fail	1.6e-1	1.4e-1	3.6e-3
	–	–	fail			

## Results for “toeppen” matrix

$n$	FHS1	FHS2	M	FHS1	FHS2	M
	mrp	mrp	mrp	time	time	time
	arp	arp	arp			
400	1.0e+0	1.0e+0	1.0e+0	1.4e+1	1.4e+1	2.8e+0
	2.0e-1	2.0e-1	1.7e-1			
600	–	1.0e+0	1.0e+0	4.1e+2	4.0e+1	7.8e+0
	–	4.1e-1	3.8e-1			
800	–	1.0e+0	1.0e+0	8.8e+2	8.8e+1	1.7e+1
	–	5.4e-1	5.1e-1			
1000	–	1.0e+0	1.0e+0	1.6e+3	1.6e+2	3.2e+1
	–	6.2e-1	6.0e-1			
1200	–	1.0e+0	1.0e+0	2.6e+3	2.7e+2	5.3e+1
	–	6.8e-1	6.6e-1			