

# Accurate Computations of Positive Polar Factors of Graded Matrices

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# OUTLINE

- ▷ Polar Decomposition
- Existing Perturbation Bounds
- Relative Perturbation Bound for Positive Polar Factors
- Accurate Computations
- Extension to the Matrix Square Root
- Concluding remarks

# Polar Decomposition

$$\boxed{\begin{array}{c} B \\ (m \times n) \end{array}} = \boxed{\begin{array}{c} \text{orth. cols} \\ Q \\ (m \times n) \end{array}} \boxed{\begin{array}{c} \text{pos. semi-def.} \\ H \\ (n \times n) \end{array}}$$

- Polar decomposition via SVD  $B = U_{m \times n} \Sigma V^*$ :

$$Q = UV^*, \quad H = V \Sigma V^*.$$

- $H$  always unique.  $Q$  unique if  $\text{rank}(B) = n$ .
- If  $\text{rank}(B) = n$ , then  $H = (B^* B)^{1/2}$  and  $Q = B(B^* B)^{-1/2}$ .

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## Existing Perturbation Bounds – $Q$ (absolute pert.)

$B = QH$  perturbed to  $\tilde{B} = \tilde{Q}\tilde{H}$ .

- (Mathias, 1993) If  $B \in \mathbb{R}^{n \times n}$ ,  $\text{rank}(B) = n$ ,  $\|\Delta B\|_2 < \sigma_n$ , then

$$\|\|\Delta Q\|\| \leq -\frac{2 \|\|\Delta B\|\|}{\|\|\Delta B\|\|_2} \ln \left( 1 - \frac{\|\|\Delta B\|\|_2}{\sigma_n + \sigma_{n-1}} \right) \approx \frac{2}{\sigma_n + \sigma_{n-1}} \|\|\Delta B\|\|,$$

where  $\|\|\cdot\|\|_2$  sum of first two largest singular values.

- (Li, 1995) If  $B \in \mathbb{C}^{n \times n}$  and  $\text{rank}(B) = n$ , then for all unitarily invariant

$$\text{norm } \|\|\Delta Q\|\| \leq \frac{2}{\sigma_n + \tilde{\sigma}_n} \|\|\Delta B\|\|.$$

- (W. Li and W. Sun, 2002) If  $B \in \mathbb{R}^{n \times n}$ ,  $\text{rank}(B) = n$ , and

$$\|\Delta B\|_2 < \sigma_n + \tilde{\sigma}_n, \text{ then } \|\Delta Q\|_F \leq \frac{4}{\sigma_{n-1} + \sigma_n + \tilde{\sigma}_{n-1} + \tilde{\sigma}_n} \|\Delta B\|_F.$$

Bounds for  $m \times n$   $B$  ( $m > n$ ) established, too.

## Existing Perturbation Bounds – $Q$ (multiplicative pert.)

- (Li, 1995) If  $B$  and  $\tilde{B} = D_L^* B D_R$  have full column rank, then

$$\begin{aligned} \|\Delta Q\|_F &\leq \sqrt{\|I - D_L^{-1}\|_F^2 + \|D_L - I\|_F^2} \\ &\quad + \sqrt{\|I - D_R^{-1}\|_F^2 + \|D_R - I\|_F^2}. \end{aligned}$$

- (Li, 1997) (Graded Case) Let  $B = GS$  and  $\tilde{B} = \tilde{G}S$  and assume that  $G$  and  $B$  have full column rank. If  $\|\Delta G\|_2 \|G^\dagger\|_2 < 1$ , then

$$\|\Delta Q\|_F \leq \gamma \|G^\dagger\|_2 \|\Delta G\|_F, \quad \gamma = \sqrt{1 + (1 - \|G^\dagger\|_2 \|\Delta G\|_2)^{-2}}.$$

## Existing Perturbation Bounds – $H$ always well-conditioned

*Kittaneh (1986):*  $\|H - \tilde{H}\|_F \leq \sqrt{2} \|B - \tilde{B}\|_F.$

Compared to  $Q$ -factor more extensively studied in the last 20 years.

$H$ 's well-conditionedness also established by *Higham (1986), Barrlund (1989), Kenney and Laub (1991), Chatelin and Gratton (2000).*

$H$ 's well-conditionedness, however, is in the *norm-wise* sense. **Relatively?** – focus of this talk.

## Graded Case – an Example

$$B = GS = \begin{pmatrix} 6 & -2 & 14 & -5 \\ 8 & 5 & -7 & -8 \\ -2 & -11 & 2 & -3 \\ 5 & -8 & -16 & 9 \end{pmatrix} \begin{pmatrix} 10^6 \\ 10^4 \\ 10^2 \\ 1 \end{pmatrix}$$

perturbed to  $\tilde{B} = \tilde{G}S$  with  $\tilde{G} = G + \Delta G$ ,  $\Delta G = 10^{-5} * \text{randn}(4)$ .

$$H = \begin{pmatrix} 1.1358 \cdot 10^7 & 8.6928 \cdot 10^3 & -4.9320 \cdot 10^2 & -3.7828 \cdot 10^0 \\ 8.6928 \cdot 10^3 & 1.4603 \cdot 10^5 & 3.1908 \cdot 10^2 & -4.4827 \cdot 10^0 \\ -4.9320 \cdot 10^2 & 3.1908 \cdot 10^2 & 2.1691 \cdot 10^3 & -7.7287 \cdot 10^0 \\ -3.7828 \cdot 10^0 & -4.4827 \cdot 10^0 & -7.7287 \cdot 10^0 & 9.2121 \cdot 10^0 \end{pmatrix}$$

## Graded Case – an Example (cont'd)

$\|H - \tilde{H}\|_F \leq \sqrt{2}\|\Delta GS\|_F \approx 24.7393$ , no assurance to smaller entries!

On the other hand,

$$\text{(by absolute pert. bound)} \quad \|\Delta Q\|_F \leq 1.90,$$

$$\text{(by relative pert. bound)} \quad \|\Delta Q\|_F \leq 4.41 \times 10^{-6}.$$

Need a better theory for  $H$ -factor!

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# Relative Perturbation Bound for Positive Polar Factors

$B = GS$ ,  $\tilde{B} = \tilde{G}S$ . Both have full column rank.  $S$  is  $n \times n$  and nonsingular, not necessarily diagonal.

**Theorem.** If  $\|\Delta G\|_2 \|G^\dagger\|_2 < 1$  then

$$\begin{aligned} \|(\tilde{H} - H)S^{-1}\|_F &\leq \sqrt{\|(\Delta G)G^\dagger\|_F^2 + \left\|I - (I + (\Delta G)G^\dagger)^{-1}\right\|_F^2} \|G\|_2 + \|\Delta G\|_F \\ &\leq \left( \sqrt{1 + \frac{1}{(1 - \|G^\dagger\|_2 \|\Delta G\|_2)^2}} \|G^\dagger\|_2 \|G\|_2 + 1 \right) \|\Delta G\|_F. \end{aligned}$$

$$\begin{aligned} \text{Example (revisited)} \quad \|(\tilde{H} - H)S^{-1}\|_F &\leq 1.4298 \cdot 10^{-4}, \\ &\leq 2.0756 \cdot 10^{-4}. \end{aligned}$$

And ...

# Relative Perturbation Bound – Analysis for Diagonal $S$

Entry-wise  $\|(\tilde{H} - H)S^{-1}\|_F \leq \varepsilon$  gives

$$|h_{ij} - \tilde{h}_{ij}| = |h_{ji} - \tilde{h}_{ji}| \leq \varepsilon \min\{|s_i|, |s_j|\}.$$

Numbers of correct significant decimal digits  $-\log_{10}(|h_{ij} - \tilde{h}_{ij}|/|h_{ij}|)$  are at least

$$-\log_{10}(\varepsilon \min\{|s_i|, |s_j|\}/|h_{ij}|).$$

Use  $\varepsilon = 1.4298 \cdot 10^{-4}$ . Numbers of correct significant decimal digits are at least, entry-wise,

$$\begin{pmatrix} 4.9 & 3.8 & 4.5 & 4.4 \\ 3.8 & 5.0 & 4.3 & 4.5 \\ 4.5 & 4.3 & 5.2 & 4.7 \\ 4.4 & 4.5 & 4.7 & 4.8 \end{pmatrix},$$

smaller entries contaminated  
nor more than larger ones.

## Relative Perturbation Bound – Proof

$$\begin{aligned}\tilde{Q}\tilde{H} &= \tilde{G}S, \\ \tilde{H}S^{-1} &= \tilde{Q}^*\tilde{G} \\ &= \tilde{Q}^*(G + \Delta G) \\ &= (Q + \Delta Q)^*G + \tilde{Q}^*\Delta G \\ &= Q^*G + \Delta Q^*G + \tilde{Q}^*\Delta G \\ &= HS^{-1} + \Delta Q^*G + \tilde{Q}^*\Delta G.\end{aligned}$$

Thus  $(\tilde{H} - H)S^{-1} = \Delta Q^*G + \tilde{Q}^*\Delta G$ , and

$$\|(\tilde{H} - H)S^{-1}\|_F \leq \|\Delta Q^*\|_F \|G\|_2 + \|\Delta G\|_F.$$

Use relative perturbation bound for  $\|\Delta Q^*\|_F$  to complete proof.

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# Accurate Computations When $S$ Diagonal

## Algorithm.

1. Compute SVD:  $B \equiv GS = U\Sigma V^*$  by one-sided Jacobi;
2. Set  $Q = UV^*$ ,  $W = Q^*G$ , and then  $H = WS$ .

*Note:*  $H$  by  $V\Sigma V^*$  does not work!

Use the following floating point arithmetic model

computed  $\alpha \odot \beta$  is  $(\alpha \odot \beta)(1 + \delta)$ ,  $|\delta| \leq \epsilon_m$  for  $\odot \in \{+, -, \times, \div\}$ ,

we have

**Theorem.** Computed  $\tilde{H}$  satisfies  $\|(\tilde{H} - H)S^{-1}\|_F = \mathcal{O}(\epsilon_m \kappa(G) \|G\|_F)$ .

# Accurate Computations – Outline of Proof

Denote computed matrices by the same symbols but with *tildes*.

1. Compute SVD:  $B \equiv GS = U\Sigma V^*$  by one-sided Jacobi;

*By Demmel, Gu, Eisenstat, Slapničar, Veselić, and Drmač (1999),*

$$\tilde{U}\tilde{\Sigma}\tilde{V}^* \equiv \tilde{B} = (I+E)B(I+F), \quad \|E\|_F = \mathcal{O}(\epsilon_m \kappa(G)), \quad \|F\|_F = \mathcal{O}(\epsilon_m \kappa(G)).$$

2. Set  $Q = UV^*$ ,  $W = Q^*G$ , and then  $H = WS$ .

*Let  $\circ$  denote the entry-wise Hadamard product. Then*

$$\tilde{Q} = \tilde{U}\tilde{V}^* + E_1, \quad \tilde{W} = \tilde{Q}^*G + E_2, \quad \tilde{H} = (\tilde{W}S) \circ M,$$

$$\|E_1\|_F = \mathcal{O}(\epsilon_m), \quad \|E_2\|_F = \mathcal{O}(\epsilon_m \|G\|_F), \quad M_{ij} = 1 + \mathcal{O}(\epsilon_m)$$

## Accurate Computations – Outline of Proof (cont'd)

*Note*  $\tilde{U}\tilde{V}^* = Q + E_3$  with  $\|E_3\|_F = \mathcal{O}(\|E\|_F + \|F\|_F) = \mathcal{O}(\epsilon_m \kappa(G))$

$$\begin{aligned}
 \tilde{H}S^{-1} &= \tilde{W} \circ M \\
 &= (\tilde{Q}^*G + E_2) \circ M \\
 &= [(\tilde{U}\tilde{V}^* + E_1)^*G + E_2] \circ M \\
 &= [(\tilde{U}\tilde{V}^*)^*G] \circ M + (E_1^*G + E_2) \circ M \\
 &= (Q^*G) \circ M + (E_3^*G + E_1^*G + E_2) \circ M \\
 &= Q^*G + (Q^*G) \circ E_4 + (E_3^*G + E_1^*G + E_2) \circ M \\
 &= HS^{-1} + (Q^*G) \circ E_4 + (E_3^*G + E_1^*G + E_2) \circ M,
 \end{aligned}$$

*which gives*

$$\|(\tilde{H} - H)S^{-1}\|_F \leq \|(Q^*G) \circ E_4 + (E_3^*G + E_1^*G + E_2) \circ M\|_F = \mathcal{O}(\epsilon_m \kappa(G) \|G\|_F).$$

## Accurate Computations – an Example

$$n = 10, B = GS, S = \text{diag}(10^3, 10^8, 10^5, 10^4, 1, 10^4, 10^9, 10^8, 10^3, 10^8),$$

$$G = \begin{pmatrix} 4.656 & 7.220 & 3.831 & 2.924 & 7.556 & 7.329 & 4.105 & 2.827 & 6.787 & 7.7860 \\ 4.187 & 2.644 & 5.986 & 9.774 & 7.660 & 7.170 & 1.042 & 2.240 & 5.235 & 6.1470 \\ 7.518 & 6.780 & 5.691 & 1.071 & 8.527 & 2.231 & 0.220 & 7.386 & 9.997 & 4.8770 \\ 3.829 & 7.801 & 2.164 & 1.158 & 1.907 & 4.992 & 7.945 & 9.693 & 7.529 & 2.1330 \\ 3.515 & 1.907 & 9.999 & 0.686 & 2.950 & 2.240 & 5.921 & 9.097 & 5.180 & 3.4390 \\ 0.871 & 4.511 & 4.406 & 3.102 & 4.731 & 1.793 & 3.100 & 9.608 & 6.400 & 6.9290 \\ 8.565 & 8.088 & 9.318 & 3.557 & 6.034 & 9.012 & 1.087 & 8.167 & 1.351 & 4.6020 \\ 1.351 & 5.289 & 3.025 & 3.591 & 1.210 & 2.123 & 8.771 & 4.650 & 8.002 & 9.6420 \\ 2.408 & 2.745 & 3.893 & 4.202 & 5.845 & 3.501 & 0.603 & 2.775 & 0.912 & 9.1680 \\ 0.231 & 2.726 & 6.898 & 9.243 & 3.813 & 5.065 & 0.594 & 8.415 & 5.140 & 1.9450 \end{pmatrix},$$

$$\|B\|_F = 3.44 \cdot 10^{10}, \kappa(B) = 5.37 \cdot 10^{10}, \kappa(G) = 4.60 \cdot 10^2, \|G\|_F = 5.02 \cdot 10^1.$$

## Accurate Computations – an Example (cont'd)

**Jacobi**: one-sided Jacobi SVD code, courtesy of Z. Drmač.

**QR**: `sgesvd` from LAPACK.

**Exact**  $Q$  and  $H$  by Maple with a 50 decimal digit arithmetic.

	$\ (\tilde{H} - H)S^{-1}\ _F$	$\ \tilde{H} - H\ _F$	$\ \tilde{Q} - Q\ _F$	$\max  \tilde{\sigma} - \sigma /\sigma$
<b>Jacobi</b>	1.19e-5	8.83e+2	1.75e-6	4.55e-6
<b>QR</b>	8.67e+1	2.47e+3	2.00e+0	2.19e+1

In both cases, norm-wise relative error  $\frac{\|\tilde{H} - H\|_F}{\|H\|_F} \approx 10^{-7}$ , good enough unless interested in knowing smaller entries.

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## Extension to the Matrix Square Root

$A = S^*TS$  perturbed to  $\tilde{A} = S^*\tilde{T}S$ ,  $S$  ( $n \times n$ ) nonsingular,  $T$  positive definite and well-conditioned.

**Theorem.** Let  $\hat{T} \stackrel{\text{def}}{=} T^{-1/2}(\tilde{T} - T)T^{-1/2}$  and  $\delta_p = \|\hat{T}\|_p$  ( $p = 2, \text{F}$ ). If  $\delta_2 < 1$ , then

$$\begin{aligned} \|(\tilde{A}^{1/2} - A^{1/2})S^{-1}\|_{\text{F}} &\leq \left( \sqrt{\frac{2 - \delta_2}{1 - \delta_2}} + 1 \right) \|T^{1/2}\|_2 \frac{\delta_{\text{F}}}{1 + \sqrt{1 - \delta_2}}, \\ &\approx \frac{\sqrt{2} + 1}{2} \|T^{1/2}\|_2 \delta_{\text{F}}. \end{aligned}$$

**Existing bound.**

$$\|\tilde{A}^{1/2} - A^{1/2}\|_{\text{F}} \leq \frac{1}{\|\tilde{A}^{-1/2}\|_2^{-1} + \|A^{-1/2}\|_2^{-1}} \|\tilde{A} - A\|_{\text{F}}.$$

# Extension to the Matrix Square Root – an Example

$A = B^*B = S^*TS$  with

$$B = GS = \begin{pmatrix} 6 & -2 & 14 & -5 \\ 8 & 5 & -7 & -8 \\ -2 & -11 & 2 & -3 \\ 5 & -8 & -16 & 9 \end{pmatrix} \begin{pmatrix} 10^6 & & & \\ & 10^4 & & \\ & & 10^2 & \\ & & & 1 \end{pmatrix}.$$

Take  $E = 10^{-5} * \text{randn}(4)$  and  $\Delta T = E + E^* + EE^*$ . Then

**Existing bound**  $\|\tilde{A}^{1/2} - A^{1/2}\|_F \leq 1.4288 \cdot 10^6,$

**New bound**  $\|(\tilde{A}^{1/2} - A^{1/2})S^{-1}\|_F \leq 1.17363 \cdot 10^{-5}.$

## Extension to the Matrix Square Root – Outline of Proof

$$A = S^* T^{1/2} T^{1/2} S = (GS)^* (GS) \equiv B^* B,$$

$$\tilde{A} = S^* (T + \Delta T) S = S^* T^{1/2} \left[ I + T^{-1/2} (\Delta T) T^{-1/2} \right] T^{1/2} S = \tilde{B}^* \tilde{B}.$$

$$\tilde{B} = \tilde{G} S \text{ with } \tilde{G} = \left( I + \hat{T} \right)^{1/2} T^{1/2} \text{ and } \hat{T} = T^{-1/2} (\Delta T) T^{-1/2}.$$

$$\text{Polar decompositions } B = QH, \tilde{B} = \tilde{Q}\tilde{H} \implies A^{1/2} = H, \tilde{A}^{1/2} = \tilde{H}.$$

Previous relative perturbation bound for  $H$ -factor applies:

$$\begin{aligned} \Delta G &= \tilde{G} - G = \left[ \left( I + \hat{T} \right)^{1/2} - I \right] G, \\ (\Delta G)G^{-1} &= \left( I + \hat{T} \right)^{1/2} - I, \\ I - \left( I + (\Delta G)G^{-1} \right)^{-1} &= I - \left( I + \hat{T} \right)^{-1/2}. \end{aligned}$$

# Extension to the Matrix Square Root – Accurate Computation

**Algorithm** (for diagonal  $S$ ).

1. Decompose  $T = G^*G$  (e.g., Cholesky decomposition);
2. Compute polar decomposition  $GS = QH$  by the previous algorithm, and return  $H \approx A^{1/2}$ .

**Theorem.** Computed  $\tilde{H} \approx A^{1/2}$  satisfies  
 $\|(\tilde{H} - A^{1/2})S^{-1}\|_F = \mathcal{O}(\epsilon_m \kappa(T) \|T^{1/2}\|_F)$ .

**Outline of Proof.** Computed  $\tilde{G}$  satisfies  $\|\Delta T\|_F = \mathcal{O}(\epsilon_m \|T\|_F)$  (*J. Sun 1992*). Let  $\hat{A} = S^* \tilde{G}^* \tilde{G} S$ .

$$\|(\hat{A}^{1/2} - A^{1/2})S^{-1}\|_F = \mathcal{O}(\epsilon_m \|T^{1/2}\|_2 \|T^{-1}\|_2 \|T\|_F) \approx \mathcal{O}(\epsilon_m \kappa(T) \|T^{1/2}\|_2),$$

$$\|(\tilde{H} - \hat{A}^{1/2})S^{-1}\|_F = \mathcal{O}(\epsilon_m \kappa(\tilde{G}) \|\tilde{G}\|_F) \approx \mathcal{O}(\epsilon_m \kappa(T^{1/2}) \|T^{1/2}\|_F).$$

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## Concluding remarks

- New perturbation bound measuring the scaled difference in positive polar factors
- In highly graded cases, smaller entries in  $H$  contaminated no more than larger ones
- Can compute  $H$  as accurately as predicted
- Extensible to matrix square roots.