

# Majorization-based convergence rate bounds for subspace iterations and the block Lanczos method

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## Outline

- The Rayleigh–Ritz method, principal angles between subspaces  $\mathcal{X}$  and  $\mathcal{Y}$ , majorization
- Weak majorization bounds on the change in Ritz values
- Bound on principal angles between  $\mathcal{X}$  and  $F\mathcal{Y}$  in terms of the principal angles between  $\mathcal{X}$  and  $\mathcal{Y}$  where  $F$  is an operator
- Subspace iteration and weak majorization bounds on Ritz values
- Convergence rate bounds for the block Lanczos method
- Conclusions

## The Rayleigh–Ritz Method

- Let  $A$  be a Hermitian matrix and  $\mathcal{X}$  a subspace
- We define an operator  $P_{\mathcal{X}}A|_{\mathcal{X}}$  on  $\mathcal{X}$ , where  $P_{\mathcal{X}}$  is the orthogonal projection onto  $\mathcal{X}$  and  $P_{\mathcal{X}}A|_{\mathcal{X}}$  denotes the restriction of  $P_{\mathcal{X}}A$  to  $\mathcal{X}$ , as discussed in Parlett [1998]. The eigenvalues  $\Lambda(P_{\mathcal{X}}A|_{\mathcal{X}})$  are called Ritz values.
- We have  $\Lambda(P_{\mathcal{X}}A|_{\mathcal{X}}) = \Lambda(X^HAX)$  where  $X$  is a matrix with orthonormal columns that span  $\mathcal{X}$
- Nonzero Ritz values are the nonzero eigenvalues of  $P_{\mathcal{X}}AP_{\mathcal{X}}$

## Bounding Eigenvalues

If  $A$  is a Hermitian matrix, we have Weyl's theorem

$$\max_{j=1,\dots,n} |\lambda_j(A) - \lambda_j(B)| \leq \|A - B\|, \quad (1)$$

and by an analog of the Hoffman–Wielandt theorem, e.g. Stewart and Sun [1990], we have

$$\sqrt{\sum_{j=1}^n (\lambda_j(A) - \lambda_j(B))^2} \leq \|A - B\|_F \quad (2)$$

How do we bound Ritz values when we vary/perturb the subspace?

## Changes in the Trial Subspace in the Rayleigh–Ritz Method

- We can vary a Hermitian matrix and see how the eigenvalues change
- Analogously for a fixed Hermitian matrix, we can vary the subspace and see how the Ritz values change. If a subspace  $\mathcal{X}$  is perturbed to give rise to another subspace  $\mathcal{Y}$ , then we can ask about how to bound

$$|\Lambda(P_{\mathcal{X}}A|_{\mathcal{X}}) - \Lambda(P_{\mathcal{Y}}A|_{\mathcal{Y}})| ?$$

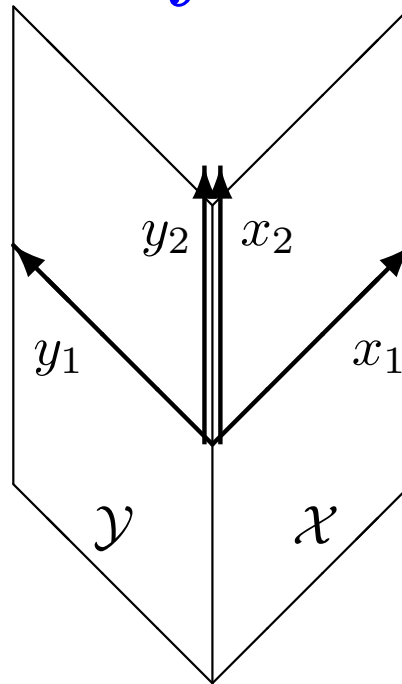
- **Answer:** We can prove some very flexible and useful bounds using principal angles between subspaces and majorization, also considering the case when  $\mathcal{X}$  or  $\mathcal{Y}$  may be invariant relative to  $A$ .

## Principal Angles Between Subspaces

- Let subspaces  $\mathcal{X}$  and  $\mathcal{Y} \subseteq \mathbf{C}^n$  with  $\dim \mathcal{X} = \dim \mathcal{Y}$ , and orthonormal bases given by the columns of the matrices  $X$  and  $Y$ . The principal angles between  $\mathcal{X}$  and  $\mathcal{Y}$  arranged in descending order are given by  $\Theta(\mathcal{X}, \mathcal{Y}) = \Theta^\downarrow(\mathcal{X}, \mathcal{Y})$ , and defined using  $\cos \Theta(\mathcal{X}, \mathcal{Y}) = S^\uparrow(X^H Y)$
- Definition is symmetric:  $\Theta(\mathcal{X}, \mathcal{Y}) = \Theta(\mathcal{Y}, \mathcal{X})$ .
- Definition of the distance between subspaces:  
 $\text{gap}(\mathcal{X}, \mathcal{Y}) = \|P_{\mathcal{X}} - P_{\mathcal{Y}}\|_2 = \sin(\Theta_{\max}(\mathcal{X}, \mathcal{Y}))$ .

*Pioneering results using angles between subspaces in the framework of unitarily invariant norms and symmetric gauge functions, equivalent to majorization, appear in Davis and Kahan [1970], which introduces many of the tools that we use here.*

## The Geometry of Principal Angles



$$\theta_1 \geq \dots \geq \theta_m$$

The cosines can be defined recursively for  $k = m, \dots, 1$  by

$$\cos(\theta_k) = \max_{x \in \mathcal{X}} \max_{y \in \mathcal{Y}} x^T y = x_k^T y_k$$

subject to

$$\|x\| = \|y\| = 1, \quad x^T x_i = 0, \quad y^T y_i = 0, \quad i = m, \dots, k + 1.$$

## Majorization and Basic Notation

- For vector  $x = [x_1, \dots, x_n]$ , we use  $x^\downarrow \equiv [x_1^\downarrow, \dots, x_n^\downarrow]$  to denote  $x$  with its elements rearranged in descending order, while  $x^\uparrow \equiv [x_1^\uparrow, \dots, x_n^\uparrow]$  denotes  $x$  with its elements rearranged in ascending order.  $|x|$  denotes the vector  $x$  with the absolute value of its components.
- We say that  $x \in \mathcal{R}^n$  is **weakly majorized** by  $y \in \mathcal{R}^n$ , written  $x \prec_w y$ , if

$$\sum_{i=1}^k x_i^\downarrow \leq \sum_{i=1}^k y_i^\downarrow, \quad 1 \leq k \leq n, \quad (3)$$

while  $x$  is (strongly) **majorized** by  $y$ , written  $x \prec y$ , if (3) holds together with

$$\sum_{i=1}^n x_i = \sum_{i=1}^n y_i. \quad (4)$$

## Why is Majorization Important?

- The notion of majorization plays an important role in mathematics, statistics and economics
- The following two conditions are equivalent:  $x \prec_w y$  and  $\sum_{i=1}^n \phi(x_i) \leq \sum_{i=1}^n \phi(y_i)$  for all nondecreasing convex functions  $\phi$ . This implies inequalities for unitarily invariant norms and certain  $p$ -norms
- For Hermitian matrices  $A$  and  $B$  let  $\Lambda(A)$  and  $S(A)$  be, respectively, the vector of eigenvalues and singular values (both nonincreasing), then

$$\text{diag}(A) \prec \Lambda(A), \quad (\text{Schur's Theorem}) \quad (5)$$

$$\Lambda(A) - \Lambda(B) \prec \Lambda(A - B) \quad \text{and} \quad |\Lambda(A) - \Lambda(B)| \prec_w S(A - B) \quad (6)$$

See Lidskii [1950], Mirsky [1960], Marshall and Olkin [1979], Horn and Johnson [1999].

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## Changes in the Trial Subspace in the Rayleigh–Ritz Method

The result of (Argentati [2003], Knyazev and Argentati [2006a]):

**Theorem 1** *Let  $A$  be a Hermitian matrix and let  $\mathcal{X}$  and  $\mathcal{Y}$  be subspaces with  $\dim\mathcal{X} = \dim\mathcal{Y} = m$ . Then*

$$\max_{j=1,\dots,m} |\lambda_j(P_{\mathcal{X}}A|_{\mathcal{X}}) - \lambda_j(P_{\mathcal{Y}}A|_{\mathcal{Y}})| \leq (\lambda_{\max} - \lambda_{\min}) \sin(\Theta_{\max}(\mathcal{X}, \mathcal{Y}))$$

where  $\lambda_{\min}$  and  $\lambda_{\max}$  are the smallest and largest eigenvalues of  $A$ .

## Main Weak Majorization Theorem for the Rayleigh–Ritz Method

**Theorem 2** *Let  $A$  be a Hermitian matrix and let  $\mathcal{X}$  and  $\mathcal{Y}$  be subspaces with  $\dim\mathcal{X} = \dim\mathcal{Y}$ , then (see Knyazev and Argentati [2006b])*

$$|\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}})| \prec_w (\lambda_{\max} - \lambda_{\min}) \sin \Theta(\mathcal{X}, \mathcal{Y}), \quad (7)$$

*where  $\lambda_{\min}$  and  $\lambda_{\max}$  are the smallest and largest eigenvalues of the  $A$ , respectively. If in addition one of the subspaces is  $A$ -invariant then (see conjecture in Argentati, Knyazev, Paige, and Panayotov [2008])*

$$|\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}})| \prec_w (\lambda_{\max} - \lambda_{\min}) \sin^2 \Theta(\mathcal{X}, \mathcal{Y}). \quad (8)$$

## Theorem 2 is Sharp

Consider the following example for (8) where  $\mathcal{X}$  is  $A$ -invariant (see Argentati, Knyazev, Paige, and Panayotov [2008]). Let  $I$  be an  $m \times m$  identity matrix and let  $A = \begin{bmatrix} I & 0 \\ 0 & -I \end{bmatrix}$ . Let  $\Sigma$  be a diagonal of given

cosines and let  $\Gamma$  be the diagonal of sines, so  $\Sigma^2 + \Gamma^2 = I$ . Let  $X = \begin{bmatrix} I \\ 0 \end{bmatrix}$ ,

and  $Y = \begin{bmatrix} \Sigma \\ \Gamma \end{bmatrix}$ . Then  $X^H AX = I$  and  $Y^H AY = 2\Sigma^2 - I$  and

$$|\Lambda(X^H AX) - \Lambda(Y^H AY)|^\downarrow = 2[\sin^2 \Theta(\mathcal{X}, \mathcal{Y})] = (\lambda_{\max} - \lambda_{\min})[\sin^2 \Theta(\mathcal{X}, \mathcal{Y})].$$

## Weak Majorization Implications of

$$|\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}})| \prec_w (\lambda_{\max} - \lambda_{\min}) \sin^2 \Theta(\mathcal{X}, \mathcal{Y})$$

The implications of weak majorization may not be obvious. Let  $\dim \mathcal{X} = \dim \mathcal{Y} = m$ , and let  $\alpha_1 \geq \dots \geq \alpha_m$  be the Ritz values of  $A$  with respect to  $\mathcal{X}$  and  $\beta_1 \geq \dots \geq \beta_m$  be the Ritz values of  $A$  with respect to  $\mathcal{Y}$ . Then using (8) we have

$$\sum_{i=1}^k |\alpha_i - \beta_i|^{\downarrow} \leq (\lambda_{\max} - \lambda_{\min}) \sum_{i=1}^k \sin^2(\Theta_i(\mathcal{X}, \mathcal{Y})), \quad k = 1, \dots, m.$$

For  $k = 1$  we have

$$\begin{aligned} \max_{j=1, \dots, m} |\alpha_j - \beta_j| &\leq (\lambda_{\max} - \lambda_{\min}) \sin^2(\Theta_{\max}(\mathcal{X}, \mathcal{Y})) \\ &= (\lambda_{\max} - \lambda_{\min}) \text{gap}^2(\mathcal{X}, \mathcal{Y}), \end{aligned}$$

## Weak Majorization Implications (Cont.)

$$|\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}})| \prec_w (\lambda_{\max} - \lambda_{\min}) \sin^2 \Theta(\mathcal{X}, \mathcal{Y}),$$

and for  $k = m$  we obtain

$$\sum_{i=1}^m |\alpha_i - \beta_i| \leq (\lambda_{\max} - \lambda_{\min}) \sum_{i=1}^m \sin^2(\Theta_i(\mathcal{X}, \mathcal{Y}))$$

We have in general

$$\left( \sum_{i=1}^m |\alpha_i - \beta_i|^p \right)^{\frac{1}{p}} \leq (\lambda_{\max} - \lambda_{\min}) \left( \sum_{i=1}^m \sin^{2p}(\Theta_i(\mathcal{X}, \mathcal{Y})) \right)^{\frac{1}{p}}, \quad 1 \leq p < \infty.$$

## Bounding Eigenvalues (Cont.)

If  $A$  is a Hermitian matrix, we have Weyl's theorem

$$\max_{j=1,\dots,n} |\lambda_j(A) - \lambda_j(B)| \leq \|A - B\|, \quad (9)$$

and by an analog of the Hoffman–Wielandt theorem, e.g. Stewart and Sun [1990], we have

$$\sqrt{\sum_{j=1}^n (\lambda_j(A) - \lambda_j(B))^2} \leq \|A - B\|_F \quad (10)$$

Note that the above formulas are two of a large class of inequalities that follow from

$$|\Lambda(A) - \Lambda(B)| \prec_w S(A - B)$$

## Matlab Example for Many Zero Angles

Let  $n = 5$  and  $\dim\mathcal{X} = \dim\mathcal{Y} = 4$

$$A = \text{diag}([1 \ .9 \ .7 \ .5 \ 0]) \quad X = \text{eye}(n, m) \quad Y = \text{orth}(\text{rand}(n, m)).$$

Then  $\mathcal{X}$  is  $A$ -invariant and

$$\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) = [1.0000 \ 0.9000 \ 0.7000 \ 0.5000],$$

$$\Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}}) = [0.9452 \ 0.7484 \ 0.6572 \ 0.0330].$$

Then

$$|\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}})|^{\downarrow} = [0.4670 \ 0.1516 \ 0.0548 \ 0.0428]$$

$$\prec_w (\lambda_{\max} - \lambda_{\min}) \sin^2 \Theta(\mathcal{X}, \mathcal{Y}) = [0.9513 \ 0 \ 0 \ 0].$$

**Conclusion:** Zero angles do not imply that  $\Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}})$  includes eigenvalues of  $A$ .

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## Subspace Iterations: $F\mathcal{Y}$ as a Possibly Improved Approximation of $\mathcal{X}$

**Theorem 3** *Let  $\dim\mathcal{X} = \dim\mathcal{Y}$  and assume  $\Theta(\mathcal{X}, \mathcal{Y}) < \pi/2$ . Let  $F$  be invariant on both  $\mathcal{X}$  and on its orthogonal complement  $\mathcal{X}^\perp$ , and assume that  $(P_{\mathcal{X}}F)|_{\mathcal{X}}$  is invertible. Then  $\dim(F\mathcal{Y}) = \dim\mathcal{Y}$  and*

$$\tan \Theta(\mathcal{X}, F\mathcal{Y}) \prec_w S \left( ((P_{\mathcal{X}}F)|_{\mathcal{X}})^{-1} \right) S \left( (P_{\mathcal{X}^\perp}F)|_{\mathcal{X}^\perp} \right) \tan \Theta(\mathcal{X}, \mathcal{Y}).$$

We can also prove the stronger form

$$\log \left( \frac{\tan \Theta(\mathcal{X}, F\mathcal{Y})}{\tan \Theta(\mathcal{X}, \mathcal{Y})} \right) \prec_w \log \left( S \left( ((P_{\mathcal{X}}F)|_{\mathcal{X}})^{-1} \right) S \left( (P_{\mathcal{X}^\perp}F)|_{\mathcal{X}^\perp} \right) \right).$$

## Subspace Iterations and Ritz Values

**Theorem 4** *Let  $\mathcal{X}$  and  $\mathcal{Y}$  be subspaces of  $\mathcal{H}$  such that  $\dim\mathcal{X} = \dim\mathcal{Y}$  and  $\Theta(\mathcal{X}, \mathcal{Y}) < \pi/2$ . Let the operator  $A$  be Hermitian, and let  $\mathcal{X}$  be an  $A$ -invariant subspace. Let  $F = f(A)$  and  $f(\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}})) \neq 0$ . Then  $\dim(F\mathcal{Y}) = \dim\mathcal{Y}$  and*

$$|\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda((P_{F\mathcal{Y}}A)|_{F\mathcal{Y}})| \\ \prec_w (\lambda_{\max} - \lambda_{\min}) \left( \frac{|f(\Lambda((P_{\mathcal{X}^\perp}A)|_{\mathcal{X}^\perp}))|^\downarrow}{|f(\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}))|^\uparrow} \right)^2 \tan^2 \Theta(\mathcal{X}, \mathcal{Y}).$$

## Subspace Iterations and Ritz Values

**Example:**  $f(A) = A^k$

Let the  $A$ -invariant subspace  $X$  correspond to the contiguous set of the largest eigenvalues of  $A$ . Let

$$\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) = [\lambda_1 \geq \dots \geq \lambda_m] \quad \text{top } m \text{ eigenvalues of } A$$

$$\Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}}) = [\tilde{\lambda}_1 \geq \dots \geq \tilde{\lambda}_m].$$

Then  $\lambda_i \geq \tilde{\lambda}_i$ ,  $i = 1, \dots, m$  and

$$\begin{aligned} & [\lambda_1 - \tilde{\lambda}_1, \dots, \lambda_m - \tilde{\lambda}_m] \\ \prec_w (\lambda_{\max} - \lambda_{\min}) & \left[ \left( \frac{\lambda_{m+1}}{\lambda_m} \right)^{2k} \tan^2 \theta_1(\mathcal{X}, \mathcal{Y}), \dots, \left( \frac{\lambda_{2m}}{\lambda_1} \right)^{2k} \tan^2 \theta_m(\mathcal{X}, \mathcal{Y}) \right] \end{aligned}$$

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## Convergence Rate Bounds of the Block Lanczos Method

**Theorem 5** *Let  $\dim \mathcal{X} = \dim \mathcal{X}_0 = m$ , operator  $A$  be Hermitian, and let the  $A$ -invariant subspace  $X$  correspond to the contiguous set of the largest eigenvalues of  $A$ . Let  $\mathcal{Y} = \mathcal{X}_0 + A\mathcal{X}_0 + \cdots + A^k \mathcal{X}_0$ , then*

$$0 \leq \frac{\Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) - \Lambda_{\dim \mathcal{X}}((P_{\mathcal{Y}}A)|_{\mathcal{Y}})}{\Lambda_{\dim \mathcal{X}}((P_{\mathcal{Y}}A)|_{\mathcal{Y}}) - \lambda_{\min}} \prec_w [\sigma_m^*, \dots, \sigma_1^*] \tan^2 \Theta(\mathcal{X}, \mathcal{X}_0).$$

where

$$\sigma_i^* = \left| T_k \left( \frac{\lambda_{m+1} + \lambda_{\min} - 2\lambda_i}{\lambda_{m+1} - \lambda_{\min}} \right) \right|^{-2} \quad i = 1, \dots, m,$$

and where  $T_k$  is the  $k$ th Chebyshev polynomial of the first kind.

See, e.g., Saad [1980], Knyazev [1987].

## Convergence Rate Bounds of the Block Lanczos Method (Cont.)

Here is a more concrete interpretation of Theorem 5. Let

$$\begin{aligned} \Lambda((P_{\mathcal{X}}A)|_{\mathcal{X}}) &= [\lambda_1 \geq \dots \geq \lambda_m] \quad \text{top } m \text{ eigenvalues of } A \\ \Lambda_{\dim \mathcal{X}}((P_{\mathcal{Y}}A)|_{\mathcal{Y}}) &= [\tilde{\lambda}_1 \geq \dots \geq \tilde{\lambda}_m] \quad \text{top } m \text{ Ritz values of } \Lambda((P_{\mathcal{Y}}A)|_{\mathcal{Y}}) \end{aligned}$$

Then

$$\begin{aligned} &\left[ \left( \frac{\lambda_1 - \tilde{\lambda}_1}{\tilde{\lambda}_1 - \lambda_{\min}} \right), \dots, \left( \frac{\lambda_m - \tilde{\lambda}_m}{\tilde{\lambda}_m - \lambda_{\min}} \right) \right] \\ &\prec_w [\sigma_m^* \tan^2 \theta_1(\mathcal{X}, \mathcal{X}_0), \dots, \sigma_1^* \tan^2 \theta_m(\mathcal{X}, \mathcal{X}_0)] \end{aligned}$$

## Conclusions

- We can prove some very flexible and useful bounds for Ritz values using principal angles and majorization
- The absolute value of the difference of Ritz values for a Hermitian matrix are majorized by  $\lambda_{\max} - \lambda_{\min}$  times the sines of the angles between the perturbed subspaces, with majorization by the squares of the sines (see conjecture in [2008]) if one of the subspaces is  $A$ -invariant
- We can characterize a possible improved approximation to  $X$  by applying a function  $F$  to  $Y$ , in terms of singular values and the tangent of principal angles
- Using this general approach we obtain a useful characterization of the eigenvalue error for the Block Lanczos Method

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# Backup Slides

## We Can Replace $\lambda_{\max} - \lambda_{\min}$ With a Reduced Constant

- As in Knyazev and Argentati [2006a, Remark 7], the constant  $\lambda_{\max} - \lambda_{\min}$  can be replaced with

$$\max_{x \in \mathcal{X} + \mathcal{Y}, \|x\|=1} (x, Ax) - \min_{x \in \mathcal{X} + \mathcal{Y}, \|x\|=1} (x, Ax),$$

which for some subspaces  $\mathcal{X}$  and  $\mathcal{Y}$  can provide a significant improvement.

- This effectively replaces  $A$  with  $P_{(\mathcal{X} + \mathcal{Y})} A P_{(\mathcal{X} + \mathcal{Y})}$ , which has the same action on  $\mathcal{X}$  and  $\mathcal{Y}$ .