

Vector estimates for the action of matrix functions on vectors

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Outline of the talk

1. Introduction
2. Vector estimates for $f(A)b$
3. Estimates for matrix functionals
4. Numerical examples
5. Concluding remarks

Introduction

Given:

- * A : a **diagonalizable** matrix of order p
- * b : a vector of order p and
- * f : an analytic function defined on the spectrum of the matrix A

Task:

- * Approximation of the **action** of $f(A)$ **on a vector** b , i.e. $f(A)b$, without computing $f(A)$.

Applications

It is not necessary either to estimate the whole matrix $f(A)$ or it is not feasible to compute $f(A)$.

- * lattice quantum chromodynamics computations in chemistry and physics
- * numerical solution of stochastic differential equations
- * sampling from a Gaussian process distribution

Ref:

- J. Chen, M. Anitescu, Y. Saad, Computing $f(A)b$ via least squares polynomial approximations, SIAM, 33 (2011), 195-222.
- P. I. Davies, N. J. Higham, Computing $f(A)b$ for Matrix Functions f , Vol. 47 (2005) of the series Lecture Notes in Computational Science and Engineering, Springer-Verlag, Berlin, 15-24.

Vector estimates for $f(A)b$

- * **Eigendecomposition** of A :

$$A = Q \Lambda Q^{-1}$$

where

- **Λ** : $p \times p$ diagonal matrix which contains the eigenvalues λ_i of A and
 - **Q** : $p \times p$ matrix which contains the corresponding linearly independent eigenvectors of A .
- * We can write:

$$f(A) = Q f(\Lambda) Q^{-1} = \sum_{i=1}^p f(\lambda_i) q_i \hat{q}_i^T$$

- * We define the **vector-moments**:

$$d_r = A^r b \quad \text{and} \quad d_f = f(A)b.$$

- * We can write:

$$d_r = A^r b = \sum_{i=1}^p \lambda_i^r q_i(\hat{q}_i^T, b) = \sum_{i=1}^p \lambda_i^r a_i q_i$$

and

$$d_f = f(A) b = \sum_{i=1}^p f(\lambda_i) q_i(\hat{q}_i^T, b) = \sum_{i=1}^p f(\lambda_i) a_i q_i$$

where:

$$\alpha_i = (\hat{q}_i^T, b).$$

One – term vector estimates

$$\begin{aligned}d_0 &= A^0 b \approx \alpha q \\d_1 &= A^1 b \approx l \alpha q = l d_0 \\d_2 &= A^2 b \approx l^2 \alpha q = l^2 d_0\end{aligned}$$

Family of one-term vector estimates for $\mathbf{f(A)b}$:

$$\varphi_N(i) = f(d_0(i)^{N(i)-1} d_1(i)^{1-2N(i)} d_2(i)^{N(i)}) d_0(i), \quad N(i) \in \mathbb{C},$$

$$i = 1, 2, \dots, p.$$

Proposition

The family of one-term vector estimates satisfies the relation

$$\varphi_N(\mathbf{i}) = f \left(\rho(\mathbf{i})^{N(\mathbf{i})} \frac{d_1(\mathbf{i})}{d_0(\mathbf{i})} \right) d_0(\mathbf{i}), \quad \mathbf{i} = 1, 2, \dots, p,$$

where $\rho(\mathbf{i}) = \frac{d_0(\mathbf{i})d_2(\mathbf{i})}{d_1(\mathbf{i})^2}$.

Lemma

Let $A \in \mathbb{R}^{p \times p}$ be a diagonalizable matrix and f an invertible function. There exists a vector $N_{\text{opt}} \in \mathbb{C}^p$ which the i -th element is given by

$$N_{\text{opt}}(i) = \frac{\log(f^{-1}(\frac{d_f(i)}{d_0(i)} \frac{d_0(i)}{d_1(i)}))}{\log(\rho(i))}, \quad \rho(i) = \frac{d_0(i)d_2(i)}{d_1(i)^2} \neq 1, \quad i = 1, 2, \dots, p,$$

such that $\varphi_{N_{\text{opt}}}$ gives the exact value of $f(A)b$.

Lemma

Let $A \in \mathbb{R}^{p \times p}$ be a diagonalizable matrix and f an increasing function. If $d_0(i) > 0$, $d_1(i) > 0$ and $\rho(i) > 1$ then a bound for the optimal value $N_{\text{opt}}(i)$ is

$$N_{\text{opt}}(i) \leq \frac{\log(f^{-1}(k(Q)f(\rho(A))\frac{\|b\|}{d_0(i)}\frac{d_0(i)}{d_1(i)}))}{\log(\rho(i))}, \quad i=1, 2, \dots, p,$$

where $k(Q)$ is the condition number of the matrix of the eigenvectors Q and $\rho(A)$ is the spectral radius of A .

Two – term vector estimates

$$\begin{aligned}d_0 &\approx \alpha_1 q_1 + \alpha_2 q_2 \\d_1 &\approx l_1 \alpha_1 q_1 + l_2 \alpha_2 q_2 \\d_2 &\approx l_1^2 \alpha_1 q_1 + l_2^2 \alpha_2 q_2\end{aligned}$$

The family of two-term vector estimates for $\mathbf{f}(\mathbf{A})\mathbf{b}$ satisfies the relation:

$$\hat{\varphi}_{n,k}(\mathbf{i}) = f(l_1(\mathbf{i})) \alpha_1 q_1(\mathbf{i}) + f(l_2(\mathbf{i})) \alpha_2 q_2(\mathbf{i}), \quad \mathbf{i} = 1, 2, \dots, p,$$

where:

$$\alpha_1 q_1(i) = \frac{1}{l_2(i) - l_1(i)} (l_2(i) d_0(i) - d_1(i)), \quad l_1(i) \neq l_2(i),$$

$$\alpha_2 q_2(i) = \frac{1}{l_2(i) - l_1(i)} (d_1(i) - l_1(i) d_0(i)), \quad l_1(i) \neq l_2(i),$$

$$l_{1,2}(i) = \frac{r(i) \pm \sqrt{r(i)^2 - 4q(i)}}{2}$$

$$r(i) = \frac{d_{n-1}(i) d_{n+2+k}(i) - d_{n+1}(i) d_{n+k}(i)}{d_{n-1}(i) d_{n+1+k}(i) - d_n(i) d_{n+k}(i)}, \quad n, k \in \mathbb{Z}$$

$$q(i) = \frac{d_n(i) d_{n+2+k}(i) - d_{n+1}(i) d_{n+1+k}(i)}{d_{n-1}(i) d_{n+1+k}(i) - d_n(i) d_{n+k}(i)}, \quad n, k \in \mathbb{Z}.$$

Estimates for matrix functionals

- * Let A be a **diagonalizable** matrix of order p and x, y be vectors of order p .
- * Estimates for matrix functionals of the form:

$$\mathbf{x}^* \mathbf{f}(A) \mathbf{y}$$

- * We define the moments:

$$c_r(\mathbf{x}, \mathbf{y}) = (\mathbf{x}, A^r \mathbf{y}) \quad \text{and} \quad c_f(\mathbf{x}, \mathbf{y}) = (\mathbf{x}, f(A) \mathbf{y}).$$

Ref: P. Fika, M. Mitrouli, Estimation of the bilinear form $\mathbf{y}^* f(A) \mathbf{x}$ for Hermitian matrices, *Linear Algebra Appl.*, 502, pp. 140-158, 2015.

- * **Family of one-term estimates:**

$$e_v = f(c_0^{v-1} c_1^{1-2v} c_2^v) c_0, \quad v \in \mathbb{C}.$$

- * The **optimal value** of v is given by the type

$$v_{\text{opt}} = \frac{\log(f^{-1}\left(\frac{c_f}{c_0}\right)\frac{c_0}{c_1})}{\log(\rho)}, \quad \rho = \frac{c_0 c_2}{c_1^2} \neq 1.$$

* **Family of two-term estimates**

$$\hat{e}_{n,k} = f(l_1) \alpha_1 q_1 + f(l_2) \alpha_2 q_2$$

where:

$$\alpha_1 q_1 = \frac{1}{l_2 - l_1} (l_2 c_0 - c_1), \quad l_1 \neq l_2,$$

$$\alpha_2 q_2 = \frac{1}{l_2 - l_1} (c_1 - l_1 c_0), \quad l_1 \neq l_2,$$

$$l_{1,2} = \frac{r \pm \sqrt{r^2 - 4q}}{2}$$

$$r = \frac{c_{n-1} c_{n+2+k} - c_{n+1} c_{n+k}}{c_{n-1} c_{n+1+k} - c_n c_{n+k}}, \quad q = \frac{c_n c_{n+2+k} - c_{n+1} c_{n+1+k}}{c_{n-1} c_{n+1+k} - c_n c_{n+k}}, \quad n, k \in \mathbb{Z}$$

Connection between $f(A)b$ and $x^T f(A)y$

* We have $x = e_i$ and $y = b$.

* Moments:

$$c_r(x,y) = (d_r)_i \quad \text{and} \quad c_f(x,y) = (d_f)_i, \quad i = 1, 2, \dots, p.$$

* We can write $f(A)b = \begin{bmatrix} \langle e_1, f(A)b \rangle \\ \langle e_2, f(A)b \rangle \\ \vdots \\ \langle e_p, f(A)b \rangle \end{bmatrix}$.

Numerical examples

We estimate the quantity $\mathbf{f}(\mathbf{A})\mathbf{b}$, for various functions f , matrices \mathbf{A} and vectors \mathbf{b} .

Complexity: $O(sp^2)$, $s \leq 8$

Ref:

- The University of Florida Sparse Matrix Collection, <http://www.cise.ufl.edu/research/sparse/matrices/>.
- The Matlab gallery, <http://www.mathworks.com/help/matlab/ref/gallery.html>.

Example 1: *Diagonalizable matrices*

We test the matrix $A = \mathbf{dw}256\mathbf{B}$ of order $\mathbf{p} = 512$.

- diagonalizable with positive eigenvalues
- well conditioned ($k(A) = 3.7328$)

We estimate the quantity $\mathbf{A}^{1/2}\mathbf{b}$ with

- \mathbf{b} is drawn from the uniform distribution
- \mathbf{b} is drawn from the normal distribution.

(n,k)	relative error [b = rand(p,1)]	relative error [b = randn(p,1)]
(1,0)	9.9679e-4	7.7755e-4
(1,1)	1.9075e-3	1.1618e-3
(1,3)	3.6696e-3	1.3892e-3
(1,-2)	3.6671e-4	8.3626e-4
(0,4)	3.5097e-3	4.4683e-3

*Estimating $A^{1/2}\mathbf{b}$ by using the family of **two-term** estimates for various values of the parameters n and k .*

* Estimating $f(A)b$ for various A , b , f .

Two-term vector estimates with $n=1$ and $k=0$.

matrix A	vector b	function $f(A)$	relative error
ex1 (p=216)	randn(216,1)	$A^{1/2}$	3.6759e-2
ex1 (p=216)	rand(216,1)	exp(A)	9.8744e-7
parter (p=800)	$b_i = \cos(i)$	exp(A)	4.0454e-2
parter (p=800)	e_5	$A^{1/2}$	4.6376e-2
rand(300)/100	randn(300,1)	exp(A)	5.2986e-3

Estimating $f(A)b$ by using the family of two-term estimates with $n=1$ and $k=0$.

Example 2: *Symmetric positive definite matrices*

We consider the **Covariance** matrix:

- symmetric positive definite

- entries: $\alpha_{ij} = \begin{cases} 1 + i, & i = j \\ \frac{1}{|i-j|^5}, & i \neq j \end{cases}$

- order $p=700$

Estimating **$\log(\mathbf{A})\mathbf{b}$** with:

- $\mathbf{b} = \text{randn}(700,1)$ and
- $b_i = \cot(i)$, $i = 1, 2, \dots, 700$.

vector \mathbf{b}	(n,k)	relative error
randn(700,1)	(1,0)	1.8048e-4
	(1,3)	8.1004e-4
	(1,5)	8.6469e-4
$\mathbf{b}_i = \cot(i)$	(1,0)	6.1381e-7
	(3,5)	6.7791e-5
	(-4,1)	3.1369e-5

Estimating $\log(\mathbf{A})\mathbf{b}$ for various values of the parameters n and k .

Concluding remarks

- **Families of vector estimates** for $f(A)b$ were produced with complexity of order $O(p^2)$.
- The **presented numerical results** show the **satisfactory behavior** of the two-term vector estimates.

In applications where a high accuracy is not required, the derived estimates are efficient and easily applicable.

References

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- * C. Brezinski, P. Fika, M. Mitrouli, Estimations of the trace of powers of positive self-adjoint operators by extrapolation of the moments, *Elec. Trans. Numer. Anal.*, 39, pp. 144-155, 2012.
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Thank you!