

Max-min and min-max approximation problems for normal matrices revisited

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Bounding GMRES residual norm

$\mathbf{A}x = b$, $\mathbf{A} \in \mathbb{C}^{n \times n}$ is nonsingular, $b \in \mathbb{C}^n$,

$x_0 = \mathbf{0}$ and $\|b\| = 1$ for simplicity.

GMRES computes $x_k \in \mathcal{K}_k(\mathbf{A}, b)$ such that $r_k \equiv b - \mathbf{A}x_k$ satisfies

$$\|r_k\| = \min_{p \in \pi_k} \|p(\mathbf{A})b\| \quad (\text{GMRES})$$

$$\leq \max_{\|b\|=1} \min_{p \in \pi_k} \|p(\mathbf{A})b\| \quad (\text{worst-case GMRES})$$

$$\leq \min_{p \in \pi_k} \|p(\mathbf{A})\| \quad (\text{ideal GMRES})$$

where $\pi_k =$ degree $\leq k$ polynomials with $p(0) = 1$.

Two bounds on the GMRES residual norm

$$\max_{\|b\|=1} \min_{p \in \pi_k} \|p(\mathbf{A})b\| \leq \min_{p \in \pi_k} \|p(\mathbf{A})\|$$

- They are **equal** if \mathbf{A} is **normal**.

[Greenbaum, Gurvits '94; Joubert '94].

- The inequality can be **strict** if \mathbf{A} is **non-normal**.

[Toh '97; Faber, Joubert, Knill, Manteuffel '96].

How to prove the equality for normal matrices?

If \mathbf{A} is normal, then

$$\max_{\|b\|=1} \min_{p \in \pi_k} \|p(\mathbf{A})b\| = \min_{p \in \pi_k} \|p(\mathbf{A})\|.$$

- [Joubert '94] Proof using analytic methods of **optimization theory**, for real or complex data, only in the GMRES context.
- [Greenbaum, Gurvits '94]: Proof based mostly on **matrix theory**, only for real data but in a more general form.
- These proofs are quite **complicated**.
- Is there a straightforward proof that uses, e.g., known classical results of **approximation theory**?

Outline

- 1 Normal matrices and classical approximation problems
- 2 Best polynomial approximation for f on Γ
- 3 Proof
- 4 Connection to results by Greenbaum and Gurvits

Link to classical approximation problems

- \mathbf{A} is normal iff $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^*$, $\mathbf{Q}^*\mathbf{Q} = \mathbf{I}$.
- $\Gamma \equiv \{\lambda_1, \dots, \lambda_n\}$ is the set of eigenvalues of \mathbf{A} .
- For any function g defined on Γ denote

$$\|g\|_{\Gamma} \equiv \max_{z \in \Gamma} |g(z)|.$$

- $p \in \pi_k$ means

$$p(z) = 1 - \sum_{i=1}^k \alpha_i z^i.$$

- Then

$$\begin{aligned} \min_{p \in \pi_k} \|p(\mathbf{A})\| &= \min_{p \in \pi_k} \|\mathbf{Q}p(\mathbf{\Lambda})\mathbf{Q}^*\| = \min_{p \in \pi_k} \max_{\lambda_i} |p(\lambda_i)| \\ &= \min_{\alpha_1, \dots, \alpha_k} \left\| 1 - \sum_{i=1}^k \alpha_i z^i \right\|_{\Gamma}. \end{aligned}$$

Generalization

- Instead of 1 we consider a general **function** f defined on Γ .
Instead of $\{z^i\}_{i=1}^k$ we consider general **basis functions** φ_i .

We ask whether

$$\max_{\|b\|=1} \min_{p \in \mathcal{P}_k} \|f(\mathbf{A})b - p(\mathbf{A})b\| = \min_{p \in \mathcal{P}_k} \|f(\mathbf{A}) - p(\mathbf{A})\|$$

where \mathbf{A} is normal and p is of the form

$$p(z) = \sum_{i=1}^k \alpha_i \varphi_i(z) \in \mathcal{P}_k.$$

- A comment on \mathbb{R} versus $\mathbb{C} \rightarrow$ coefficients α_i .
- As in the previous

$$\min_{p \in \mathcal{P}_k} \|f(\mathbf{A}) - p(\mathbf{A})\| = \min_{p \in \mathcal{P}_k} \|f(z) - p(z)\|_{\Gamma}.$$

A polynomial of best approximation for f on Γ

Definition and notation

$p_* \in \mathcal{P}_k$ is a **polynomial of best approximation** for f on Γ when

$$\|f - p_*\|_{\Gamma} = \min_{p \in \mathcal{P}_k} \|f - p\|_{\Gamma}.$$

For $p \in \mathcal{P}_k$, define

$$\Gamma(p) \equiv \{z \in \Gamma : |f(z) - p(z)| = \|f - p\|_{\Gamma}\}.$$

Characterization of best approximation for f on Γ

[Chebyshev, Bernstein, de la Vallée Poussing, Haar, Remez, Zuhovickiĭ, Kolmogorov]

[Rivlin, Shapiro '61], [Lorentz '86]

Characterization theorem (complex case)

$p_* \in \mathcal{P}_k$ is **a polynomial of best approximation** for f on Γ

if and only if

there exist ℓ points $\mu_i \in \Gamma(p_*)$ where $1 \leq \ell \leq 2k + 1$, and ℓ real numbers $\omega_1, \dots, \omega_\ell > 0$ with $\omega_1 + \dots + \omega_\ell = 1$, such that

$$\sum_{j=1}^{\ell} \omega_j \overline{p(\mu_j)} [f(\mu_j) - p_*(\mu_j)] = 0, \quad \forall p \in \mathcal{P}_k.$$

Denote

$$\delta \equiv \|f - p_*\|_{\Gamma} = |f(\mu_j) - p_*(\mu_j)|, \quad j = 1, \dots, \ell.$$

Proof I

It suffices to prove that

$$\begin{aligned} \max_{\|b\|=1} \min_{p \in \mathcal{P}_k} \|f(\mathbf{A})b - p(\mathbf{A})b\| &\geq \min_{p \in \mathcal{P}_k} \|f(\mathbf{A}) - p(\mathbf{A})\| \\ &= \min_{p \in \mathcal{P}_k} \|f(z) - p(z)\|_{\Gamma}. \end{aligned}$$

Suppose that the eigenvalues of \mathbf{A} are sorted such that

$$\lambda_j = \mu_j, \quad j = 1, \dots, \ell.$$

Define the vector w

$$w = \mathbf{Q} \xi, \quad \xi \equiv [\sqrt{\omega_1}, \dots, \sqrt{\omega_\ell}, 0, \dots, 0]^T.$$

Then

$$\begin{aligned} 0 &= \sum_{j=1}^{\ell} \omega_j \overline{p(\mu_j)} [f(\mu_j) - p_*(\mu_j)] \\ &= \xi^H p(\mathbf{A})^H [f(\mathbf{A}) - p_*(\mathbf{A})] \xi \\ &= w^H p(\mathbf{A})^H [f(\mathbf{A}) - p_*(\mathbf{A})] w. \end{aligned}$$

Proof II

In other words,

$$f(\mathbf{A})b - p_*(\mathbf{A})w \perp p(\mathbf{A})w, \quad \forall p \in \mathcal{P}_k,$$

or, equivalently,

$$\|f(\mathbf{A})w - p_*(\mathbf{A})w\| = \min_{p \in \mathcal{P}_k} \|f(\mathbf{A})w - p(\mathbf{A})w\|.$$

Moreover

$$\begin{aligned} \|f(\mathbf{A})w - p_*(\mathbf{A})w\|^2 &= \|[f(\mathbf{A}) - p_*(\mathbf{A})]\xi\|^2 \\ &= \sum_{j=1}^{\ell} \xi_j^2 |f(\mu_j) - p_*(\mu_j)|^2 \\ &= \sum_{j=1}^{\ell} \omega_j \delta^2 = \delta^2 \\ &= \|f(\mathbf{A}) - p_*(\mathbf{A})\|^2. \end{aligned}$$

In summary, **for** $p_* \in \mathcal{P}_k$ we have constructed $w \in \mathbb{C}^n$ such that

$$\begin{aligned} \min_{p \in \mathcal{P}_k} \|f(\mathbf{A}) - p(\mathbf{A})\| &= \|f(\mathbf{A}) - p_*(\mathbf{A})\| \\ &= \|f(\mathbf{A})w - p_*(\mathbf{A})w\|^2 \\ &= \min_{p \in \mathcal{P}_k} \|f(\mathbf{A})w - p(\mathbf{A})w\| \\ &\leq \max_{\|b\|=1} \min_{p \in \mathcal{P}_k} \|f(\mathbf{A})b - p(\mathbf{A})b\|. \end{aligned}$$

The proof for **complex** \mathbf{A} is finished.

A note on the real case

- Assume that \mathbf{A} , $f(\mathbf{A})$ and $\varphi_i(\mathbf{A})$ are **real**. We look for a polynomial of a best approximation with **real coefficients**.
- **Technical problem**: \mathbf{A} can have **complex eigenvalues** but we look for a **real vector** b that maximizes

$$\min_{p \in \mathcal{P}_k} \|f(\mathbf{A})b - p(\mathbf{A})b\|.$$

- Γ is a set of points that appear in **complex conjugate pairs**.
- This **symmetry** with respect to the real axes has been used to find a real b and to prove the equality [Liesen, T. 2013].

Results by Greenbaum and Gurvits, Horn and Johnson

Theorem

[Greenbaum, Gurvits '94]

Let $\mathbf{A}_0, \mathbf{A}_1, \dots, \mathbf{A}_k$ be **normal matrices** that **commute**. Then

$$\max_{\|v\|=1} \min_{\alpha_1, \dots, \alpha_k} \left\| \mathbf{A}_0 v - \sum_{i=1}^k \alpha_i \mathbf{A}_i v \right\| = \min_{\alpha_1, \dots, \alpha_k} \left\| \mathbf{A}_0 - \sum_{i=1}^k \alpha_i \mathbf{A}_i \right\|.$$

Theorem

[Theorem 2.5.5, Horn, Johnson '90]

Commuting normal matrices **can be simultaneously unitarily diagonalized**, i.e., there exists a unitary \mathbf{U} so that

$$\mathbf{U}^H \mathbf{A}_i \mathbf{U} = \Lambda_i, \quad i = 0, 1, \dots, k.$$

Connection to results by Greenbaum and Gurvits

Using the theorem by Horn and Johnson we can equivalently rewrite the problem

$$\min_{\alpha_1, \dots, \alpha_k} \left\| \mathbf{A}_0 - \sum_{i=1}^k \alpha_i \mathbf{A}_i \right\|$$

in our notation

$$\min_{\alpha_1, \dots, \alpha_k} \left\| f(\mathbf{A}) - \sum_{i=1}^k \alpha_i \varphi_i(\mathbf{A}) \right\|$$

where \mathbf{A} is any diagonal matrix with distinct eigenvalues and f and φ_i are any functions satisfying

$$f(\mathbf{A}) = \mathbf{A}_0, \quad \varphi_i(\mathbf{A}) = \mathbf{A}_i, \quad i = 1, \dots, k.$$

Summary

- Inspired by the convergence analysis of GMRES
we formulated two general approximation problems involving normal matrices.
- We used a **direct link** between
 - approximation problems involving normal matrices,
 - classical approximation problemsand **proved** that

$$\max_{\|b\|=1} \min_{p \in \mathcal{P}_k} \|f(\mathbf{A})b - p(\mathbf{A})b\| = \min_{p \in \mathcal{P}_k} \|f(\mathbf{A}) - p(\mathbf{A})\|.$$

- Our results
 - represent a **generalization** of results by [Joubert '94],
 - offer **another point of view** to [Greenbaum, Gurvits '94].

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Thank you for your attention!