

Upper and Lower Bounds on Norms of Functions of Matrices

Given an n by n matrix A , find a set $S \subset \mathbf{C}$ that can be associated with A to give more information than the spectrum alone can provide about the 2-norm of functions of A .

- Field of values:

$$W(A) = \{\langle Aq, q \rangle : \langle q, q \rangle = 1\}.$$

- ϵ -pseudospectrum:

$$\sigma_\epsilon(A) = \{z \in \mathbf{C} : z \text{ is an eigenvalue of } A + E$$

$$\text{for some } E \text{ with } \|E\| < \epsilon\}.$$

- Polynomial numerical hull of degree k :

$$\mathcal{H}_k(A) = \{z \in \mathbf{C} : \|p(A)\| \geq |p(z)| \ \forall p \in \mathcal{P}_k\}.$$

Find a set S and scalars m and M with M/m of moderate size such that for all polynomials (or analytic functions) p :

$$m \cdot \sup_{z \in S} |p(z)| \leq \|p(A)\| \leq M \cdot \sup_{z \in S} |p(z)|.$$

- $S = \sigma(A)$, $m = 1$, $M = \kappa(V)$.

If A is normal then $m = M = 1$, but if A is nonnormal then $\kappa(V)$ may be huge. Moreover, if columns of V have norm 1, then $\kappa(V)$ is close to smallest value that can be used for M .

- If A is nonnormal, might want S to contain more than the spectrum. BUT...

If S contains more than $\sigma(A)$, must take $m = 0$ since if p is minimal polynomial of A then $p(A) = 0$ but $p(z) = 0$ only if $z \in \sigma(A)$.

- How to modify the problem?

$$m \cdot \sup_{z \in S} |p_{r-1}(z)| \leq \|p(A)\| \leq M \cdot \sup_{z \in S} |p(z)|$$

- If degree of minimal polynomial is r , then any $p(A) = p_{r-1}(A)$ for a certain $(r - 1)$ st degree polynomial – the one that matches p at the eigenvalues, and whose derivatives of order up through $t - 1$ match those of p at an eigenvalue corresponding to a t by t Jordan block.
- The largest set S where above holds with $m = 1$ is called the **polynomial numerical hull of degree $r - 1$** . In general, however, we do not know good values for M ($\ll \kappa(V)$).

Given a set S , for each p consider

$$\inf\{\|f\|_{\mathcal{L}^\infty(S)} : f(A) = p(A)\}. \quad (*)$$

Find scalars m and M such that for all p :

$$m \cdot (*) \leq \|p(A)\| \leq M \cdot (*).$$

- $f(A) = p(A)$ if $f(z) = p_{r-1}(z) + \chi(z)h(z)$ for some $h \in H^\infty(S)$. Here χ is the minimal polynomial (of degree r) and p_{r-1} is the polynomial of degree $r - 1$ satisfying $p_{r-1}(A) = p(A)$.
- $(*)$ is a **Pick-Nevanlinna interpolation problem**.

Given $S \subset \mathbb{C}$, $\lambda_1, \dots, \lambda_n \in S$, and w_1, \dots, w_n , find

$$\inf\{\|f\|_{\mathcal{L}^\infty(S)} : f(\lambda_j) = w_j, j = 1, \dots, n\}.$$

- If S is the open unit disk, then infimum is achieved by a function \tilde{f} that is a scalar multiple of a finite **Blaschke product**:

$$\begin{aligned}\tilde{f}(z) &= \mu \prod_{k=0}^{n-1} \frac{z - \alpha_k}{1 - \bar{\alpha}_k z}, \quad |\alpha_k| < 1 \\ &= \mu \frac{\gamma_0 + \gamma_1 z + \dots + \gamma_{n-1} z^{n-1}}{\bar{\gamma}_{n-1} + \bar{\gamma}_{n-2} z + \dots + \bar{\gamma}_0 z^{n-1}}.\end{aligned}$$

- Using second representation, Glader and Lindström showed how to compute \tilde{f} and $\|\tilde{f}\|_{\mathcal{L}^\infty(\mathcal{D})}$ by solving a simple eigenvalue problem.

Given $S \subset \mathbf{C}$, $\lambda_1, \dots, \lambda_n \in S$, and w_1, \dots, w_n ,
find

$$\inf\{\|f\|_{\mathcal{L}^\infty(S)} : f(\lambda_j) = w_j, j = 1, \dots, n\}.$$

- If S is a simply-connected open set, it can be mapped onto the open unit disk \mathcal{D} via a one-to-one analytic mapping g .

$$\inf\{\|F\|_{\mathcal{L}^\infty(S)} : F(\lambda_j) = w_j\} =$$

$$\inf\{\|f \circ g\|_{\mathcal{L}^\infty(S)} : (f \circ g)(\lambda_j) = w_j\} =$$

$$\inf\{\|f\|_{\mathcal{L}^\infty(\mathcal{D})} : f(g(\lambda_j)) = w_j\}.$$

- Some results also known when S is multiply-connected.

The Field of Values and 2 by 2 Matrices

- Suppose $S = W(A)$. Crouzeix showed that $M_{opt}(A, W(A)) \leq 11.08$ and he conjectures that $M_{opt}(A, W(A)) \leq 2$. (He proved this if A is 2 by 2 or if $W(A)$ is a disk.) In most cases, do not have good estimates for $m_{opt}(A, W(A))$, but...
- If A is a 2 by 2 matrix, since $W(A) = \mathcal{H}_1(A)$, the polynomial numerical hull of degree 1, and since any function of a 2 by 2 matrix A can be written as a first degree polynomial in A ,

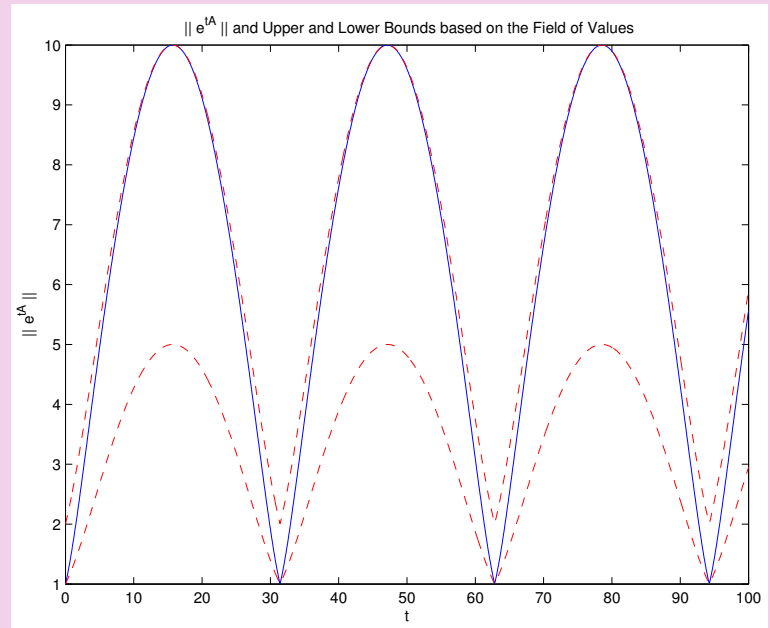
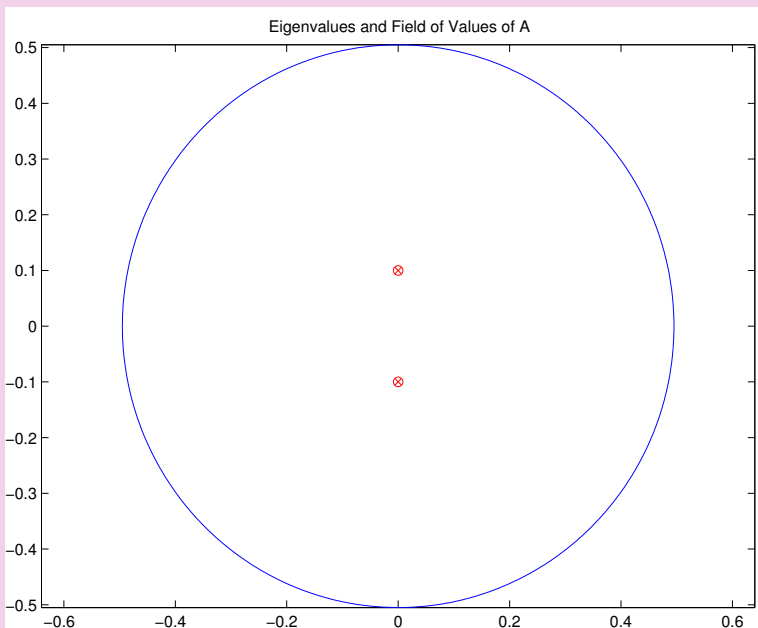
$$\begin{aligned} \|p(A)\| &\geq \|p_1\|_{\mathcal{L}^\infty(W(A))} \\ &\geq \inf\{\|f\|_{\mathcal{L}^\infty(W(A))} : f(A) = p(A)\}. \end{aligned}$$

Hence for 2 by 2 matrices:

$$m_{opt}(A, W(A)) = 1 \text{ and } M_{opt}(A, W(A)) \leq 2.$$

Example:

$$A = \begin{pmatrix} 0 & 1 \\ -.01 & 0 \end{pmatrix}$$



The Unit Disk and Perturbed Jordan Blocks

$$A = \begin{pmatrix} 0 & 1 & & \\ & \ddots & \ddots & \\ & & 0 & 1 \\ \nu & & & 0 \end{pmatrix}, \quad \nu \in (0, 1).$$

- The eigenvalues of A are the n th roots of ν : $\lambda_j = \nu^{(1/n)} e^{2\pi i j/n}$.
- For $\nu = 1$, A is a normal matrix with eigenvalues uniformly distributed about the unit circle. $W(A)$ is the convex hull of the eigenvalues. $\mathcal{H}_{n-1}(A)$ consists of the eigenvalues and the origin. The ϵ -pseudospectrum consists of disks about the eigenvalues of radius ϵ .

- For $\nu = 0$, A is a Jordan block with eigenvalue 0. $W(A)$ is a disk about the origin of radius $\cos(\pi/(n+1))$. $\mathcal{H}_{n-1}(A)$ is a disk of radius $1 - \log(2n)/n + \log(\log(2n))/n + o(1/n)$, and this is equal to the ϵ -pseudospectrum for $\epsilon \approx \log(2n)/(2n) - \log(\log(2n))/(2n)$.

$$A = \begin{pmatrix} 0 & 1 & & \\ & \ddots & \ddots & \\ & & 0 & 1 \\ \nu & & & 0 \end{pmatrix}, \quad \nu \in (0, 1).$$

Theorem. For any polynomial p ,

$$\|p(A)\| = \inf\{\|f\|_{\mathcal{L}^\infty(\mathcal{D})} : f(A) = p(A)\}.$$

Thus $M_{opt}(A, \mathcal{D}) = m_{opt}(A, \mathcal{D}) = 1$.

Proof: $A = V\Lambda V^{-1}$, where V^T is the Vandermonde matrix for the eigenvalues:

$$V^T = \begin{pmatrix} 1 & \lambda_1 & \dots & \lambda_1^{n-1} \\ 1 & \lambda_2 & \dots & \lambda_2^{n-1} \\ \vdots & \vdots & & \vdots \\ 1 & \lambda_n & \dots & \lambda_n^{n-1} \end{pmatrix}$$

How do we compute the minimal-norm interpolating function \tilde{f} ?

As noted earlier, it has the form

$$\tilde{f}(z) = \mu \frac{\gamma_0 + \gamma_1 z + \dots + \gamma_{n-1} z^{n-1}}{\bar{\gamma}_{n-1} + \bar{\gamma}_{n-2} z + \dots + \bar{\gamma}_0 z^{n-1}},$$

and satisfies $\tilde{f}(\lambda_j) = p(\lambda_j)$, $j = 1, \dots, n$.

If $\gamma = (\gamma_0, \dots, \gamma_{n-1})^T$, and Π is the permutation matrix with 1's on its skew diagonal (running from top right to bottom left), then these conditions are:

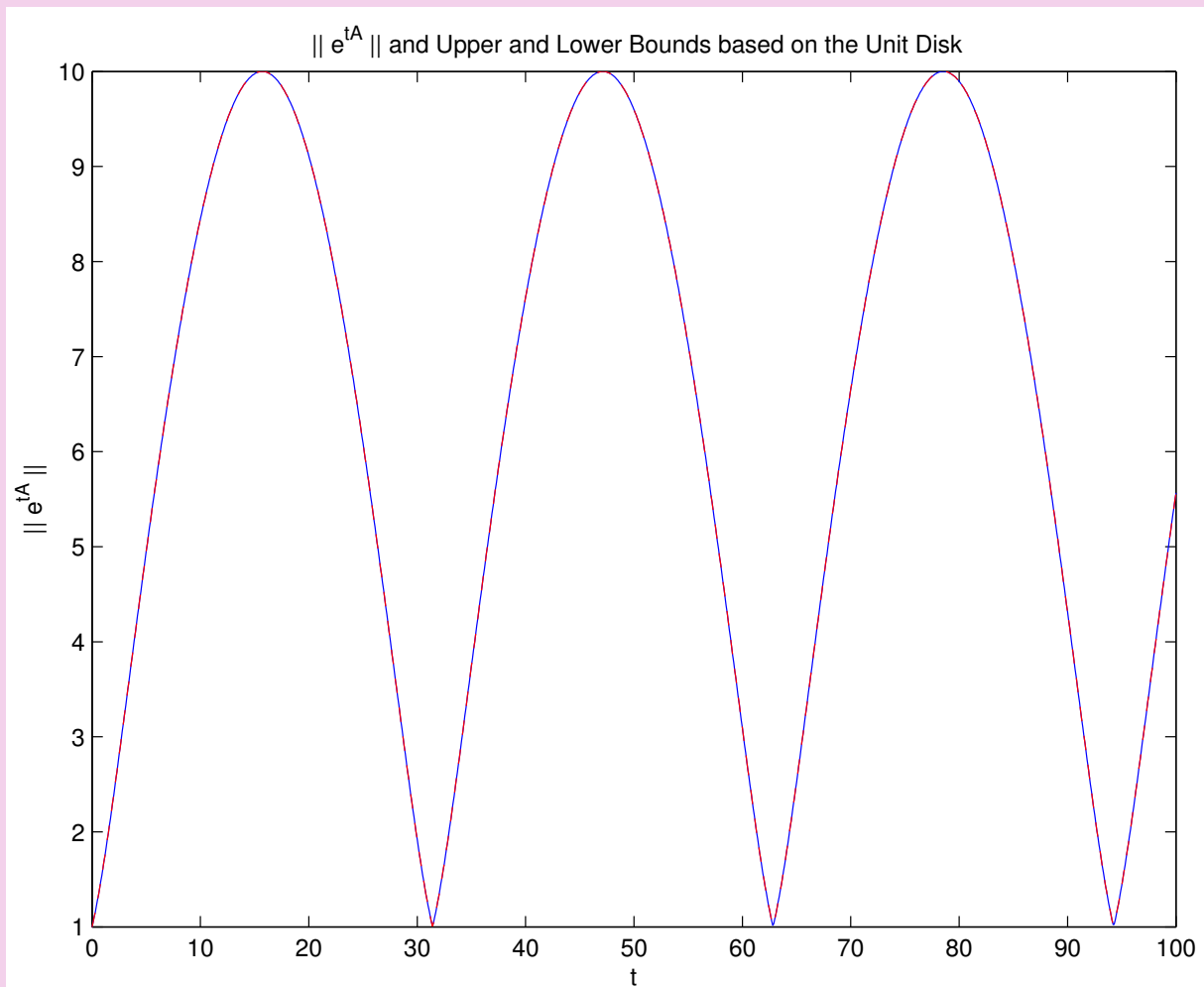
$$V^{-T} p(\Lambda) V^T \Pi \bar{\gamma} = (p(A))^T \Pi \bar{\gamma} = \mu \gamma.$$

Glader and Lindström showed that there is a real scalar μ for which this equation has a nonzero solution vector γ and that the largest such μ is $\|\tilde{f}\|_{\mathcal{L}^\infty(\mathcal{D})}$.

Since $(p(A))^T \Pi$ is (complex) symmetric, it has an SVD of the form $X \Sigma X^T$. The solutions to above equation are: $\gamma = \mathbf{x}_j$, $\mu = \sigma_j$; and $\gamma = i \mathbf{x}_j$, $\mu = -\sigma_j$. \square

Example:

$$A = \begin{pmatrix} 0 & 1 \\ -.01 & 0 \end{pmatrix}$$



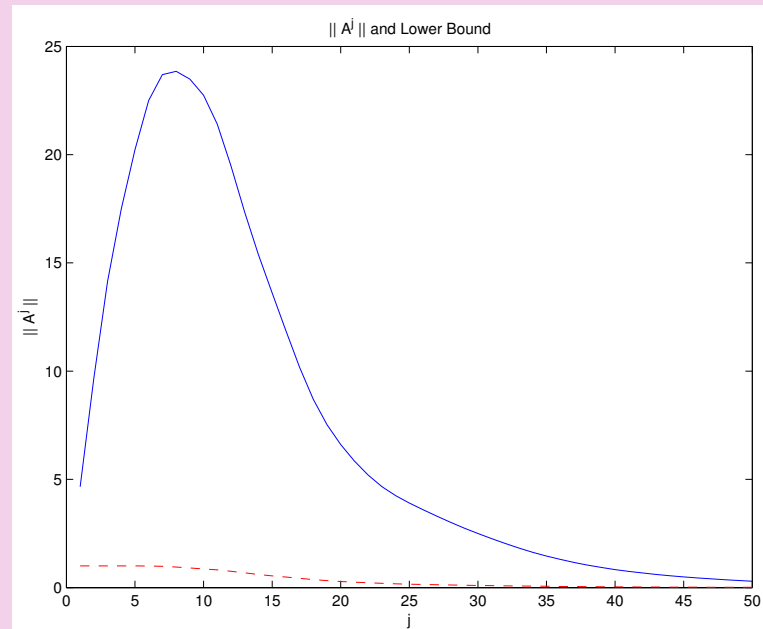
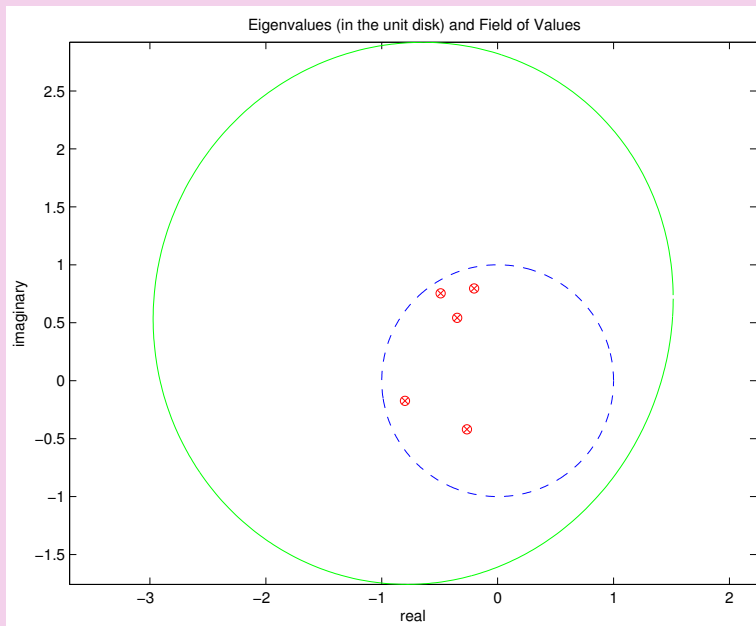
Corollary. If $A = V\Lambda V^{-1}$ where V^T is the Vandermonde matrix for Λ ; i.e., if A is a companion matrix with eigenvalues in \mathcal{D} , then $m_{opt}(A, \mathcal{D}) = 1$.

Proof:

$$V^{-T}p(\Lambda)V^T\Pi\bar{\gamma} = (p(A))^T\Pi\bar{\gamma} = \mu\gamma,$$

so $\|p(A)\| \geq |\mu|$. \square

Example: Companion matrix with 5 random eigenvalues in the unit disk. $\|A^j\|$ and lower bound.



The Unit Disk and More General Matrices

Map S to \mathcal{D} via g and study $A = g(\mathcal{A})$. Then

$$m_{opt}(A, \mathcal{D}) = \inf_{\|\mathbf{w}\|_\infty=1} \frac{\|V \text{diag}(\mathbf{w}) V^{-1}\|}{\inf\{\|f\|_{\mathcal{L}^\infty(\mathcal{D})} : f(\lambda_j) = w_j \ \forall j\}},$$

$$M_{opt}(A, \mathcal{D}) = \sup_{\|\mathbf{w}\|_\infty=1} \frac{\|V \text{diag}(\mathbf{w}) V^{-1}\|}{\inf\{\|f\|_{\mathcal{L}^\infty(\mathcal{D})} : f(\lambda_j) = w_j \ \forall j\}},$$

- Unless all (but a few) eigenvalues of A are very close to $\partial\mathcal{D}$, for certain \mathbf{w} 's the minimal norm interpolating function will be **huge**. If $\kappa(V)$ is very large, but not as large as the *constant of interpolation*:

$\sup_{\|\mathbf{w}\|_\infty=1} \inf\{\|f\|_{\mathcal{L}^\infty(\mathcal{D})} : f(\lambda_j) = w_j \ \forall j\}$,
then m_{opt} will be tiny.

- For other w 's, the minimal norm interpolating function is **well-behaved**. For example, $w_j = \lambda_j^k$ shows $M_{opt} \geq \max_k \|A^k\|$, which may be much greater than 1, especially if ill-conditioned eigenvalues are close to $\partial\mathcal{D}$.

In many cases, it appears difficult/impossible to find a set S where both m_{opt} and M_{opt} are of moderate size.

Summary and Thoughts

Given an n by n matrix A , we looked for a set $S \subset \mathbb{C}$ and scalars m and M with M/m of moderate size ($\ll \kappa(V)$ if $\kappa(V)$ is large) such that for all polynomials p :

$$\begin{aligned} m \cdot \inf\{\|f\|_{\mathcal{L}^\infty(S)} : f(A) = p(A)\} &\leq \|p(A)\| \\ &\leq M \cdot \inf\{\|f\|_{\mathcal{L}^\infty(S)} : f(A) = p(A)\}. \end{aligned}$$

- In a few exceptional cases (2 by 2 matrices, and perturbed Jordan blocks), we found such a set.

- In general, it seems difficult, perhaps impossible, to find such a set. The problem is that interpolating Blaschke products (like interpolating polynomials) can (but do not always) do wild things between the interpolation points. Hence to get a good value for m , need S to contain little more than $\sigma(A)$. But if $\kappa(V)$ is large, to get a good value for M , need S to contain significantly more than $\sigma(A)$.
- Perhaps the problem should be changed. Limit the class of polynomials. Or look for two different sets $S_m \subset \mathbf{C}$ for lower bounds on $\|p(A)\|$ and $S_M \subset \mathbf{C}$ for upper bounds.