### Polynomially isometric matrices in low dimensions Based on joint work with C. Brooks and N. Seguin

#### A.A. Condori

#### Florida Gulf Coast University, Fort Myers, FL

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A.A. Condori (FGCU)

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# Outline

- Functional calculus of a matrix A and its importance
- Notions of size/norm of a matrix A
- Olynomially isometric matrices
- Pseudospectra

- Partial bad news
- Partial good news
- We end with better news!

### Functional calculus of a matrix

Given an  $d \times d$  matrix A, it is of importance to consider matrices that can be generated by A, e.g.

$$A^2$$
,  $A^3 + 4A^2$ ,  $I + A + \frac{1}{2}A^2 + \frac{1}{6}A^3 + \frac{1}{24}A^4 + \dots$ , etc.

- I How can all such matrices be written concisely?
- Why is this important?

### First consumer: Linear Algebra

Eigenvalues: Every matrix A has an eigenvalue λ, i.e. there is a vector v ≠ 0 so that Av = λv.

One-line proof: Given  $w \neq 0$ , there is a polynomial p so that p(A)w = 0.

• Numerical methods: The convergence of algorithms to compute solutions to Ax = b rely on estimates on the size of certain p(A)'s and on their rate of decay, e.g. GMRES,

$$||Ax_m - b|| = \min\{||p(A)x_0|| : \deg p \le m, p(0) = 1\}.$$

### Second consumer: Differential Equations

**O** Computation of solutions: How to solve the DE below?

$$y''(t) + c_1 y'(t) + c_0 y(t) = f(t)$$

*Easy:* If  $p(z) = z^2 + c_1 z + c_0 = (z - \lambda_1)(z - \lambda_2)$  and let *D* denotes the differentiation operator, the DE above states p(D)y = f.

Obscay estimates: Solutions to the system of DEs x' = Ax depend on || f(A) ||, where f(t) = exp(tA).

### What is the size of a matrix?

For a vector  $x = (x_1, x_2, ..., x_n)$ , we use the *Euclidean norm* 

$$|x| = \sqrt{x_1^2 + x_2^2 + \ldots + x_n^2}$$

Likewise, one may use the **Frobenius/Euclidean norm**  $||A||_F$  which is easy to compute, but not useful in practice.

In practice, one needs to measure the "largest distortion" among unit vectors. The spectral norm ||A|| of A is defined as  $||A|| = \max_{|v|=1} |Av|$ .



**Exercises:** (1) Show that  $||A||_F = \sqrt{10}$  but  $||A|| = 2\sqrt{2}$ . (2) Show that the "smallest distortion" is  $\min_{|v|=1} |Av| = \frac{1}{||A^{-1}||}$ .

#### Our main question

Given a pair of square matrices A and B, what set of invariants (e.g. eigenvalues, Frobenius norms, etc.) are necessary and sufficient to ensure that A and B are "polynomially isometric"?

We say that A and B are polynomially isometric if

||p(A)|| = ||p(B)|| for all polynomials p.

What quantity may capture the "spirit" of a matrix?

The spectrum of A

The **spectrum**  $\sigma(A)$  of A is the set of its eigenvalues.

If you are familiar with the Spectral Theorem: What do you think?

# Moral: Spectral analysis is ineffective

The matrices  $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$  have equal spectra, but ||A|| = 1 and  $||B|| = \sqrt{2}$ , i.e. A and B are not polynomially isometric.

More generally, it is now well-known to numerical analysts that the spectrum is rarely sufficient to analyze a matrix. If so, what is a more effective replacement?

#### Pseudospectra

Roughly, a **pseudospectral plot** for a matrix A consists of contour plots of the norm  $||(zI - A)^{-1}||$  of its resolvent  $z \mapsto (zI - A)^{-1}$ . A recent book by Trefethen and Embree illustrates that pseudospectra may capture the "spirit" of a (non-normal) matrix more effectively.

As before, 
$$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$
 and  $B = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$ . Are their graphs *identical*?





Figure: Sketch of  $||(zI - A)^{-1}||$ 

Figure: Sketch of  $||(zI - B)^{-1}||$ 

#### Main Question

For A and B to be polynomially isometric, is it necessary and sufficient that A and B have identical pseudospectra?

Recall that A and B are **polynomially isometric** if

 $\|p(A)\| = \|p(B)\|$  for all polynomials p. (1)

Also, A and B have identical pseudospectra if

$$\|(zI - A)^{-1}\| = \|(zI - B)^{-1}\|$$
 for all  $z \in \mathbb{C}$ . (2)

The implication  $1 \Longrightarrow 2$  is due to Greenbaum & Trefethen (1993).

Moreover,  $1 \iff 2$  holds if at least one of the matrices is normal; this is a consequence of previous work of Brooks and Condori (2018).

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### Negative results

- An example due to Greenbaum and Trefethen (1993) shows that there are 5 × 5 matrices having identical pseudospectra for which the condition in (1) fails with p(z) = z.
- An example due to Fortier Bourque and Ransford (2009) shows that there are 4 × 4 matrices having identical pseudospectra but whose squares have distinct norms, i.e. (1) fails with p(z) = z<sup>2</sup>.

What about matrices of *lower* dimensions?

# The case of $2 \times 2$ matrices

#### Theorem

The following statements are equivalent for  $A, B \in \mathbb{M}_2$ .

A and B have identical pseudospectra.

**2** A and B are polynomially isometric.

3 A and B are unitarily similar.

"1  $\iff$  2" was observed by Greenbaum & Trefethen (1993).

A former FGCU student, N. Camacho and I also worked out a proof for his Senior Seminar project (2016); however, we did not realize the equivalence to 3 at time.

### Translation to an "easy-to-check" criterion

#### Corollary

For A,  $B \in \mathbb{M}_2$  to be polynomially isometric, it is necessary and sufficient that

tr 
$$A^*A = \text{tr } B^*B$$
, tr  $A = \text{tr } B$ , and tr  $A^2 = \text{tr } B^2$ .

**Exercise 1:** Why are 
$$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$
 and  $B = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$  not polynomially isometric?

**Exercise 2:** Construct matrices  $A, B \in \mathbb{M}_2$  which are similar but do not have identical pseudospectra.

# The case of $3\times 3$ matrices - Part 1

#### Theorem

The following statements are equivalent for  $A, B \in \mathbb{M}_3$ .

**()** A and B have identical pseudospectra and characteristic polynomials.

**2** tr 
$$(A^*A) =$$
tr  $(B^*B)$ , tr  $(A^*A^2) =$ tr  $(B^*B^2)$ ,  
tr  $(A^{*2}A^2) =$ tr  $(B^{*2}B^2)$ , tr  $A^k =$ tr  $B^k$  for  $k = 1, 2, 3$ 

In particular, A and B must be polynomially isometric.

Exercise: Consider the matrices

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} \text{ and } B = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 2 & 0 \end{bmatrix}.$$
(3)

Show that the matrices in (3) are polynomially isometric.

Are they unitarily similar?

The case of  $3 \times 3$  matrices - Part 2

#### Question

In the previous theorem, is the requirement that A and B have the same characteristic polynomials really important?

To address this question, consider the matrices

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ and } B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

Where does this leave us in our quest to characterize polynomially isometric matrices? Is having identical pseudospectra good enough?

The case of  $3 \times 3$  matrices - Part 3

#### Theorem

The following statements are equivalent for  $A, B \in \mathbb{M}_3$ .

- A and B have identical pseudospectra.
- A and B are polynomially isometric.

Moreover, if A and B also have the same minimal polynomial of degree 2, then the above statements are equivalent to

3.  $||A - \gamma_A I||_F = ||B - \gamma_B I||_F$ , where  $\gamma_A$  and  $\gamma_B$  are the eigenvalues corresponding to A and B, respectively, of largest multiplicity.

# Recap

- Functional calculus of a matrix A and its importance
- Output Notions of size/norm of a matrix A
- Olynomially isometric matrices
- Pseudospectra (as a more effective replacement to spectra)
  - Partial bad news (counterxamples in dimensions 4 & 5)
  - Partial good news (success in the 2 × 2 case)
  - We now have better news: success in the 3 × 3 case with easy-to-check criteria!

#### Challenge: Test your knowledge!

**Only two** of the following seven matrices are polynomially isometric but *not* unitarily similar. Find them!

$$\begin{aligned} A_1 &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad A_2 &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \quad A_3 &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}, \\ A_4 &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \quad A_5 &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}, \quad A_6 &= \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \\ A_7 &= \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}. \end{aligned}$$

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A.A. Condori (FGCU)

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