

# Pseudozeros and Pseudospectra

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- 1 Pseudozeros
- 2 Application of pseudozeros
- 3 Pseudozeros of interval polynomials
- 4 Pseudozeros of multivariate polynomials
- 5 Pseudosectra and structures
- 6 Open problems

# Motivations

- Polynomials appear in almost all areas in **scientific computing and engineering**
- The relationships between **industrial applications** and **polynomial systems solving** studied by the European Community Project FRISCO
- Applications in Computer Aided Design and Modeling, Mechanical Systems Design, Signal Processing and Filter Design, Civil Engineering, Robotics, Simulation
- The wide range of use of polynomial systems needs to have **fast and reliable** methods to solve them
  - **symbolic approach** based either on the theory of Gröbner basis or on the theory of resultants
  - **numeric approach** based on iterative methods like Newton's method or homotopy continuation methods
  - recently, **hybrid methods**, combining both symbolic and numeric methods

# Dealing with uncertainties

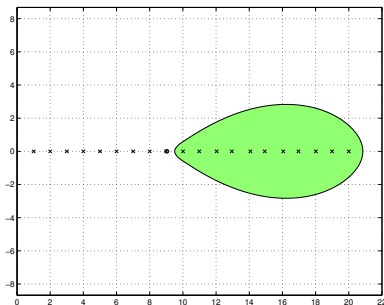
- In practice, from situations arising in science or engineering, the data are known only to a limited accuracy
- Analytical sensitivity analysis introduces a **condition number** that bounds the magnitudes of the (first order) changes of the roots with respect to the coefficient perturbations
- Continuous sensitivity analysis, introduced by Ostrowski, considers the uncertainty of the coefficients as a continuity problem. The most powerful tool of this last type of methods seems to be the **pseudozero set** of a polynomial

# An example for the univariate case

Computing the zeros of the Wilkinson polynomial of degree 20

$$\begin{aligned}W(x) &= (x-1)(x-2)\cdots(x-20) \\ &= x^{20} - 210x^{19} + \cdots + 20!\end{aligned}$$

Uncertainty of  $2^{-23}$  on the coefficient of  $x^{19}$



# Pseudozero set: definition

## **Perturbation :**

Neighborhood of polynomial  $p$

$$N_\varepsilon(p) = \{\hat{p} \in \mathbb{C}_n[z] : \|p - \hat{p}\| \leq \varepsilon\}.$$

## **Definition of the $\varepsilon$ -pseudozero set:**

$$Z_\varepsilon(p) = \{z \in \mathbb{C} : \hat{p}(z) = 0 \text{ for } \hat{p} \in N_\varepsilon(p)\}.$$

$\|\cdot\|$  a norm on the vector of the coefficients of  $p$

This set is formed by the zeros of polynomials “near  $p$ ”.

# Pseudozeros: brief survey of existing references

- ▶ Mosier (1986): Definition and study form the  $\infty$ -norm.
- ▶ Hinrichsen and Kelb: *spectral value sets*
- ▶ Trefethen and Toh (1994): Study for the 2-norm.  
pseudozeros  $\approx$  pseudospectra of the companion matrix.
- ▶ Chatelin and Frayssé (1996): propose a Synthesis in *Lectures on Finite Precision Computations* (SIAM)
- ▶ Stetter (1999,2004): *Numerical polynomial algebra*. Generalization of the previous works.
- ▶ Zhang (2001): Study of the influence of the basis for the 2-norm (condition number of the evaluation).
- ▶ Karow (2003): thesis on *Spectral value sets*

# Pseudozeros are easily computable

## Theorem 1

*The  $\varepsilon$ -pseudozeros set satisfies*

$$Z_\varepsilon(p) = \left\{ z \in \mathbb{C} : |g(z)| := \frac{|p(z)|}{\|\underline{z}\|_*} \leq \varepsilon \right\},$$

*where  $\underline{z} = (1, z, \dots, z^n)$  and  $\|\cdot\|_*$  is the dual norm of  $\|\cdot\|$ ,*

$$\|y\|_* = \sup_{x \neq 0} \frac{|y^* x|}{\|x\|}$$



# The nearest polynomial with a given root $p_u$

Let  $p$  be in  $\mathbb{C}_n[z]$  and  $u \in \mathbb{C}$ .

## Statement of the problem:

*Find a polynomial  $p_u \in \mathbb{C}_n[z]$  satisfying  $p_u(u) = 0$  and such that if there exists a polynomial  $q \in \mathbb{C}_n[z]$  with  $q(u) = 0$  then we get  $\|p - p_u\| \leq \|p - q\|$ .*

## We are looking for:

- an **expression** of  $p_u$ ;
- **uniqueness** of  $p_u$ .

# Computation of $p_u$

Let us denote  $\underline{u} := (1, u, u^2, \dots, u^n) \in \mathbb{C}^{n+1}$ .

There exists  $d \in \mathbb{C}^{n+1}$  satisfying  ${}^t d \underline{u} = \|\underline{u}\|_*$  et  $\|d\| = 1$ .

Let us define the polynomials  $r$  and  $p_u$  by

$$r(z) = \sum_{k=0}^n r_k z^k \quad \text{with} \quad r_k = d_k,$$

$$p_u(z) = p(z) - \frac{p(u)}{r(u)} r(z).$$

$p_u$  is the nearest polynomial of  $p$  with root  $u$ .

# Uniqueness of $p_u$

A sufficient condition for uniqueness :

## Theorem 2

*If the norm  $\|\cdot\|$  is strictly convex then  $p_u$  is unique.*

It is the case, for example, for the norms  $\|\cdot\|_p$  for  $1 < p < \infty$ .

We do not have unicity for  $\|\cdot\|_1$  and  $\|\cdot\|_\infty$ . For  $p(z) = 1 + z$

	$\ \cdot\ _1, u = 1$		$\ \cdot\ _\infty, u = 0$	
$p_u$	$p_1^{(1)}(z) = 0$	$p_1^{(2)}(z) = \frac{1}{3}(1 - z)$	$p_0^{(1)}(z) = z$	$p_0^{(2)}(z) = \frac{1}{2}z$
$p - p_i$	$z - 1$	$\frac{4}{3}z - \frac{2}{3}$	1	$\frac{1}{2}z + 1$
$\ p - p_i\ $	2	2	1	1

# Algorithm of computation

## Algorithm to draw the $\varepsilon$ -pseudozero set:

- 1 We mesh a square containing all the roots of  $p$  (MATLAB command: `meshgrid`).
- 2 We compute  $g(z) := \frac{|p(z)|}{\|z\|_*}$  for all the nodes  $z$  in the grid.
- 3 We draw the contour level  $|g(z)| = \varepsilon$  (MATLAB command: `contour`).

## Problems :

- Find a square containing **all the roots of  $p$  and all the pseudozeros**.
- Find a grid step that **separates all the roots**.

# Choice of the grid

Let  $p$  be a unitary polynomial of degree  $n$  and  $\{z_i\}$  the set of its  $n$  roots. Let us denote  $r = \max_{i=1;\dots;n} |z_i|$ . We have

$$r \leq \max\{1, \sum_{k=1}^n |p_k|\}.$$

Let us denote  $R := \max\{1, \sum_{i=1}^n |p_i| + n\varepsilon\}$ . We can prove (in  $\|\cdot\|_p$ )

$Z_\varepsilon(p) \subset B(0, R)$  the closed ball of centre 0 and radix  $R$ .

# Complexity of drawing pseudozero set

Let  $L$  be the length of the square and  $h$  the step of discretization. The evaluation of  $g(z) = \frac{|p(z)|}{\|z\|_*}$  needs

- the evaluation of polynomial  $p$ , that can be done in  $\mathcal{O}(n)$ ,
- the computation of the norm of a vector (the complexity depends on the norm).

Let us denote  $\mathcal{O}(\|\cdot\|_*)$  this complexity. The complexity of the algorithm to draw the pseudozero set is

$$\mathcal{O}\left(\left(\frac{L}{h}\right)^2 (n + \|\cdot\|_*)\right).$$

$L$  and  $h$  depend on  $n$  but also on the polynomial coefficients.

# A famous example

Pseudozero set of the *Wilkinson* polynomial

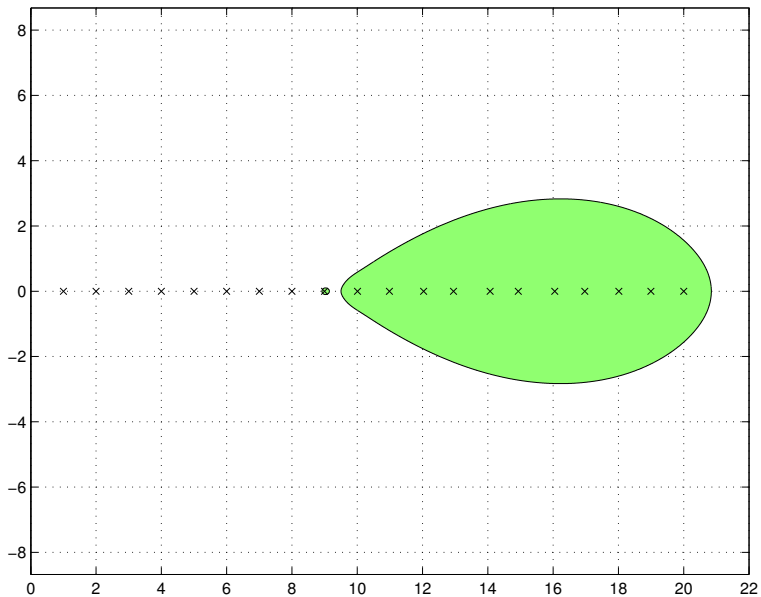
$$\begin{aligned}W_{20} &= (z-1)(z-2)\cdots(z-20), \\ &= z^{20} - 210z^{19} + \cdots + 20!.\end{aligned}$$

We perturb only the coefficient of  $z^{19}$  with  $\varepsilon = 2^{-23}$ .

One use the weighted-norm  $\|\cdot\|_\infty$ :

$$\|p\|_\infty = \max_i \frac{|p_i|}{m_i} \text{ with } m_i \text{ non negative}$$

with  $m_{19} = 1$ ,  $m_i = 0$  otherwise and the convention  $m/0 = \infty$  if  $m > 0$  and  $0/0 = 0$ .





# Interests of pseudozeros

Pseudozero set provides:

- a qualitative study of polynomials
- a better understanding of the results of polynomial algorithms
- a use of polynomials with coefficients known to a certain accuracy.

Drawback

- the cost

# Pseudozeros of real polynomials

If  $p \in \mathbb{R}_n[x]$ , we define

$$N_\varepsilon(p) := \{q \in \mathbb{R}_n[x] : \|p - q\| \leq \varepsilon\}.$$

Two cases :

- we seek the real pseudozeros: the same as the complex case;
- we seek all the complex non real pseudozeros.

We define the pseudozero set by

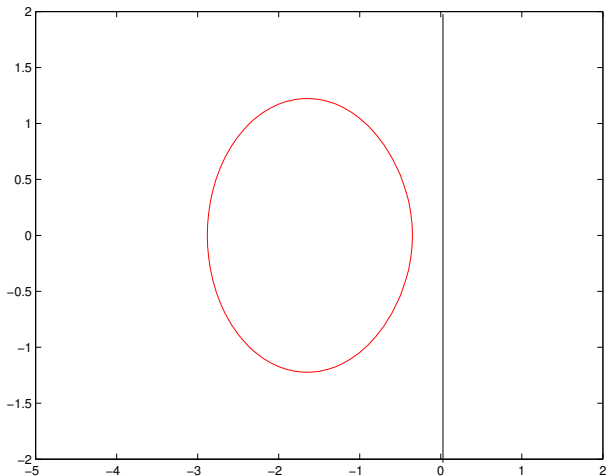
$$Z_\varepsilon(p) := \{z \in \mathbb{C} : \hat{p}(z) = 0 \text{ for } \hat{p} \in N_\varepsilon(p)\}.$$

$Z_\varepsilon(p)$  is symmetrical with respect to the real axis.

# Other applications of pseudozeros

# Hurwitz robust stability in control theory

**Hurwitz stability:** Real part of roots of  $p < 0$ .  
 $\varepsilon$ -pseudozero set of  $p(z) = (z + 1)^2$  for  $\varepsilon = 0.4$ .



# Computation of stability radius

$\mathcal{P}_n$ : polynomials of  $\mathbb{C}[X]$  of degree less or equal than  $n$

$\mathcal{M}_n$ : monic polynomials of  $\mathcal{P}_n$

$\|\cdot\|$ : the 2-norm of the coefficients of a polynomial

## Definition 1

A polynomial is said to be *stable* if all the roots have negative real part and *unstable* otherwise (Hurwitz stability).

The function *abscissa*  $a: \mathcal{P} \rightarrow \mathbb{R}$  is defined by

$$a(p) = \max\{\operatorname{Re}(z) : p(z) = 0\}.$$

A polynomial  $p$  is stable  $\iff a(p) < 0$

# Motivation

En **control theory**, a **transfer function** can be written as  $H(p) = \frac{N(p)}{D(p)}$  where  $N$  and  $D$  are polynomials.

The system is stable if  $D$  is a stable polynomial .

Question : if  $D$  is stable, is it far from unstable system?

Problem : Find the distance to the nearest unstable system.  
(we assume that  $D$  is monic)

# Statement of the problem

**Stability radius  $\beta(p)$**  : distance of the polynomial  $p \in \mathcal{M}_n$  from the set of monic unstable polynomials.

$$\beta(p) = \min\{\|p - q\| : q \in \mathcal{M}_n \text{ and } a(q) \geq 0\}.$$

**Statement of the problem:**

*Given a polynomial  $p \in \mathcal{M}_n$ , compute  $\beta(p)$ .*

# Solution

## Tools

- an explicit formula giving the **pseudozeros**
- the **continuous dependency** of the roots with respect to the polynomial **coefficients**
- the **Sturm sequences** to count the real roots

## The results

- a **algorithm** calculating  $\beta(p)$  with an arbitrary tolerance  $\tau$
- a **drawing** showing the pseudozeros at the distance  $\beta(p)$ 
  - enable a **qualitative analysis** of the result
  - **visualization** of the result



# Another characterization of $Z_\varepsilon(p)$

Let us denote  $h_{p,\varepsilon} : \mathbb{R}^2 \rightarrow \mathbb{R}$  the function defined by

$$h_{p,\varepsilon}(x, y) = |p(x + iy)|^2 - \varepsilon^2 \sum_{j=0}^{n-1} (x^2 + y^2)^j.$$

Then one has

$$Z_\varepsilon(p) = \{(x, y) \in \mathbb{R}^2 : h_{p,\varepsilon}(x, y) \leq 0\}$$

$\implies h_\varepsilon(\cdot, y)$  et  $h_\varepsilon(x, \cdot)$  are polynomials of degree  $2n$ .

# Theoretical results

## Proposition 1

*The function abscissa*

$$a: \mathcal{P}_n \rightarrow \mathbb{R}$$

*defined by  $a(p) = \max\{\operatorname{Re}(z) : p(z) = 0\}$  is continuous on  $\mathcal{M}_n$ .*

## Proposition 2

*One has the following relation*

$$\beta(p) = \min\{\|p - q\| : q \in \mathcal{M}_n \text{ and } a(q) = 0\}.$$

## Theorem 3

*The equation  $h_{p,\varepsilon}(0, y) = 0$  has a real solution  $y$  if and only if  $\beta(p) \leq \varepsilon$ .*

# Algorithm (bisection)

**Require:** a stable polynomial  $p$  and a tolerance  $\tau$

**Ensure:** a number  $\alpha$  such that  $|\alpha - \beta(p)| \leq \tau$

- 1:  $\gamma := 0, \quad \delta := \|p - z^n\|$
- 2: **while**  $|\gamma - \delta| > \tau$  **do**
- 3:    $\varepsilon := \frac{\gamma + \delta}{2}$
- 4:   **if** the equation  $h_{p,\varepsilon}(0, y) = 0, y \in \mathbb{R}$  has a solution **then**
- 5:      $\delta := \varepsilon$
- 6:   **else**
- 7:      $\gamma := \varepsilon$
- 8:   **end if**
- 9: **end while**
- 10: **return**  $\alpha = \frac{\gamma + \delta}{2}$

# Numerical simulation

For the polynomial  $p(z) = z + 1$ , the algorithm gives  $\beta(p) \approx 0.999996$

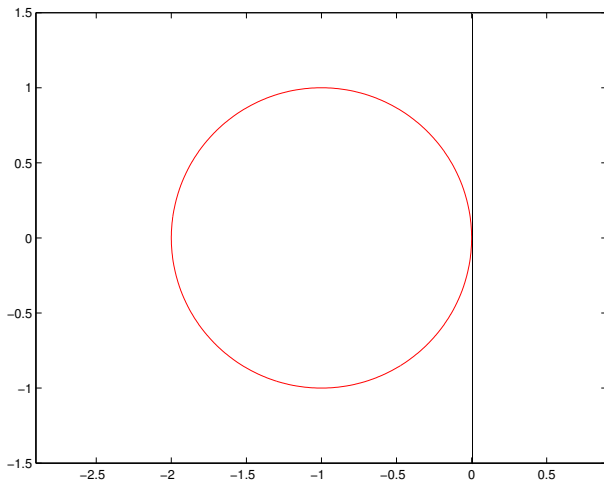


Figure :  $\beta(p)$ -pseudozero set of  $p(z) = z + 1$

# Numerical simulation (contd)

For the polynomial  $p(z) = z^3 + 4z^2 + 6z + 4$ , the algorithm gives  $\beta(p) \approx 2.61022$

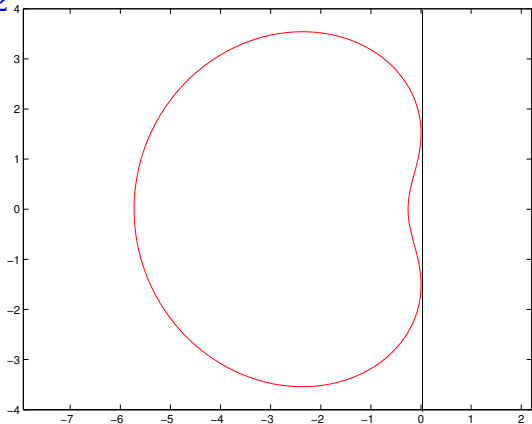


Figure :  $\beta(p)$ -pseudozero set of  $p(z) = z^3 + 4z^2 + 6z + 4$

# Pseudozero set of interval polynomials

# Interval polynomial

An **interval polynomial**: polynomial whose coefficients are real intervals.

We denote by  $\mathbb{IR}[z]$  the set of interval polynomials and by  $\mathbb{IR}_n[z]$  the set of interval polynomials with degree at most  $n$ .

Let  $p \in \mathbb{IR}_n[z]$ . We can write

$$p(z) = \sum_{i=0}^n [a_i, b_i] z^i.$$

The **zeros of an interval polynomial** is the set

$$\mathbb{Z}(p) := \{z \in \mathbb{C} : \text{there exist } m_i \in [a_i, b_i], i = 0 : n \text{ such that } \sum_{i=0}^n m_i z^i = 0\}.$$

$\Rightarrow$  Compute  $\mathbb{Z}(p)$ .

# Definition of real pseudozero set

Let  $p = \sum_{i=0}^n p_i z^i$  be a polynomial of  $\mathbb{R}_n[z]$

## **Perturbations :**

Real neighborhood of  $p$

$$N_\varepsilon^R(p) = \{\hat{p} \in \mathbb{R}_n[z] : \|p - \hat{p}\| \leq \varepsilon\}.$$

## **Definition of the real $\varepsilon$ -pseudozero set**

$$Z_\varepsilon^R(p) = \{z \in \mathbb{C} : \hat{p}(z) = 0 \text{ for } \hat{p} \in N_\varepsilon^R(p)\}.$$



# Computation of the real pseudozero set

## Theorem:

The real  $\varepsilon$ -pseudozero set satisfies

$$Z_\varepsilon^R(p) = Z(p) \cup \left\{ z \in \mathbb{C} \setminus Z(p) : h(z) := d(G_R(z), \mathbb{R}G_I(z)) \geq \frac{1}{\varepsilon} \right\},$$

where  $d$  is defined for  $x, y \in \mathbb{R}^{n+1}$  by

$$d(x, \mathbb{R}y) = \inf_{\alpha \in \mathbb{R}} \|x - \alpha y\|_*$$

and where  $G_R(z)$ ,  $G_I(z)$  are the real and imaginary part of

$$G(z) = \frac{1}{p(z)} (1, z, \dots, z^n)^T, \quad z \in \mathbb{C} \setminus Z(p)$$

Can be viewed as a special case of *spectral value set* [Karow 03]

# What for $\mathbb{R} \cap Z_\varepsilon^R(p)$ ?

## Lemma 1

*Given  $z \in \mathbb{R}$ ,  $z$  belongs to  $Z_\varepsilon^R(p)$  if and only if  $z$  belongs to  $Z_\varepsilon(p)$ .*

Draw the complex pseudozero set or the real pseudozero set on the real axis is similar.

# Some properties

The function  $d$  defined for  $x, y \in \mathbb{R}^{n+1}$  by

$$d(x, \mathbb{R}y) = \inf_{\alpha \in \mathbb{R}} \|x - \alpha y\|_*$$

satisfies

$$d(x, \mathbb{R}y) = \begin{cases} \sqrt{\|x\|_2^2 - \frac{\langle x, y \rangle^2}{\|y\|_2^2}} & \text{if } y \neq 0, \\ \|x\|_2 & \text{if } y = 0 \end{cases} \quad \text{for the norm } \|\cdot\|_2$$

$$d(x, \mathbb{R}y) = \begin{cases} \min_{\substack{i=0:n \\ y_i \neq 0}} \|x - (x_i/y_i)y\|_1 & \text{if } y \neq 0, \\ \|x\|_1 & \text{if } y = 0 \end{cases} \quad \text{for the norm } \|\cdot\|_\infty$$

# Some properties (cont'd)

## Proposition 3

The real  $\varepsilon$ -pseudozero set  $Z_\varepsilon^R(p)$  is *symmetric* with respect to the real axis.

## Proposition 4

The real  $\varepsilon$ -pseudozero set  $Z_\varepsilon^R(p)$  is *included* in the complex  $\varepsilon$ -pseudozero set.

# Algorithm to draw real pseudozero set

## Drawing of real $\varepsilon$ -pseudozero set:

- 1 We mesh a square containing all the roots of  $p$  (MATLAB command: `meshgrid`).
- 2 We compute  $h(z) := d(G_R(z), \mathbb{R}G_I(z))$  for all the nodes  $z$  in the grid.
- 3 We draw the contour level  $|h(z)| = \frac{1}{\varepsilon}$  (MATLAB command: `contour`).

# Pseudozero set with weighted norm

$$p(z) = \sum_{i=0}^n p_i z^i.$$

- identification of  $p$  with the vector  $(p_0, p_1, \dots, p_n)^T$
- $d := (d_0, \dots, d_n)^T \in \mathbb{R}^{n+1}$  represents the weight of the coefficients of  $p$
- $\|\cdot\|_{\infty, d}$  defined by

$$\|p\|_{\infty, d} = \max_{i=0:n} \{|p_i|/|d_i|\}.$$

- Its dual norm is

$$\|x\|_{1, d} := \sum_{i=0}^n |d_i| |x_i|.$$

# Zeros of interval polynomials and real pseudozero set

Let us denote  $p_c$  the central polynomial defined by

$$p_c(z) = \sum_{i=0}^n c_i z^i,$$

with  $c_i = (a_i + b_i)/2$ .

Let us denote  $d_i := (b_i - a_i)/2$ .

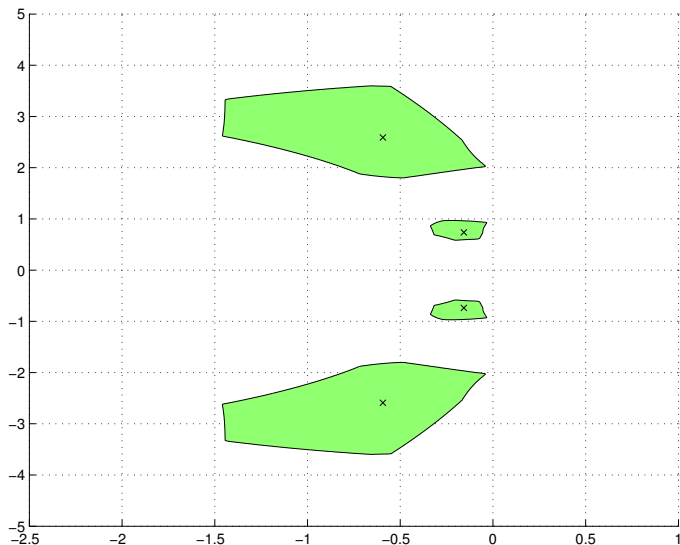
## **Proposition:**

With the notation above, we have

$$\mathbb{Z}(p) = Z_\varepsilon^R(p_c) \text{ with } \varepsilon = 1.$$

# Example 1

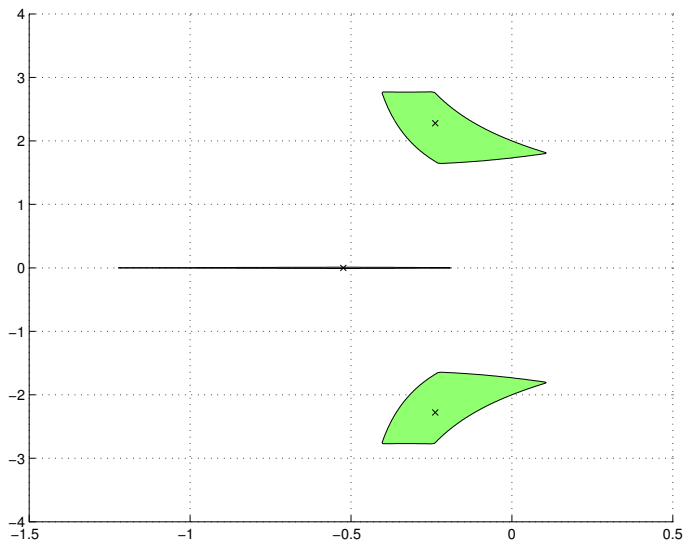
$$p(z) = [1, 2]z^4 + [3.2, 3]z^3 + [10, 14]z^2 + [3, 5\sqrt{2}]z + [5, 7]$$





# Example 2

$$p(z) = z^3 + z^2 + [3, 8]z + [1.5, 4]$$



# Problem: choice of the grid

## Lemma :

Let  $p(z) = \sum_{i=0}^n [a_i, b_i] z^i$  an interval polynomial and

$$R := 1 + \frac{\max_{i=0:n} \{\max\{|a_i|, |b_i|\}\}}{\min\{|a_n|, |b_n|\}}.$$

Then

$$\mathbb{Z}(p) \subset B(O, R),$$

where  $B(O, R)$  the ball in  $\mathbb{C}$  of centre  $O$  and radius  $R$ .

# Problems: discontinuities

## Lemma [Hinrichsen et Kelb]:

The function

$$d: \mathbb{R}^{n+1} \times \mathbb{R}^{n+1} \rightarrow \mathbb{R}_+, \quad (x, y) \mapsto d(x, \mathbb{R}y)$$

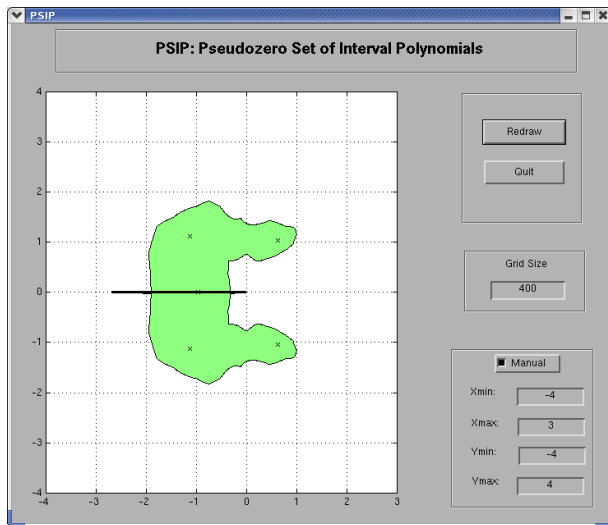
is **continue** for all  $(x, y)$  with  $y \neq 0$  or  $x = 0$  and **discontinue** for all  $(x, 0) \in \mathbb{R}^{n+1} \times \mathbb{R}^{n+1}$ ,  $x \neq 0$ .

$\Rightarrow$  Those discontinuities imply some difficulties for drawing near the real axis.

**Solution :** on the real axis, we draw complex pseudozero set.

# Presentation of PSIP

A tool to draw zeros of interval polynomials



# Presentation of PSIP (cont'd)

- a graphical interface for MATLAB (version 6.5 (R13))
- computation of grid that contains all the zeros
- possibilities of zoom and mesh refinement

## **Limitations :**

- problem if the leading interval contains 0
- problems with discontinuities

# Pseudozero set of multivariate polynomials

# Definitions (1/3)

A **monomial** in the  $n$  variables  $z_1, \dots, z_n$  is the power product

$$z^j := z_1^{j_1} \cdots z_n^{j_n}, \quad \text{with } j = (j_1, \dots, j_n) \in \mathbb{N}^n;$$

$j$  is the **exponent** and  $|j| := \sum_{\sigma=1}^n j_\sigma$  the *degree* of the monomial  $z^j$ .

## Definition 2

A *complex (real) polynomial* in  $n$  variables is a finite linear combination of monomials in  $n$  variables with coefficients from  $\mathbb{C}$  (from  $\mathbb{R}$ ),

$$p(z) = p(z_1, \dots, z_n) = \sum_{(j_1, \dots, j_n) \in J} a_{j_1 \dots j_n} z_1^{j_1} \cdots z_n^{j_n} = \sum_{j \in J} a_j z^j.$$

$\mathcal{P}^n(\mathbb{C})$  ( $\mathcal{P}^n(\mathbb{R})$ ) represents the set of all complex (real) polynomials in  $n$  variables.

## Definitions (2/3)

Given  $p = \sum_{j \in J} a_j z^j \in \mathcal{P}^n(\mathbb{K})$  with  $\mathbb{K} = \mathbb{R}$  or  $\mathbb{C}$

→  $|J|$  the number of elements of  $J$

If  $|J| = M$  and let  $\|\cdot\|$  be a norm on  $\mathbb{K}^M$

→  $\|p\|$  is the norm of the vector  $a = (\dots, a_j, \dots, j \in J)$

Given a norm  $\|\cdot\|$  on  $\mathbb{K}^N$  with  $\mathbb{K} = \mathbb{R}$  or  $\mathbb{C}$ , the **dual norm** is defined by

$$\|x\|_* := \sup_{\|y\|=1} |y^T x|.$$

Given a vector  $x \in \mathbb{K}^N$ , there exists a **dual vector**  $y \in \mathbb{K}^N$  with  $\|y\| = 1$  satisfying  $x^T y = \|x\|_*$ .

Norms	Dual norms
$\ x\ _1 := \sum_j  x_j $	$\ x\ _1^* = \max_j  x_j  = \ x\ _\infty$
$\ x\ _2 := (\sum_j  x_j ^2)^{1/2}$	$\ x\ _2^* = (\sum_j  x_j ^2)^{1/2} = \ x\ _2$
$\ x\ _\infty := \max_j  x_j $	$\ x\ _\infty^* = \sum_j  x_j  = \ x\ _1$



# Definitions (3/3)

Given  $\varepsilon > 0$ , the  $\varepsilon$ -neighborhood  $N_\varepsilon(p)$  of the polynomial  $p \in \mathcal{P}^n(\mathbb{K})$  is the set of all polynomials of  $\mathcal{P}^n(\mathbb{K})$  with  $\tilde{p} = \sum_{j \in \tilde{J}} \tilde{a}_j z^j \in \mathcal{P}^n(\mathbb{K})$  with support  $\tilde{J} \subset J$  and  $\|\tilde{p} - p\| \leq \varepsilon$ .

## Definition 3

A value  $z \in \mathbb{K}^n$  is an  $\varepsilon$ -pseudozero of a polynomial  $p \in \mathcal{P}^n$  if it is a zero of some polynomial  $\tilde{p}$  in  $N_\varepsilon(p)$ .

## Definition 4

The  $\varepsilon$ -pseudozero set of a polynomial  $p \in \mathcal{P}^n$  (denoted by  $Z_\varepsilon(p)$ ) is the set of all the  $\varepsilon$ -pseudozeros,

$$Z_\varepsilon(p) := \{z \in \mathbb{K}^n : \exists \tilde{p} \in N_\varepsilon(p), \tilde{p}(z) = 0\}.$$

# Pseudozeros of complex multivariate polynomials (1/2)

## Theorem 4 (Stetter)

The complex  $\varepsilon$ -pseudozero set of  $p = \sum_{j \in J} a_j z^j \in \mathcal{P}^n(\mathbb{C})$  verifies

$$Z_\varepsilon(p) = \left\{ z \in \mathbb{C}^n : g(z) := \frac{|p(z)|}{\|\mathbf{z}\|_*} \leq \varepsilon \right\}$$

where  $\mathbf{z} := (\dots, |z|^j, \dots, j \in J)^T$ .

# Pseudozeros of complex multivariate polynomials (2/2)

## Corollary 1 (Stetter)

The complex  $\varepsilon$ -pseudozero set of  $P = \{p_1, \dots, p_k\}$ ,  $k \in \mathbb{N}$  verifies

$$Z_\varepsilon(P) = \left\{ z \in \mathbb{C}^n : \frac{|p_l(z)|}{\|\mathbf{z}_l\|_*} \leq \varepsilon \text{ for } l = 1, \dots, k \right\},$$

where  $\mathbf{z}_l := (\dots, |z|^j, \dots, j \in J_l)^T$ .

We restrict our attention to situations where  $P$  as well as all the systems in  $N_\varepsilon(P)$  are 0-dimensional, that is, if the solution of the system is non-empty and finite.

## Theorem 5 (Stetter)

Each system  $\tilde{P} \in N_\varepsilon(P)$  has the same number of zeros (counting multiplicities) in a fixed pseudozero set connected component of  $Z_\varepsilon(P)$ .

# Pseudozeros of real multivariate polynomials: definition

A **real  $\varepsilon$ -neighborhood** of  $p$  is the set of all polynomials of  $\mathcal{P}^n(\mathbb{R})$ , close enough to  $p$ , that is to say,

$$N_\varepsilon^R(p) = \{\tilde{p} \in \mathcal{P}^n(\mathbb{R}) : \|p - \tilde{p}\| \leq \varepsilon\}.$$

The **real  $\varepsilon$ -pseudozero set** of  $p$  is defined to include all the zeros of the real  $\varepsilon$ -neighborhood of  $p$ :

$$Z_\varepsilon^R(p) = \{z \in \mathbb{C}^n : \tilde{p}(z) = 0 \text{ for } \tilde{p} \in N_\varepsilon^R(p)\}.$$

For  $\varepsilon = 0$ , the pseudozero set  $Z_0^R(p)$  is the set of the roots of  $p$  we denote  $Z(p)$ .

# Pseudozeros of real multivariate polynomials: computation

Distance of a point  $x \in \mathbb{R}^N$  from the linear subspace  $\mathbb{R}y = \{\alpha y, \alpha \in \mathbb{R}\}$

$$d(x, \mathbb{R}y) = \inf_{\alpha \in \mathbb{R}} \|x - \alpha y\|_*,$$

## Theorem 6

The real  $\varepsilon$ -pseudozero set of  $p = \sum_{j \in J} a_j z^j \in \mathcal{P}^n(\mathbb{R})$  verifies

$$Z_\varepsilon^R(p) = Z(p) \cup \left\{ z \in \mathbb{C}^n \setminus Z(p) : h(z) := d(G_R(z), \mathbb{R}G_I(z)) \geq \frac{1}{\varepsilon} \right\},$$

where  $G_R(z)$  and  $G_I(z)$  are the real and imaginary parts of

$$G(z) = \frac{1}{p(z)} (\dots, z^j, \dots, j \in J)^T, \quad z \in \mathbb{C}^n \setminus Z(p).$$

# Computing the distance

- computing real  $\varepsilon$ -pseudozero set  $Z_\varepsilon^R(p)$  needs to evaluate the distance  $d(G_R(z), \mathbb{R}G_I(z))$ .
- the 2-norm  $\|\cdot\|_2$  and  $\langle \cdot, \cdot \rangle$  the corresponding inner product

$$d(x, \mathbb{R}y) = \begin{cases} \sqrt{\|x\|_2^2 - \frac{\langle x, y \rangle^2}{\|y\|_2^2}} & \text{if } y \neq 0, \\ \|x\|_2 & \text{if } y = 0. \end{cases}$$

- the  $\infty$ -norm,

$$d(x, \mathbb{R}y) = \begin{cases} \min_{\substack{i=0:n \\ y_i \neq 0}} \|x - (x_i/y_i)y\|_1 & \text{if } y \neq 0, \\ \|x\|_1 & \text{if } y = 0. \end{cases}$$

- other  $p$ -norm with  $p \neq 2, \infty$ , no easy computable formula to calculate  $d(x, \mathbb{R}y)$ .

# Real pseudozeros of polynomial systems

## Corollary 2

The real  $\varepsilon$ -pseudozero set of  $P = \{p_1, \dots, p_k\}$ ,  $k \in \mathbb{N}$  verifies

$$Z_\varepsilon^R(P) = \bigcap_{l=1}^k \left( Z(p_l) \cup \left\{ z \in \mathbb{C}^n \setminus Z(p_l) : d(G_R^l(z), \mathbb{R}G_I^l(z)) \geq \frac{1}{\varepsilon} \right\} \right)$$

where  $G_R^l(z)$  and  $G_I^l(z)$  are the real and imaginary parts of

$$G^l(z) = \frac{1}{p_l(z)} (\dots, z^j, \dots, j \in J_l)^T, \quad z \in \mathbb{C}^n \setminus Z(p_l).$$

# Visualization of pseudozero sets (1/5)

- The descriptions of  $Z_\varepsilon(P)$  and  $Z_\varepsilon^R(P)$  given previously make it possible to **compute, plot and visualize** pseudozero set of multivariate polynomials.
- The pseudozero set is a subset of  $\mathbb{C}^n$  which can only be seen by its **projections on low dimensional spaces** that is often  $\mathbb{C}$ .

We have written a MATLAB program to compute and visualize these projections. This program requires the Symbolic Math Toolbox.



## Visualization of pseudozero sets (2/5)

For a given  $v \in \mathbb{C}^n$ , let  $Z_\varepsilon(P, j, v)$  be the projection of  $Z_\varepsilon(P)$  onto the  $z_j$ -space around  $v$ . Then, it follows that for  $P = \{p_1, \dots, p_k\}$ ,

$$Z_\varepsilon(P, j, v) = \left\{ z \in \mathbb{C}^n : z_i = v_i, i \neq j, \max_{l=1, \dots, k} \frac{|p_l(z)|}{\|\mathbf{z}_1\|_*} \leq \varepsilon \right\},$$

where  $\mathbf{z}_1 := (\dots, |z|^j, \dots, j \in J_l)^T$ .

One way for visualizing  $Z_\varepsilon(P, j, v)$  is to plot the values of the projection of

$$\text{ps}(z) := \log_{10} \left( \max_{l=1, \dots, k} \frac{|p_l(z)|}{\|\mathbf{z}_1\|_*} \right)$$

over a set of grid points around  $v$  in  $z_j$ -space.

## Visualization of pseudozero sets (3/5)

In the same way, we define for a given  $v \in \mathbb{C}^n$ ,  $Z_\varepsilon^R(P, j, v)$  by the projection of  $Z_\varepsilon^R(P)$  onto the  $z_j$ -space around  $v$ . It follows that for  $P = \{p_1, \dots, p_k\}$ ,

$$Z_\varepsilon^R(P, j, v) = \left\{ z \in \mathbb{C}^n : z_i = v_i, i \neq j, \max_{l=1, \dots, k} d(G_R^l(z), \mathbb{R}G_I^l(z))^{-1} \leq \varepsilon \right\}$$

where  $G_R^l(z)$  and  $G_I^l(z)$  are the real and imaginary parts of

$$G^l(z) = \frac{1}{p_l(z)} (\dots, z^j, \dots, j \in J_l)^T, z \in \mathbb{C}^n \setminus Z(p_l).$$

One way for visualizing  $Z_\varepsilon^R(P, j, v)$  is still to plot the values of the projection of

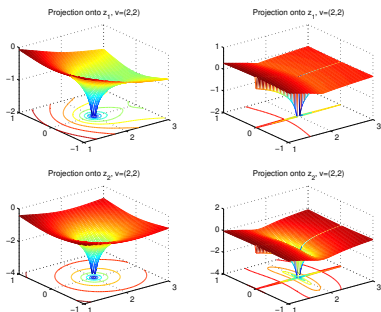
$$\text{ps}^R(z) := \log_{10} \left( \max_{l=1, \dots, k} d(G_R^l(z), \mathbb{R}G_I^l(z))^{-1} \right)$$

over a set of grid points around  $v$  in  $z_j$ -space.

# Visualization of pseudozero sets (4/5)

We examine the following system using the 2-norm: two unit balls intersection at (2,2),

$$P_1 = \begin{cases} p_1 = (z_1 - 1)^2 + (z_2 - 2)^2 - 1, \\ p_2 = (z_1 - 3)^2 + (z_2 - 2)^2 - 1. \end{cases}$$



Projections of the complex pseudozero set (on the left) and the real pseudozero set (on the right) of  $P_1$

# Visualization of pseudozero sets (5/5)

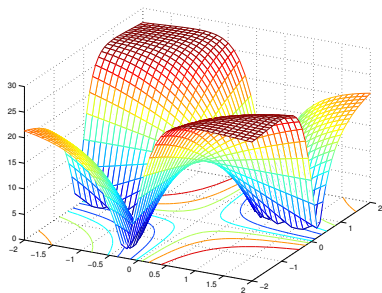
We can be only interested in the **real zeros** of a polynomial systems. In this case, we can only draw  $\mathbb{R}^n \cap Z_{\varepsilon}^R(P)$ .

$$P_2 = \begin{cases} p_1 = z_1^2 + z_2^2 - 1, \\ p_2 = 25z_1z_2 - 12. \end{cases}$$

We have computed the function

$$g(x, y) = \max_{l=1,2} \frac{p_l(x, y)}{\|\mathbf{z}_1\|_*},$$

with  $\mathbf{z}_1 := (\dots, |x + iy|^j, \dots, j \in J_l)^T$ .



Projection of the real pseudozero set of  $P_2$

# Pseudospectra of matrices

# Why structured matrices?

- Structured matrices are used in various fields such as signal processing, etc.
- Using the structure of a matrix, we get some better properties
- Substantial interest in algorithms for structured problems in recent years
- Growing interest in structured perturbation analysis
- In general perturbation and error analysis for structured solvers are performed with *general* perturbations: for a structured solver nothing else but structured perturbations are *possible*

# Our structures

Toeplitz matrices $(t_{i-j})_{i,j=0}^{n-1}$	$\begin{pmatrix} t_0 & t_{-1} & \cdots & t_{1-n} \\ t_1 & t_0 & \ddots & \vdots \\ \vdots & \ddots & \ddots & t_{-1} \\ t_{n-1} & \cdots & t_1 & t_0 \end{pmatrix}$
Hankel matrices $(h_{i,j})_{i,j=0}^{n-1}$	$\begin{pmatrix} h_0 & h_1 & \cdots & h_{n-1} \\ h_1 & h_2 & \ddots & h_n \\ \vdots & \ddots & \ddots & \vdots \\ h_{n-1} & h_n & \cdots & h_{2n-2} \end{pmatrix}$
Circulant matrices $(v_i)_{i=0}^{n-1}$	$\begin{pmatrix} v_0 & v_{n-1} & \cdots & v_1 \\ v_1 & v_0 & \ddots & \vdots \\ \vdots & \ddots & \ddots & v_{n-1} \\ v_{n-1} & \cdots & v_1 & v_0 \end{pmatrix}$

# Number of independant parameters

- In the following table,  $k$  represents the number of independant parameters for the different structures

Structure	general	Toeplitz	circulant	Hankel
$k$	$n^2$	$2n - 1$	$n$	$2n - 1$



# Notations

In this talk, we will use the following notation:

struct	Toeplitz, circulant or Hankel
$M_n(\mathbb{C})$	set of complex $n \times n$ matrices
$M_n^{\text{struct}}(\mathbb{C})$	set of structured complex $n \times n$ matrices
$\ \cdot\ $	spectral norm
$I, I_n$	identity matrix (with $n$ rows and columns)
$\sigma_{\min}(A)$	smallest singular value of $A$
$\Lambda(A)$	spectrum of $A$

# Definition of pseudospectra

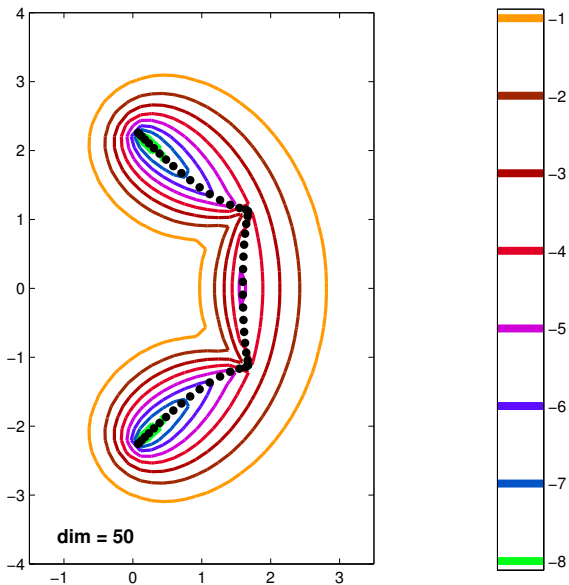
The  $\varepsilon$ -pseudospectrum of a matrix  $A$ , denoted  $\Lambda_\varepsilon(A)$ , is the subset of complex numbers consisting of all eigenvalues of all complex matrices within a distance  $\varepsilon$  of  $A$

## Definition 5

*For a real  $\varepsilon > 0$ , the  $\varepsilon$ -pseudospectrum of a matrix  $A \in M_n(\mathbb{C})$  is the set*

$$\Lambda_\varepsilon(A) = \{z \in \mathbb{C} : z \in \Lambda(X) \text{ where } X \in M_n(\mathbb{C}) \text{ and } \|X - A\| \leq \varepsilon\}.$$

# Example of pseudospectra



# Distance to singularity

## Definition 6

*Given a nonsingular matrix  $A \in M_n(\mathbb{C})$ , we define the distance to singularity by*

$$d(A) = \min\{\|\Delta A\| : A + \Delta A \text{ singular}, \Delta A \in M_n(\mathbb{C})\}.$$

## Lemma 2 (Gastinel)

*Let nonsingular  $A \in M_n(\mathbb{C})$ . Then we have*

$$d(A) = \|A^{-1}\|^{-1}.$$

## Theorem 7 (Trefethen)

*The following assertions are equivalent*

- (i)  $\Lambda_\varepsilon(A)$  is the  $\varepsilon$ -pseudospectrum of a matrix  $A$
- (ii)  $\Lambda_\varepsilon(A) = \{z \in \mathbb{C} : \|(zI - A)^{-1}\| \geq \varepsilon^{-1}\}$
- (iii)  $\Lambda_\varepsilon(A) = \{z \in \mathbb{C} : \sigma_{\min}(zI - A) \leq \varepsilon\}$
- (iv)  $\Lambda_\varepsilon(A) = \{z \in \mathbb{C} : d(zI - A) \leq \varepsilon\}$

# Definition of structured pseudospectra

The structured  $\varepsilon$ -pseudospectrum of a matrix  $A$ , denoted  $\Lambda_\varepsilon^{\text{struct}}(A)$ , is the subset of complex numbers consisting of all eigenvalues of all complex structured matrices within a distance  $\varepsilon$  of  $A$

## Definition 7

*For a real  $\varepsilon > 0$ , the structured  $\varepsilon$ -pseudospectrum of a matrix  $A \in M_n^{\text{struct}}(\mathbb{C})$  is the set*

$$\Lambda_\varepsilon^{\text{struct}}(A) = \{z \in \mathbb{C} : z \in \Lambda(X) \text{ where } X \in M_n^{\text{struct}}(\mathbb{C})$$

*and  $\|X - A\| \leq \varepsilon\}$ .*

# Structured distance to singularity

## Definition 8

*Given a nonsingular matrix  $A \in M_n^{\text{struct}}(\mathbb{C})$ , we define the structured distance to singularity by*

$$d^{\text{struct}}(A) = \min\{\|\Delta A\| : A + \Delta A \text{ singular}, \Delta A \in M_n^{\text{struct}}(\mathbb{C})\}.$$

## Theorem 8 (Rump)

*Let nonsingular  $A \in M_n^{\text{struct}}(\mathbb{C})$  with *struct* being Toeplitz, Hankel or circulant. Then we have*

$$d^{\text{struct}}(A) = d(A) = \|A^{-1}\|^{-1}.$$

# Characterisation of structured pseudospectra

## Lemma 3

*Given  $\varepsilon > 0$  and  $A \in M_n^{\text{struct}}(\mathbb{C})$  with struct Toeplitz or circulant, the structured  $\varepsilon$ -pseudospectrum satisfies*

$$\Lambda_\varepsilon^{\text{struct}}(A) = \{z \in \mathbb{C} : d^{\text{struct}}(A - zI) \leq \varepsilon\}.$$

## Theorem 9

*Given  $\varepsilon > 0$  and  $A \in M_n^{\text{struct}}(\mathbb{C})$  with struct Toeplitz or circulant, the  $\varepsilon$ -pseudospectrum and the structured  $\varepsilon$ -pseudospectrum satisfy*

$$\Lambda_\varepsilon^{\text{struct}}(A) = \Lambda_\varepsilon(A).$$



# What for others linear structures?

We do not have equality for Hermitian and skew-Hermitian structures.

For example for Hermitian structure we always have  $\Lambda_\varepsilon^{\text{herm}}(A) \subsetneq \mathbb{R}$  whereas one can find an Hermitian matrix such that  $\Lambda_\varepsilon(A) \not\subseteq \mathbb{R}$ .

# The polynomial eigenvalue problem

## Problem 10

Find the solutions  $(x, \lambda) \in \mathbb{C}^n \times \mathbb{C}$  of

$$P(\lambda)x = 0,$$

where

$$P(\lambda) = \lambda^m A_m + \lambda^{m-1} A_{m-1} + \cdots + A_0,$$

with  $A_k \in M_n(\mathbb{C})$ ,  $k = 0 : m$

If  $x \neq 0$  then  $\lambda$  is called an eigenvalue and  $x$  the corresponding eigenvector. The set of eigenvalues of  $P$  is denoted  $\Lambda(P)$ . We assume that  $P$  has only finite eigenvalues (and pseudoeigenvalues)

# Definition of pseudospectra

Let us define

$$\Delta P(\lambda) = \lambda^m \Delta A_m + \lambda^{m-1} \Delta A_{m-1} + \cdots + \Delta A_0,$$

where  $\Delta A_k \in M_n(\mathbb{C})$ .

## Definition 9

*For a given  $\varepsilon > 0$ , the  $\varepsilon$ -pseudospectrum of  $P$  is the set*

$$\Lambda_\varepsilon(P) = \{\lambda \in \mathbb{C} : (P(\lambda) + \Delta P(\lambda))x = 0 \text{ for some } x \neq 0 \\ \text{with } \|\Delta A_k\| \leq \alpha_k \varepsilon, k = 0 : m\}.$$

*The nonnegative parameters  $\alpha_1, \dots, \alpha_m$  allow freedom in how perturbations are measured*

## Lemma 4 (Tisseur and Higham (2001))

$$\Lambda_\varepsilon(P) = \{\lambda \in \mathbb{C} : d(P(\lambda)) \leq \varepsilon p(|\lambda|)\},$$

where  $p(x) = \sum_{k=0}^m \alpha_k x^k$ .

# Definition of structured pseudospectra

We suppose that  $\Delta A_k$  have a structure belonging to  $\text{struct}$ . We also suppose that all the matrices  $A_k$  and  $\Delta A_k$ ,  $k = 0 : n$ , belong to  $M_n^{\text{struct}}(\mathbb{C})$  for a given structure  $\text{struct}$ . Let

$$P(\lambda) = \lambda^m A_m + \lambda^{m-1} A_{m-1} + \cdots + A_0,$$

with  $A_k \in M_n^{\text{struct}}(\mathbb{C})$ ,  $k = 0 : m$  and

$$\Delta P(\lambda) = \lambda^m \Delta A_m + \lambda^{m-1} \Delta A_{m-1} + \cdots + \Delta A_0,$$

where  $\Delta A_k \in M_n^{\text{struct}}(\mathbb{C})$ .  $P(\lambda)$  and  $\Delta P(\lambda)$  belong to  $M_n^{\text{struct}}(\mathbb{C})$ .

## Definition 10

*We define the structured  $\varepsilon$ -pseudospectrum of  $P$  by*

$$\Lambda_\varepsilon^{\text{struct}}(P) = \{\lambda \in \mathbb{C} : (P(\lambda) + \Delta P(\lambda))x = 0 \text{ for some } x \neq 0 \\ \text{with } \Delta A_k \in M_n^{\text{struct}}(\mathbb{C}), \|\Delta A_k\| \leq \alpha_k \varepsilon, k = 0 : n\}.$$

# Characterisation of structured pseudospectra

## Lemma 5

For  $\text{struct} \in \{\text{Toep}, \text{circ}, \text{Hankel}\}$ , we have

$$\Lambda_{\varepsilon}^{\text{struct}}(P) = \{\lambda \in \mathbb{C} : d^{\text{struct}}(P(\lambda)) \leq \varepsilon p(|\lambda|)\},$$

where  $p(x) = \sum_{k=0}^n \alpha_k x^k$ .

## Theorem 11

Given  $\varepsilon > 0$  and  $P(\lambda) \in M_n^{\text{struct}}(\mathbb{C})$  a matrix polynomial with  $\text{struct} \in \{\text{Toep}, \text{circ}, \text{Hankel}\}$ , the  $\varepsilon$ -pseudospectrum and the structured  $\varepsilon$ -pseudospectrum satisfy

$$\Lambda_{\varepsilon}^{\text{struct}}(P) = \Lambda_{\varepsilon}(P).$$

# Real structured perturbations

Consider

$$P(\lambda) = \lambda^m A_m + \lambda^{m-1} A_{m-1} + \cdots + A_0,$$

with  $A_k \in M_n(\mathbb{R})$ ,  $k = 0 : m$  and

$$\Delta P(\lambda) = \lambda^m \Delta A_m + \lambda^{m-1} \Delta A_{m-1} + \cdots + \Delta A_0,$$

where  $\Delta A_k \in M_n(\mathbb{R})$ . Suppose that  $P(\lambda)$  is subject to structured perturbations:

$$[\Delta A_0, \dots, \Delta A_m] = D\Theta[E_0, \dots, E_m],$$

with  $D \in M_{n,1}(\mathbb{R})$ ,  $\Theta \in M_{1,t}(\mathbb{R})$  and  $E_k \in M_{t,n}(\mathbb{R})$ ,  $k = 0 : m$ .

For notational convenience, we introduce

$$E(\lambda) = E[I_n, \lambda I_n, \dots, \lambda^m I_n]^T = \lambda^m E_m + \lambda^{m-1} E_{m-1} + \cdots + E_0,$$

and

$$G(\lambda) = E(\lambda)P(\lambda)^{-1}D = G_R(\lambda) + iG_I(\lambda), \quad G_R(\lambda), G_I(\lambda) \in \mathbb{R}^t.$$

# Definition and characterisation of pseudospectra

## Definition 11

*The structured  $\varepsilon$ -pseudospectrum is defined by*

$$\Lambda_\varepsilon(P) = \{\lambda \in \mathbb{C} : (P(\lambda) + D\Theta E(\lambda))x = 0 \text{ for some } x \neq 0, \|\Theta\| \leq \varepsilon\}$$

We denote for  $x, y \in \mathbb{R}^t$ ,

$$d(x, \mathbb{R}y) = \inf_{\alpha \in \mathbb{R}} \|x - \alpha y\|,$$

the distance of the point  $x$  from the linear subspace  $\mathbb{R}y = \{\alpha y, \alpha \in \mathbb{R}\}$ .

## Theorem 12

$$\Lambda_\varepsilon(P) = \{\lambda \in \mathbb{C} \setminus \Lambda(P) : d(G_R(\lambda), \mathbb{R}G_I(\lambda)) \geq 1/\varepsilon\} \cup \Lambda(P)$$



# Conclusion

We have

- The structured pseudospectrum is equal to the pseudospectrum for the two following structures: Toeplitz and circulant
- This result is false for structures Hermitian and skew-Hermitian
- We have generalized these results to pseudospectra of matrix polynomials.
- We have given a formula for structured pseudospectra of real matrix polynomials

# Open problems

# Pseudozeros of interval polynomials

## Problem

Given

- an ball polynomial  $p(x) = \sum_{i=0}^n B(a_i, r_i)x^i$  with  $a_i \in \mathbb{C}$ ,  $r_i \geq 0$  and
- $z \in \mathbb{C}$

does there exist  $c_i \in B(a_i, r_i)$  such that  $p_c(z) := \sum_{i=0}^n c_i z^i = 0$

## Solution [Mosier (1986)]

The  $c_i$  exist if and only if

$$\frac{|p(z)|}{r_0 + r_1|z| + \cdots + r_n|z|^n} \leq 1$$

# Pseudozeros of interval polynomials

## Problem

Given

- an interval polynomial  $p(x) = \sum_{i=0}^n [a_i; b_i] x^i$  with  $a_i, b_i \in \mathbb{R}$ ,  $a_i \leq b_i$  and
- $z \in \mathbb{C}$

does there exist  $c_i \in [a_i, b_i]$  such that  $p_c(z) := \sum_{i=0}^n c_i z^i = 0$

Given a vector  $d := (d_0, \dots, d_n)^T$  in  $\mathbb{C}^{n+1}$ , we consider the weighted norms

$$\|x\|_{\infty, d} = \max_{i=0:n} \{|p_i| / |d_i|\} \quad \text{and} \quad \|x\|_{1, d} := \sum_{i=0}^n |d_i| |x_i|.$$

# Pseudozeros of interval polynomials

We define

$$\text{dist}_d(x, \mathbb{R}y) = \inf_{\alpha \in \mathbb{R}} \|x - \alpha y\|_{1,d},$$
$$\text{dist}_d(x, \mathbb{R}y) = \begin{cases} \min_{\substack{i=0:n \\ y_i \neq 0}} \|x - (x_i/y_i)y\|_{1,d} & \text{if } y \neq 0, \\ \|x\|_{1,d} & \text{if } y = 0. \end{cases}$$

and  $G_R(p, z)$  and  $G_I(p, z)$  being the real and imaginary parts of

$$G(p, z) = \frac{1}{p(z)} (1, z, \dots, z^n)^T, \quad z \in \mathbb{C} \text{ with } p(z) \neq 0.$$

# Pseudozeros of interval polynomials

Let  $p_m(x) = \sum_{i=0}^n m_i x^i$  with  $m_i = (a_i + b_i)/2$  et  $d_i := (b_i - a_i)/2$ .

## Solution

The  $c_i$  exist if and only if either  $p(z) = 0$  or

$$\text{dist}_d(G_R(p_m, z), \mathbb{R}G_I(p_m, z)) \geq 1$$

# Pseudozeros of interval polynomials

## Problem

Given

- an ball polynomial  $p(x) = \sum_{i=0}^n ([a_j; b_j] + i[c_j; d_j])x^j$  with  $a_j, b_j, c_j, d_j \in \mathbb{C}$ ,  $a_j \geq b_j$ ,  $c_j \geq d_j$  and
- $z \in \mathbb{C}$

does there exist  $\alpha_j \in [a_j; b_j]$  and  $\beta_j \in [c_j; d_j]$  such that  $p_c(z) := \sum_{i=0}^n (\alpha_j + i\beta_j)z^i = 0$

For the moment, no closed formula ! Maybe NP-hard ?

Thank you for your attention