

Condition-based

Low-Degree Approximation

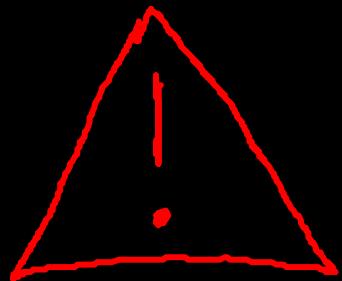
of Real Polynomial Systems

Josué TONELLI-CUETO  
UT San Antonio

Elias TSIGARIDAS  
Inria Paris & IMJ-PRG

Slides at:

[https://tonellicueto.xyz/pdf/JMM2023\\_slides.pdf](https://tonellicueto.xyz/pdf/JMM2023_slides.pdf)



# WARNING

There will be one result  
on Eigenvalue Computations,  
the talk will focus  
on real polynomial systems

# PROBLEM

Given a real polynomial system

$$g_1(x) = \dots = g_q(x) = 0$$

in  $n$  variables,

- 1) what can we say about the conditioning of solving?
- 2) what does the condition say about the zero set?

# Conditioning

- The condition number depends on the metric
  - how we measure errors —
- The condition number depends on the encoding
  - how we write the problem —

# Condition à la Demmel-Renegar — conic framework

$\mathcal{X}$  input space

$\Sigma \subseteq \mathcal{X}$  ill-posed inputs

$$c(i) := \frac{1}{d(i, \Sigma)}$$



Set Up

$$\mathcal{H}_d := \prod_{i=1}^n \mathbb{R}[x_0, \dots, x_n]_d.$$

$\mathbb{P}^n$  Real Projective Space

$$\Sigma = \{ f \in \mathcal{H}_d \mid f \text{ has a singular zero} \}$$

# Weyl Norm

$$\|g\|_W := \sqrt{\sum_{i=1}^n \sum_{|\alpha|=d_i} \left( \frac{d_i}{\alpha} \right)^{-1} |g_{i,\alpha}|^2}$$

where

$$g_i = \sum g_{i,\alpha} x^\alpha \quad (x := x_0^{\alpha_0} \cdots x_n^{\alpha_n})$$

$$\left( \frac{d_i}{\alpha} \right) = \frac{d_i!}{\alpha_0! \cdots \alpha_n!}$$

# Weyl Condition Number

(Cucker, Krick, Malajovich, Wschebor)

$$\kappa_w(\gamma) := \frac{\|\gamma\|_w}{\text{dist}_w(\gamma, \Sigma)}$$

$$= \sup_{x \in \mathbb{S}^n} \frac{\|\gamma\|_w}{\sqrt{\|\gamma(x)\|^2 + \|D_x \gamma^{-1} \Delta^{1/2}\|^2}}$$

where  $\Delta = \text{diag}(d_i)$  &  $D_x \gamma: T_x \mathbb{S}^n \rightarrow \mathbb{R}^n$

# Main Theorem (Simple Form)

$$\#Z(g, P^n)$$

$$\leq D^{n/2} \text{poly}(n, \log D)^n \log^n(2\chi_w(g))$$

where  $D = \max d_i$

# MAIN THEOREM (COMPLEX FORM)

There is a cover

$\{B(x, 1/c\sqrt{D})\}_{x \in G}$  of size  $O(D^{n/2})$

of  $S^n$  s.t. for all  $x \in G$  &  $g \in \mathcal{H}_d$ ,

there is  $\Phi_{x,g} \in \mathbb{R}[X_0, \dots, X_n]^n$  of degree

$$\leq \text{poly}(n, \log D) \log(2K_w(g))$$

s.t.  $\#\mathcal{Z}(g, T_x S^n \cap B(x, 1/c\sqrt{D})) \leq \#\mathcal{Z}(\Phi_{x,g}, T_x S^n)$ .

Moreover, for all  $g \in \mathcal{Z}(g, T_x S^n \cap B(x, 1/c\sqrt{D}))$

there is  $z \in \mathcal{Z}(\Phi_{x,g}, T_x S^n)$  converging quadratically under Newton's method.

# Corollary 1 of MAIN RESULT

- If  $\#\mathcal{Z}(\gamma) \geq D^n$ , then  
 $\kappa_w(\gamma) \geq 2^{D^{n/2}/\text{poly}(n, \log D)^n}$
- Real systems with many zeros  
are badly-conditioned

## COROLLARY 2 OF MAIN THEOREM

Let  $f \in \mathcal{F}^d$  be random such that for all  $x \in S^n$ , the  $f_i(x)$  are independent, subgaussian and with anti-concentration. Then:

$$\left( \mathbb{E} \#Z(f, P^n)^e \right)^{1/e} \leq D^{n/2} \text{poly}(n, \log D)^n e^n$$

## COROLLARY 2 OF MAIN THEOREM

Let  $f \in \mathcal{J}_d$ . Under very general random hypotheses,

$$\#\mathcal{E}(f, P^n)^{1/n}$$

is subexponential with constant

$$D^{1/2} \text{poly}(n, \log D)$$

i.e.

$$P(\#\mathcal{E}(f, P^n)^{1/n} \geq t) \leq e^{-\frac{t}{D^{1/2} \text{poly}(n, \log D)}}$$

## COROLLARY 2 OF MAIN THEOREM

A KSS random polynomial system

$$f \in \mathcal{A}_{(0, \dots, D)}$$

has its number of real zeros  
concentrated around

$$D^{n/2} = \mathbb{E} \# Z(f, P^n)$$

# COMPARISON WITH LERARIO ET.AL.

## OUR APPROACH

Taylor Approx.

Many local approx.

Control the moments

Robust

Exploits analyticity!

## LERARIO'S APPROACH

S. Harmonic Approx.

A global approx.

Control only probability

Only KSS

(Lerario, Diatta; '22) (Breiding, Keneshlou, Lerario; '22)



Set Up

$$\mathcal{P}_A := \left\{ f \in \mathbb{R}[x_0, \dots, x_n] \mid 0 \in \text{supp } f \subseteq A \right\}^n$$

$$I^n := [-1, 1]^n$$

$$\Sigma = \left\{ f \in \mathcal{P}_A \mid f \text{ has a singular zero in } I^n \right\}$$

# 1-Norm

$$\|g\|_1 := \max_{\alpha \in A} \sum |g_{i,\alpha}|$$

where

$$g_i = \sum g_{i,\alpha} x^\alpha \quad (x := x_1 \cdots x_n)$$

# Cubic Condition Number (TC, Tsigaridas)

$$C_1(\varphi) := \sup_{x \in I^n} \frac{\|\varphi\|_1}{\max \left\{ \|\varphi(x)\|_\infty, \|D_x \varphi^{-1} \Delta\|_{\infty, \infty}^{-1} \right\}}$$
$$\sim \frac{\|\varphi\|_1}{\text{dist}_1(\varphi, \Sigma)}$$

where  $\Delta = \text{diag}(d_i)$  &  $D_x \varphi: \mathbb{R}^n \rightarrow \mathbb{R}^n$

# Main Theorem (Simple Form)

$$\#\mathcal{Z}(g, I^n) \leq \text{poly}(n)^n \log^{2n}(2D) \log^n(2C_1(g))$$

where  $D = \max d_i$

# MAIN THEOREM (COMPLEX FORM)

There is a partition into boxes

$$\{B\}_{B \in \mathcal{B}} \text{ of size } O(\log^n(2D))$$

of  $\mathbb{T}^n$  s.t. for all  $B \in \mathcal{B}$  &  $g \in \mathcal{H}_d$ ,

there is  $\Phi_{B,g} \in \mathbb{R}[x_0, \dots, x_n]^n$  of degree

$$\leq \text{poly}(n) \log D \log(2C_1(g))$$

s.t.  $\#\mathcal{Z}(g, B) \leq \#\mathcal{Z}(\Phi_{B,g})$ .

Moreover, for all  $g \in \mathcal{Z}(g, B)$ , there is  $z \in \mathcal{Z}(\Phi_{B,g})$  converging quadratically under Newton's method.

# Corollary 1 of MAIN RESULT

If  $\#\mathcal{Z}(\gamma, I^n) \geq D^K$ , then

$$C_1(\gamma) \geq 2^{D^K / \text{poly}(n, \log D)^n}$$

Real systems with many zeros  
are badly-conditioned

## COROLLARY 2 OF MAIN THEOREM

Let  $f \in \mathcal{P}_A$  be random  
such that for all  $x \in I^n$ ,  
the  $f_i(x)$  are independent, subgaussian  
and with anti-concentration. Then:

$$\left( \mathbb{E} \#Z(f, I^n)^\epsilon \right)^{1/\epsilon} \leq \text{poly}(n)^\epsilon \log^{2n} (2D) \epsilon^n$$

## COROLLARY 2 OF MAIN THEOREM

Let  $f \in \mathcal{P}_A$ . Under very general random hypotheses,

$$\#\mathcal{E}(f, I^n)^{1/n}$$

is subexponential with constant  
 $\log^2(2D) \text{poly}(n)$

i.e.

$$P(\#\mathcal{E}(f, P^n)^{1/n} \geq t) \leq e^{1 - \frac{t}{\log^2(2D) \text{poly}(n)}}$$



# Main Tricks

- Smale's  $\alpha$ -theory  
is stable under analytic truncation
- Well-conditioned polynomials  
converge fast around zeros  
— as fast as a geometric series

# Smale's $\alpha$ -Theory

$$\alpha(g, x) := \beta(g, x) \gamma(g, x)$$

$$\beta(g, x) := \|D_x g^{-1} g(x)\| = \|x - N_g(x)\|$$

$$\gamma(g, x) := \max \left\{ 1, \sup_{K \geq 2} \left\| D_x^{g^{-1}} \frac{1}{K!} D_x^K g \right\|^{\frac{1}{K-1}} \right\}$$

Smale's  $\alpha$ -Theorem There is absolute  $\alpha_* > 0$ , s.t. if  $\alpha(g, x) < \alpha_*$ , then the Newton method at  $x$  converges quadratically.  
More concretely,  $\text{dist}(N_g^K(x), z(g)) = O(2^{-2^K})$

# Truncation Theorem (One version)

Let  $g \in \mathbb{R}[x_1, \dots, x_n]^n$ ,  $\delta \in \mathbb{N}$ ,  $x \in \mathbb{B}^n$  &

$$\tau(g, x; \delta) := \sup_{k \geq \delta+1} \left\| D_x^{g^{-1}} \frac{2^k}{k!} D_0^k g \right\|$$

Consider

$$g|_\delta(x) := \sum_{k=0}^{\delta} \frac{1}{k!} D_0^k g(x, \dots, x)$$

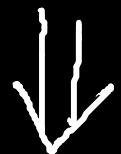
Then for

$$\delta - \log(\delta + 2) \geq \log \tau(g, x; \delta),$$

$$\alpha(g|_\delta, x) \leq \frac{2 \alpha(g, x) + 2^{1-\delta} g(g, x) \tau(g, x; \delta)}{(1 - 2^{-\delta} (\delta + 2) \tau(g, x; \delta))^2}$$

I.e.

approximate zero of  $\delta$  à la Smale



approximate zero of  $\delta_{IS}$  à la Smale

+ reverse & move ineqs.

# MOKOZ'S Lemma

W-Lemma: For  $g \in \mathbb{R}[x_0, x_1]^d$  and  $(x_0, x_1) \in \mathbb{S}^1$ ,

$$\left\| \frac{1}{k!} \frac{d^k}{dt^k} \Big|_{t=0} g((x_0, x_1) + t(x_1 - x_0)) \right\| \leq \sqrt{\binom{d}{k}} \|g\|_W$$

1-Lemma: For  $g \in \mathbb{R}[x]_{\leq d}$ ,  $a \in \mathbb{I}$

and  $\rho > 0$ , if

either  $2|a| < 1 - \rho$  or  $\rho < 1/2d$

then  $\left\| \frac{1}{k!} \frac{d^k}{dt^k} \Big|_{t=0} g(a + \rho t) \right\| \leq \frac{1}{2^k} \|g\|_1$

# Multivariate Moroz's 1-Lemma

Let  $g \in R[X_1, \dots, X_n]_{\leq D}^n$ ,  $a \in I^n$  &  $\rho \in (0, 1]^n$

Consider

$$g_{a, \rho} := \left( g_i(a + \rho X) / \|g_i\|_1 \right)$$

where  $P = \text{diag}(\rho)$ .

If for all  $i$ ,

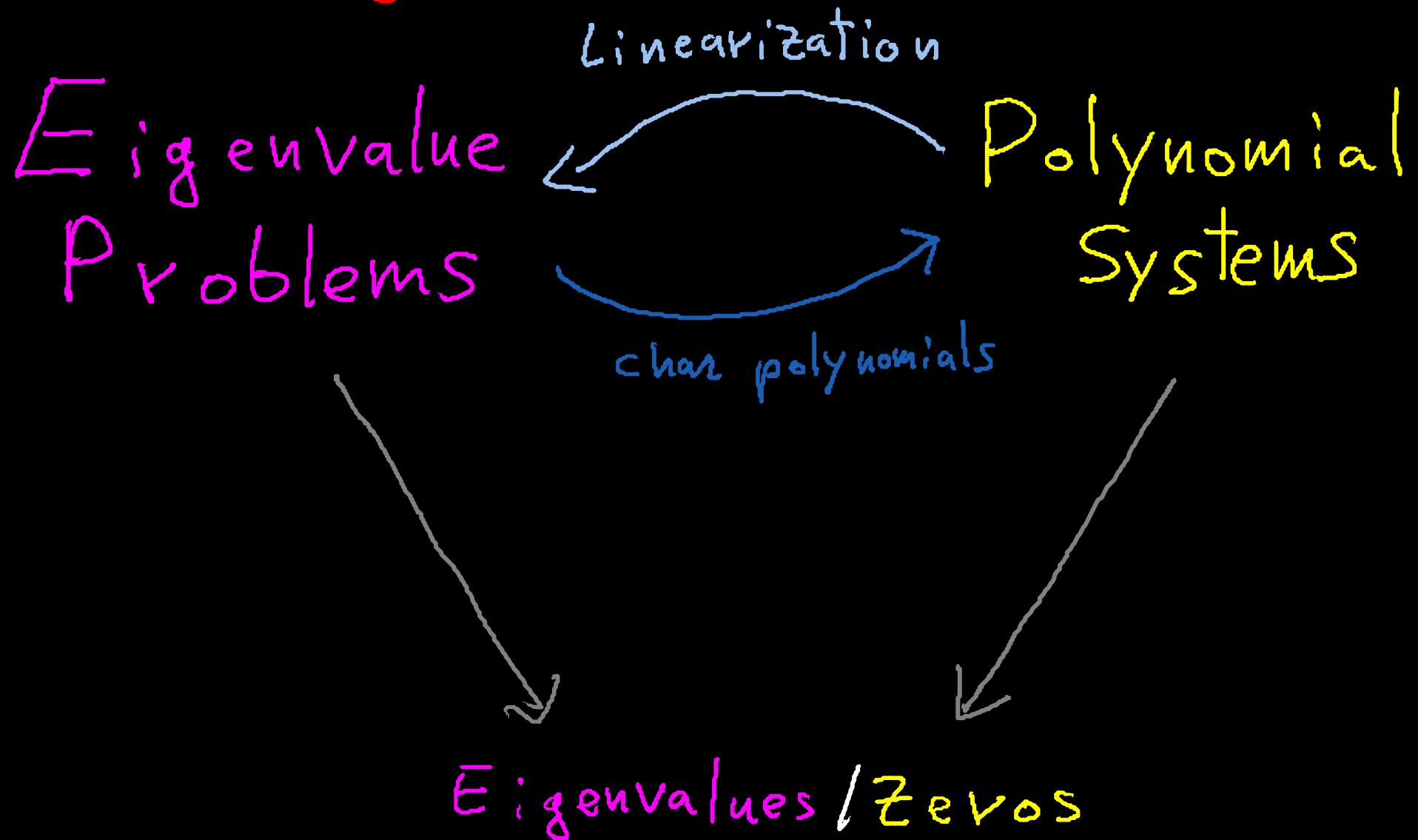
either  $|a_i| \leq 1 - \rho_i$  or  $\rho_i \leq \frac{1}{2D}$

then for all  $e$ ,

$$\left\| \frac{2^e}{e!} D_o^e g_{a, \rho} \right\|_{\infty, \infty} \leq \binom{D+n-1}{n-1} \leq \left( 1 + \frac{\ln(n-1)}{n-1} D \right)^{n-1}$$



# Triangle of competition



Numerical Analyst's Rule

NEVER USE  
CHARACTERISTIC  
POLYNOMIALS  
TO COMPUTE  
EIGENVALUES

# A Formalization For Hermitian Matrices

THM. Let  $A \in \text{Herm}_d$ . Then

$$\kappa_w(\chi_A) \geq 2^{\frac{\sqrt{d}}{\text{polylog}(d)}}$$

$$\& C_1(\chi_A) \geq 2^{\frac{d}{\text{polylog}(d)}}$$

I.e. characteristic polynomials  
of Hermitian matrices are  
badly conditioned.

Future  
Work

- Can we make all this into fast algorithms?
  - avoid condition estimation
- Generalize it beyond zero-dim systems
  - volume & Betti numbers

Thank You For your attention!

