

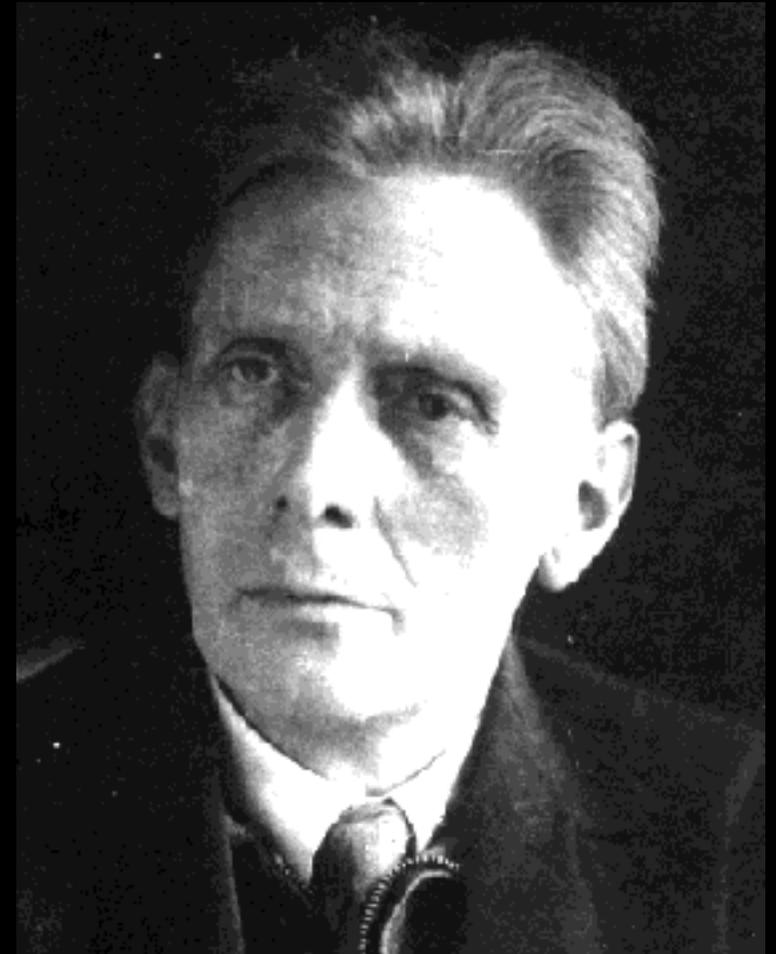
Today, 130 years ago,

the Soviet mathematician

Aleksandr KHINCHIN

was born

Known for his contributions
to probability theory



Some

Lower Bounds

on the

Reach

of an

Algebraic Variety

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ISSAC'24

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Talk supported by

grant 24-30

of the

Acheson J. Duncan

Fund

for the Advancement

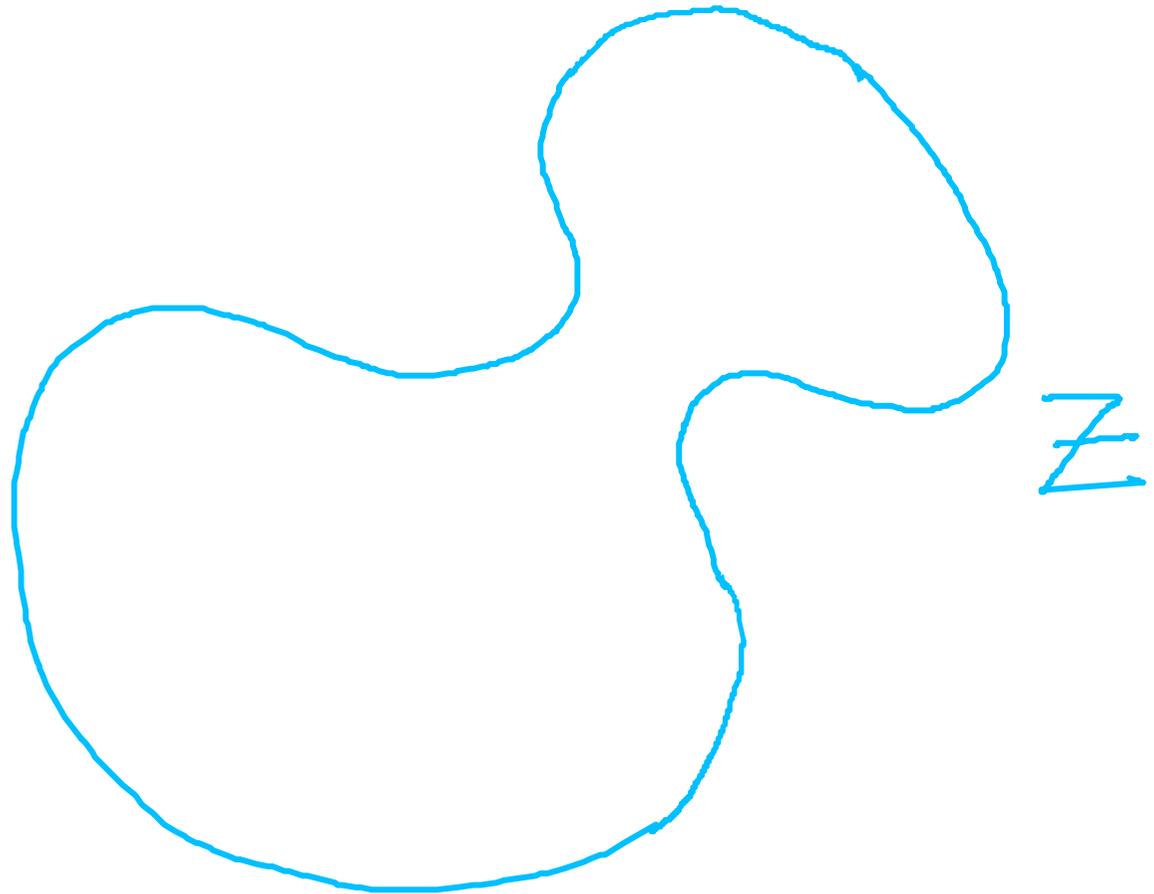
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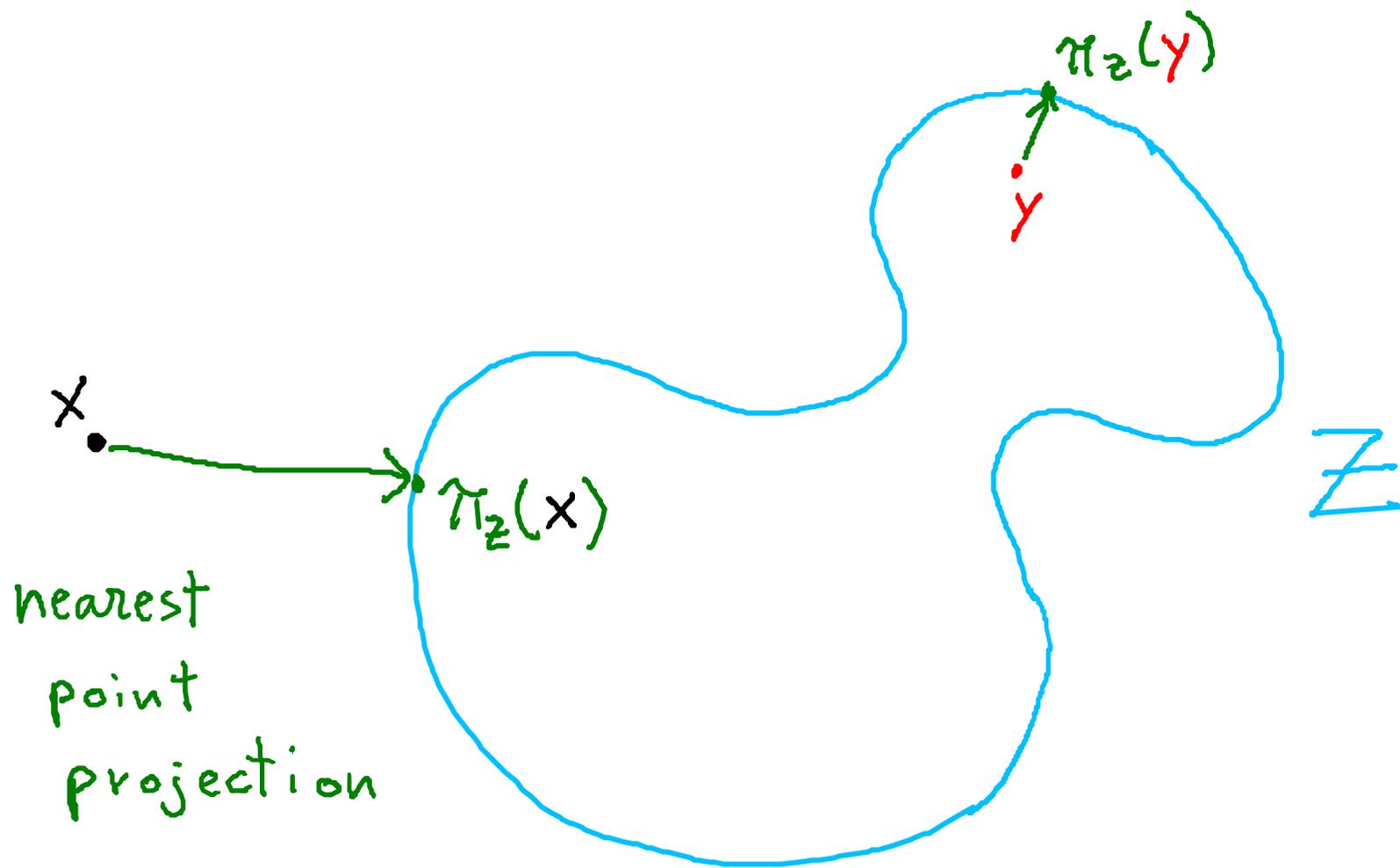
Research in Statistics

of

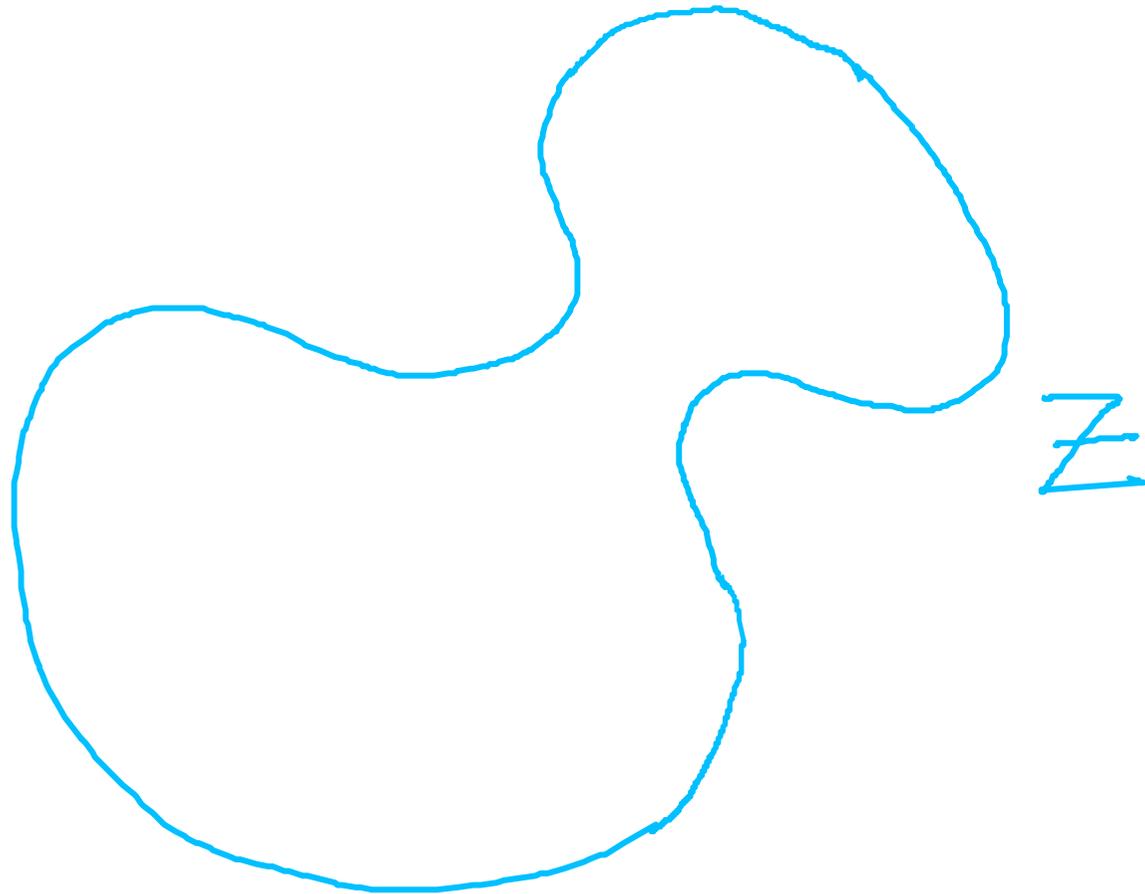
Soledad VILLAR

What is
the Reach?

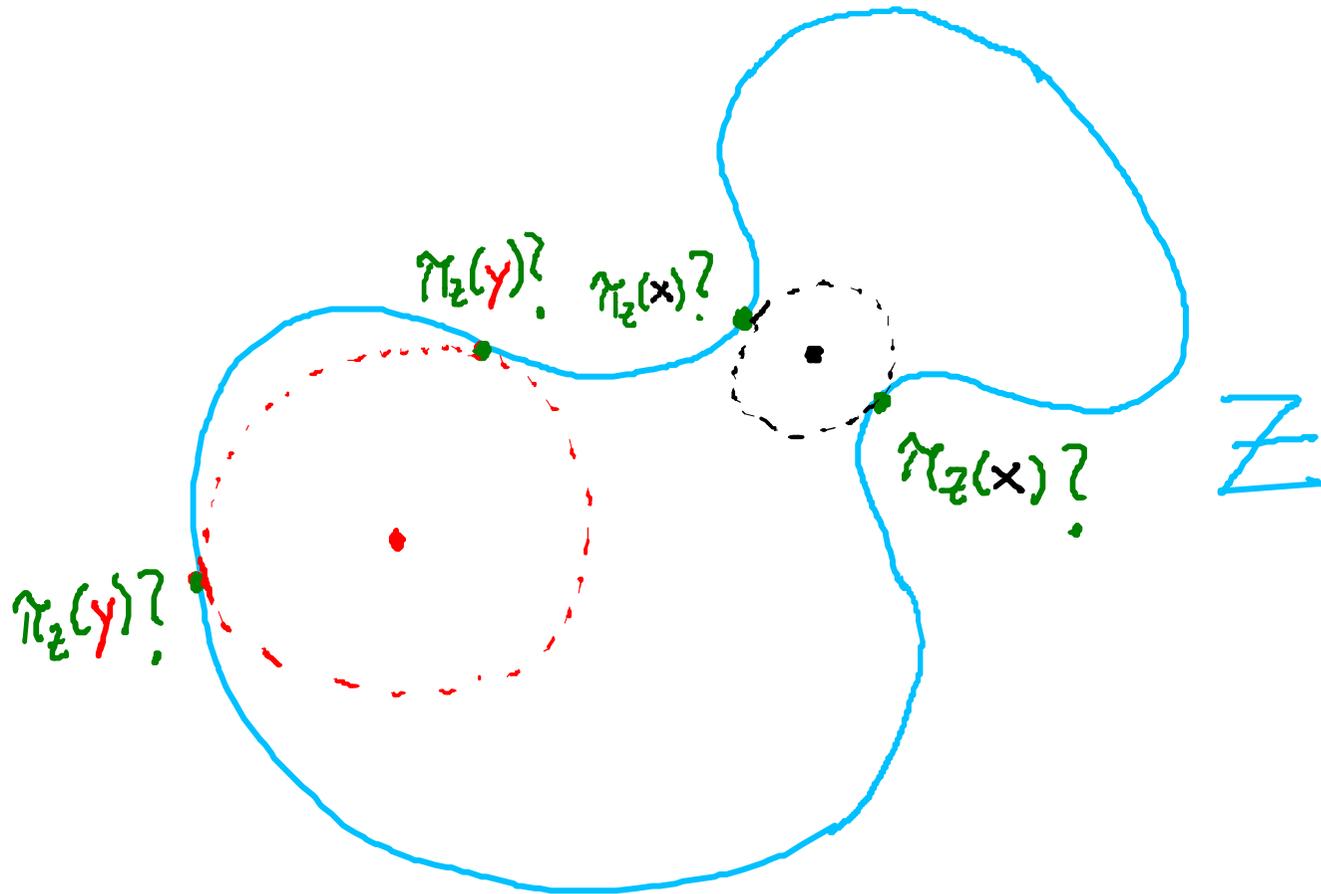




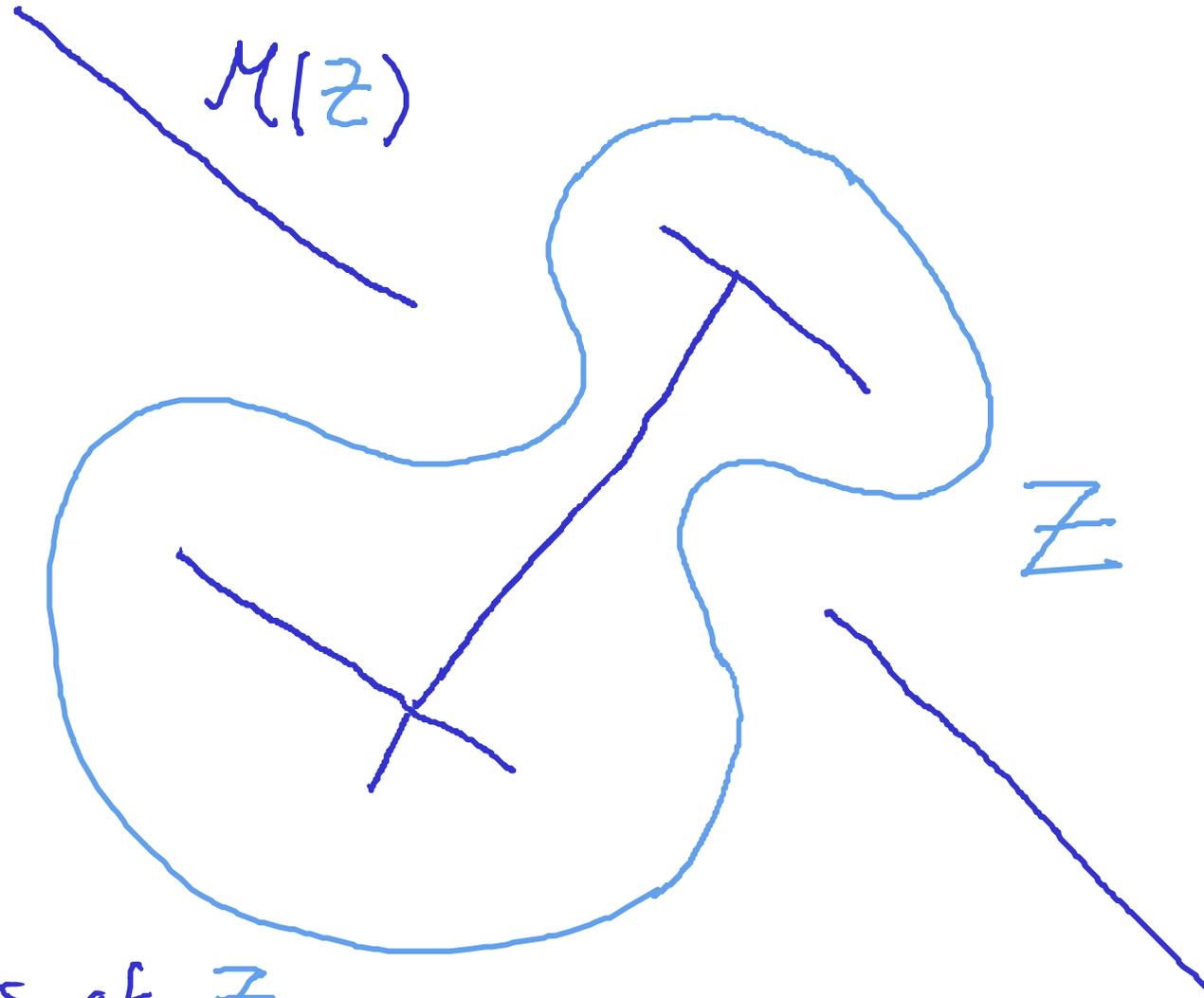
What if there is not only one nearest point?



What if there is not only one nearest point?



What if there is not only one nearest point?



Medial Axis of Z

$$M(Z) := \{x \mid \exists \tilde{g}, \tilde{g}' \in Z; \tilde{g} \neq \tilde{g}', \text{dist}(x, \tilde{g}) = \text{dist}(x, \tilde{g}') = \text{dist}(x, Z)\}$$

Local Reach:

$$\rho(z, g) := \text{dist}(g, \mathcal{M}(z))$$

Reach:

$$\rho(z) := \min_{g \in Z} \rho(z, g) = \text{dist}(z, \mathcal{M}(z))$$

And...

why do we care

about the reach?

TDA

Topological Data Analysis

Reach

is

the positive-dimensional

separation bound

Our Results

in a friendly version

Integer Worst Case:

$$R \in \mathbb{N} \quad \mathcal{Z} \in \mathbb{Z}[X_1, \dots, X_n]_{\leq D}^q \quad |\mathcal{Z}_{i,a}| \leq 2^\gamma$$

$$e_R(\mathcal{Z}) = 0 \quad (\mathcal{Z} \text{ singular})$$

OR

$$\log \frac{1}{e_R(\mathcal{Z})} \leq \mathcal{O}\left(n(2D)^{q+2n}\right) (\gamma + \log R + n \log D)$$

$$e_R(Z) := \min\{e(Z, \mathcal{G}) \mid \mathcal{G} \in Z, \max |g_i| \leq R\}$$

$$e(Z, \mathcal{G}) := \text{dist}(\mathcal{G}, \mathcal{M}(Z)) \quad \mathcal{M}(Z): \text{Medial Axis of } Z$$

Integer Probabilistic Case

$$R \in \mathbb{N} \quad F \in \mathbb{Z}[X_1, \dots, X_n]_{\leq D}^q$$

$$F_{i,a} \stackrel{\text{iid}}{\sim} \mathcal{U}(\mathbb{Z} \cap [-2^\tau, 2^\tau]) \leftarrow \text{Also for random bit polynomials}$$

$$\log \frac{1}{\mathcal{P}_R(Z(F))} \leq \mathcal{O}(\log R + n \log n + (q+n)n \log D) + s$$

$$\text{with prob.} \geq 1 - 2^{-s}$$

$$\text{For } s \leq \mathcal{O}(\tau) \text{ and } \tau = \Omega(\log R + n \log n + (q+n)n \log D)$$

$$\mathcal{P}_R(Z) := \min\{\mathcal{P}(Z, \mathfrak{g}) \mid \mathfrak{g} \in \mathbb{Z}, \max |g_i| \leq R\}$$

$$\mathcal{P}(Z, \mathfrak{g}) := \text{dist}(\mathfrak{g}, \mathcal{M}(Z)) \quad \mathcal{M}(Z): \text{Medial Axis of } Z$$

Continuous Probabilistic Case

$$R \in \mathbb{R}_+ \quad F \in \mathbb{R} [X_1, \dots, X_n]_{\leq D}^q$$

$F_{i,\alpha} \stackrel{\text{iid}}{\sim} \mathcal{U}([-1, 1])$ ← Also for
zintzo random polynomials!

$$\log \frac{1}{\mathcal{P}_R(Z(F))} \leq \mathcal{O}(\log R + n \log n + (q+n)n \log D) + s/2$$

with prob. $\geq 1 - 2^{-s}$

$$\mathcal{P}_R(Z) := \min \{ \mathcal{P}(Z, \mathfrak{g}) \mid \mathfrak{g} \in Z, \max |g_i| \leq R \}$$

$$\mathcal{P}(Z, \mathfrak{g}) := \text{dist}(\mathfrak{g}, \mathcal{M}(Z)) \quad \mathcal{M}(Z): \text{Medial Axis of } Z$$

Our Techniques

Federer's

Lower Bound

$Z \subseteq \mathbb{R}^m$ closed

$g \in Z$

$r, t > 0$

IF $Z \cap B(g, r) \subseteq B(g, r)$ closed submanifold

& for all $z, \tilde{z} \in Z \cap B(g, r)$,

$$\text{dist}(\tilde{z} - z, T_z Z) \leq \frac{\|\tilde{z} - z\|^2}{2t}$$

THEN:

$$e(z, g) \geq \min\{r, t\}$$

Improvement of Thm 3.3 of (Bürgisser, Cucker, Laird; 2019)

$$\rho(z(\xi), \xi) \geq \frac{1}{5 \gamma(\xi, \xi)}$$

↑ instead of 14

where

$$\gamma(\xi, \xi) := \sup_{k \geq 2} \left\| D_{\xi} \xi^{\dagger} \frac{1}{k!} D_x^k \xi \right\|^{\frac{1}{k-1}}$$

is Smale's γ .

Condition

Number

$$\|g\|_1 := \max_i \sum |g_{i,x}|$$

$$C(g, x) \propto \frac{\|g\|_1}{\text{dist}_1(g, \Sigma_x)}$$

$$\Sigma_x := \{g \mid x \in Z(g) \text{ singular}\}$$

Copernican turn:

We don't ask about the quality of the zero,
but about the quality of the system around the zero.

A new condition-based bound

$$C(\mathcal{Z}(g), g) \geq \frac{\max_i \{1, |g_i|\}}{\max\{D-2, C(g, g)\}}$$

$$C(\mathcal{L}, x) = \sup_v \frac{\|\mathcal{L}\|_1}{\|\mathcal{L}_{x,v}\mathcal{L}\|}$$

linear projection
polynomially depending on x
and linearly on v

Geometric

Functional Analysis

- Anti-concentration of linear projections

x anti-concentrated
+ ind. comp. $\Rightarrow Ax$ anti-concentrated

- Ball's smoothing

x disc.
 y cont.
well-chosen $\Rightarrow x+y$ cont. (+ details)
anti-concentrated

Thank You

for your Attention!