

Beyond
the CURSE

of INFINITE AVERAGE RUN-TIME

in NUMERICAL REAL ALGEBRAIC GEOMETRY,

can

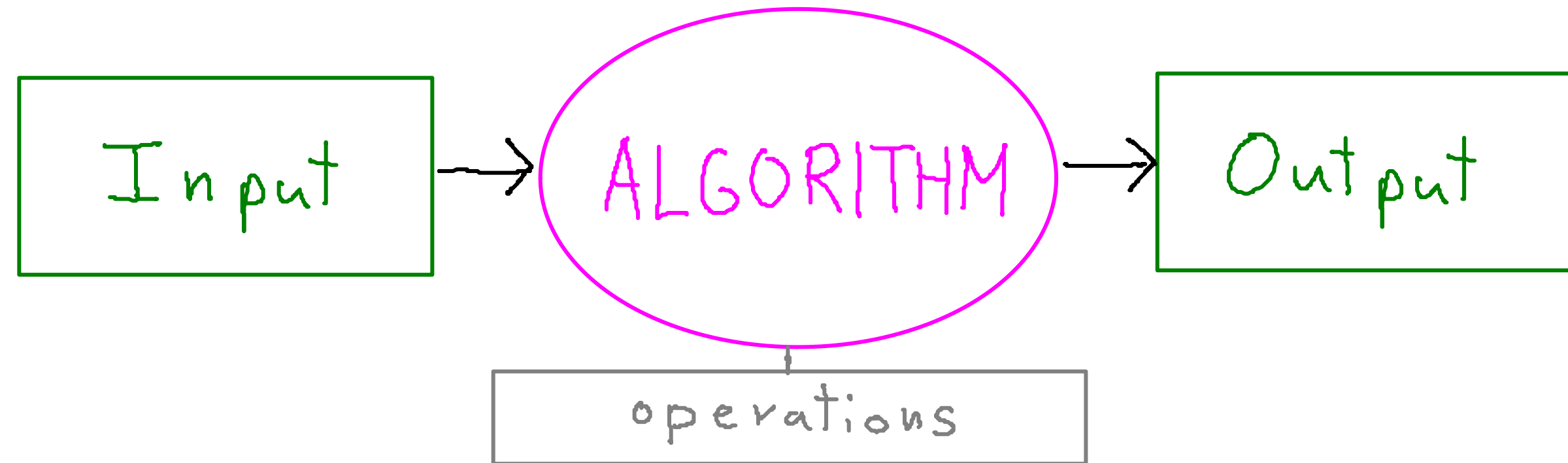
the Adaptive Grid Method

do the job?

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SIAM AG21

Complexity of (Traditional) Algorithms

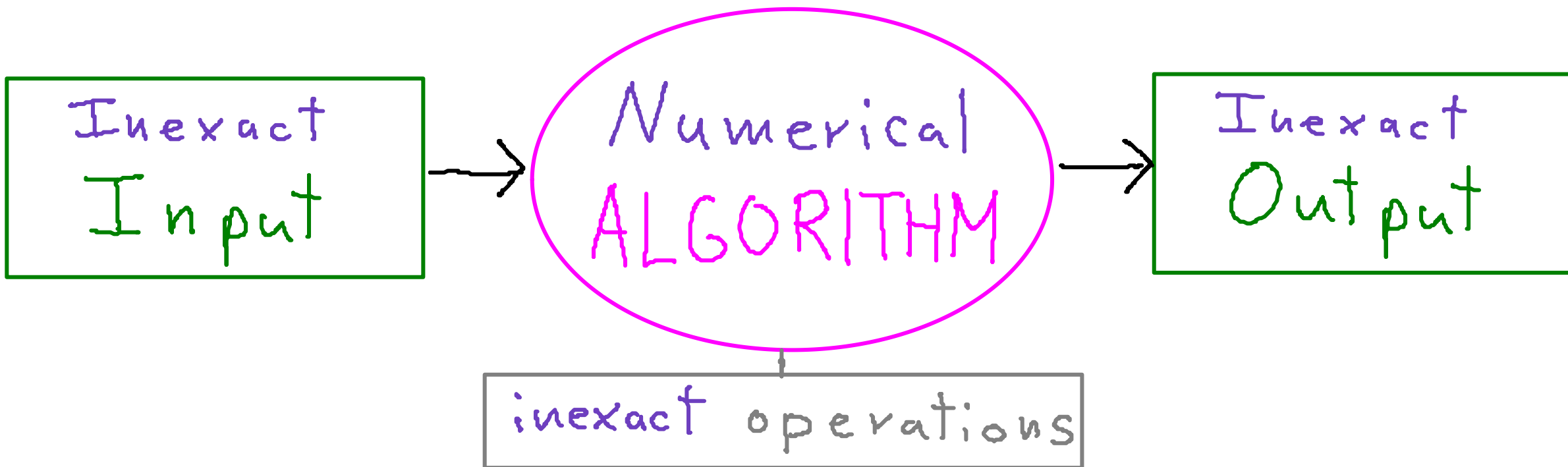


Worst-case form of complexity estimate:

$$\text{run-time}(\text{ALGORITHM}, \text{Input}) \leq f(\text{size}(\text{Input}))$$

⚠ sometimes **size** has several parameters
(e.g. #variables, degree...)

Complexity of Numerical Algorithms I



⚠ usual form of complexity fails!

ALL INPUTS OF THE SAME SIZE ARE EQUAL,
BUT SOME INPUTS ARE MORE EQUAL
THAN OTHERS

Complexity of Numerical Algorithms II

Condition-based complexity (Turing)
(Goldstine, von Neumann)


$\text{cond}(\text{Input})$: measures numerical sensitivity of Input

cond big \Rightarrow

Small variations of Input
 \rightarrow big variations of Output

cond small \Rightarrow

'big' variations of Input
 \rightarrow small variations of Output

 cond is a property of the computational problem,
not of the algorithm!

Condition-based form of complexity estimates

$\text{run-time}(\text{ALGORITHM}, \text{Input}) \leq f(\text{size}(\text{Input}), \text{cond}(\text{Input}))$

Complexity of Numerical Algorithms III

Probabilistic complexity I (Goldstine, von Neumann)
(Smale) (Demmel)

Can we have a complexity estimate
of a numerical algorithm only depending on size?

Yes, if we randomize the Input

How do we randomize the Input?

We choose the probability distribution
depending on the context!

Statistical complexity might have been a better name

Complexity of Numerical Algorithms II

Probabilistic complexity II (Goldstine, von Neumann)
(Smale) (Demmel)

Probabilistic form of complexity estimates

$$\mathbb{P}_{\text{input}}[\text{run-time}(\text{ALGORITHM}, \text{input}) \geq t] \leq \delta(s, t)$$

where $\text{size}(\text{input}) \leq s$

... and if we are lucky

$$\mathbb{E}_{\text{input}}[\text{run-time}(\text{ALGORITHM}, \text{input}) \geq t] \leq \delta(s)$$

Complexity of Numerical Algorithms

Smoothed complexity (Spielman, Teng)

Smoothed form of complexity estimates

$$\sup_{\substack{\text{Input} \\ \text{size}(\text{Input})=S}} \mathbb{P}_{\text{noise}} \left[\text{runtime}(\text{ALGORITHM}, \text{Input} + \sigma \text{noise}) \geq t \right] \leq f(S, t) / \sigma$$

... and if we are lucky

$$\sup_{\substack{\text{Input} \\ \text{size}(\text{Input})=S}} \mathbb{E}_{\text{noise}} \left[\text{runtime}(\text{ALGORITHM}, \text{Input} + \sigma \text{noise}) \right] \leq f(S) / \sigma$$

- As $\sigma \rightarrow \infty$, we recover the average form
- As $\sigma \rightarrow 0$, we recover the worst-case form

Computing the Homology of Algebraic Sets I

Input

$$\mathcal{g} \in \mathcal{H}_d[q]$$

q -tuple of homogeneous real polynomials in variables X_0, \dots, X_n

Output

$$H_*(\mathbb{Z}_p(\mathcal{g}))$$

Homology of projective real zero set

Size

$$N = \sum_{i=1}^q \binom{n+d_i}{n}$$

Total number of coefficients
(including the zero ones)

$$D = \max d_i$$

Maximum degree

Desired complexity

$$\text{poly}(q, D)^{\text{poly}(n)} N$$

Computing the Homology of Algebraic Sets II

Condition number (local version)

$$\kappa(\mathfrak{g}, x) := \frac{\|\mathfrak{g}\|_w}{\sqrt{\|\mathfrak{g}(x)\|^2 + \sigma_q(\Delta^{-1/2} D_x \mathfrak{g})}}$$

where

$$\|\mathfrak{g}\|_w := \sqrt{\sum_{i=1}^q \sum_{|\alpha|=d_i} \binom{d_i}{\alpha}^{-1/2} \mathfrak{g}_{i,\alpha}^2} \quad (\text{Weyl norm})$$

σ_q q th singular value

$$\Delta := \text{diag}(d_1, \dots, d_q)$$

$D_x \mathfrak{g}: T_x \mathbb{P}^n \rightarrow \mathbb{R}^q$ tangent map (upto signs!)

Computing the Homology of Algebraic Sets II

Condition number (global version)

$$\kappa(\mathfrak{g}) := \max_{x \in \mathbb{P}^n} \kappa(\mathfrak{g}, x)$$

Condition Number Theorem:

$$\kappa(\mathfrak{g}) = \frac{\|\mathfrak{g}\|_w}{\text{dist}_w(\mathfrak{g}, \Sigma)}$$

where Σ is the discriminant variety

Cor. IF $\frac{\|\tilde{\mathfrak{g}} - \mathfrak{g}\|_w}{\|\mathfrak{g}\|_w} \leq \frac{1}{\kappa(\mathfrak{g})}$, then $H_*(Z_{\mathbb{P}}(\mathfrak{g})) = H_*(Z_{\mathbb{P}}(\tilde{\mathfrak{g}}))$

↑
Topology does not change!

Computing the Homology of Algebraic Sets III

Grid Method
(A bird's view)

(Cucker, Krick, Malajovich, Wschebor)

(Cucker, Krick, Shub)

1. Cover with a sufficiently fine grid \mathbb{P}^n
2. Select points in the grid sufficiently near to $\mathbb{Z}_p(\mathcal{g})$ (meaning evaluation of \mathcal{g} 'small')
3. Use TDA to compute $H_*(\mathbb{Z}(\mathcal{g}))$ out of the selected points (the sample)

Advantages:

- Numerically stable
- Parallelizable

For semialgebraic: (Bürgisser, Cucker, Laires), (Bürgisser, Cucker, Tonelli-Cueto)

Computing the Homology of Algebraic Sets IV

Complexity of the Grid Method

Condition-based complexity estimate

$$\text{run-time}(\text{GRID}, \mathcal{g}) \leq (n D \chi(\mathcal{g}))^{O(n^2)}$$

Probabilistic complexity estimate

$$\mathbb{P}_F \left[\text{run-time}(\text{GRID } F) \geq (n D)^{O(n^2)} \frac{10^{n(n+1)}}{\epsilon} \right] \leq \frac{1}{\epsilon}$$

where $f_i = \sum \binom{d_i}{\alpha} f_{i,\alpha} X^\alpha$ with i.i.d. $f_{i,\alpha} \sim N(0,1)$

also for smoothed & more general distributions (Ergür, Paouris, Rojas)

... unfortunately

$$\mathbb{E}_F \left[\text{run-time}(\text{GRID } F) \right] = \infty$$

Computing the Homology of Algebraic Sets V

the Curse

$$\sum_F \kappa(F) = \infty$$

... so $\kappa(F)^{O(n^2)}$ is really non-finite

Obs. $\sum_F \log^e \kappa(F) < \infty$ for all $e \geq 1$

Are there numerical algorithms with condition-based complexity $(n \log \kappa(F))^{\text{poly}(n)}$?

This would be amazing!

Computing the Homology of Algebraic Sets VI

Adaptive Grid Method

- Like the grid method,
but having a **non-uniform** grid
- Local mesh around x controlled by $\kappa(\delta, x)$
- More details in my MEGA 21 talk
(only the zero dimensional case)

For obtaining sample (not homology),
the condition-based complexity is:

$$(nD)^{O(n^2)} \sum_{x \in \mathbb{P}^n} \kappa(\delta, x)^n$$

Computing the Homology of Algebraic Sets VII

where...

$$\sum_f \sum_{x \in \mathbb{P}^n} \chi(f, x)^n \leq (nD)^{O(n^3)}$$

is FINITE!

Unfortunately, postprocessing the sample brings expected complexity back to ∞ ,

can we escape this situation?

Better analysis?

New algorithms?

Only time will say...

Esmerikh
asho
zure avretagatik!