

Solving Polynomial Systems

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Outline of this Lecture

1. What does solving mean?
2. Numerical Polynomial Algebra
3. Numerical Algebraic Geometry
4. What is Symbolic-Numeric Computing?

What does solving mean?

Two vastly different views on polynomial systems.

Computer algebra views the polynomials in a system as generators of an ideal. A **Gröbner basis** allows to compute with the ideal and solve the system.

Numerical analysis views a polynomial system as a system of nonlinear equations. **Newton's method**, often with continuation, approximates the solutions.

Where and when do the algebra and analysis meet?

Ideals generated by Polynomial Systems

$P(\mathbf{x}) = \mathbf{0}$ polynomial system:

$$P = (p_1, p_2, \dots, p_N), p_k \in \mathbb{C}[\mathbf{x}], \mathbf{x} = (x_1, x_2, \dots, x_n).$$

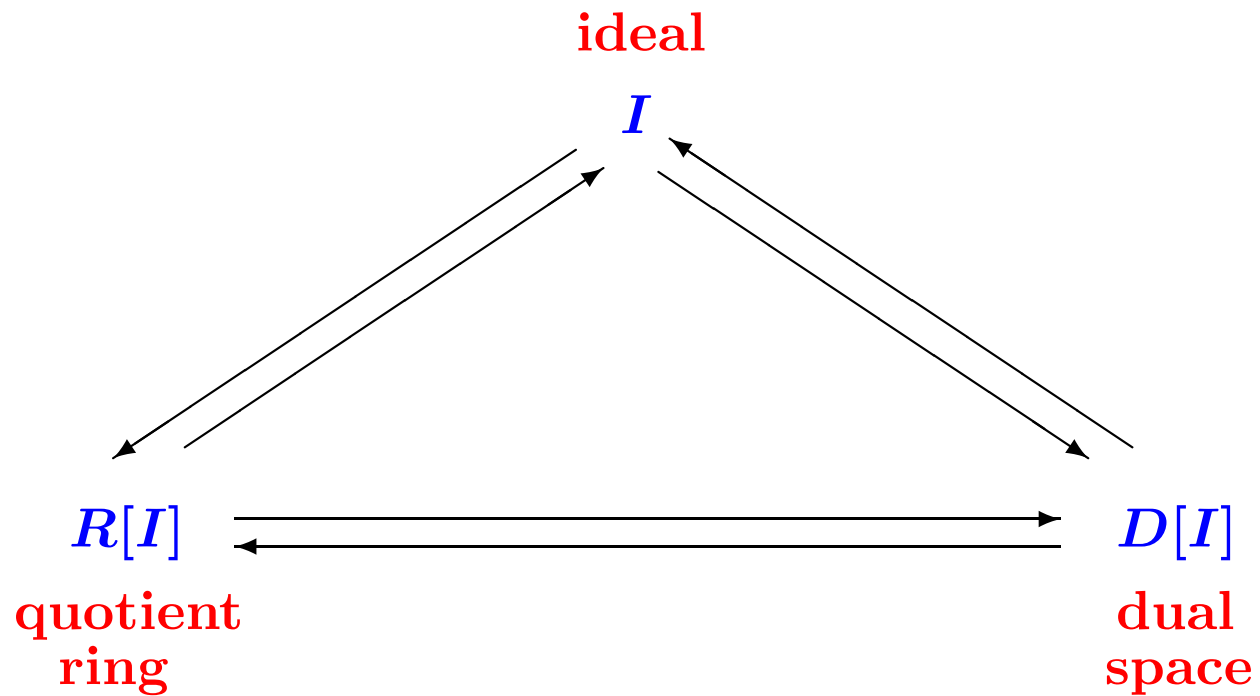
$I = \langle P \rangle$ ideal generated by P :

$$I = \{ c_1(\mathbf{x})p_1 + c_2(\mathbf{x})p_2 + \dots + c_N(\mathbf{x})p_N \mid c_k \in \mathbb{C}[\mathbf{x}] \}.$$

$Z[I]$ solution set:

$$Z[I] = \{ \mathbf{z} \in \mathbb{C}^n \mid p(\mathbf{z}) = 0, \forall p \in I \}.$$

A Commutative Diagram



Ideals and Quotient Rings

The quotient ring $R[I]$ of a ideal I allows to calculate modulo I :

$R[I] = \{ [p]_I \mid p \in \mathbb{C}[\mathbf{x}] \}$ is the residue class ring,

with $[p]_I$ the **residue class of p mod I** :

$$\begin{aligned} [p]_I &= \{ r \in \mathbb{C}[\mathbf{x}] \mid p - r \in I \} \\ &= \{ r \in \mathbb{C}[\mathbf{x}] \mid p(\mathbf{z}) = r(\mathbf{z}), \forall \mathbf{z} \in Z[I] \}. \end{aligned}$$

Theorem: **If $\#Z[I] = m < \infty$, then $\dim(R[I]) = m$.**

$I \rightarrow R[I]$: if $\dim(Z[I]) = 0$, the residue classes of Lagrange polynomials interpolating at $Z[I]$ give a basis for $R[I]$.

$R[I] \rightarrow I$: for some basis \mathbf{b} of $R[I]$: $A_k \mathbf{b} = x_k \mathbf{b}$, $k = 1, 2, \dots, n$, means $x_k \mathbf{b} = A_k \mathbf{b} \bmod I$, or $x_k \mathbf{b} - A_k \mathbf{b} = 0$ over $Z[I]$, and $x_k \mathbf{b} - A_k \mathbf{b} \in I$ leads to a border basis for I .

Example One

$$P(x, y) = \begin{cases} x^2 + 4xy + 4y^2 - 4 = 0 \\ 4x^2 - 4xy + y^2 - 4 = 0 \end{cases} \quad \begin{array}{l} \text{we expect } 2 \times 2 \\ = 4 \text{ solutions} \end{array}$$

$$\xrightarrow[\text{on coefficients}]{\text{row reduction}} \begin{cases} 15x^2 - 20xy - 12 = 0 \\ 20y^2 + 20xy - 12 = 0 \end{cases}$$

A natural **basis for $R[I]$** is $\mathbf{b} = (1, x, y, xy)$.

$$\mathbf{x}\mathbf{b} = x \begin{pmatrix} 1 \\ x \\ y \\ xy \end{pmatrix} = \begin{pmatrix} x \\ x^2 \\ xy \\ x^2y \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ \frac{12}{15} & 0 & 0 & \frac{20}{15} \\ 0 & 0 & 0 & 1 \\ 0 & \frac{48}{125} & \frac{36}{125} & 0 \end{pmatrix} \begin{pmatrix} 1 \\ x \\ y \\ xy \end{pmatrix}$$

Observe: $\mathbf{A}_x \mathbf{b} = \mathbf{x}\mathbf{b}$ is an eigenvalue problem.

Dual Spaces of Quotient Rings

The **dual space** of the quotient ring $R[I]$:

$$(R[I])^* = \{ l : R[I] \rightarrow \mathbb{C} : [p]_I \mapsto l(p) := l(r), r \in R[I], p - r \in I \}.$$

The dual of I : $D[I] = \{ l : \mathbb{C}[\mathbf{x}] \rightarrow \mathbb{C} : l(p) = 0, \forall p \in I \}$. $I \rightarrow D[I]$

Theorem: $D[I] = (R[I])^*$ and $R[I] = (D[I])^*$.

The ideal of the dual:

$$D[I] \rightarrow I$$

$$I[D[I]] = \ker(D[I]) = \{ p \in \mathbb{C}[\mathbf{x}] : l(p) = 0, \forall l \in D[I] \} = I.$$

Note: more satisfactory than $I[Z[I]] = \sqrt{I}$.

The dual defines the multiplicity structure of a multiple zero.

Multiplicity of an Isolated Zero

An isolated zero of multiplicity m occurs in numerical analysis as a cluster of m (ill-conditioned) regular zeros.

Problem: geometrical significance for overdetermined systems?
→ perturbed overdetermined system has no zeros!

Analogy with Univariate Case: z_0 is m -fold zero of $f(x) = 0$:

$$\underbrace{f(z_0) = 0, \frac{\partial f}{\partial x}(z_0) = 0, \frac{\partial^2 f}{\partial x^2}(z_0) = 0, \dots, \frac{\partial^{m-1} f}{\partial x^{m-1}}(z_0) = 0}_{m} .$$

m = number of linearly independent polynomials annihilating z_0

The dual space D_0 at z_0 is spanned by m linear independent differentiation functionals annihilating z_0 .

D_0 is the multiplicity structure of the m -fold zero z_0 .

Example Two

Consider

$$f(x, y) = \begin{cases} x^2 = 0 \\ xy = 0 \\ y^2 = 0 \end{cases} \quad \mathbf{z}_0 = (0, 0).$$

The **multiplicity of \mathbf{z}_0 is 3** because

$$D_0 = \text{span}\{\partial_{00}[\mathbf{z}_0], \partial_{10}[\mathbf{z}_0], \partial_{01}[\mathbf{z}_0]\}$$

with

$$\partial_{ij}[\mathbf{z}_0] = \frac{1}{i!j!} \frac{\partial^{i+j}}{\partial x^i \partial y^j} f(\mathbf{z}_0).$$

Solving means to compute \mathbf{z}_0 and D_0 .

Newton and Continuation Methods

Consider a family of systems $h_k(x(t), y(t), t) = 0$, $k = 1, 2$.

By $\frac{\partial}{\partial t}$ on h_k : $\frac{\partial h_k}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial h_k}{\partial y} \frac{\partial y}{\partial t} + \frac{\partial h_k}{\partial t} \frac{\partial t}{\partial t} = 0$, $k = 1, 2$.

Set $\Delta x := \frac{\partial x}{\partial t}$, $\Delta y := \frac{\partial y}{\partial t}$, and $\frac{\partial t}{\partial t} = 1$.

Increment $t := t + \Delta t$

Solve
$$\begin{bmatrix} \frac{\partial h_1}{\partial x} & \frac{\partial h_1}{\partial y} \\ \frac{\partial h_2}{\partial x} & \frac{\partial h_2}{\partial y} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} = - \begin{bmatrix} \frac{\partial h_1}{\partial t} \\ \frac{\partial h_2}{\partial t} \end{bmatrix} \quad (\text{Newton})$$

Update
$$\begin{cases} x := x + \Delta x \\ y := y + \Delta y \end{cases}$$

fails when the Jacobian matrix is singular!

An Application: cyclic 9-roots

The system

$$f(\mathbf{x}) = \begin{cases} f_i = \sum_{j=0}^8 \prod_{k=1}^i x_{(k+j) \bmod 9} = 0, & i = 1, 2, \dots, 8 \\ f_9 = x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_8 - 1 = 0 \end{cases}$$

has 333×18 **isolated regular** zeros, 164 **isolated 4-fold** zeros, and 6 **cubic 2-dimensional** irreducible solution components.

Newton's method with 64 decimal places, tolerance is 10^{-60} :

regular : 4 iterations (**quadratic convergence**)

4-fold : 79 iterations (**> 1 step for one correct decimal place**)

about 20 times slower to reach same magnitude of residual ...

Newton's method for Overdetermined Systems

Singular Value Decomposition of N -by- n Jacobian matrix J_f :

$$J_f = U\Sigma V^T, \quad U \text{ and } V \text{ are orthogonal: } U^T U = I_N, V^T V = I_n,$$

and singular values $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n$ as the only nonzero elements on the diagonal of the N -by- n matrix Σ ($N > n$).

Moore-Penrose inverse: $J_f^+ = V\Sigma^+U^T$, with $\frac{1}{\sigma_1}, \frac{1}{\sigma_2}, \dots, \frac{1}{\sigma_n}$ on the diagonal of the n -by- N matrix Σ^+ .

Then $\Delta \mathbf{z} = -J_f(\mathbf{z})^+ f(\mathbf{z})$ is the least squares solution.

The **condition number** $\text{cond}(J_f(\mathbf{z})) = \frac{\sigma_1}{\sigma_n}$.

$$\text{Rank}(J_f(\mathbf{z})) = R \iff \Sigma = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_R, 0, \dots, 0).$$

At a **multiple root** \mathbf{z}_0 : $\text{Rank}(J_f(\mathbf{z}_0)) = R < n$.

Close to \mathbf{z}_0 , $\mathbf{z} \approx \mathbf{z}_0$: $\sigma_{R+1} \approx 0$.

Newton with Deflation – Example Two revisited

$$f(x, y) = \begin{cases} x^2 = 0 \\ xy = 0 \\ y^2 = 0 \end{cases} \quad J_f(x, y) = \begin{bmatrix} 2x & 0 \\ 1 & 1 \\ 0 & 2y \end{bmatrix} \quad \begin{array}{l} \mathbf{z}_0 = (0, 0), m = 3 \\ \text{Rank}(J_f(\mathbf{z}_0)) = 1 \end{array}$$

A nontrivial linear combination of the columns of $J_f(\mathbf{z}_0)$ is zero.

$$G(x, y, \lambda_1, \lambda_2) = \begin{cases} f(x, y) = 0 \\ \begin{bmatrix} 2x & 0 \\ 1 & 1 \\ 0 & 2y \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \\ c_1 \lambda_1 + c_2 \lambda_2 = 1, \quad \text{random } c_1, c_2 \in \mathbb{C} \end{cases}$$

The system $G(x, y, \lambda_1, \lambda_2) = 0$ has $(0, 0, \lambda_1^*, \lambda_2^*)$ as **regular** zero!

The Deflation Operator Dfl

Suppose $\text{Rank}(J_f(\mathbf{z}_0)) = R$ for \mathbf{z}_0 an isolated zero of $f(\mathbf{x}) = 0$.

Choose $\mathbf{h} \in \mathbb{C}^{R+1}$ and $B \in \mathbb{C}^{n \times (R+1)}$ at random.

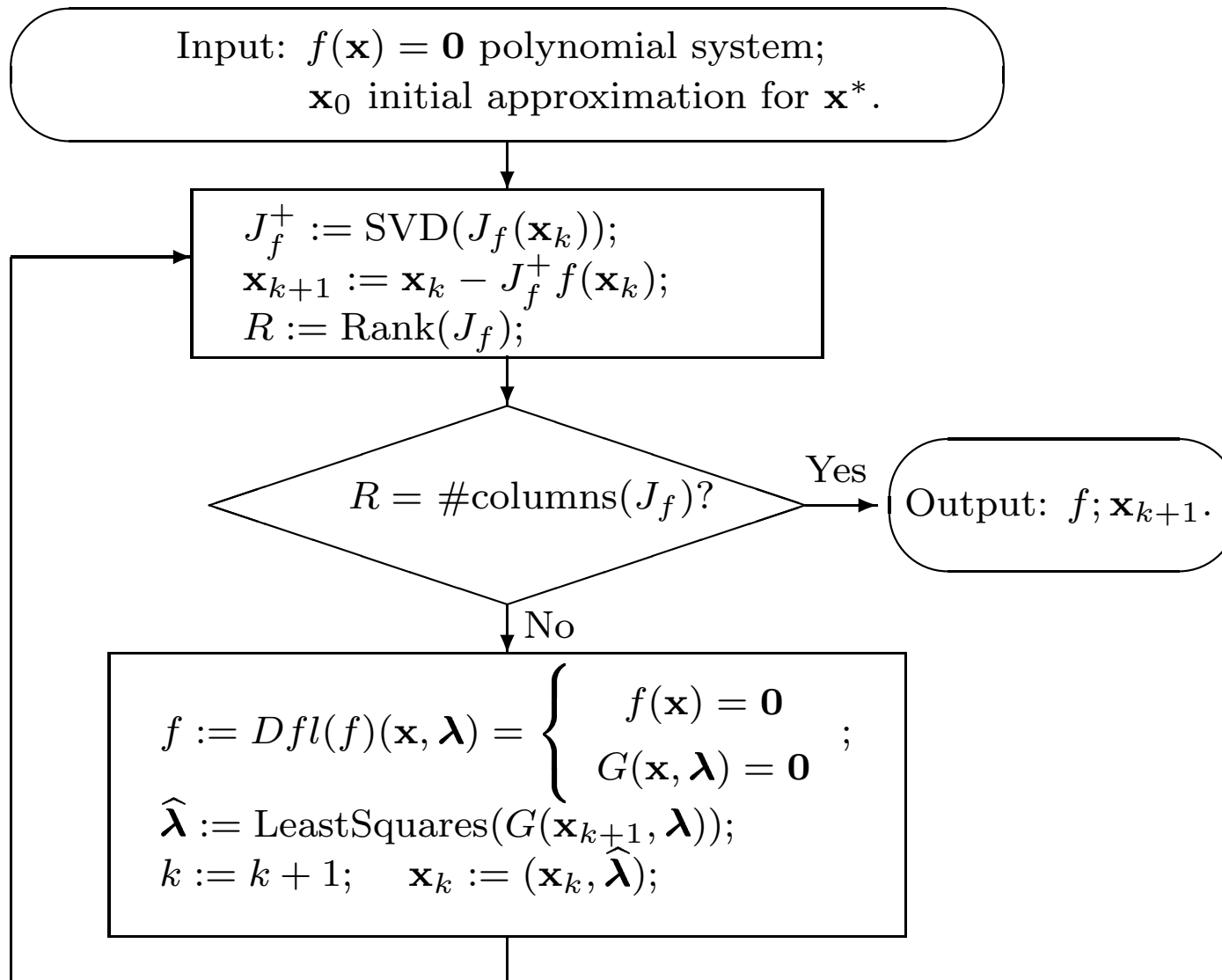
Introduce $R + 1$ new multiplier variables $\boldsymbol{\lambda} = (\lambda_1, \lambda_2, \dots, \lambda_{R+1})$.

$$Dfl(f)(\mathbf{x}, \boldsymbol{\lambda}) := \begin{cases} f(\mathbf{x}) & = \mathbf{0} \\ J_f(\mathbf{x})B\boldsymbol{\lambda} & = \mathbf{0} \\ \mathbf{h}\boldsymbol{\lambda} & = 1 \end{cases}$$

Theorem (Anton Leykin, JV, Ailing Zhao):

The number of deflations needed to restore the quadratic convergence of Newton's method converging to an isolated solution is strictly less than the multiplicity.

Newton's method with Deflation



cyclic 9-roots revisited

Recall:

$$f(\mathbf{x}) = \begin{cases} f_i = \sum_{j=0}^8 \prod_{k=1}^i x_{(k+j) \bmod 9} = 0, & i = 1, 2, \dots, 8 \\ f_9 = x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_8 - 1 = 0 \end{cases}$$

has 164 solutions of multiplicity 4.

One deflation suffices to restore quadratic convergence.

The **condition number** drops from 1.8E+9 to 5.6E+2.

→ **deflation reconditions the system**

Computing the Multiplicity Structure

following B.H. Dayton and Z. Zeng

Looking for differentiation functionals $d[\mathbf{z}_0] = \sum_{\mathbf{a}} c_{\mathbf{a}} \partial_{\mathbf{a}}[\mathbf{z}_0]$,

$$\text{with } \partial_{\mathbf{a}}[\mathbf{z}_0](p) = \frac{1}{a_1! a_2! \cdots a_n!} \left(\frac{\partial^{a_1 + a_2 + \cdots + a_n}}{\partial x_1^{a_1} \partial x_2^{a_2} \cdots \partial x_n^{a_n}} p \right) (\mathbf{z}_0).$$

Membership criterium for $d[\mathbf{z}_0]$:

$$d[\mathbf{z}_0] \in D_0 \Leftrightarrow d[\mathbf{z}_0](p f_i) = \mathbf{0}, \forall p \in \mathbb{C}[\mathbf{x}], i = 1, 2, \dots, N.$$

To turning this criterium into an **algorithm**, observe:

1. since $d[\mathbf{z}_0]$ is linear, restrict p to $\mathbf{x}^{\mathbf{k}} = x_1^{k_1} x_2^{k_2} \cdots x_n^{k_n}$; and
2. limit degrees $k_1 + k_2 + \cdots + k_n \leq a_1 + a_2 + \cdots + a_n$,
as $\mathbf{z}_0 = \mathbf{0}$ vanishes trivially if not annihilated by $\partial_{\mathbf{a}}$.

Computing the Multiplicity Structure – An Example

$$f_1 = x_1 - x_2 + x_1^2, \quad f_2 = x_1 - x_2 + x_2^2$$

following B.H. Dayton and Z. Zeng

		$\overbrace{\partial_{00}}^{ a =0}$	$\overbrace{\partial_{10} \quad \partial_{01}}^{ a =1}$		$\overbrace{\partial_{20} \quad \partial_{11} \quad \partial_{02}}^{ a =2}$			$\overbrace{\partial_{30} \quad \partial_{21} \quad \partial_{12} \quad \partial_{03}}^{ a =3}$			
S_1	f_1	0	1	-1	1	0	0	0	0	0	0
	f_2	0	1	-1	0	0	1	0	0	0	0
S_2	$x_1 f_1$	0	0	0	1	-1	0	1	0	0	0
	$x_1 f_2$	0	0	0	1	-1	0	0	0	1	0
	$x_2 f_1$	0	0	0	0	1	-1	0	1	0	0
	$x_2 f_2$	0	0	0	0	1	-1	0	0	0	1
	$x_1^2 f_1$	0	0	0	0	0	0	1	-1	0	0
	$x_1^2 f_2$	0	0	0	0	0	0	1	-1	0	0
	$x_1 x_2 f_1$	0	0	0	0	0	0	0	1	-1	0
	$x_1 x_2 f_2$	0	0	0	0	0	0	0	1	-1	0
S_3	$x_2^2 f_1$	0	0	0	0	0	0	0	0	1	-1
	$x_2^2 f_2$	0	0	0	0	0	0	0	0	1	-1

Nullity(S_2) = Nullity(S_3) \Rightarrow stop algorithm

$D_0 = \text{span}\{ \partial_{00}, \partial_{10} + \partial_{01}, -\partial_{10} + \partial_{20} + \partial_{11} + \partial_{02} \} \Rightarrow \text{multiplicity} = 3$

cyclic 9-roots once more

Recall:

$$f(\mathbf{x}) = \begin{cases} f_i = \sum_{j=0}^8 \prod_{k=1}^i x_{(k+j) \bmod 9} = 0, & i = 1, 2, \dots, 8 \\ f_9 = x_0 x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_8 - 1 = 0 \end{cases}$$

has 164 solutions of multiplicity 4.

Running the algorithm of Dayton and Zeng:

$$H[1] = 1, H[2] = 2, H[3] = 1, H[4] = 0,$$

$$\text{with } H[i] = \text{Nullity}(S_i) - \text{Nullity}(S_{i-1}), i > 0,$$

so we compute the multiplicity as 4.

What is Symbolic-Numeric Computing?

A puristic point of view:

Computer algebra rewrites the problem, producing “easier” equations of the ideal, but **“easier” \neq numerically better**.

Numerical analysis produces approximate numbers for a fixed system of equations, but **many problems are “ill-posed”**.

The synergistic approach

Symbolic-Numeric Computing rewrites an ill-conditioned numerical problem into a well-conditioned formulation.

works very well in Newton’s method with deflation

Chronological List of References

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