

*Solving Nonlinear Polynomial Systems via  
Symbolic-Numeric Elimination Method*

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Joint work with Greg Reid and Xiaoli Wu

# Zero Dimensional Polynomial System Solving

Consider a polynomial system  $F \in \mathbb{C}[x_1, \dots, x_s]$  of degree  $d$ ,

$$F : \begin{cases} f_1(x_1, \dots, x_s) = 0, \\ f_2(x_1, \dots, x_s) = 0, \\ \vdots \\ f_t(x_1, \dots, x_s) = 0. \end{cases}$$

We are going to find **all** the common solutions of  $F$ .

- Symbolic Algorithms: Gröbner bases, Characteristic sets, Involutive bases, H-bases, Border bases...

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- Numeric Algorithms: Newton's method, Homotopy continuation, Optimization methods...
- Symbolic-Numeric Hybrid Approaches: Gröbner bases, Involutive system, Border bases, Resultant...

# Matrix Eigenproblems

$F$  can be written in terms of its coefficient matrix  $M_d^{(0)}$  as

$$M_d^{(0)} \cdot \begin{pmatrix} x_1^d \\ x_1^{d-1} x_2 \\ \vdots \\ x_s^2 \\ x_1 \\ \vdots \\ x_s \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{pmatrix}$$

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**Remark:**  $[\xi_1, \xi_2, \dots, \xi_s]$  is a **solutions** of the polynomial system  $F$   
 $\iff [\xi_1^d, \xi_1^{d-1} \xi_2, \dots, \xi_s^2, \xi_1, \dots, \xi_s, 1]^T$  is a **null vector** of the  
 coefficient matrix  $M_d^{(0)}$ .

# Translates $F$ into a System of PDEs $R$

The bijection

$$\phi : x_i \leftrightarrow \frac{\partial}{\partial x_i}, \quad 1 \leq i \leq s,$$

$\phi$  maps the system  $F$  to an equivalent system of linear PDES  $R$ :

$$M_d^{(0)} \cdot \begin{pmatrix} \frac{\partial^d u}{\partial x_1^d} \\ \frac{\partial^d u}{\partial x_1^{d-1} \partial x_2} \\ \vdots \\ \frac{\partial^2 u}{\partial x_s^2} \\ \frac{\partial u}{\partial x_1} \\ \vdots \\ \frac{\partial u}{\partial x_s} \\ u \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{pmatrix}$$



## Why Differential Equations? [Sturmfels'02]

- We do not lose any information by doing so.

**Remark:** A vector  $\xi = [\xi_1, \xi_2, \dots, \xi_s] \in \mathbb{C}^s$  is a solution to polynomial system  $F$  if and only if the exponential function  $\exp(\xi \cdot x) = \exp(\xi_1 x_1 + \dots + \xi_s x_s)$  is a solution of the differential equations  $R$ .

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- PDE formulation reveals more information than the polynomial formulation.

## Example 1 [Sturmfels'02]

Consider the system of three polynomial equations

$$x^3 = yz, \quad y^3 = xz, \quad z^3 = xy.$$

We translate them to the three differential equations

$$\frac{\partial^3 u}{\partial x^3} = \frac{\partial^2 u}{\partial y \partial z}, \quad \frac{\partial^3 u}{\partial y^3} = \frac{\partial^2 u}{\partial x \partial z}, \quad \frac{\partial^3 u}{\partial z^3} = \frac{\partial^2 u}{\partial x \partial y}.$$

## Example 1 (continued)

A solution basis for the PDEs is given by

$$\begin{aligned}
 & \exp(x + y + z), \exp(x - y - z), \exp(y - x - z), \exp(z - x - y), \\
 & \exp(x + iy - iz), \exp(x - iy + iz), \exp(y + ix - iz), \exp(y - ix + iz), \\
 & \exp(z + ix - iy), \exp(z - ix + iy), \exp(-x + iy + iz), \exp(-x - iy - iz), \\
 & \exp(-y + ix + iz), \exp(-y - ix - iz), \exp(-z + iy + ix), \exp(-z - iy - ix), \\
 & 1, x, y, z, z^2, y^2, x^2, x^3 + 6yz, y^3 + 6xz, z^3 + 6xy, x^4 + y^4 + z^4 + 24xyz
 \end{aligned}$$

- The system has 17 distinct complex zeros, 5 are real.

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 & \exp(z + ix - iy), \exp(z - ix + iy), \exp(-x + iy + iz), \exp(-x - iy - iz), \\
 & \exp(-y + ix + iz), \exp(-y - ix - iz), \exp(-z + iy + ix), \exp(-z - iy - ix), \\
 & 1, x, y, z, z^2, y^2, x^2, x^3 + 6yz, y^3 + 6xz, z^3 + 6xy, x^4 + y^4 + z^4 + 24xyz
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- The multiplicity of the origin  $(0, 0, 0)$  is eleven.

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 & \exp(-y + ix + iz), \exp(-y - ix - iz), \exp(-z + iy + ix), \exp(-z - iy - ix), \\
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 \end{aligned}$$

- The system has 17 distinct complex zeros, 5 are real.
- The multiplicity of the origin  $(0, 0, 0)$  is eleven.
- Polynomial solutions gotten from  $x^4 + y^4 + z^4 + 24xyz$  by taking successive derivatives describe the multiplicity structure of the polynomial system at  $(0, 0, 0)$ .

# Symbolic-Numeric Completion of PDEs

We study the linear mapping:

$$R : v \mapsto M_d^{(0)} v$$

here  $v = [u_d, u_{d-1}, \dots, u_1, u]^T$ , and  $u_j$  denotes the formal jet coordinates corresponding to derivatives of order exactly  $j$ .

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Jet space approaches study the jet variety

$$V(R) = \left\{ (u, u, \dots, u, u) \in J^d : R(u, u, \dots, u, u) = 0 \right\}$$

where  $J^d \approx C^{s_d}$  is a jet space of order  $d$  and  $s_d = \binom{s+d}{d}$ .



## Symbolic Prolongation

A single **prolongation** of a system  $R$  is defined as:

$$DR := \{w \in J^{d+1} : R(w) = 0, \quad D_{x_1}R(w) = 0, \dots, D_{x_s}R(w) = 0\}$$

The prolonged system  $DR$  has order  $d + 1$ ,

$$M_d^{(1)} \cdot \begin{pmatrix} u \\ d+1 \\ u \\ d \\ \vdots \\ u \\ 1 \\ u \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 0 \end{pmatrix}$$

**Remark:**  $M_d^{(1)}$  is the coeff. matrix of the prolonged system

$$F^{(1)} = F \cup x_1 F \cup \dots \cup x_s F$$

# Geometric Projection

A single geometric **projection** is defined as

$$\pi(R) := \left\{ \left( \begin{array}{c} u \\ \dots \\ u \\ \dots \\ u, u \end{array} \right) \in J^{d-1} : R \left( \begin{array}{c} u \\ \dots \\ u \\ \dots \\ u, u \end{array} \right) = 0 \right\}$$

**Remark:** For polynomial system, the projection is equivalent to eliminating the monomials of the highest degree  $d$ .

- **Symbolic elimination** method using Gröbner basis algorithms or Ritt-Wu's characteristic algorithms.

**Remark:** variables have to be **well ordered**.

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- **Symbolic elimination** method using Gröbner basis algorithms or Ritt-Wu's characteristic algorithms.

**Remark:** variables have to be **well ordered**.

- **Numerical projection** via singular value decomposition.

**Remark:** variables are **partially ordered by total degrees**.

# Numeric Projection via SVD

For a chosen tolerance  $\tau$ :

- Compute SVD of  $M_d^{(0)}$ , obtain a **basis** for the **null space** of  $R$  and  $\dim R = \dim \text{Nullspace}(M_d^{(0)})$ ;

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- Estimate  $\dim \pi(R)$  by applying SVD to the **spanning set** of  $\pi(R)$ .
- Proceeding in the same way, we can estimate  $\dim \pi^\ell(D^k R)$ .

## Symbol Involutive

The **Symbol matrix** of PDEs is Jacobian matrix of the system w.r.t. its **highest order** jet coordinates,

$$\dim \left( \text{Symbol } \pi^\ell(D^k R) \right) = \dim \pi^\ell(D^k R) - \dim \pi^{\ell+1}(D^k R)$$

**Remark:** In case of polynomials, the **Symbol matrix** is the submatrix of the coefficient matrix of the system corresponding to **highest degree** monomials.

**Theorem 1.** [Seiler 2002] For *finite* type PDEs (i.e., *zero dimensional* polynomial systems), the **Symbol** of  $\pi^\ell(D^k R)$  is *involutive* if

$$\dim \pi^\ell(D^k R) = \dim \pi^{\ell+1}(D^k R)$$



## Involutive System

The system  $R = 0$  is said to be **involutive** at prolonged order  $k$  and projected order  $l$ , if  $\pi^l(D^k(R))$  satisfies:

$$\dim \pi^l(D^k R) = \dim \pi^{\ell+1}(D^{k+1} R)$$

and the Symbol of  $\pi^l(D^k R)$  is involutive.

**Theorem 2.** *[Cartan-Kuranishi 1957] Two integers  $k, l \geq 0$  exist for every regular differential equation  $R$  such that  $\pi^l(D^k R)$  is involutive.*

## Zero Dimensional Polynomial System

**Theorem 3.** [Zhi and Reid 2004] *If the polynomial system has only finite number of solutions, then  $\pi^\ell(D^k R)$  is involutive if and only if it satisfies:*

$$\dim \pi^\ell(D^k R) = \dim \pi^{\ell+1}(D^{k+1} R) \text{ (projected elimination test)}$$

$$\dim \pi^\ell(D^k R) = \dim \pi^{\ell+1}(D^k R) \text{ (symbol involutive test)}$$

and by the bijection  $\phi$ , we have

$$\text{numsols}(F) = \dim \pi^\ell(D^k R)$$

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- The number of solutions of polynomial system  $F$  is  **$\dim \pi^\ell(D^k R)$** .
- Apply eigenvalue method to the **null space of  $\pi^\ell(D^k R)$ ,  $\pi^{\ell+1}(D^k R)$**  to obtain the solutions of  $F$ .

## Example 1 (continue)

**Table 1:**  $\dim \pi^\ell(D^k R)$

	$k = 0$	$k = 1$	$k = 2$	$k = 3$	$k = 4$	$k = 5$
$\ell = 0$	17	23	26	27	27	27
$\ell = 1$	10	17	23	26	27	27
$\ell = 2$	4	10	17	23	26	27
$\ell = 3$	1	4	10	17	23	26
$\ell = 4$		1	4	10	17	23
$\ell = 5$			1	4	10	17
$\ell = 6$				1	4	10
$\ell = 7$					1	4
$\ell = 8$						1

## Example 1 (continued): Eigenvalue Method

We obtain 16 simple solutions:

$$\begin{aligned} &(1, 1, 1), (1, -1, -1), (1, -1, -1), (-1, -1, 1), \\ &(1, i, -i), (1, -i, i), (i, 1, -i), (-i, 1, i), \\ &(i, -i, 1), (-i, i, 1), (-1, i, i), (-1, -i, -i), \\ &(i, -1, i), (-i, -1, -i), (i, i, -1), (-i, -i, -1) \end{aligned}$$

and one multiple root

$$(0, 0, 0)$$

with multiplicity **11**.



## Computing Multiplicity Structure [Wu and Zhi'08]

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- Construct differential operators
- Refine an approximate singular solution

**Theorem 4.** *[Van Der Waerden 1970] Suppose the polynomial ideal  $I = (f_1, \dots, f_t)$  has an isolated primary component  $Q$  whose associated prime  $P$  is maximal, and  $\rho$  is the index of  $Q$ , i.e., the minimal nonnegative integer s.t.  $\sqrt{Q}^\rho \subset Q$ .*

- *If  $\sigma \leq \rho$ , then*

$$\dim(\mathbb{C}[\mathbf{x}]/(I, P^{\sigma-1})) < \dim(\mathbb{C}[\mathbf{x}]/(I, P^\sigma)).$$

- *If  $\sigma > \rho$ , then*

$$Q = (I, P^\rho) = (I, P^\sigma).$$

**Corollary 5.** *The index is less than or equal to the multiplicity  $\mu$ :*

$$\rho \leq \mu = \dim(\mathbb{C}[\mathbf{x}]/Q).$$

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- Compute  $d_k = \dim(\mathbb{C}[\mathbf{x}]/(I, P^k))$  by **SNEPSolver** for a given tolerance  $\tau$  until  $d_k = d_{k-1}$ , set

$$\rho = k - 1, \mu = d_\rho, Q = (I, P^\rho).$$

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- Compute the multiplication matrices  $M_{x_1}, \dots, M_{x_s}$  of  $\mathbb{C}[\mathbf{x}]/Q$  to obtain the primary component.

## Example 2 [Ojika 1987]

$$I = (f_1 = x_1^2 + x_2 - 3, f_2 = x_1 + 0.125x_2^2 - 1.5)$$

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$$\dim(\mathbb{C}[\mathbf{x}]/(I, P^2)) = 2,$$

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So we have index  $\rho = 3$ , multiplicity  $\mu = 3$ .

## Example 2 (continued)

The multiplication matrices (**local ring**) w.r.t.  $\{x_1, x_2, 1\}$ :

$$M_{x_1} = \begin{bmatrix} 0 & -1 & 3 \\ 6 & 3 & -10 \\ 1 & 0 & 0 \end{bmatrix}, \quad M_{x_2} = \begin{bmatrix} 6 & 3 & -10 \\ -8 & 0 & 12 \\ 0 & 1 & 0 \end{bmatrix}.$$

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The primary component of  $I$  associated to  $(1, 2)$  is

$$(x_1^2 + x_2 - 3, x_2^2 + 8x_1 - 12, x_1x_2 - 6x_1 - 3x_2 + 10).$$

## Differential Operators

- Let  $D(\alpha) = D(\alpha_1, \dots, \alpha_s) : \mathbb{C}[\mathbf{x}] \rightarrow \mathbb{C}[\mathbf{x}]$  denote the differential operator defined by:

$$D(\alpha_1, \dots, \alpha_s) = \frac{1}{\alpha_1! \cdots \alpha_s!} \partial x_1^{\alpha_1} \cdots \partial x_s^{\alpha_s},$$

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- Let  $\mathfrak{D} = \{D(\alpha) \mid |\alpha| \geq 0\}$ , we define the space associated to  $I$  and  $\hat{\mathbf{x}}$  as

$$\Delta_{\hat{\mathbf{x}}} := \{L \in \text{Span}_{\mathbb{C}}(\mathfrak{D}) \mid L(f)|_{\mathbf{x}=\hat{\mathbf{x}}} = 0, \forall f \in I\}.$$

## Construct Differential Operators I

- Write Taylor expansion of  $h \in \mathbb{C}[x]$  at  $\hat{x}$ :

$$T_{\rho-1}h(x_1, \dots, x_s) = \sum_{\alpha \in \mathbb{N}^s, |\alpha| < \rho} c_\alpha (x_1 - \hat{x}_1)^{\alpha_1} \cdots (x_s - \hat{x}_s)^{\alpha_s}.$$

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- Compute  $\text{NF}(h)$ , and expand it at  $\hat{\mathbf{x}}$

$$\text{NF}(h(x)) = \sum_{\beta} d_\beta (\mathbf{x} - \hat{\mathbf{x}})^\beta,$$

and find scalars  $a_{\alpha\beta} \in \mathbb{C}$  such that  $d_\beta = \sum_{\alpha} a_{\alpha\beta} c_\alpha$ .



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and find scalars  $a_{\alpha\beta} \in \mathbb{C}$  such that  $d_\beta = \sum_{\alpha} a_{\alpha\beta} c_\alpha$ .

- For each  $\beta$  such that  $d_\beta \neq 0$ , return

$$L_\beta = \sum_{\alpha} a_{\alpha\beta} \frac{1}{\alpha_1! \cdots \alpha_s!} \partial x_1^{\alpha_1} \cdots \partial x_s^{\alpha_s} = \sum_{\alpha} a_{\alpha\beta} D(\alpha).$$

$L = \{L_1, \dots, L_\mu\}$  is a set of basis for  $\Delta_{\hat{\mathbf{x}}}$ .

[Damiano, Sabadini, Struppa '07]

## Example 2 (continued)

Write Taylor expansion at  $(1, 2)$  up to degree  $\rho - 1 = 2$ ,

$$h(\mathbf{x}) = c_{0,0} + c_{1,0}(x_1 - 1) + c_{0,1}(x_2 - 2) + c_{2,0}(x_1 - 1)^2 + c_{1,1}(x_1 - 1)(x_2 - 2) + c_{0,2}(x_2 - 2)^2.$$

Obtain the normal form of  $h$  by replacing  $x_1^2, x_1x_2, x_2^2$  with

$$x_1^2 = -x_2 + 3, x_1x_2 = 6x_1 + 3x_2 - 10, x_2^2 = -8x_1 + 12.$$

The differential operators are:

$$\begin{cases} L_1 &= D(0, 0), \\ L_2 &= D(0, 1) - D(2, 0) + 2D(1, 1) - 4D(0, 2), \\ L_3 &= D(1, 0) - 2D(2, 0) + 4D(1, 1) - 8D(0, 2). \end{cases}$$

## Criterion of Involution of $F_k$

Let  $\mathbf{T}_k(f_i) = \sum_{|\alpha| < k} f_{i,\alpha}(\mathbf{x} - \hat{\mathbf{x}})^\alpha$ , and

$$F_k = \{ \mathbf{T}_k(f_1), \dots, \mathbf{T}_k(f_t), (x_1 - \hat{x}_1)^{\alpha_1} \cdots (x_s - \hat{x}_s)^{\alpha_s}, \sum \alpha_i = k \}.$$

Symbol matrices of  $F_k$  and prolongations are of **full column rank**.

$M_k^{(j)}$  denotes coeff. matrices of the truncated prolonged system

$\mathbf{T}_k(F^{(j)})$  with  $\binom{k+s-1}{s}$  columns,  $d_k^{(j)} = \dim \text{Nullspace}(M_k^{(j)})$ .

**Theorem 6.**  $F_k$  is involutive at prolongation order  $m$  if and only if

$$d_k^{(m)} = d_k^{(m+1)}$$

and  $d_k = \dim(\mathbb{C}[\mathbf{x}] / (I, P^k)) = d_k^{(m)}$ .

## Compute Primary Component II

- Form the matrix  $M_k^{(0)}$  by computing the truncated Taylor series expansions of  $f_1, \dots, f_t$  at  $\hat{x}$  to order  $k$ . The prolonged matrix  $M_k^{(j)}$  is computed by shifting  $M_k^{(0)}$  accordingly.

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- If  $d_k = d_{k-1}$ , then set  $\rho = k - 1$  and  $\mu = d_\rho$ .
- Compute the multiplication matrices  $M_{x_1}, \dots, M_{x_s}$  from the null vectors of  $M_{\rho+1}^{(m)}$ .

### Example 3 [Leykin et al. 2006]

$$\{f_1 = x_1^3 + x_2^2 + x_3^2 - 1, f_2 = x_1^2 + x_2^3 + x_3^2 - 1, f_3 = x_1^2 + x_2^2 + x_3^3 - 1\}$$

has a 4-fold solution  $\hat{\mathbf{x}} = (1, 0, 0)$ . Transform it to the origin:

$$\begin{cases} g_1 &= y_1^3 + 3y_1^2 + 3y_1 + y_2^2 + y_3^2, \\ g_2 &= y_1^2 + 2y_1 + y_2^3 + y_3^2, \\ g_3 &= y_1^2 + 2y_1 + y_2^2 + y_3^3. \end{cases}$$

has the 4-fold solution  $\hat{\mathbf{y}} = (0, 0, 0)$ . Let  $I = (g_1, g_2, g_3)$ ,  
 $P = (y_1, y_2, y_3)$ .

$$[\mathbf{T}_3(g_1), \mathbf{T}_3(g_2), \mathbf{T}_3(g_3)]^T = M_3^{(0)} \cdot [y_1^2, \dots, y_3, 1]^T,$$

$$M_3^{(0)} = \begin{bmatrix} 3 & 0 & 0 & 1 & 0 & 1 & 3 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 2 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 2 & 0 & 0 & 0 \end{bmatrix}$$



$$M_3^{(1)} = \begin{bmatrix} 3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 3 & 0 & 0 & 1 & 0 & 1 & 3 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 2 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 2 & 0 & 0 & 0 \end{bmatrix}.$$

## Example 3 (continued)

- $d_3^{(0)} = 7, d_3^{(1)} = d_3^{(2)} = 4 \implies d_3 = \dim(\mathbb{C}[\mathbf{y}]/(I, P^3)) = 4.$
- $d_4^{(0)} = 17, d_4^{(1)} = 8, d_4^{(2)} = d_4^{(3)} = 4,$   
 $\implies d_4 = \dim(\mathbb{C}[\mathbf{y}]/(I, P^4)) = 4.$
- $d_3 = d_4 = 4$ , then index  $\rho = 3$ , multiplicity  $\mu = 4.$

## Construct Differential Operators II

**Theorem 7.** *Let  $Q = (I, P^\rho)$  be an isolated primary component of  $I$  at  $\hat{\mathbf{x}}$  and  $\mu \geq 1$ . Suppose  $F_\rho = T_\rho(F) \cup P^\rho$  is involutive after  $m$  prolongations, the null space of the matrix  $M_\rho^{(m)}$  is generated by  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_\mu$ . Then differential operators are:*

$$L_j = \mathbf{L} \cdot \mathbf{v}_j, \quad \text{for } 1 \leq j \leq \mu,$$

$$\mathbf{L} = [D(\rho - 1, 0, \dots, 0), D(\rho - 2, 1, 0, \dots, 0), \dots, D(0, \dots, 0)].$$

See also [Dayton and Zeng 2005].

## Example 3 (continued)

Since  $\rho = 3$ ,  $\mu = 4$ , and  $d_3^{(0)} = 7$ ,  $d_3^{(1)} = d_3^{(2)} = 4$ , the null space of  $M_3^{(1)}$  is:

$$N_3^{(1)} = [e_{10}, e_9, e_8, e_5],$$

Multiplying the diff. operators of order less than 3:

$$\{D(0,0,0), D(0,0,1), D(0,1,0), D(0,1,1)\}.$$

## Example 1 (continued)

Since  $\rho = 5$ ,  $\mu = 11$ , and

$$d_5^{(0)} = 23, d_5^{(1)} = d_5^{(2)} = 11,$$

Multiplying the diff. operators of order less than 5 w.r.t. to the null vectors of  $M_5^{(1)}$  ( $35 \times 30$ ), we get

$$D(0,0,0), D(1,0,0), D(0,0,1), D(0,1,0),$$

$$D(2,0,0), D(0,2,0), D(0,0,2),$$

$$D(0,0,3) + D(1,1,0), D(0,3,0) + D(1,0,1), D(3,0,0) + D(0,1,1),$$

$$D(0,0,4) + D(0,4,0) + D(4,0,0) + D(1,1,1)$$

# Complexity for Computing Differential Operators

The complexity of our algorithm is:

$$O\left(t \binom{\rho + s - 1}{s}^3\right).$$

The complexity of algorithm in [Mourrain MEGA'96] is:

$$O((s^2 + t)\mu^3).$$

Notice  $\mu \leq \binom{\rho + s - 1}{s}$ .

## Approximate Singular Solution

- Suppose  $\hat{\mathbf{x}}$  is an approximate singular solution of  $F$ :

$$\hat{\mathbf{x}} = \hat{\mathbf{x}}_{\text{exact}} + \hat{\mathbf{x}}_{\text{error}}.$$

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- Transform  $\hat{\mathbf{x}}$  to the origin, and we get a new system  $G = \{g_1, \dots, g_t\}$ , where  $g_i = f_i(y_1 + \hat{x}_1, \dots, y_s + \hat{x}_s)$ .



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- $\hat{\mathbf{y}} = -\hat{\mathbf{x}}_{\text{error}} = (-\hat{x}_{1,\text{error}}, \dots, -\hat{x}_{s,\text{error}})$  is an exact solution of the system  $G$ .

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- Construct multiplication matrices locally to refine the solution.

## Refining Approximate Singular Solution(RASS)

- For approximate  $\hat{\mathbf{x}}$  and tolerance  $\tau$ , the prime ideal  $P = (x_1 - \hat{x}_1, \dots, x_s - \hat{x}_s)$ , estimate  $\mu$  and  $\rho$ .

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- Set  $\hat{\mathbf{x}} = \hat{\mathbf{x}} + \hat{\mathbf{y}}$  and run the first two steps for the refined solution and smaller  $\tau$ .
- If  $\hat{\mathbf{y}}$  converges to the origin, we get  $\hat{\mathbf{x}}$  with high accuracy.

## Example 3 (continued)

Given an approximate solution  $\hat{\mathbf{x}} = (1.001, -0.002, -0.001 i)$ .

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Given an approximate solution  $\hat{\mathbf{x}} = (1.001, -0.002, -0.001 i)$ .

Set  $\tau = 10^{-2}$ , we compute the singular solution of  $G$ :

$$\hat{\mathbf{y}} = (-0.0009994 - 7.5315 \times 10^{-10} i, \\ 0.002001 + 2.8002 \times 10^{-8} i, \\ -1.4949 \times 10^{-6} + 0.0010000 i).$$



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$$\hat{\mathbf{x}} = (1 + 0.6 \times 10^{-6} - 7.5315 \times 10^{-10} i, \\ 0.1 \times 10^{-5} + 2.8002 \times 10^{-8} i, \\ -1.4949 \times 10^{-6}).$$

## Example 3 (continued)

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Apply twice for  $\tau = 10^{-5}, 10^{-8}$  respectively, we get:

$$\hat{\mathbf{x}} = (1 + 7.0405 \times 10^{-18} - 7.8146 \times 10^{-19} i, \\ 1.0307 \times 10^{-16} - 1.9293 \times 10^{-17} i, \\ 1.5694 \times 10^{-16} + 7.9336 \times 10^{-17} i).$$

## Algorithm Performance

System	Zero	$\rho$	$\mu$	RASS
cmbs1	$(0, 0, 0)$	5	11	$3 \rightarrow 11 \rightarrow 15$
cmbs2	$(0, 0, 0)$	4	8	$3 \rightarrow 13 \rightarrow 15$
mth191	$(0, 1, 0)$	3	4	$4 \rightarrow 9 \rightarrow 15$
LVZ	$(0, 0, -1)$	7	18	$5 \rightarrow 10 \rightarrow 14$
KSS	$(1, 1, 1, 1, 1, 1)$	5	16	$5 \rightarrow 11 \rightarrow 14$
Caprasse	$(2, -i\sqrt{3}, 2, i\sqrt{3})$	3	4	$4 \rightarrow 12 \rightarrow 15$
DZ1	$(0, 0, 0, 0)$	11	131	$5 \rightarrow 14$
DZ2	$(0, 0, -1)$	8	16	$4 \rightarrow 7 \rightarrow 14$
D2	$(0, 0, 0)$	5	5	$5 \rightarrow 10 \rightarrow 15$
Ojika1	$(1, 2)$	3	3	$3 \rightarrow 6 \rightarrow 18$
Ojika2	$(0, 1, 0)$	2	2	$5 \rightarrow 10 \rightarrow 14$

Examples cited from <http://www.math.uic.edu/~jan/>,

[Dayton, Zeng '05, Dayton '07].

Thank you!

Grazie mille!