

Differential elimination for dynamical systems

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MAX team, LIX, CNRS, École Polytechnique, Institut Polytechnique de Paris

MEGA (Effective Methods in Algebraic Geometry)

June 10, 2021



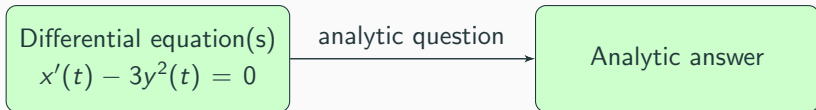
Plan

- I (Some) algebra of differential equations
Differential elimination: what and why?
- II Elimination for dynamical system in theory & action
- III Open problems and conclusions
Including very specific problems

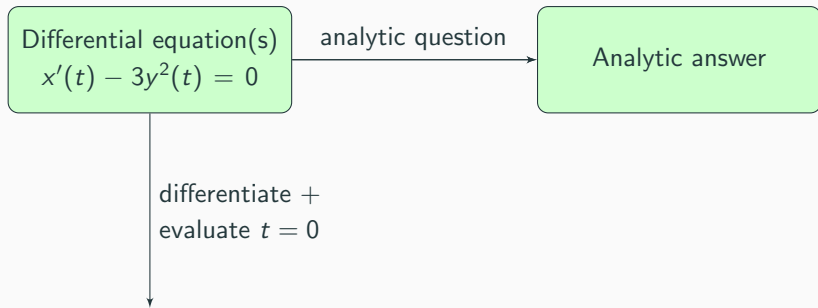
Part I

(Some) Algebra of differential equations

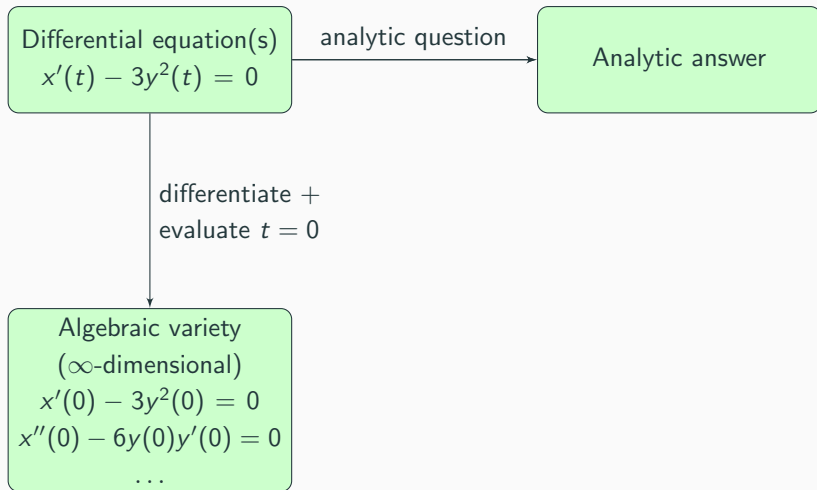
Differential equations meet algebraic geometry



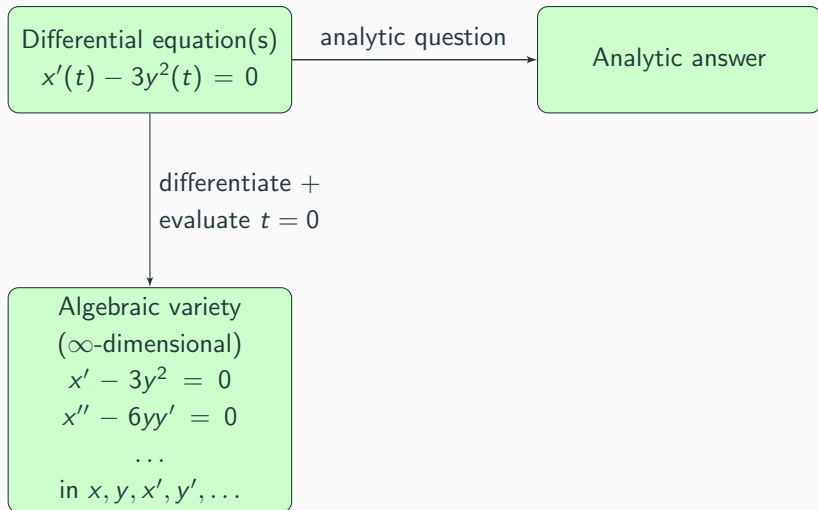
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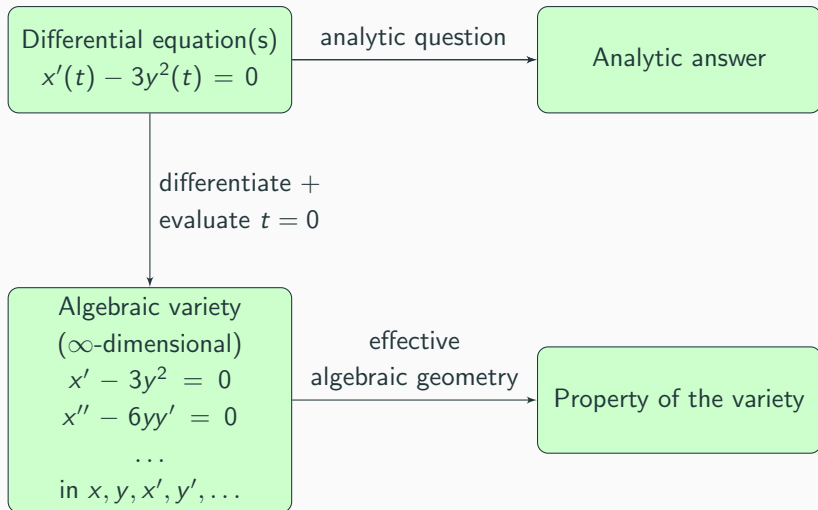
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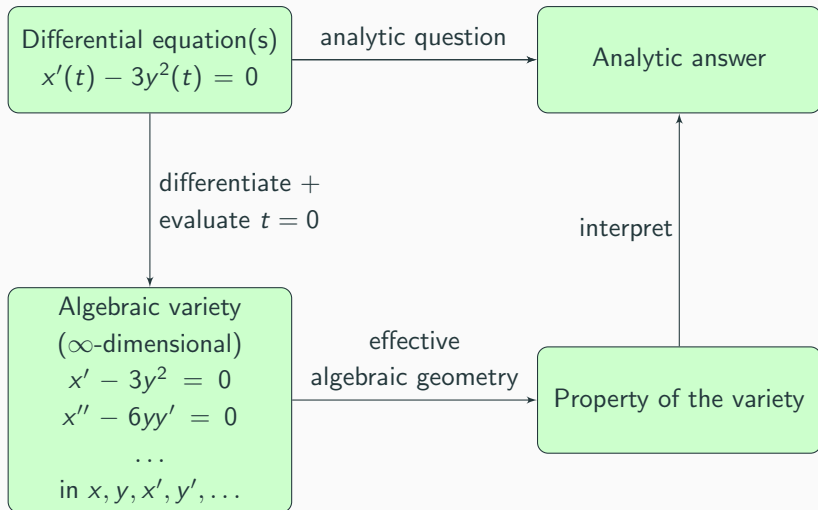
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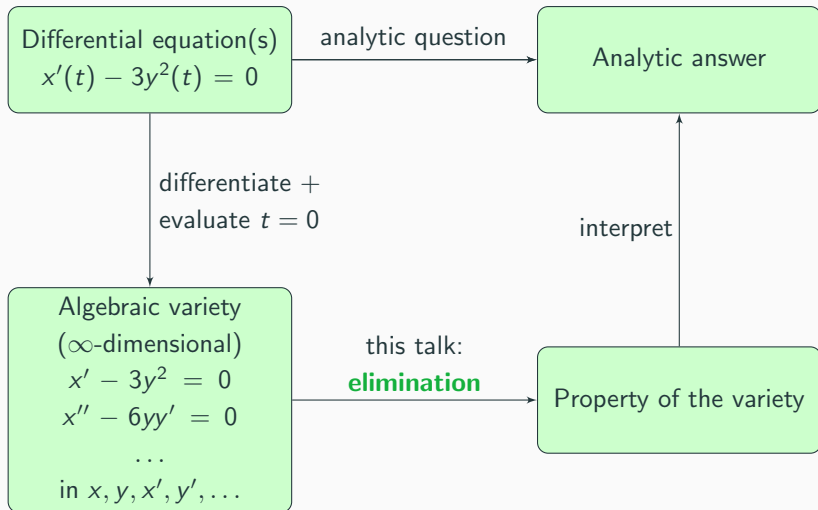
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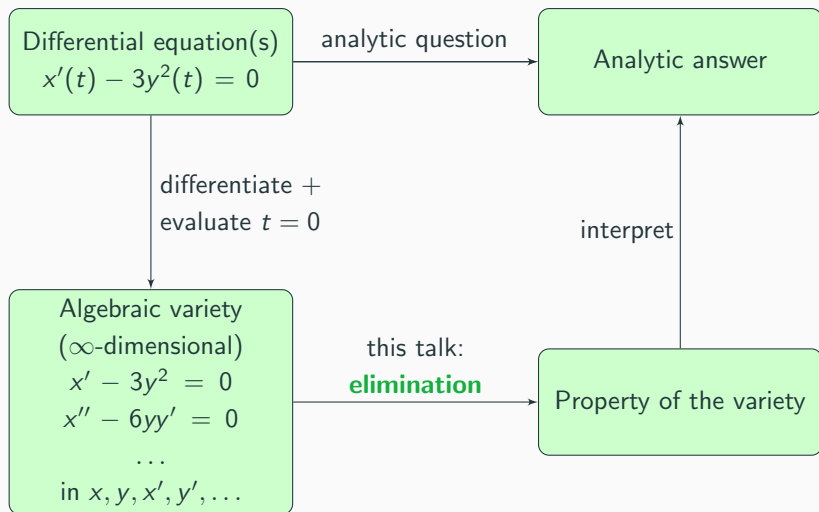
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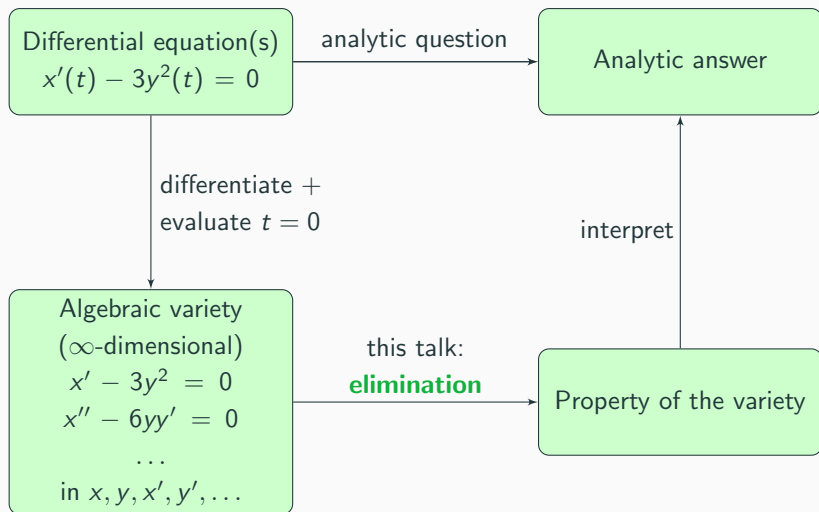


Differential equations meet algebraic geometry



Convergence questions: talk by S. Falkensteiner, J. Cano, R. Sendra on Tuesday

Differential equations meet algebraic geometry



Other connections of algebra and differential equations: talk by D. Agostini, C. Fevola, Y. Mandelshtam, B. Sturmfels on Friday

Differential elimination: general

Fix a ground field k .

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Semantic: Relations satisfied by the \mathbf{y} -component of any power series solution of $f_1 = f_2 = \dots = f_s = 0$.

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Toy example

$$\begin{cases} f_1 = x' - y, \\ f_2 = y' - x \end{cases} \implies y^{(2)} - y = (x' - y) + (y' - x)' \in \langle f_1, f_1', \dots, f_2, f_2', \dots \rangle$$

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Moreover $\langle f_1, f_1', \dots, f_2, f_2', \dots \rangle \cap k[y^{(\infty)}] = \langle y^{(2)} - y, y^{(3)} - y', \dots \rangle$.

Differential elimination: why?

- Eliminate **non-observable** variables from models.

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- Eliminate **auxiliary** (non-important) variables.

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Differential elimination goes back to Ritt (1930-s).

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For every elimination method X there exists at least one “differential X ”.

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Their relative standing is different, e.g:

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- in practice, characteristic sets are the most popular
BLAD library by F. Boulier, available in MAPLE

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Existing algorithms are **general** \implies
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efficient computation in practice is still a **challenge**.

And the plan is?

Restrict to dynamical systems and use **Effective Algebraic Geometry**.

Part II

Elimination for dynamical systems:
in theory and in action

What do we mean by dynamical system?

System

$$\begin{cases} x_1' = f_1(x_1, \dots, x_n), \\ \dots \\ x_n' = f_n(x_1, \dots, x_n), \end{cases}$$

where $f_1, \dots, f_n \in k[x_1, \dots, x_n]$.

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Corresponding ideal

$$\mathcal{I} := \langle \mathbf{x}' - \mathbf{f}, \mathbf{x}^{(2)} - \mathbf{f}', \dots \rangle \subset k[\mathbf{x}^{(\infty)}].$$

Algebraic properties

- \mathcal{I} is prime
- generators \rightarrow Gröbner basis
- the variety is rational
(parameters = initial conditions)

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Notation fixed for the rest of the section: n, \mathcal{I}

Elimination in dynamical systems: setup

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- dynamical system

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- integer $1 \leq s \leq n$
(“to keep x_1, \dots, x_s ”)

Output:

A description of

$$\mathcal{I}_{\text{elim}} := \mathcal{I} \cap k[x_1^{(\infty)}, \dots, x_s^{(\infty)}]$$

(recall $\mathcal{I} := \langle \mathbf{x}' - \mathbf{f}, \mathbf{x}^{(2)} - \mathbf{f}', \dots \rangle$)

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What kind of description?

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What kind of **description**? \Leftarrow depends on the questions we ask!

Questions: identifiability

Given: parametric dynamical system

$$\text{observed} \left\{ \begin{array}{l} x'_1 = f_1(\boldsymbol{\mu}, \mathbf{x}), \\ \dots \\ x'_s = f_s(\boldsymbol{\mu}, \mathbf{x}) \end{array} \right.$$

$$\text{hidden} \left\{ \begin{array}{l} x'_{s+1} = f_{s+1}(\boldsymbol{\mu}, \mathbf{x}), \\ \dots \\ x'_n = f_n(\boldsymbol{\mu}, \mathbf{x}), \end{array} \right.$$

$\boldsymbol{\mu}$: unknown scalar parameters
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Informal definition

μ_1 is **identifiable** if (generically) μ_1 is uniquely determined by functions $x_1(t), \dots, x_s(t)$.

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Example: $s = 1$

$$\begin{cases} x_1' = x_2, \\ x_2' = \mu x_1. \end{cases} \implies \text{YES } (\mu = \frac{x_1''}{x_1})$$

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(Ollivier'90)

μ_1 is identifiable \iff

$\mu_1 \in$ **the field of definition** of $\mathcal{I}_{\text{elim}}$

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μ_1 is identifiable \iff

$\mu_1 \in$ **the field of definition** of $\mathcal{I}_{\text{elim}}$

(under a mild condition, more in

Ovchinnikov, Pillay, P., Scanlon'20)

Assessing identifiability via elimination

Joint with Ruiwen Dong, Christian Goodbrake, and Heather Harrington.

Overall: we compute the field of definition of $\mathcal{I}_{\text{elim}}$.

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In the next slides

- How we represent $\mathcal{I}_{\text{elim}}$?
- How we compute the representation?
- How we extract the field of definition?
- Performance

Representation: infinite \rightarrow finite

How many Taylor coefficients are enough?

A tuple $(h_1, \dots, h_s) \in \mathbb{Z}^s$ is called **profile** if

- $x_1^{(<h_1)}, \dots, x_s^{(<h_s)}$ are algebraically independent modulo \mathcal{I} ,
where $x_j^{(<i)} := (x_j, x_j', \dots, x_j^{(i-1)})$

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- (h_1, \dots, h_s) is maximal with this property.

Intuition: $x_1^{(<h_1)}(0), \dots, x_s^{(<h_s)}(0)$ define $x_1(t), \dots, x_s(t)$ up to finitely many choices (generically).

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Example

$$\begin{cases} x'_1 = -x_2, \\ x'_2 = x_1 \end{cases} \quad \& \quad s = 1 \implies h_1 = 2$$

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Propositions on profiles

- (Hong, Ovchinnikov, P., Yap, 2020) $x_1^{(\leq h_1)}, \dots, x_s^{(\leq h_s)}$ generate the fraction field of $k[x_1^{(\infty)}, \dots, x_s^{(\infty)}] / \mathcal{I}_{\text{elim}}$.
(used for assessing identifiability and observability)

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- (Dong, Goodbrake, Harrington, P., 2021) The **fields of definition** of $\mathcal{I}_{\text{elim}}$ and $\mathcal{I}_{\text{elim}} \cap k[x_1^{(\leq h_1)}, \dots, x_s^{(\leq h_s)}]$ are the same.

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Talk about this now!

Representation: profiles + projections

Representation

- Profile: $(h_1, \dots, h_s) \in \mathbb{Z}^s$ such that
 - $x_1^{(<h_1)}, \dots, x_s^{(<h_s)}$ are algebraically independent modulo $\mathcal{I}_{\text{elim}}$.
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- **Projections:** for every $1 \leq i \leq s$, the generator of the principal ideal

$$\mathcal{I}_{\text{elim}} \cap \mathbb{C}(\mu)[x_1^{(<h_1)}, \dots, x_i^{(\leq h_i)}, \dots, x_s^{(<h_s)}]$$

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$$\mathcal{I}_{\text{elim}} \cap \mathbb{C}(\mu)[x_1^{(<h_1)}, \dots, x_i^{(\leq h_i)}, \dots, x_s^{(<h_s)}]$$

Example (contd.)

$$\begin{cases} x_1' = -x_2, \\ x_2' = x_1 \end{cases} \quad \& \quad s = 1 \quad \& \quad h_1 = 2 \quad \implies \text{Projection: } x_1'' + x_1$$

Representation: profiles + projections

Representation

- **Profile:** $(h_1, \dots, h_s) \in \mathbb{Z}^s$ such that
 - $x_1^{(<h_1)}, \dots, x_s^{(<h_s)}$ are algebraically independent modulo $\mathcal{I}_{\text{elim}}$.
 - (h_1, \dots, h_s) is maximal with this property.
- **Projections:** for every $1 \leq i \leq s$, the generator of the principal ideal

$$\mathcal{I}_{\text{elim}} \cap \mathbb{C}(\boldsymbol{\mu})[x_1^{(<h_1)}, \dots, x_i^{(\leq h_i)}, \dots, x_s^{(<h_s)}]$$

Observation

Original system is **already** in this form with $s = n$ and $h_1 = \dots = h_n = 1$.

Computing the representation: reprofiling!

Profile point of view on elimination

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$$\underbrace{(1, 1, \dots, 1)}_{n \text{ times}}$$

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- Transition: a sequence of sokoban-type steps:

$$\left. \begin{array}{c} (\dots, \underbrace{h}_{x_i, i \leq s}, \dots, \underbrace{1}_{x_j, j > s}, \dots) \\ \text{the projection for } x_i \text{ involves } x_j \end{array} \right\}$$

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Computing the representation: efficiency

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- profile \mathbf{h} is built dynamically by the socoban algorithm (and can be chosen in a more efficient way)
- special variable change to simplify resultant computation.

How to find the field of definition?

Subtlety

The coefficients of the projections belong to the field of definition but not necessarily **generate** it.

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The coefficients of the projections belong to the field of definition but not necessarily **generate** it.

Solution

Taking coefficients of one more projection on a generic plane **is enough** (but typically not needed and can be avoided)

Performance

The resulting algorithm is implemented in **julia** using OSCAR library
<https://github.com/pogudingleb/StructuralIdentifiability.jl>

Runtimes below are on a laptop, 16 GB RAM, 1.6 GHz.

Elimination

Model	MAPLE	Our
Cholera	> 5 h.	3 s.
Pharm	OOM	18 s.
MAPK	OOM	28 s.
SEAIJRC	OOM	29 s.

Identifiability

Model	DAISY	SIAN	Our
Cholera	OOM	> 5 h.	18 s.
Pharm	> 5 h.	> 5 h.	7 min.
MAPK	> 5 h.	> 5 h.	1 min.
SEAIJRC	OOM	> 5 h.	2 min.
NF κ B	OOM	33 min.	> 5 h.

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SIAN (Hong, Ovchinnikov, **P.**, Yap, 2020) is also based in elimination for dynamical systems!

Part III

Open problems and conclusions

Problem 1: Degree of a prolongation variety

Consider a variety V_1

$$\begin{cases} x_1' = x_2^2, \\ x_2' = x_1^2 \end{cases}$$

$\deg V_1 = ?$

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$$\begin{cases} x_1' = x_2^2, \\ x_2' = x_1^2, \\ x_1'' = 2x_2x_2', \\ x_2'' = 2x_1x_1' \end{cases}$$

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m	1	2	3	4	5	6	7	8
$\deg V_m$	4	7	16	25	34	49	64	79

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Guessed formula

$$\deg V_m = \begin{cases} (m+1)^2, & \text{if } m \equiv 0, 1 \pmod{3}, \\ (m+1)^2 - 2, & \text{if } m \equiv 2 \pmod{3} \end{cases}$$

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Similar to Jan Draisma's talk at the NASO seminar...

Problem 2: Computing the support

Predator-prey model

$$\begin{cases} x_1' = a_1 x_1 - a_2 x_1 x_2, \\ x_2' = -a_3 x_2 + a_4 x_1 x_2. \end{cases}$$

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Result: $a_1 a_4 x_1^3 - a_4 x_1^2 x_1' - a_1 a_3 x_1^2 + a_3 x_1 x_1' + x_1 x_1^{(2)} - (x_1')^2 = 0$

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Tropical methods \implies Newton polytope of the implicit equation

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$$x_1'' = 1 \cdot a_2^2 x_1 x_2^2 + 1 \cdot a_2 a_3 x_1 x_2 +$$

$$1 \cdot a_1^2 x_1 + (-1) \cdot a_2 a_4 x_1^2 x_2 + (-2) \cdot a_1 a_2 x_1 x_2$$

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$$a_1 a_4 x_1^3 - a_4 x_1^2 x_1' + 7 a_1^2 x_1^2 + 3 a_1 a_3 x_1^2 - 7 a_1 x_1 x_1' - 3 a_3 x_1 x_1' - x_1 x_1'' + 2 (x_1')^2 = 0$$

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Goal: eliminate x_2

Problem

For a system

$$\begin{cases} x_1' = f(\boldsymbol{\mu}, x_1, x_2), \\ x_2' = g(\boldsymbol{\mu}, x_1, x_2), \end{cases}$$

where $f, g \in \mathbb{C}[\boldsymbol{\mu}, x_1, x_2]$, given Newton polytopes of f and g , predict the Newton polytope of the minimal differential equation for x_1 .

Summary

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- Equations defining dynamical systems are ubiquitous in applications and have interesting and useful geometry.
- Although differential elimination has been studied for ~ 100 years, one can still compute much more.
- Open problems in effective algebraic geometry of practical interest.

Acknowledgements

This work was partially supported by the Paris Ile-de-France Region and National Science Foundation.

