

Numeric-symbolic resolution Of Differential Equations

ANR PRME “NODE” — 320 730 € — 5 years — 150 p. months + 2 PhDs

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Solving differential equations

$$\Phi_1(f_1(t), f_1'(t), \dots, f_n(t), f_n'(t), \dots) = 0$$

...

$$\Phi_{n'}(f_1(t), f_1'(t), \dots, f_n(t), f_n'(t), \dots) = 0$$

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State of the art

- Numerical solvers for machine precision
- Symbolic “solvers”: closed form, simplified equations

Part I: numerical resolution

Problem

Computational cost of solving a system of differential equations ?

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As a function of

- Precision (of the computations or of the guaranteed end-result)
- Type of equations (linear, special functions, stiff, singularities, ...)
- Computational model (sequential, parallel, hardware, ...)

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Applications

- New library for the fast and reliable integration of systems of odes
- Differential Galois theory
- Control theory, flatness



Part II: symbolic resolution

Goal (see also OCCAM project by Gleb Pogudin)

Faster algorithms to solve/reduce systems of differential equations

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Representation through power series solutions (instead of equations)

$$f_1(t) = c_{1,0} + c_{1,1}t + c_{1,2}t^2 + c_{1,3}t^3 + c_{1,4}t^4 + \dots$$

$$f_2(t) = c_{2,0} + c_{2,1}t + c_{2,2}t^2 + c_{2,3}t^3 + c_{2,4}t^4 + \dots$$

$$f_3(t) = c_{3,0} + c_{3,1}t + c_{3,2}t^2 + c_{3,3}t^3 + c_{3,4}t^4 + \dots$$

$c_{i,j}$ → free parameters (initial conditions)

$c_{i,j}$ → constrained parameters (by polynomial equations)

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Homotopy continuation for polynomial equations

- Solve simpler system that is similar (e.g. same degrees)
- Continuously deform into target system and follow solutions

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Sparse interpolation

- Magic device to recover a sparse polynomial from evaluations