An overview of Non Smooth Dynamical Systems Higher order systems, numerical methods and links with Optimization

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Activities of the BipOp Project. http://www.inrialpes.fr/bipop

- * Team: 4 Permanent members + 5 PhD students + 1 post doc
 - Bernard Brogliato (Head of the Project)
 - Claude Lemaréchal
 - Pierre-Brice Wieber
 - Vincent Acary
- * The core of our activities is in the field of the Non-Smooth Analysis :
 - Non smooth optimization (CL)
 - Modeling of Non Smooth Dynamical Systems (NSDS) (VA, BB)
 - Control of NSDS (PBW, BB)
 - Simulation of NSDS (VA, PB)

Favorite applications :

- Mechanical systems with contact and friction (Multi-body dynamics, Granular materials, Buildings made of masonry, ..) with possibly real-time constraints (Haptic feedback)
- Electrical networks with idealized components (Diodes, transistors, switch, ...)
- Walking robot and bipedal Locomotion

European Projects :

- FP5 project SICONOS coordinated by B. Brogliato.
 - → Main outcome : Open source software platform for simulation, modeling and control of NSDS (Python, C++, F77) http://siconos.inrialpes.fr/software
- The FP6 Network of Excellence HyCon (Hybrid Control)

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Outline

- → 1 Introduction on Non Smooth Dynamical systems
 - 1.1 What is a Non Smooth Dynamical System (NSDS) ?
 - 1.2 Linear Complementarity Systems (LCS)
 - 1.3 Lagrangian systems with Contact and Coulomb's Friction
 - 1.4 Optimal Control with state constraints
 - 1.5 Applications
- 2 Historical background on low order systems
- 3 Higher-order systems: Formulation and Time-discretization
- 4 Higher-order systems: Numerical Methods, Applications and links with Optimization
- 5 Conclusions and Perspectives

Introduction Low order systems Higher order systems Numerical Methods Conclusions
 What is a Non Smooth Dynamical System (NSDS) ?

A NSDS is a dynamical system is characterized by two correlated features :

- * a non smooth evolution with the respect to time, for instance :
 - Jumps in the state and/or in its derivatives wrt. time
 - Generalized solutions (distributions)
- * a set of non smooth laws (Generalized equations) between the state x and a set of Lagrange multipliers λ

A typical example is the finite-dimensional unilateral dynamics :

$$\begin{cases} \dot{x} = f(x,t) + \lambda, x \in \mathbb{R}^n \\ x \ge 0, \lambda \ge 0, x \cdot \lambda = 0 \end{cases}$$
(1)

that can be written as a (unbounded) differential inclusion :

$$-\dot{x} + f(x,t) = -\lambda \in \partial \Psi_{\mathbb{R}_+}(x) = \mathcal{N}_{\mathbb{R}_+}(x)$$
(2)

where - $\Psi_{\rm I\!R_+}$ is the indicatrix function of $\rm I\!R_+$ - $\mathcal{N}_{\rm I\!R_+}$ the normal cone to $\rm I\!R_+$, i.e, $\rm I\!R_-$

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Typical examples

Linear Complementarity Systems (LCS)

$$\begin{cases} \dot{x} = Ax + B\lambda, & x \in \mathbb{R}^{n}, \lambda \in \mathbb{R}^{m} \\ y = Cx + D\lambda & (3) \\ 0 \le y \perp \lambda \ge 0 \end{cases}$$
with $A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, C \in \mathbb{R}^{m \times n}, D \in \mathbb{R}^{m \times m}, for m \text{ constraints.} \end{cases}$

у

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 λ

% Piecewise linear systems



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Lagrangian dynamical system :

 $M(q)\ddot{q} + Q(\dot{q},q) + F(\dot{q},q,t) = F_{ext}(t) + R$

- $q \in \mathbb{R}^n$: generalized coordinates vector.
- $M \in {\rm I\!R}^{n \times n}$: the inertia matrix
- $Q(\dot{q},q)$: The non linear inertial term (Coriolis)
- $F(\dot{q},q,t)$: the internal forces
- $F_{ext}(t): \mathbb{R} \mapsto \mathbb{R}^n$: given external load,
- $R \in \mathbb{R}^n$ is the force due the non smooth law.

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- $R \in \mathbb{R}^n$ is the force due the non smooth law.
- Linear relations.
 - Kinematic laws from the generalized coordinates to the local coordinates at contact.

$$y = H^T q + b, \dot{y} = H^T \dot{q}$$

Mapping H: Restriction mapping composed with a change of frame

• By duality,

$$R = H\lambda$$

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* Local frame at contact : (n, t)

$$y = y_n n + y_t, \quad \dot{y} = \dot{y}_n n + \dot{y}_t$$

$$\lambda = \lambda_n n + \lambda_t,$$



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* Coulomb's Friction, μ Coefficient of friction, $C(\mu\lambda_n) = \{\lambda_t, \|\lambda_t\| \le \mu\lambda_n\}$

$$\begin{cases} \dot{y}_{t} = 0, \|\lambda_{t}\| \leq \mu\lambda_{n} \\ \dot{y}_{t} \neq 0, \lambda_{t} = -\mu\lambda_{n} \operatorname{sign}(\dot{y}_{t}) \end{cases} \iff \dot{y}_{t} \in \partial\Psi_{\mathcal{C}(\mu\lambda_{n})}(-\lambda_{t}) \Longleftrightarrow -\lambda_{t} \in \partial\Psi_{\mathcal{C}(\mu\lambda_{n})}^{*}(\dot{y}_{t}) \end{cases}$$

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(Newton) Impact law, if necessary, e coefficient of restitution

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 Optimal Control with state constraints

% Quadratic optimal control Problem

$$\min_{u} I(u) = \frac{1}{2} \int_{0}^{T} (x^{T}Qx + u^{T}Ru) dt + \frac{1}{2}x^{T}(T)Fx(T)$$

(s.t.) $\dot{x}(t) = Ax(t) + Bu(t)$
 $x(0) = x_{0}, \quad x(T) = x_{T}$
 $w(t) = Cx(t) + D \ge 0$

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* Necessary conditions \implies LCS with Boundary conditions :

$$\begin{bmatrix} \dot{x} \\ \dot{\eta} \end{bmatrix} = \begin{bmatrix} A & BR^{-1}B^T \\ Q & -A^T \end{bmatrix} \begin{bmatrix} x \\ \eta \end{bmatrix} + \begin{bmatrix} 0 \\ -C^T \end{bmatrix} \lambda$$
$$0 \leq Cx(t) + D \perp \lambda \geq 0$$
$$x(0) = x_0, \quad x(T) = x_T$$
$$\eta(0) = \eta_0, \quad \eta(T) = Fx(T) + C^T \gamma + \beta = \eta_T$$

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Applications

Simulation, modeling and control of electrical networks with idealized components (diodes, transistors, switch, ...)



DC-DC Boost Converter with Sliding mode control

IntroductionApplications

Simulation, modeling and control of electrical networks with idealized components (diodes, transistors, switch, ...)

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Simulation, modeling and control of mechanical systems Simulation of Circuit breakers (INRIA/Schneider Electric)



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- Simulation, modeling and control of electrical networks with idealized components (diodes, transistors, switch, ...)
- Simulation, modeling and control of mechanical systems

Bipedal Robot INRIA BIPOP

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Optimization and its app





- Simulation, modeling and control of electrical networks with idealized components (diodes, transistors, switch, ...)
- Simulation, modeling and control of mechanical systems



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- Simulation, modeling and control of electrical networks with idealized components (diodes, transistors, switch, ...)
- Simulation, modeling and control of mechanical systems



* There are also applications in biology, macro-economics, ..

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 - 2.2 Approaches
 - 2.3 Moreau's Sweeping Process of order 1
 - 2.4 Moreau's Sweeping Process of order 2
 - 2.5 Moreau's Sweeping Process. Discretization
 - 2.6 Summary of the algorithm
 - 2.7 The Bouncing ball example with time-stepping
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 - 2.8 Open Problems and links with optimization
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Difficulties and Approaches

Two major difficulties :

- ✤ Time integration of non smooth evolution
- * Solving a optimization problem together with a dynamical equilibrium constraint

Two major approaches :

* Hybrid multi-modal dynamical system : Event–Driven Approach

For a set of unilateral constraints, $y_{\alpha} = h_{\alpha}(x) \ge 0$, $\alpha = 1 \dots \nu$, we define the index set of active constraints as : $I = \{\alpha, y_{\alpha} = 0\}$ and associated modes. An Event is a change in the index set of active constraints and a change of mode

- Advantages
 - Easy to handle from the computational point of view : smooth integration between two events (ODE/DAE). At event, a optimization problem is solved without time evolution.
- Disadvantages :
 - Need an accurate event detection
 - Accumulation of events
 - No existence or uniqueness results
- Lead to Numerical Event–Driven schemes suitable :
 - Small systems with a small number of events

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- Difficulties and Approaches (continued ...)
 - * Unbounded Differential inclusion and Sweeping process
 - Advantages

- Compact formulation which allow existence and uniqueness results
- Dissipativity and monotonicity properties
- Disadvantages :
 - More difficult mathematical framework
 - Low order accuracy
- Lead to Time-stepping integration schemes (without event-handling) suitable :
 - Large systems with a large number of events
 - Accumulation of events in finite time
 - Convergence results and Existence proofs

Moreau's Sweeping Process of order 1

The Moreau's Sweeping Process is a kind of unbounded differential inclusion (Moreau 1971, 1977, Brezis 1973) :

$$\begin{cases} x(0) \in K(0) \subset \mathbb{R}^n \\ \dot{x} \in \mathcal{N}_{K(t)}(x(t)) \end{cases} \iff \begin{cases} x(0) \in K(0) \subset \mathbb{R}^n \\ \dot{x} = \lambda \\ K \ni x \perp \lambda \in N_{K(t)}(x(t)) \end{cases}$$

where K(t) is a convex set Major results :

K(t) is bounded and a Lipschitz-continuous multi-function (Hausdorff distance) then there exists a unique solution, which is Lipschitz-continuous with the respect to time.

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* If K(t) is a multi-function with right continuous bounded variation then there exists a unique solution, which is of bounded variation and right continuous (Monteiro–Marques, 1987)

References on seminal works :

- Moreau, J.J. (1971) Rafle par un convexe variable, Séminaire d'analyse convexe
- Moreau, J.J. (1977) Evolution problem associated with a moving convex set in a Hilbert space, J. of Differential Equation, pp. 347–374
- Brezis, H. (1973) Maximal Monotone operators, North–Holland Publishing.

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 Moreau's Sweeping Process of order 1 (Continued...)

Equivalent Formulations

Other equivalent formulations with Projected dynamical system, Differential inclusion and variational inequalities may be found in : V.Acary, B. Brogliato, C. Lemaréchal and A. Daniilidis, Inria Research Report, RR-5107 to appear Systems and Control Letters, 2005

The Time discretization is given by the *Catching up algorithm* for instance for $K = \mathbb{R}^+$:

$$\begin{cases} x_{k+1} - x_k = h\lambda_{k+1} \\ 0 \le x_{k+1} \perp \lambda_{k+1} \ge 0 \end{cases}$$
(1)

Remarks :

- Implicit type scheme (necessary for the unilateral constraints) but low order =1
- Resolution of a LCP at each time step
- Convergence Proofs => Existence of solution due to the monotonicity of the operator

This algorithm may be used for LCS system for which *D* is P-matrix.

- The velocity is no longer a smooth function of time but a function of bounded variations. This is the case of Lagrangian systems
- * Lagrangian dynamical system is reformulated as a measure differential equation.

$$M(q)dv + (Q(v,q) + F(v,q,t)) dt = F_{ext}(t) dt + R$$

where

- dt is the Lebesgue measure on ${\rm I\!R}$
- dv is the Stieltjes measure (Differential measure) associated with the right continuous function v(t) of bounded variations, such that :

$$dv((a,b]) = \int_{(a,b]} dv = v(b^+) - v(a^+)$$

- R is a measure due to the non smooth law
- q(t) is the absolutely continuous displacement given by :

$$q(t) = q(t_0) + \int_{t_0}^t v(s) \, ds$$

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 Moreau's Sweeping Process of order 2 (Continued ...)

Reformulation of the constraints as a measure inclusion

* Reformulation of the unilateral constraints in terms of derivatives :

 $-\lambda \in \partial \Psi_{V(y)}(\dot{y})$

where V(y) is the tangent cone of K at y which can be stated equivalently for $K = \mathbb{R}^+$ as for

If
$$y(t) = 0$$
, then $0 \le \dot{y} \perp \lambda \ge 0$ (2)

It's noteworthy that the (switched off) constraints is now on the velocity \dot{y} and depend on the value of y(t) = 0

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 Moreau's Sweeping Process of order 2 (Continued ...)

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% If λ is a measure, the inclusion is extended considering the Radon-Nykodym derivative

$$\lambda'(t) = \frac{d\lambda}{d\nu} \in \partial \Psi_{V(y)}(\dot{y})$$

where $d\nu$ is a nonnegative measure and λ is absolutely continuous with respect to $d\nu$

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 ≪ Given a subdivision of a time interval, $\{t_0, t_1, ..., t_i, ..., t_N\}$, we evaluate of the measure differential equation on a time interval $(t_i, t_{i+1}]$ of length *h* :

$$Mdv((t_i, t_{i+1}]) = \int_{(t_i, t_{i+1}]} M \, dv = M(v(t_{i+1}^+) - v(t_i^+)) = \int_{t_i}^{t_{i+1}} F_{ext}(t) \, dt + \int_{(t_i, t_{i+1}]} R$$

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***** Evaluation of the displacement $q(t_{i+1}) = q(t_i) + \int_{t_i}^{t_{i+1}} v(s) ds$

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- ***** Evaluation of the displacement $q(t_{i+1}) = q(t_i) + \int_{t_i}^{t_{i+1}} v(s) ds$
- * The measure $R((t_i, t_{i+1}])$ of the time-interval $(t_i, t_{i+1}]$ is kept as primary unknown :

$$R_{i+1} = R((t_i, t_{i+1}])$$

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Interpretation : The measure R may be decomposed as follows :

$$R = R_a \, dt + R_s$$

where $R_a dt$ is the abs. continuous part of the measure R and R_s the singular part.

- Impulse : If $R_a = 0$ and $R_s = P\delta_{t_{i+1}}$ then $R_{i+1} = P$
- Continuous multiplier : If $R_a(t) = f(t)$ and $R_s = 0$ then $R_{i+1} = \int_{t_i}^{t_i+1} f(t) dt$

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Notations :

$$v_i \approx v(t_i^+), \quad q_i \approx q(t_i)$$

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***** Approximation of the integral of functions : θ -method

$$\int_{t_i}^{t_{i+1}} F_{ext}(t) dt \approx h \left[\theta F_{ext}(t_{i+1}) + (1-\theta) F_{ext}(t_i) \right]$$

$$q_{i+1} = q_i + h \left[\theta v_{i+1} + (1-\theta)v_i\right]$$

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Complete set of discrete equations:

$$\begin{cases} M(v_{i+1} - v_i) = h \left[\theta F_{ext}(t_{i+1}) + (1 - \theta)(F_{ext}(t_i)) \right] + R_{i+1} \\ q_{i+1} = q_i + h \left[\theta v_{i+1} + (1 - \theta) v_i \right] \end{cases}$$

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* One step linear system : $v_{i+1} = v_{free} + hWR_{i+1}$ with

$$W = M^{-1}, \quad v_{free} = v_i + W \left[h \left[\theta F_{ext}(t_{i+1}) + (1 - \theta) F_{ext}(t_i) \right] \right]$$

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***** Relations at t_{k+1} :

$$y_{i+1} = H^T q_{i+1} + b$$
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 Discretization of the Dynamics Continued

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Discretization of an unilateral constraint : A natural way :

$$0 \le y_{i+1} \perp \lambda_{i+1} \ge 0$$

in terms of velocity

If
$$y^p \leq 0$$
, then $0 \leq \dot{y}_{i+1} \perp \lambda_{i+1} \geq 0$

where y^p is a prediction of the position at time t_{i+1} , for instance, $y^p = y_i + \frac{h}{2}\dot{y}_i$.

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If
$$y^p \leq 0$$
, then $0 \leq \dot{y}_{i+1} \perp \lambda_{i+1} \geq 0$

where y^p is a prediction of the position at time t_{i+1} , for instance, $y^p = y_i + \frac{h}{2}\dot{y}_i$.

***** Newton Impact law $\dot{y}_{i+1}^e = \dot{y}_{i+1} + e\dot{y}_i$

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One step linear problem	$\begin{cases} v_{i+1} = v_{free} + hWR_{i+1} \\ q_{i+1} = q_i + h \left[\theta v_{i+1} + (1-\theta)v_i\right] \end{cases}$
Relations	$\begin{cases} \dot{y}_{i+1} = H^T v_{i+1} \\ R_{i+1} = H\lambda_{i+1} \end{cases}$
Non Smooth Law	$\begin{cases} If \; y^p = y_i + \frac{h}{2} \dot{y}_i \\ then \; 0 \leq \dot{y}_{i+1}^e \perp \lambda_{i+1} \geq 0 \end{cases}$

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$$\begin{array}{ll} \text{One step linear problem} & \begin{cases} v_{i+1} = v_{free} + hWR_{i+1} \\ q_{i+1} = q_i + h \left[\theta v_{i+1} + (1-\theta) v_i \right] \\ \\ \text{Relations} & \begin{cases} \dot{y}_{i+1} = H^T v_{i+1} \\ R_{i+1} = H\lambda_{i+1} \end{cases} \\ \\ \text{Non Smooth Law} & \begin{cases} \text{If } y^p = y_i + \frac{h}{2} \dot{y}_i \\ \text{then } 0 \leq \dot{y}_{i+1}^e \perp \lambda_{i+1} \geq 0 \end{cases} \end{cases}$$

 \rightarrow One step Quasi-LCP in terms of \dot{y}_{i+1}^e and λ_{i+1} :

$$\begin{split} \dot{y}_{i+1}^{e} &= H^{T} \dot{q}_{free} + h H^{T} W H \lambda_{i+1} + e \dot{y}_{i} \\ y^{p} &= y_{i} + \frac{h}{2} \dot{y}_{i} \\ \text{If} \quad y^{p} &\leq 0, \text{ then } 0 \leq \dot{y}_{i+1}^{e} \perp \lambda_{i+1} \geq 0 \end{split}$$

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 A simple example : A bouncing ball

The Bouncing ball example with time-stepping



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- Introduction Low order systems Higher order systems Numerical Methods Conclusions
 Open Problems and links with optimization
 - * Efficient algorithm for the LCP with a switched-off constraints :

$$y = A\lambda + b$$

 $v = h(y)$
If $y \leq 0$, then $0 \leq v \perp \lambda \geq 0$

- Issue ? : Reformulation in terms of QP with a additional slack variable (MIP)?
- \rightarrow Good BVP solvers for which a prediction of y is not reasonable

Introduction Low order systems Higher order systems Numerical Methods Conclusions
 Open Problems and links with optimization

* Efficient algorithm for the LCP with a switched-off constraints :

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- Subscription Structure Structure

$$\begin{array}{rcl} y &=& A\lambda + b \\ y &=& [y_n, y_t], \quad \lambda = [\lambda_n, \lambda_t] \\ v &=& h(y) = [\lambda_n, \lambda_t] \\ \end{array}$$

If $y_n &\leq& 0, {\rm then} \begin{pmatrix} 0 \leq v_n \perp \lambda_n \geq 0 \\ -\lambda_t \in \partial \Psi^*_{C(\mu\lambda_n)}(v_n) \end{pmatrix}$

We use basic and robust iterative scheme (Gauss-Seidel like) and (Non smooth) Generalized Newton Method (Alart and Curnier, 1990)

- Issue ? : Try to find a good potential to minimize and/or a good Lagrangian relaxation ?
 - NLCP solvers ? Bundle Methods ?
 - Good line searches/trust regions for Generalized Newton Method ?

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Energetic coefficient of restitution e and multiple impact law



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Introduction Low order systems Higher order systems Numerical Methods Conclusions
 Open Problems and links with optimization (Continued)

* Energetic coefficient of restitution e and multiple impact law Find $(u, v, \lambda) \in \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^m$ giving $M \succ 0, H, \Theta, b$:

$$\begin{array}{rcl} M(u-b) &=& H\lambda \\ u^T M u &=& e \, b^T M b^T, & (\text{Energy dissipation}) \\ v &=& H^T u \geq 0 \\ \lambda &\geq& 0 \\ \Theta\lambda &=& 0, & (\text{Multiple Impact Law at distance}) \end{array}$$

We can also add friction : Find $(u, v, \lambda \in \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^m$ giving $M \succ 0, H, \Theta, b$:

$$\begin{split} M(u-b) &= H[\lambda_{n},\lambda_{t}]^{T} \\ u^{T}Mu &= e b^{T}Mb^{T}, & \text{Energy dissipation} \\ v &= [v_{n},v_{t}]^{T} = H^{T}u \geq 0, \\ \lambda_{n} &\geq 0 \\ \Theta\lambda_{n} &= 0 \quad \text{(Multiple Impact Law at distance)} \\ -\lambda_{t} &\in \partial\Psi_{C(\mu\lambda_{n})}^{*}(v_{t}) \end{split}$$

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 Open Problems and links with optimization (Continued)

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$$\begin{split} M(u-b) &= H[\lambda_n, \lambda_t]^T \\ u^T M u &= e \, b^T M b^T, & \text{Energy dissipation} \\ v &= [v_n, v_t]^T = H^T u \ge 0, \\ \lambda_n &\ge 0 \\ \Theta \lambda_n &= 0 \quad (\text{Multiple Impact Law at distance}) \\ -\lambda_t &\in \partial \Psi^*_{C(\mu\lambda_n)}(v_t) \end{split}$$

***** Extension to Non linear mechanical behavior $y = f(\lambda)$ We use only outer linearization with Newton-Raphson scheme

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- → 3 Higher-order systems: Formulation and Time-discretization
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 - 3.2 Preliminary example on LCS
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 - 3.4 Issues to be fixed
 - 3.5 Canonical form : The Zero Dynamical form
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 - 3.7 Measure differential dynamics
 - 3.8 Reinitialization mapping
 - 3.9 Well posedness results
 - 3.10 Time-discretization
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 - 4 Higher-order systems: Numerical Methods, Applications and links with Optimization
- 5 Conclusions and Perspectives

Introduction

Joint Work with :

- Bernard Brogliato, Head of the Bipop Project, INRIA Rhône-Alpes
- Daniel Goeleven, IREMIA, University of La Réunion

References :

- V. Acary and B. Brogliato, *Higher Order Moreau's sweeping process*, Colloquium in the honor of the 80th Birthday of J.J. Moreau, to appear in "Non smooth Mechanics and Analysis: theoretical and numerical advances", Kluwer, 2005
- V. Acary, B. Brogliato and D. Goeleven, Higher Order Moreau's sweeping process: Mathematical formulation and numerical simulation, INRIA Research Report RR-5236, submitted to MPA
- J.S Pang and D. Stewart, Differential Variational Inequalities, preprint, submitted to MPA
 - Elegant Formulations of Unbounded Differential inclusion as Variational Inequalities
 - IVP and BVP
 - New proof of convergence for time-stepping scheme
 - But only for low order systems (\leq 1)

Preliminary example on LCS

Linear complementarity system :

$$\begin{cases} \dot{x} = Ax + B\lambda, & x \in \mathbb{R}^n, \lambda \in \mathbb{R}^m \\ y = Cx + D\lambda \\ 0 \le y \perp \lambda \ge 0 \end{cases}$$

* Let us consider the very simple example :

$$\begin{cases} \ddot{x} = \lambda, & x \in \mathbb{R}, \lambda \in \mathbb{R} \\ 0 \le y = x \perp \lambda \ge 0 \end{cases}$$
(-8)

Naive Remarks:

- If x(t) = 0 and $\dot{x}(t^-) < 0$, $\ddot{x}(t^-) < 0$, $\ddot{x}(t^-) < 0$ then all of the derivatives must jump.
- If x have a jump, x is a measure (Dirac) and x a derivative (in the sense of distribution) of a Dirac.
- In this case, λ is also a derivative of a Dirac and then there is no sense to require that $\lambda \ge 0$

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Introduction Low order systems Higher order systems Numerical Methods Conclusions Notion of Relative degree

* Definition : Defining the Markov Parameters as $(D, CB, CAB, CA^2B, ...)$, the relative degree r is the rank of the first non zero Markov Parameter.

℁ Remarks

- the Relative degree r is the number of differentiation of y to obtain explicitly y in function of λ .
- Clear Analogy with the differential index in DAE ($\delta = r + 1$)

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 Notion of Relative degree(Continued ...)

- **※** Relative degree r = 0, *D* ≻ 0, Trivial case
 - The multiplier $\lambda = \max(0, -D^{-1}Cx)$ is a Lipschitz continuous function of x
 - The numerical integration may be performed with any standard ODE solvers.

- Notion of Relative degree(Continued ...)
 - ***** Relative degree r = 0, $D \succ 0$, Trivial case
 - The multiplier $\lambda = \max(0, -D^{-1}Cx)$ is a Lipschitz continuous function of x
 - The numerical integration may be performed with any standard ODE solvers.
 - ***** Relative degree r = 1, D = 0, $CB \succ 0$
 - The multiplier λ is a function of time t, not necessarily continuous, for instance, of bounded variations (BV).
 - The numerical integration have to be performed with specific solvers (Event–Driven or Moreau's Catching up algorithm)

- Notion of Relative degree(Continued ...)
 - ***** Relative degree r = 0, $D \succ 0$, Trivial case
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 - The multiplier λ is a function of time t, not necessarily continuous, for instance, of bounded variations (BV).
 - The numerical integration have to be performed with specific solvers (Event–Driven or Moreau's Catching up algorithm)
 - ***** Relative degree r = 2, D = 0, CB = 0, $CAB \succ 0$
 - The system is not self-consistent : Need a re-initialization mapping
 - The multiplier λ is a real measure.
 - Specific solvers (Event–Driven or Moreau's Time–stepping) as for Lagrangian dynamical system with constraints

- Notion of Relative degree(Continued ...)
 - ***** Relative degree r = 0, $D \succ 0$, Trivial case
 - The multiplier $\lambda = \max(0, -D^{-1}Cx)$ is a Lipschitz continuous function of x
 - The numerical integration may be performed with any standard ODE solvers.
 - ***** Relative degree r = 1, D = 0, $CB \succ 0$
 - The multiplier λ is a function of time t, not necessarily continuous, for instance, of bounded variations (BV).
 - The numerical integration have to be performed with specific solvers (Event–Driven or Moreau's Catching up algorithm)
 - ***** Relative degree r = 2, D = 0, CB = 0, $CAB \succ 0$
 - The system is not self-consistent : Need a re-initialization mapping
 - The multiplier λ is a real measure.
 - Specific solvers (Event–Driven or Moreau's Time–stepping) as for Lagrangian dynamical system with constraints
 - * Higher Relative degree $r \ge 3$ $D = 0, CB = 0, CA^{r-2} = 0, \dots, CA^{r-1}B \succ 0$
 - The multiplier λ is a distribution of order r-1.
 - Dedicated time-stepping scheme and nested complementarity problems

Issues to be fixed

- Reformulation of the problem as :
 - Canonical form (Zero-Dynamics)
 - Distributional dynamical systems
 - Measure differential equations (also possibly Measure variational Inequalities)
 - "Good" Reinitialization mapping (Monotone mapping)
- * Characterization of solutions
- Mathematical results Existence and uniqueness
- Time-stepping scheme for IVP and BVP
- # Efficient Algorithm for Nested Complementarity Problems

Assumptions :

- * Autonomous and linear time invariant systems
- Homogeneous relative degree

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 Canonical form: The Zero Dynamical form (ZD)

Let us consider the following LTI system :

$$\begin{cases} \dot{x} = Ax + B\lambda, & x \in \mathbb{R}^n, \lambda \in \mathbb{R}^m \\ w = Cx + D\lambda \end{cases}$$

We perform a state-space transformation $z = Wx, z^T = (w, \dot{w}, \dots, w^{(r-1)}, \xi)$ such that :

$$\begin{cases} \dot{z}_{1}(t) = z_{2}(t) \ (t \ge 0) \\ \dot{z}_{2}(t) = z_{3}(t) \ (t \ge 0) \\ \dot{z}_{3}(t) = z_{4}(t) \ (t \ge 0) \\ \vdots \\ \dot{z}_{r-1}(t) = z_{r}(t) \ (t \ge 0) \\ \dot{z}_{r}(t) = CA^{r}W^{-1}z(t) + CA^{r-1}B\lambda(t) \ (t \ge 0) \\ \dot{\xi}(t) = A_{\xi}\xi(t) + B_{\xi}z_{1}(t) \ (t \ge 0) \\ w(t) = z_{1}(t) \ (t \ge 0) \end{cases}$$

This transformation always exists for controllable systems

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Introduction Low order systems Higher order systems Distributional Dynamics

Let us consider a system of equality distributions of Class, $\cup_{n \in \mathbb{N}} \mathcal{T}_n(I)$,

$$\begin{cases}
Dz_{1} = z_{2} \\
Dz_{2} = z_{3} \\
Dz_{3} = z_{4}
\end{cases}$$

$$\Rightarrow
\begin{cases}
Dz_{1} = \{z_{2}\} + \nu_{1} \\
Dz_{2} = \{z_{3}\} + D\nu_{1} + \nu_{2} \\
Dz_{3} = \{z_{4}\} + D^{2}\nu_{1} + D\nu_{2} + \nu_{3}
\end{cases}$$

$$\Rightarrow
\begin{bmatrix}
Dz_{r-1} = z_{r} \\
Dz_{r} = CA^{r}W^{-1}z + CA^{r-1}B\lambda \\
D\xi = A_{\xi}\xi + B_{\xi}z_{1}.
\end{cases}$$

$$\Rightarrow
\begin{cases}
Dz_{r-1} = \{z_{r}\} + D^{(r-1)}\nu_{1} + D^{(i-2)}\nu_{2} + \dots + D\nu_{i-1} + \nu_{i} \\
\vdots \\
Dz_{r-1} = \{z_{r}\} + D^{(r-2)}\nu_{1} + \dots + \nu_{r-1} \\
Dz_{r} = CA^{r}W^{-1}\{z\} + CA^{r-1}B\lambda. \\
D\xi = A_{\xi}\xi + B_{\xi}z_{1}.
\end{cases}$$
(-7)

where ν_i the measure part of the distribution Dz_i This now possible to give a meaning to the positivity of λ :

$$\lambda = (CA^{r-1}B)^{-1} [D^{(r-1)}\nu_1 + \ldots + D\nu_{r-1}] + \nu_r$$

by imposing some constraints of positivity to ν_i

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Measure differential dynamics

Stronger Assumption("weaker" formalism) : requiring that the solutions z_i of the distributional dynamics are regular distributions z_i generated by right continuous functions of special locally bounded variation.

More precisely, $\xi_1, ..., \xi_{n-r} \in \mathcal{F}_{\infty}(\mathrm{I\!R}^+; \mathrm{I\!R})$ such that

$$\begin{cases} dz_1 = z_2(t)dt + d\nu_1 \\ dz_2 = z_3(t)dt + d\nu_2 \\ dz_3 = z_4(t)dt + d\nu_3 \\ \vdots \\ dz_i = z_{i+1}(t)dt + d\nu_i \\ \vdots \\ dz_{r-1} = z_r(t)dt + d\nu_{r-1} \\ dz_r = CA^r W^{-1} z(t)dt + CA^{r-1} Bd\nu_r \\ \dot{\xi}(t) = A_{\xi}\xi(t) + B_{\xi}z_1(t) \end{cases}$$

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Reinitialization mapping

Introduction Low order systems

Befinition of tangent cone to Φ : Let Φ be a nonempty closed convex subset of \mathbb{R} . We denote by $T_{\Phi}(x)$ the tangent cone of Φ at $x \in \mathbb{R}$ defined by

$$T_{\Phi}(x) = \overline{\text{cone}}(\Phi - \{x\}) \tag{-6}$$

Higher order systems Numerical Methods Conclusions

where $cone(\Phi - \{x\})$ denotes the cone generated by $\Phi - \{x\}$. This definition allows us to take into account constraints violations. Note that

$$T_{\mathrm{I\!R}^+}(x) = \begin{cases} \mathrm{I\!R} & \mathrm{if} \quad x > 0 \\ \mathrm{I\!R}^+ & \mathrm{if} \quad x \le 0 \end{cases} \quad \mathrm{and} T_{\mathrm{I\!R}}(x) = \mathrm{I\!R}.$$

* Definition of nested tangent cones: Let us now set $\Phi := \mathbb{R}^+$. For $z \in \mathbb{R}^r$, we set $Z_i = (z_1, z_2, ..., z_i), \quad (1 \le i \le r)$. We define

$$T^{0}_{\Phi}(Z_{1}) = \Phi, \quad T^{1}_{\Phi}(Z_{1}) = T_{\Phi}(z_{1}), \quad T^{2}_{\Phi}(Z_{2}) = T_{T^{1}_{\Phi}(Z_{1})}(z_{2}), \dots T^{i}_{\Phi}(Z_{i}) = T_{T^{i-1}_{\Phi}(Z_{i-1})}(z_{i}).$$

Definition of the Reinitialization mapping :

$$d\nu_i \in -\partial \psi_{T_{\Phi}^{i-1}(\{Z_{i-1}\}(t^-))}(\{z_i\}(t^+)) \quad \text{on } \tilde{I}, \quad (1 \le i \le r)$$
(-5)

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Reinitialization mapping Continued ...

Interpretation of this inclusion

If $T_{\Phi}^{i-1}(\{Z_{i-1}\}(t^{-})) = \mathbb{R}^{+}$, i.e., if $z_1 \leq 0, z_2 \leq 0, ..., z_i \leq 0$ then one gets a complementarity condition :

$$0 \le d\nu_i \perp \{z_i\}(t^+) \ge 0$$

otherwise

 $d\nu_i = 0$

 \rightarrow we obtain a set of nested complementarity conditions (Generalization of r = 2):

$$\begin{array}{l} 0 \leq d\nu_1 \perp \{z_1\}(t^+) \geq 0 \\ \text{if } z_1 \leq 0 \text{ then } 0 \leq d\nu_2 \perp \{z_2\}(t^+) \geq 0 \\ \text{if } z_1 \leq 0 \text{ and } z_2 \leq 0 \text{ then } 0 \leq d\nu_3 \perp \{z_3\}(t^+) \geq 0 \\ \text{f } z_1 \leq 0 \text{ and } z_2 \leq 0 \text{ and } z_3 \leq 0 \text{ then } 0 \leq d\nu_4 \perp \{z_4\}(t^+) \geq 0 \end{array}$$

Well posedness results

Definition of Regular solution :

Let $0 \le a < b \in \mathbb{R} \cup \{+\infty\}$ be given. We say that a solution $z \in (\mathcal{T}_{r-1}(\mathbb{R}^+))^n$ of Measure differential Inclusions is regular on [a, b) if for each $t \in [a, b)$, there exists a right neighborhood $[t, \sigma)$ ($\sigma > 0$) such that the restriction of $\{z\}$ to $[t, \sigma)$ is analytic.

* Global Existence and Uniqueness of a Regular Solution

Suppose that $CA^{r-1}B \succ 0$. For each $z_0 \in \mathbb{R}^n$, the system of Measure differential Inclusions has at least one regular solution.

Moreover:

i)
$$z_1\equiv\{z_1\}\geq 0$$
 on ${
m I\!R}^+$

ii)
$$\{\bar{z}\}(0^+) = \bar{z}_0'$$

iii) $\|\{z\}(t)\| \le e^{\|WAW^{-1}\|t}\|z_0\|, \ \forall t \in \mathbb{R}^+$

iv) If z^1 and z^2 are two regular solutions then $\langle z^1, \varphi \rangle = \langle z^2, \varphi \rangle, \ \forall \varphi \in C_0^{\infty}(\mathbb{R}^+; \mathbb{R}^n).$

Time-discretization

Summary of the Measure Differential inclusion :

$$\begin{cases} dz_i - z_{i+1}(t)dt = d\nu_i, 1 \le i \le r - 1\\ dz_r - CA^r W^{-1} z(t)dt = (CA^{r-1}B)^{-1} d\nu_r\\ d\nu_i \in -\partial \psi_{T_{\Phi}^{i-1}(z_1(t^{-1}), \dots, z_{i-1}(t^{-1}))}(z_i(t^{+1}))\\ \dot{\xi}(t) dt = A_{\xi}\xi(t) + B_{\xi}z_1(t) dt \end{cases}$$

Time discretization :

We denote by $0 = t_0 < t_1 < \ldots < t_k < t_N = T$ a finite partition (or a subdivision) of the time interval [0, T], T > 0 and the time step is $h = t_{k+1} - t_k$

The values of the measures $dz_i((t_k, t_{k+1}])$ and $\mu_{i,k+1} = d\nu_i((t_k, t_{k+1}])$ are kept as primary variables and this fact is crucial for the consistency of the method for the non smooth evolutions.

$$\begin{cases} z_{i,k+1} - z_{i,k} - hz_{i+1,k+1} = \mu_{i,k+1}.\\ z_{r,k+1} - z_{r,k} - hCA^{r}W^{-1}z_{k+1} = CA^{r-1}B \ \mu_{r,k+1}\\ \mu_{i,k+1} \in -\partial \psi_{T_{\Phi}^{i-1}(z_{1,k},\dots,z_{i-1,k})}(z_{i,k+1})\\ \xi_{k+1} - \xi_{k} = hA_{\xi}\xi_{k+1} + hB_{\xi}z_{1,k+1} \end{cases}$$

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Properties of the Time-discretization

Introduction Low order systems

***** Proposition 1: Boundedness of the sequences of approximation (z_k, ξ_k) :

 $||z_n|| \le \alpha, \quad ||\mu_k||| \le M$

* Proposition 2: Local Bounded Variation of step function $Z_i(t)$ generated by the approximation z_i on a interval [0, T]

$$\operatorname{var}(z_i^N, [0, T]) \le \frac{1}{2R}(|z_{i,0} - a| + h\alpha)^2 + \frac{\alpha^2}{2R}T^2 + \alpha T(1 + \frac{1}{R}|z_{i,0} - a|) \qquad \text{for all } 1 \le i \le r - 1$$

$$/\mathrm{Or}(z_r^N, [0, T]) \le \frac{1}{2R}(|z_{r,0} - a| + h\beta\alpha)^2 + \frac{\beta^2 \alpha^2}{2R}T^2 + \beta\alpha T(1 + \frac{1}{R}|z_{1,0} - a|)$$

 $\mathrm{Var}(\xi^{\scriptscriptstyle N},[0,T]) \leq (\gamma+\delta)\alpha T$

→ Helly's Theorem : There is a subsequence of (z_k, ξ_k) that converges point-wisely towards to some function $z(t), \xi(t)$ which is of Local Bounded variations

Still to be done:

- Prove that this limit is a solution of Measure differential inclusion
- Choose and define a correct topology to measure convergence between two filled in graphs of BV functions. (Hausdorff distance)
- After that, the convergence of the scheme and the order are straightforward corollaries due to the existence and uniqueness properties of the problem

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 - 4.2 Applications
 - 4.3 Empirical Order
 - 4.4 Open Problems
 - 5 Conclusions and Perspectives

A simple example with a non trivial zero-dynamics:

$$\begin{aligned} z(0) &= (1, 0, 0, 0, 0)^T \\ \dot{z}_1(t) &= z_2(t) \\ \dot{z}_2(t) &= z_3(t) \\ \dot{z}_3(t) &= -z_1(t) - z_2(t) - z_3(t) - d_{\xi}^T \xi(t) + \lambda(t) \\ \dot{\xi}_1(t) &= \alpha \xi_2(t) \\ \dot{\xi}_2(t) &= -\omega \xi_1(t) + z_1(t) \\ w(t) &= z_1(t) \ge 0 \end{aligned}$$

A simple example with a non trivial zero-dynamics:

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For $d_{\xi} = (0, 1)$ the zero dynamic by plays role in the global dynamics $\alpha = 1$ and $\omega = 1$.

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Applications

- * Electrical and Mechanical systems with feedback Control Loop
 - The Feedback loop may increase the relative degree of the system
- Indirect Methods for Optimal control with state constraints : Finite difference BVP solvers
 - We can prove that the relative degree of the Necessary condition system is twice the original one of the system to be controlled.
 For a mechanical system (r=2), the necessary conditions for Optimality leads to a dynamical system of relative equal to r = 4.
- Advantages of the approach :
 - Take into account accumulation of events.
 - Do not need any first guess for the algorithm
 - Theoretical results

Empirical Order

* This error is measure using the l_{∞} norm between the step function generated by the sequences of approximation



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Empirical Order

* This error is now measure using a Hausdorff distance between filled-in graph of BV function.



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Introduction Low order systems Higher order systems Numerical Methods Conclusions Open Problems * Efficient Algorithms for the Multi-level nested Complementarity problem $d\nu_i \in -\partial \psi_{T_{\Phi}^{i-1}(\{Z_{i-1}\}(t^-))}(\{z_i\}(t^+)) \text{ on } \tilde{I}, \quad (1 \leq i \leq r)$



$$0 \le d\nu_1 \perp \{z_1\}(t^+) \ge 0$$

if $z_1 \le 0$ then $0 \le d\nu_2 \perp \{z_2\}(t^+) \ge 0$
if $z_1 \le 0$ and $z_2 \le 0$ then $0 \le d\nu_3 \perp \{z_3\}(t^+) \ge 0$
if $z_1 \le 0$ and $z_2 \le 0$ and $z_3 \le 0$ then $0 \le d\nu_4 \perp \{z_4\}(t^+) \ge 0$

- Non linear and Non autonomous systems
- # Higher order Time integration scheme



- ✓ 1 Introduction on Non Smooth Dynamical systems
- ✓ 2 Historical background on low order systems
- ✓ 3 Higher-order systems: Formulation and Time-discretization
- ✓ 4 Higher-order systems: Numerical Methods, Applications and links with Optimization
- \rightarrow 5 Conclusions and Perspectives

Conclusion

There is a lot of stuff to do in the field of Non Smooth Dynamical systems

I would be very grateful if someone could provide some advises and references which come from the optimization community

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