

# About Nonlinear Sampled-data Dynamics Representations

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*Présentation Groupe ShY 10 décembre 2014*

## Motivations

### *Continuous-time Dynamics:*

$$\dot{x}(t) = f(x(t), u(t)) \quad y(t) = h(x(t)); \quad f : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n; h : \mathbb{R}^n \rightarrow \mathbb{R}^p$$

### *Discrete -Time Dynamics*

$$x_{k+1} = F(x_k, u_k); \quad y_k = H(x_k); \quad F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n; H : \mathbb{R}^n \rightarrow \mathbb{R}^p$$

### *Sampled-data Dynamics*

$$x_{k+1} = F^\delta(x_k, u_k); \quad y_k = H^\delta(x_k) \quad F^\delta, H^\delta \text{ parameterized by } \delta \in ]0, T^*]$$

$$u(t) = u_k = Cst; t \in [k\delta, (k+1)\delta[ \text{ ZOH}; k = (0, 1, \dots); \delta : \text{sampling period}$$

**Find equivalent representations : A unified mathematical framework**

## Hybrid Dynamics as

- *CT and DT Dynamics with Switches or Jumps*
- *Interconnection of CT and DT Dynamics - Networks*
- *Sampled-data Dynamics in a Computer Aided Environment*

$$\begin{aligned} \dot{x}(t) &= f(x(t), u_k) & u(t) &= u_k = Cst; t \in [k\delta, (k+1)\delta[ \\ x^+ &= x_k & t &= k\delta; k = 0, 1, \dots \end{aligned}$$

- *Variable Sampling : Event-triggered dynamics*

**vast literature, variety of concepts, approaches, interdisciplinary**

## Sampling : a Unifying Tool

- *To get a unified representation and unified mathematical framework*
- *To deal with Computer Aided analysis and design - unavoidable*
- *To make feasible the CT solutions - further to the Zero Order Holding device*
- *To provide new solutions taking advantage of sampling - multirate approaches*
- *To provide computable solutions through constructive algorithms*

## Outline of the presentation

- *Nonlinear sampling and equivalent sampled-data representations*
- *Exponential form representations of the behaviors*
- *A new Differential-Difference representation of DT Dynamics - DDR*
- *Some Impacts in Control Theory*
- *Some Computational Open Questions*

— *From chronological calculus to exponential representations of continuous and discrete-time dynamics: a Lie algebraic approach, TAC, 52, 2227-2241, (2007). S. Monaco, D. Normand-Cyrot and C. Califano*

— *Advanced Tools for Nonlinear Sampled-Data Systems' Analysis and Control, Special Issue "Fundamental Issues in Control" , EJC, 13 (2,3), 221-241, (2007). S. Monaco, D. Normand-Cyrot*

## Nonlinear sampling and equivalent sampled-data representations

*Given:*  $\dot{x}(t) = f(x(t), u(t))$  *CD*

*with:*  $u(t) = u_k = Cst$  *ZOH*

(over  $[k\delta, (k+1)\delta[$ ;  $k = (0, 1, \dots)$ ;  $\delta$  : sampling period)

*Find a Nonlinear Difference Equation:*

$$x_{k+1} = F^\delta(x_k, u_k)$$

*such that:*  $x_k = x(t = k\delta)$  *when*  $x_0 = x(0)$

(input-state reproduction at the sampling instants)

*(sampled- data equivalent model )*

● *Extensions: Multi-Input, Higher Order Holding device, Parameter or Time Varying Dynamics, ...*

- *formulate sampling as formal ODE Integration:  $\dot{x}(t) = f(x(t), u(t))$*

- *derive the sampled-data equivalent model*

→ in a computable state-space form through formal power series expansions in  $\delta$  → easy approximations

→ in terms of the VF  $(f, g)$  and their Lie brackets  $(ad_f^i g, ad_g ad_f^i g, \dots)$  : basic tools for characterizing structural and control properties in continuous-time

- *in the form of two coupled differential difference equations rather than in the form of a map*

→ more suitable to characterize structural and control properties.

## *The mathematical framework*

→ *Formal power series*

→ *Exponential series and Lie algebraic methods*

→ *Chronological calculus and combinatorics ...*

→ in Mathematics : *Ree (1958), Magnus (1954), Chen (1961), Grobner (1973), ...*

→ in the Control theory ( from about 40 years in CT - more recently in DT): *Agrachev, Gamkrelidze (1978), Sussmann (1984), Fliess (1986), Tretyak (1998), Kawski (1999), Sarychev (2001), ....*

→ in Computer Sciences



## some notations

- Vector fields (complete) and functions smooth in  $x$  and analytic  $t$  and/or  $u$
- the Lie derivative of  $\lambda$  along  $\theta$  at  $x$ :

$$L_\theta \lambda|_x := \frac{\partial \lambda}{\partial x}|_x \theta(x) = \theta(\lambda)(x)$$

- $L_\theta^p \lambda := L_{\theta \dots \theta} \lambda$ ;  $p$  – times  $L_\theta^0 = 1$  the identity operator
- the exponential Lie series wrt  $L_\theta$  :

$$e^{L_\theta} = 1 + \sum_{p \geq 1} \frac{1}{p!} L_\theta^p = e^\theta; \quad e^\theta h|_x = h|_{x=e^\theta x}$$

- the Lie bracket:  $[\theta, \sigma] := \text{ad}_\theta \sigma = L_\theta \circ L_\sigma - L_\sigma \circ L_\theta$
- for  $p > 1$ :  $\text{ad}_\theta^p(\sigma) := [\theta, \text{ad}_\theta^{p-1}(\sigma)]$
- all the results are formal, no specific convergence study

*single Input-Affine dynamics for simplicity:*

$$\dot{x}(t) = f(x(t)) + u_k g(x(t))$$

Uncontrolled Dynamics -  $u = 0$

$$\begin{aligned} \dot{x}(t) &= f(x(t)); & x(t = k\delta) &= x_k \\ x_{k+1} &= x_k + \int_0^\delta f(x(\tau)) d\tau = e^{\delta f} x_k \\ e^{\delta f} x_k &= F^\delta(x_k) = x_k + \sum_{i \geq 1} \frac{\delta^i}{i!} F_i(x_k) \end{aligned}$$

The Linear Case :  $\dot{x}(t) = Ax(t) \quad \rightarrow \quad x_{k+1} = A^\delta x_k = e^{\delta A} x_k$

## Idem for Controlled Dynamics under ZOH

$$\begin{aligned}\dot{x}(t) &= f(x(t), u_k) & u(t) &= u_k, t \in [k\delta, (k+1)\delta[ \\ x_{k+1} &= x_k + \int_{k\delta}^{(k+1)\delta} f(x(\tau), u_k) d\tau \\ e^{\delta f(\cdot, u_k)} x_k &= F^\delta(x_k, u_k) = F_0^\delta(x_k) + \sum_{i \geq 1} \frac{u_k^i}{i!} F_i^\delta(x_k)\end{aligned}$$

- $F^\delta(\cdot, u_k)$  : *exact sampled equivalent to  $f(\cdot, u_k)$  (asymptotic expansion in  $\delta$ )*
- *closed form solutions* do not exist in general

Sampled equivalent to I-A dynamics as a map

$$\dot{x}(t) = f(x(t)) + u_k g(x(t))$$

$$x_{k+1} = F^\delta(x_k, u_k) = e^{\delta(f+u_k g)} x_k$$

→ *nonlinearity in*  $u$  :  $F^\delta(\cdot, u) = F_0^\delta + \sum_{p \geq 1} \frac{u^p}{p!} F_p^\delta$

→ *integro-differential formula of Poincaré* - (MNC, 1984)

$$F_0^\delta(x) = e^{\delta f} x$$

$$F_p^\delta(x) = p \int_0^\delta e^{(\delta-s)f} L_g F_{p-1}^s(x) ds \quad p \geq 1$$

## The Linear Case

$$\begin{aligned}\dot{x}(t) = Ax(t) + Bu_k &\quad \rightarrow \quad x_{k+1} = A^\delta x_k + B^\delta u_k \\ A^\delta = e^{\delta A} &= I + \delta A + \frac{\delta^2}{2!}A^2 + \dots + \frac{\delta^i}{i!}A^i + \dots \\ B^\delta = \int_0^\delta e^{\tau A} B d\tau &= \delta B + \frac{\delta}{2!}AB + \dots + \frac{\delta^i}{i!}A^{i-1}B + \dots\end{aligned}$$

linearity in  $x$  and in  $u$  preserved

exponential of matrices : linear algebra

computational facilities

## time-varying dynamics

$$\dot{x}(t) = f(t, x(t)) = f_0(x) + \sum_{i \geq 1} \frac{t^i}{i!} f_i(x)$$

*integration*    ↓    *over*  $[0, \delta[$

$$x(\delta) = \int_0^\delta f(\tau, x(\tau)) d\tau = \overrightarrow{\exp} \int_0^\delta f(\tau, \cdot) d\tau |_{x_0}$$

*the chronological exponential series solution :*

$$\overrightarrow{\exp} \int_0^\delta f(\tau, \cdot) d\tau = 1 + \int_0^\delta f(\tau_1, \cdot) d\tau_1 + \sum_{m \geq 1} \int_0^\delta d\tau_1 \int_0^{\tau_1} \dots \int_0^{\tau_{m-1}} f(\tau_m, \cdot) \circ \dots \circ f(\tau_1, \cdot) d\tau_m$$

how to compute the chronological exponential series ?

how to express the  $F_i^\delta$ 's in terms of the  $f_i$ 's ?

**R1** *an ordinary exponential series solution :*

$$\overrightarrow{\exp} \int_0^\delta f(\tau, \cdot) d\tau = e^{\delta \mathcal{F}(f^\delta)} = F^\delta = 1 + \sum_{i \geq 1} \frac{\delta^i}{i!} F_i$$

## the exponential Lie series solution

- $F^\delta$  is described by an ordinary exponential of operators :  
from *Integro/Differential (Poincaré)* to *Formal Lie series (Gröbner, 1973)*

$$\overrightarrow{\exp} \int_0^\delta f(\tau, \cdot) d\tau = e^{\delta \mathcal{F}(f^\delta)} = F^\delta = 1 + \sum_{i \geq 1} \frac{\delta^i}{i!} F_i$$

↓↑

$$\delta \mathcal{F}(f^\delta) = \text{Log}\left(1 + \sum_{i \geq 1} \frac{\delta^i}{i!} F_i\right)$$

- $\delta \mathcal{F}(f^\delta)$  admits a Lie series expansion :  $\rightarrow$  Lie brackets of VF :  $\text{ad}_{f_i} f_j = [f_i, f_j]$

$$\delta \mathcal{F}(f^\delta) = \sum_{i \geq 1} \frac{\delta^i}{i!} B_i(f_0, \dots, f_{i-1}) = \delta B_1 + \frac{\delta^2}{2!} B_2 + \frac{\delta^3}{3!} B_3 + \dots$$

with:  $B_1 = f_0;$

$B_2 = f_1$

$B_3 = f_2 + 1/2 \text{ad}_{f_0} f_1$

$B_4 = f_3 + \text{ad}_{f_0} f_2;$   $B_5 = f_4 + \frac{3}{2} \text{ad}_{f_0} f_3 + \text{ad}_{f_1} f_2 + \frac{1}{2} \text{ad}_{f_1}^2 f_0 + \frac{1}{6} \text{ad}_{f_0}^2 f_2 - \frac{1}{6} \text{ad}_{f_0}^3 f_1$

- Each  $B_i$  is a Lie polynomial of degree  $i > 1$  in the  $f_j$ 's :

$$B_{i+1} = f_i + \sum_{0 \leq q < i} \sum_{\substack{l_1, \dots, l_j \geq 1 \\ \sum p(l_p) + q = i}} \frac{i! (-1)^j b_j}{j! q! l_1! \dots l_j!} \text{ad}_{B_{l_1}} \circ \dots \circ \text{ad}_{B_{l_j}} f_q.$$

## exact and approximate sampled models

- exact sampled dynamics in the form of asymptotic series

$$1 + \delta F_1 + \frac{\delta^2}{2!} F_2 + \frac{\delta^3}{3!} F_3 + \dots = e^{\delta \mathcal{F}(f^\delta)} = e^{\delta f_0 + \frac{\delta^2}{2!} f_1 + \frac{\delta^3}{3!} (f_2 + 1/2 \text{ad}_{f_0} f_1) + \dots}$$

→ be carefull !  $e^{\delta \mathcal{F}(f^\delta)} \neq e^{\int_0^\delta f(\tau, \cdot) d\tau}$

→ conditions on the Lie algebra generated by the  $f_i$  directly modify the series exponent

- usual truncation :

$$\begin{aligned} F_a^\delta &= 1 + \delta F_1 + \frac{\delta^2}{2!} F_2 + \frac{\delta^3}{3!} F_3 + O(\delta^4) \\ &= 1 + \delta f_0 + \frac{\delta^2}{2!} (f_0^2 + f_1) + \frac{\delta^3}{3!} (f_0^3 + 2f_0 \cdot f_1 + f_1 \cdot f_0 + f_2) + O(\delta^4) \\ \tilde{F}_a^\delta &= 1 + \delta f_0 \qquad \text{Euler scheme at the order 1 in } \delta \end{aligned}$$

- truncation of the exponent series :

$$\begin{aligned} F_b^\delta &= e^{\delta f_0 + \frac{\delta^2}{2!} f_1 + \frac{\delta^3}{3!} (f_2 + 1/2 \text{ad}_{f_0} f_1) + O(\delta^4)} \\ \tilde{F}_b^\delta = e^{\delta f_0 + O(\delta^2)} &\leftarrow \dot{x}(t) = f_0(x(t)) \qquad \text{Euler type scheme} \end{aligned}$$

- polynomial approximations in  $x$



## why to look for an exponential form

$$\dot{x}(t) = f(t, x(t)) \longleftrightarrow x_{k+1} = e^{\delta \mathcal{F}_k^\delta}(x_k)$$

...

- exponential structure maintained for any mapping  $y = h(x)$  :

$$y_{k+1} = h(x_{k+1}) = e^{\delta \mathcal{F}_k^\delta} h|_{x_k}$$

- exponential suitable for composition over several steps :

$$x_{k+1} = e^{\delta \mathcal{F}_0^\delta} \circ \dots \circ e^{\delta \mathcal{F}_k^\delta}(x_0)$$

...

**Part 2** : Discrete-Time dynamics as coupled Differential/Difference equations - DDR

$$x_{k+1} = F(x_k, u_k)$$

Differential/Difference Representation DDR  $(F_0, G(., u))$

*ODE in  $u$  for the forced evolution :*

$$\frac{dx^+(u)}{du} = G(x^+(u), u)$$

*with initial condition given by the discrete mapping - jump:*

$$x \rightarrow x^+(0) = F_0(x)$$

## *Key novel ideas:*

- $x^+(u)$  is a curve in  $R^n$  parameterized by  $u \in \mathcal{U}$
- $G(\cdot, u)$ : analytic and complete VF on  $R^n$  parameterized by  $u$ , specifies the tangent VF of the curve  $x^+(u)$  (continuous  $u$ -dependency); forced evolution imbedded into an ODE
- $F_0$ : analytic on  $R^n$ ; the free dynamics induces a jump
- Nonlinearity in  $u$  of the ODE:  $G(\cdot, u) = G_1 + \sum_{i \geq 1} \frac{u^i}{i!} G_i$

→ Hybrid state space representation of NL DT D!

→ Unified framework: Continuous-Time, Discrete-Time and Sampled versus Hybrid Dynamics

How to compute the evolution ?

$$\frac{dx^+(u)}{du} = G(x^+(u), u) \quad (1)$$

$$x^+(0) = F_0(x) \quad (\text{initial jump}) \quad (2)$$

*step 0: initial conditions*  $(x_0, u_0, \dots, u_p)$

*step 1: compute*  $F(x_0)$  *according to* (2) *(jump) - integrate* (1) *between* 0 *and*  $u_0$  *with initialization at*  $x^+(0) = F(x_0)$  *to get*

$$x_1 = x^+(u_0)$$

*step 2: compute*  $x^+(0) = F_0(x_1)$  *from*  $x_1$  *according to* (2), *integrate* (1) *between* 0 *and*  $u_1$  *with initialization at*  $x^+(0) = F(x_1)$  *to get*

$$x_2 = x^+(u_0, u_1)$$

.. *step p: Iterate the construction to find for*  $p \geq 3$

$$x_p = x^+(u_0, \dots, u_{p-1})$$

## About the existence of the DDR ?

- *Directly from modeling*
- *Under the existence of  $G(., u)$  on  $R^n \times \mathcal{U}$  s.t.*

$$G(x, u)|_{F(x, u)} = \frac{\partial F(x, u)}{\partial u} \quad (A) \text{ satisfied (locally) for :}$$

- *Drift invertible dynamics (equivalently there exists  $u_0$  s.t.  $F(., u_0)$  invertible)*
- *Submersive dynamics (equivalent under feedback to drift invertible ones)*

- *Sampled dynamics OK :*

$$F(x, 0) = e^{\delta f} x \quad F^{-1}(x, 0) = e^{-\delta f} x$$

## Two Equivalent State-Space Representations of DT Dynamics

Given a mapping :  $(x, u) \rightarrow F(x, u)$

derivation  $\downarrow$  with respect to  $u$

Let:  $G(., u)$  such that:  $G(., u)|_{F(x,u)} = \frac{\partial F(x, u)}{\partial u}$

(provided  $F(x, u)$  invertible (OK under sampling))

$\updownarrow$

*Reciprocally given a DDR :*

$$\frac{dx^+(u)}{du} = G(x^+(u), u)$$

$$x^+(0) = F_0(x)$$

*integration ↓ with respect to u*

$$x^+(u) = \overrightarrow{\exp} \int_0^u G(x^+(0), v) dv \quad (\text{chronological exponential})$$

$$x^+(u) = x^+(0) + \sum_{m \geq 1} \int_0^u dv_1 \dots \int_0^{v_{m-1}} dv_m G(., v_m) \circ \dots \circ G(x^+(0), v_1)$$

$$F(x, u) = e^{u\mathcal{G}(.,u)} x|_{F_0(x)}$$

$u\mathcal{G}(., u)$  : *explicitely computable Lie series* - (MNCC,2002)



## More about integration

Theorem :

$$x^+(u) = \phi(u, 0, x^+(0)) = \Phi(u, 0)(x^+(0))$$

- $\Phi(u, 0)$  : *(flow) admits an exponential representation:*

$$\Phi(u, 0) = e^{u\mathcal{G}(\cdot, u)}$$

$$u\mathcal{G}(\cdot, u) = \sum_{i \geq 1} \frac{u^i}{i!} \mathcal{B}_i = u\mathcal{B}_1 + \frac{u^2}{2} \mathcal{B}_2 + \frac{u^3}{3!} \mathcal{B}_3 + \frac{u^4}{4!} \mathcal{B}_4 + O(u^5)$$

- $\mathcal{B}_i$  : *homogeneous Lie polynomial of degree  $i$  in the  $G_i$*

$$\mathcal{B}_{i+1} = G_{i+1} + \sum_{0 \leq k < i} \sum_{\substack{l_1, \dots, l_j \geq 1 \\ \sum_p (l_p) + k = i}} \frac{(-1)^{j_i} b_j}{j! k! l_1! \dots l_j!} \text{ad}_{\mathcal{B}_{l_1}} \circ \dots \circ \text{ad}_{\mathcal{B}_{l_j}} G_{k+1}$$

$$\mathcal{B}_1 = G_1; \quad \mathcal{B}_2 = G_2; \quad \mathcal{B}_3 = G_3 + 1/2[G_1, G_2]; \quad \mathcal{B}_4 = G_4 + [G_1, G_3]$$

$$\begin{aligned} \mathcal{B}_5 &= G_5 + \frac{3}{2}[G_1, G_4] + [G_2, G_3] + \frac{1}{6}[G_1[G_1, G_3]] \\ &\quad - \frac{1}{2}[G_2[G_1, G_2]] - \frac{1}{6}[G_1[G_1[G_1, G_2]]] \end{aligned}$$



Geometric differential technics and tools can be definitively used to  
handle NLDTD and thus NLSD

## DDR under Sampling

## DDR under Sampling

**Theorem** : *Given*  $\dot{x}(t) = f(x(t)) + u_k g(x(t))$

*There exists*  $(F_0^\delta, G^\delta(\cdot, u))$  *such that*

$$\begin{aligned}\frac{dx^+(u)}{du} &= G^\delta(x^+(u), u) \\ x^+(0) &= F_0^\delta(x) = e^{\delta f} x\end{aligned}$$

$$G^\delta(\cdot, u) = \int_0^\delta e^{-s a d_{f+ug}} g ds = \delta g + \sum_{i \geq 1} \frac{(-1)^i \delta^{i+1}}{(i+1)!} a d_{f+ug}^i g$$

$$G^\delta(\cdot, u) = G_1^\delta + \sum_{i \geq 1} \frac{u^i}{i!} G_i^\delta$$

*(linearity in  $u$  lost under sampling)*

Iterative computations of the  $G_i^\delta$ :  $G^\delta(., u) = G_1^\delta + \sum_{i \geq 1} \frac{u^i}{i!} G_i^\delta$

$$G_1^\delta = \int_0^\delta e^{-s \operatorname{ad}_f} g ds = \delta g + \sum_{i \geq 1} \frac{(-1)^i \delta^{i+1}}{(i+1)!} \operatorname{ad}_f^i g \quad (u = 0)$$

$$G_1^\delta = \delta g - \delta \operatorname{ad}_f g + \frac{\delta^2}{2!} \operatorname{ad}_f^2 g - \frac{\delta^3}{3!} \operatorname{ad}_f^3 g + O(\delta^4)$$

$$G_1^\delta = \sum_{i \geq 0} \frac{\delta^{i+1}}{(i+1)!} X_i \in \mathcal{L}_0 \quad (\text{setting } X_i = (-1)^i \operatorname{ad}_f^i g)$$

$$\begin{aligned} G_2^\delta &= - \int_0^\delta ds_1 \int_0^{s_1} e^{-s \operatorname{ad}_f} \llbracket \operatorname{ad}_g g \rrbracket ds \in [\mathcal{L}_0, \mathcal{L}_0] \\ &= \sum_{i \geq 2} \frac{(-1)^i \delta^{i+1}}{(i+1)!} \operatorname{ad}_f^{i-1} \llbracket \operatorname{ad}_g g \rrbracket \end{aligned}$$

$\mathcal{L}_0$  : Lie ideal of  $\mathcal{L}(f, g)$  generated by  $g$  (combinatoric proofs)

Iterative computations of the  $G_i^\delta$ : (follow)

$$\begin{aligned}
 G_{p+1}^\delta &= (-1)^p p! \int_0^\delta ds_p \int_0^{s_p} \dots \int_0^{s_1} e^{-s \text{ad}_f} \lrcorner \text{ad}_g^p g ds \\
 &= p! \sum_{i \geq p+1} \frac{(-1)^i \delta^{i+1}}{(i+1)!} \text{ad}_f^{i-p} \lrcorner \text{ad}_g^p g \in \mathcal{L}_0^{p+1}
 \end{aligned}$$

→ *Combinatorial proofs*

the shuffle product  $\lrcorner$  (integrations by parts) is defined in a recursive way:

$$\mathbf{1} \lrcorner \zeta_i = \zeta_i \lrcorner \mathbf{1} = \zeta_i, \quad \zeta_i \lrcorner \zeta_j = \zeta_j \lrcorner \zeta_i = \zeta_i \cdot \zeta_j + \zeta_j \cdot \zeta_i$$

$$\begin{aligned}
 \zeta_{i_1} \dots \zeta_{i_m} \lrcorner \zeta_{j_1} \dots \zeta_{j_p} &:= \zeta_{i_1} (\zeta_{i_2} \dots \zeta_{i_m} \lrcorner \zeta_{j_1} \dots \zeta_{j_p}) \\
 &+ \zeta_{j_1} (\zeta_{i_1} \dots \zeta_{i_m} \lrcorner \zeta_{j_2} \dots \zeta_{j_p})
 \end{aligned}$$

**More about the need of formal calculus**

## The integro-differential method

$$\frac{\partial G^\delta(\cdot, u)}{\partial u} = \int_0^\delta \left[ \frac{\partial G^s(\cdot, u)}{\partial s}, G^s(\cdot, u) \right] ds$$

$$G_2^\delta = \int_0^\delta \left[ \frac{\partial G_1^s}{\partial s}, G_1^s \right] ds = -G_1^\delta * G_1^\delta \quad (u = 0)$$

\* : *chronological product (non associative)*

$$G_2^\delta = \sum_{i \geq 2} \frac{\delta^{i+1}}{(i+1)!} \sum_{k=0}^{i-1} C_i^k [X_k, X_{i-k-1}] \quad \text{with} \quad C_i^k := \frac{i!}{k!(i-k)!}$$

$$G_2^\delta = -\frac{\delta^3}{3!} [X_0, X_1] - \frac{2\delta^4}{4!} [X_0, X_2] - \frac{3\delta^5}{5!} [X_0, X_3] - \frac{2\delta^5}{5!} [X_1, X_2] + o(\delta^6)$$



Iteratively for the other vector fields :

$$G_{p+1}^\delta = (G_p^\delta)^{(+)}$$

(the formal rule  $(\cdot)^{(\cdot)}$  "mimics" derivation with respect to  $u$ )

$$G_3^\delta = (G_2^\delta)^{(+)} = \int_0^\delta \left[ \frac{\partial (G_1^s)^{(\cdot)}}{\partial s}, G_1^s \right] ds + \int_0^\delta \left[ \frac{\partial G_1^s}{\partial s}, (G_1^s)^{(\cdot)} \right] ds$$

$$G_3^\delta = \int_0^\delta \left[ \frac{\partial G_2^s}{\partial s}, G_1^s \right] ds + \int_0^\delta \left[ \frac{\partial G_1^s}{\partial s}, G_2^s \right] ds = \frac{3}{2} G_1^\delta * (G_1^\delta * G_1^\delta)$$

$$G_3^\delta = \frac{2\delta^4}{4!} [X_0[X_0, X_1]] + \frac{6\delta^5}{5!} [X_0[X_0, X_2]] + \frac{2\delta^5}{5!} [X_1[X_0, X_1]] + 0(\delta^6)$$

A formal recursive law for computing the  $G_i^\delta$ 's

Starting from:  $G_1^\delta = \sum_{i \geq 0} \frac{\delta^{i+1}}{(i+1)!} X_i \in \mathcal{L}_0$

Setting:  $X_0^+ = X_1^+ = 0;$

$$X_i^+ := \sum_{k=0}^{i-1} C_i^k [X_k, X_{i-k-1}]$$

Compute iteratively :

$$\begin{aligned} G_2^\delta &= \sum_{i \geq 0} \frac{\delta^{i+1}}{(i+1)!} X_i^+ = \sum_{i \geq 2} \frac{\delta^{i+1}}{(i+1)!} \sum_{k=0}^{i-1} C_i^k [X_k, X_{i-k-1}] \in [\mathcal{L}_0, \mathcal{L}_0] \\ &= -\frac{\delta^3}{3!} [X_0, X_1] - \frac{2\delta^4}{4!} [X_0, X_2] - \frac{3\delta^5}{5!} [X_0, X_3] - \frac{2\delta^5}{5!} [X_1, X_2] + o(\delta^6) \end{aligned}$$

$$\begin{aligned}
G_3^\delta &= \sum_{i \geq 2} \frac{\delta^{i+1}}{(i+1)!} X_i^{+++} \\
&= \frac{2\delta^4}{4!} [X_0[X_0, X_1]] + \frac{6\delta^5}{5!} [X_0[X_0, X_2]] + \frac{2\delta^5}{5!} [X_1[X_0, X_1]] + o(\delta^6) \\
G_4^\delta &= \sum_{i \geq 2} \frac{\delta^{i+1}}{(i+1)!} X_i^{++++} \\
&= -\frac{6\delta^5}{5!} [X_0[X_0, [X_0, X_1]]] + o(\delta^6) \\
&\dots
\end{aligned}$$

→ *algorithmic procedure for computing exact models*

→ *approximated solutions computed through series truncations are sufficient in practice*

**some impact in analysis and control design**

*A link between the map  $e^{\delta f + \delta u g}$  and the DDR*

*Given :*

$$\frac{dx^+(u)}{du} = G^\delta(x^+(u), u)$$

$$x^+(0) = e^{\delta f} x$$

*under integration*      $\downarrow$      *with respect to  $u$*

$$x^+(u) = e^{u\mathcal{G}^\delta(.,u)}(x^+(0))$$

*with series exponent :  $u\mathcal{G}^\delta(.,u) = \sum_{p \geq 1} \frac{u^p}{p!} \mathcal{B}_p^\delta(G_1^\delta, \dots, G_p^\delta)$*

$\mathcal{B}_p^\delta(G_1^\delta, \dots, G_p^\delta)$  : *homogeneous Lie polynomial of degree  $p$ .*

$$\mathcal{B}_1^\delta = G_1^\delta; \quad \mathcal{B}_2^\delta = G_2^\delta; \quad \mathcal{B}_3^\delta = G_3^\delta + 1/2[G_1^\delta, G_2^\delta]; \quad \mathcal{B}_4^\delta = G_4^\delta + [G_1^\delta, G_3^\delta]$$

$$\begin{aligned} \mathcal{B}_5^\delta &= G_5^\delta + \frac{3}{2}[G_1^\delta, G_4^\delta] + [G_2^\delta, G_3^\delta] + \frac{1}{6}[G_1^\delta[G_1^\delta, G_3^\delta]] \\ &\quad - \frac{1}{2}[G_2^\delta[G_1^\delta, G_2^\delta]] - \frac{1}{6}[G_1^\delta[G_1^\delta[G_1^\delta, G_2^\delta]]] \end{aligned}$$

$$\begin{aligned} x^+(u) &= e^{\delta f} \circ e^{u\mathcal{G}^\delta(\cdot, u)} x = e^{\delta f + \delta u g} x \\ &= e^{\delta f} \circ e^{uG_1^\delta + \frac{u^2}{2}G_2^\delta + \frac{u^3}{3!}(G_3^\delta + 1/2[G_1^\delta, G_2^\delta]) + O(u^4)} x \end{aligned}$$

*Decomposition of the dynamics around the free evolution*

## The behaviour over several time-steps as Multi-Input SD

*Given* :  $x^+(u_0, u_1) = e^{\delta f + u_0 \delta g} \circ e^{\delta f + u_1 \delta g} x_0$ , *the DDR is* :

$$\begin{aligned} \frac{\partial x^+(u_0, u_1)}{\partial u_0} &= G_{u_1}^\delta(x^+(u_0, u_1), u_0) \\ \frac{\partial x^+(u_0, u_1)}{\partial u_1} &= G^\delta(x^+(u_0, u_1), u_1) \\ x^+(0) &= e^{2\delta f} x_0 \end{aligned}$$

→ *A set of forced partial derivatives equations*

*with*  $G_{u_1}^\delta(\cdot, u_0)$  : *transport of*  $G^\delta(\cdot, u_0)$  *along*  $e^{\delta f + \delta u_1 g}$

*idem*  $Ad_{G_i}^{p\delta f}$  : *transport of*  $G_i^\delta$  *along*  $e^{p\delta f}$  MNC (1984), Jakubczyk-NC(1984),

Sontag Albertini(1990)

In this sampled context, we get :

$$\begin{aligned}
 G_{u_1}^\delta(\cdot, u_0) &:= e^{-ad_{\delta f + \delta u_1 g}} G^\delta(\cdot, u_0) \\
 Ad_{G_i}^{p\delta f} &:= e^{-ad_{p\delta f}} G_i^\delta = e^{-p\delta f} \circ G_i^\delta \circ e^{p\delta f} \\
 &= G_i^{(p+1)\delta} - G_i^{p\delta}
 \end{aligned}$$

- under successive compositions (1<sup>st</sup> order approximations in  $u$  of the series exponent):

$$x_p = e^{p\delta f} \circ e^{u_0 Ad_{\delta f}^p G_1^\delta} \circ \dots \circ e^{u_p G_1^\delta} x_0$$



$\mathcal{L}(G_1^\delta, \dots, Ad_{G_1}^{p\delta f}, \dots)$ : defines the Lie algebra of controllability

$$\begin{pmatrix} G_1^\delta \\ Ad_{G_1}^{\delta f} \\ \dots \\ Ad_{G_1}^{(p-1)\delta f} \end{pmatrix} = \begin{pmatrix} \delta & \delta^2/2! & \dots & \delta^p/p! \\ 2\delta & (2\delta)^2/2! & \dots & (2\delta)^p/p! \\ \dots & \dots & \dots & \dots \\ p\delta & (p\delta)^2/2! & \dots & (p\delta)^p/p! \end{pmatrix} \begin{pmatrix} X_0 \\ X_1 \\ \dots \\ X_{p-1} \end{pmatrix}$$

when  $X_i = (-1)^i ad_f^i g$

## *A fundamental tool to study*

- Local Controllability/Observability criteria → State-Space decomposition
- Characterization of Invariant Structures, Parameterized Canonical forms, Passivity or Losslessness
- Control Geometric Properties as Feedback Linearizability, Decoupling with Stability under Static/Dynamic feedback, Regulation, Disturbance Rejection
- Homological-like Equations → Approximated Normal Forms and approximated Feedback Linearizability
- Strong parallelism between discrete-time dynamics

$$\frac{dx^+(u)}{du} = G_1(x^+(u)) \quad x^+(0) = F_0(x)$$

and input-affine continuous-time dynamics

$$\dot{x}(t) = f(x(t)) + u(t)g(x(t))$$

## About some extensions

## About some extensions

- Nonlinearity in  $u$  :  $\dot{x}(t) = f(x(t), u_k)$  (ZOH)

$$G^\delta(\cdot, u_k) = \int_0^\delta e^{-sA} \frac{\partial f(\cdot, u_k)}{\partial u_k} ds$$

- Multi-input case :  $\dot{x}(t) = f(x(t), u_{1k}, \dots, u_{mk})$

$${}_m G^\delta(\cdot, \underline{u}_k) = \int_0^\delta e^{-sA} \frac{\partial f(\cdot, \underline{u}_k)}{\partial u_{mk}} ds$$

## About some extensions ( follow)

- Time-varying case :  $\dot{x}(t) = f(x(t), t, u_k)$

$$G^\delta(\cdot, u_k) = \int_0^\delta e^{-s\mathcal{A}(\cdot, s, u_k)} \frac{\partial f(\cdot, s, u_k)}{\partial u_k} ds$$

with :  $x(\delta) = e^{\delta\mathcal{F}(\cdot, \delta, u_k)} x_0$

closed-loop, parameter dependent or perturbed dynamics, ...

- Higher Order Holding Device  $\rightarrow$  Multi-Input SD

$$\dot{x}(t) = f(x(t), u_k + t\dot{u}_k) \quad \dot{u}_k := \left. \frac{du(t)}{dt} \right|_{t=k\delta} \quad \text{1st order device}$$

$\rightarrow$  The sampled dynamics depends on  $u_k$  and  $\dot{u}_k$

$\rightarrow$  Digital design over  $u_k$  and  $\dot{u}_k$  (piecewise discontinuous control)

## Summaries : Sampling is useful :

- *To discuss how structural and control properties are (can be) maintained under sampling*
- *To design multirate control strategies for maintaining under sampling control properties or solving difficult problems (no standard smooth continuous-time design) ...*
- *To set state space representation of hybrid dynamics*
- *Favorite examples: motion planning or articulated mechanical structures (nonholonomic constraints) and walking robots (switching dynamics)*
- *Computational aspects → Approximated models through series truncations*

## Some Examples

*Chained dynamics under sampling (ex. on  $R^3$ )*

$$\dot{x}_1(t) = u_1 \quad \dot{x}_2(t) = u_2 \quad \dot{x}_3(t) = x_2 u_1$$

$\mathcal{L}(g_1, g_2)$  non involutive but  $\rho(g_1, g_2, [g_1, g_2]) = 3$

*Exact Driftless sampled equivalent ( $x_i^+(0) = x_i$ )*

$$\begin{pmatrix} \frac{dx_1^+(u)}{du_1} \\ \frac{dx_2^+(u)}{du_1} \\ \frac{dx_3^+(u)}{du_1} \end{pmatrix} = \begin{pmatrix} \delta \\ 0 \\ \delta x_2^+(u) - \frac{\delta^2 u_2}{2} \end{pmatrix}; \quad \begin{pmatrix} \frac{dx_1^+(u)}{du_2} \\ \frac{dx_2^+(u)}{du_2} \\ \frac{dx_3^+(u)}{du_2} \end{pmatrix} = \begin{pmatrix} 0 \\ \delta \\ \frac{\delta^2 u_1}{2} \end{pmatrix}$$



*with partial derivatives forced equations*

$$G_1^\delta(\cdot, u) = (\delta, 0, \delta x_2 - \frac{\delta^2}{2}u_2)^T \quad G_2^\delta(\cdot, u) = (0, \delta, \frac{\delta^2 u_1}{2})^T$$

↓

$$\rho(G_1^\delta(\cdot, u), G_2^\delta(\cdot, u), [G_1^\delta(\cdot, u), G_2^\delta(\cdot, u)]) = 3$$

↓

*Full inversion under multirate of order 2*

*→ exact state steering over 2 time steps*

*Idem for various extensions of chained/driftless dynamics*

*→ exact finite sampling*

*Polynomial Normal Forms (ex. quadratic on  $R^3$ )*

$$\dot{x}_1(t) = x_2 + ax_3^2 \quad \dot{x}_2(t) = x_3 \quad \dot{x}_3(t) = u$$

$$X_0 = g = (0, 0, 1)^T; \quad X_1 = -ad_f g = (2ax_3, 1, 0)^T$$

$$X_2 = ad_f^2 g = (1, 0, 0)^T; \quad [X_0, X_1] = (0, 0, 2a)^T; \quad X_{i \geq 3} = 0$$

*Sampled equivalent:*

$$\begin{pmatrix} \frac{dx_1^+(u)}{du} \\ \frac{dx_2^+(u)}{du} \\ \frac{dx_3^+(u)}{du} \end{pmatrix} = \begin{pmatrix} \delta^2 ax_3^+(u) + \frac{\delta^3}{3!} \\ \frac{\delta^2}{2!} \\ \delta \end{pmatrix} - u \begin{pmatrix} \frac{\delta^3 a}{3} \\ 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} x_1^+ \\ x_2^+ \\ x_3^+ \end{pmatrix} = \begin{pmatrix} x_1 + \delta x_2 + \delta a x_3^2 + \frac{\delta^2 x_3}{2} \\ x_2 + \delta x_3 \\ x_3 \end{pmatrix}$$

→ *Exact sampled equivalent of finite degree in  $u$  and  $\delta$*

→ *Quadraticity in  $(x, u)$  is maintained*

$$\begin{aligned} G_1^\delta(x) &= \delta X_0 + \frac{\delta^2}{2!} X_1 + \frac{\delta^3}{3!} X_2 \\ G_2^\delta(x) &= -\frac{\delta^3}{3!} [X_0, X_1]; \quad G_{i \geq 3}^\delta(x) = 0 \\ Ad_{G_1}^{\delta f}(x) &= \delta X_0 + \frac{3\delta^2}{2!} X_1 + \frac{7\delta^3}{3!} X_2 = G_1^{2\delta}(x) - G_1^\delta(x) \\ Ad_{G_1}^{2\delta f}(x) &= 2\delta X_0 + \frac{3\delta^2}{2!} X_1 + \frac{7\delta^3}{3!} X_2 = G_1^{3\delta}(x) - G_1^{2\delta}(x) \end{aligned}$$