

# Developer's Notes

Siconos Development Team

January 11, 2011

# Chapter 1

## First Order Nonlinear Relation

|         |              |
|---------|--------------|
| author  | O. Bonnefon  |
| date    | July, 1 2009 |
| version | Kernel 3.0.0 |

## Chapter 2

# OneStepNSProblem formalisation for several interactions

|         |              |
|---------|--------------|
| author  | F. Pérignon  |
| date    | May 16, 2006 |
| version | ?            |

### 2.1 LinearDS - Linear Time Invariant Relations

#### 2.1.1 General notations

We consider  $n$  dynamical systems of the form:

$$\dot{x}_i = A_i x_i + R_i \quad (2.1)$$

Each system is of dimension  $n_i$ , and we denote  $N = \sum_{i=1}^n n_i$ .

An interaction,  $I_\alpha$  is composed with a non smooth law,  $nslaw_\alpha$  and a relation:

$$y_\alpha = C_\alpha X_\alpha + D_\alpha \lambda_\alpha \quad (2.2)$$

The “dimension” of the interaction, ie the size of vector  $y_\alpha$ , is denoted  $m_\alpha$  and we set:

$$M = \sum_{\alpha=1}^m m_\alpha$$

$m$  being the number of interactions in the Non Smooth Dynamical System.

$X_\alpha$  is a vector that represents the DS concerned by the interaction. Its dimension is noted  $N_\alpha$ , this for  $n_\alpha$  systems in the interaction.

$C_\alpha$  is a  $m_\alpha \times N_\alpha$  row-blocks matrix and  $D_\alpha$  a  $m_\alpha \times m_\alpha$  square matrix.

$$C_\alpha = \begin{bmatrix} C_\alpha^i & C_\alpha^j & \dots \end{bmatrix} \quad (2.3)$$

with  $i, j, \dots \in \mathcal{DS}_\alpha$  which is the set of DS belonging to interaction  $\alpha$ .

We also have the following relation:

$$\begin{bmatrix} R_\alpha^i \\ R_\alpha^j \\ \dots \end{bmatrix} = B_\alpha \lambda_\alpha = \begin{bmatrix} B_\alpha^i \\ B_\alpha^j \\ \dots \end{bmatrix} \lambda_\alpha \quad (2.4)$$

$R_\alpha^i$  represents the contribution of interaction  $\alpha$  on the reaction of the dynamical system  $i$ , and  $B_\alpha^i$  is a  $n_i \times m_\alpha$  block matrix.

And so:

$$R_i = \sum_{\beta \in \mathcal{I}_i} R_\beta^i = \sum_{\beta \in \mathcal{I}_i} B_\beta^i \lambda_\beta \quad (2.5)$$

with  $\mathcal{I}_i$  the set of interactions in which dynamical system number  $i$  is involved.

Introducing the time discretisation, we get:

$$x_i^{k+1} - x_i^k = hA_i x_i^{k+1} + hR_i^{k+1} \quad (2.6)$$

$$y_\alpha^{k+1} = C_\alpha X_\alpha^{k+1} + D_\alpha \lambda_\alpha^{k+1} \quad (2.7)$$

$$R_i^{k+1} = \sum_{\beta \in \mathcal{I}_i} B_\beta^i \lambda_\beta^{k+1} \quad (2.8)$$

ie, with  $W_i = (I - hA_i)^{-1}$ :

$$x_i^{k+1} = W_i x_i^k + hW_i R_i^{k+1} \quad (2.9)$$

$$y_\alpha^{k+1} = C_\alpha W_\alpha X_\alpha^k + C_\alpha hW_\alpha \sum_{\beta \in \mathcal{I}_i} B_\beta^i \lambda_\beta^{k+1} + D_\alpha \lambda_\alpha^{k+1} \quad (2.10)$$

$$= C_\alpha W_\alpha X_\alpha^k + (C_\alpha hW_\alpha B_\alpha + D_\alpha) \lambda_\alpha^{k+1} + \sum_{\beta \neq \alpha} \left( \sum_{i \in \mathcal{DS}_\alpha \cap \mathcal{DS}_\beta} hC_\alpha^i W_i B_\beta^i \lambda_\beta^{k+1} \right) \quad (2.11)$$

with

$$W_\alpha = \begin{bmatrix} W_i & 0 & \dots \\ 0 & W_j & \dots \\ 0 & \dots & \dots \end{bmatrix} \quad (2.12)$$

the block-diagonal matrix of all the  $W$  for the dynamical systems involved in interaction  $\alpha$ .

The global-assembled  $Y$  vector, of dimension  $M$ , composed by  $m$   $y_\alpha$  subvectors, is given by:

$$Y_{k+1} = q_{OSNSP} + M_{OSNSP} \Lambda_{k+1} \quad (2.13)$$

or,

$$Y_{k+1} = \begin{bmatrix} y_1 \\ \dots \\ y_m \end{bmatrix}_{k+1} = \begin{bmatrix} C_1^1 & \dots & C_1^n \\ \vdots & \dots & \vdots \\ C_m^1 & \dots & C_m^n \end{bmatrix} \begin{bmatrix} W_1 & 0 & \dots & 0 \\ 0 & W_2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & W_n \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ \vdots \\ x_n \end{bmatrix}_k \quad (2.14)$$

$$+ \begin{bmatrix} D_1 + h \sum_{j \in \mathcal{DS}_1} C_1^j W_j B_1^j & h \sum_{j \in \mathcal{DS}_1 \cap \mathcal{DS}_2} C_1^j W_j B_2^j & \dots \\ \vdots & \ddots & \\ h \sum_{j \in \mathcal{DS}_m} C_m^j W_j B_{m-1}^j & D_m + h \sum_{j \in \mathcal{DS}_m \cap \mathcal{DS}_{m-1}} C_m^j W_j B_m^j \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_m \end{bmatrix}_k$$

To sum it up, the block-diagonal term of matrix  $M_{OSNSP}$ , for block-row  $\alpha$  is:

$$D_\alpha + h \sum_{j \in \mathcal{DS}_\alpha} C_\alpha^j W_j B_\alpha^j \quad (2.15)$$

This is an  $m_\alpha \times m_\alpha$  square matrix. The extra-diagonal block term, in position  $(\alpha, \beta)$  is:

$$h \sum_{j \in \mathcal{DS}_\alpha \cap \mathcal{DS}_\beta} C_\alpha^j W_j B_\beta^j \quad (2.16)$$

and is a  $m_\alpha \times m_\beta$  matrix. This matrix differs from 0 when interactions  $\alpha$  and  $\beta$  are coupled, ie have common DS.

Or, for the relation  $l$  of interaction  $\alpha$ , we get:

$$D_{\alpha,l} + h \sum_{j \in \mathcal{DS}_\alpha} C_{\alpha,l}^j W_j B_\alpha^j \quad (2.17)$$

for the diagonal, and

$$h \sum_{j \in \mathcal{DS}_\alpha \cap \mathcal{DS}_\beta} C_{\alpha,l}^j W_j B_\beta^j \quad (2.18)$$

for extra-diagonal terms.

$D_{\alpha,l}$ , row number  $l$  of  $D_\alpha$ , the same for  $C_{\alpha,l}$

Finally, the linked-Interaction map provides, for each interaction (named “current interaction”), the list of all the interactions (named “linked interaction”) that have common dynamical system with the “current interaction”.

### 2.1.2 A simple example

We consider  $n = 3$  dynamical systems and  $m = 2$  interactions:

$$\begin{aligned} I_\mu &\rightarrow \mathcal{DS}_\mu = \{DS_1, DS_3\}, m_\mu = 3 \\ I_\theta &\rightarrow \mathcal{DS}_\theta = \{DS_2, DS_3\}, m_\theta = 1 \end{aligned}$$

The linked-interaction map is :

$$\begin{aligned} I_\mu &\rightarrow I_\theta, \text{commonDS} = DS_3 \\ I_\theta &\rightarrow I_\mu, \text{commonDS} = DS_3 \end{aligned}$$

And:

$$M = 4, N = \sum_{i=1}^3 n_i$$

$$\mathcal{I}_1 = \{I_\mu\}$$

$$\mathcal{I}_2 = \{I_\theta\}$$

$$\mathcal{I}_3 = \{I_\mu, I_\theta\}$$

$$y_1 = \begin{bmatrix} C_1^1 & C_1^3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_3 \end{bmatrix} + D_1 \lambda_1 \quad (2.19)$$

$$y_2 = \begin{bmatrix} C_2^2 & C_2^3 \end{bmatrix} \begin{bmatrix} x_2 \\ x_3 \end{bmatrix} + D_2 \lambda_2 \quad (2.20)$$

$$\begin{bmatrix} R_1 \\ R_2 \\ R_3 \end{bmatrix} = \begin{bmatrix} B_1^1 \lambda_1 \\ B_2^2 \lambda_2 \\ B_1^3 \lambda_1 + B_2^3 \lambda_2 \end{bmatrix} \quad (2.21)$$

$$M_{OSNSP} = \begin{bmatrix} D_1 + hC_1^1 W_1 B_1^1 + hC_1^3 W_3 B_1^3 & hC_1^3 W_3 B_2^3 \\ hC_2^3 W_3 B_1^3 & D_2 + hC_2^2 W_2 B_2^2 + hC_2^3 W_3 B_2^3 \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix}_{k+1} \quad (2.22)$$

### 2.1.3 relative degree

Let us consider the global vector

$$Y = \begin{bmatrix} y_1 \\ \dots \\ y_M \end{bmatrix} = CX + D\Lambda \quad (2.23)$$

We denote  $r_j$  the relative degree of equation  $j$ ,  $j \in [1..M]$ . We have:

$$y_j = \sum_{i=1}^n C_j^i x_i + D_{j,j} \lambda_j + \sum_{i \neq j, i=1}^m D_{j,i} \lambda_i \quad (2.24)$$

$D_{j,i}$  a scalar and  $C_j^i$  a  $1 \times n_i$  line-vector.

If  $D_{jj} \neq 0$ , then  $r_j = 0$ . Else, we should consider the first derivative of  $y_j$ .

Before that, recall that:

$$R_i = \sum_{k=1}^M B_k^i \lambda_k \quad (2.25)$$

Through many of the  $B_j^i$  are equal to zero, we keep them all in the following lines.

Then:

$$\dot{y}_j = \sum_{i=1}^n C_j^i (A_i x_i + \sum_{k=1}^M B_k^i \lambda_k) + f(\lambda_k)_{k \neq j} \quad (2.26)$$

$$= \sum_{i=1}^n C_j^i (A_i x_i + B_j^i \lambda_j + \sum_{k=1, k \neq j}^M B_k^i \lambda_k) + \dots \quad (2.27)$$

So, if  $\sum_{i=1}^n C_j^i B_j^i \neq 0$  (note that this corresponds to the product between line  $j$  of  $C$  and column  $j$  of  $B$ ) then  $r_j = 1$  else we consider the next derivative, and so on.  
In derivative  $r$ , the coefficient of  $\lambda_j$  will be:

$$coeff_j = \sum_{i=1}^n C_j^i (A_i)^{r-1} B_j^i \quad (2.28)$$

if  $coeff_j \neq 0$  then  $r_j = r$ .

## 2.2 LagrangianDS - Lagrangian Linear Relations

### 2.2.1 General notations

We consider  $n$  dynamical systems, lagrangian and non linear, of the form:

$$M_i(q_i) \ddot{q}_i + N_i(\dot{q}_i, q_i) = F_{Int,i}(\dot{q}_i, q_i, t) + F_{Ext,i}(t) + p_i \quad (2.29)$$

Each system if of dimension  $n_i$ , and we denote  $N = \sum_{i=1}^n n_i$ .

An interaction,  $I_\alpha$  is composed with a non smooth law,  $nslaw_\alpha$  and a relation:

$$y_\alpha = H_\alpha Q_\alpha + b_\alpha \quad (2.30)$$

The “dimension” of the interaction, ie the size of vector  $y_\alpha$ , is denoted  $m_\alpha$  and we set:

$$M_y = \sum_{\alpha=1}^m m_\alpha$$

$m$  being the number of interactions in the Non Smooth Dynamical System.

$Q_\alpha$  is a vector that represents the DS concerned by the interaction. Its dimension is noted  $N_\alpha$ , this for  $n_\alpha$  systems in the interaction.

$H_\alpha$  is a  $m_\alpha \times N_\alpha$  row-blocks matrix and  $b_\alpha$  a  $m_\alpha$  vector.

$$H_\alpha = \begin{bmatrix} H_\alpha^i & H_\alpha^j & \dots \end{bmatrix} \quad (2.31)$$

with  $i, j, \dots \in \mathcal{DS}_\alpha$  which is the set of DS belonging to interaction  $\alpha$ .

We also have the following relation:

$$\begin{bmatrix} R_\alpha^i \\ R_\alpha^j \\ \dots \end{bmatrix} = {}^t H_\alpha \lambda_\alpha = \begin{bmatrix} {}^t H_\alpha^i \\ {}^t H_\alpha^j \\ \dots \end{bmatrix} \lambda_\alpha \quad (2.32)$$

$R_\alpha^i$  represents the contribution of interaction  $\alpha$  on the reaction of the dynamical system  $i$ , and  ${}^t H_\alpha^i$  is a  $n_i \times m_\alpha$  block matrix.

And so:

$$R_i = \sum_{\beta \in \mathcal{I}_i} R_\beta^i = \sum_{\beta \in \mathcal{I}_i} H_\beta^i \lambda_\beta \quad (2.33)$$

with  $\mathcal{I}_i$  the set of interactions in which dynamical system number  $i$  is involved.

Introducing the time discretisation, we get:

$$\begin{aligned} \dot{q}_i^{k+1} &= \dot{q}_{free,i} + W_i R_i^{k+1} \\ \dot{y}_\alpha^{k+1} &= H_\alpha Q_\alpha^{k+1} \end{aligned} \quad (2.34)$$

$$R_i^{k+1} = \sum_{\beta \in \mathcal{I}_i} H_\beta^i \lambda_\beta^{k+1} \quad (2.35)$$

ie,

$$y_\alpha^{k+1} = H_\alpha Q_\alpha^{free} + H_\alpha W_\alpha {}^t H_\alpha \lambda_\alpha + \sum_{i \in \mathcal{DS}_\alpha} \sum_{\beta \in \mathcal{I}_i, \alpha \neq \beta} H_\alpha^i W_i H_\beta^j \lambda_\beta \quad (2.36)$$

with  $W_\alpha$  given by (2.12).

The global-assembled  $Y$  vector, of dimension  $M$ , composed by  $m$   $y_\alpha$  subvectors, is given by:

$$Y_{k+1} = q_{OSNSP} + M_{OSNSP} \Lambda_{k+1} \quad (2.37)$$

with:

$$q_{OSNSP}^\alpha = H_\alpha Q_\alpha^{free} \quad (2.38)$$

and for  $M_{OSNSP}$ , the block-diagonal term for block-row  $\alpha$  is

$$\sum_{j \in \mathcal{DS}_\alpha} H_\alpha^j W_j {}^t H_\alpha^j \quad (2.39)$$

an  $m_\alpha \times m_\alpha$  square matrix. The extra-diagonal block term, in position  $(\alpha, \beta)$  is:

$$\sum_{j \in \mathcal{DS}_\alpha \cap \mathcal{DS}_\beta} H_\alpha^j W_j {}^t H_\beta^j \quad (2.40)$$

and is a  $m_\alpha \times m_\beta$  matrix. This matrix differs from 0 when interactions  $\alpha$  and  $\beta$  are coupled, ie have common DS.

Or, for the relation  $l$  of interaction  $\alpha$ , we get:

$$\sum_{j \in \mathcal{DS}_\alpha} H_{\alpha,l}^j W_j^t H_\alpha^j \quad (2.41)$$

for the diagonal, and

$$\sum_{j \in \mathcal{DS}_\alpha \cap \mathcal{DS}_\beta} H_{\alpha,l}^j W_j^t H_\beta^j \quad (2.42)$$

for extra-diagonal terms.

$H_{\alpha,l}$ , row number  $l$  of  $H_\alpha$ .

WARNING: depending on linear and non linear case for the DS, there should be a factor  $h$  ahead  $W$ . See Bouncing Ball template.

## 2.3 Block matrix approach

The built of the OSNSProblem matrix could be computed using block matrix structure. This section describe these matrices. It is not implemented in Siconos. Using previous notations,  $n$  is the number of DS.  $m$  the number of interactions.

### 2.3.1 Block matrix of DS

$$M\dot{X} = AX + R$$

where  $M = \text{diag}(M_1, \dots, M_n)$  and  $A = \text{diag}(A_1, \dots, A_n)$ .

$$R = B\lambda$$

$$B = \begin{pmatrix} B_1^1 \dots B_m^1 \\ \vdots \\ B_1^n \dots B_m^n \end{pmatrix}$$

Some of  $B_j^i$  doesn't exist.

### 2.3.2 Block matrix of interaction

$$Y = CX + D\lambda$$

with  $D = \text{diag}(D_1 \dots D_m)$  and

$$C = \begin{pmatrix} C_1^1 \dots C_m^1 \\ \vdots \\ C_m^1 \dots C_m^n \end{pmatrix}$$

Some of  $C_j^i$  doesn't exist.

### 2.3.3 OSNSProblem using block matrices

The Matrix of the OSNS Problem could be assambled using the following block-product-matrices  $CWB$ .



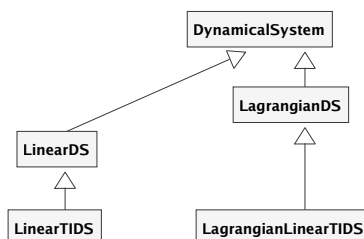
## Chapter 3

# Dynamical Systems formulations in Siconos.

|         |                |
|---------|----------------|
| author  | F. P rignon    |
| date    | March 22, 2006 |
| version | Kernel 1.1.4   |

### 3.1 Class Diagram

There are four possible formulation for dynamical systems in Siconos, two for first order systems and two for second order Lagrangian systems. The main class is `DynamicalSystem`, all other derived from this one, as shown in the following diagram:



### 3.2 General non linear first order dynamical systems

→ **class** *DynamicalSystem*

This is the top class for dynamical systems. All other systems classes derived from this one.

A general dynamical systems is described by the following set of  $n$  equations, completed with initial conditions:

$$\begin{aligned}\dot{x} &= f(x, t) + T(x)u(x, \dot{x}, t) + r \\ x(t_0) &= x_0\end{aligned}\tag{3.1}$$

(3.2)

- $x$ : state of the system - Vector of size  $n$ .
- $f(x, t)$ : vector field - Vector of size  $n$ .

- $u(x, \dot{x}, t)$ : control term - Vector of size  $uSize$ .
- $T(x)$ :  $n \times uSize$  matrix, related to control term.
- $r$ : input due to non-smooth behavior - Vector of size  $n$ .

The Jacobian matrix,  $\nabla_x f(x, t)$ , of  $f$  according to  $x$ ,  $n \times n$  square matrix, is also a member of the class.

Initial conditions are given by the member  $x_0$ , vector of size  $n$ . This corresponds to  $x$  value when simulation is starting, ie after a call to `strategy->initialize()`.

There are plug-in functions in this class for  $f$  (`vectorField`),  $jacobianX$ ,  $u$  and  $T$ . All of them can handle a vector of user-defined parameters.

### 3.3 First order linear dynamical systems $\rightarrow$ class *LinearDS*

Derived from `DynamicalSystem`, described by the set of  $n$  equations and initial conditions:

$$\dot{x} = A(t)x(t) + Tu(t) + b(t) + r \quad (3.3)$$

$$x(t_0) = x_0 \quad (3.4)$$

With:

- $A(t)$ :  $n \times n$  matrix, state independent but possibly time-dependent.
- $b(t)$ : Vector of size  $n$ , possibly time-dependent.

Other variables are those of `DynamicalSystem` class.  
 $A$  and  $B$  have corresponding plug-in functions.

Warning: time dependence for  $A$  and  $b$  is not available at the time in the simulation part for this kind of dynamical systems.

Links with `vectorField` and its Jacobian are:

$$f(x, t) = A(t)x(t) + b(t) \quad (3.5)$$

$$jacobianX = \nabla_x f(x, t) = A(t) \quad (3.6)$$

### 3.4 Second order non linear Lagrangian dynamical systems $\rightarrow$ class *LagrangianDS*

Lagrangian second order non linear systems are described by the following set of  $nDof$  equations + initial conditions:

$$M(q)\ddot{q} + NNL(\dot{q}, q) + F_{Int}(\dot{q}, q, t) = F_{Ext}(t) + p \quad (3.7)$$

$$q(t_0) = q_0 \quad (3.8)$$

$$\dot{q}(t_0) = velocity_0 \quad (3.9)$$

With:

- $M(q)$ :  $nDof \times nDof$  matrix of inertia.
- $q$ : state of the system - Vector of size  $nDof$ .
- $\dot{q}$  or *velocity*: derivative of the state according to time - Vector of size  $nDof$ .

- $NNL(\dot{q}, q)$ : non linear terms, time-independent - Vector of size  $nDof$ .
- $F_{Int}(\dot{q}, q, t)$ : time-dependent linear terms - Vector of size  $nDof$ .
- $F_{Ext}(t)$ : external forces, time-dependent BUT do not depend on state - Vector of size  $nDof$ .
- $p$ : input due to non-smooth behavior - Vector of size  $nDof$ .

The following Jacobian are also member of this class:

- $jacobianQFInt = \nabla_q F_{Int}(t, q, \dot{q})$  -  $nDof \times nDof$  matrix.
- $jacobianVelocityFInt = \nabla_{\dot{q}} F_{Int}(t, q, \dot{q})$  -  $nDof \times nDof$  matrix.
- $jacobianQNNL = \nabla_q NNL(q, \dot{q})$  -  $nDof \times nDof$  matrix.
- $jacobianVelocityNNL = \nabla_{\dot{q}} NNL(q, \dot{q})$  -  $nDof \times nDof$  matrix.

There are plug-in functions in this class for  $F_{int}$ ,  $F_{Ext}$ ,  $M$ ,  $NNL$  and the four Jacobian matrices. All of them can handle a vector of user-defined parameters.

Links with first order dynamical system are:

$$n = 2nDof \quad (3.10)$$

$$x = \begin{bmatrix} q \\ \dot{q} \end{bmatrix} \quad (3.11)$$

$$f(x, t) = \begin{bmatrix} \dot{q} \\ M^{-1}(F_{Ext} - F_{Int} - NNL) \end{bmatrix} \quad (3.12)$$

$$\nabla_x f(x, t) = \begin{bmatrix} 0_{nDof \times nDof} & I_{nDof \times nDof} \\ \nabla_q(M^{-1})(F_{Ext} - F_{Int} - NNL) - M^{-1}\nabla_q(F_{Int} + NNL) & -M^{-1}\nabla_{\dot{q}}(F_{Int} + NNL) \end{bmatrix} \quad (3.13)$$

$$r = \begin{bmatrix} 0_{nDof} \\ p \end{bmatrix} \quad (3.14)$$

$$u(x, \dot{x}, t) = u_L(\dot{q}, q, t) \text{ (not yet implemented)} \quad (3.15)$$

$$T(x) = \begin{bmatrix} 0_{nDof} \\ T_L(q) \end{bmatrix} \text{ (not yet implemented)} \quad (3.16)$$

$$(3.17)$$

$$(3.18)$$

With  $0_n$  a vector of zero of size  $n$ ,  $0_{n \times m}$  a  $n \times m$  zero matrix and  $I_{n \times n}$ , identity  $n \times n$  matrix.

Warning: control terms ( $Tu$ ) are not fully implemented in Lagrangian systems. This will be part of future version.

### 3.5 Second order linear and time-invariant Lagrangian dynamical systems $\rightarrow$ class *LagrangianLinearTIDS*

$$M\ddot{q} + C\dot{q} + Kq = F_{Ext}(t) + p \quad (3.19)$$

With:

- $C$ : constant viscosity  $nDof \times nDof$  matrix
- $K$ : constant rigidity  $nDof \times nDof$  matrix

And:

$$F_{Int} = C\dot{q} + Kq \quad (3.20)$$

$$NNL = 0_{nDof} \quad (3.21)$$

## Chapter 4

# Dynamical Systems implementation in Siconos.

|         |                  |
|---------|------------------|
| author  | F. P rignon      |
| date    | November 7, 2006 |
| version | Kernel 1.3.0     |

### 4.1 Introduction

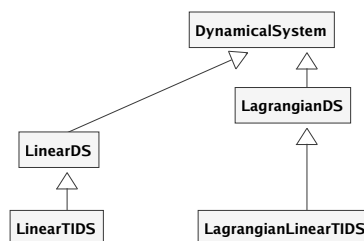
This document is only a sequel of notes and remarks on the way Dynamical Systems are implemented in Siconos.

It has to be completed, reviewed, reorganized etc etc for a future Developers' Guide.

See also documentation in Doc/User/DynamicalSystemsInSiconos for a description of various dynamical systems types.

### 4.2 Class Diagram

There are four possible formulation for dynamical systems in Siconos, two for first order systems and two for second order Lagrangian systems. The main class is `DynamicalSystem`, all other derived from this one, as shown in the following diagram:



### 4.3 Construction

Each constructor must:

- initialize all the members of the class and of the top-class if it exists

- allocate memory and set value for all required inputs
- allocate memory and set value for optional input if they are given as argument (in xml for example)
- check that given data are coherent and that the system is complete (for example, in the LagrangianDS if the internal forces are given as a plug-in, their Jacobian are also required. If they are not given, this leads to an exception).

No memory allocation is made for unused members  $\Rightarrow$  requires if statements in simulation. (if!=NULL ...).

### 4.3.1 DynamicalSystem

**Required data:**

**n, x0, f, jacobianXF**

**Optional:**

**T,u**

**Always allocated in constructor:**

**x, x0, xFree, r, rhs, jacobianXF**

**Warning:** default constructor is always private or protected and apart from the others and previous rules or remarks do not always apply to it. This for DS class and any of the derived ones.

### 4.3.2 LagrangianDS

**Required data:**

**ndof, q0, velocity0, mass**

**Optional:**

**fInt and its Jacobian, fExt, NNL and its Jacobian.**

**Always allocated in constructor:**

**mass, q, q0, qFree, velocity, velocity0, velocityFree, p.**

All other pointers to vectors/matrices are set to NULL by default.

Memory vectors are required but allocated during call to initMemory function.

Various rules:

- fInt (NNL) given as a plug-in  $\Rightarrow$  check that JacobianQ/Velocity are present (matrices or plug-in)
- any of the four Jacobian present  $\Rightarrow$  allocate memory for block-matrix jacobianX (connectToDS function)
- 

check: end of constructor or in initialize?

computeF and JacobianF + corresponding set functions: virtual or not?

## 4.4 Specific flags or members

- **isAllocatedIn:** to check inside-class memory allocation
- **isPlugin:** to check if operators are computed with plug-in or just directly set as a matrix or vector
- **workMatrix:** used to save some specific matrices in order to avoid recomputation if possible (inverse of mass ...)

## 4.5 plug-in management

**DynamicalSystem** class has a member named **parameterList** which is a *map*  $\langle \text{string}, \text{SimpleVector}^* \rangle$ , ie a list of pointers to **SimpleVector**\*, with a string as a key to identified them. For example, *parametersList*["mass"] is a **SimpleVector**\*, which corresponds to the last argument given in mass plug-in function.

By default, each parameters vectors must be initialized with a **SimpleVector** of size 1, as soon as the plug-in is declared. Moreover, to each vector corresponds a flag in *isAllocatedIn* map, to check if the corresponding vector has been allocated inside the class or not.

For example, in **DynamicalSystem**, if *isPlugin*["vectorField"] == true, then, during call to constructor or set function, it is necessary to defined the corresponding parameter:

*parametersList*["vectorField"] = newSimpleVector(1)

**and to complete the *isAllocatedIn* flag:**

*isAllocatedIn*["parameter\_for\_vectorField"] = true.

# Chapter 5

## Interactions

|         |                  |
|---------|------------------|
| author  | F. Pérignon      |
| date    | November 7, 2006 |
| version | Kernel 1.3.0     |

### 5.1 Introduction

This document is only a sequel of notes and remarks on the way Interactions are implemented in Siconos.

It has to be completed, reviewed, reorganized etc etc for a future *Developpers' Guide*.  
See also documentation in *Doc/User/Interaction*.

### 5.2 Class Diagram

### 5.3 Description

#### 5.3.1 Redaction note F. PERIGNON

review of interactions (for EventDriven implementation) 17th May 2006.

- variable *nInter* renamed in *interactionSize*: represents the size of  $y$  and  $\lambda$ . NOT the number of relations !!
- add a variable *nsLawSize* that depends on the non-smooth law type.  
Examples:
  - NewtonImpact -> *nsLawSize* = 1
  - Friction 2D -> *nsLawSize* = 2
  - Friction 3D -> *nsLawSize* = 3
  - ...
  - *nsLawSize* = n with n dim of matrix D in :  $y = Cx + D\lambda$ , D supposed to be a full-ranked matrix.  
Warning: this case is represented by only one relation of size n.
- *numberOfRelations*: number of relations in the interaction,  $numberOfRelations = \frac{interactionSize}{nsLawSize}$ .

## Chapter 6

# Notes on the Non Smooth Dynamical System construction

|         |                  |
|---------|------------------|
| author  | F. Pérignon      |
| date    | November 7, 2006 |
| version | Kernel 1.3.0     |

### 6.1 Introduction

### 6.2 Class Diagram

### 6.3 Description

Objects must be constructed in the following order:

- **DynamicalSystems**
- **NonSmoothLaw**: depends on nothing
- **Relation**: no link with an interaction during construction, this will be done during initialization.
- **Interaction**: default constructor is private and copy is forbidden. Two constructors: xml and from data. Required data are a DSSet, a NonSmoothLaw and a Relation (+ dim of the Interaction and a number).  
Interaction has an initialize function which allocates memory for y and lambda, links correctly the relation and initializes it .... This function is called at the end of the constructor. That may be better to call it in simulation->initialize? Pb: xml constructor needs memory allocation for y and lambda if they are provided in the input xml file.
- **NonSmoothDynamicalSystem**: default is private, copy forbidden. Two constructors: xml and from data. Required data are the DSSet and the InteractionsSet. The topology is declared and constructed (but empty) during constructor call of the nsds, but initialize in the Simulation, this because it can not be initialize until the nsds has been fully described (ie this to allow user to add DS, Inter ...) at any time in the model, but before simulation initialization).



## 6.4 misc

- no need to keep a number for Interactions? Only used in xml for OSI, to know which Interactions it holds.
- pb: the number of saved derivatives for  $y$  and  $\lambda$  in Interactions is set to 2. This must depends on the relative degree which is computes during Simulation initialize and thus too late. It is so not available when memory is allocated (Interaction construction). Problem-> to be reviewed.

## Chapter 7

# OneStepIntegrator and derived classes.

|         |                  |
|---------|------------------|
| author  | F. Pérignon      |
| date    | November 7, 2006 |
| version | Kernel 1.3.0     |

### 7.1 Introduction

This document is only a sequel of notes and remarks on the way OneStepIntegrators are implemented in Siconos.

It has to be completed, reviewed, reorganized etc etc for a future Developers' Guide.

See also documentation in Doc/User/OneStepIntegrator for a description of various OSI.

### 7.2 Class Diagram

### 7.3 Misc

OSI review for consistency between Lsodar and Moreau:

- add set of DynamicalSystem\*
- add set of Interaction\*
- add link to strategy that owns the OSI
- remove td object in OSI -> future: replace it by a set of td (one per ds)
- add strat in constructors arg list

osi -> strat -> Model -> nsds -> topology

osi -> strat -> timeDiscretisation

let a timeDiscretisation object in the OSI? set of td (one per ds)?  
create a class of object that corresponds to DS on the simulation side ?  
will contain the DS, its discretization, theta for Moreau ... ?  
Allow setStrategyPtr operation? Warning: need reinitialisation.

Required input by user:

- list of DS or list of Interactions ?
- pointer to strategy
- ...

## 7.4 Construction

Each constructor must:

- 

### 7.4.1 Moreau

Two maps: one for  $W$ , and one for  $\theta$ . To each DS corresponds a  $\theta$  and a  $W$ .  
Strategy arg in each constructor.

Required data:

Optional:

Always allocated in constructor:

Warning: default constructor is always private or protected and apart from the others and previous rules or remarks do not always apply to it.

### 7.4.2 Lsodar

Required data:

Optional:

Always allocated in constructor:

## Chapter 8

# Newton's linearization

Last update : 2009/07/07 This section is devoted to the implementation and the study of the algorithm. The interval of integration is  $[0, T]$ ,  $T > 0$ , and a grid  $t_0 = 0$ ,  $t_{k+1} = t_k + h$ ,  $k \geq 0$ ,  $t_N = T$  is constructed. The approximation of a function  $f(\cdot)$  on  $[0, T]$  is denoted as  $f^N(\cdot)$ , and is a piecewise constant function, constant on the intervals  $[t_k, t_{k+1})$ . We denote  $f^N(t_k)$  as  $f_k$ . The time-step is  $h > 0$ .

### 8.0.3 Time-discretization of the general case (??)

This fully general case is not yet implemented in Siconos.

Starting from (??), let us introduce anew notation,

$$\begin{aligned}\dot{x}(t) &= f(x(t), t) + r(t) \\ y(t) &= h(t, x(t), \lambda(t)) \\ r(t) &= g(t, x(t), \lambda(t))\end{aligned}\tag{8.1}$$

where  $s(t) \in \mathbb{R}^m$  and  $y(t) \in \mathbb{R}^m$  are complementary variables related through the *Sgn* multi-valued mapping. According to the class of systems (??), (??) or (??), we are studying, the function  $f$  and  $g$  are defined by a fully nonlinear framework or by affine functions. We have decided to present the time-discretization in its full generality and specialize the algorithms for each cases in Section 8.0.5.

Let us now proceed with the time discretization of (8.48) by a fully implicit scheme :

$$\begin{aligned}Mx_{k+1} &= x_k + h\theta f(x_{k+1}, t_{k+1}) + h(1 - \theta)f(x_k, t_k) + h\theta_r r(t_{k+1}) + h(1 - \theta_r)r(t_k) \\ y_{k+1} &= h(t + 1, x_{k+1}, \lambda_{k+1}) \\ r_{k+1} &= g(x_{k+1}, \lambda_{k+1}, t + 1)\end{aligned}\tag{8.2}$$

where  $\theta \in [0, 1]$  and  $\theta_r \in [0, 1]$ . As in ?, we call the problem (8.49) the "one-step nonsmooth problem".

This time-discretization is slightly more general than a standard implicit Euler scheme. The main discrepancy lies in the choice of a  $\theta$ -method to integrate the nonlinear term. For  $\theta = 0$ , we retrieve the explicit integration of the smooth and single valued term  $f$ . Moreover for  $\gamma = 0$ , the term  $g$  is explicitly evaluated. The flexibility in the choice of  $\theta$  and  $\gamma$  allows the user to improve and control the accuracy, the stability and the numerical damping of the proposed method. For instance, if the smooth dynamics given by  $f$  is stiff, or if we have to use big step sizes for practical reasons, the choice of  $\theta > 1/2$  offers better stability with the respect to  $h$ .

### 8.0.4 Newton's linearization

Due to the fact that two of the studied classes of systems that are studied in this paper are affine functions in terms of  $f$  and  $g$ , we propose to solve the "one-step nonsmooth problem" (8.49) by performing an external Newton linearization, which yields a Mixed Linear Complementarity Problems (MLCP).

**Newton's linearization** The first line of the problem (8.49) can be written under the form of a residue  $\mathcal{R}$  depending only on  $x_{k+1}$  and  $r_{k+1}$  such that

$$\mathcal{R}(x_{k+1}, r_{k+1}) = 0 \quad (8.3)$$

with  $\mathcal{R}(x, r) = M(x - x_k) - h\theta f(x, t_{k+1}) - h(1 - \theta)f(x_k, t_k) - h\theta_r r - h(1 - \theta_r)r_k$ . The solution of this system of nonlinear equations is sought as a limit of the sequence  $\{x_{k+1}^\alpha, r_{k+1}^\alpha\}_{\alpha \in \mathbb{N}}$  such that

$$\begin{cases} x_{k+1}^0 = x_k \\ \mathcal{R}_L(x_{k+1}^{\alpha+1}, r_{k+1}^{\alpha+1}) = \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha) + [\nabla_x \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha)] (x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) + [\nabla_r \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha)] (r_{k+1}^{\alpha+1} - r_{k+1}^\alpha) = 0 \end{cases} \quad (8.4)$$

The residu free is also defined (useful for implementation only):

$$\mathcal{R}_{free}(x) = M(x - x_k) - h\theta f(x, t_{k+1}) - h(1 - \theta)f(x_k, t_k)$$

$$\mathcal{R}(x, r) = \mathcal{R}_{free}(x) - h\theta_r r - h(1 - \theta_r)r_k$$

$$\boxed{\mathcal{R}_{k+1}^\alpha = \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha) = \mathcal{R}_{free}(x_{k+1}^\alpha, r_{k+1}^\alpha) - h\theta_r r_{k+1}^\alpha - h(1 - \theta_r)r_k}$$

$$\mathcal{R}_{freek+1}^\alpha = \mathcal{R}_{free}(x_{k+1}^\alpha, r_{k+1}^\alpha) = \boxed{M(x_{k+1}^\alpha - x_k) - h\theta f(x_{k+1}^\alpha, t_{k+1}) - h(1 - \theta)f(x_k, t_k)}$$

The computation of the Jacobian of  $\mathcal{R}$  with respect to  $x$ , denoted by  $M(x, \lambda)$  leads to

$$W_{k+1}^\alpha = \nabla_x \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha) = M - h\theta \nabla_x f(x_{k+1}^\alpha, t_{k+1}). \quad (8.5)$$

At each time-step, we have to solve the following linearized problem,

$$\mathcal{R}_{k+1}^\alpha + W_{k+1}^\alpha (x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) - h\theta_r (r_{k+1}^{\alpha+1} - r_{k+1}^\alpha) = 0, \quad (8.6)$$

that is

$$h\theta_r r_{k+1}^{\alpha+1} = r_c + W_{k+1}^\alpha x_{k+1}^{\alpha+1}. \quad (8.7)$$

with

$$r_c = h\theta_r r_{k+1}^\alpha - W_{k+1}^\alpha x_{k+1}^\alpha + \mathcal{R}_{k+1}^\alpha = -W_{k+1}^\alpha x_{k+1}^\alpha + \mathcal{R}_{freek+1}^\alpha - h(1 - \theta_r)r_k \quad (8.8)$$

$$\mathcal{R}_{k+1}^\alpha = M(x_{k+1}^\alpha - x_k) - h\theta f(x_{k+1}^\alpha, t_{k+1}) - h(1 - \theta)f(x_k, t_k) - h\theta_r r_{k+1}^\alpha - h(1 - \theta_r)r_k$$

$$x_{k+1}^{\alpha+1} = h(W_{k+1}^\alpha)^{-1} r_{k+1}^{\alpha+1} + (W_{k+1}^\alpha)^{-1} (\mathcal{R}_{freek+1}^\alpha) + x_{k+1}^\alpha$$

Using 8.7

$$\boxed{x_{k+1}^{\alpha+1} = h(W_{k+1}^\alpha)^{-1} r_{k+1}^{\alpha+1} + x_{free}}$$

with :

$$x_{free} = x_{k+1}^\alpha - (W_{k+1}^\alpha)^{-1} \mathcal{R}_{freek+1}^\alpha$$

The matrix  $W$  is clearly non singular for small  $h$ . The same operation is performed with the second equation of (8.49)

$$\mathcal{R}_y(x, y, \lambda) = y - h(t_{k+1}, x, \lambda) = 0$$

$$\mathcal{R}_{Ly}(x_{k+1}^{\alpha+1}, y_{k+1}^{\alpha+1}, \lambda_{k+1}^{\alpha+1}) = \mathcal{R}_y(x_{k+1}^\alpha, y_{k+1}^\alpha, \lambda_{k+1}^\alpha) + (y_{k+1}^{\alpha+1} - y_{k+1}^\alpha) - C_{k+1}^\alpha (x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) - D_{k+1}^\alpha (\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^\alpha) = 0 \quad (8.9)$$

$\mathcal{R}_y(x_{k+1}^{\alpha+1}, y_{k+1}^{\alpha+1} + 1, \lambda_{k+1}^{\alpha+1})$  leading to the following linearized equation

$$y_{k+1}^{\alpha+1} = y_{k+1}^\alpha - \mathcal{R}_{yk+1}^\alpha + C_{k+1}^\alpha (x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) + D_{k+1}^\alpha (\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^\alpha). \quad (8.10)$$

with,

$$C_{k+1}^\alpha = \nabla_x h(t_{k+1}, x_{k+1}^\alpha, \lambda_{k+1}^\alpha) \quad (8.11)$$

$$D_{k+1}^\alpha = \nabla_\lambda h(t_{k+1}, x_{k+1}^\alpha, \lambda_{k+1}^\alpha)$$

and

$$\boxed{\mathcal{R}_{y_{k+1}}^\alpha = y_{k+1}^\alpha - h(x_{k+1}^\alpha, \lambda_{k+1}^\alpha)} \quad (8.12)$$

The same operation is performed with the third equation of (8.49)

$$\mathcal{R}_r(r, x, \lambda) = r - g(x, \lambda, t_{k+1}) = 0$$

$$\mathcal{R}_{L\lambda}(r_{k+1}^{\alpha+1}, x_{k+1}^{\alpha+1}, \lambda_{k+1}^{\alpha+1}) = \mathcal{R}_{rk+1}^\alpha + (r_{k+1}^{\alpha+1} - r_{k+1}^\alpha) - K_{k+1}^\alpha(x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) - B_{k+1}^\alpha(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^\alpha) = 0$$

$$r_{k+1}^{\alpha+1} = r_1 + K_{k+1}^\alpha x_{k+1}^{\alpha+1} + B_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} r_{k+1}^{\alpha+1} = \boxed{g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha + B_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} - K_{k+1}^\alpha x_{k+1}^\alpha + K_{k+1}^\alpha x_{k+1}^{\alpha+1}} \quad (8.13)$$

with,

$$r_1 = g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) - K_{k+1}^\alpha x_{k+1}^\alpha - B_{k+1}^\alpha \lambda_{k+1}^\alpha \quad (8.14)$$

and,

$$K_{k+1}^\alpha = \nabla_x g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) \quad (8.15)$$

$$B_{k+1}^\alpha = \nabla_\lambda g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1})$$

and the r residue:

$$\mathcal{R}_{rk+1}^\alpha = r_{k+1}^\alpha - g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) \quad (8.16)$$

Inserting (8.7), we get the following linear relation between  $x_{k+1}^{\alpha+1}$  and  $\lambda_{k+1}^{\alpha+1}$ ,

$$\begin{aligned} (W_{k+1}^\alpha) x_{k+1}^{\alpha+1} &= h\theta_r \left[ g(r_{k+1}^\alpha, t_{k+1}) + B_{k+1}^\alpha(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^\alpha) + K_{k+1}^\alpha(x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) \right] - r_c \\ x_{k+1}^{\alpha+1} &= h\theta_r (W_{k+1}^\alpha)^{-1} \left[ g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) + B_{k+1}^\alpha(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^\alpha) + K_{k+1}^\alpha(x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) \right] - \\ &\quad (W_{k+1}^\alpha)^{-1} (-W_{k+1}^\alpha x_{k+1}^\alpha + \mathcal{R}_{freek+1}^\alpha - h(1 - \theta_r)r_k) \\ x_{k+1}^{\alpha+1} &= x_p + h\theta_r (W_{k+1}^\alpha)^{-1} (B_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} + K_{k+1}^\alpha x_{k+1}^{\alpha+1}) \\ (I - h\theta_r (W_{k+1}^\alpha)^{-1} K_{k+1}^\alpha) x_{k+1}^{\alpha+1} &= x_p + h\theta_r (W_{k+1}^\alpha)^{-1} B_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} \end{aligned} \quad (8.17)$$

We now define:

$$\tilde{K}_{k+1}^\alpha = (I - h\theta_r (W_{k+1}^\alpha)^{-1} K_{k+1}^\alpha)$$

with  $\theta_r = 1$ :

$$\begin{aligned} (W_{k+1}^\alpha) x_{k+1}^{\alpha+1} &= h r_{k+1}^{\alpha+1} - \mathcal{R}_{freek+1}^\alpha + W_{k+1}^\alpha x_{k+1}^\alpha \\ x_{k+1}^{\alpha+1} &= h (W_{k+1}^\alpha)^{-1} r_{k+1}^{\alpha+1} - (W_{k+1}^\alpha)^{-1} \mathcal{R}_{freek+1}^\alpha + x_{k+1}^\alpha \\ x_{k+1}^{\alpha+1} &= h (W_{k+1}^\alpha)^{-1} r_{k+1}^{\alpha+1} + x_{free} \end{aligned}$$

with, using ??

$$x_p - x_{k+1}^\alpha = h (W_{k+1}^\alpha)^{-1} (g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha - K_{k+1}^\alpha x_{k+1}^\alpha) + \tilde{x}_{free} \quad (8.18)$$

$$\tilde{x}_{free} = -(W_{k+1}^\alpha)^{-1} \mathcal{R}_{freek+1}^\alpha$$

$$x_{free} = \tilde{x}_{free} + x_{k+1}^\alpha = \boxed{-W^{-1} \mathcal{R}_{freek+1}^\alpha + x_{k+1}^\alpha}$$

$$x_p = x_{free} + h(W_{k+1}^\alpha)^{-1}(g(x_{k+1}^\alpha, \lambda_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha - K_{k+1}^\alpha x_{k+1}^\alpha)$$

Inserting (8.10), we get the following linear relation between  $y_{k+1}^{\alpha+1}$  and  $\lambda_{k+1}^{\alpha+1}$ ,

$$y_{k+1}^{\alpha+1} = y_p + [hC_{k+1}^\alpha (\tilde{K}_{k+1}^\alpha)^{-1} (W_{k+1}^\alpha)^{-1} B_{k+1}^\alpha + D_{k+1}^\alpha] \lambda_{k+1}^{\alpha+1} \quad (8.19)$$

with

$$y_p = y_{k+1}^\alpha - \mathcal{R}_{y_{k+1}}^\alpha + C_{k+1}^\alpha ((\tilde{K}_{k+1}^\alpha)^{-1} x_p - x_{k+1}^\alpha) - D_{k+1}^\alpha \lambda_{k+1}^\alpha \quad (8.20)$$

**Mixed linear complementarity problem (MLCP)** To summarize, the problem to be solved in each Newton iteration is:

$$\begin{cases} y_{k+1}^{\alpha+1} = W_{mlcpk+1}^\alpha \lambda_{k+1}^{\alpha+1} + b_{k+1}^\alpha \\ -y_{k+1}^{\alpha+1} \in N_{[l,u]}(\lambda_{k+1}^{\alpha+1}). \end{cases} \quad (8.21)$$

with  $W_{mlcpk+1} \in \mathbb{R}^{m \times m}$  and  $b \in \mathbb{R}^m$  defined by

$$\begin{aligned} W_{mlcpk+1}^\alpha &= hC_{k+1}^\alpha (\tilde{K}_{k+1}^\alpha)^{-1} (W_{k+1}^\alpha)^{-1} B_{k+1}^\alpha + D_{k+1}^\alpha \\ b_{k+1}^\alpha &= y_p \end{aligned} \quad (8.22)$$

The problem (8.21) is equivalent to a Mixed Linear Complementarity Problem (MLCP) which can be solved under suitable assumptions by many linear complementarity solvers such as pivoting techniques, interior point techniques and splitting/projection strategies. The reformulation into a standard MLCP follows the same line as for the MCP in the previous section. One obtains,

$$\begin{aligned} y_{k+1}^{\alpha+1} &= -W_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} + b_{k+1}^\alpha \\ (y_{k+1}^{\alpha+1})_i &= 0 \quad \text{for } i \in \{1..n\} \\ 0 &\leq (\lambda_{k+1}^{\alpha+1})_i \perp (y_{k+1}^{\alpha+1})_i \geq 0 \quad \text{for } i \in \{n..n+m\} \end{aligned} \quad (8.23)$$

**MLCP solvers.** As for MCP, there exists numerous methods to numerically solve MLCP. In the worst case when the matrix  $W_{k+1}^{\alpha+1}$  has no special properties, the MCLP can be always solved by enumerative solvers for which various implementation can be found. With positivity, P-matrix or co-positivity properties, standard methods for LCP? can be straightforwardly extended. Among these methods, we can cite the family of projection/splitting methods, interior point methods and semi-smooth Newton methods (see ? for an overview of various types of methods.).

### 8.0.5 The special cases of the affine systems (??) and (??)

In this section, we specify the time-discretization of the fully nonlinear case for the two other classes of systems (??) and (??) and for particular value of  $\theta$  and  $\gamma$ .

### 8.0.6 Time-discretization of the the case (??)

This case is implemented in Siconos with the relation `FirstOrderType2R`.

Starting from (??), let us introduce a new notation,

$$\begin{aligned} M\dot{x}(t) &= f(x(t), t) + r(t) \\ y(t) &= h(t, x(t), \lambda(t)) \\ r(t) &= g(t, \lambda(t)) \end{aligned} \quad (8.24)$$

Let us now proceed with the time discretization of (8.48) by a fully implicit scheme :

$$\begin{aligned} Mx_{k+1} &= Mx_k + h\theta f(x_{k+1}, t_{k+1}) + h(1 - \theta)f(x_k, t_k) + h\theta_r r(t_{k+1}) + h(1 - \theta_r)r(t_k) \\ y_{k+1} &= h(t + 1, x_{k+1}, \lambda_{k+1}) \\ r_{k+1} &= g(\lambda_{k+1}, t + 1) \end{aligned} \quad (8.25)$$

### 8.0.7 Newton's linearization

**Newton's linearization** The first line of the problem (8.49) can be written under the form of a residue  $\mathcal{R}$  depending only on  $x_{k+1}$  and  $r_{k+1}$  such that

$$\mathcal{R}(x_{k+1}, r_{k+1}) = 0 \quad (8.26)$$

with  $\mathcal{R}(x, r) = M(x - x_k) - h\theta f(x, t_{k+1}) - h(1 - \theta)f(x_k, t_k) - h\theta_r r - h(1 - \theta_r)r_k$ . The solution of this system of nonlinear equations is sought as a limit of the sequence  $\{x_{k+1}^\alpha, r_{k+1}^\alpha\}_{\alpha \in \mathbb{N}}$  such that

$$\begin{cases} x_{k+1}^0 = x_k \\ \mathcal{R}_L(x_{k+1}^{\alpha+1}, r_{k+1}^{\alpha+1}) = \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha) + [\nabla_x \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha)] (x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) + [\nabla_r \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha)] (r_{k+1}^{\alpha+1} - r_{k+1}^\alpha) = 0 \end{cases} \quad (8.27)$$

The residu free is also defined (useful for implementation only):

$$\mathcal{R}_{free}(x) = M(x - x_k) - h\theta f(x, t_{k+1}) - h(1 - \theta)f(x_k, t_k)$$

$$\mathcal{R}(x, r) = \mathcal{R}_{free}(x) - h\theta_r r - h(1 - \theta_r)r_k$$

$$\mathcal{R}_{k+1}^\alpha = \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha) = \mathcal{R}_{free}(x_{k+1}^\alpha, r_{k+1}^\alpha) - h\theta_r r_{k+1}^\alpha - h(1 - \theta_r)r_k$$

$$\mathcal{R}_{freek+1}^\alpha = \mathcal{R}_{free}(x_{k+1}^\alpha, r_{k+1}^\alpha) = M(x_{k+1}^\alpha - x_k) - h\theta f(x_{k+1}^\alpha, t_{k+1}) - h(1 - \theta)f(x_k, t_k)$$

The computation of the Jacobian of  $\mathcal{R}$  with respect to  $x$ , denoted by  $M(x, \lambda)$  leads to

$$W_{k+1}^\alpha = \nabla_x \mathcal{R}(x_{k+1}^\alpha, r_{k+1}^\alpha) = M - h\theta \nabla_x f(x_{k+1}^\alpha, t_{k+1}). \quad (8.28)$$

At each time-step, we have to solve the following linearized problem,

$$\mathcal{R}_{k+1}^\alpha + W_{k+1}^\alpha (x_{k+1}^{\alpha+1} - x_{k+1}^\alpha) - h\theta_r (r_{k+1}^{\alpha+1} - r_{k+1}^\alpha) = 0, \quad (8.29)$$

that is

$$h\theta_r r_{k+1}^{\alpha+1} = r_c + W_{k+1}^\alpha x_{k+1}^{\alpha+1}. \quad (8.30)$$

with

$$r_c = h\theta_r r_{k+1}^\alpha - W_{k+1}^\alpha x_{k+1}^\alpha + \mathcal{R}_{k+1}^\alpha = -W_{k+1}^\alpha x_{k+1}^\alpha + \mathcal{R}_{freek+1}^\alpha - h(1 - \theta_r)r_k \quad (8.31)$$

so:

$$x_{k+1}^{\alpha+1} = h(W_{k+1}^\alpha)^{-1} r_{k+1}^{\alpha+1} + (W_{k+1}^\alpha)^{-1} (\mathcal{R}_{freek+1}^\alpha) + x_{k+1}^\alpha$$



$$x_{k+1}^{\alpha+1} = h(W_{k+1}^{\alpha})^{-1}g_{k+1}^{\alpha+1} + x_{free}$$

The matrix  $W$  is clearly non singular for small  $h$ . The same operation is performed with the second equation of (8.49)

$$\begin{aligned} \mathcal{R}_y(x, y, \lambda) &= y - h(t_{k+1}, x, \lambda) = 0 \\ \mathcal{R}_{Ly}(x_{k+1}^{\alpha+1}, y_{k+1}^{\alpha+1}, \lambda_{k+1}^{\alpha+1}) &= \mathcal{R}_y(x_{k+1}^{\alpha}, y_{k+1}^{\alpha}, \lambda_{k+1}^{\alpha}) + (y_{k+1}^{\alpha+1} - y_{k+1}^{\alpha}) - \\ &\quad C_{k+1}^{\alpha}(x_{k+1}^{\alpha+1} - x_{k+1}^{\alpha}) - D_{k+1}^{\alpha}(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^{\alpha}) = 0 \end{aligned} \quad (8.32)$$

$\mathcal{R}_y(x_{k+1}^{\alpha+1}, y_{k+1}^{\alpha+1}, \lambda_{k+1}^{\alpha+1})$  leading to the following linearized equation

$$y_{k+1}^{\alpha+1} = y_{k+1}^{\alpha} - \mathcal{R}_{y_{k+1}}^{\alpha} + C_{k+1}^{\alpha}(x_{k+1}^{\alpha+1} - x_{k+1}^{\alpha}) + D_{k+1}^{\alpha}(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^{\alpha}). \quad (8.33)$$

with,

$$C_{k+1}^{\alpha} = \nabla_x h(t_{k+1}, x_{k+1}^{\alpha}, \lambda_{k+1}^{\alpha}) \quad (8.34)$$

$$D_{k+1}^{\alpha} = \nabla_{\lambda} h(t_{k+1}, x_{k+1}^{\alpha}, \lambda_{k+1}^{\alpha})$$

and

$$\mathcal{R}_{y_{k+1}}^{\alpha} = y_{k+1}^{\alpha} - h(x_{k+1}^{\alpha}, \lambda_{k+1}^{\alpha}, t_{k+1}) \quad (8.35)$$

The same operation is performed with the third equation of (8.49)

$$\begin{aligned} \mathcal{R}_r(r, \lambda) &= r - g(\lambda, t_{k+1}) = 0 \\ \mathcal{R}_{L\lambda}(r_{k+1}^{\alpha+1}, \lambda_{k+1}^{\alpha+1}) &= \mathcal{R}_{r_{k+1}}^{\alpha} + (r_{k+1}^{\alpha+1} - r_{k+1}^{\alpha}) - B_{k+1}^{\alpha}(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^{\alpha}) = 0 \\ r_{k+1}^{\alpha+1} &= -\mathcal{R}_{r_{k+1}}^{\alpha} + r_{k+1}^{\alpha} + B_{k+1}^{\alpha}(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^{\alpha}) \end{aligned} \quad (8.36)$$

$$r_{k+1}^{\alpha+1} = g(\lambda_{k+1}^{\alpha}, t_{k+1}) - B_{k+1}^{\alpha}\lambda_{k+1}^{\alpha} + B_{k+1}^{\alpha}\lambda_{k+1}^{\alpha+1}$$

because the  $r$  residue is:

$$\mathcal{R}_{r_{k+1}}^{\alpha} = r_{k+1}^{\alpha} - g(r_{k+1}^{\alpha}, t_{k+1}) \quad (8.37)$$

with,

$$B_{k+1}^{\alpha} = \nabla_{\lambda} g(r_{k+1}^{\alpha}, t_{k+1}) \quad (8.38)$$

Inserting (8.30), we get the following linear relation between  $x_{k+1}^{\alpha+1}$  and  $\lambda_{k+1}^{\alpha+1}$ ,

$$\begin{aligned} (W_{k+1}^{\alpha})x_{k+1}^{\alpha+1} &= h\theta_r \left[ g(r_{k+1}^{\alpha}, t_{k+1}) + B_{k+1}^{\alpha}(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^{\alpha}) \right] - r_c \\ x_{k+1}^{\alpha+1} &= h\theta_r (W_{k+1}^{\alpha})^{-1} \left[ g(r_{k+1}^{\alpha}, t_{k+1}) + B_{k+1}^{\alpha}(\lambda_{k+1}^{\alpha+1} - \lambda_{k+1}^{\alpha}) \right] - (W_{k+1}^{\alpha})^{-1}(-W_{k+1}^{\alpha}x_{k+1}^{\alpha} + \mathcal{R}_{freek+1}^{\alpha} - h(1 - \theta_r)r_k) \\ x_{k+1}^{\alpha+1} &= x_p + h\theta_r (W_{k+1}^{\alpha})^{-1}B_{k+1}^{\alpha}\lambda_{k+1}^{\alpha+1} \end{aligned} \quad (8.39)$$

with  $\theta_r = 1$ :

$$\begin{aligned} (W_{k+1}^{\alpha})x_{k+1}^{\alpha+1} &= hr_{k+1}^{\alpha+1} - \mathcal{R}_{freek+1}^{\alpha} + W_{k+1}^{\alpha}x_{k+1}^{\alpha} \\ x_{k+1}^{\alpha+1} &= h(W_{k+1}^{\alpha})^{-1}r_{k+1}^{\alpha+1} - (W_{k+1}^{\alpha})^{-1}\mathcal{R}_{freek+1}^{\alpha} + x_{k+1}^{\alpha} \\ x_{k+1}^{\alpha+1} &= h(W_{k+1}^{\alpha})^{-1}r_{k+1}^{\alpha+1} + x_{free} \end{aligned}$$

with

$$x_p - x_{k+1}^\alpha = (W_{k+1}^\alpha)^{-1} [-r_c + h\theta_r(g(r_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha)] \quad (8.40)$$

$$x_p - x_{k+1}^\alpha = \tilde{x}_{free} + h\theta_r(W_{k+1}^\alpha)^{-1}(g(r_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha)$$

$$\tilde{x}_{free} = (W_{k+1}^\alpha)^{-1}(-\mathcal{R}_{freek+1}^\alpha + h(1 - \theta_r)r_k)$$

$$x_{free} = \tilde{x}_{free} + x_{k+1}^\alpha = \boxed{-W^{-1}R_{freek+1}^\alpha + x_{k+1}^\alpha}$$

$$\boxed{x_p = x_{free} + h\theta_r(W_{k+1}^\alpha)^{-1}(g(\lambda_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha)}$$

Inserting (8.33), we get the following linear relation between  $y_{k+1}^{\alpha+1}$  and  $\lambda_{k+1}^{\alpha+1}$ ,

$$y_{k+1}^{\alpha+1} = y_p + [h\theta_r C_{k+1}^\alpha (W_{k+1}^\alpha)^{-1} B_{k+1}^\alpha + D_{k+1}^\alpha] \lambda_{k+1}^{\alpha+1} \quad (8.41)$$

with

$$\boxed{y_p = y_{k+1}^\alpha - \mathcal{R}_{yk+1}^\alpha + C_{k+1}^\alpha (x_p - x_{k+1}^\alpha) - D_{k+1}^\alpha \lambda_{k+1}^\alpha} \quad (8.42)$$

$$y_p = h(x_{k+1}^\alpha, \lambda_{k+1}^\alpha) + C_{k+1}^\alpha (x_p - x_{k+1}^\alpha) - D_{k+1}^\alpha \lambda_{k+1}^\alpha \quad (8.43)$$

$$y_p = h(x_{k+1}^\alpha, \lambda_{k+1}^\alpha) - C_{k+1}^\alpha x_{k+1}^\alpha + C_{k+1}^\alpha x_{free} + hC(W_{k+1}^\alpha)^{-1}(g(\lambda_{k+1}^\alpha, t_{k+1}) - B_{k+1}^\alpha \lambda_{k+1}^\alpha) - D_{k+1}^\alpha \lambda_{k+1}^\alpha \quad (8.44)$$

**Mixed linear complementarity problem (MLCP)** To summarize, the problem to be solved in each Newton iteration is:

$$\begin{cases} y_{k+1}^{\alpha+1} = \tilde{W}_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} + b_{k+1}^\alpha \\ -y_{k+1}^{\alpha+1} \in N_{[l,u]}(\lambda_{k+1}^{\alpha+1}). \end{cases} \quad (8.45)$$

with  $W \in \mathbb{R}^{m \times m}$  and  $b \in \mathbb{R}^m$  defined by

$$\begin{aligned} \tilde{W}_{k+1}^\alpha &= h\theta_r C_{k+1}^\alpha (W_{k+1}^\alpha)^{-1} B_{k+1}^\alpha + D_{k+1}^\alpha \\ b_{k+1}^\alpha &= y_p \end{aligned} \quad (8.46)$$

The problem (8.45) is equivalent to a Mixed Linear Complementarity Problem (MLCP) which can be solved under suitable assumptions by many linear complementarity solvers such as pivoting techniques, interior point techniques and splitting/projection strategies. The reformulation into a standard MLCP follows the same line as for the MCP in the previous section. One obtains,

$$\begin{aligned} y_{k+1}^{\alpha+1} &= -W_{k+1}^\alpha \lambda_{k+1}^{\alpha+1} + b_{k+1}^\alpha \\ (y_{k+1}^{\alpha+1})_i &= 0 \quad \text{for } i \in \{1..n\} \\ 0 &\leq (\lambda_{k+1}^{\alpha+1})_i \perp (y_{k+1}^{\alpha+1})_i \geq 0 \quad \text{for } i \in \{n..n+m\} \end{aligned} \quad (8.47)$$

**MLCP solvers.** As for MCP, there exists numerous methods to numerically solve MLCP. In the worst case when the matrix  $W_{k+1}^{\alpha+1}$  has no special properties, the MCLP can be always solved by enumerative solvers for which various implementation can be found. With positivity, P-matrix or co-positivity properties, standard methods for LCP? can be straightforwardly extended. Among these methods, we can cite the family of projection/splitting methods, interior point methods and semi-smooth Newton methods (see ? for an overview of various types of methods.).

### 8.0.8 Implementation In siconos

The initial residu x:

$$\begin{aligned}\mathcal{R}_{k+1}^0 &= \mathcal{R}_{freek+1}^0 - hr_{k+1}^0 \\ \mathcal{R}_{k+1}^0 &= M(x_{k+1}^0 - x_k) - h\theta f(x_k, t_{k+1}) - h(1 - \theta)f(x_k, t_k) - hr_{k+1}^0 \\ \mathcal{R}_{k+1}^0 &= -h\theta f(x_k, t_{k+1}) - h(1 - \theta)f(x_k, t_k) - hr_{k+1}^0\end{aligned}$$

The initial residu y:

$$\mathcal{R}_{yk+1}^0 = y_k - h(x_k, \lambda_k, t_{k+1})$$

The initial residu r:

$$\mathcal{R}_{rk+1}^0 = r_k - g(\lambda_k, t_{k+1})$$

In the case of the system (??) with a affine function  $f$  or  $\theta = 0$ , the the MLCP matrix  $W$  can be computed before the beginning of the time loop saving a lot of computing effort. In the case of the system (??) with  $\theta = \gamma = 0$ , the MLCP matrix  $W$  can be computed before the beginning of the Newton loop.

### 8.0.9 Time-discretization of the linear case (??)

Starting from (??), let us introduce anew notation,

$$\begin{aligned} M\dot{x}(t) &= Ax(t) + r(t) + b(t) \\ y(t) &= h(x(t), \lambda(t), z) = Cx + Fz + D\lambda \\ r(t) &= g(t, \lambda(t)) = B\lambda \end{aligned} \quad (8.48)$$

Let us now proceed with the time discretization of (8.48) by a fully implicit scheme :

$$\begin{aligned} Mx_{k+1}^{\alpha+1} &= Mx_k + h\theta Ax_{k+1}^{\alpha+1} + h(1-\theta)Ax_k + h\theta_r r_{k+1}^{\alpha+1} + h(1-\theta_r)r(t_k) + hb \\ y_{k+1}^{\alpha+1} &= Cx_{k+1}^{\alpha+1} + D\lambda_{k+1}^{\alpha+1} + Fz + e \\ r_{k+1}^{\alpha+1} &= B\lambda_{k+1}^{\alpha+1} \end{aligned} \quad (8.49)$$

$$\begin{aligned} R_{free} &= M(x_{k+1}^{\alpha} - x_k) - h\theta Ax_{k+1}^{\alpha} - h(1-\theta)Ax_k - hb_{k+1} \\ R_{free} &= W(x_{k+1}^{\alpha} - x_k) - hAx_k - hb_{k+1} \end{aligned}$$

### 8.0.10 Resulting Newton step (only one step)

suppose:  $\theta_r = 1$

$$\begin{aligned} (M - h\theta A)x_{k+1}^{\alpha+1} &= Mx_k + h(1-\theta)Ax_k + hr_{k+1}^{\alpha+1} + hb \\ y_{k+1}^{\alpha+1} &= Cx_{k+1}^{\alpha+1} + D\lambda_{k+1}^{\alpha+1} + Fz + e \\ r_{k+1}^{\alpha+1} &= B\lambda_{k+1}^{\alpha+1} \end{aligned} \quad (8.50)$$

that lead to with:  $(M - h\theta A) = W$

$$\begin{aligned} x_{k+1}^{\alpha+1} &= W^{-1}(Mx_k + h(1-\theta)Ax_k + r_{k+1}^{\alpha+1} + hb) = x_{free} + W^{-1}(r_{k+1}^{\alpha+1}) \\ y_{k+1}^{\alpha+1} &= (D + hCW^{-1}B)\lambda_{k+1}^{\alpha+1} + Fz + CW^{-1}(Mx_k + h(1-\theta)Ax_k + hb) + e \end{aligned} \quad (8.51)$$

with  $x_{free} = x_{k+1}^{\alpha} + W^{-1}(-R_{free}) = x_{k+1}^{\alpha} - W^{-1}(W(x_{k+1}^{\alpha} - x_k) - hAx_k - hb_{k+1}) = W^{-1}(Mx_k + h(1-\theta)Ax_k + hb_{k+1})$

$$\begin{aligned} y_{k+1}^{\alpha+1} &= (D + hCW^{-1}B)\lambda_{k+1}^{\alpha+1} + Fz + Cx_{free} + e \\ r_{k+1}^{\alpha+1} &= B\lambda_{k+1}^{\alpha+1} \end{aligned} \quad (8.52)$$

### 8.0.11 coherence with previous formulation

$$\begin{aligned} y_p &= y_{k+1}^{\alpha} - \mathcal{R}_{y_{k+1}}^{\alpha} + C_{k+1}^{\alpha}(x_p - x_{k+1}^{\alpha}) - D_{k+1}^{\alpha}\lambda_{k+1}^{\alpha} \\ y_p &= Cx_k + D\lambda_k + C(\tilde{x}_{free}) - D\lambda_k + Fz + e \\ y_p &= Cx_k + C(\tilde{x}_{free}) + Fz + e \\ y_p &= Cx_k + C(\tilde{x}_{free}) + Fz + e \\ y_p &= C(x_{free}) + Fz + e \end{aligned}$$

## Chapter 9

# NewtonEuler

This section is about the Newton-Euler equations:

### 9.0.12 The dynamical system

$$\begin{aligned} m\dot{V}(t) &= F_{ext} \\ I\dot{\Omega} + \Omega \wedge I\Omega &= M_{ext} \end{aligned} \quad (9.1)$$

Where  $V$  is the velocity of the center of mass and  $\Omega$  is the angular speed in the referential attached to the object.

### 9.0.13 The relation

$X$  is the position and the orientation, we don't focus on its representation.

$$\begin{aligned} Y &= H(X) \\ R &= G(X, \lambda) \end{aligned} \quad (9.2)$$

The first equation is derived:

$$\dot{Y} = C\dot{X} + \dot{H}$$

It exists an operator  $T$  :

$$T : \begin{pmatrix} V \\ \Omega \end{pmatrix} \rightarrow \dot{X}$$

Using  $T$  leads to :

$$\dot{Y} = CT \begin{pmatrix} V \\ \Omega \end{pmatrix} + \dot{H}$$

### 9.0.14 discretization $t_k \rightarrow t_{k+1}$ , and implementation in Siconos

The goal of this section is to describe the computation done in Siconos. The unknown are rename using the Siconos convention.

#### 9.0.14.a the unknowns

About the DS:

$$_v_k = \begin{pmatrix} V_k \\ \Omega_k \end{pmatrix}$$

And  $_q_k$  represent the system, usually it could be the coordinate of the center of mass and a representation of the orientation.

### 9.0.14.b explicit case

The dynamical system 11.2 leads to the system:

$$\begin{pmatrix} m & 0 \\ 0 & I \end{pmatrix} (_{v_{k+1}} - _{v_k}) = h_{Fl_k} + {}^t(CT)h\lambda_{k+1} \quad (9.3)$$

With

$$_{Fl_k} = \begin{pmatrix} Fext_k \\ Mext_k - \Omega_k \wedge I\Omega_k \end{pmatrix}$$

Using  $W = \begin{pmatrix} m & 0 \\ 0 & I \end{pmatrix}^{-1}$

$$\boxed{_{v_{k+1}} = W(h_{Fl_k}) + W^t(CT)h\lambda_{k+1} + _{v_k}} \quad (9.4)$$

This computation is done in Moreau::updateState, using:

$$_{ResiduFree_k} = -h_{Fl_k}$$

$$Xfree_k = -W_{ResiduFree_k} + _{v_k}$$

The relation 9.2 leads to the system:

$$\dot{Y}_{k+1} = CT_{v_{k+1}} + \dot{H}_k$$

Substitute  $_{v_{k+1}}$  using 9.4 leads:

$$\dot{Y}_{k+1} = CT[Wh_{Fl_k} + hW^t(CT)\lambda_{k+1} + _{v_k}] + \dot{H}_k = CT[hW^t(CT)\lambda_{k+1} + Xfree_k] + \dot{H}_k$$

Ones gets:

$$\boxed{\dot{Y}_{k+1} = CTW^t(CT)(h\lambda_{k+1}) + CTXfree_k + \dot{H}_k}$$

Solving the one step problem gives  $h\lambda_{k+1}$ , and from 9.4 we get  $_{v_{k+1}}$ . At least,  $_{v_{k+1}}$  is used to compute  $\dot{q}_{k+1}$ , provided  $_{q_{k+1}}$ .

### 9.0.15 Quaternion case

Working in 3D, we chose  $_{q} = \begin{pmatrix} X_g \\ q \end{pmatrix}$ .  $X_g$  are the 3 coordinates of the center of mass, and  $q$  is a quaternion represented the orientation of solid. It means :

$$q_k(0, GM_0)q_k^c = (0, GM_k)$$

Where G is the center of mass, and M any point of the solid.

This section describes the  $T$  operator in this case. Computation using quaternion leads to the relation:

$$\dot{q} = \frac{1}{2}q(0, \Omega)$$

So using the matrix formulation:

$$\dot{q} = \frac{1}{2} \begin{pmatrix} q_0 & -q_1 & -q_2 & -q_3 \\ q_1 & q_0 & -q_3 & q_2 \\ q_2 & q_3 & q_0 & -q_1 \\ q_3 & -q_2 & -q_1 & q_0 \end{pmatrix} \begin{pmatrix} 0 \\ \Omega \end{pmatrix} = T_q \Omega$$

That lead to :

$$\dot{q} = \begin{pmatrix} I_3 & 0 \\ 0 & T_q \end{pmatrix} \begin{pmatrix} V \\ \Omega \end{pmatrix} = T \begin{pmatrix} V \\ \Omega \end{pmatrix} = T_{v_{k+1}}$$

It is noteworthy that  $T$  must be updated at each step.

### 9.0.16 Case of the Relation between many DS

We focus on relation using 2 DS:

$$Y = H(_q1, _q2)$$

## Chapter 10

# NewtonEulerR: computation of $\nabla_q H$

### 10.0.17 Gradient computation, case of NewtonEuler with quaternion

In the section,  $q$  is the quaternion of the dynamical system.

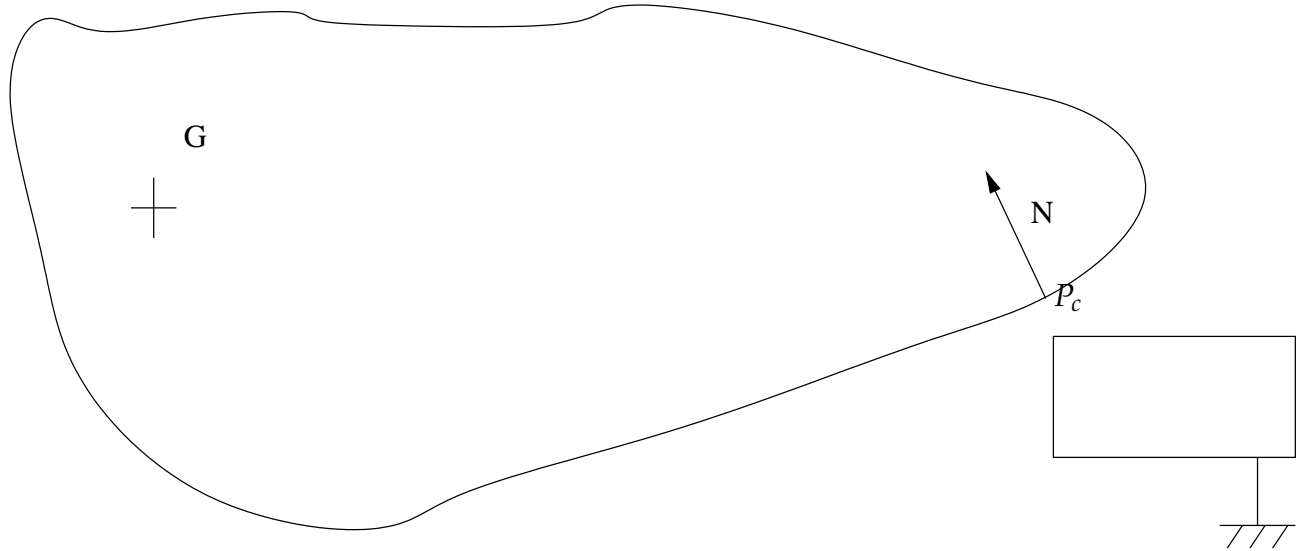


Figure 10.1: Impact of one DS.

$\nabla_q h$  consist in computing  $P_c(q + \delta q) - P_c(q)$ .

$$GP(q) = qG_0P_0 {}^c q$$

$$\begin{aligned} GP(q + \delta q) &= (q + \delta q)G_0P_0 {}^c (q + \delta q) \\ &= (q + \delta q) {}^c q GP(q) q {}^c (q + \delta q) \\ &= (1, 0, 0, 0) + \delta q {}^c q GP(q) (q {}^c q + q {}^c \delta q) \\ &= GP(q) + \delta q {}^c q GP(q) + GP(q) q {}^c \delta q + 0(\delta q)^2 \end{aligned}$$

So, because  $G$  is independant of  $q$ :

$$P(q + \delta q) - P(q) = qGP(q + \delta q) - GP(q) = \delta q {}^c q GP(q) + GP(q) q {}^c \delta q + 0(\delta q)^2$$

For the directional derivation, we chose  $\delta q = \epsilon * (1, 0, 0, 0)$

$$\lim_{\epsilon \rightarrow 0} \frac{P(q + \delta q) - P(q)}{\epsilon} = {}^c q GP(q) + GP(q) q$$



For the directional derivation, we chose  $\delta q = \epsilon * (0, 1, 0, 0) = \epsilon * e_i$

$$\lim_{\epsilon \rightarrow 0} \frac{P(q + \delta q) - P(q)}{\epsilon} = e_i \cdot {}^c q GP(q) - GP(q) q e_i$$

Application to the NewtonEulerRImpact:

$$\begin{aligned} H : \mathbb{R}^7 &\rightarrow \mathbb{R} \\ \nabla_q H &\in \mathcal{M}^{1,7} \\ \nabla_q H &= \begin{pmatrix} N_x \\ N_y \\ N_z \\ ({}^c q GP(q) + GP(q) q) \cdot N \\ (e_2 \cdot {}^c q GP(q) - GP(q) q e_2) \cdot N \\ (e_3 \cdot {}^c q GP(q) - GP(q) q e_3) \cdot N \\ (e_4 \cdot {}^c q GP(q) - GP(q) q e_4) \cdot N \end{pmatrix} \end{aligned}$$

### 10.0.18 Ball case

It is the case where  $GP = -N$ : for  $e_2$ :

$$\begin{aligned} (0, 1, 0, 0) \cdot (q_0, -\underline{p}) \cdot (0, -N) &= \\ \left( \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \cdot \underline{p}, \begin{pmatrix} q_0 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * \underline{p} \right) \cdot (0, -N) &= \\ \left( ?, -\underline{p}_x N - \left( \begin{pmatrix} q_0 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * \underline{p} \right) * N \right) &= \end{aligned}$$

and:

$$\begin{aligned} (0, -N) \cdot (q_0, \underline{p}) \cdot (0, 1, 0, 0) &= \\ (N \cdot \underline{p}, -q_0 N - N * \underline{p}) \cdot (0, 1, 0, 0) &= \\ \left( ?, (N \cdot \underline{p}) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * (q_0 N + N * \underline{p}) \right) &= \\ \left( ?, (N \cdot \underline{p}) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + q_0 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * N + \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * (N * \underline{p}) \right) &= \end{aligned}$$

sub then and get the resulting vector.N:

$$\begin{aligned} \left[ -\underline{p}_x N - N \cdot \underline{p} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + () * N - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * (N * \underline{p}) \right] \cdot N &= \\ -\underline{p}_x - N_x N \cdot \underline{p} + 0 - \left( \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} * (N * \underline{p}) \right) \cdot N &= \end{aligned}$$

using  $a * (b * c) = b(a \cdot c) - c(a \cdot b)$  leads to

$$-q_1 - N_x N \cdot \underline{p} - (q_1 N - N_x \underline{p}) \cdot N =$$

$$-q_1 - N_x N \cdot \underline{p} - q_1 + N_x N \cdot \underline{p} = -2q_1$$

for  $e1 = (1, 0, 0, 0)$ :

$$(q_0, -\underline{p}) \cdot (0, -N) = (?, -q_0 N + \underline{p} * N)$$

$$(0, -N) \cdot (q_0, \underline{p}) = (?, -q_0 N - \underline{p} * N)$$

So

$$\nabla_q H = \begin{pmatrix} N_x \\ N_y \\ N_z \\ -2q_0 \\ -2q_1 \\ -2q_2 \\ -2q_3 \end{pmatrix}$$

The question is  $(\nabla_q H)^t \delta q$  where  $\delta q$  is tangential to the sphere:

$$\delta q = (\cos((x + \delta x)/2), \sin((x + \delta x)/2)(p \delta p)) - (\cos(x/2), \sin(x/2)p)$$

$$\delta q = (\cos(x/2)\cos(\delta x/2) - \sin(x/2)\sin(\delta x/2), (\sin(x/2)\cos(\delta x/2) + \cos(x/2)\sin(\delta x/2)(p + \delta p) - (\cos(x/2), \sin(x/2)p))$$

$$\delta q = \left( \frac{\delta x \sin(x/2)}{2} + 0(\delta x)^2, \delta x/2 \cos(x/2)p + \delta p \sin(x/2) + 0(\delta x)^2 \right)$$

$$\delta q = (-\delta x/2 \sin(x/2) + 0(\delta x)^2, \delta x/2 \cos(x/2)p + \delta p \sin(x/2) + 0(\delta)^2)$$

and

$$(\nabla_q H)^t \delta q = -\delta x \cos(x/2) \sin(x/2) + \delta x \cos(x/2) \sin(x/2) + 2\delta p \sin(x/2) + O(\delta)^2$$

Moreover  $\delta p \cdot p = 0$

$$(\nabla_q H)^t \delta q = O(\delta)^2$$

That shows that this method is a second order approximation.

### 10.0.19 Case FC3D: using the local frame

$$\begin{pmatrix} m\dot{V} \\ I\dot{\omega} + \omega I\omega \end{pmatrix} = \begin{pmatrix} Fect + R \\ Mext + R * PG \end{pmatrix}$$

with \* vectoriel product,  $R$  reaction in the globla frame.  $P$  the point of contact.  $r$  is the reaction in the local frame.  $M^t r = R$  with:

$$M^t = \begin{pmatrix} nx & t_1x & t_2x \\ ny & t_1y & t_2y \\ nz & t_1z & t_2z \end{pmatrix}$$

we have :

$$\begin{pmatrix} R \\ R * PG \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & -PG_z & PG_y \\ PG_z & 0 & -PG_x \\ -PG_y & PG_x & 0 \end{pmatrix} \cdot R := N^t R = N^t M^t r$$

we want:

$$\begin{pmatrix} m\dot{V} \\ I\dot{\omega} + \omega I\omega \end{pmatrix} = jachq T^t r$$

So  $jachqt = MN$

# Chapter 11

## Projection On constraints

### 11.0.20 Velocity formulation

The first step consists in doing a velocity formulation of the system:

$$\begin{aligned} M\dot{v} &= F_{ext} + B\lambda \\ \dot{q} &= Tv \\ y &= h(q) \\ NSLAW(y, \lambda, \dots) \end{aligned} \tag{11.1}$$

The constraint  $\dot{q} = Tv$  is sufficient to keep a normal quaternion. Because of the speed formulation,  $h(q)$  could violate the NSLAW. A solution could be to add a formulation in position. We must underline that the constraints  $|Q| = 1$  is implicit in this system. Indeed, the direction  $\dot{q} = Tv$  is tangential to the sphere.

### 11.0.21 Position formulation

It consists in writing a position formulation on the system:

$$h(q) = \begin{pmatrix} HI(q) \\ HE(q) \end{pmatrix} \tag{11.2}$$

We are looking for  $q_1$  from  $q_0$ :

$$q_1 = q_0 + \nabla HI \Lambda_I + \nabla HE \Lambda_E \tag{11.3}$$

Assume that  $h(q_0)$  doesn't satisfy the constraints, ie  $HI(q_0) \not\perp 0$  or  $HE(q_0) \neq 0$ . Linearize  $h$  leads to:

$$0 \leq HI(q_0) + \nabla^t HI (\nabla HI \Lambda_I + \nabla HE \Lambda_E) \perp \Lambda_I \geq 0 \tag{11.4}$$

$$0 = HE(q_0) + \nabla^t HE (\nabla HI \Lambda_I + \nabla HE \Lambda_E) \tag{11.5}$$

The getting system could be written has a MLCP:

$$C \ni h(q_0) + \nabla^t (\nabla \Lambda), \Lambda \in C^* \tag{11.6}$$

In the case of a quaternion  $Q$  for the rotation representation, it is noteworthy that this system doesn't deal with the constraints  $|Q| = 1$ . Thus, the direction  $(q_1, q_0)$  can be normal to this constraint, in that case this approach doesn't work. (It happens in practice) The solution that consists in normaliaed  $q$  after this formulation is not convenient because, it could be incompatible with  $|Q| = 1$ . A better approach is to add this constraint.

### 11.0.22 Position formulation with the quaternion constraints

It consists in adding the constraint  $|Q| = 1$  in the system HE:

$$\tilde{H}E(q) = \begin{pmatrix} HE(q) \\ |Q| - 1 \end{pmatrix} \quad (11.7)$$

The formulation described above can be done.

## Chapter 12

# Simulation of a Cam Follower System

**Main Contributors:** *Mario di Bernardo, Gustavo Osorio, Stefania Santini*  
*University of Naples Federico II, Italy.*

The free body dynamics can be described by a linear second order system. An external input is considered acting directly on the follower. This input is a non linear forcing component coming from the valve. The follower motion is constrained to a phase space region bounded by the cam position. The non conservative Newton restitution law is used for the computation of the post impact velocity. The cam is assumed to be massive therefore only rotational displacement is allowed. Under these assumptions, the free body dynamics of the follower can be described by

$$\mu \frac{d^2 u(t)}{dt^2} + \zeta \frac{du(t)}{dt} + \kappa u(t) = f_v(t), \quad \text{if } u(t) > c(t). \quad (12.1)$$

where  $\mu$ ,  $\zeta$  and  $\kappa$  are constant parameters for the follower mass, friction viscous damping and spring stiffness respectively. The state of the follower is given by the position  $u(t)$  and velocity  $v(t) = \frac{du}{dt}$ . The external forcing is given by  $f_v(t)$ . The cam angular position determines  $c(t)$  that defines the holonomic (i.e. constraint only on the position) rheonomic (i.e. time varying) constraint. The dynamic behavior when impacts occurs (i.e.  $u(t) = c(t)$ ) is modelled via Newton's impact law that in this case is given by

$$v(t^+) = \frac{dc}{dt} - r \left( v(t^-) - \frac{dc}{dt} \right) = (1+r) \frac{dc}{dt} - rv(t^-), \quad \text{if } u(t) = c(t). \quad (12.2)$$

where  $v(t^+)$  and  $v(t^-)$  are the post and pre impact velocities respectively,  $\frac{dc}{dt}$  is the velocity vector of the cam at the contact point with the follower, and  $r \in [0, 1]$  is the restitution coefficient to model from plastic to elastic impacts. In Figure 12.1 is presented the schematic diagram of the physical cam-follower system. In Figure 12.1.a for  $t = 0$ , 12.1.b for  $t = \beta$ , and 12.1.c the profile of the constraint position  $\delta c(t)$ , velocity  $\frac{dc}{dt}(t)$  and acceleration  $\frac{d^2 c}{dt^2}(t)$ . It is possible to visualize the follower displacement as a function of the cam position. It is also important to notice that different types of cams and followers profiles are used in practical applications.

### 12.0.23 The cam-follower as a Lagrangian NSDS.

It is possible to completely describe the cam-follower system as a driven impact oscillator into the framework of *Lagrangian NSDS* using a translation in space. Setting  $\hat{u}(t) = u(t) - c(t)$  and  $\hat{v}(t) = v(t) - dc/dt$ , then equations (12.1) and (12.2) can be expressed as (the argument  $t$  will not be explicitly written)

$$\mu \frac{d^2 \hat{u}}{dt^2} + \zeta \frac{d\hat{u}}{dt} + \kappa \hat{u} = f_v - \left( \mu \frac{d^2 c}{dt^2} + \zeta \frac{dc}{dt} + \kappa c \right) \equiv \hat{f}, \quad \text{if } \hat{u} > 0. \quad (12.3)$$

$$\hat{v}^+ = -r\hat{v}^-, \quad \text{if } \hat{u} = 0. \quad (12.4)$$

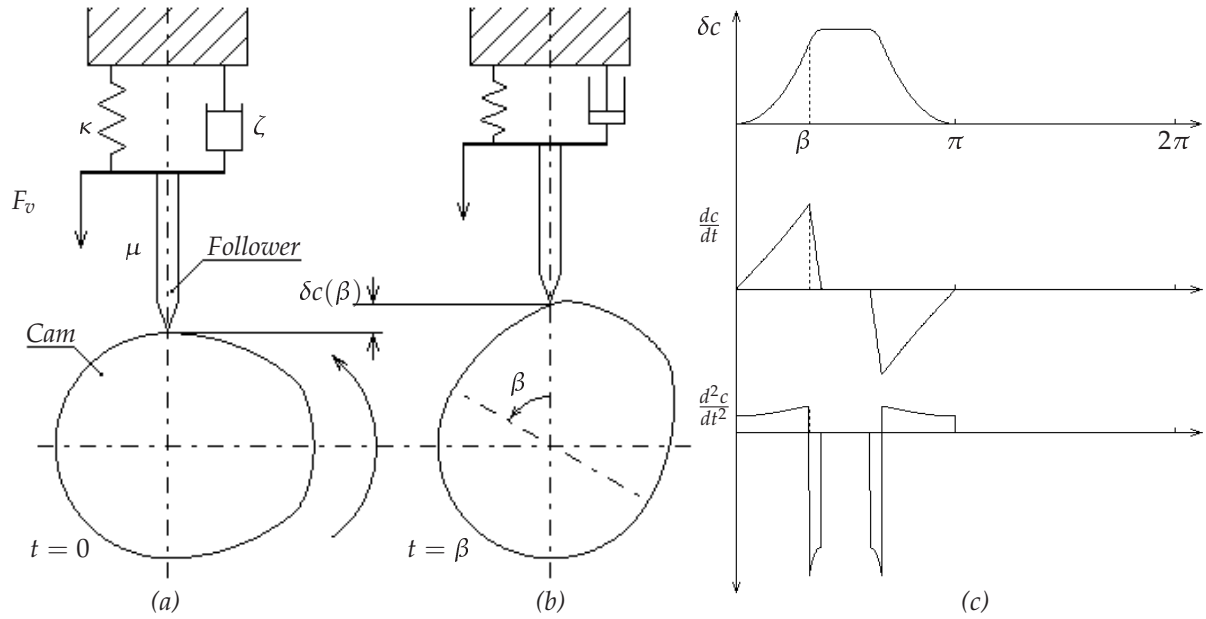


Figure 12.1: Cam-Shaft's schematics. (a)  $t=0$ . (b)  $t=\beta$ . (c) Constraint position  $\delta c(t)$ , velocity  $\frac{d\delta c}{dt}(t)$  and acceleration  $\frac{d^2\delta c}{dt^2}(t)$ .

Using the framework presented in [2] we have that the equation of motion of a Lagrangian system may be stated as follows :

$$M(q)\ddot{q} + Q(q, \dot{q}) + F(\dot{q}, q, t) = F_{ext}(t) + R \quad (12.5)$$

From the (12.3) we can derive all of the terms which define a Lagrangian NSDS. In our case the model is completely linear:

$$\begin{aligned} q &= \begin{bmatrix} \hat{u} \end{bmatrix} \\ M(q) &= \begin{bmatrix} \mu \end{bmatrix} \\ Q(q, \dot{q}) &= \begin{bmatrix} 0 \end{bmatrix} \\ F(q, \dot{q}) &= \begin{bmatrix} \zeta \end{bmatrix} \dot{q} + \begin{bmatrix} \kappa \end{bmatrix} q \\ F_{ext} &= \begin{bmatrix} \hat{f} \end{bmatrix} \end{aligned} \quad (12.6)$$

The unilateral constraint requires that:

$$\hat{u} \geq 0$$

so we can obtain

$$\begin{aligned} y &= H^T q + b \\ H^T &= \begin{bmatrix} 1 \end{bmatrix} \\ b &= 0 \end{aligned} \quad (12.7)$$

In the same way, the reaction force due to the constraint is written as follows:

$$R = H\lambda, \quad \text{with } H = \begin{bmatrix} 1 \end{bmatrix}$$

The unilatara contact law may be formulated as follow:

$$0 \leq y \perp \lambda \geq 0 \quad (12.8)$$

and the Newton's impact law:

$$\text{If } y = 0, \dot{y}^+ = -r\dot{y}^- \quad (12.9)$$

## 12.0.24 Implementation in the platform

For the simulation of the cam follower system follow the steps

1. Move to the working directory `sample/CamFollower`  
`$cd sample/CamFollower`
2. Clean the directory from binary files using the `siconos` command  
`$siconos -c`
3. Compile the file `CamFollowerNoXml.cpp` in the sample folder (See the code at the end of the section)  
`$siconos CamFollowerNoXml.cpp`
4. Change the simulation parameters (*i.e.* Follower initial position and velocity, cam initial angle, simulations time, cam rotational speed in rpm, etc.) in the file `CamFollowerNoXml.cpp`.

Next we present the sample code for the `CamFollowerNoXml.cpp` file:

```
int main(int argc, char* argv[]) {
    {
        // ===== Creation of the model =====
        // User-defined main parameters
        double rpm=358;
        double phi_0=0;
        unsigned int dsNumber = 1; // the Follower and the ground
        unsigned int nDof = 1; // degrees of freedom for the Follower
        double t0 = 0; // initial computation time
        double T = 5; // final computation time
        double h = 0.0001; // time step
        int Kplot;
        Kplot=(int)(Tplot/h);
        double position_init = 0.4; // initial position for lowest bead.
        double velocity_init = 0.4; // initial velocity for lowest bead.

        // ===== Dynamical systems =====

        vector<DynamicalSystem*> vectorDS; // the list of DS
        vectorDS.resize(dsNumber,NULL);

        SiconosMatrix *Mass, *K, *C; // mass/rigidity/viscosity
        Mass = new SiconosMatrix(nDof,nDof);
        (*Mass)(0,0) = 1.221;
        K = new SiconosMatrix(nDof,nDof);
        (*K)(0,0) = 1430.8;
        C = new SiconosMatrix(nDof,nDof);
        (*C)(0,0) = 0;

        // Initial positions and velocities
        vector<SimpleVector*> position_0;
        vector<SimpleVector*> velocity_0;
        position_0.resize(dsNumber,NULL);
        velocity_0.resize(dsNumber,NULL);
        position_0[0] = new SimpleVector(nDof);
        velocity_0[0] = new SimpleVector(nDof);
    }
}
```

```

(*position_0[0])(0) = position_init;
(*velocity_0[0])(0) = velocity_init;

vectorDS[0] =
new LagrangianLinearTIDS(0,nDof,*(position_0[0]),*(velocity_0[0]),*Mass,*K,*C);

static_cast<LagrangianDS*>(vectorDS[0])
    ->setComputeFExtFunction("FollowerPlugin.so", "FollowerFExt");

// Example to set a list of parameters in FExt function.
// 1 - Create a simple vector that contains the required parameters.

// Here we set two parameters, the DS number.
SimpleVector * param = new SimpleVector(2);

(*param)(0)=rpm;
(*param)(1)=phi_0;
// 2 - Assign this param to the function FExt
static_cast<LagrangianDS*>(vectorDS[0])>setParametersListPtr(param,2);
// 2 corresponds to the position of FExt in the stl vector of possible parameters.
// 0 is mass, 1 FInt.
// Now the cam rotational velocity in rpms will be available in FExt plugin.

// ===== Interactions =====

vector<Interaction*> interactionVector;
interactionVector.resize(1,NULL);
vector<DynamicalSystem*> *dsConcerned =
    new vector<DynamicalSystem*>(dsNumber);

// ===== Non Smooth Law =====
double e = 0.8;
// Interaction Follower-floor
SiconosMatrix *H = new SiconosMatrix(1,nDof);
(*H)(0,0) = 1.0;
NonSmoothLaw * nslaw = new NewtonImpactLawNSL(e);
Relation * relation = new LagrangianLinearR(*H);
(*dsConcerned)[0] = vectorDS[0];
interactionVector[0] = new Interaction("Follower-Ground",0,1, dsConcerned);
interactionVector[0]>setRelationPtr(relation);
interactionVector[0]>setNonSmoothLawPtr(nslaw);
// ===== Interactions =====

// ===== NonSmoothDynamicalSystem =====
bool isBVP =0;
NonSmoothDynamicalSystem * nsds =
    new NonSmoothDynamicalSystem(isBVP);

// Set DS of this NonSmoothDynamicalSystem
nsds->setDynamicalSystems(vectorDS);
// Set interactions of the NonSmoothDynamicalSystem
nsds->setInteractions(interactionVector);

// ===== Model =====

```



```

Model * Follower = new Model(t0,T);
// set NonSmoothDynamicalSystem of this model
Follower->setNonSmoothDynamicalSystemPtr(nsds);

// ===== Strategy =====

double theta = 0.5;      // theta for Moreau integrator
string solverName = "QP" ;

Strategy* S = new TimeStepping(Follower);

// - Time discretisation -
TimeDiscretisation * t = new TimeDiscretisation(h,S);

// - OneStepIntegrators -
vector<OneStepIntegrator *> vOSI;
vOSI.resize(dsNumber,NULL);
vOSI[0] = new Moreau(t,vectorDS[0],theta);
S->setOneStepIntegrators(vOSI);

// - OneStepNsProblem -
OneStepNSProblem * osnspb = new LCP(S,solverName,101, 0.0001,"max",0.6);
S->setOneStepNSProblemPtr(osnspb); // set OneStepNSProblem of the strategy
cout << "=== End of model loading === " << endl;
// ==== End of model definition=====

// ===== Computation=====

// — Strategy initialization —
S->initialize();
cout << "End of strategy initialisation" << endl;

int k = t->getK();           // Current step
int N = t->getNSteps();      // Number of time steps

// — Get the values to be plotted —
// -> saved in a matrix dataPlot
unsigned int outputSize = 8;

SiconosMatrix DataPlot(Kplot+1,outputSize );
// For the initial time step:

// time
DataPlot(k,0) = k*t->getH();

DataPlot(k,1) = static_cast<LagrangianDS*>(vectorDS[0])->getQ()(0);
DataPlot(k,2) = static_cast<LagrangianDS*>(vectorDS[0])->getVelocity()(0);
DataPlot(k,3) = (Follower->getNonSmoothDynamicalSystemPtr()->
    getInteractionPtr(0)->getLambda(1))(0);
DataPlot(k,4) = static_cast<LagrangianDS*>(vectorDS[0])->getFExt()(0);

```

```

// State of the Cam
double CamEqForce,CamPosition,CamVelocity,CamAcceleration;
CamEqForce=
    CamState(k*t->getH(),rpm,CamPosition,CamVelocity,CamAcceleration);
// Position of the Cam
DataPlot(k, 5) = CamPosition;
// Velocity of the Cam
DataPlot(k, 6) = CamVelocity;
// Acceleration of the Cam
DataPlot(k, 7) =
    CamPosition+static_cast<LagrangianDS*>(vectorDS[0])->getQ(0);

// — Time loop —
cout << "Start computation ... " << endl;
while(k < N)
{
    // — Get values to be plotted —
    DataPlot(k,0) = k*t->getH();

    DataPlot(k,1) =
        static_cast<LagrangianDS*>(vectorDS[0])->getQ(0);
    DataPlot(k,2) =
        static_cast<LagrangianDS*>(vectorDS[0])->getVelocity(0);
    DataPlot(k,3) =
        (Follower->getNonSmoothDynamicalSystemPtr()->
        getInteractionPtr(0)->getLambda(1))(0);
    DataPlot(k,4) = static_cast<LagrangianDS*>(vectorDS[0])->getFExt(0);

    CamEqForce=
    CamState(k*t->getH(),rpm,CamPosition,CamVelocity,CamAcceleration);

    DataPlot(k, 5) = CamPosition;
    DataPlot(k, 6) = CamVelocity;
    DataPlot(k, 7) = CamPosition+
        static_cast<LagrangianDS*>(vectorDS[0])->getQ(0);
    // transfer of state i+1 into state i and time incrementation
    S->nextStep();
    // get current time step
    k = t->getK();
    // solve ...
    S->computeFreeState();
    S->computeOneStepNSProblem();
    // update
    S->update();
}
// — Output files —
DataPlot.rawWrite("result.dat", "ascii");
// — Free memory —
delete osnspb;
delete vOSI[0];
delete t;
delete S;
delete Follower;

```

```
    delete nsds;  
    delete interactionVector[0];  
    delete relation;  
    delete nslaw;  
    delete H;  
    delete dsConcerned;  
    delete vectorDS[0];  
    delete position_0[0];  
    delete velocity_0[0];  
    delete C;  
    delete K;  
    delete Mass;  
}
```

### 12.0.25 Simulation

We have perform the simulation of the cam follower system for different values of the cam rotational speed with the SICONOS software package using a time-stepping numerical scheme with step size ( $h = 1e^{-4}$ ) and an event-driven scheme with minimum step size ( $h_{min} = 1e^{-12}$ ). Fig. 12.2 and 12.3 show the time simulations for different values of the cam rotational speed and Fig. 12.4 show the chaotic attractor at  $rpm = 660$  for impact and stroboscopic Poincarè sections.

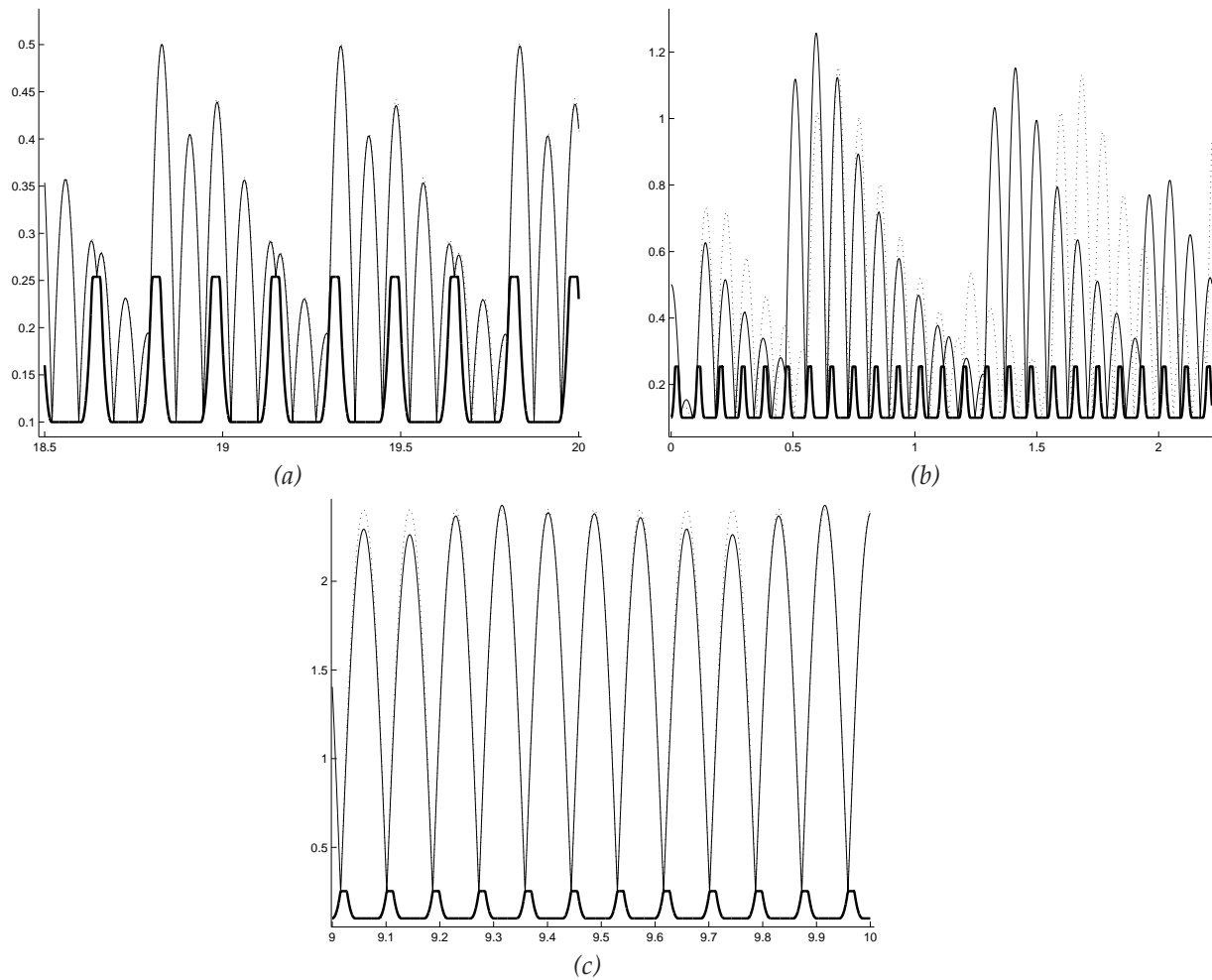


Figure 12.2: Time series using SICONOS platform. Time-stepping scheme (continuous line). Event-driven scheme (dashed line) (a) rpm=358. (b) rpm=660. (c) rpm=700.

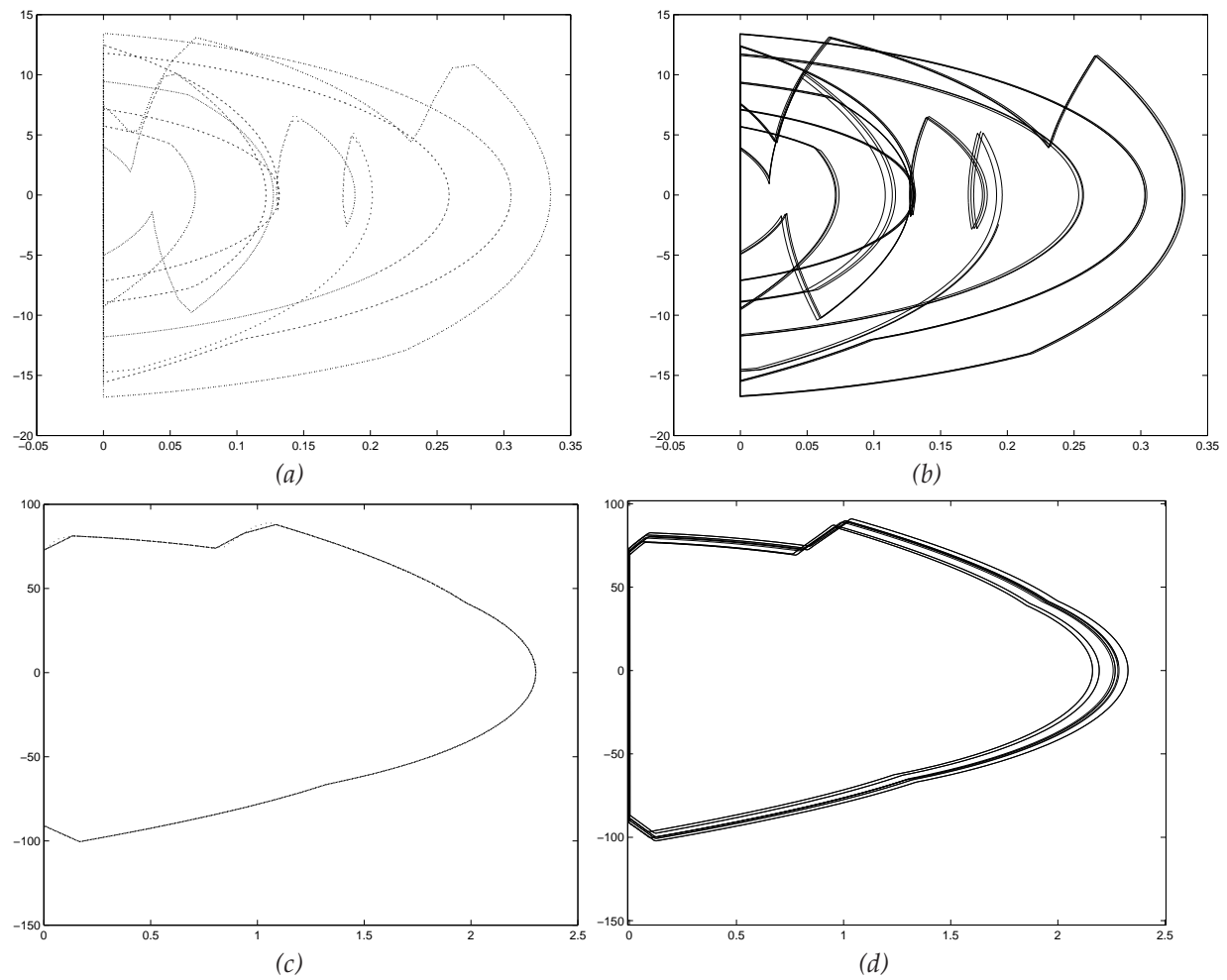


Figure 12.3: State space comparison using SICONOS platform. (a) rpm=358. Event Driven (b) rpm=358. Time Stepping ( $h = 1e^{-4}$ ) (c) rpm=700. Event Driven (d) rpm=700. Time Stepping ( $h = 1e^{-4}$ )

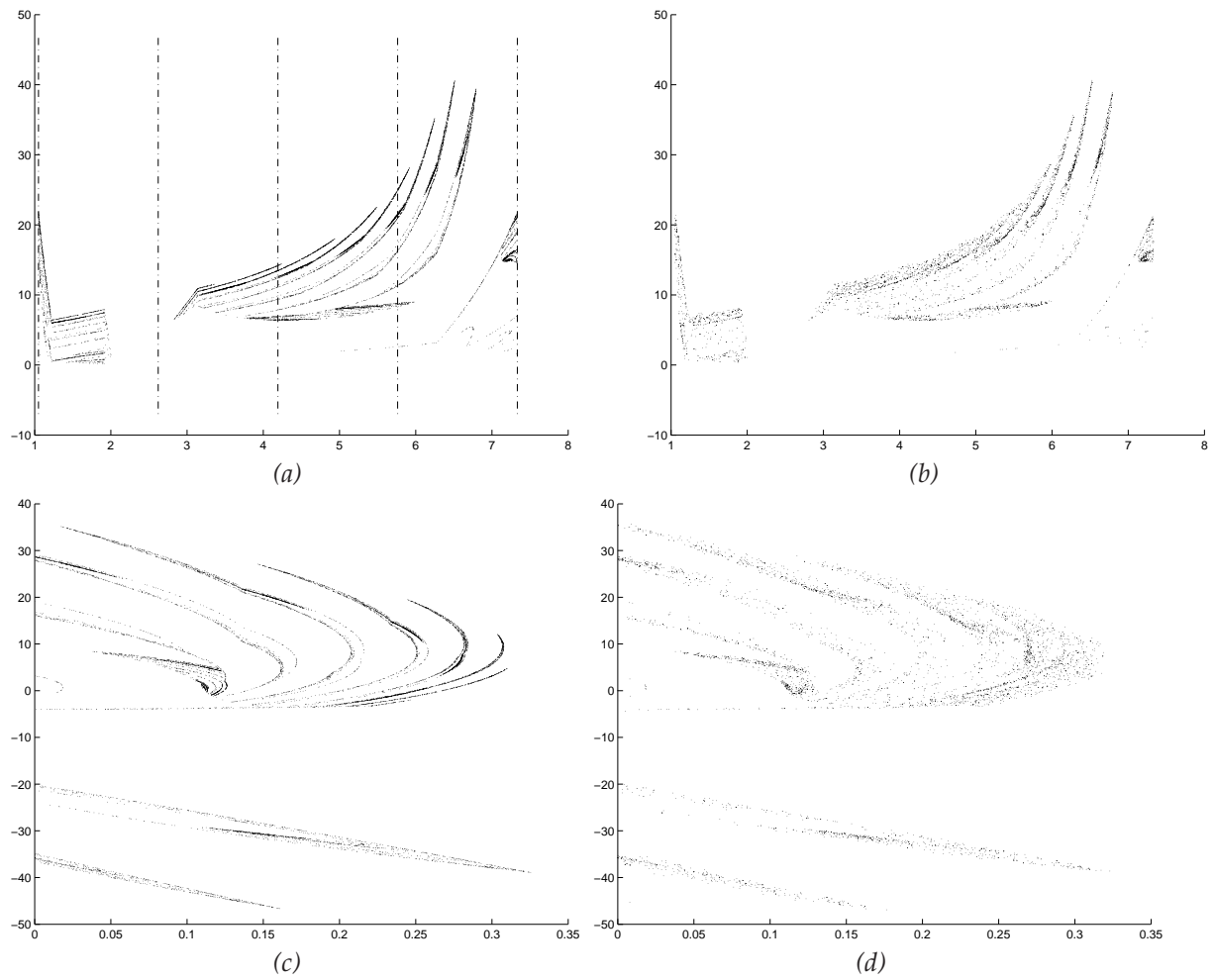


Figure 12.4: Attractors comparison using SICONOS platform at rpm=660. (a) Impact map. (Event Driven) (b) Impact Map. Time Stepping ( $h = 1e^{-4}$ ) (c) Stroboscopic map. (Event Driven) (d) Stroboscopic Map. Time Stepping ( $h = 1e^{-4}$ )

## Chapter 13

# Quartic Formulation

### 13.0.26 Slidding ?

It consists in finding  $\alpha > 0$  and  $R \in \partial K_\mu$  such that  $-\alpha \begin{pmatrix} 0 \\ R_T \end{pmatrix} = MR + q$ . That is :

$$\left[ M + \begin{pmatrix} 0 & 0 & 0 \\ 0 & \alpha & 0 \\ 0 & 0 & \alpha \end{pmatrix} \right] R + q = 0 \quad (13.1)$$

#### 13.0.26.a $R_T$ is on a conic

The first line of the system 13.1 and the  $R \in \partial K_\mu$  is the intersection between a plan and a cone in  $\mathbb{R}^3$ , ended:

$$\begin{aligned} \mu R_N &= \| R_T \| \\ \frac{M_{11}}{\mu} \| R_T \| &= -q_1 - M_{12}R_{T1} - M_{13}R_{T2} \end{aligned} \quad (13.2)$$

That is:

$$\begin{aligned} \mu^2 R_N^2 &= (R_{T1}^2 + R_{T2}^2) \\ \frac{M_{11}^2}{\mu^2} (R_{T1}^2 + R_{T2}^2) &= (-q_1 - M_{12}R_{T1} - M_{13}R_{T2})^2 \end{aligned} \quad (13.3)$$

That means that  $R_T$  is contained in a conic, focus and directrice are:

$$\begin{aligned} \mathcal{D} : q_1 + M_{12}R_{T1} + M_{13}R_{T2} &= 0 \\ \text{focus} : \mathcal{O} \\ \frac{M_{11}^2}{\mu^2} \text{Dist}(\mathcal{O}, R_T)^2 &= \text{Dist}(\mathcal{D}, R_T)^2 (M_{12}^2 + M_{13}^2) \\ \frac{\text{Dist}(\mathcal{O}, R_T)}{\text{Dist}(\mathcal{D}, R_T)} &= \frac{\mu \sqrt{(M_{12}^2 + M_{13}^2)}}{M_{11}} = e \end{aligned} \quad (13.4)$$

The parametric equation is:

$$\begin{aligned} R_{T1} &= r \cos(\theta) \\ R_{T2} &= r \sin(\theta) \\ r &= \frac{p}{1 + e \cos(\theta - \phi)} \end{aligned} \quad (13.5)$$

With  $p$  an simple expression of  $M_{11}, M_{12}, M_{13}$ , and  $\phi$  a constant angle between  $\mathcal{D}$  and  $(O, R_{T1})$

#### 13.0.26.b The two last line of the system 13.1

$$\frac{\| R_T \|}{\mu} \tilde{M}_1 + \left( \tilde{M} + \begin{pmatrix} \alpha & 0 \\ 0 & \alpha \end{pmatrix} \right) R_T + \tilde{q} = 0 \quad (13.6)$$

$\tilde{M}$  is symetric, so it exists a unitary matrix  $V$  such that  $V\tilde{M}V^T = \begin{pmatrix} d_1 & 0 \\ 0 & d_2 \end{pmatrix}$ . One can get:

$$\frac{\|R_T\|}{\mu} V\tilde{M}_1 + V \left( \tilde{M} + \begin{pmatrix} \alpha & 0 \\ 0 & \alpha \end{pmatrix} \right) V^T V R_T + V\tilde{q} = 0 \quad (13.7)$$

Rename:

$$\frac{\|\bar{R}_T\|}{\mu} \bar{M}_1 + \begin{pmatrix} d_1 + \alpha & 0 \\ 0 & d_2 + \alpha \end{pmatrix} \bar{R}_T + \bar{q} = 0 \quad (13.8)$$

In the plan, either  $V$  is a rotation or a symetrie. So  $\bar{R}_T = V R_T$  is a conic with the same focus and a rotated directrice, it means that it exists  $\phi_1$  such that :

$$\begin{aligned} \bar{R}_{T1} &= r \cos(\theta) \\ \bar{R}_{T2} &= r \sin(\theta) \\ r &= \frac{p}{1 + e \cos(\theta - \phi_1)} \end{aligned} \quad (13.9)$$

The equation 13.8 is :

$$\begin{aligned} (d_1 + \alpha) \bar{R}_{T1} &= -\bar{q}_1 + a_1 \|\bar{R}_T\| \\ (d_2 + \alpha) \bar{R}_{T2} &= -\bar{q}_2 + a_2 \|\bar{R}_T\| \end{aligned} \quad (13.10)$$

The case ( $\bar{R}_{T1} = 0$  or  $\bar{R}_{T2} = 0$ ) has to be examine. We try to eliminate  $\alpha$ :

$$\begin{aligned} d_1 \bar{R}_{T1} \bar{R}_{T2} + \alpha \bar{R}_{T1} \bar{R}_{T2} &= -\bar{q}_1 \bar{R}_{T2} + a_1 \bar{R}_{T2} \|\bar{R}_T\| \\ d_2 \bar{R}_{T1} \bar{R}_{T2} + \alpha \bar{R}_{T1} \bar{R}_{T2} &= -\bar{q}_2 \bar{R}_{T1} + a_2 \bar{R}_{T1} \|\bar{R}_T\| \end{aligned} \quad (13.11)$$

that leads to:

$$(d_1 - d_2) \bar{R}_{T1} \bar{R}_{T2} = -\bar{q}_1 \bar{R}_{T2} + \bar{q}_2 \bar{R}_{T1} + (a_1 \bar{R}_{T2} - a_2 \bar{R}_{T1}) \|\bar{R}_T\| \quad (13.12)$$

The parametric expression of  $\bar{R}_T$  leads to:

$$\begin{aligned} (d_1 - d_2) r^2 \cos(\theta) \sin(\theta) &= -\bar{q}_1 r \sin(\theta) + \bar{q}_2 r \cos(\theta) + r(a_1 r \sin(\theta) - a_2 r \cos(\theta)) \\ \text{ie: } (d_1 - d_2) r \cos(\theta) \sin(\theta) &= -\bar{q}_1 \sin(\theta) + \bar{q}_2 \cos(\theta) + r(a_1 \sin(\theta) - a_2 \cos(\theta)) \end{aligned} \quad (13.13)$$

with the expression of  $r$ :

$$\begin{aligned} (d_1 - d_2) \frac{p}{1 + e \cos(\theta - \phi_1)} \cos(\theta) \sin(\theta) &= \\ -\bar{q}_1 \sin(\theta) + \bar{q}_2 \cos(\theta) + \frac{p}{1 + e \cos(\theta - \phi_1)} (a_1 \sin(\theta) - a_2 \cos(\theta)) \\ \text{ie: } (d_1 - d_2) p \cos(\theta) \sin(\theta) &= \\ (1 + e \cos(\theta - \phi_1)) (-\bar{q}_1 \sin(\theta) + \bar{q}_2 \cos(\theta)) + p(a_1 \sin(\theta) - a_2 \cos(\theta)) \end{aligned} \quad (13.14)$$

$$\begin{aligned} \text{ie: } (d_1 - d_2) p \cos(\theta) \sin(\theta) &= \\ (1 + e(\cos(\theta) \cos(\phi_1) + \sin(\theta) \sin(\phi_1))) (-\bar{q}_1 \sin(\theta) + \bar{q}_2 \cos(\theta)) + p(a_1 \sin(\theta) - a_2 \cos(\theta)) \\ \text{ie: } (d_1 - d_2) p \cos(\theta) \sin(\theta) + \\ (1 + e \cos(\theta) \cos(\phi_1) + e \sin(\theta) \sin(\phi_1)) (\bar{q}_1 \sin(\theta) - \bar{q}_2 \cos(\theta)) + p(-a_1 \sin(\theta) + a_2 \cos(\theta)) &= 0 \end{aligned}$$

rename :

$$A \cos(\theta)^2 + B \sin(\theta)^2 + C \sin(\theta) \cos(\theta) + D \sin(\theta) + E \cos(\theta) = 0 \quad (13.15)$$

with

$$\begin{aligned} A &= -e \bar{q}_2 \cos(\phi_1) \\ B &= e \bar{q}_1 \sin(\phi_1) \\ C &= (d_1 - d_2) p + e \cos(\phi_1) \bar{q}_1 - e \sin(\phi_1) \bar{q}_2 \\ D &= \bar{q}_1 - p a_1 \\ E &= -\bar{q}_2 + p a_2 \end{aligned} \quad (13.16)$$



rename : Using the following set of unknown :

$$\begin{aligned} t &= \tan(\theta/2) \\ \sin(\theta) &= \frac{2t}{1+t^2} \\ \cos(\theta) &= \frac{1-t^2}{1+t^2} \end{aligned} \quad (13.17)$$

leads to:

$$\begin{aligned} A \frac{(1-t^2)^2}{1+t^2} + B \frac{4t^2}{1+t^2} + C \frac{2t(1-t^2)}{1+t^2} + D2t + E(1-t^2) &= 0 \\ \text{ie: } A(1-t^2)^2 + 4Bt^2 + C2t(1-t^2) + 2Dt(1+t^2) + E(1-t^2)(1+t^2) &= 0 \end{aligned} \quad (13.18)$$

$$\text{ie: } P_4 = A - E \quad P_3 = -2C + 2D \quad P_2 = 4B - 2A \quad P_1 = 2C + 2D \quad P_0 = A + E$$

Finally, we get 4 possible values for  $R_T$ , checking the sign of  $\alpha$  and  $R_N$  selects the solutions.

**13.0.26.c** case  $R_{T12} = 0$

From 13.10,  $R_{T1}$  leads to:

$$\begin{aligned} \| R_T \| &= |\bar{R}_{T2}| = \frac{\bar{q}_1}{a_1} \\ \bar{R}_T &= \begin{pmatrix} 0 \\ \pm \frac{\bar{q}_1}{a_1} \end{pmatrix} \end{aligned} \quad (13.19)$$

From 13.10,  $R_{T2}$  leads to:

$$\begin{aligned} \| R_T \| &= |\bar{R}_{T1}| = \frac{\bar{q}_2}{a_2} \\ \bar{R}_T &= \begin{pmatrix} \pm \frac{\bar{q}_2}{a_2} \\ 0 \end{pmatrix} \end{aligned} \quad (13.20)$$

From  $\bar{R}_T$ , we have to check the coherence with the equation 13.9. If it is on the conic, we compute  $R$ , and the sign condition of the equation 13.1 must be check.