### Recent developments in simulation, optimization and control of flexible multibody systems

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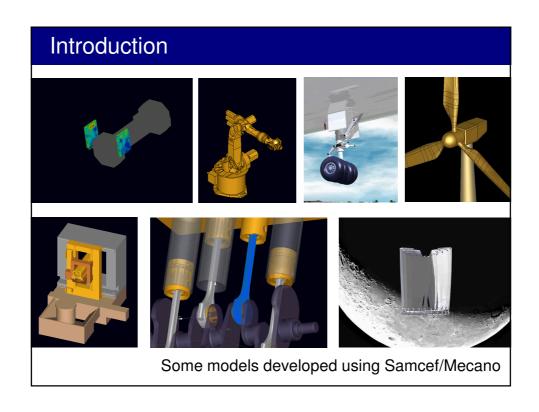


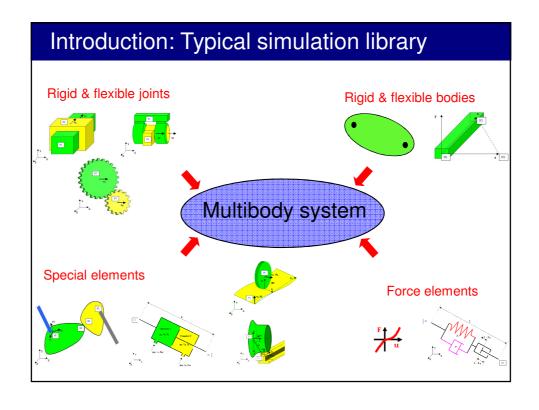
Katholieke Universiteit Leuven, January 28, 2010



### Outline

- Introduction
- ☐ Modelling of multibody & mechatronic systems
- ☐ Time integration algorithms
- ☐ Topology optimization of structural components





### Introduction: Commercial simulation tools

Multibody dynamics approach

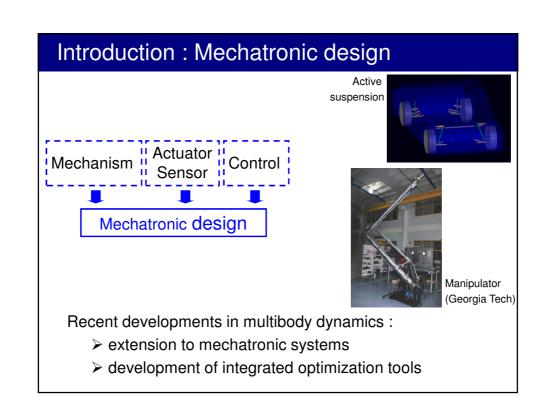
- > MSC ADAMS
- > LMS VIRTUAL LAB MOTION
- > SIMPACK
- > RECURDYN

Linear flexibility effects can be represented

Finite element approach

- > SAMCEF MECANO
- > OOFELIE

Most general approach for nonlinear flexible systems



### Outline

- Introduction
- ☐ Modelling of multibody & mechatronic systems
  - > Modelling of flexible multibody systems
  - > Modelling of coupled mechatronic systems
  - > Application to a semi-active car suspension
- ☐ Time integration algorithms
- ☐ Topology optimization of structural components

### Modelling of flexible multibody systems





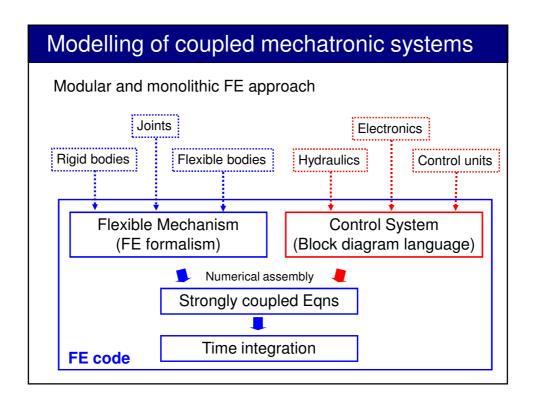
- > translations & rotations
- geometric nonlinearities

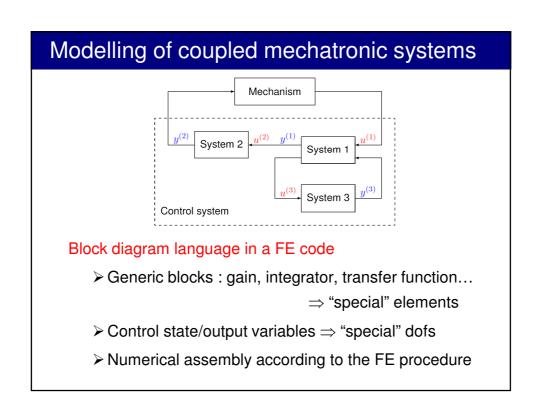
Kinematic joints

➤ algebraic constraints

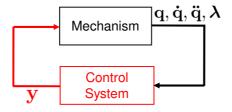
$$\mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{g}_{gyr}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{g}_{int}(\mathbf{q}) + \mathbf{\Phi}_{\mathbf{q}}^T \boldsymbol{\lambda} = \mathbf{g}_{ext}$$
  
 $\mathbf{\Phi}(\mathbf{q}, t) = \mathbf{0}$ 

index-3 DAE with rotation variables





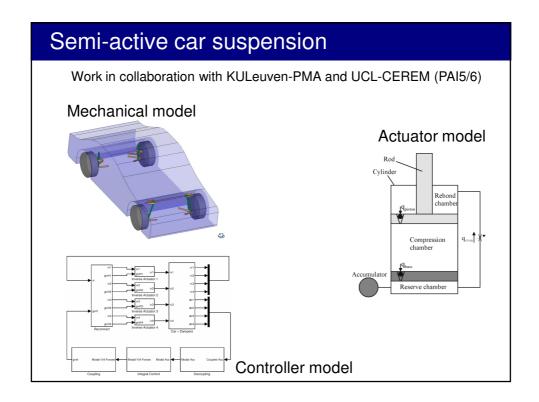
### Modelling of coupled mechatronic systems

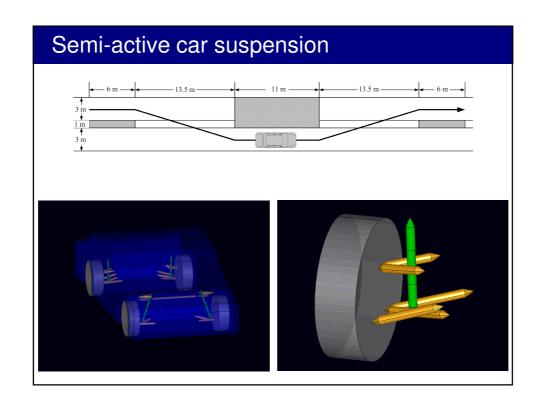


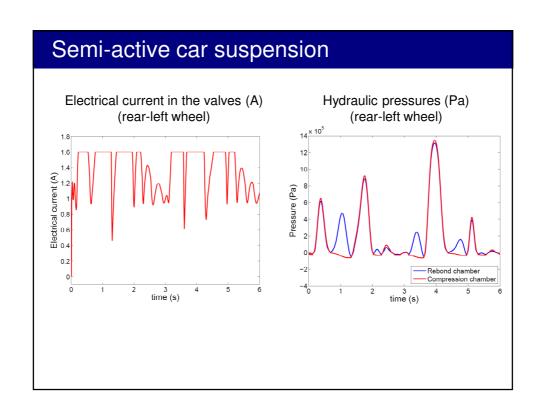
Coupled equations:  $\begin{aligned} \mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} &= \mathbf{g}(\mathbf{q},\dot{\mathbf{q}},t) - \mathbf{\Phi}_{\mathbf{q}}^T \boldsymbol{\lambda} + \mathbf{L}\mathbf{y} \\ \mathbf{0} &= \mathbf{\Phi}(\mathbf{q},t) \\ \dot{\mathbf{x}} &= \mathbf{f}(\mathbf{q},\dot{\mathbf{q}},\ddot{\mathbf{q}},\boldsymbol{\lambda},\mathbf{x},\mathbf{y},t) \\ \mathbf{y} &= \mathbf{h}(\mathbf{q},\dot{\mathbf{q}},\ddot{\mathbf{q}},\boldsymbol{\lambda},\mathbf{x},\mathbf{y},t) \end{aligned}$ 

Time-integration scheme for coupled 1st/2nd order DAE?

- ➤ Classical ODE solvers : multistep & Runge-Kutta methods
- ightharpoonup Generalized- $\alpha$  time integration scheme







## Strongly coupled simulation of mechatronic systems: Mechanical equations are obtained using the finite element technique (rigid bodies, elastic bodies & kinematic joints) Control equations are formulated in the FE code using the block diagram language The generalized-α time integrator is used to solve the strongly coupled problem

# Outline Introduction Modelling of multibody & mechatronic systems Time integration algorithms Generalized-α method Kinematic constraints Treatment of rotation variables Controller dynamics Topology optimization of structural components

### Generalized- $\alpha$ method

Numerical time-integration methods

- > Standard integrators: multistep, Runge-Kutta
- Methods from structural dynamics (Newmark, HHT, g-α)
- Energy conserving schemes

Generalized-α method [Chung & Hulbert 1993]

- ➤ One step method for 2<sup>nd</sup> ODEs
- ➤ 2<sup>nd</sup> order accuracy
- Unconditional stability (A-stability) for linear problems
- Controllable numerical damping at high frequencies
- > Computational efficiency for large and stiff problems
- $\Rightarrow$  Extensions of the g- $\alpha$  method to deal with kinematic constraints, rotation variables and controller dynamics?

### Generalized- $\alpha$ method

2nd order ODE system:  $\mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} = \mathbf{g}(\mathbf{q}, \dot{\mathbf{q}}, t)$ 

Newmark implicit formulae:

$$\mathbf{q}_{n+1} = \mathbf{q}_n + h\dot{\mathbf{q}}_n + h^2(0.5 - \beta)\mathbf{a}_n + h^2\beta\mathbf{a}_{n+1}$$
  
$$\dot{\mathbf{q}}_{n+1} = \dot{\mathbf{q}}_n + h(1 - \gamma)\mathbf{a}_n + h\gamma\mathbf{a}_{n+1}$$

 $\begin{center} Generalized-$\alpha$ method [Chung \& Hulbert, 1993] \end{center}$ 

$$(1 - \alpha_m)\mathbf{a}_{n+1} + \alpha_m\mathbf{a}_n = (1 - \alpha_f)\ddot{\mathbf{q}}_{n+1} + \alpha_f\ddot{\mathbf{q}}_n$$

To be solved with :  $\mathbf{M}(\mathbf{q}_{n+1})\ddot{\mathbf{q}}_{n+1} = \mathbf{g}(\mathbf{q}_{n+1},\dot{\mathbf{q}}_{n+1},t_{n+1})$ 

- ightharpoonup Two kinds of acceleration variables:  $\mathbf{a}_n 
  eq \ddot{\mathbf{q}}_n$
- > Algorithmic parameters:  $\gamma$ ,  $\beta$ ,  $\alpha_f$ ,  $\alpha_m$  2nd order accuracy & numerical damping

### Kinematic constraints

$$\mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} = \mathbf{g}(\mathbf{q}, \dot{\mathbf{q}}, t) - \mathbf{\Phi}_{\mathbf{q}}^{T} \boldsymbol{\lambda}$$
$$\mathbf{0} = \mathbf{\Phi}(\mathbf{q}, t)$$

Direct integration of the index-3 DAE problem using  $g-\alpha$ 

- Linear stability analysis demonstrates the importance of numerical damping [Cardona & Géradin 1989]
- Scaling of equations and variables reduces the sensitivity to numerical errors [Bottasso, Bauchau & Cardona 2007]
- ➤ Global convergence is demonstrated [Arnold & B. 2007]

### Reduced index formulations

[Lunk & Simeon 2006; Jay & Negrut 2007; Arnold 2009]

### Treatment of rotations

It is impossible to have a global 3-dimensional parameterization of rotations without singular points. [Stuelpnagel 1964]

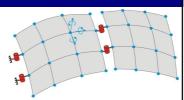
### Possible strategies

- ➤ 3-dimensional parameterization + reparameterization to avoid singularities [Cardona & Géradin 1989]
- ➤ Higher dimensional parameterization + kinematic constraints [Betsch & Steinmann 2001]
- Rotationless formulation, e.g. ANCF [Shabana]
- ➤ Lie group time integrator: no parameterization of the manifold is required *a priori* [Simo & Vu-Quoc 1988; B. & Cardona 2010]

### Treatment of rotations

### Nodal coordinates

$$q = (\mathbf{x}_1, \dots, \mathbf{x}_N, \mathbf{R}_1, \dots, \mathbf{R}_N)$$



The configuration evolves on the n-dimensional Lie group

$$G = \mathbb{R}^3 \times \ldots \times \mathbb{R}^3 \times SO(3) \times \ldots \times SO(3)$$

with the composition  $q_{tot} = q_1 \circ q_2$  such that

$$\mathbf{x}_{i,tot} = \mathbf{x}_{i,1} + \mathbf{x}_{i,2}$$
 and  $\mathbf{R}_{i,tot} = \mathbf{R}_{i,1}\mathbf{R}_{i,2}$ 

Constrained equations of motion:

$$\begin{aligned} \dot{q} &= DL_q(e) \cdot \widetilde{\mathbf{v}} \\ \mathbf{M}(q)\dot{\mathbf{v}} + \mathbf{g}(q, \mathbf{v}, t) + \mathbf{B}^T(q)\boldsymbol{\lambda} &= \mathbf{0} \\ \boldsymbol{\Phi}(q) &= \mathbf{0} \end{aligned}$$

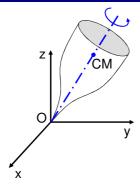
### Treatment of rotations

$$\begin{split} \mathbf{M}(q_{n+1})\dot{\mathbf{v}}_{n+1} + \mathbf{g}(q_{n+1},\mathbf{v}_{n+1},t_{n+1}) + \mathbf{B}^T(q_{n+1})\boldsymbol{\lambda}_{n+1} &= \mathbf{0} \\ \boldsymbol{\Phi}(q_{n+1}) &= \mathbf{0} \\ q_{n+1} &= \varphi_h(q_n,\mathbf{v}_n,\mathbf{a}_n,\mathbf{a}_{n+1}) \\ \mathbf{v}_{n+1} &= \mathbf{v}_n + (1-\gamma)h\mathbf{a}_n + \gamma h\mathbf{a}_{n+1} \\ (1-\alpha_m)\mathbf{a}_{n+1} + \alpha_m\mathbf{a}_n &= (1-\alpha_f)\dot{\mathbf{v}}_{n+1} + \alpha_f\dot{\mathbf{v}}_n \\ \mathbf{with, e.g.,} & \varphi_h^{(1)}(q_n,\mathbf{v}_n,\mathbf{a}_n,\mathbf{a}_{n+1}) &= q_n \circ \exp\left(h\widetilde{\mathbf{v}_n} + h^2(0.5-\beta)\widetilde{\mathbf{a}_n} + \beta h^2\widetilde{\mathbf{a}_{n+1}}\right) \end{split}$$

### Lie group generalized- $\alpha$ solver [B. & Cardona 2010]

- 1. No rotation parameterization is explicitly involved
- 2. The solution inherently remains on the manifold
- 3. The integration formulae are nonlinear
- 4. The classical generalized- $\alpha$  algorithm is a special case when G is a linear vector space
- 5. Second-order accuracy + numerical damping
- 6. Much simpler than parameterization-based methods

### Treatment of rotations: benchmark

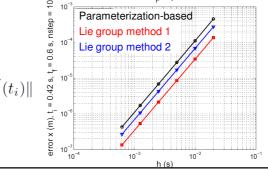


Spinning top with spherical ellipsoid of inertia and constant follower torque

$$\dot{\mathbf{R}} = \mathbf{R}\widetilde{\Omega}$$

$$\mathbf{J}\dot{\mathbf{\Omega}} + \mathbf{\Omega} \times \mathbf{J}\mathbf{\Omega} = \mathbf{C}$$

⇒ analytical solution [Romano 2008]



error = 
$$\frac{1}{n_e} \sum_{i=1}^{n_e} \|\mathbf{x}(t_i) - \mathbf{x}^{ref}(t_i)\|$$

### Controller dynamics

Coupled dynamic equations:  $\begin{cases} \ddot{q} = g \\ \dot{r} = f \end{cases}$ 

$$\begin{cases} \ddot{\mathbf{q}} = \mathbf{g}(\mathbf{q}, \dot{\mathbf{q}}, \mathbf{x}, t) \\ \dot{\mathbf{x}} = \mathbf{f}(\mathbf{q}, \dot{\mathbf{q}}, \mathbf{x}, t) \end{cases}$$

$$\mathbf{q}_{n+1} = \mathbf{q}_n + h\dot{\mathbf{q}}_n + h^2(0.5 - \beta)\mathbf{a}_n + h^2\beta\mathbf{a}_{n+1}$$

$$\dot{\mathbf{q}}_{n+1} = \dot{\mathbf{q}}_n + h(1-\gamma)\mathbf{a}_n + h\gamma\mathbf{a}_{n+1}$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + h(1-\theta)\mathbf{w}_n + h\theta\mathbf{w}_{n+1}$$

$$\begin{cases} (1 - \alpha_m)\mathbf{a}_{n+1} + \alpha_m \mathbf{a}_n &= (1 - \alpha_f)\ddot{\mathbf{q}}_{n+1} + \alpha_f \ddot{\mathbf{q}}_n \\ (1 - \delta_m)\mathbf{w}_{n+1} + \delta_m \mathbf{w}_n &= (1 - \delta_f)\dot{\mathbf{x}}_{n+1} + \delta_f \dot{\mathbf{x}}_n \end{cases}$$

To be solved with : 
$$\begin{cases} \ddot{\mathbf{q}}_{n+1} &= \mathbf{g}(\mathbf{q}_{n+1}, \dot{\mathbf{q}}_{n+1}, \mathbf{x}_{n+1}, t_{n+1}) \\ \dot{\mathbf{x}}_{n+1} &= \mathbf{f}(\mathbf{q}_{n+1}, \dot{\mathbf{q}}_{n+1}, \mathbf{x}_{n+1}, t_{n+1}) \end{cases}$$

Order conditions: 
$$\left\{ \begin{array}{lcl} \gamma & = & 0.5 + \alpha_f - \alpha_m \\ \theta & = & 0.5 + \delta_f - \delta_m \end{array} \right.$$

### Summary

The generalized- $\alpha$  method combines

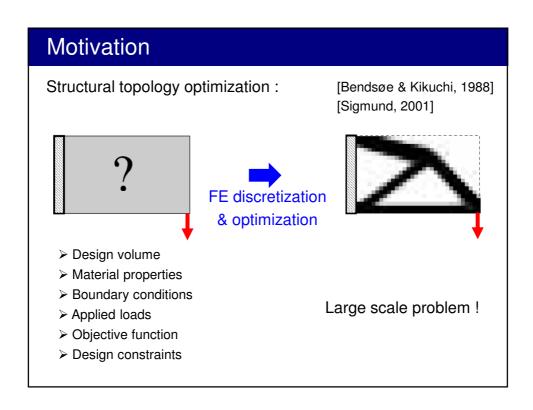
- Second-order accuracy (demonstrated for ODEs)
- > Adjustable numerical damping
- Computational efficiency for large and stiff problems

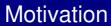
Extension to coupled DAEs on Lie groups with a consistent treatment of:

- > Kinematic constraints
- > Rotational variables
- > Control state variables

### Outline

- ☐ Introduction
- ☐ Modelling of multibody & mechatronic systems
- ☐ Time integration algorithms
- ☐ Topology optimization of structural components
  - ➤ Motivation
  - > Method
  - > Two-dofs robot arm



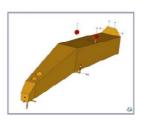


Achievements in structural topology optimization

- ➤ Gradient-based algorithms (CONLIN, MMA, GCMMA...)
- > Relevant problem formulations (SIMP penalization...)

### A powerful design tool:





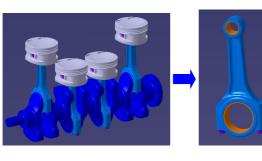


[Poncelet et al., 2005]

### Motivation

Our objective: Topology optimization for the design of components of multibody systems

Equivalent static load approach, see e.g. [Kang & Park, 2005]



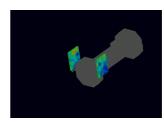
Equivalent static problem

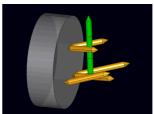
- ➤ boundary conditions?
- ➤ load case(s) ?
- ➤ objective function?
- ⇒ experience and intuition are required
- ⇒ optimal solution for a wrong problem!

### Motivation

Topology optimization based on the actual dynamic response

- > Flexible multibody model (FE)
- $\triangleright$  Time integrator (g- $\alpha$ )
- > Sensitivity analysis
- > Coupling with an optimizer

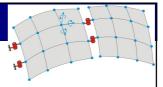




### Advantages:

- ⇒ Systematic approach
- $\Rightarrow$  More realistic objective function

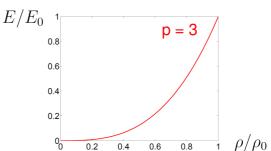
### Topology optimization



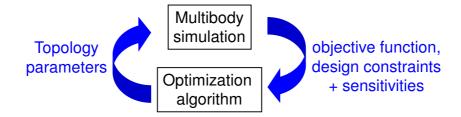
Parameterization of the topology: for each element,

- $\triangleright$  one density variable is defined  $x = \rho/\rho_0, x \in [0, 1]$
- ➤ the Young modulus is computed according to the SIMP law

 $E = x^p E_0$ 



### Global optimization framework



Coupled industrial software

- > OOFELIE (simulation and sensitivity analysis)
- ➤ CONLIN (gradient-based optimization) [Fleury 1989]

Efficient and reliable sensitivity analysis?

Direct differentiation technique

### Sensitivity analysis

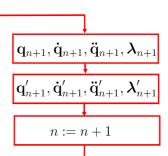
For one design variable x, direct differentiation leads to

$$\mathbf{M}\ddot{\mathbf{q}}' + \mathbf{C}_t\dot{\mathbf{q}}' + \mathbf{K}_t\mathbf{q}' + \mathbf{\Phi}_{\mathbf{q}}^T \boldsymbol{\lambda}' + \mathbf{r}_{,x} = \mathbf{0} \ \mathbf{\Phi}_{\mathbf{q}}\mathbf{q}' + \mathbf{\Phi}_{,x} = \mathbf{0}$$

 $\begin{aligned} \mathbf{M}\ddot{\mathbf{q}}' + \mathbf{C}_t \dot{\mathbf{q}}' + \mathbf{K}_t \mathbf{q}' + \mathbf{\Phi}_{\mathbf{q}}^T \boldsymbol{\lambda}' + \mathbf{r}_{,x} &= 0 \\ \Phi_{\mathbf{q}} \mathbf{q}' + \Phi_{,x} &= 0 \\ \end{aligned} \quad \text{pseudo-loads} \\ \text{Inertia forces} &\propto \boldsymbol{\rho} \\ \text{Elastic forces} &\propto E \end{aligned} \Rightarrow \text{Analytical expressions for} \quad \mathbf{r}_{,x} \end{aligned}$ 

Integration of the sensitivities

- > iteration matrix already computed and factorized
- > one linear pseudo-load case for each design variable



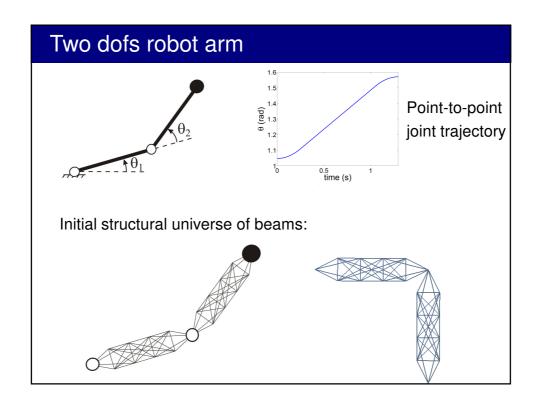
### Sensitivity analysis

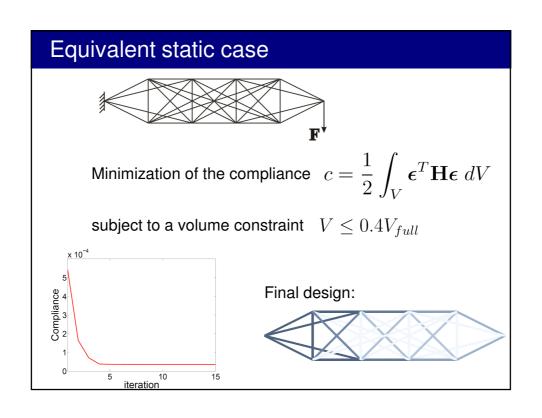
### Importance of an efficient sensitivity analysis:

- > Test problem with (only) 60 design variables
- Finite difference (61 simulations)
  - $\Rightarrow$  CPU time = 141 s
- Direct differentiation (1 extended simulation)
  - $\Rightarrow$  CPU time = 16 s

Moreover, the direct differentiation method leads to higher levels of accuracy

[B. & Eberhard 2008]

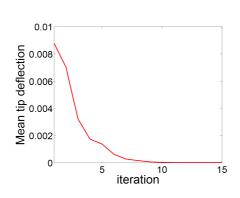




### Optimization based on multibody simulations

Minimization of the tip deflection  $\ \, \frac{1}{t_f} \int_0^{t_f} \| {f r} - {f r}_{rigid} \|^2 \, dt$ 

subject to a volume constraint  $V_{(i)} \leq 0.4\,V_{full,(i)}$ 



### Final design:

### Summary

- ☐ Topology optimization of mechanisms components
- ☐ Equivalent static load ⇒ multibody dynamics approach
  - ➤ flexible multibody simulation
  - > sensitivity analysis
  - > coupling with an optimization code
- ☐ Application to a two dofs robot arm with truss linkages
  - > importance of problem formulation
- Perspectives
  - Complex 3D mechanisms
  - ➤ Mechatronic systems

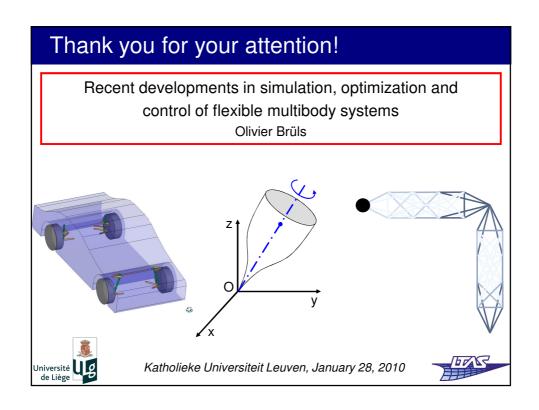
### Future work

### Optimal control methods?

- inverse dynamics problems: identification of applied loads from kinematic data
- control-structure design optimization: the trajectory can be tuned for a given structural design the structural design can be tuned for a given trajectory

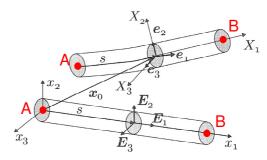
### How to exploit our background for OC? [Bottasso et al, 2004]

- ➤ Indirect OC approach: an existing simulation framework cannot be reused easily
- ➤ Direct OC approach: reuse of an existing simulation framework is more feasible, but one needs to solve large NLP problems



### Modelling of flexible multibody systems

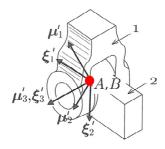
Flexible beam element



- > Timoshenko-type geometrically exact model
- > Two nodes A and B
- ightharpoonup Nodal translations  $(\mathbf{x}_A,\mathbf{x}_B)$  and rotations  $(\mathbf{R}_A,\mathbf{R}_B)$
- > Strain energy: bending, torsion, traction and shear
- > Kinetic energy: translation and rotation

### Modelling of flexible multibody systems

Hinge element



- > Two nodes A (on body 1) and B (on body 2)
- ightharpoonup Nodal translations  $(\mathbf{x}_A,\mathbf{x}_B)$  and rotations  $(\mathbf{R}_A,\mathbf{R}_B)$
- > 5 kinematic constraints

$$\mathbf{x}_A - \mathbf{x}_B = \mathbf{0}$$

$$\boldsymbol{\mu}_1'(\mathbf{R}_A) \cdot \boldsymbol{\xi}_3'(\mathbf{R}_B) = 0$$

$$\boldsymbol{\mu}_2'(\mathbf{R}_A) \cdot \boldsymbol{\xi}_3'(\mathbf{R}_B) = 0$$

Multibody dynamics approach 
$$\frac{1}{t_f} \int_0^{t_f} \sum_{i=1}^{n_c} c_{(i)} \, dt$$
 subject to a volume constraint 
$$V_{(i)} \leq 0.4 \, V_{full,(i)}$$
 Final design:

