

Recent developments in optimization of flexible components of multibody systems

P. Duysinx
LTAS - Automotive Engineering
Aerospace and Mechanics Department
University of Liège

OUTLINE

- Introduction
- Finite Element approach of Multibody system dynamics
- Sensitivity Analysis
- Formulation of optimization problem
- Numerical application: 2-dof robot arm
- Conclusion & Perspectives

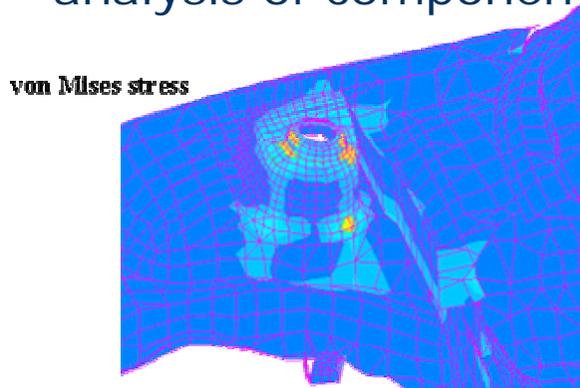


INTRODUCTION

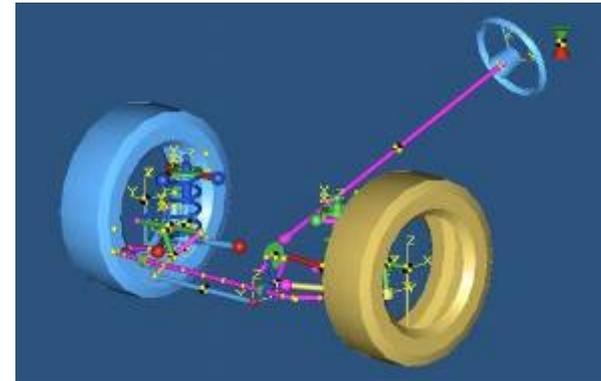


EVOLUTION OF VIRTUAL PROTOTYPING

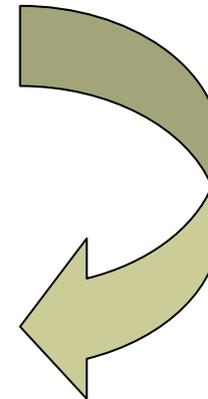
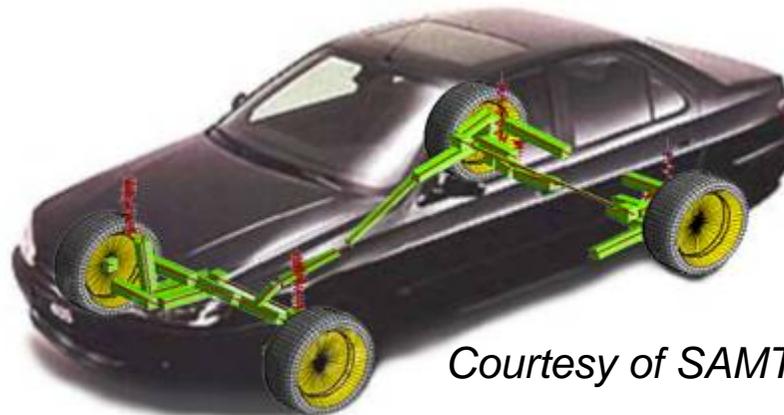
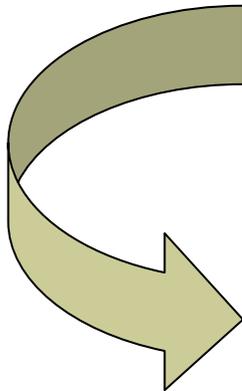
- Finite Element: structural analysis of components



- Multibody system: mechanism of rigid bodies



- Flexible Multibody systems:
System approach (MBS)
& structural dynamics (FEM)

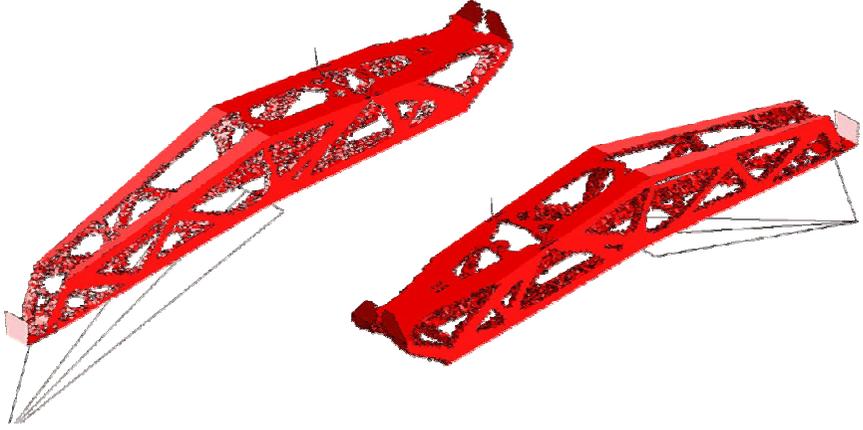


Courtesy of SAMTECH

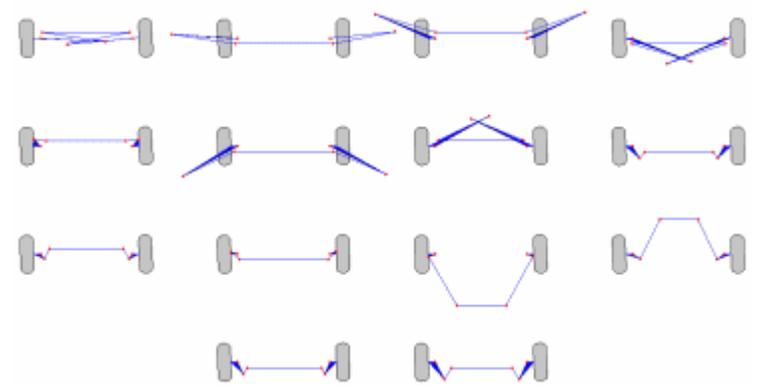


EVOLUTION OF OPTIMIZATION

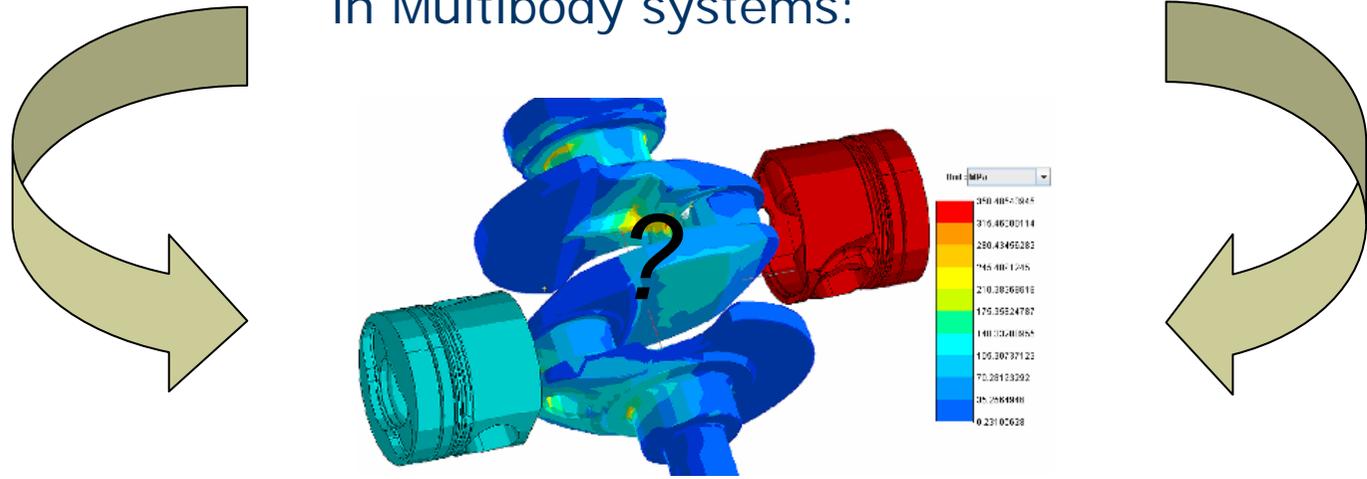
■ Structural optimization



■ Mechanism synthesis

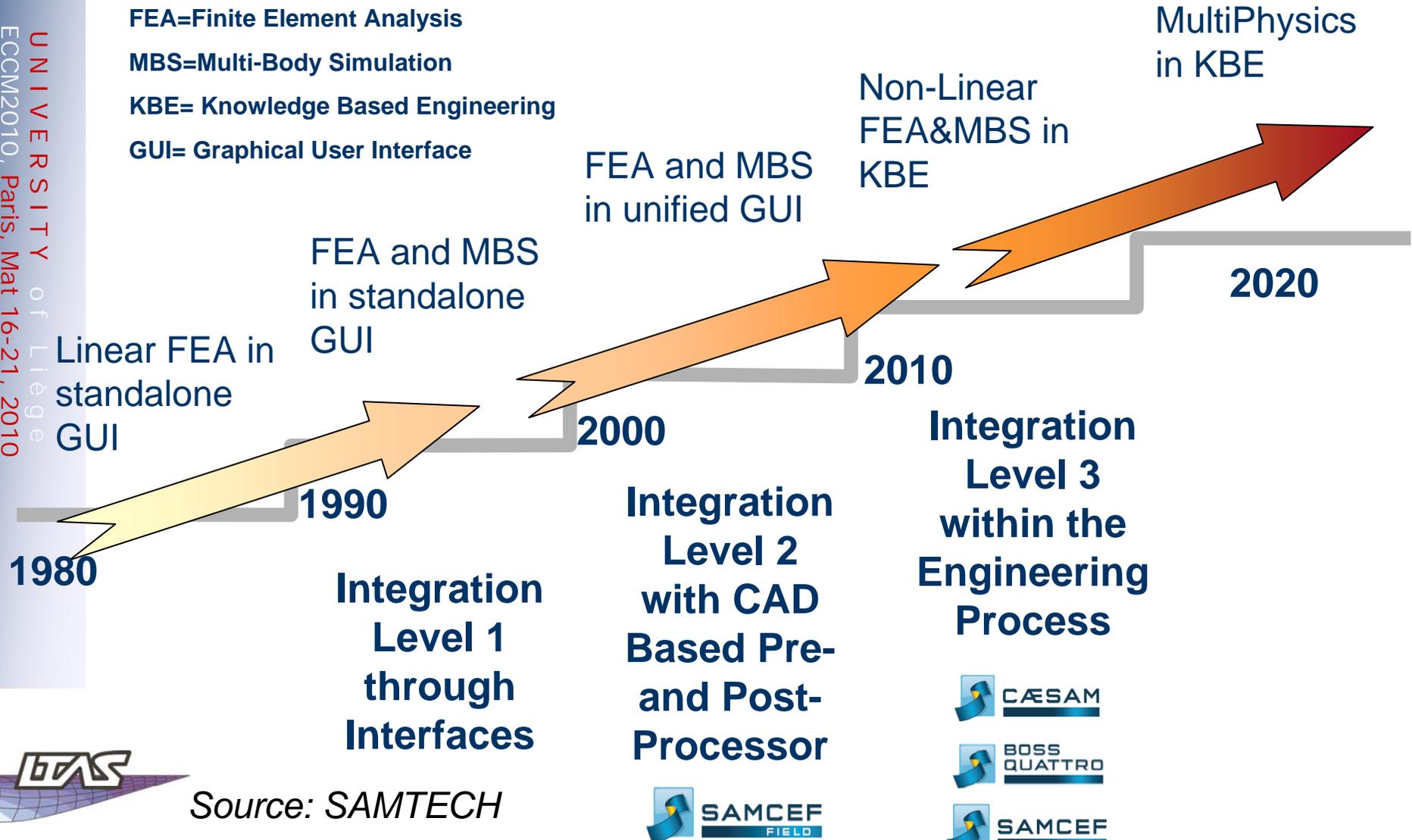


■ Optimization of flexible components in Multibody systems:



AN INDUSTRIAL PERSPECTIVE...

FEA=Finite Element Analysis
 MBS=Multi-Body Simulation
 KBE= Knowledge Based Engineering
 GUI= Graphical User Interface



Source: SAMTECH

INTRODUCTION

- Classical FEM approach: static load cases (empirical)
- Weak coupling between FEM and MBS:
 - Coupling with pre / post processing
 - Define equivalent quasi-static load cases
 - Optimization of isolated components
- Optimization of flexible components in multibody system dynamics
 - Define realistic dynamic loading
 - Take care of the coupling between large overall rigid-body motions and deformations
 - FEM based approach of MBS
 - Sensitivity of MBS
 - Coupling with optimization



FINITE ELEMENT APPROACH OF MULTIBODY SYSTEM DYNAMICS



EQUATION OF MBS DYNAMICS

- Motion of the flexible body is represented by absolute nodal coordinates \mathbf{q}

- Dynamic equation of multibody system

$$\mathbf{M}\ddot{\mathbf{q}} = \mathbf{g}(\dot{\mathbf{q}}, \mathbf{q}, t) = \mathbf{g}^{\text{ext}} - \mathbf{g}^{\text{int}}$$

- \mathbf{M} masse matrix of the system
 - \mathbf{g} external and internal forces
- The motion of the mechanical system is subject to some kinematic constraints (joints, admissible or prescribed motion)

$$\Phi(\mathbf{q}, t) = 0$$



EQUATION OF MBS DYNAMICS

- Solution based on an augmented Lagrangian approach of total energy and two additional terms related to the constraints including a penalty and the Lagrangian multipliers λ .
- Stationary conditions

$$\begin{cases} \mathbf{M}\ddot{\mathbf{q}} + \mathbf{B}^T (k\lambda + p\Phi) = \mathbf{g}(\dot{\mathbf{q}}, \mathbf{q}, t) \\ k\Phi(\mathbf{q}, t) = 0 \end{cases}$$

If

$$\mathbf{B} = \frac{\partial \Phi}{\partial \mathbf{q}}$$

- With initial conditions

$$\mathbf{q}'(0) = \mathbf{q}'_0 \text{ and } \dot{\mathbf{q}}'(0) = \dot{\mathbf{q}}_0$$



TIME INTEGRATION

- The set of nonlinear differential and algebraic equations can be solved using the generalized- α method by Chung and Hulbert (1993)

- Define pseudo acceleration \mathbf{a} :

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

- Newmark integration formulae

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

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

- Solve the dynamic equation system (Newton-Raphson)

$$\begin{cases} \mathbf{M}\Delta\ddot{\mathbf{q}} + \mathbf{C}_t\Delta\dot{\mathbf{q}} + \mathbf{K}_t\Delta\mathbf{q} + \mathbf{B}^T\Delta\boldsymbol{\lambda} = \Delta\mathbf{r} & \mathbf{r} = \mathbf{M}\ddot{\mathbf{q}} - \mathbf{g} + \mathbf{B}^T\boldsymbol{\lambda} \\ \mathbf{B} = \mathbf{0} \end{cases}$$

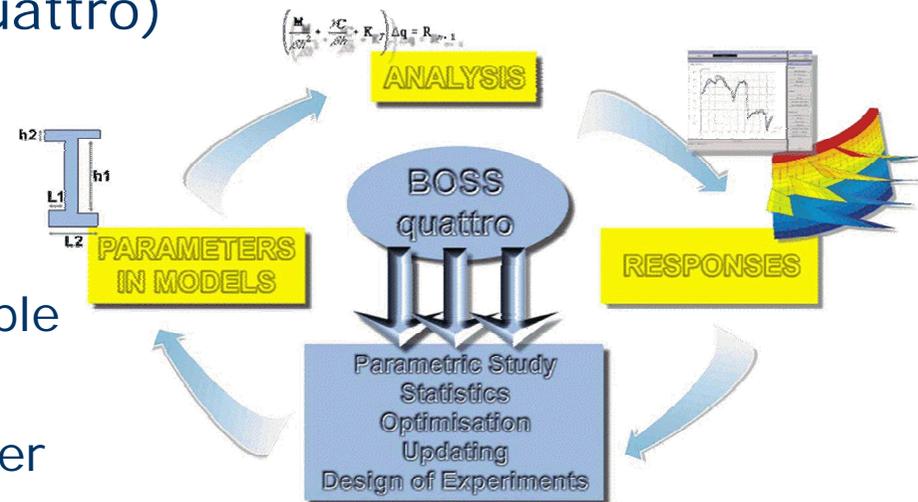


SENSITIVITY ANALYSIS

- Gradient-based optimization methods requires the first order derivatives of the responses
- Finite differences (Boss Quattro)

$$\frac{\partial f}{\partial x} \approx \frac{f(x + \delta x) - f(x)}{\delta x}$$

Perturbation of design variable
Additional call to MBS code
→ Boss Quattro task manager



- Semi-analytical approach

$$\frac{\partial \mathbf{r}}{\partial x} \approx \frac{\mathbf{r}(x + \delta x) - \mathbf{r}(x)}{\delta x}$$

$$\frac{\partial \Phi}{\partial x} \approx \frac{\Phi(x + \delta x) - \Phi(x)}{\delta x}$$



- Derivative of response function f

$$f' = f'_{,z} \mathbf{z}'(\mathbf{x}, t) + f'_{,x}$$

- → Evaluate the derivatives of the state variables

- Derivation of the state equations

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

Pseudo loads

- With the initial conditions

$$\mathbf{q}'(0) = \mathbf{q}'_0 \text{ and } \dot{\mathbf{q}}'(0) = \dot{\mathbf{q}}_0$$



SENSITIVITY ANALYSIS Bruls and Eberhard (2008)

- One observes that the sensitivity equations are linear with respect to derivatives of q and λ .
- At time step $n+1$, the sensitivities can be computed using the same integration algorithm as for the dynamic response except for the residuals
- The iteration (tangent) matrix is the same as for the original problem. Hence this matrix should be computed and factorized only once for the sensitivity analysis at time step $n+1$.
- See Bruls and Eberhard (2008) for details



FORMULATION AND SOLUTION OF OPTIMIZATION PROBLEM



GENERAL FORM OF THE OPTIMIZATION PROBLEM

- Design problem is cast into a mathematical programming problem

$$\begin{aligned} \min_{\mathbf{x}} \quad & g_0(\mathbf{x}) \\ \text{s.t.} \quad & g_j(\mathbf{x}) \leq \bar{g}_j \quad j = 1 \dots m \\ & \underline{x}_i \leq x_i \leq \bar{x}_i \quad i = 1 \dots n \end{aligned}$$

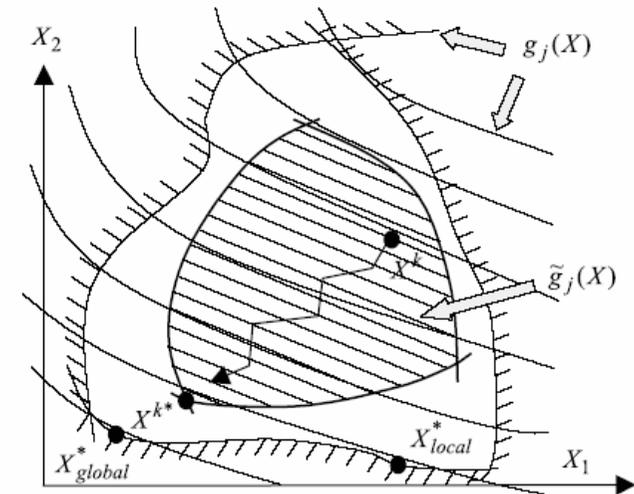
- Provides a general and robust framework to the solution procedure
- Take benefit of the available efficient solvers :
 - CONLIN (Fleury, 1989); MMA (Svanberg, 1987)
 - GA, PSO, etc.



SOLUTION ALGORITHMS

■ GRADIENT-BASED ALGORITHMS (SCP)

- Structural approximation + Efficient solvers
- CONLIN (Fleury, 1989), MMA (Svanberg, 1987), GCM (Bruyneel, Duysinx, and Fleury, 2002), SQP



■ METAHEURISTIC ALGORITHMS: GA

- Require only the computation of the function values
- Global convergence
- Large number of function evaluations

■ SURROGATE BASED OPTIMIZATION

- Replace the direct evaluation of the simulation model by a global approximation of the responses (Artificial Neural Network, etc.)



Design variables

- Optimization of flexible components
 - ➔ Sizing variables: Plate thickness, Bar and beams cross sections, lumped properties (stiffness, mass, etc.)
 - ➔ Shape variables: geometrical parameters of flexible body shape: e.g. control node positions
 - ➔ Topology: Pseudo density variables, e.g. SIMP $E = \mu^3 E^0$
- **BUT NOT**
 - Synthesis variables of mechanisms (Hansen, 2002)
 - Links and joints connectivity (Kawamoto et al. 2004)



Objective function and restrictions

- Mass of the mechanism
 - Source of inertia loads
 - Cost of material
- Stiffness → Compliance
 - Compliance of component i at time step t

$$C_{(i)}(t) = \int_{V_E} \boldsymbol{\varepsilon}^T \mathbf{D} \boldsymbol{\varepsilon} dV$$

- Averaged compliance along the motion (Bruls et al. 2007)

$$\bar{C} = \frac{1}{T} \int_0^T \sum_i C_{(i)}(\tau) d\tau$$



Objective function and restrictions

- For robot motion: stiffness \rightarrow trajectory tracking
 - Perfect trajectory = rigid body trajectory $\mathbf{r}_{rigid}(t)$
 - Flexible trajectory $\mathbf{r}(t)$
 - Average deviation of the flexible trajectory: mean tip deflection (Bruls et al. 2007)

$$\bar{d} = \frac{1}{T} \int_0^T \left\| \mathbf{r}(\tau) - \mathbf{r}_{rigid}(\tau) \right\|^2 d\tau$$

- Norm 2 of deviation
- Loose control of instantaneous deviation?
- Relation between mean square deviation and max deviation?



Objective function and restrictions

- For robot motion: stiffness \rightarrow trajectory tracking

- Instantaneous deviation of the flexible trajectory

$$-|d_{\min}| \leq \text{dist}(\mathbf{r}_{\text{rigid}}(t) - \mathbf{r}(t)) \leq d_{\max}$$

- Two constraints per time step \rightarrow large scale optimization problem

- Global maximum/minimum deviation constraint

$$[\max_k \text{dist}(\mathbf{r}_{\text{rigid}}(t_k) - \mathbf{r}(t_k))] \leq d_{\max}$$

- One single constraint
- Non smooth and strongly non linear

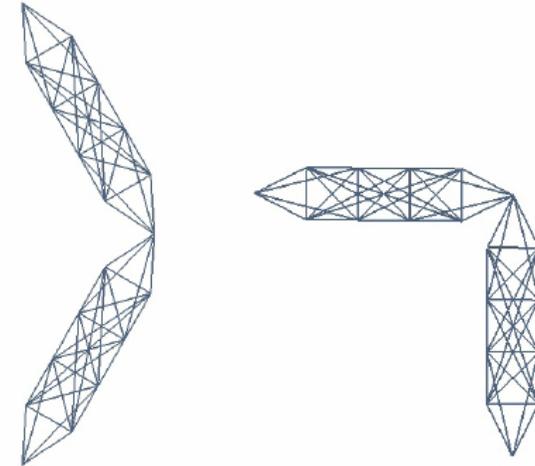
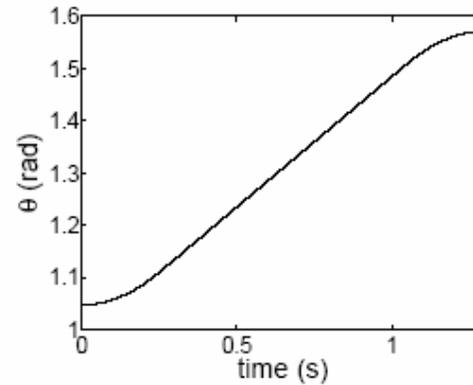
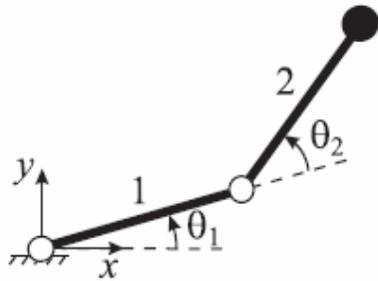


NUMERICAL APPLICATIONS



TOPOLOGY OPTIMIZATION OF ROBOT COMPONENTS *(Bruls et al. 2007, 2010)*

■ Two-dof robot-arm



■ Topology optimization of truss-like components

- SIMP model $E = \mu^3 E^0$
- CONLIN
- Semi-analytical sensitivity

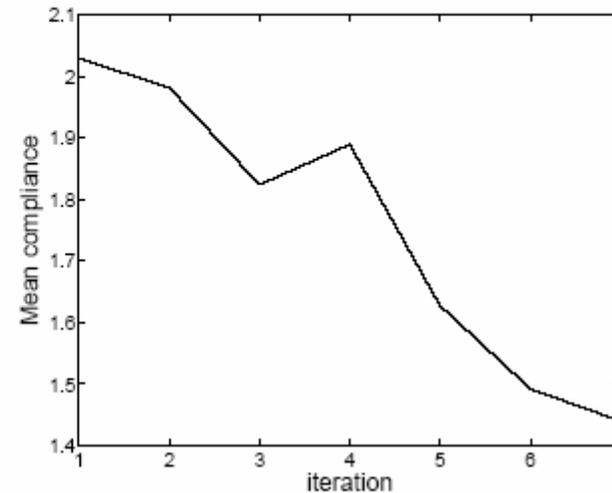
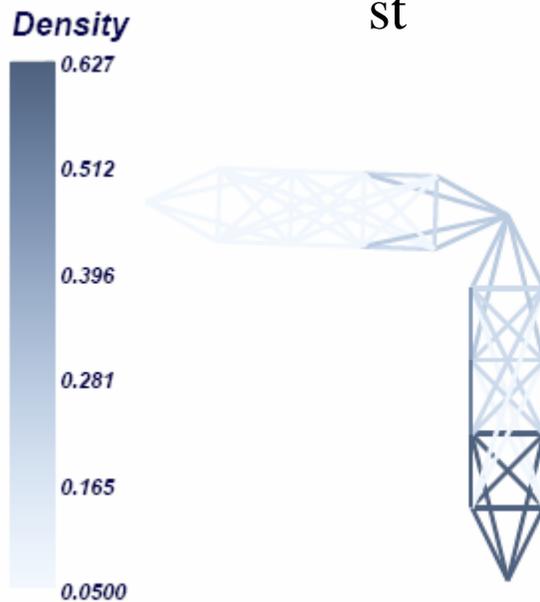


TOPOLOGY OPTIMIZATION OF ROBOT COMPONENTS *(Bruls et al. 2007, 2010)*

■ Minimum compliance

$$\min_x \bar{C} = \frac{1}{T} \int_0^T \sum_i C_{(i)}(\tau) d\tau$$

$$\text{st} \quad V \leq 0.4 \bar{V}$$



Non convergence at iteration 7

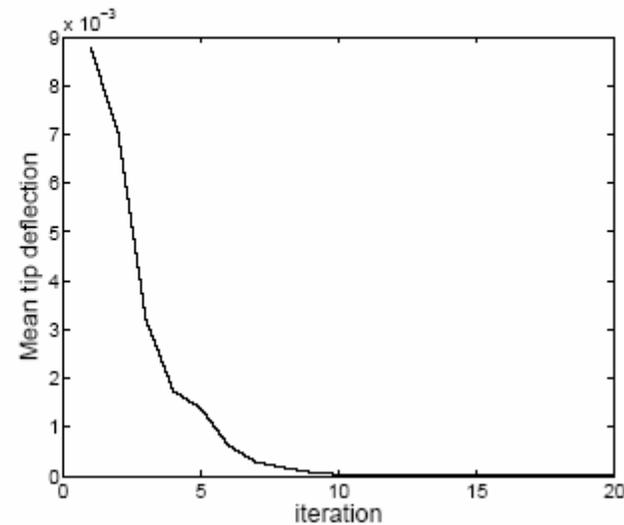
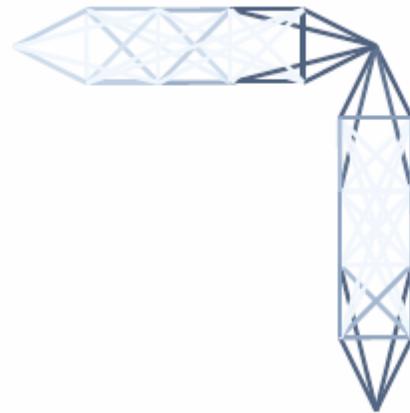
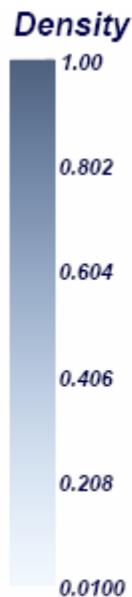


TOPOLOGY OPTIMIZATION OF ROBOT COMPONENTS *(Bruls et al. 2007, 2010)*

- Minimum average deviation

$$\min_x \quad \bar{d} = \frac{1}{T} \int_0^T \|\mathbf{r}(\tau) - \mathbf{r}_{rigid}(\tau)\|^2 d\tau$$

$$st \quad V \leq 0.4\bar{V}$$

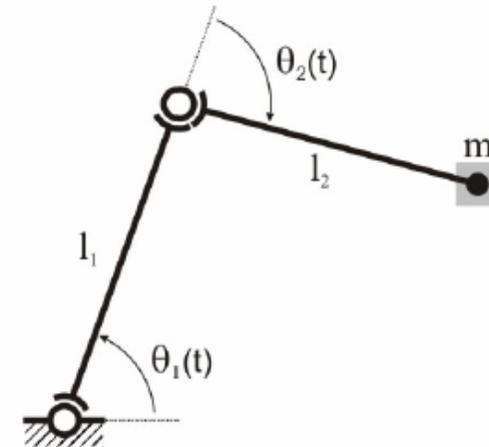


Convergence OK!

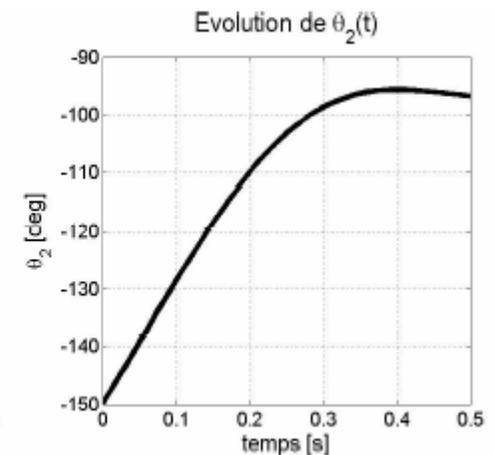
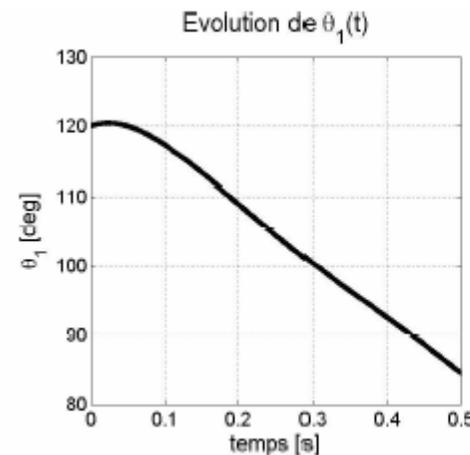
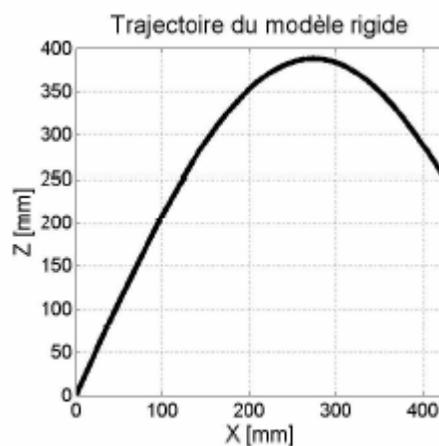


COMPONENT OPTIMIZATION OF A 2-DOF ROBOT ARM

- Two-dof robot arm (Ata, 2007)
trajectory time: 0,5s
 $l_1=l_2=600$ mm
 $M= 1$ kg
Aluminum: 72 GPa, $\nu=0.3$
 $\rho=2700\text{kg/m}^3$

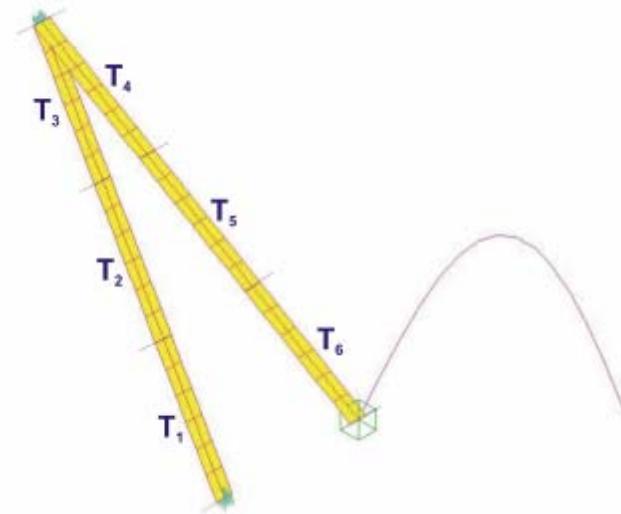


- Trajectory followed by rigid robot \rightarrow prescribed joint law



SIZING OPTIMIZATION OF ROBOT COMPONENTS

- Modeling of the flexible components as plates of variable thicknesses: $T_1 \dots T_6$.
- Prescribe the joint motion from rigid body simulation



- Optimization problem

$$\begin{aligned} \min \quad & \text{mass} \\ \text{s.t.} \quad & \text{deviation} \leq 10 \text{ mm} \\ & 5 \leq T_i \leq 200 \text{ mm} \end{aligned}$$

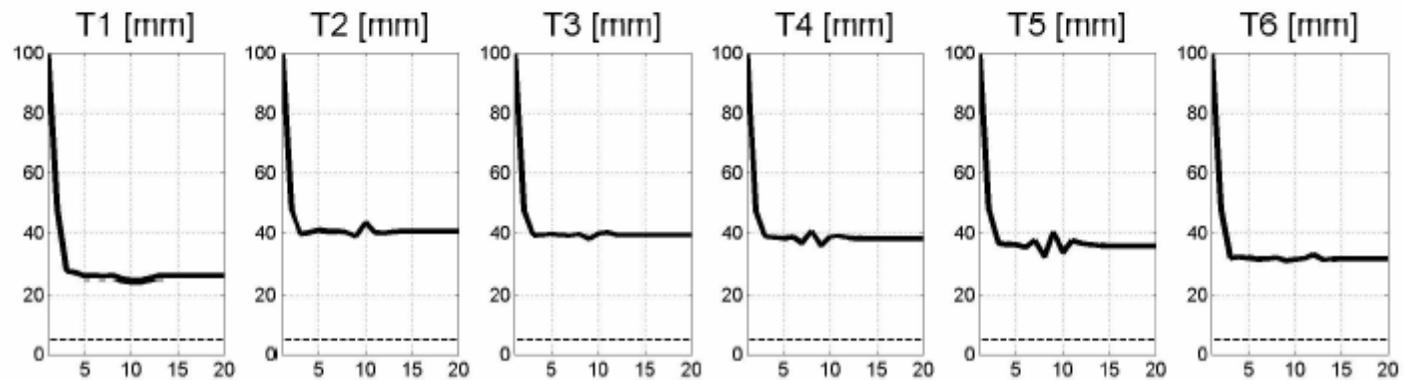
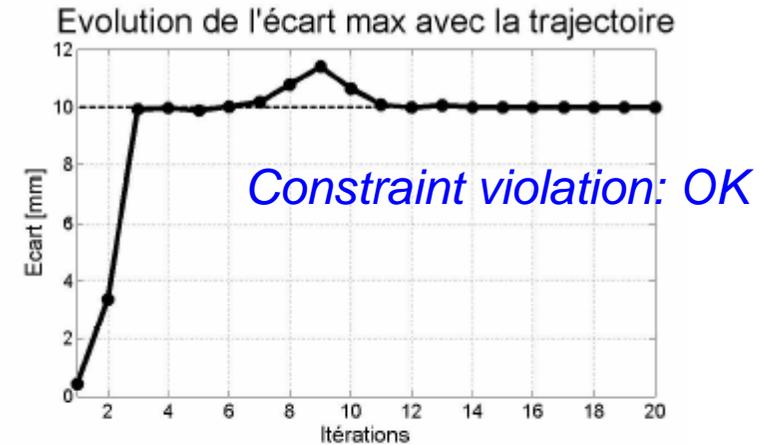
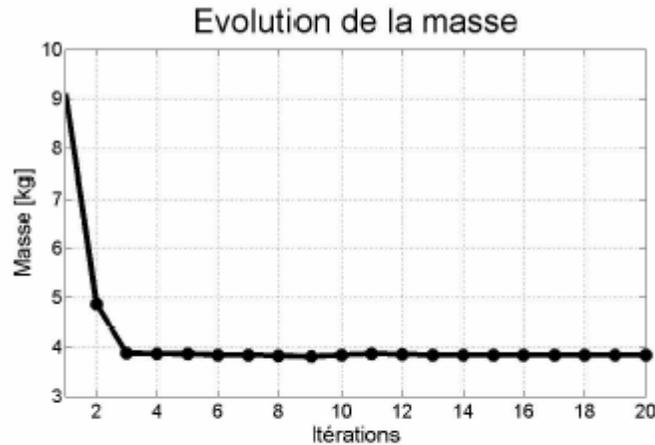
- Deviation each time step

$$-|d_{\min}| \leq \text{dist}(\mathbf{r}_{\text{rigid}}(t) - \mathbf{r}(t)) \leq d_{\max}$$

- Solution GCM (Bruyneel et al. 2002)



SIZING OPTIMIZATION OF ROBOT COMPONENTS

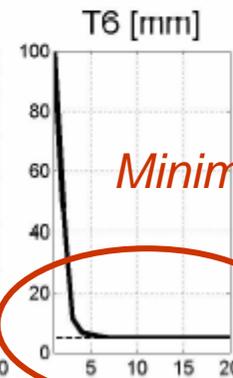
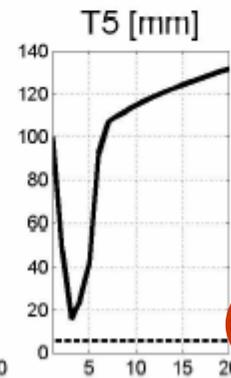
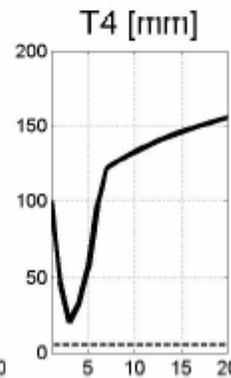
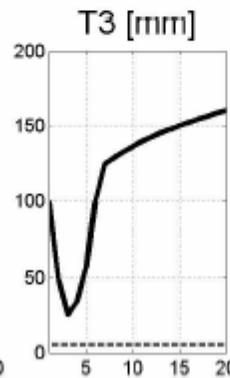
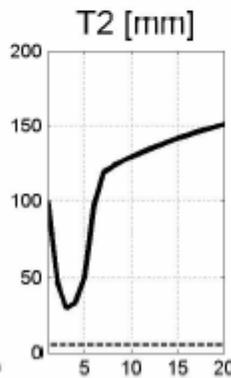
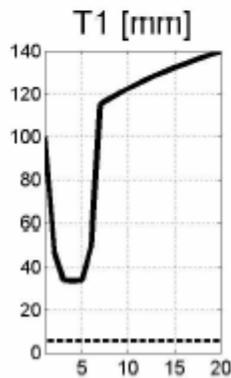
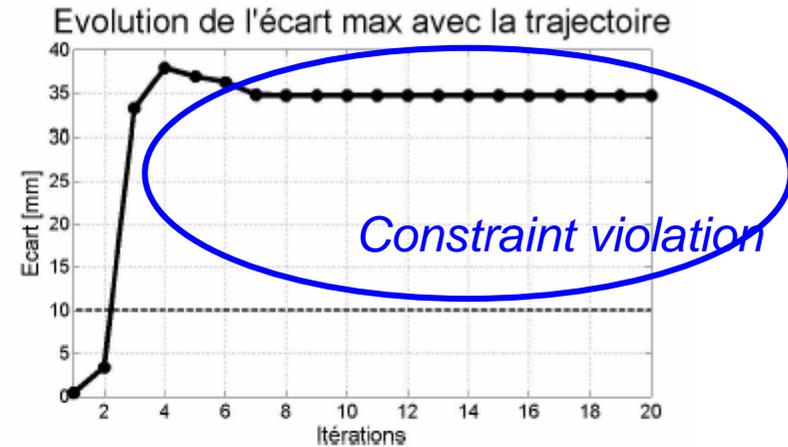
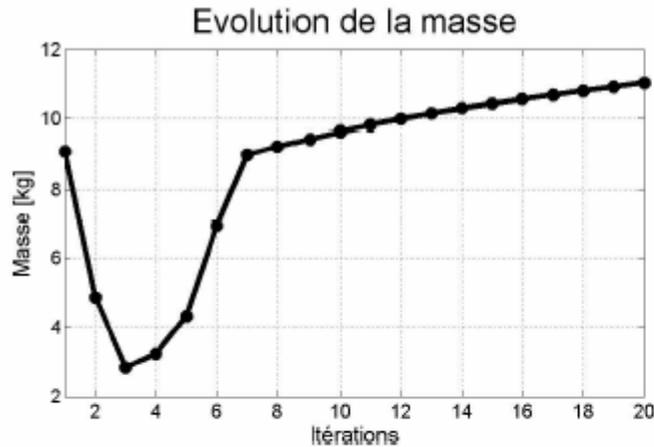


T6 → 31 mm



By chance, one feasible solution using finite difference sensitivity analysis

SIZING OPTIMIZATION OF ROBOT COMPONENTS



Generally, optimization procedure fails
(here semi-analytical sensitivity analysis)

SIZING OPTIMIZATION OF ROBOT COMPONENTS

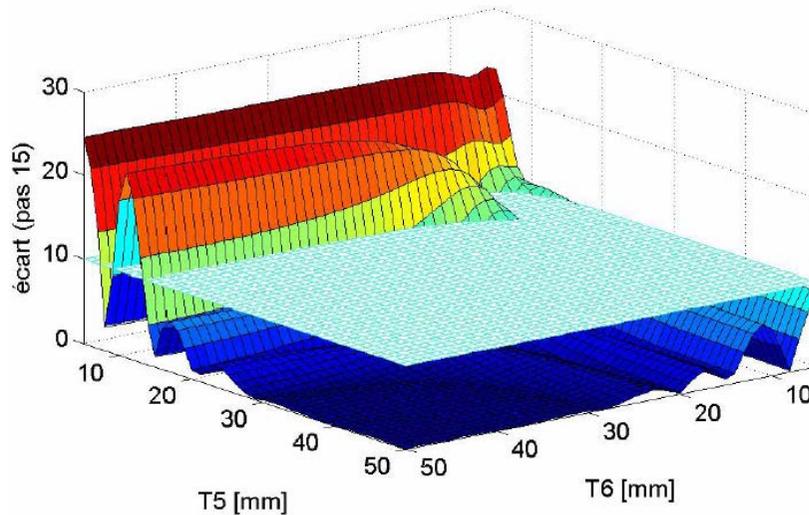
- Open questions
 - Origin of the problem?
 - Suitable formulation?
 - Deviation at each time step
 - Average deviation
 - Global deviation constraint (max)
 - Choice of optimization algorithm?
 - Gradient-based algorithm
 - GA, surrogate (ANN)

- Answer:
 - Plot design space (slices for selected design variables)
 - Iteration points trajectory in the design space



SIZING OPTIMIZATION OF ROBOT COMPONENTS

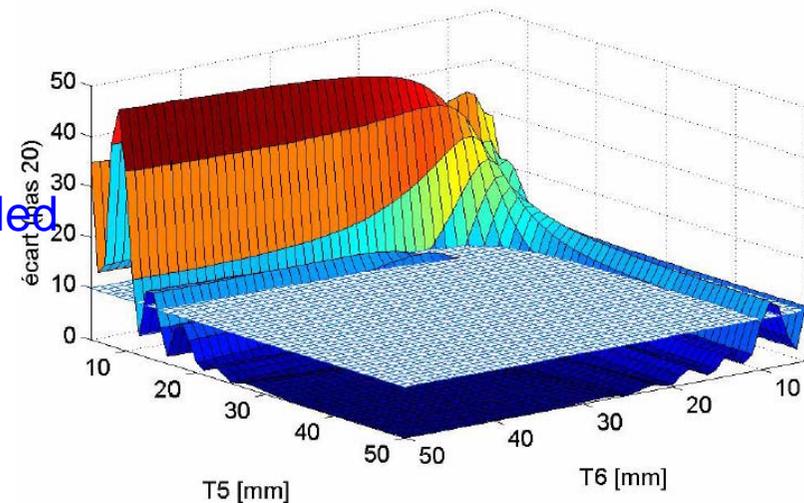
- One deviation constraint per time step
 - Complex feasible design domain



$$-|d_{\min}| \leq \text{dist}(\mathbf{r}_{\text{rigid}}(t) - \mathbf{r}(t)) \leq d_{\max}$$

Is structural approximation procedure adapted to such problems?

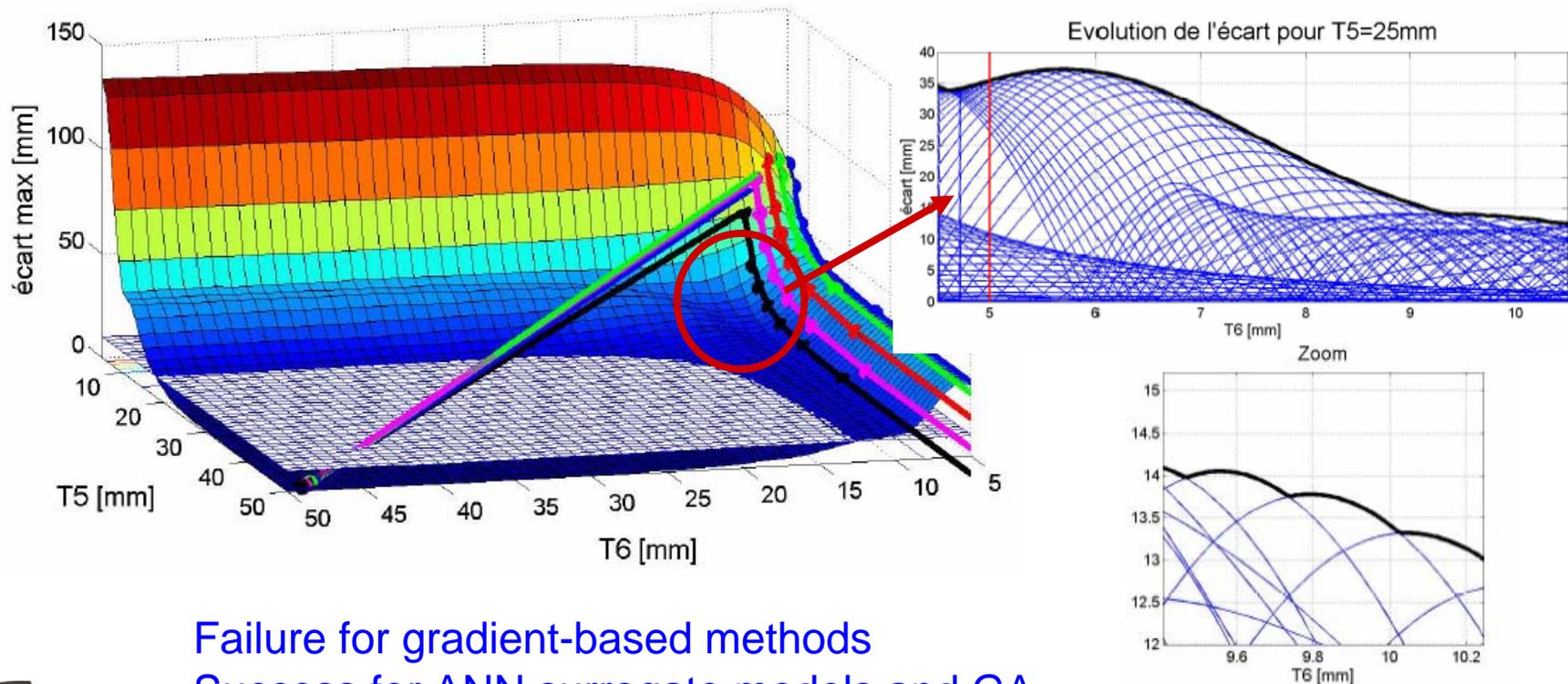
CONLIN, GCM, MMA, SQP failed
Surrogate (ANN) failed
GA: very heavy solution



SIZING OPTIMIZATION OF ROBOT COMPONENTS

- Global maximum deviation constraint
- Non smooth but less complex topology of design space

$$[\max_k \text{dist}(\mathbf{r}_{\text{rigid}}(t_k) - \mathbf{r}(t_k))] \leq d_{\text{max}}$$

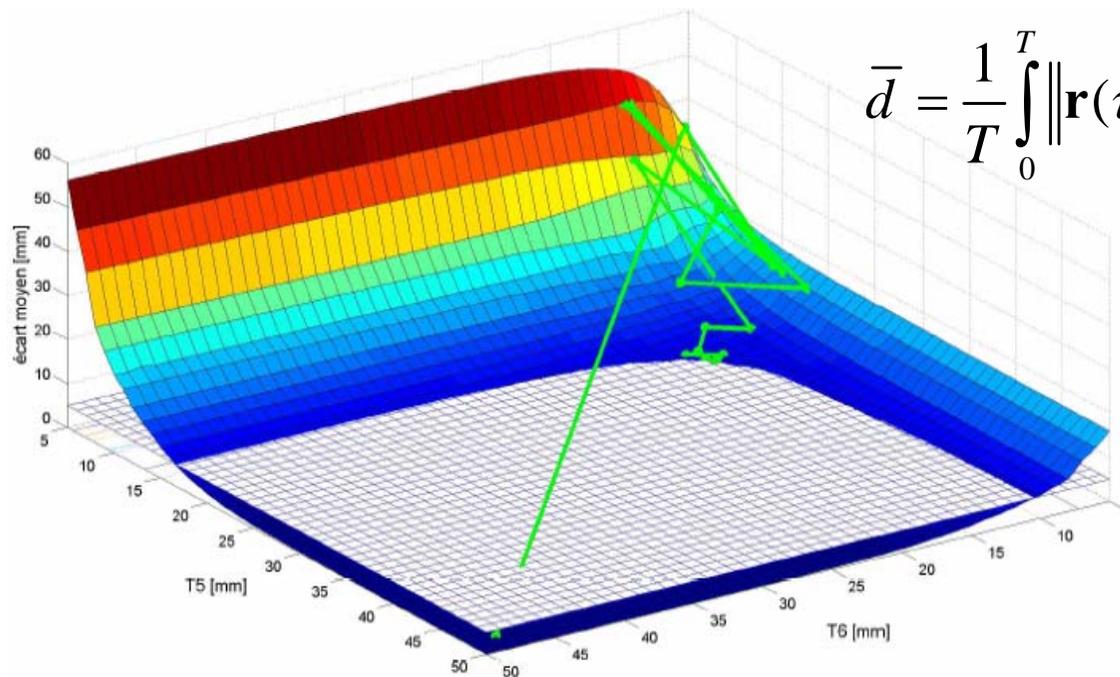


Failure for gradient-based methods
Success for ANN surrogate models and GA



SIZING OPTIMIZATION OF ROBOT COMPONENTS

- Average deviation constraint ($d < 5$ mm)
 - Smooth and less complex topology of design space
 - Loose control of the instantaneous deviation



$$\bar{d} = \frac{1}{T} \int_0^T \|\mathbf{r}(\tau) - \mathbf{r}_{rigid}(\tau)\|^2 d\tau$$

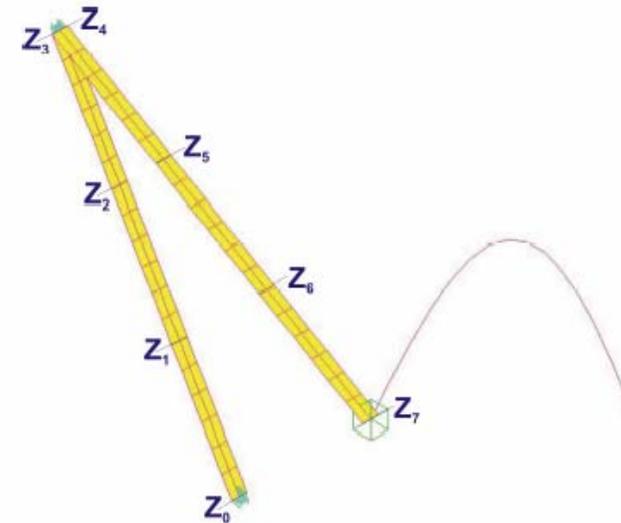
Success for GCM, SQP
Success for ANN surrogate models and GA



SHAPE OPTIMIZATION OF ROBOT COMPONENTS

- Modeling of the flexible components as plates of variable width: $Z_1 \dots Z_8$.
- Prescribe joint motion from rigid body simulation
- Optimization problem

$$\begin{aligned} \min \quad & \text{mass} \\ \text{s.t.} \quad & \text{deviation} \leq 10 \text{ mm} \\ & 5 \leq T_i \leq 200 \text{ mm} \end{aligned}$$



- Semi-analytical sensitivities
- Deviation each time step

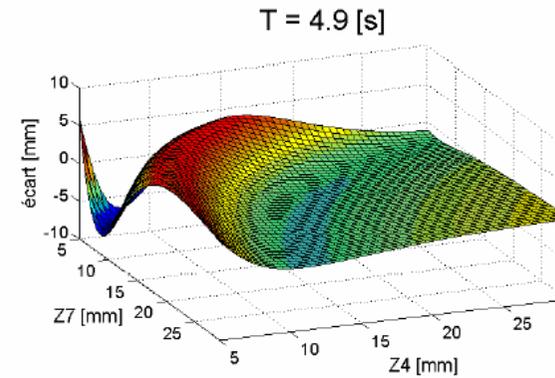
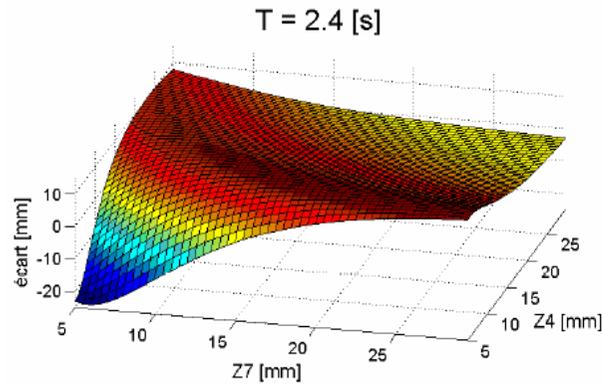
$$-|d_{\min}| \leq \text{dist}(\mathbf{r}_{\text{rigid}}(t) - \mathbf{r}(t)) \leq d_{\max}$$

- Solution GCM (Bruyneel et al. 2002)

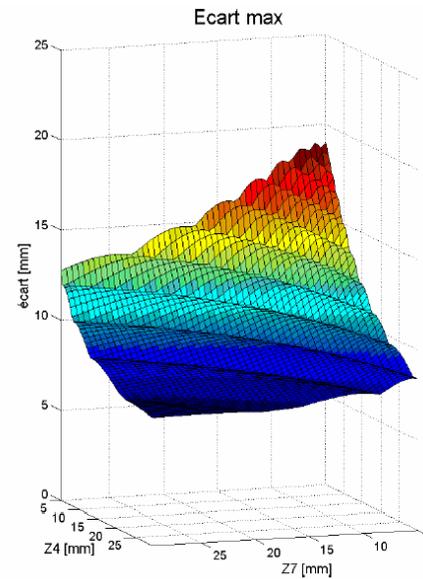
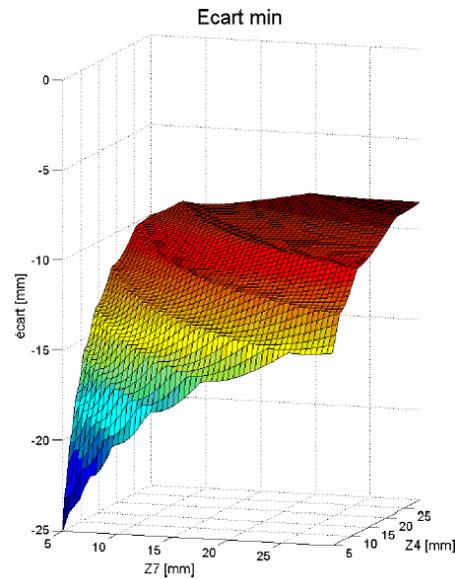


SHAPE OPTIMIZATION OF ROBOT COMPONENTS

- Nature of deviation constraints at $t=15$ and $t=20$

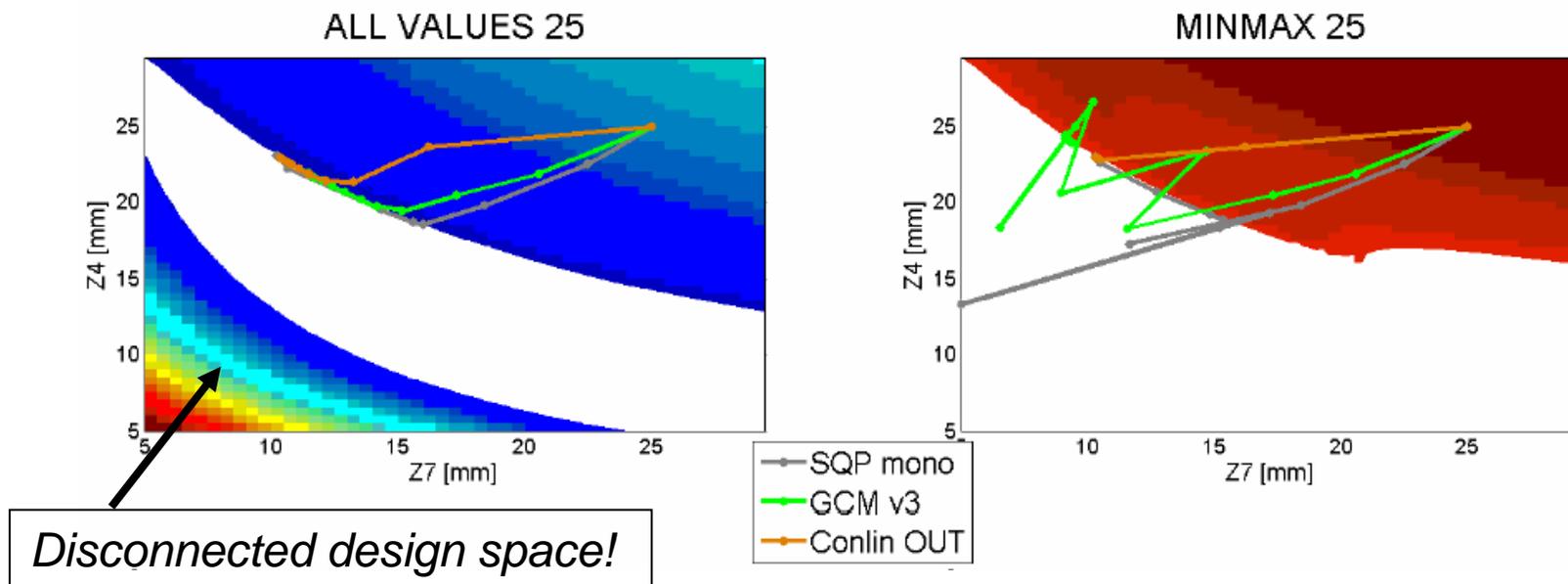


- Global max /min deviation



SHAPE OPTIMIZATION OF ROBOT COMPONENTS

- Comparison between: CONLIN, GCM, SQP
 - Deviation at each time step v.s. max deviation

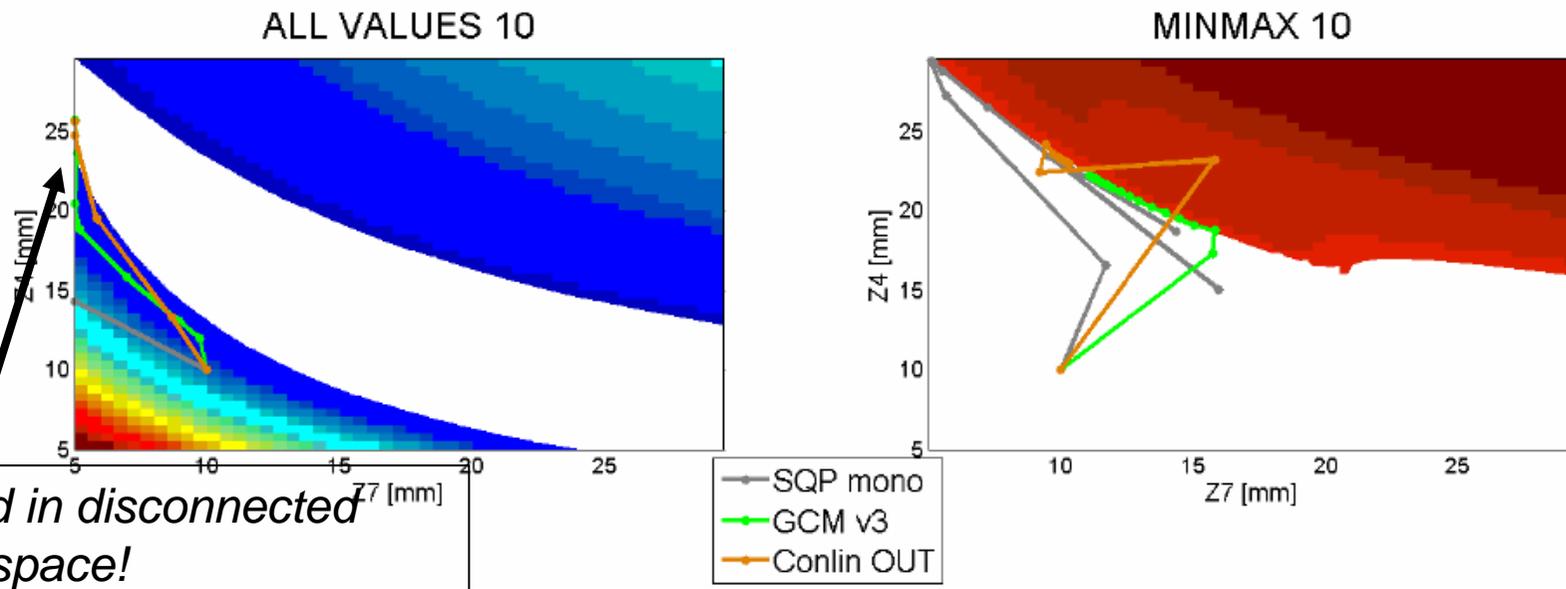


- All gradient-based algorithms converge when starting from a feasible design point



SHAPE OPTIMIZATION OF ROBOT COMPONENTS

- Comparison between: CONLIN, GCM, SQP
 - Deviation at each time step v.s. max deviation

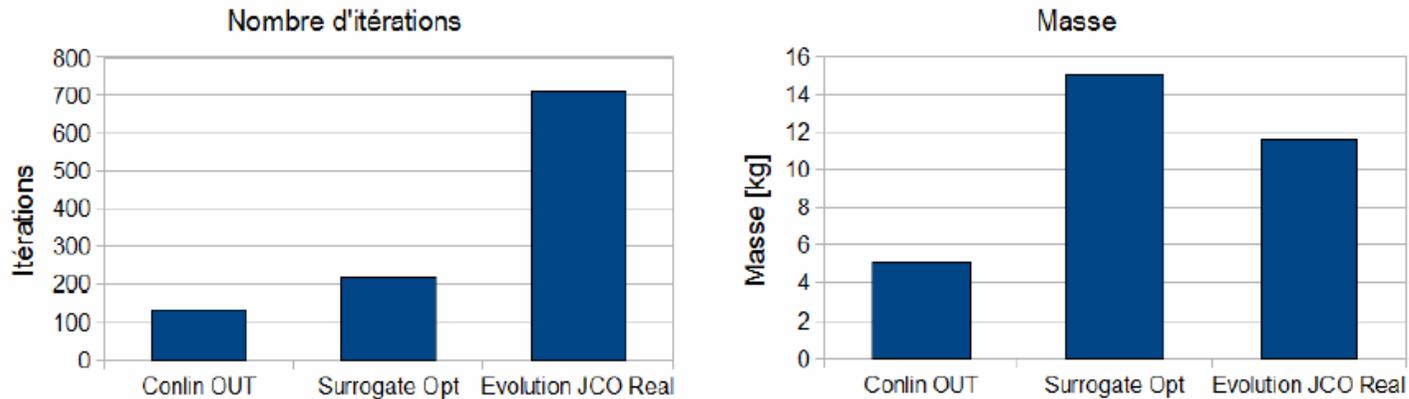


- Gradient-based algorithms + local minima and unfeasible starting point
- Global criteria improves the situation and helps to find global optima

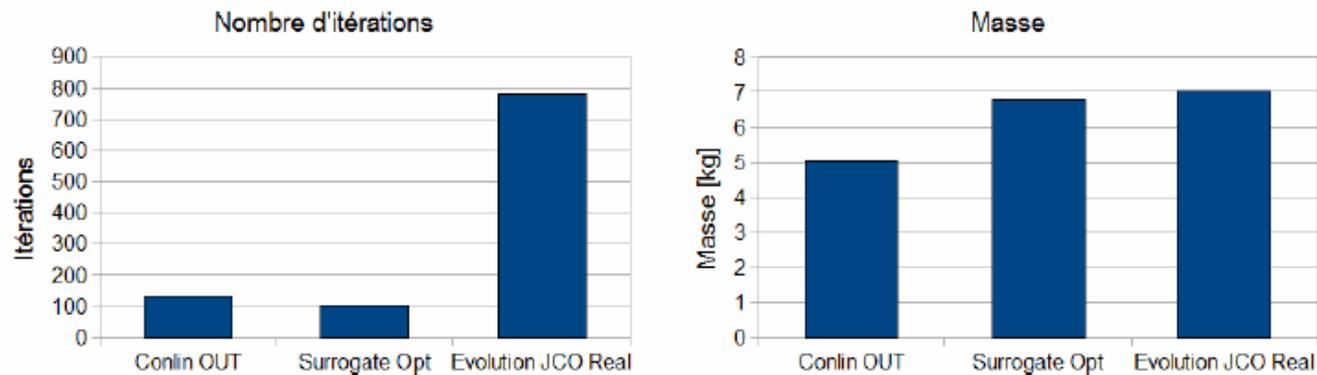


SHAPE OPTIMIZATION OF ROBOT COMPONENTS

- Comparison between: CONLIN, SURROGATE+GA, GA



Constraint at each time step – feasible starting point

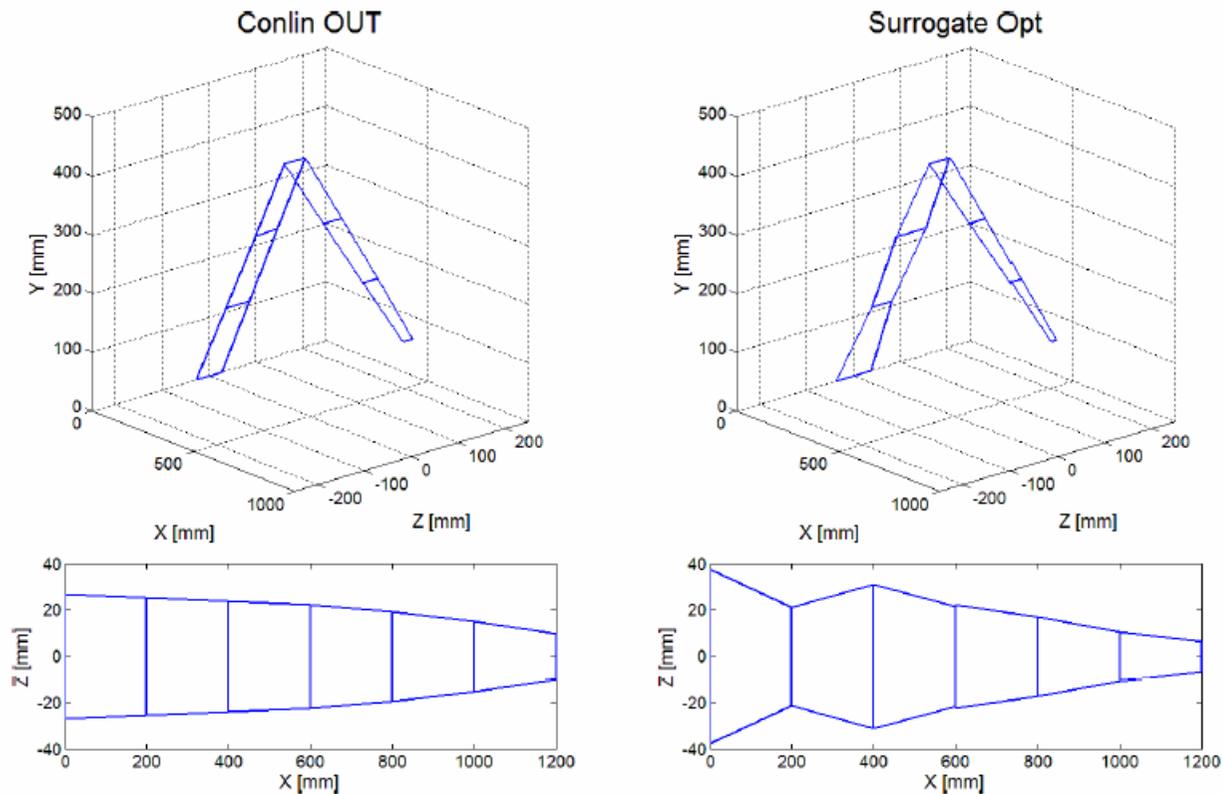


Constraint max deviation – unfeasible starting point



SHAPE OPTIMIZATION OF ROBOT COMPONENTS

- Comparison between: CONLIN, SURROGATE+GA



Constraint at each time step – feasible starting point

CONCLUSIONS



CONCLUSIONS

- Optimization of structural components can be carried out in the framework of flexible multibody simulations
 - Dynamic coupling between large overall rigid-body motions and deformations
 - Single dynamic analysis instead of a patchwork of (empirical) static analyses
 - Design constraints defined with respect to the actual dynamic problem

- Lack of past experiences in
 - Formulation of optimization problem
 - Solution strategies



CONCLUSIONS

- Design problem is complex: naïve implementations do generally not work!
- Trajectory following
 - Time step deviation constraint are very difficult but may work when starting from feasible starting point
 - Global max constraints are non smooth → GA + surrogate
 - Average deviation are the easiest ones but what is the link to local deviations?
- Solvers
 - Gradient-based solvers can work efficiently when starting from feasible design
 - GA give poor results
 - Surrogate models + GA are reliable and robust



PERSPECTIVES

- Mixed formulations including global (average) constraints and time step constraints should be the best compromise
- Revisiting solution algorithms for dynamic problems!
 - Structural approximations: local / global: trust regions?
 - Reliability and robustness when starting from unfeasible design points
- Other design criteria for time domain analysis of dynamic systems:
 - ISE, ISA
 - Rise time
 - Controller and actuator activity...



*THANK YOU VERY MUCH
FOR YOUR ATTENTION*



CONTACT



Pierre DUYSINX

Automotive Engineering
Aerospace and Mechanics Department
of University of Liège

Chemin des Chevreuils, 1 building B52
4000 Liège Belgium

Email: P.Duysinx@ulg.ac.be

Tel +32 4 366 9194

Fax +32 4 366 9159

url: www.ingveh.ac.be

