

# **Multibody System Dynamics: MBDyn Overview**



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#### **Outline**



- Multibody dynamics
- Software architectures
- Problems
- Arbitrary motion description
- Deformable components
- Solving the problem
- Extracting useful information
- Examples of multibody modeling with MBDyn
- Future development
- Documentation and support

## **Multibody dynamics**



#### **Basic equations:**

$$\mathbf{M}(\mathbf{x})\ddot{\mathbf{x}} = \mathbf{f}(\mathbf{x},\dot{\mathbf{x}},t)$$

- mechanics of unconstrained system of bodies
- subjected to configuration-dependent loads

Can be obtained from many (equivalent!) approaches:

- Newton-Euler: linear/angular equilibrium of each body
- d'Alembert-Lagrange: virtual work of active forces/moments
- Gauss, Hertz, Hamilton, ...: variational principles



### Constrained system: kinematic constraints

holonomic

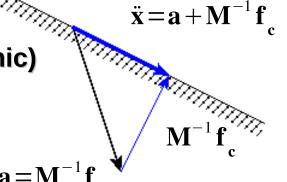
$$\phi(\mathbf{x},t)=0$$

non-holonomic (not integrable to holonomic)

$$\psi(\mathbf{x},\dot{\mathbf{x}},t)=0$$

usually

$$\mathbf{A}(\mathbf{x},t)\dot{\mathbf{x}}=\mathbf{b}(\mathbf{x},t)$$



- algebraic relationship between kinematic variables
- explicitly dependent on time: rheonomic
- scleronomous otherwise



#### Minimal set:

$$\mathbf{x} = \mathbf{x}(\mathbf{q}, t)$$

usually, this relationship:

- is not known in advance, or
- cannot be easily made explicit with respect to q
   Coordinate partitioning is required, e.g.:
  - direct elimination from derivative of constraint equation
  - QR or similar decomposition

Results in Maggi-Kane equations and similar approaches

Small system is obtained by expensive numerical reduction (unless topology knowledge can be exploited)

## **Multibody dynamics**

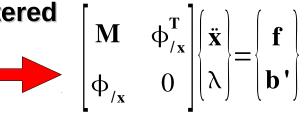


**Redundant set:** 
$$\delta(\lambda \cdot \phi) = \delta \lambda \cdot \phi + \delta \mathbf{x} \cdot \phi_{\mathbf{x}}^{\mathbf{T}} \lambda$$

$$\delta (\mu \cdot \psi) = \delta \mu \cdot \psi + \delta \mathbf{x} \cdot \psi_{/\dot{\mathbf{x}}}^{T} \mu$$

### By Lagrange multipliers:

- dynamics of constrained system using physical coordinates
- constraint reactions applied to equations of motion
- algebraic constraints explicitly added to the system
- Multiple bodies with few actual dofs:
  - system size nearly doubles
- Multiple bodies with few constraints:
  - system size not significantly altered
- Sparsity is almost preserved





#### Constraint equations written "as is":

$$\phi(\mathbf{x},t)=0$$

problem becomes differential algebraic (DAE); issues:

- needs specific care to be solved: (nearly) L-stable integration, i.e.
  - unconditionally stable, and
  - $\Delta \mathbf{x}_{\mathbf{k}+1} \rightarrow 0$  for  $\Delta t \rightarrow \infty$
- the constraint equation implies the additional constraints

$$\phi(\mathbf{x}, t) = 0$$
$$\ddot{\phi}(\mathbf{x}, t) = 0$$

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

but they are not explicitly enforced: may need constraint stabilization (Gear et al.)

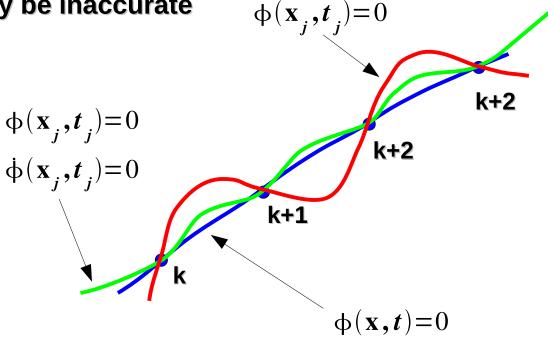
# **Multibody dynamics**



### **Constraint equations:**

- x is correct
- derivatives may be inaccurate

multipliers may be inaccurate





### Alternative: constraint equations differentiated to second order:

$$\phi_{/x}\ddot{x} = b'$$

problem remains ordinary differential (ODE);

- can be solved by conditionally stable algorithms
- the constraint equation does not imply the original constraints

$$\phi(\mathbf{x}, t) = 0$$
$$\phi(\mathbf{x}, t) = 0$$

#### definitely needs constraint stabilization!

common technique: Baumgarte

$$\phi_{/\mathbf{x}}\ddot{\mathbf{x}} = \mathbf{b}' - 2\alpha\dot{\phi} - \beta^2\phi$$

(violation governed by asymptotically stable linear differential eq.)

#### **Software architectures**



- Monolithic:
  - user prepares specific model using built-in library elements
  - general-purpose solver swallows model and spits results
- Library:
  - user writes specific solver using library elements
    - usually needs programming skills; the solver must be compiled
  - specific solver solves the problem and spits results
- Symbolic manipulators:
  - user writes equations
  - symbolic manipulation engine solves equations and spits results
- Modelica (and Modelica-like):
  - user prepares model using a modeling language and libs
  - general-purpose interpreter generates specific solver
  - specific solver solves the problem and spits results

#### **Software architectures**



#### Free software examples (surely there are more):

- Monolithic:
  - MBDyn

non-free counterparts omitted

- Library:
  - DynaMechs (C++)
  - Mbs3d (requires Matlab)
  - Open Dynamics Engine (ODE) (C++)
- Symbolic manipulators:
  - 3D\_MEC
  - EasyDyn (MuPad)
  - RoboTran (requires Matlab)
- Modelica:
  - OpenModelica?

frequently architectures overlap

#### **Software architectures**



- MBDyn is monolithic
- Input consists in a text file
- The input syntax allows some flexibility, e.g.:
  - math expressions evaluation
  - variables definition
  - "rigorous" syntax checking, but free style, indentation, ...
- Relevant portions of the code are modular and can be extended by:
  - writing run-time loadable modules
  - hacking the code (it's free, all in all!)
- There is no built-in pre-post processing facility
- Help in this area is warmly appreciated!
  - MBDyn output can be translated into EasyAnim
  - there is an independent, partial customization based on Blender



## Equations of motion: for each node (purely geometrical entity),

Newton-Euler, written as first-order system of equations:

$$\mathbf{M} \dot{\mathbf{x}} = \mathbf{p}$$

$$\dot{\mathbf{p}} = \mathbf{f}(\mathbf{x}, \dot{\mathbf{x}}, t)$$

- Momentum and momenta moment instead of pseudo-velocities
- allows multiple contributions to inertia of a single node

Constrained equations in differential-algebraic form:

$$\mathbf{M} \dot{\mathbf{x}} = \mathbf{p}$$

$$\dot{\mathbf{p}} + \boldsymbol{\phi}_{/\mathbf{x}}^{\mathbf{T}} \lambda = \mathbf{f}(\mathbf{x}, \dot{\mathbf{x}}, t)$$

$$\boldsymbol{\phi}(\mathbf{x}, t) = 0$$



- Fundamental problem:
  - Integration of Initial Value Problem (IVP) in time
- Static analysis as degeneration of IVP dynamic analysis:
  - momentum and momenta moment definitions omitted
  - only gravity is considered
  - system determination only provided by kinematic constraints and deformable components
- Kinematic analysis as degeneration of IVP dynamic analysis:
  - inertia elements omitted
  - system determination only provided by kinematic constraints
  - deformable components can act as "regularization"

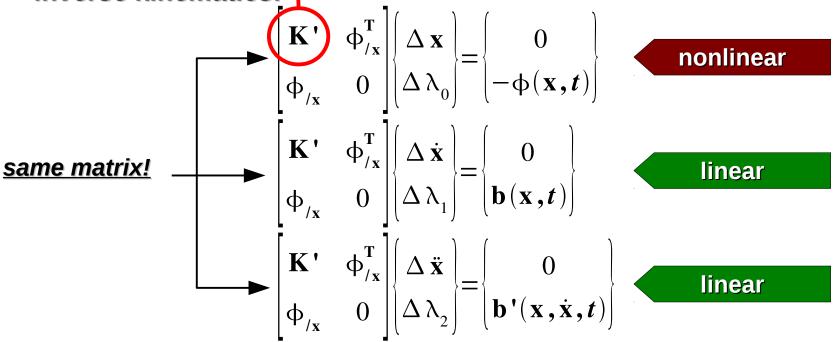
#### **Problems**



#### K'=I: Moore-Penrose pseudo-inverse

# Experimental inverse dynamics problem

inverse kinematics:



- the RHS contains the desired motion and its derivatives
- the (regularized) static analysis provides the kinematic inversion  $\phi^T \wedge \lambda = \mathbf{f}'(\mathbf{x}, \dot{\mathbf{x}}, \ddot{\mathbf{x}}, \mathbf{t})$

$$\phi_{/\mathbf{x}}^{\mathbf{T}} \Delta \lambda = \mathbf{f}'(\mathbf{x}, \dot{\mathbf{x}}, \ddot{\mathbf{x}}, t)$$

linear



- Experimental direct eigenanalysis
  - issues with constraints formulation (mainly with rotations)
  - issues with equations implementation (matrices not available)
- Relative coordinate frame dynamics
  - imposed frame motion: modifications only to RHS inertia elems
  - instrumental for many helicopter rotor dynamics problems



- Mechanical degrees of freedom:
  - structural node positions in the absolute reference frame
  - structural node orientation with respect to the absolute frame
- Kinematics is always written with respect to the absolute frame
- Newton-Euler equations are written in the absolute frame
  - moment equilibrium (Euler) equations are written with respect to the respective (moving) node
- Special elements may introduce further approximations
  - e.g. Component Mode Synthesis (CMS) element



### Orientation handling:

- orientation variables: Cayley-Gibbs-Rodrigues parameters
- orientation matrix:

$$\mathbf{R} = \mathbf{R}(\mathbf{g})$$

orthonormality:

$$\mathbf{R}^{\mathbf{T}} = \mathbf{R}^{-1}$$

derivative:

$$\dot{\mathbf{R}} \, \mathbf{R}^{\mathrm{T}} = \omega \times = (\mathbf{G}(\mathbf{g}) \, \dot{\mathbf{g}}) \times$$

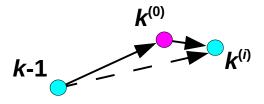
 incremental approach from step k to k+1 to eliminate the orientation parameters singularity issue (increments are necessarily small for accuracy):

$$\mathbf{R}_{k} = \mathbf{R}(\mathbf{g}_{k}) \mathbf{R}_{k-1}$$

# **Arbitrary motion description**



- Orientation handling:
  - the actual orientation variables are the Cayley-Gibbs-Rodrigues parameters relative to the correction phase of each step
  - k: time step counter
  - i: correction iteration counter (0: predicted value)



Orientation matrix:

$$\mathbf{R}_{\mathbf{k}}^{(i)} = \mathbf{R}(\mathbf{g}_{\Delta}^{(i)}) \mathbf{R}_{\mathbf{k}}^{(0)}$$

Derivative:

$$\dot{\mathbf{R}}_{k}^{(i)}(\mathbf{R}_{k}^{(i)})^{\mathrm{T}} = \boldsymbol{\omega}_{k}^{(i)} \times = (\mathbf{R}(\mathbf{g}_{\Delta}^{(i)}) \boldsymbol{\omega}_{k}^{(0)}) \times + (\mathbf{G}(\mathbf{g}_{\Delta}^{(i)}) \dot{\mathbf{g}}_{\Delta}^{(i)}) \times$$



- Incremental orientation from previous step:
  - Orientation parameters order of magnitude:

$$\mathbf{g} \sim \mathbf{O}(\|\mathbf{\omega}\| \Delta \mathbf{t})$$

- Incremental orientation from prediction (as in MBDyn):
  - Orientation parameters order of magnitude:

$$\mathbf{g}_{\wedge} \sim \mathbf{O}(\Delta \mathbf{t}^{n+1})$$

where n is the min between the order of the predictor and of the integration method (MBDyn: 3 and 2, respectively, so n = 2)

As a consequence:

$$\mathbf{R}(\mathbf{g}_{\Lambda}) \approx \mathbf{I}$$

(only in Jacobian)

$$G(g_{\wedge}) \approx I$$

$$\dot{\mathbf{G}}(\mathbf{g}_{\wedge}) \approx 0$$

## **Deformable components**



- Lumped deformable components
  - rod (1D)
  - linear, angular components (3D)
  - linear & angular component (6D)
- Intrinsic, composite-ready Finite-Volume beam element
  - arbitrary constitutive law
  - piezoelectric constitutive law
  - aerodynamic beam element
- Intrinsic, composite-ready shell and membrane elements
- Component Mode Synthesis (CMS)
  - attached to a floating frame (a node)
  - linear state-space representation of unsteady aerodynamics



### **Lumped deformable components (3D, 6D):**

$$\theta = \mathbf{a} \mathbf{x} (\exp^{-1} (\mathbf{R}_{\mathbf{a}}^{\mathsf{T}} \mathbf{R}_{\mathbf{b}}))$$
$$\mathbf{m} = \mathbf{R} (\xi \theta) \tilde{\mathbf{m}} (\theta)$$

- Attached form:  $\xi=0, \xi=1$ 
  - Constitutive properties referred to either of the connected nodes
- Intrinsic form (invariant:  $\xi=1/2$  ):
  - Constitutive properties referred to a floating reference frame
  - Intrinsically handles geometrical nonlinearity related to rotations
  - Correctly captures bending-torsion buckling behavior
  - Essential for anisotropic deformable components

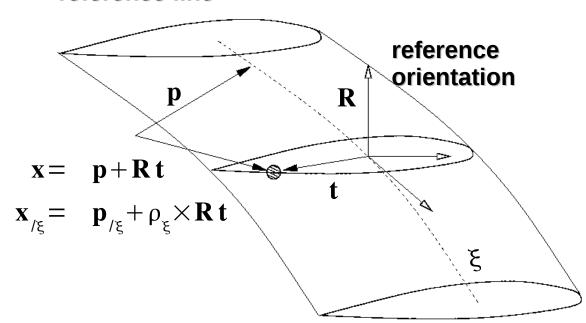
# **Deformable components**



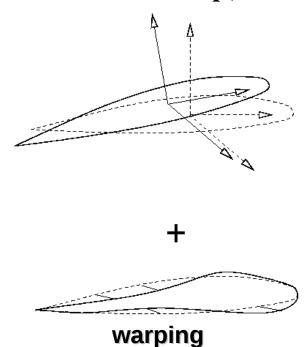
### Intrinsic, composite-ready beam

- Topology:
  - 1D reference line p, 1D reference structure R
  - 2D section characterization

#### reference line



## reference motion: p, R



## **Deformable components**



## Intrinsic, composite-ready beams

strain measure:

$$\mathbf{v} = \mathbf{R}^{\mathbf{T}} \mathbf{p}_{/\xi} - \mathbf{R}_{0}^{\mathbf{T}} \mathbf{p}_{0/\xi}$$

$$\mathbf{\kappa} = \mathbf{R}^{\mathbf{T}} \mathbf{p}_{\xi} - \mathbf{R}_{0}^{\mathbf{T}} \mathbf{p}_{\xi 0}$$

equilibrium (from VWP):

$$\mathbf{f}_{/\xi} = \tau$$

$$\mathbf{m}_{/\xi} + \mathbf{p}_{/\xi} \times \mathbf{f} = \mu$$

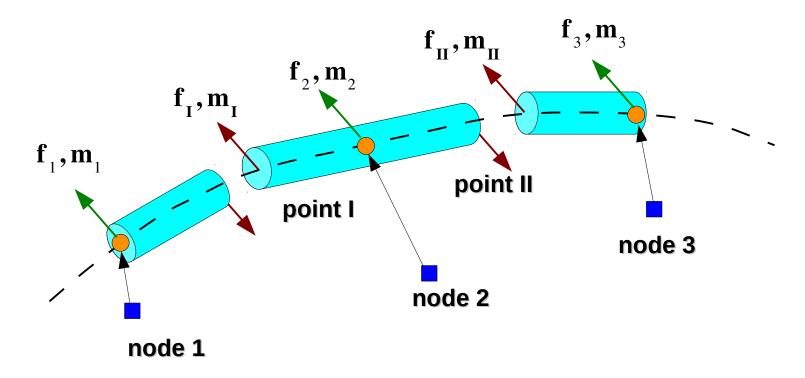
constitutive properties:

$$\mathbf{f} = \mathbf{f}(\mathbf{v}, \mathbf{\kappa})$$
$$\mathbf{m} = \mathbf{m}(\mathbf{v}, \mathbf{\kappa})$$



## Intrinsic, composite-ready beams: 3-node discretization

- Finite Volume approach: equilibrium of finite portions of beam
- internal forces function of node kinematics thru constitutive laws
- warping goes into constitutive properties computation





### **Numerical integration**

implicit, (quasi-)L stable 2 step algorithm

$$\mathbf{y}_{k} = \mathbf{a}_{1} \mathbf{y}_{k-1} + \mathbf{a}_{2} \mathbf{y}_{k-2} + \Delta t (\mathbf{b}_{0} \dot{\mathbf{y}}_{k} + \mathbf{b}_{1} \dot{\mathbf{y}}_{k-1} + \mathbf{b}_{2} \dot{\mathbf{y}}_{k-2})$$

- tunable algorithmic dissipation: asymptotic spectral radius  $1\rightarrow 0$ 
  - asymptotic spectral radius = 0: 2<sup>nd</sup> order BDF
  - "optimal" dissipation: spectral radius ~ 0.6
- second-order accurate, with third-order accurate predictor
- variable time step
- not ideal for non-smooth problems (multi-step)
- different integrators can be used; new ones can be implemented

## **Solving the problem**



Prediction:

$$\dot{\mathbf{y}}_{k}^{(0)} = (\mathbf{m}_{1} \mathbf{y}_{k-1} + \mathbf{m}_{2} \mathbf{y}_{k-2}) / \Delta t + \mathbf{n}_{1} \dot{\mathbf{y}}_{k-1} + \mathbf{n}_{2} \dot{\mathbf{y}}_{k-2}$$

$$\mathbf{y}_{k}^{(0)} = \mathbf{a}_{1} \mathbf{y}_{k-1} + \mathbf{a}_{2} \mathbf{y}_{k-2} + \Delta t (\mathbf{b}_{0} \dot{\mathbf{y}}_{k}^{(0)} + \mathbf{b}_{1} \dot{\mathbf{y}}_{k-1} + \mathbf{b}_{2} \dot{\mathbf{y}}_{k-2})$$

Correction iteration:

$$\mathbf{f}_{\dot{\mathbf{y}}} \Delta \dot{\mathbf{y}}^{(i)} + \mathbf{f}_{\dot{\mathbf{y}}} \Delta \mathbf{y}^{(i)} = -\mathbf{f}(\dot{\mathbf{y}}_{k}^{(i-1)}, \mathbf{y}_{k}^{(i-1)}, t_{k})$$

but

$$\Delta \mathbf{y}^{(i)} = \Delta t \mathbf{b}_0 \Delta \dot{\mathbf{y}}^{(i)}$$

the problem becomes algebraic

$$\begin{split} &(\mathbf{f}_{/\dot{\mathbf{y}}} + \Delta \, t \, \mathbf{b}_0 \, \mathbf{f}_{/\mathbf{y}}) \Delta \, \dot{\mathbf{y}}^{(i)} = -\mathbf{f} \, (\dot{\mathbf{y}}_k^{(i-1)}, \mathbf{y}_k^{(i-1)}, t_k) \\ &\dot{\mathbf{y}}_k^{(i)} = \dot{\mathbf{y}}_k^{(i-1)} + \Delta \, \dot{\mathbf{y}}^{(i)} \\ &\mathbf{y}_k^{(i)} = \mathbf{y}_k^{(i-1)} + \Delta \, t \, \mathbf{b}_0 \, \Delta \, \dot{\mathbf{y}}^{(i)} \end{split}$$



### **Model assembly**

- model could be input incorrectly
- initial values of the state (position, velocity, reactions) are needed
- this might not be a trivial task
- initial state values must comply with constraints:

$$\phi(\mathbf{x}_{0}, \mathbf{t}_{0}) = 0$$
$$\phi(\mathbf{x}_{0}, \mathbf{t}_{0}) = 0$$

a dummy static nonlinear problem is solved (regularization):

$$\mathbf{K'}(\mathbf{x} - \mathbf{x}_0) + \boldsymbol{\phi}_{/\mathbf{x}}^{\mathbf{T}} \boldsymbol{\lambda}' = \mathbf{f'}$$

$$\mathbf{C'}(\dot{\mathbf{x}} - \dot{\mathbf{x}}_0) + \boldsymbol{\phi}_{/\mathbf{x}}^{\mathbf{T}} \boldsymbol{\mu}' = \dot{\mathbf{f}'}$$

$$\boldsymbol{\phi}(\mathbf{x}, \boldsymbol{t}_0) = 0$$

$$\boldsymbol{\dot{\phi}}(\mathbf{x}, \boldsymbol{t}_0) = 0$$

# **Solving the problem**



### Solution initialization (so-called "derivatives")

explicit problem:

$$\dot{\mathbf{y}} = \mathbf{f}(\mathbf{y}, t)$$

implicit problem:

$$0 = \mathbf{f}(\mathbf{y}, \dot{\mathbf{y}}, t)$$

modified correction phase to initialize solution:

$$(\mathbf{f}_{/\dot{\mathbf{y}}} + c \, \mathbf{f}_{/\mathbf{y}}) \Delta \, \dot{\mathbf{y}}^{(i)} = -\mathbf{f} \, (\dot{\mathbf{y}}_{0}^{(i-1)}, \mathbf{y}_{0}, t_{0})$$

$$\dot{\mathbf{y}}_{0}^{(i)} = \dot{\mathbf{y}}_{0}^{(i-1)} + \Delta \, \dot{\mathbf{y}}^{(i)}$$

$$\mathbf{y}_{0} = \mathbf{y}_{0}$$

- convergence no longer quadratic, but saves lots of code duplication
- Setting (c=0) might not work (problem can be structurally singular)

# **Extracting useful information**



- Detailed analysis requires detailed models, but...
- excessive details endanger the chance to extract useful information
- Proper Orthogonal Decomposition allows to extract information from redundant measures
- Consider a set of N measurements X for n time steps; their SVD:

$$\mathbf{X}^{\mathrm{T}}(\in \mathbb{R}^{n \times N}) = \mathbf{U} \Sigma \mathbf{V}^{\mathrm{T}}$$

The singular values allow to determine the m most relevant signals

 $\mathbf{X}_{1:m,n}^{\mathbf{T}} = \mathbf{U}_{n,1:m} \boldsymbol{\Sigma}_{1:m,1:m} \mathbf{V}_{N,1:m}^{\mathbf{T}}$ 

Note that

$$\mathbf{X}^{\mathsf{T}}\mathbf{X} = \mathbf{U} \, \boldsymbol{\Sigma}^{2} \mathbf{U}^{\mathsf{T}} \qquad \mathbf{X} \, \mathbf{X}^{\mathsf{T}} = \mathbf{V} \, \boldsymbol{\Sigma}^{2} \mathbf{V}^{\mathsf{T}}$$

$$\mathbf{U}^{\mathsf{T}}\mathbf{X}^{\mathsf{T}} = \boldsymbol{\Sigma} \, \mathbf{V}^{\mathsf{T}} \qquad \mathbf{X}^{\mathsf{T}}\mathbf{V} = \mathbf{U} \, \boldsymbol{\Sigma}$$

This allows to efficiently compute the singular values and the POMs

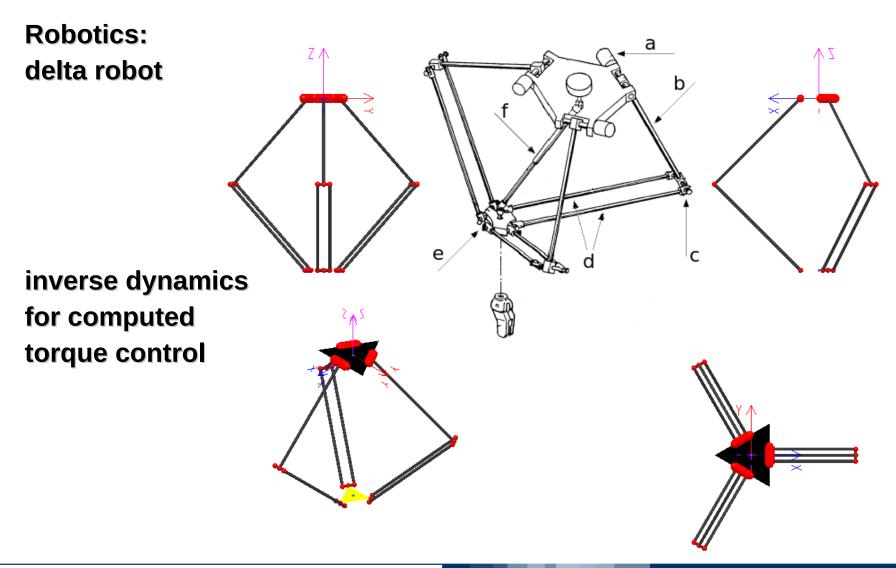


The POMs can be used to identify a transition matrix

$$\mathbf{X}^{(k+1)} = \Phi \, \mathbf{X}^{(k)}$$

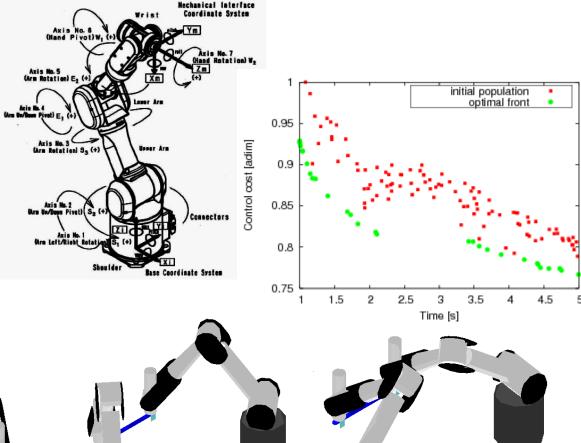
- If X contains the free response of the system, the transition matrix allows to estimate the relevant eigenvalues (AR model)
- More sophisticated system identification techniques can be used (model order reduction is an open research field)
- A technique based on covariance estimates from time histories has been recently proposed; works for:
  - free response
  - forced response
  - unmeasured forced response

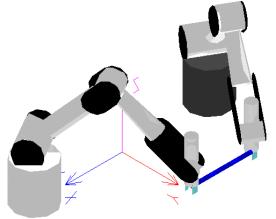


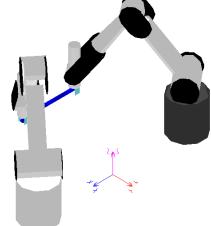


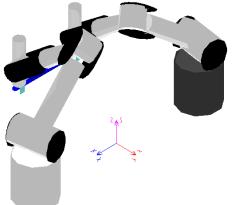
RT-MBDyn

**Robotics: PA-10** inverse kinematics with path optimization of cooperating robots



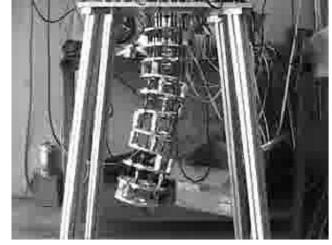


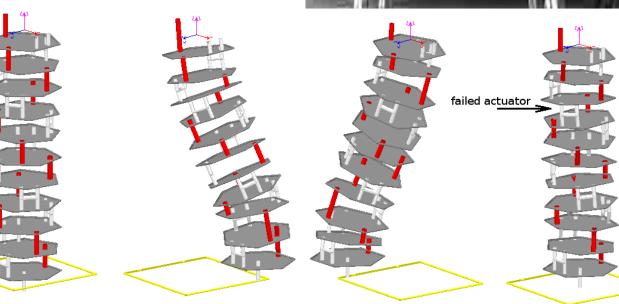


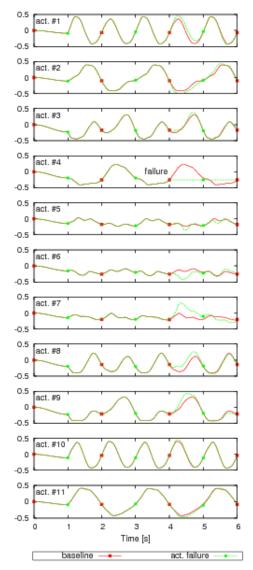




Robotics:
biomimetic robot
real-time motion
planning by inverse
kinematics with
fault detection









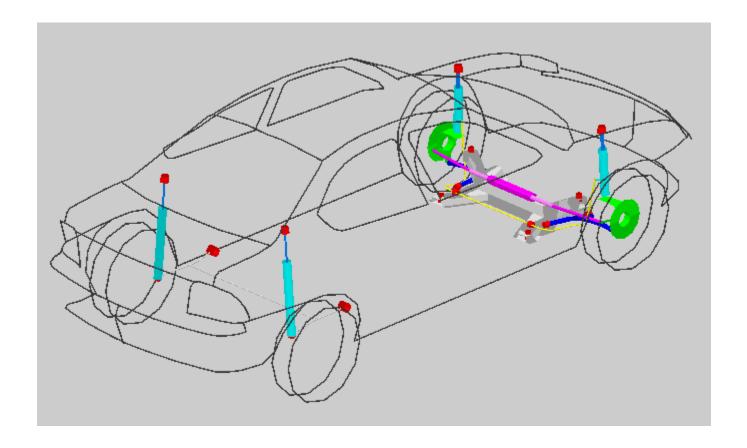
#### **Industrial processes:**

- simulation of automotive components assembly (car brake pipe) to:
  - check stresses introduced during assembly
  - check loads on supports introduced during assembly
  - check interference with other parts during assembly
  - check interference with other parts during operation
- the model has been developed by a rubber manufacturer
- it is used for product design and certification
- it required the development of specific features for solution partitioning, which are now part of MBDyn



**Automotive: mechanical modeling of suspensions** 

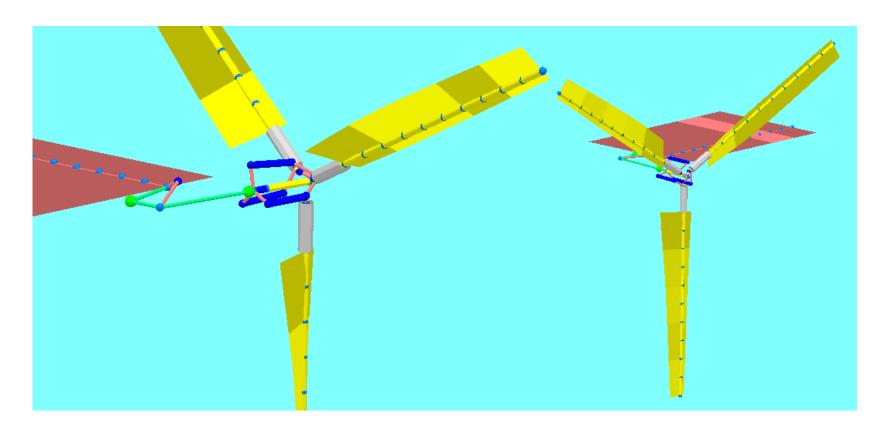
purpose: determine loads in rubber bushings and other components





### **Rotorcraft dynamics and aeroservoelasticity:**

WRATS (NASA/Army) tiltrotor aeromechanics

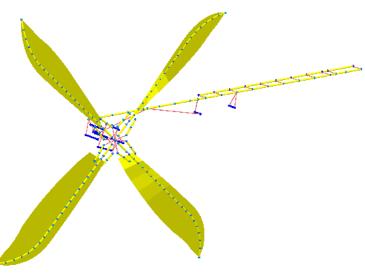


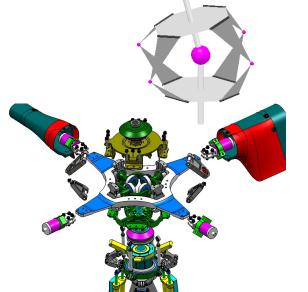
RT-MBDyn

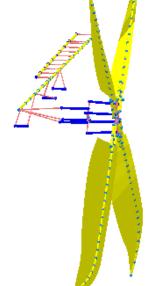
Rotorcraft dynamics and aeroservoelasticity:

ERICA (AgustaWestland)
 tiltrotor aeromechanics
 (ADYN, NICETRIP)











- Multiscale handling of submodels with different dynamics
  - aircraft flight mechanics (~1 to 5 Hz: very slow)
  - main rotor dynamics (~5 to 40 Hz: intermediate)
  - tail rotor dynamics (~25 to >100 Hz: fast)
- Interfacing with different domains
  - Fluid-structure (Lagrangian/Eulerian modeling of workflows)
  - structure-structure
  - active control of large deformable systems
- Better abstraction/modularization of components/solution phases
  - more freedom in model customization
  - tight integration into nonlinear structural analysis (Aster?)
- More...

## **Documentation and support**



- Theory manual:
  - Incomplete; needs lots of work
- User manual
  - available and up to date
  - Tutorials
    - available, but reportedly too simple; need work
- Applications manual
  - available, but only few applications so far
- Installation manual
  - available, incomplete and outdated (not critical)
- Mailing lists
  - available: announce, users, devel
  - the "users" list also serves as issue tracking provision

## **Documentation and support**



- Another important item that is missing is an automated test suite
  - can be run automatically after building the software
  - allows to check build errors
  - allows to check regressions in new releases
  - serves as example of modeling functionalities
- The rest is underway (always a work in progress)

Given the nature of the project, contributions are always welcome!



