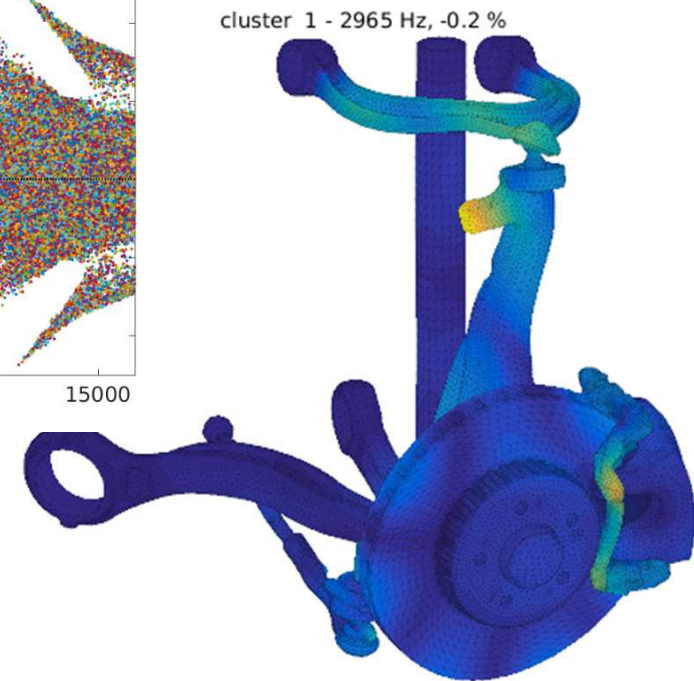
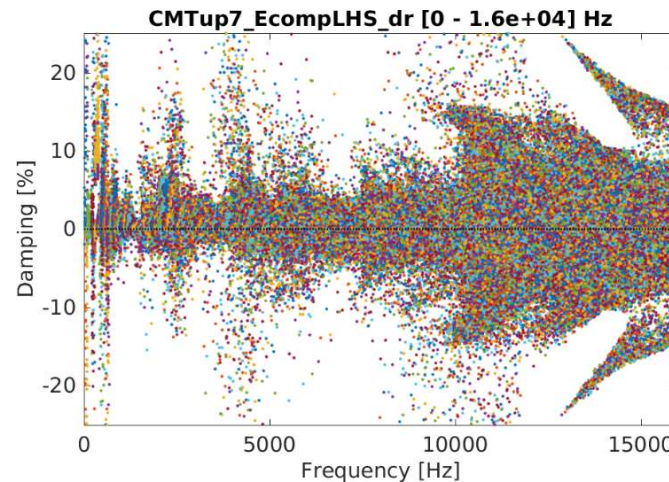
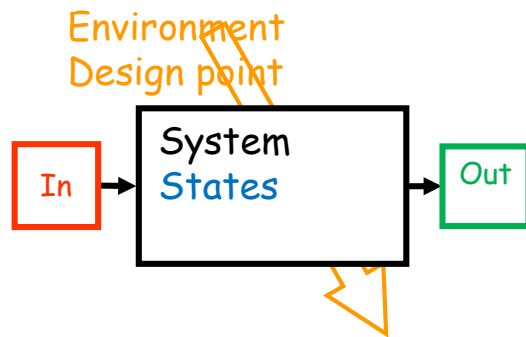


Course 7-8 parametric problems



- Parameters
- Sensitivity
- Reanalysis
- Updating

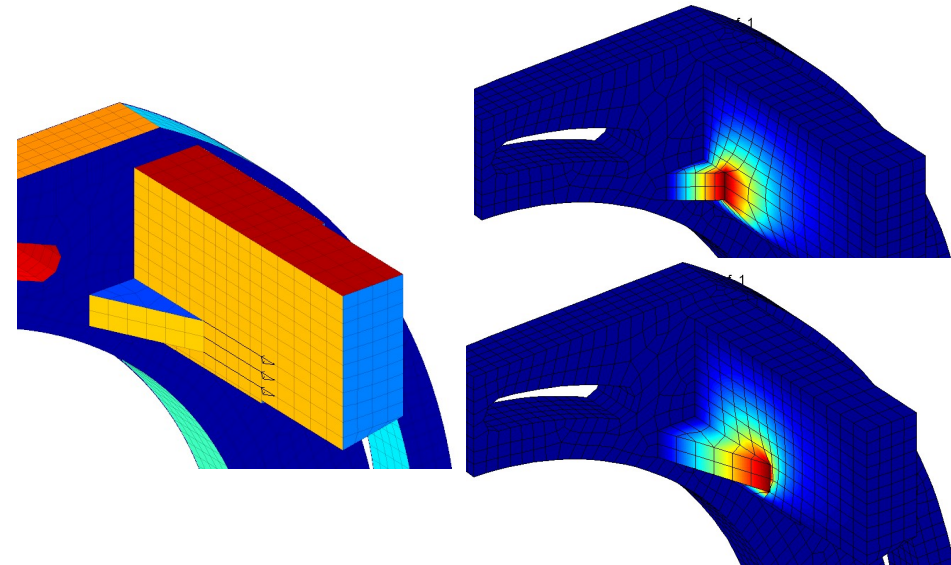
Geometry parametrization / morphing

- Shape optimization / morphing

$$P(x) = \sum_i \{p_i(x)\} q_{imaster}$$

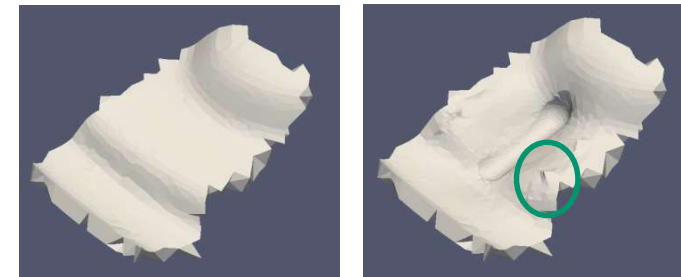
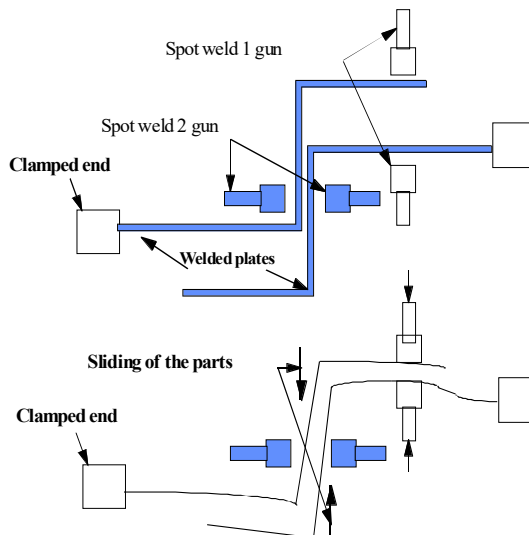
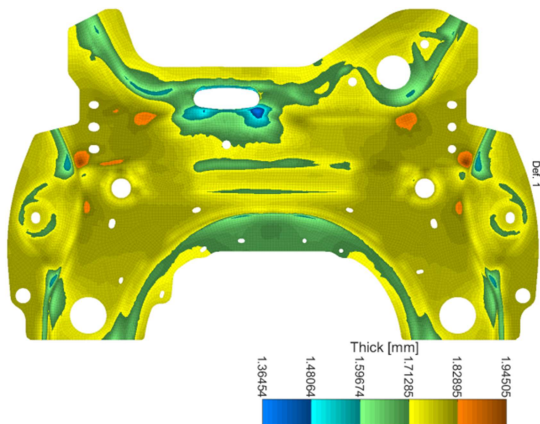
fe_shapeoptim BuildFromSel

- 1) fix bottom face, 2) prescribe edge motion
- 3) deform edges (straight)
- 4) deform faces (flat)
- 5) deform interior (*good elements ?*)



- Process simulation + field projection

fe_shapeoptim Interp



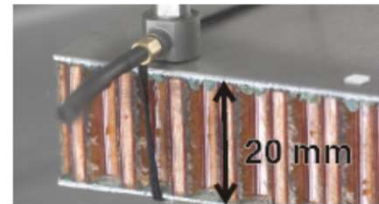
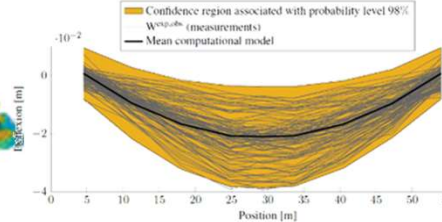
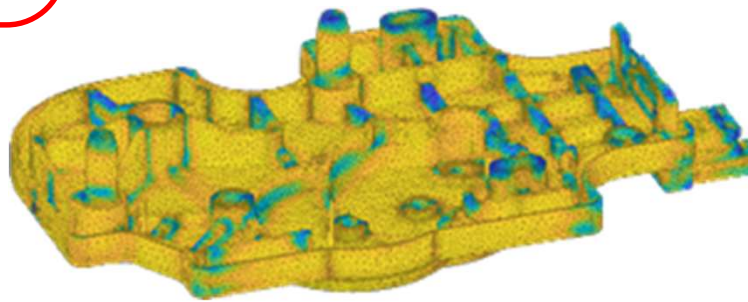
- [1] de Paula, Rejdych, Chancelier, Vermot, Balmes, « On the influence of geometry updating on modal correlation of brake components. », in *Vibrations, Chocs & Bruit*, 2012.
- [2] G. Vermot Des Roches, E. Balmes, et S. Nacivet, « Error localization and updating of junction properties for an engine cradle model », in *ISMA*, Leuven, Belgium, 2016, p. ID 372.
- [3] E. Blain, « Etudes expérimentales et numériques de la dispersion vibratoire d'assemblages soudés par points », Ph.D. thesis, Ecole Centrale de Paris, 2000.

Direct problems : material parameters

- Uniform
- Field
- Equivalent
(at certain scales)



- Geometry
- **Material parameters**
- Junction representation
- Equivalent parameters



- Basic parametrization tool : dependence on **constitutive parameters** C_{ij}

$$K = \int_{\Omega} \{\epsilon\}^T [C_{ij}] \{\epsilon\} = \sum_g B^T [C_{ij}] B w_g = \sum_{ij} C_{ij} \left[\sum_g B^T [C_{ij}^u] B w_g \right] = \sum_{ij} C_{ij} [K_{ij}]$$

Constitutive law parameterization

- Example : bar stiffness $K = \frac{EA}{L} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$ proportional to E
- Classical problem shell thickness parameterized with

$$\beta_1 = t, \beta_2 = t^3, \beta_3 = t^2 \quad \begin{Bmatrix} N_{xx} \\ N_{yy} \\ N_{xy} \\ M_{xx} \\ M_{yy} \\ M_{xy} \end{Bmatrix} = \begin{bmatrix} A & B \\ B^T & D \end{bmatrix} \begin{Bmatrix} \epsilon_{xx} \\ \epsilon_{yy} \\ \epsilon_{xy} \\ \kappa_{xx} \\ \kappa_{yy} \\ \kappa_{xy} \end{Bmatrix}$$

- Current practice : weighted element sum

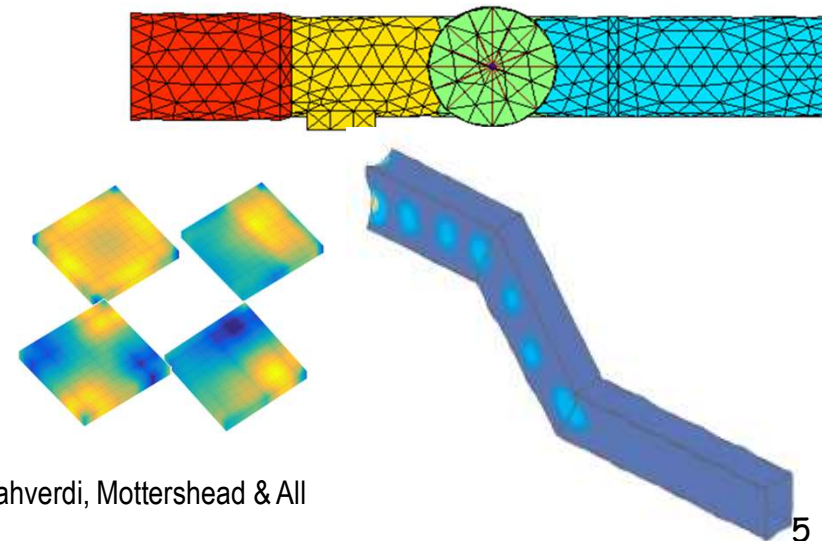
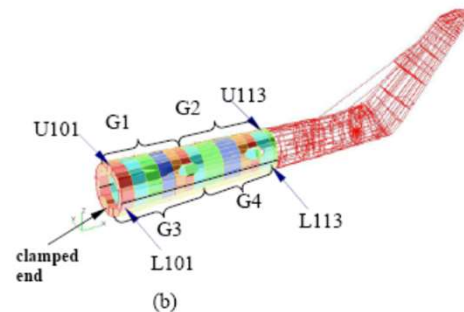
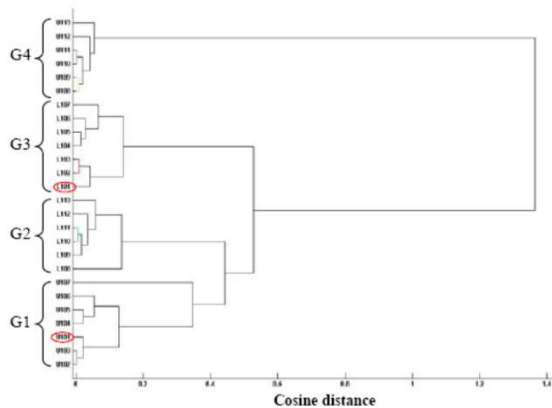
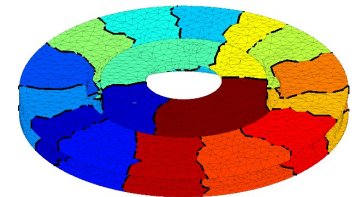
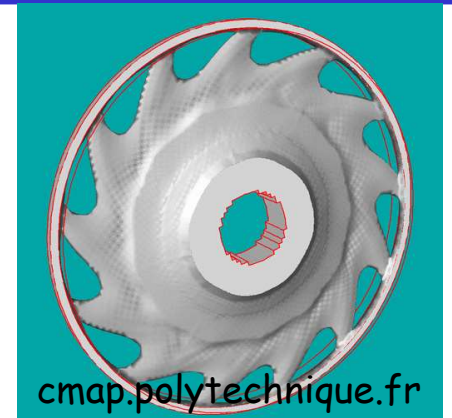
$$[M(p)] = \sum_{j=1}^{NE} \alpha_k(p) [M_k^e] \quad [K(p)] = \sum_{j=1}^{NE} \beta_k(p) [K_k^e]$$

Element/model weights

$$K(p) = \sum_e \alpha_e(p) [K^e]$$

Weighted element matrices = standard

- Element-wise (topology optimization)
- Field/groupwise
 - parameter groups, ...
 - solution of eigenvalue problem ([polynomial chaos](#), Ghanem, Soize, ...)
 - [clustering](#) (for example [k-means](#) a usual data-mining algorithm)



Clustering of Parameter Sensitivities: Examples from a Helicopter Airframe Model Updating Exercise. Shahverdi, Mottershead & All

SDT implementations : upcom / zCoef / stressCut

- Element wise $K(p) = \sum_e \alpha_k^e [K^e]$

www.sdtools.com/help/upcom.html

$$\text{mind} = \begin{bmatrix} M_s & M_e & K_s & K_e & \alpha_m & \alpha_k \\ \vdots & & & & & \\ \text{elt} & & & & & \end{bmatrix}$$

- Group wise $K(p) = \sum_g \alpha_g [K^g]$

www.sdtools.com/help/zCoef.html

```
zCoef={'Klab','mCoef','zCoef0','zCoefFcn';
      'M'    1      0      '-w.^2';
      'Ke'    0      1      '1+i*fe_def('DefEta',[ ]);
      'Kv'    0      1      'par(1)'};
```

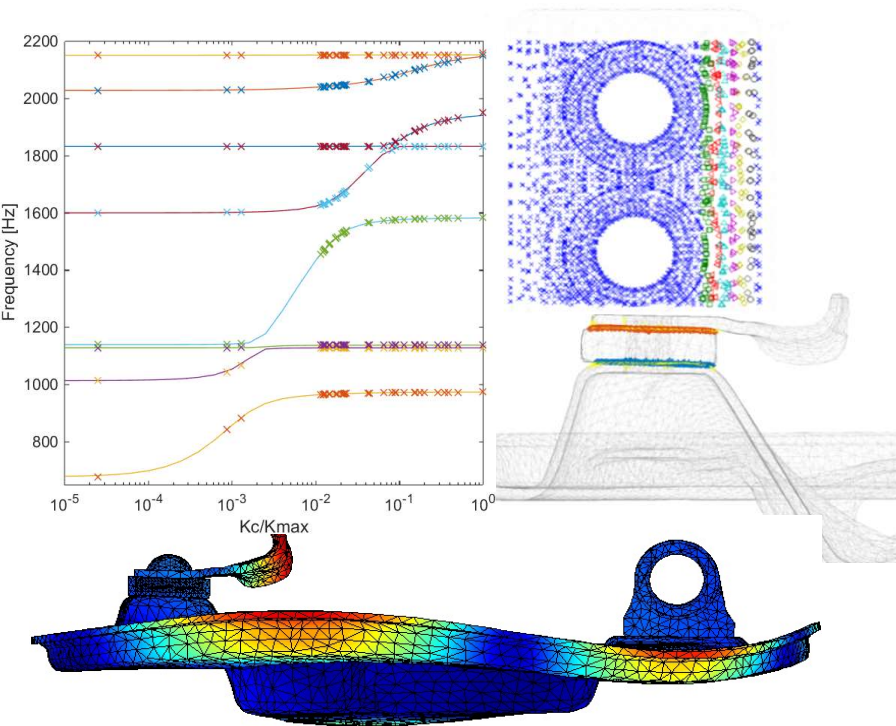
- Disassembly

www.sdtools.com/help/corstress.html

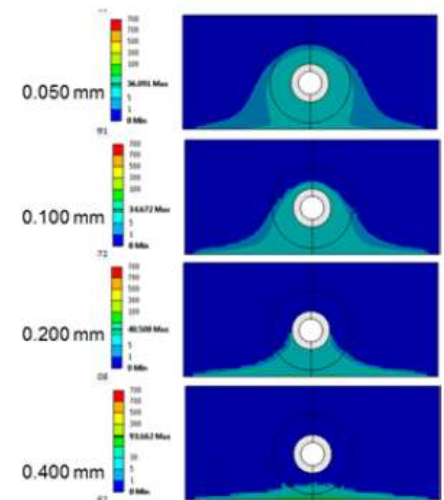
$$K(p) = [b][C_{constit}][c] = \begin{bmatrix} \ddots & & \\ & w_g J_g N_i^g & \\ & & \ddots \end{bmatrix}^T \begin{bmatrix} \ddots & & \\ & C_{ij}^g & \\ & & \ddots \end{bmatrix} \begin{bmatrix} \ddots & & \\ & N_i^g & \\ & & \ddots \end{bmatrix}_{(Ng \times Nstrain) \times N}$$

Parametrization of contact/sliding

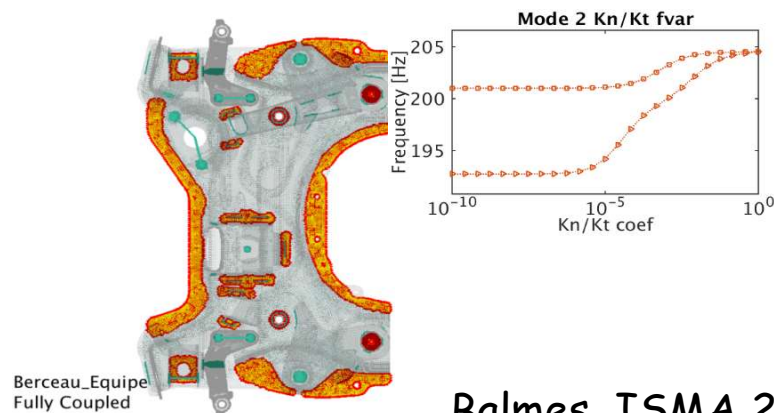
- Variable contact **surface**, **contact**, **sliding**



Chassis Brakes International
Eurobrake 2014



Goth, ISMA 2016



Balmes, ISMA 2016

Sensitivity

- Static direct and adjunct ([poly section 10.2](#))
- Frequency and shape sensitivity ([poly section 10.3](#))
- Exact and Fox-Kapoor (change to modal coordinates)

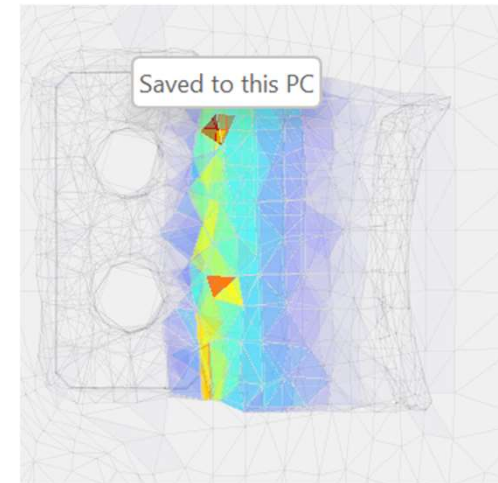
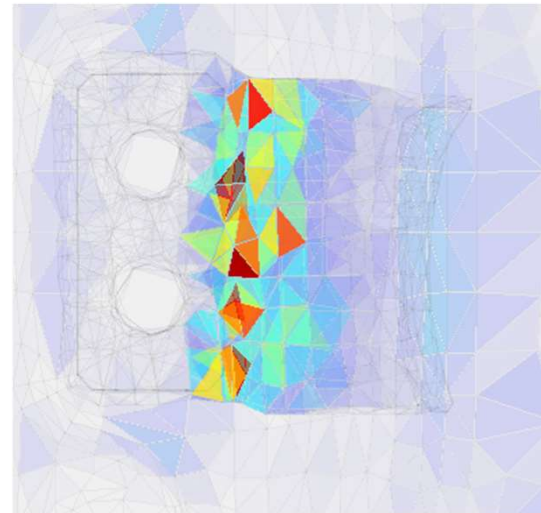
Tricks with energy display

$$\frac{\partial \omega_j^2}{\partial p} = \{\phi_j\}^T \left[\frac{\partial K}{\partial p} - \omega_j^2 \frac{\partial M}{\partial p} \right] \{\phi_j\}$$
$$\omega_j = \sqrt{\frac{\sum_e q^T K^e q}{\sum_e q^T M^e q}}$$

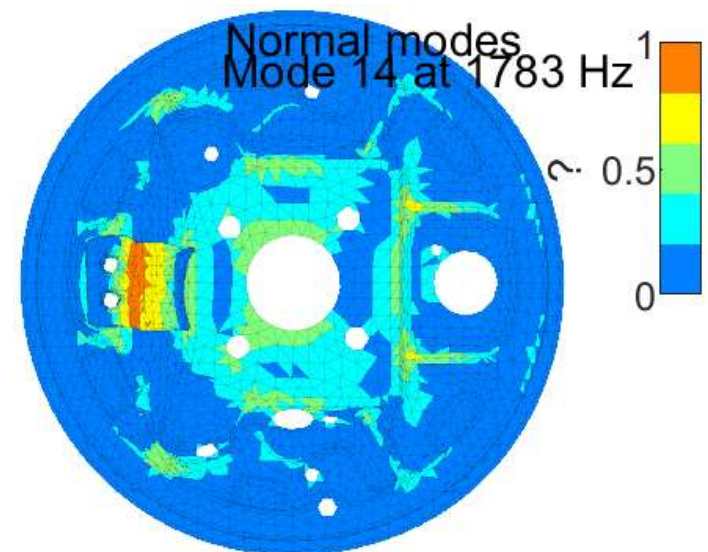
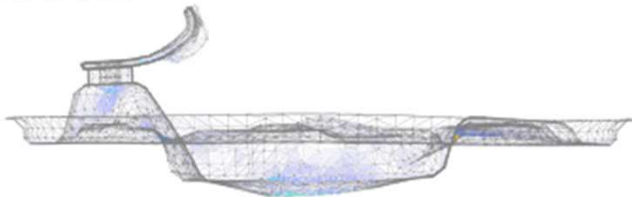
Views by

- Element energy (Abaqus ELSE)
- Energy density (ESEDEN)
- Energy in group

Give different perspectives

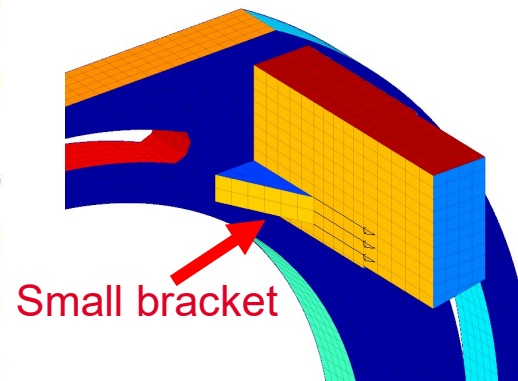
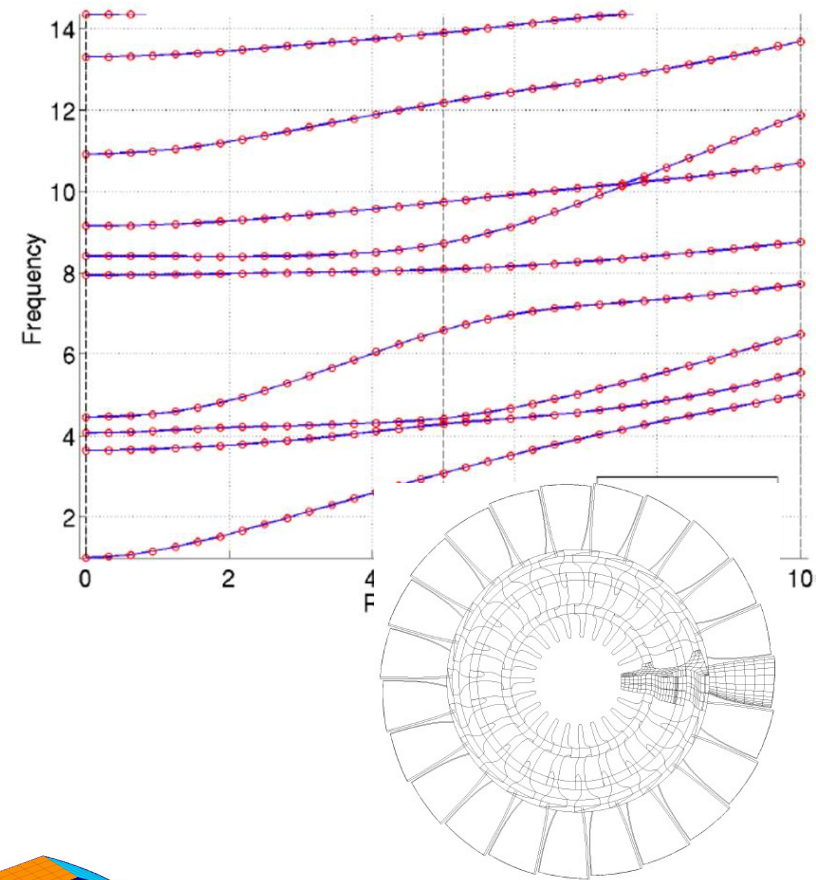
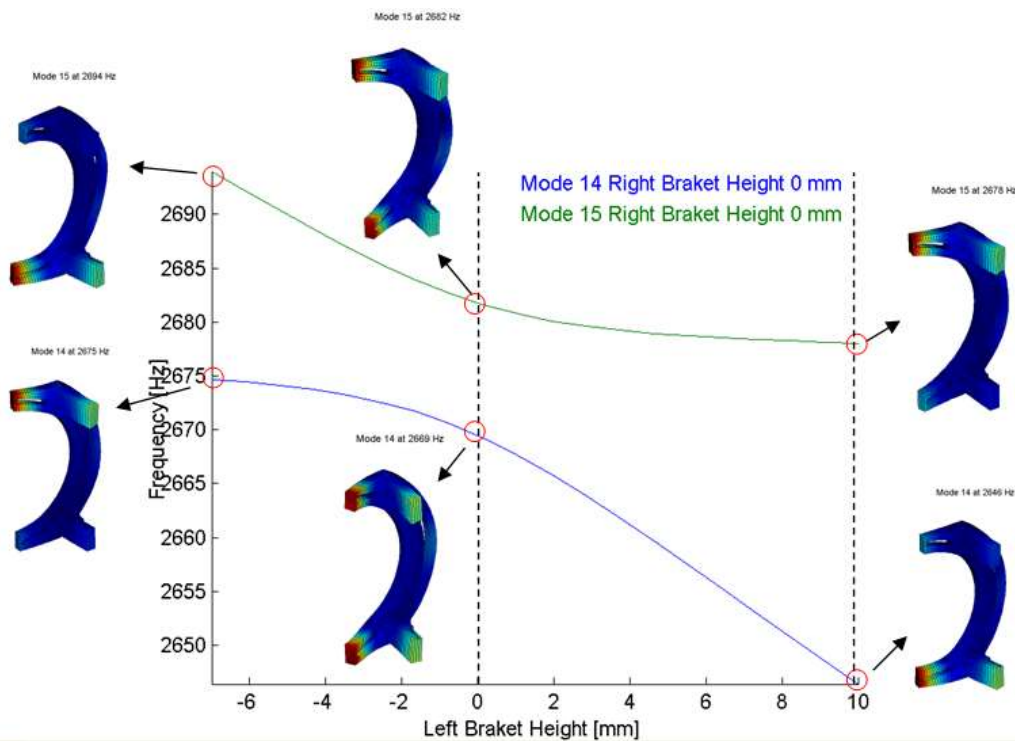


Normal modes
Mode 14 at 1783 Hz



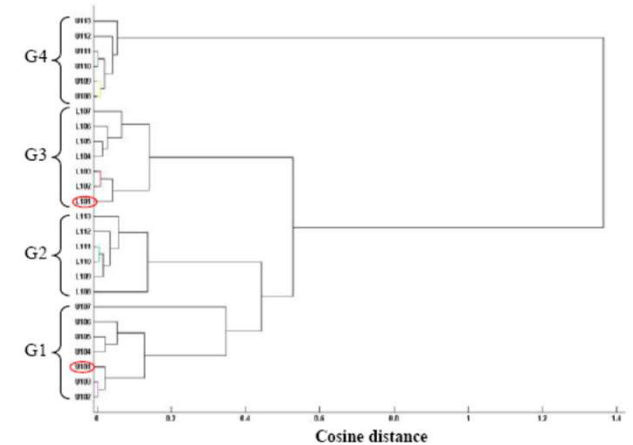
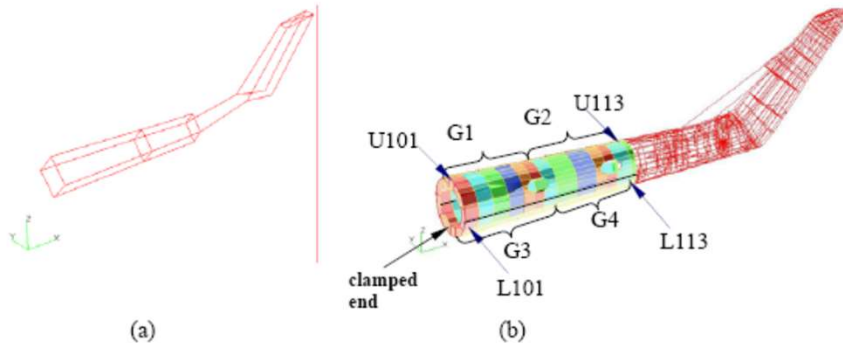
Mode crossing

- High sensitivity for close modes associated with mode crossing

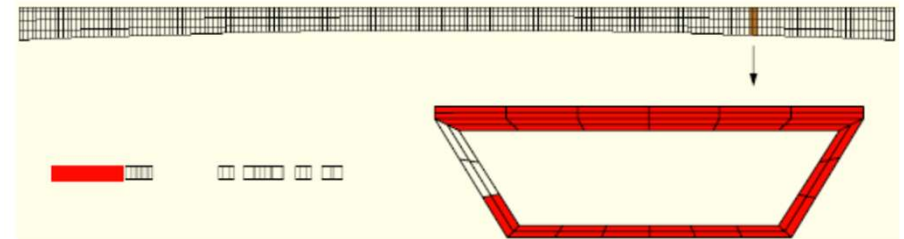
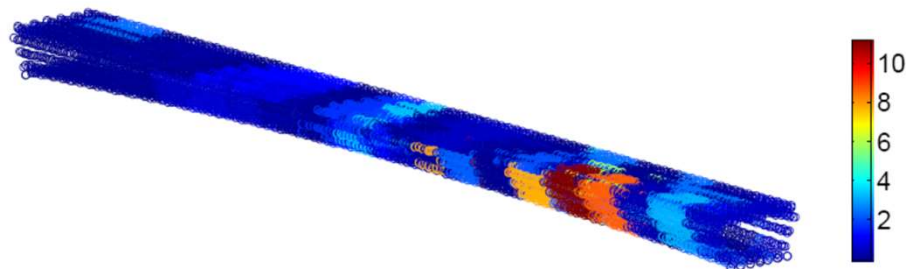


Clustering examples

- Helicopter frame



- Bridge deck



Clustering of Parameter Sensitivities: Examples from a Helicopter Airframe Model Updating Exercise.

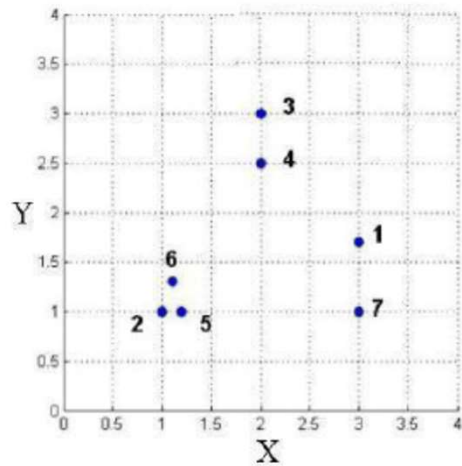
Shahverdi, Mottershead & All

Statistical Model-Based Damage Localization: a Combined Subspace-Based and Substructuring Approach

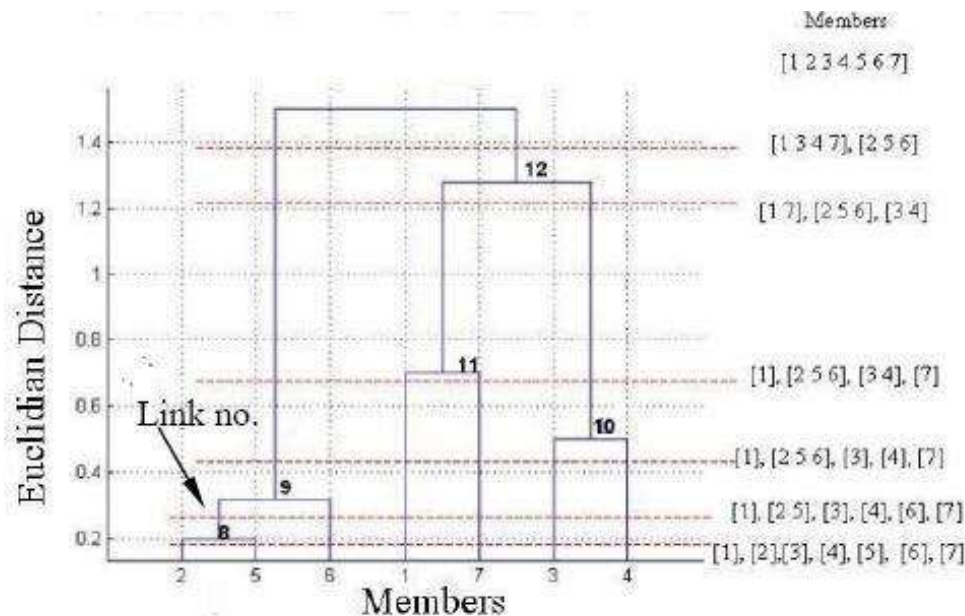
Balmes, Basseville & All

Sensitivity / clustering

- Clustering techniques (**k-means for example**) can be used to group elements with similar effects
- Key mathematical notion : cosine distance (subspace in MATLAB)



$$d_{ij} = 1 - \frac{\sum_{k=1}^p x_{ik} x_{jk}}{\left(\sum_{k=1}^p x_{ik}^2 \sum_{k=1}^p x_{jk}^2 \right)^{1/2}}$$

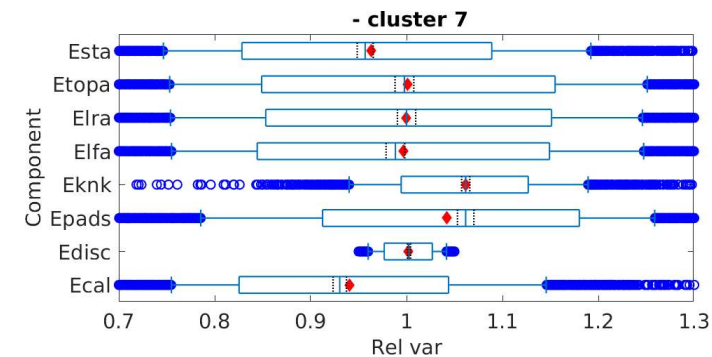
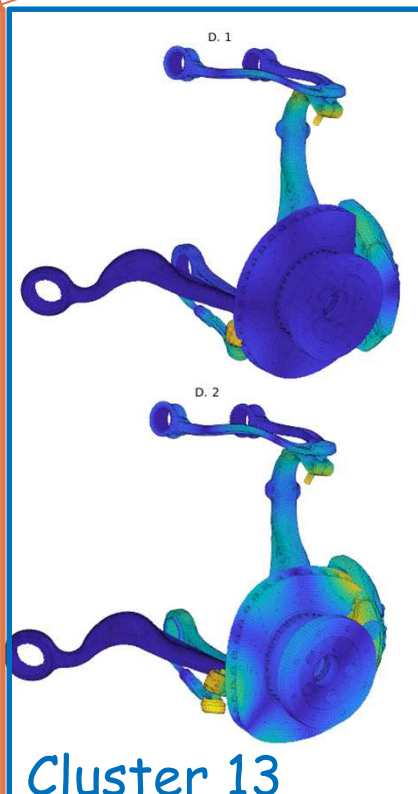
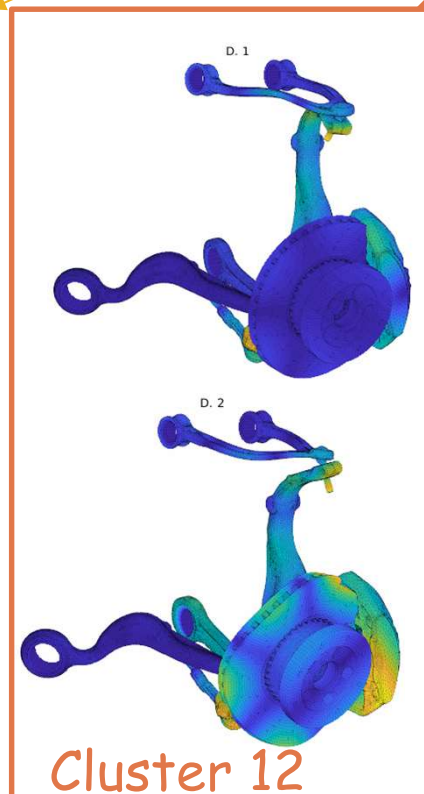
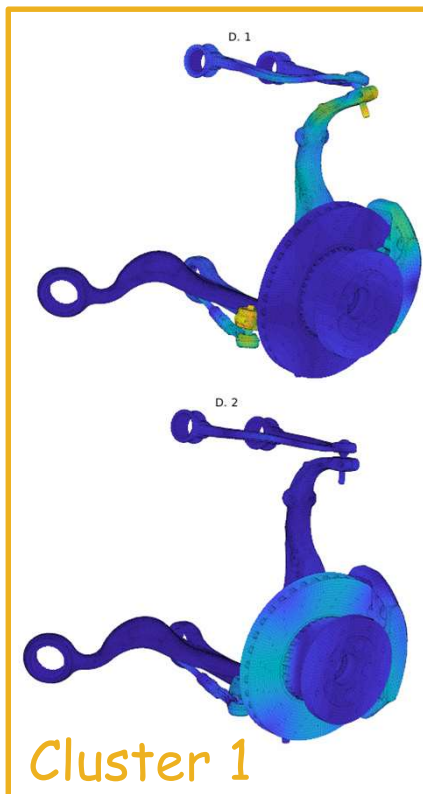
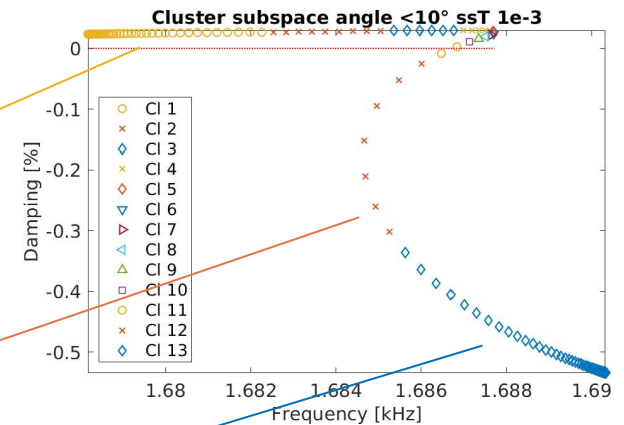


Link no.	linked objects		distance
8	2	5	0.2
9	6	8	0.32
10	3	4	0.5
11	1	7	0.7
12	10	11	1.28
	9	12	1.5

Clustering of Parameter Sensitivities: Examples from a Helicopter Airframe Model Updating Exercise.
Shahverdi, Mottershead & All

Subspace clustering example

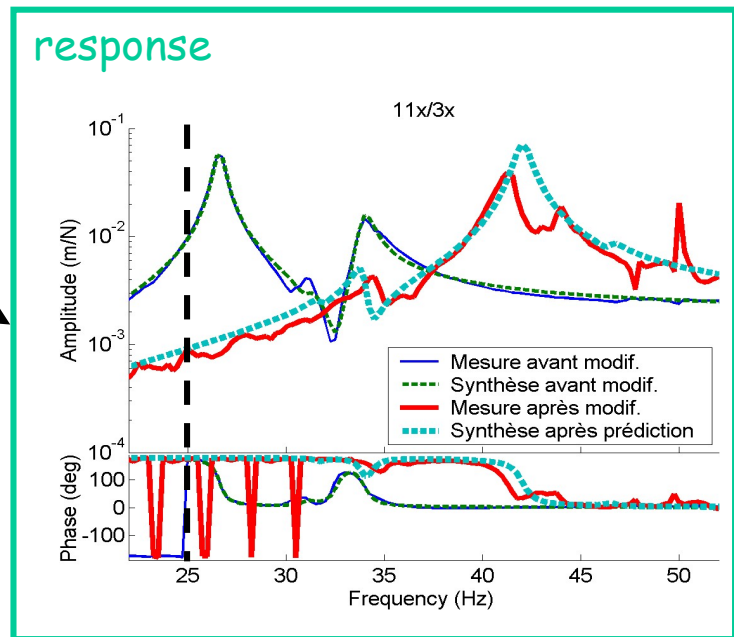
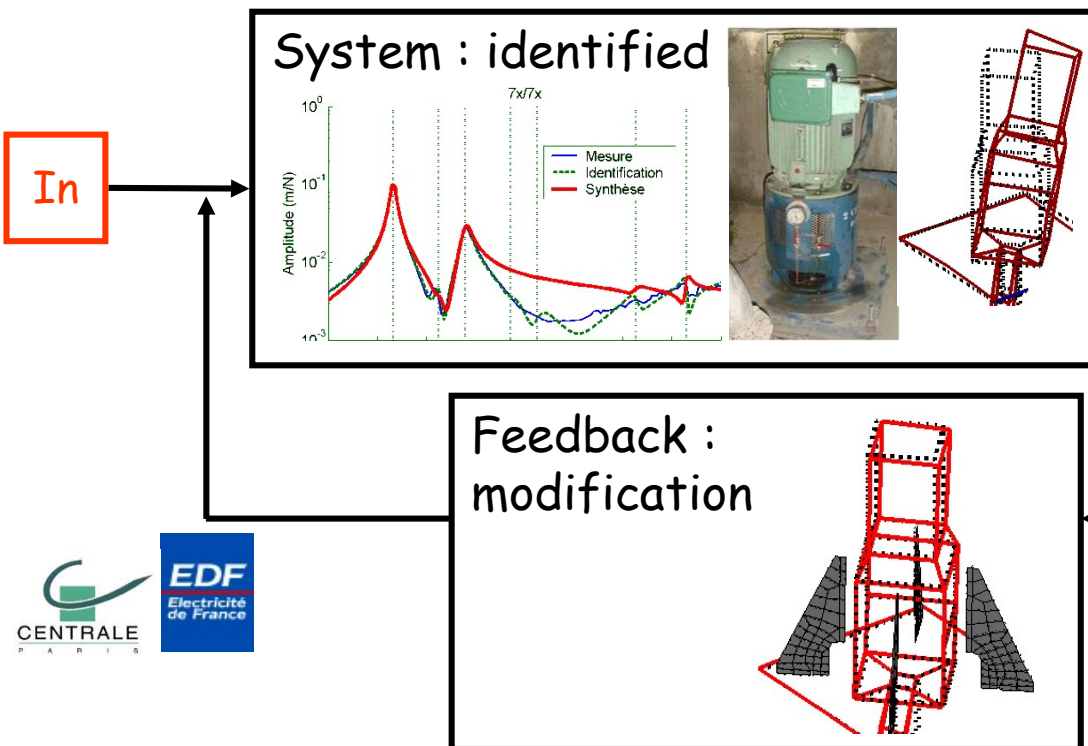
- Clustering
 - 2D complex mode basis
 - Subspace **angle tolerance**
- Parameter variations per cluster



Topic 2 Reanalysis

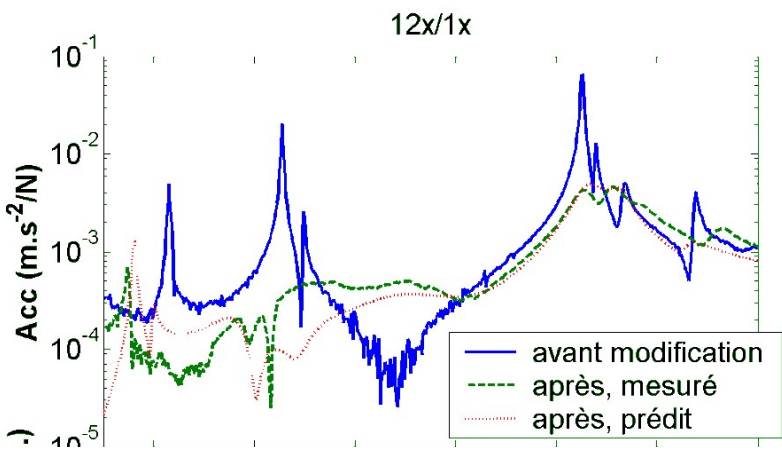
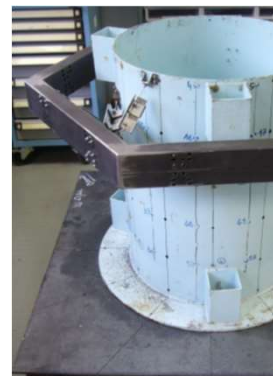
- Rather than sensitivity use "reanalysis"

Reanalysis : model 1 config -> prediction other



Test : example

- SDM : structural dynamics modification
- distributed mass, stiffness or damping modifications
 - Restricted to low frequency global modes



Reanalysis : computational example

Squeal applications

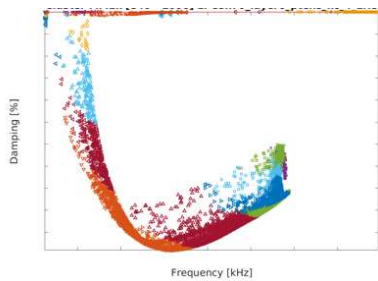
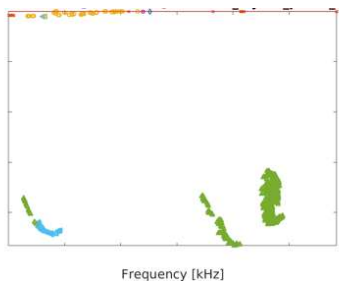
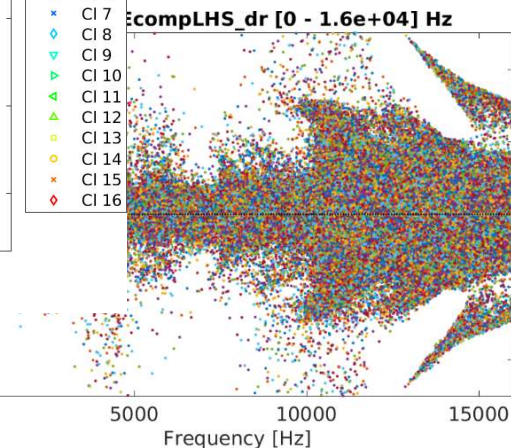
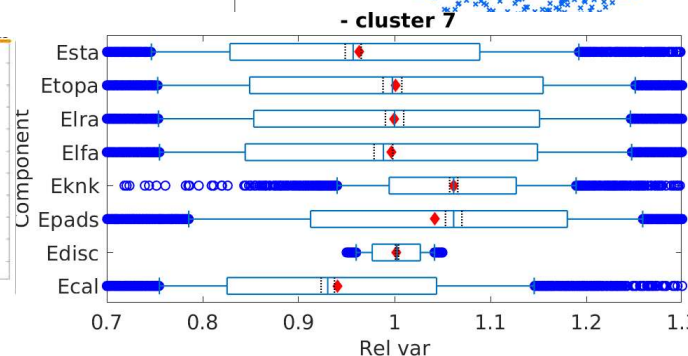
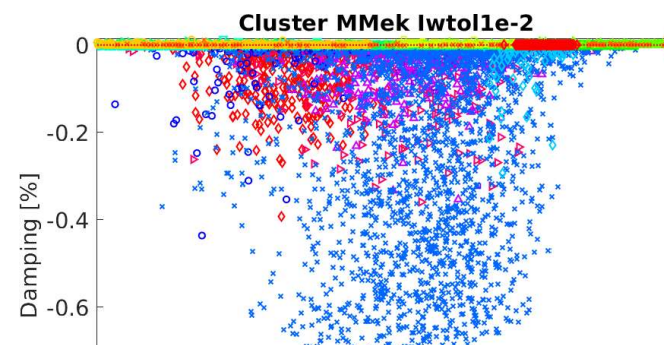
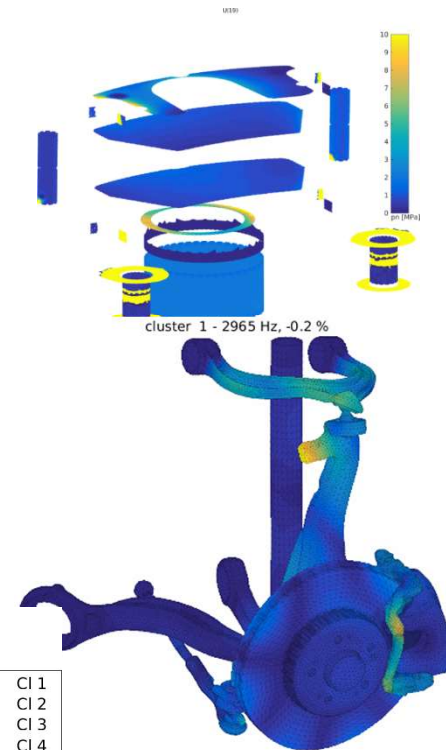
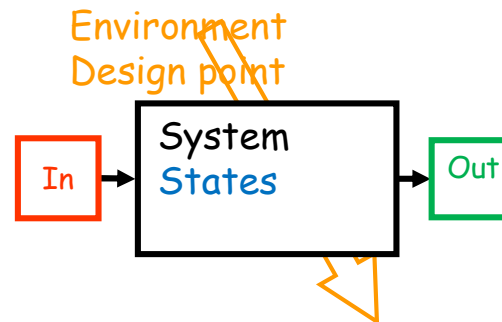
- 8-15 components
- Multiple interfaces/parameters
- 300-600 modes

Design exploration 1000 points

- Full 80 days CPU, 22 TB
- CMT a few hours off-line learning, <1h exploration

Exploitation

- DOE : basic edges, LHS
- Clustering & parameters



Families of reduced models

p physical parameters (geometry, constit.)

$Z(p,s)=[M(p)s^2+K(p)]$ finite element model

$Z_R(p,s)=T^T Z T$ reduced dynamic model

$T(p)=[\phi_1 \dots \phi_{NR} K^{-1}b]$ modal model

$\{y\}=[H]\{u\}$ transfer functions

Kinematic reduction $\Leftrightarrow$ choice fixed

Ritz basis $\{q\} = [T]\{q_R\}$

First order enhancement to MSE

$[Z(E(s),s)] \{q\} = \{F\}$ Damped viscoelastic resp. rewritten as

$$[Z(E_o,s)] \{q\} = \{F\} - \left[\sum (E(s) - E_o) / E_o [\text{Im}(Z - Z_o)] \right] \{q\}$$

Tangent system, internal loads

Basis contains

- Modes to represent nominal resonances
- Flexibility to viscoelastic loads associated with nominal modes

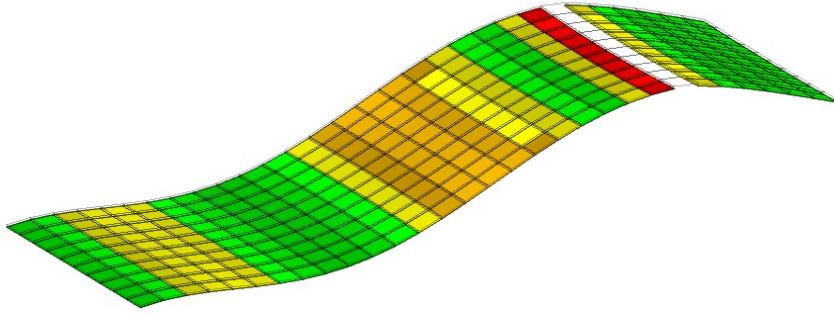
$$T = \left[\phi_{1:NM} K_o^{-1} [\text{Im}(Z - Z_o)] \phi_{1:NM} \right]$$

Modes MSE static response to unit load

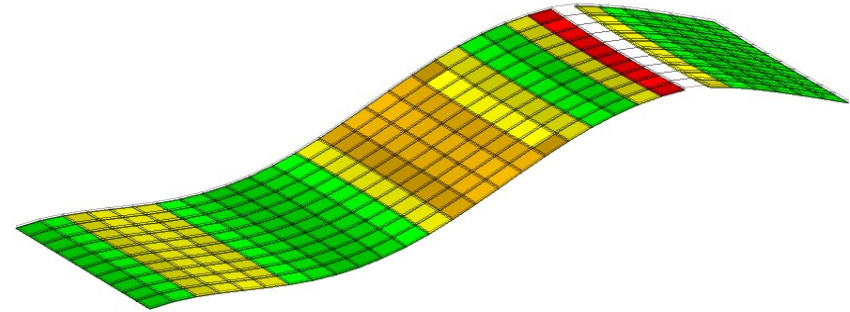
Principle of reduction
(assumptions on excitation space & freq) unchanged

Non proportional damping model

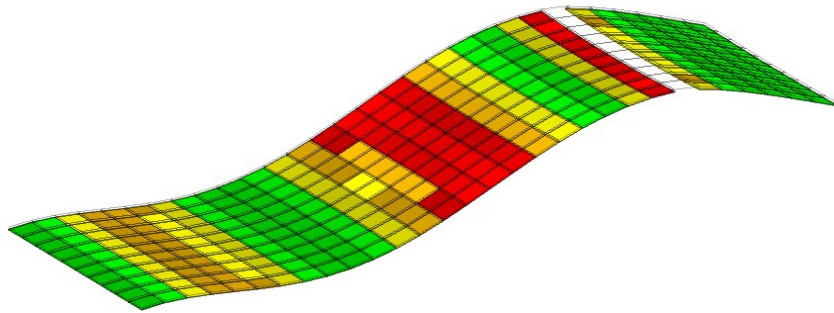
Modal



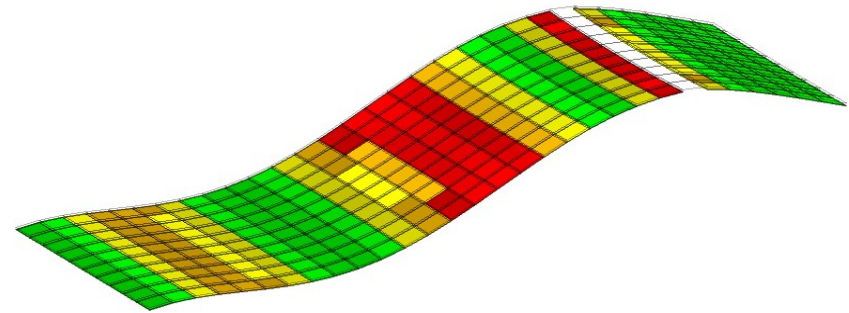
MSE



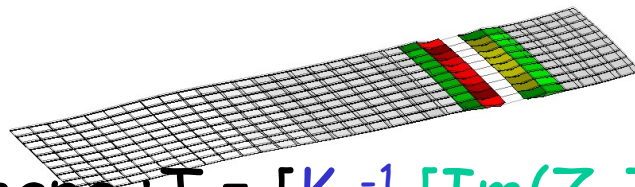
First order



Exact



• Forced response around a resonance



First order shape : $T = [K_o^{-1} [\text{Im}(Z-Z_o)] \phi_4]_{\text{orth}}$

Reanalysis

Reanalysis

Multi-model bases

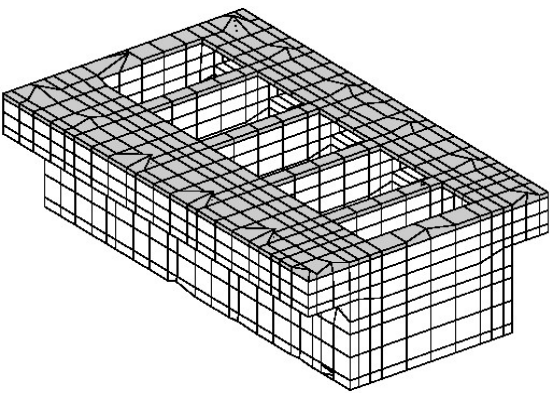
- Modes
- Multi-model
- Modes + sensitivities

$$T = [\Phi(p_0)]$$

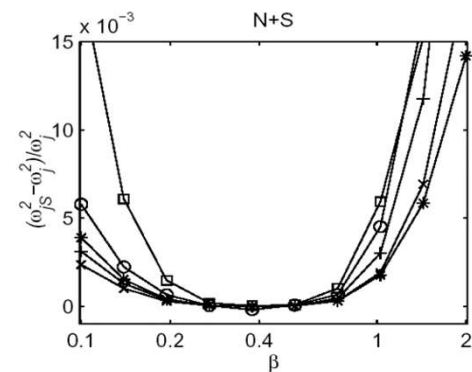
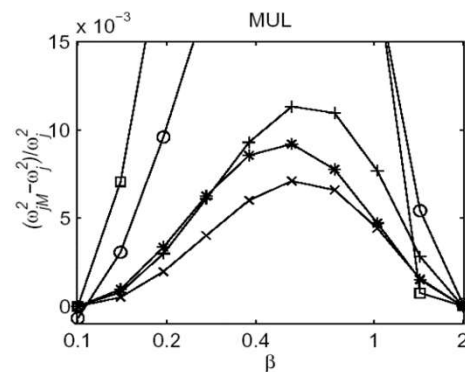
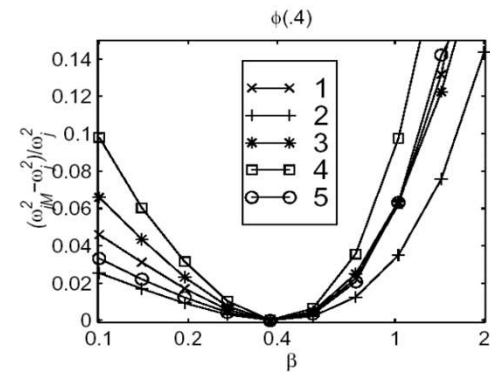
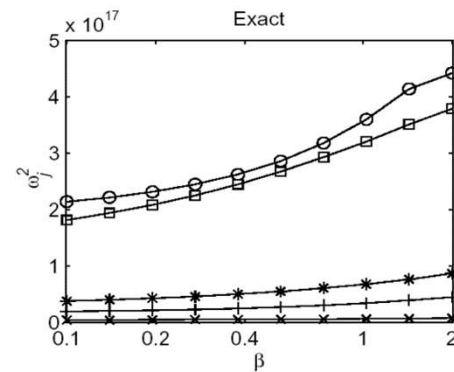
$$T = [\Phi(p_1) \ \Phi(p_2)]$$

$$T = [\Phi(p_0) \ \partial\Phi(p_0)/\partial p]$$

$$\left[\left[T^T K(p) T \right] - \omega_{jR}^2(p) \left[T^T M(p) T \right] \right] \{ \phi_{jR}(p) \} = \{ 0 \}$$

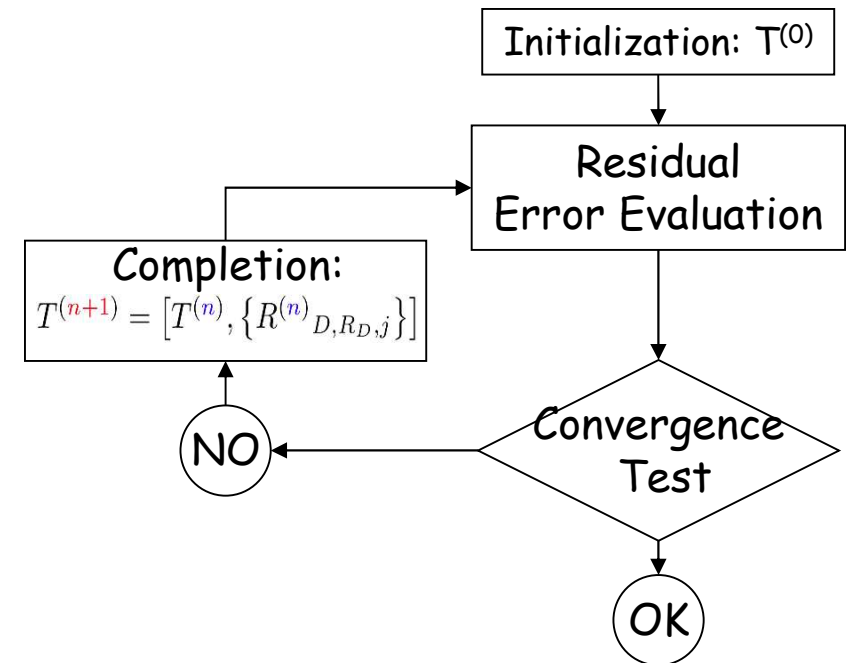
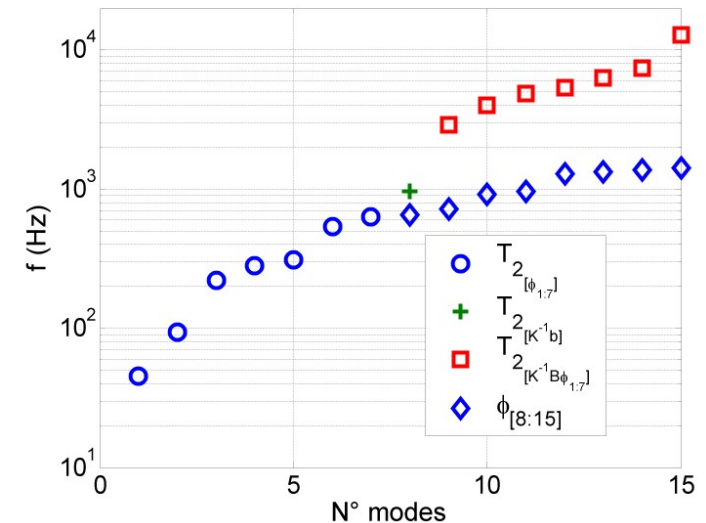


Frequencies



Reduction Bases/Starting Point

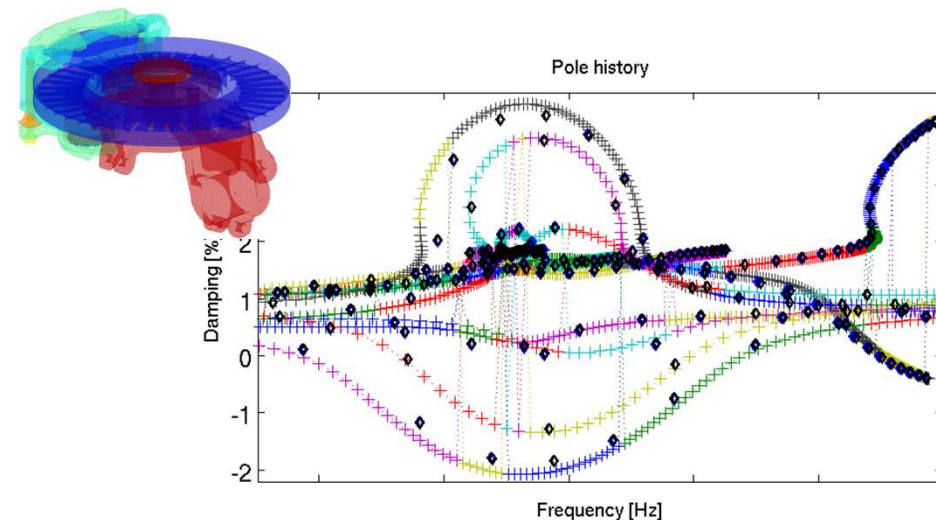
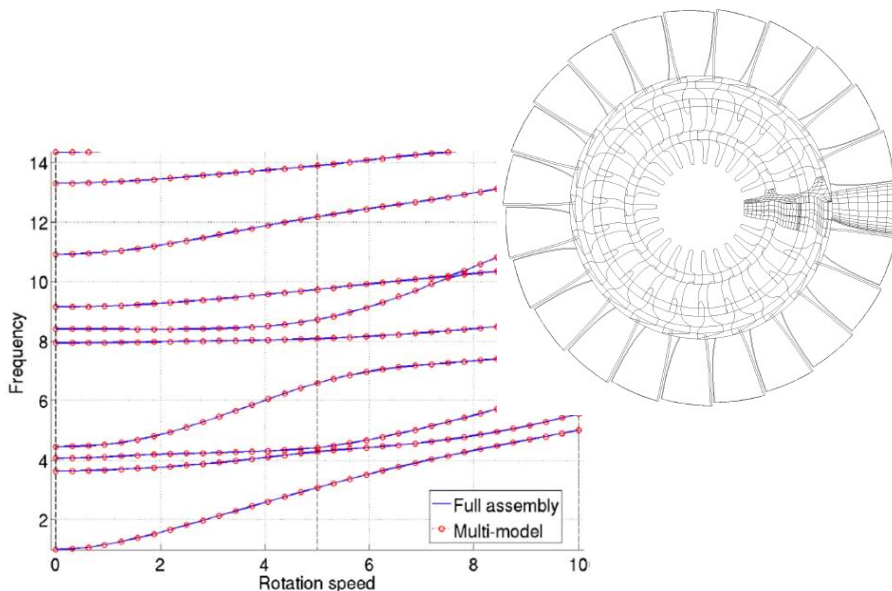
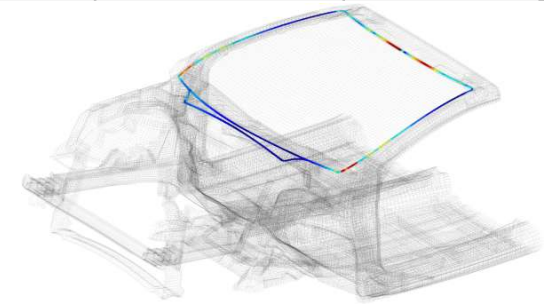
- Modal $T=[\Phi(p_0)]$
- Modal + static Responses to input $T=[\Phi(p_0) K^{-1}b]$
- Modal + static responses to representative loads
 $T=[\Phi(p_0) K^{-1}[b R_j] \Phi(p_k)]$
- Frequencies of corrections $\square$ differ from modes $\diamond$: the added information is VERY different
- Extension : residue iteration



Fixed basis : enormous cost reduction

- **Windshield joint complex modes** at 500 design points for $\frac{1}{2}$ cost of direct solver
- **Campbell diagram** : 200 rotations speeds for the cost of 4.
- **Squeal instabilities as function of pressure** : few pressures sufficient for interpolation

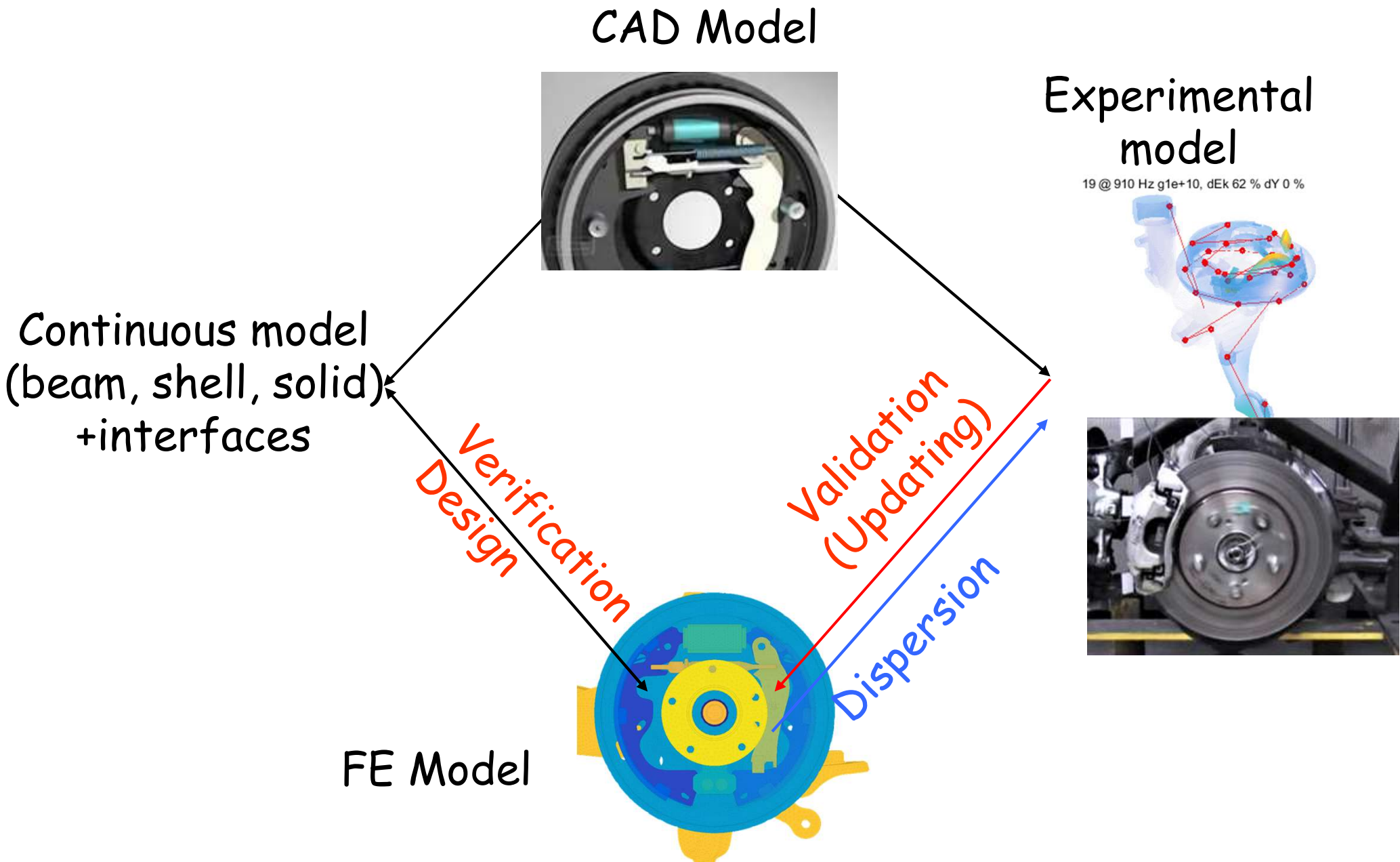
Ψ, λ	SOL107	2200s
Φ, ω	SOL103	300s
Ψ, λ Reduced	First order Error <4%	490s
$\Psi, \lambda(500 \times T)$	SOL107	~ 12 days
$\Psi, \lambda(500 \times T)$ reduced	First order Error small	~1000s



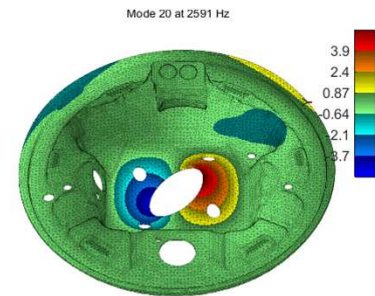
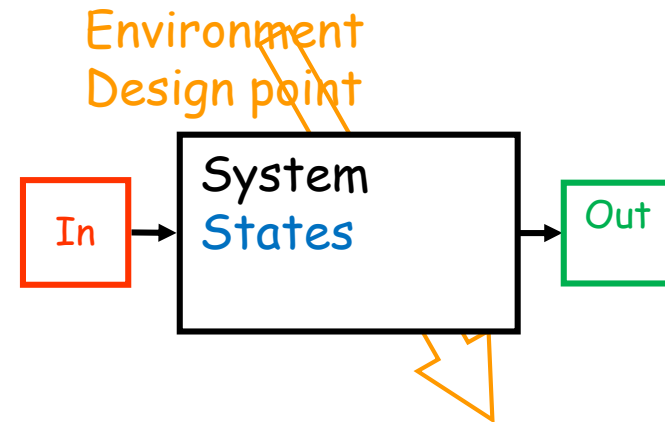
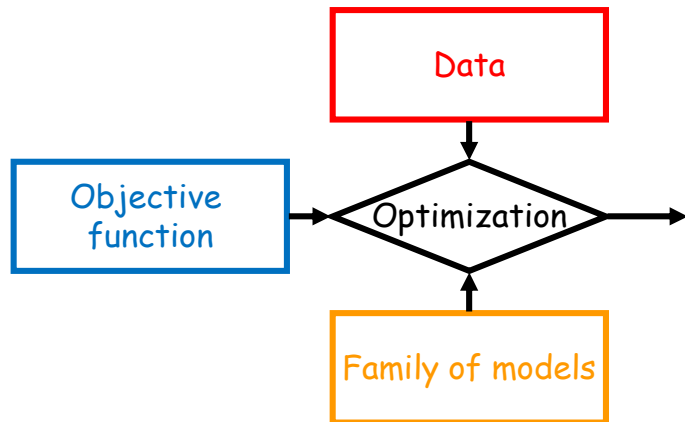
Updating

- Principles
- Matrix or parameter updating
- Objective function summary
- Parameter selection
- Optimization tools

Model validation and verification



Parametric inverse problems



FEM error

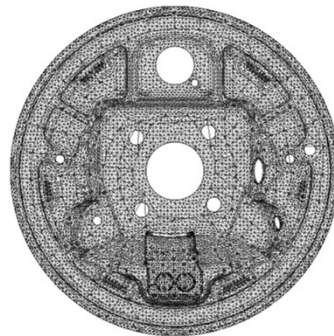
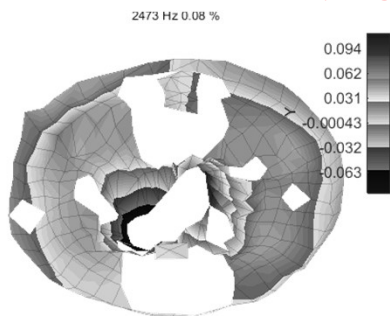
- Geometry
- Material parameters
- Multi-scale problems & equivalent parameters
- Junction representation
- Design change & modal property tracking

Identification error

- Noisy measurements
- Identification bias
- NL, time varying, ...

Topology errors

- sensor/act position
- matching

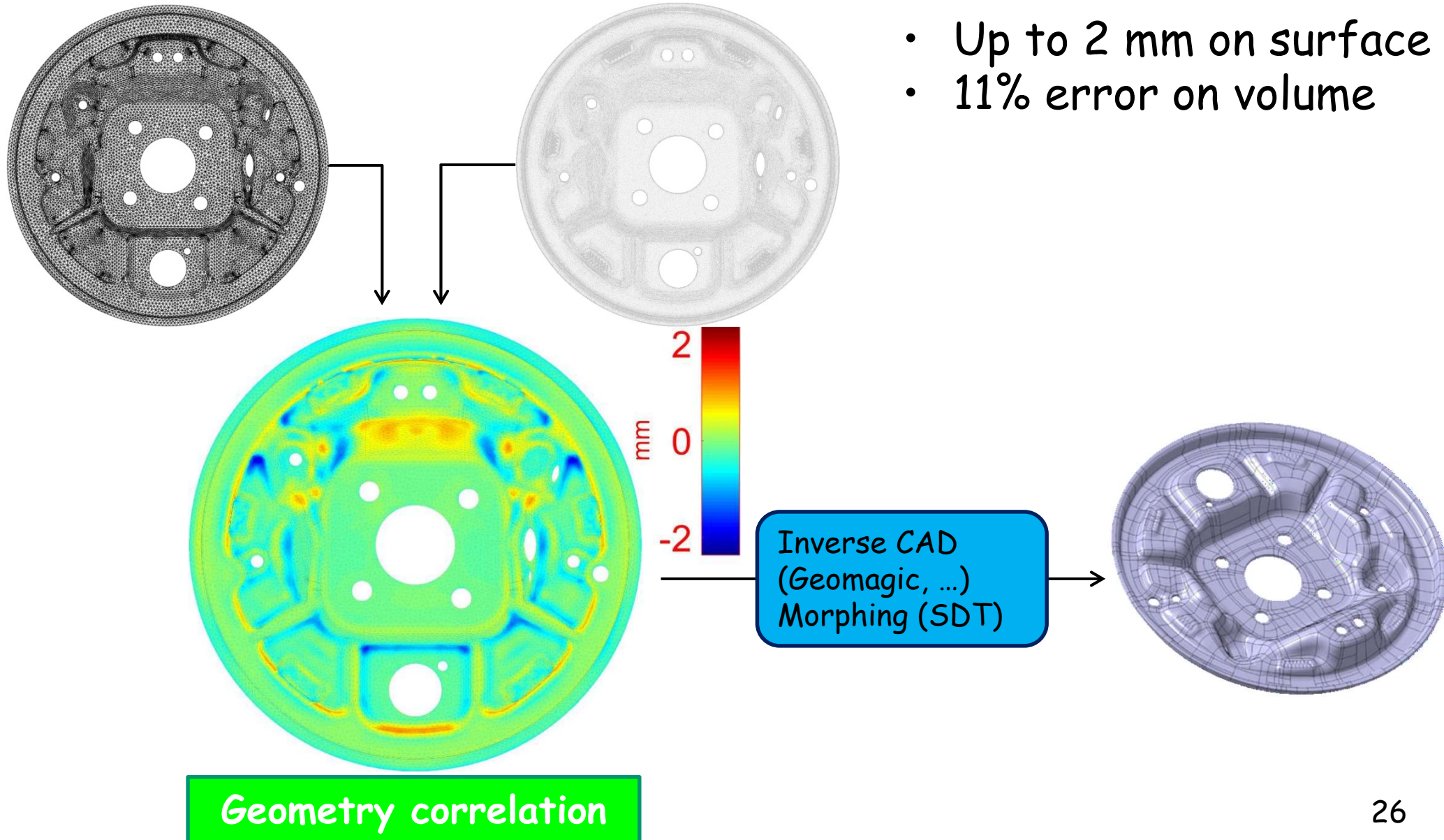


Geometry updating

Nominal geometry

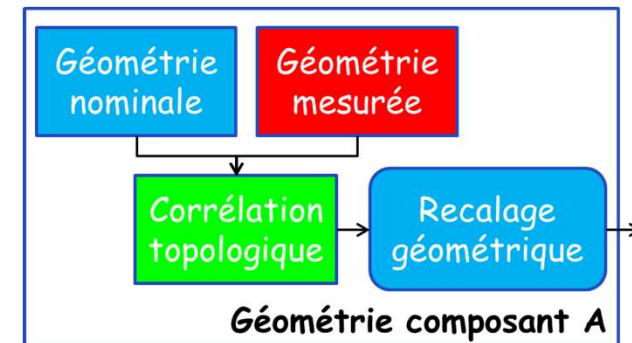
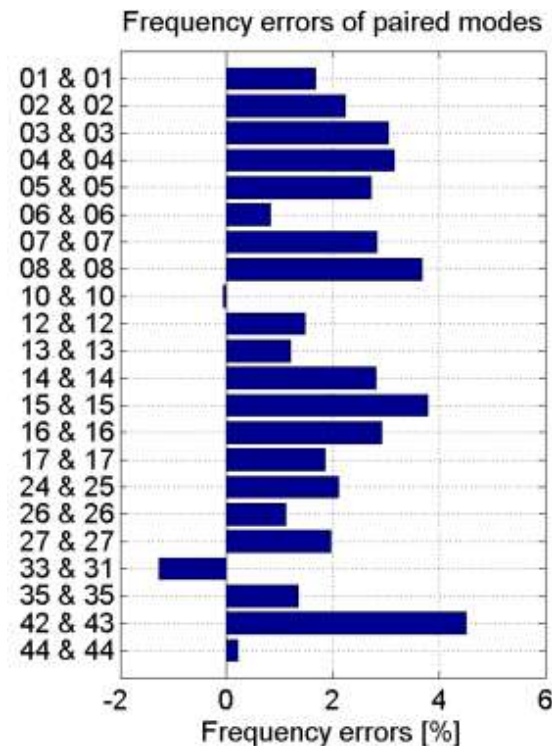
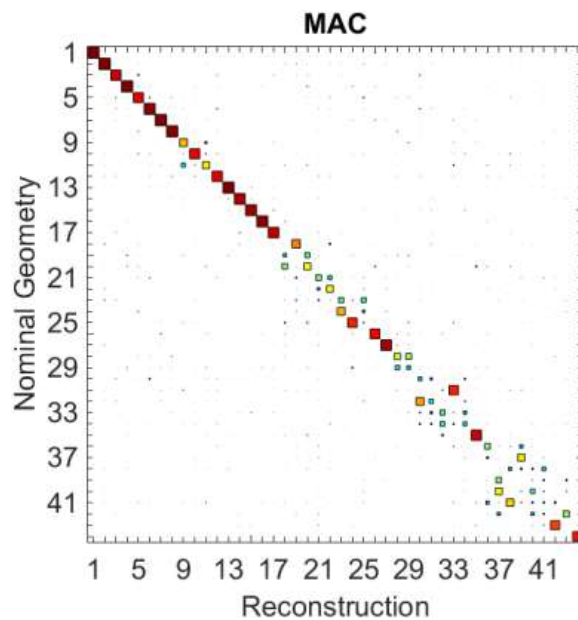
Measured geometry

- Up to 2 mm on surface
- 11% error on volume



Geometry updating

- Frequency band : 0 - 6kHz
- Notable shape difference >3 kHz
- Frequency error
 - 2 % mean
 - 5 % max



- Checking **geometry** should always be the **first step**
- Shells **strong effect** of geometry errors $\approx$ **thickness**

Invariance : influence of object

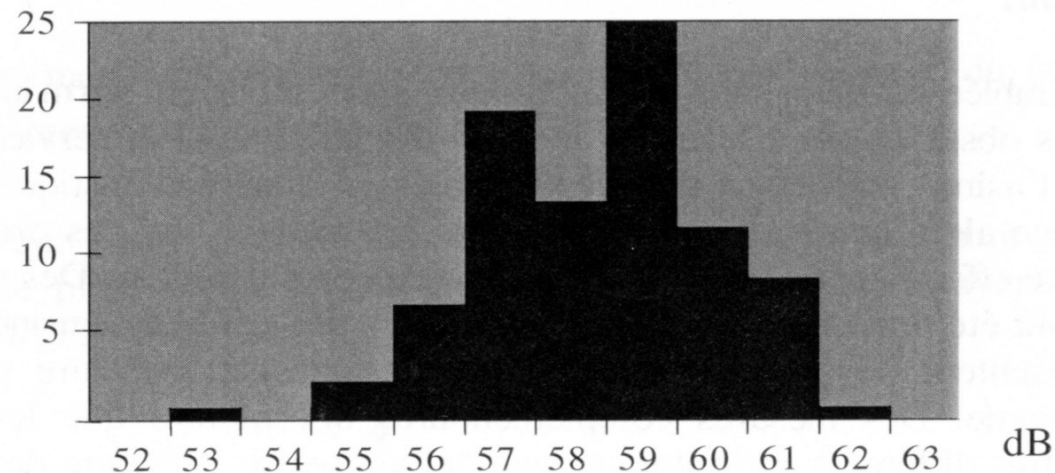


- Multiple measurements show significant variability

Spot welding manufacturing process has inherent variability due to

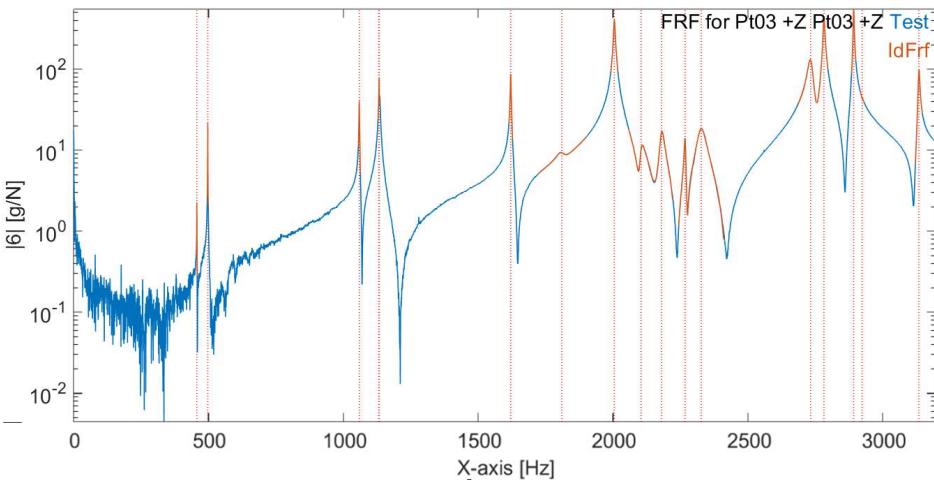
- Geometric tolerances
- Weld failure (several %)

Nombre de voitures



Updating of material properties

Modal frequencies

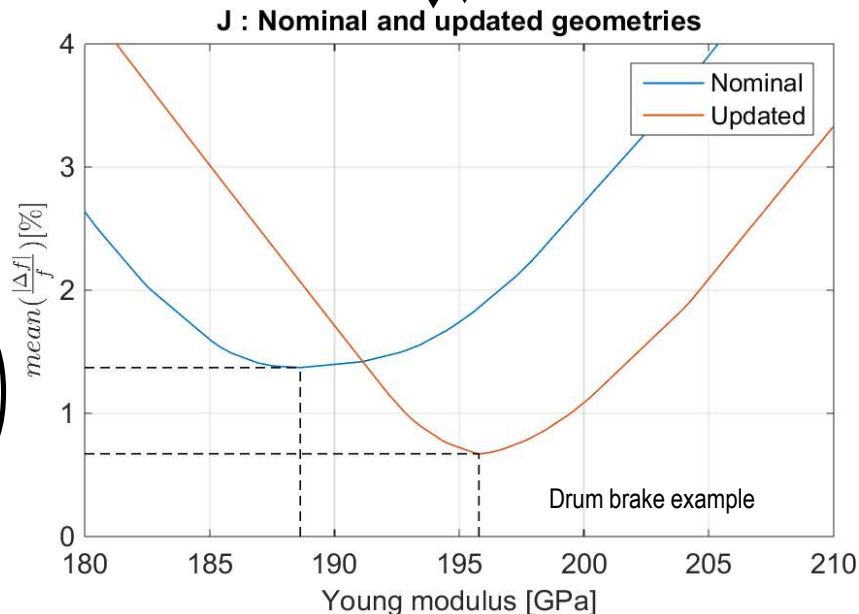


Weight

- Weight / volume $\Rightarrow$ density
- Frequencies $\Rightarrow$ modulus
- Test/FEM distance below physical dispersion ($df/f \approx 2\%$)

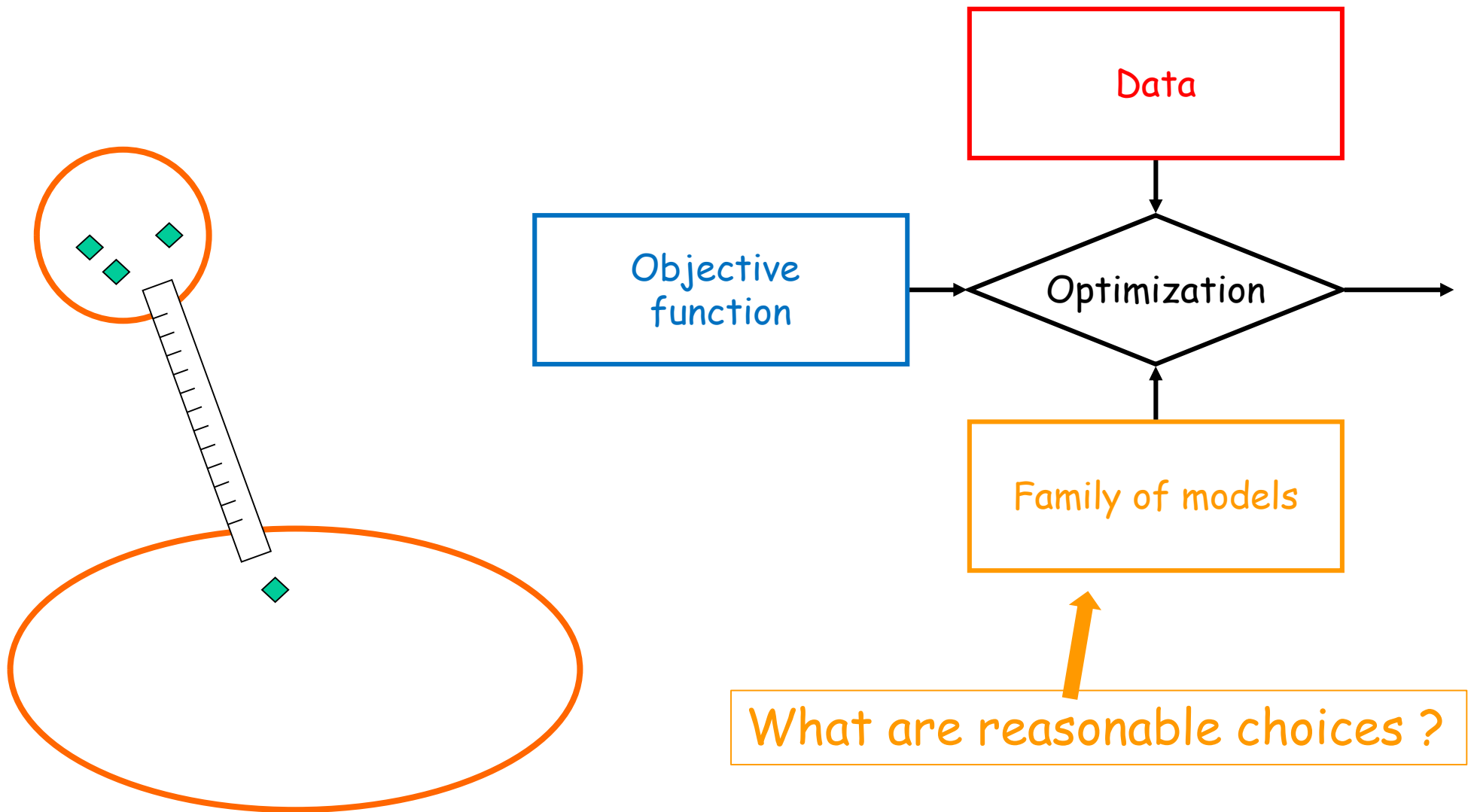
Frequency correlation

$$J = \min_E \left(\sum_i \frac{|\Delta f_i|}{f_i} \right)$$



Wrong volume = bias

Inverse problem : family of models



Equivalent materials / homogeneization

Family

- **Micro** : cell walls, glue, face-sheet, viscoelastic material
- **Macro** : shell/ **orthotropic** volume/ shell

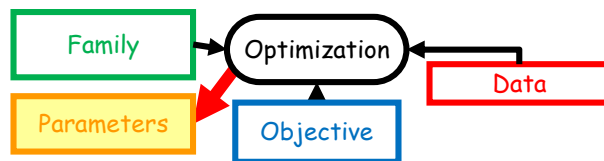
Objective

- Equivalence: waves/modes

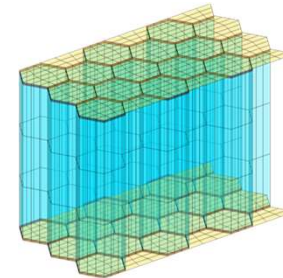
Parameters

- **orthotropic law**

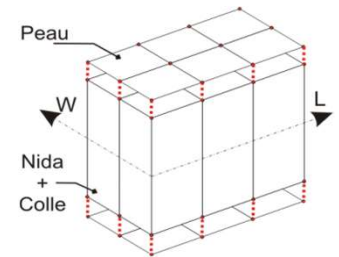
- **Scale separation**



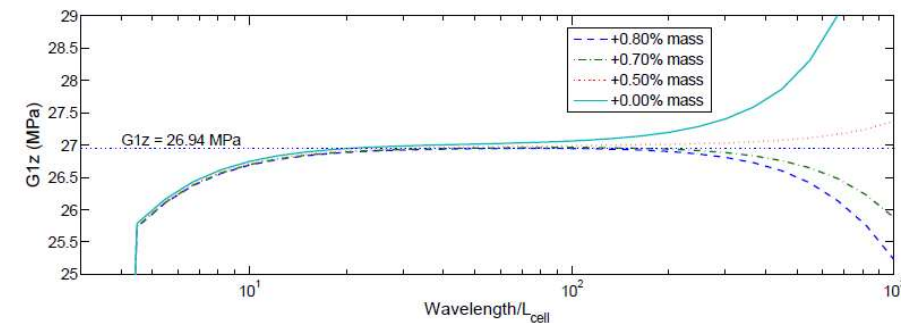
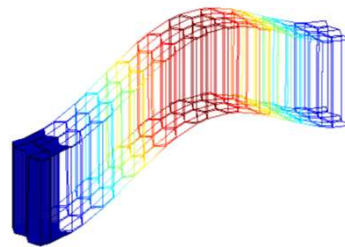
Detailed 3D
honeycomb



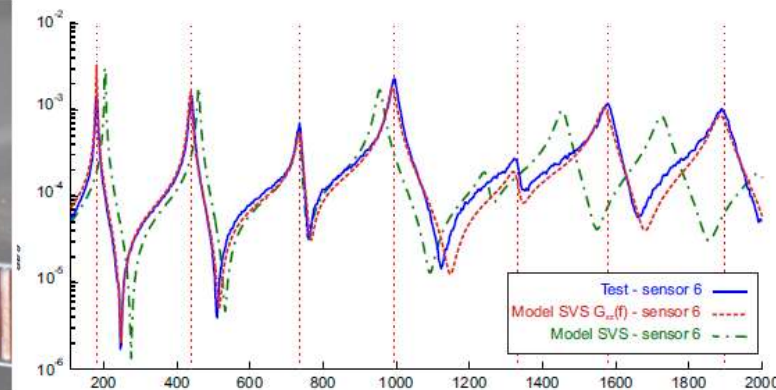
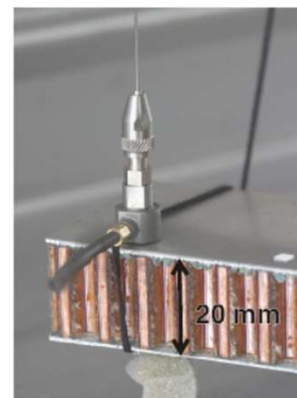
Shell/volume/shell



Numerical **homogeneization**



Updating from test

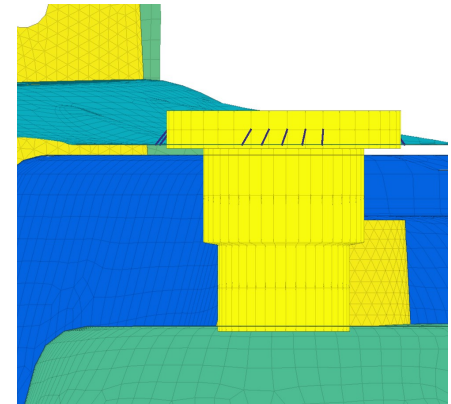
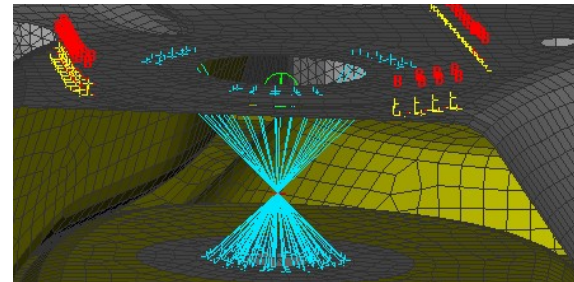
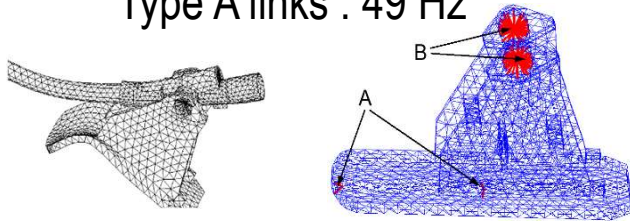


Junctions : contact / OD strategies

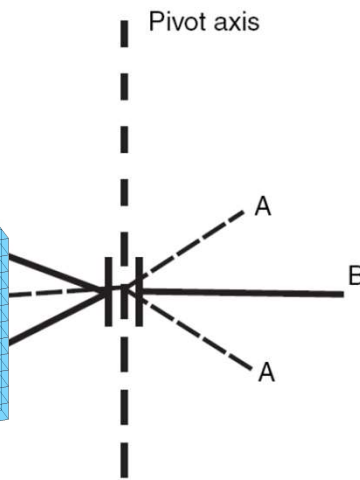
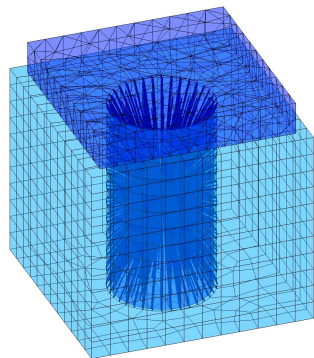
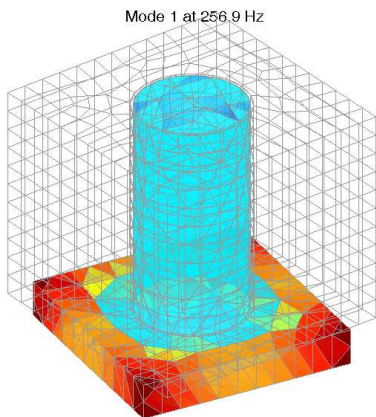
Master point : mass & stiffness error

Type B links : 82 Hz

Type A links : 49 Hz

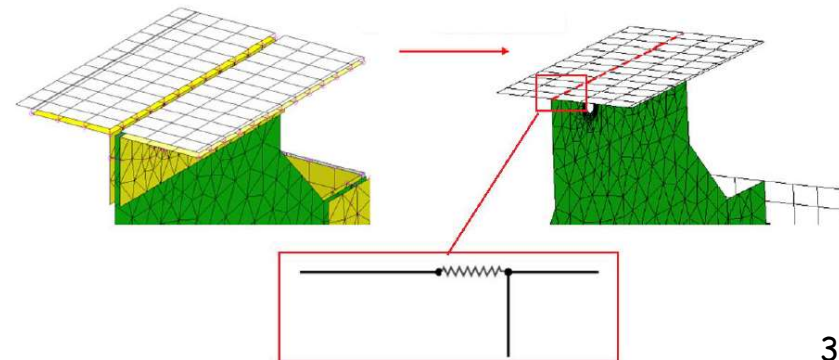


Kinematic junctions (pivots, ...)
often difficult with 2D



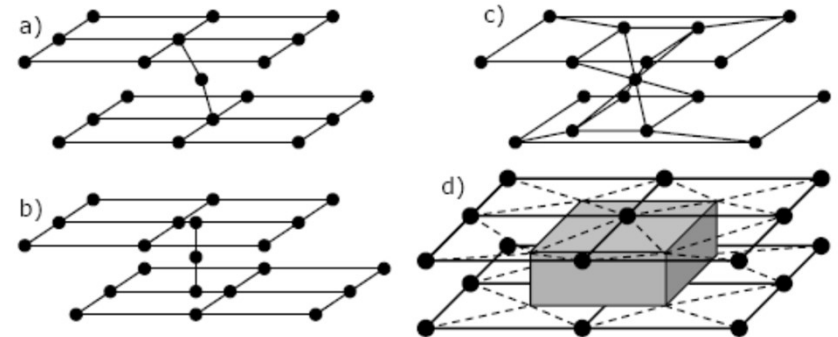
Contact : pivot mode at non zero frequency

Detail simplification (equivalent
spring) may require complex
numeric/test identification

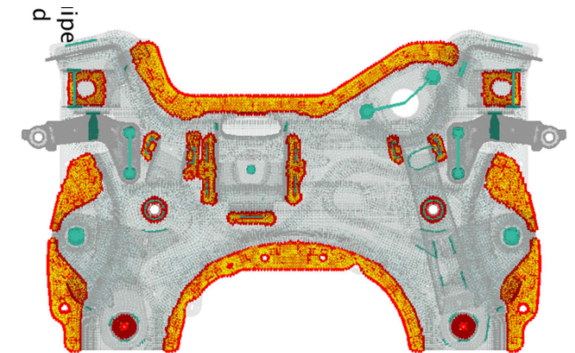


Junction problems : weld spots

- Weld spots [1]



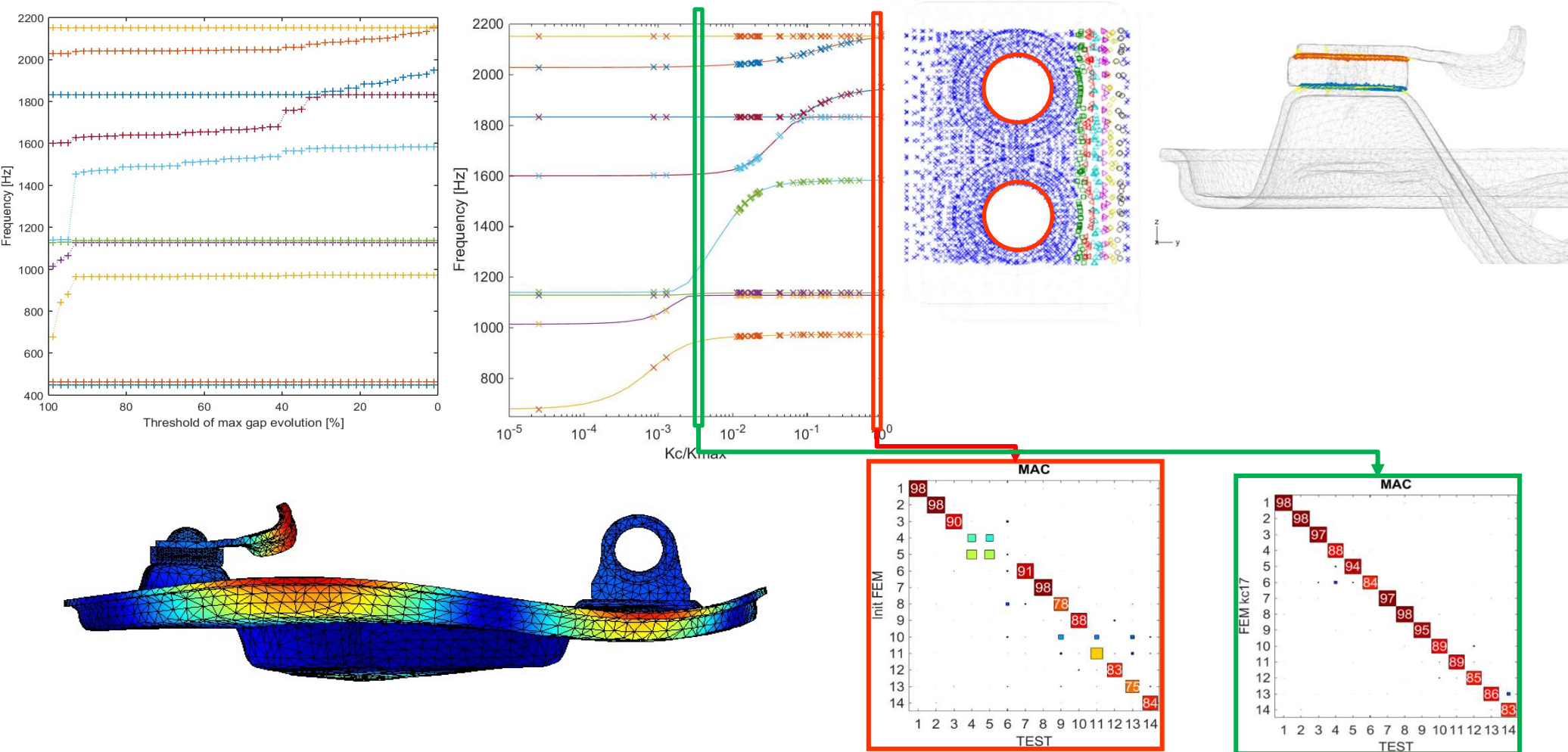
- Contact around WS [2]



- [1] Lardeur & All .: Spot weld modelling techniques and performances of finite element models for the vibrational behavior of automotive structures, ISMA 2000.
[2] G. Vermot Des Roches, E. Balmes, et S. Nacivet, « Error localization and updating of junction properties for an engine cradle model », in ISMA, 2016

Junction : equivalent contact surface/stiffness

- Variable contact **surface** and **stiffness**
- Comparable impact **frequencies** and **shapes**

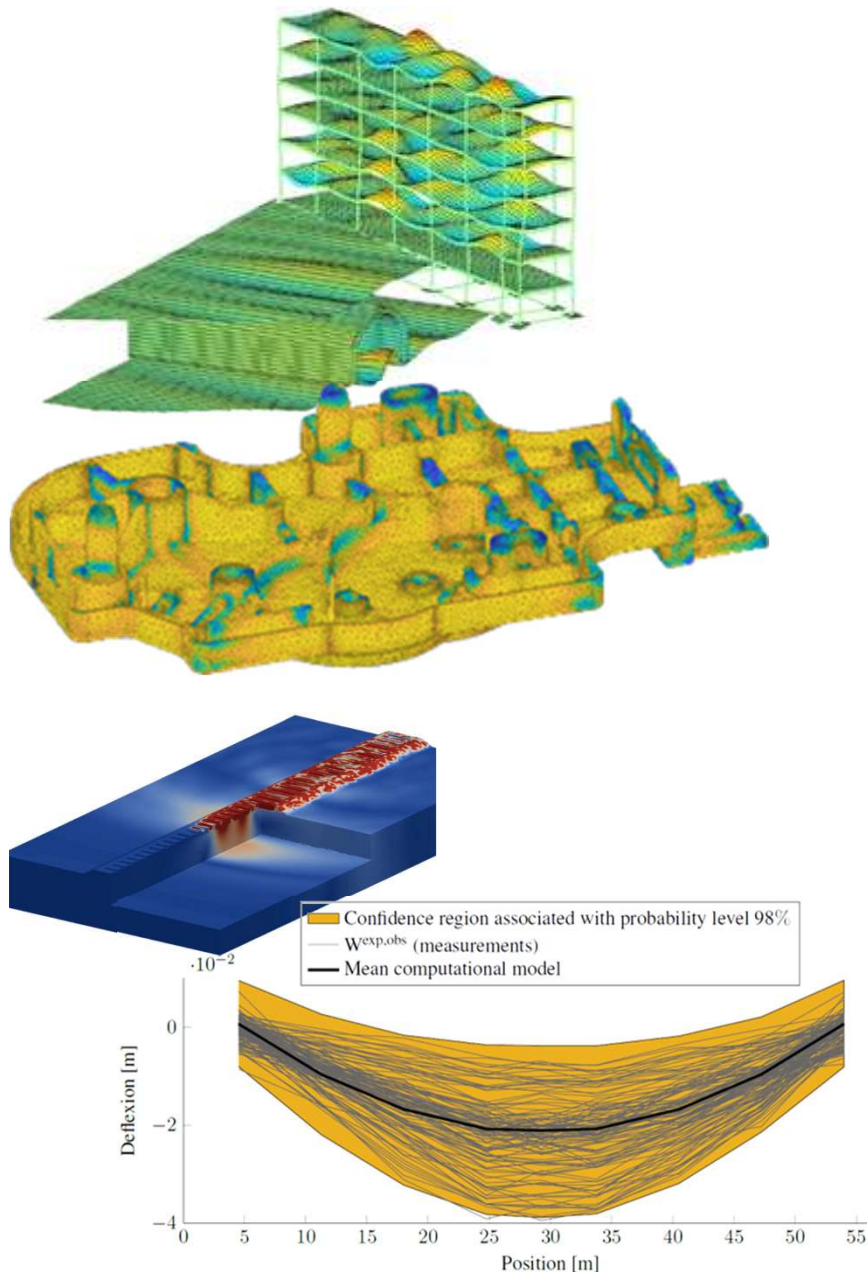


Unknown/random/variable characteristics

- Properties : soils, concrete, composite microstructure, ballast, ...
- Loads (wind, waves, boiling water, ...)

Needs

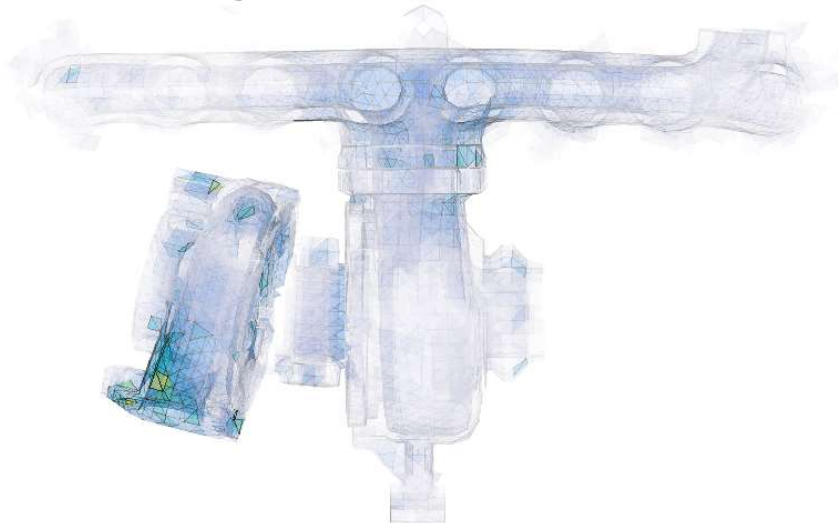
- Better knowledge (reduce uncertainties)
- A statistical description of uncertain fields
 - parametric (polynomial chaos, ...)
 - Modal (non-parametric, see **Soize & al.**)
- A propagation method : from random model to random response



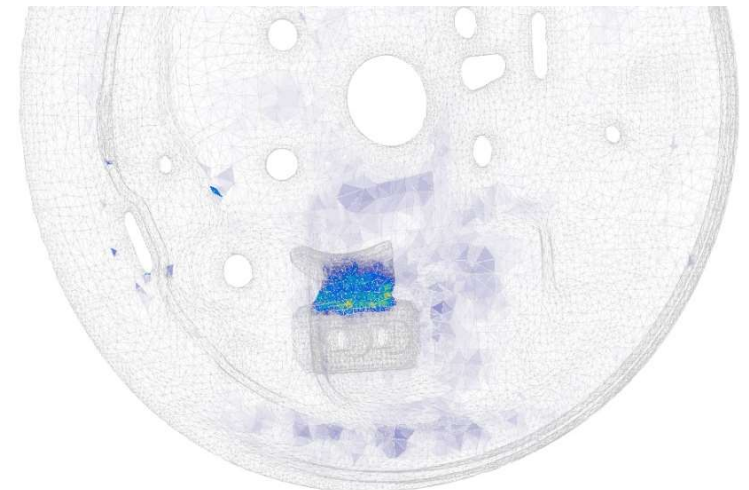
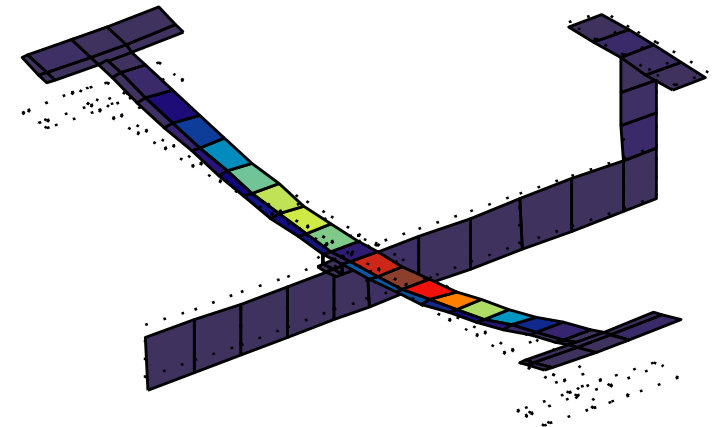
Measurement error -> localization

In MDRE expansion the energy distribution of the displacement residual ($R_d^T K^{(e)} R_d$) gives a localization

4 @ 573.4 Hz, dEk 4e-07 % dY 15.60 %



Mode 1 at 6.376 Hz



Updating & modeling errors

Usual **modeling errors** (solution : detailed model/equivalent par.)

- Incorrect model (joints, geometry, ...)
- Variability of physical parameters : geometry, properties
- Un-modeled physics (damping, contact, non-linearities material and geometric, ...)
- Discretization / mesh convergence / FEM parameter
- User input errors

Legitimate **updating parameters**

- Material parameters
- Equivalent macro-parameters : junction/contact stiffness
- Geometry

Next :

- Objectives
- Numerical tools for selection of parameters

Numerical methods for selection

- Objective is often of the form $J = ||R||^2$
example frequency sensitivity
- Main effects
 - Parameter scaling
 - Parameter equivalence (poor conditionning)
 - Saturation (range where parameter is insensitive)
 - Parameter grouping (element divided in 2 $\Rightarrow$ sensitivity too)

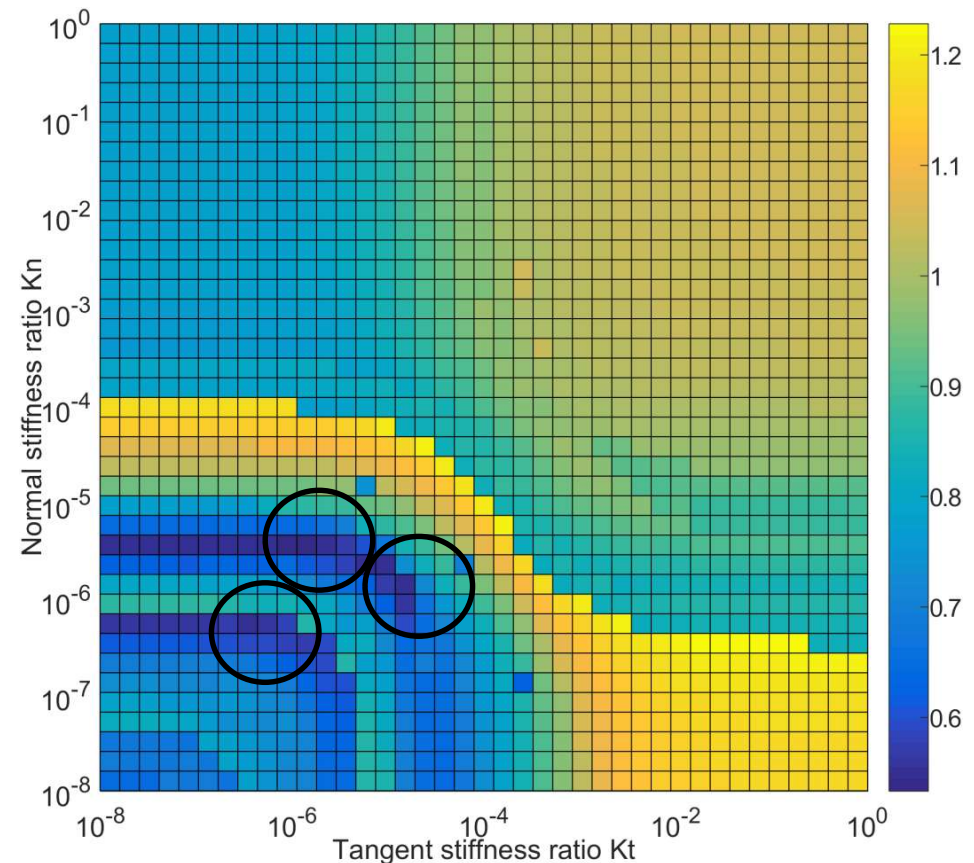
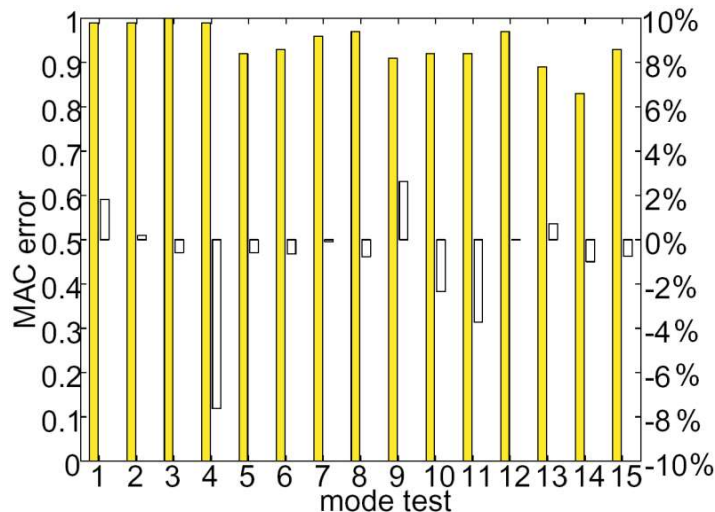
Freq-MAC objective

$$R_j(p) = \left\{ \begin{array}{c} \frac{f_{EF,i} - f_{Test,j}}{f_{Test,j}} \\ \beta(1 - MAC_{i,j}) \end{array} \right\}$$

$$J_{freq-MAC}(p) = \sqrt{\sum_{j=1}^{NM} \|R_j(p)\|^2}$$

Problems

- $J_{freq-MAC}$ irregular
- Jumps on mode pairing.
- Can be difficult to minimize



Updating based on frequency and modes shapes

$$J(p) = \sum_{j=1}^{NR} a_j (\Delta\omega_j^2)^2 + \sum_{j=1}^{NR} b_j \left(\{\Delta c\phi_j\}^T \{\Delta c\phi_j\} \right)$$

$$\begin{bmatrix} \frac{\partial\omega_j^2}{\partial p_1}(p^n) & \dots & \frac{\partial\omega_j^2}{\partial p_{N_p}}(p^n) \\ \frac{\partial c\phi_j}{\partial p_1}(p^n) & \dots & \frac{\partial c\phi_j}{\partial p_{N_p}}(p^n) \end{bmatrix} \begin{Bmatrix} \delta p_1^{n+1} \\ \vdots \\ \delta p_{N_p}^{n+1} \end{Bmatrix} = \begin{Bmatrix} \omega_{\text{id}_j}^2 - \omega_j^2(p^n) \\ c\phi_{\text{id}_j} - c\phi_j(p^n) \end{Bmatrix}$$

- Variant MAC rather than relative error
- Problems : pairing, sensitivity, numerical conditioning, systematic test errors

Non linear least squares

$$J(p) = \|R\|_2^2 = \text{Trace} (R^T R) = \sum_{i,j} \bar{R}_{ij} R_{ij}$$

- First derivative $\frac{\partial J(p)}{\partial p} = 2\text{Trace} \left(\frac{\partial R^T}{\partial p} R \right)$
- Second derivative $\frac{\partial^2 J(p)}{\partial p^2} = 2\text{Trace} \left(\frac{\partial R^T}{\partial p} \frac{\partial R}{\partial p} + R^T \frac{\partial^2 R}{\partial p^2} \right)$
- Newton method $p^{n+1} = p^n + \delta p^{n+1}$ with

- Convergence $\left[\frac{\partial^2 J(p)}{\partial p^2} \right] \{ \delta p^{n+1} \} + \left\{ \frac{\partial J(p)}{\partial p} \right\} = \{0\}$

$$\lambda_{\max} \left(\left[\frac{d^2 J}{dp^2} (p) \right] \right) < 1$$

Sensitivity method

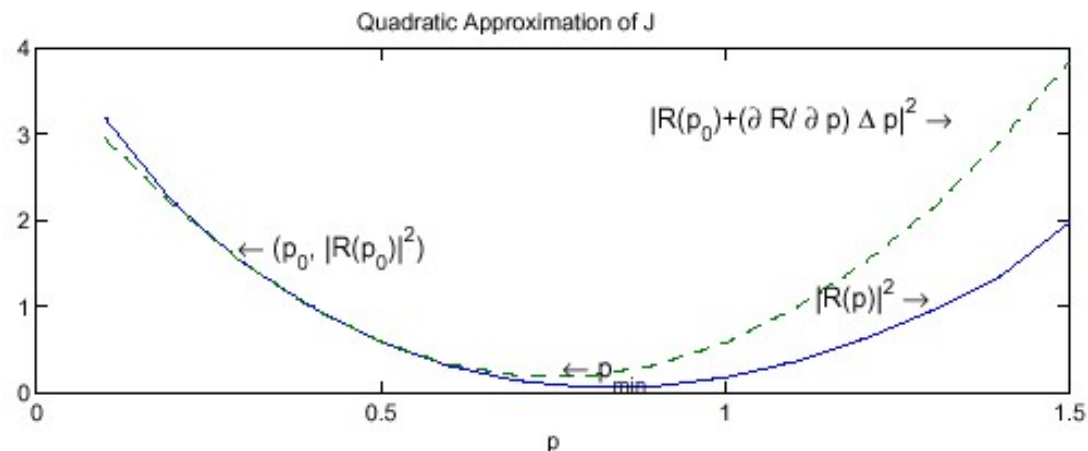
- Approximation of second order derivative $\Rightarrow$

$$\text{Trace} \left(\left\{ \frac{\partial R}{\partial p} \right\}^T \left(\frac{\partial R}{\partial p} \{ \delta p^{n+1} \} - R(p^n) \right) \right) = 0$$

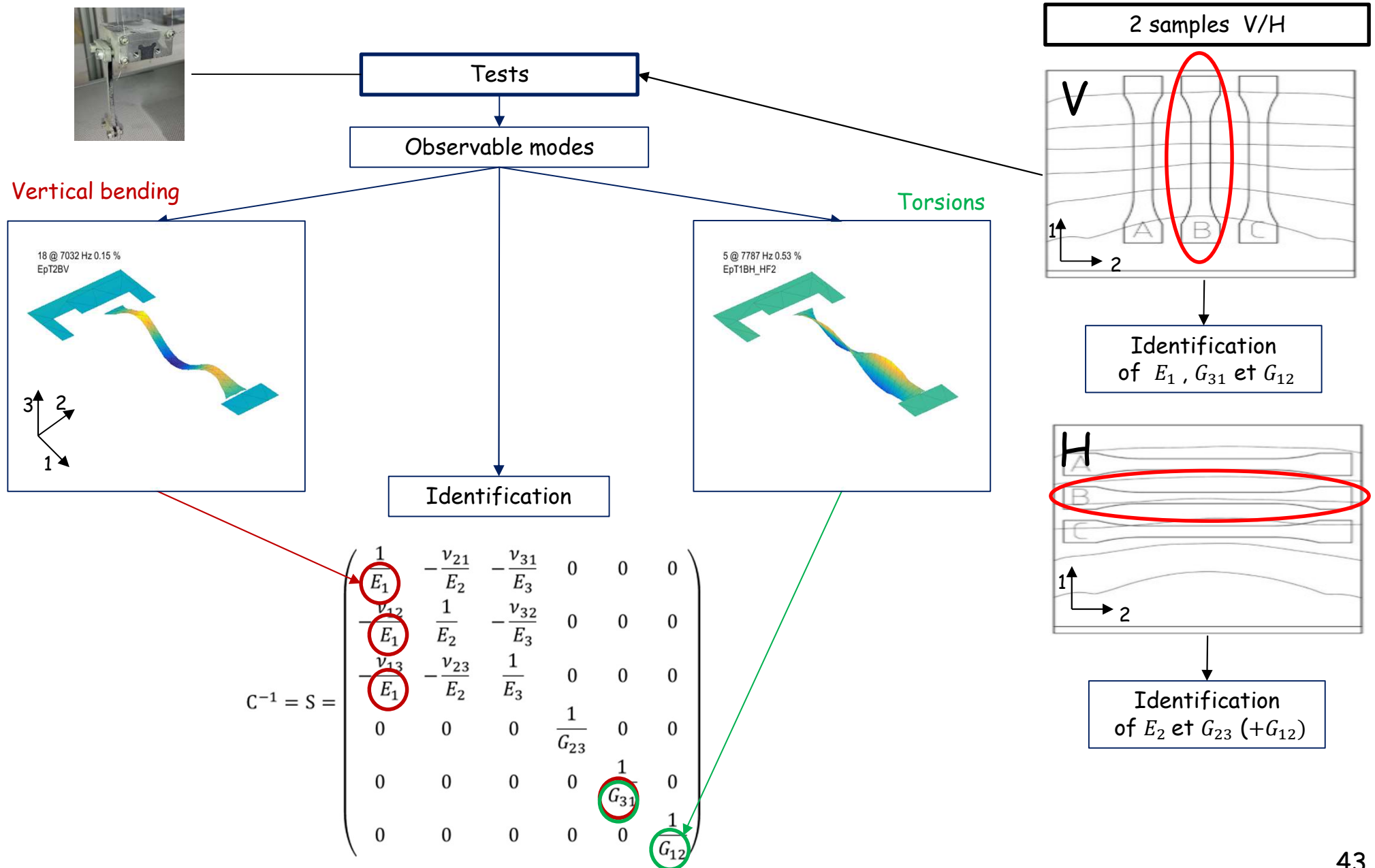
- Equivalent to (with appropriate norm)

$$\min_{\{ \delta p^{n+1} \}} \left\| \left\{ \frac{\partial R}{\partial p} \right\} \{ \delta p^{n+1} \} - \{ R(p^n) \} \right\|$$

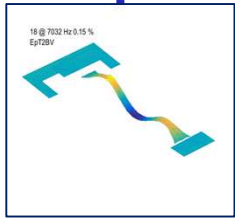
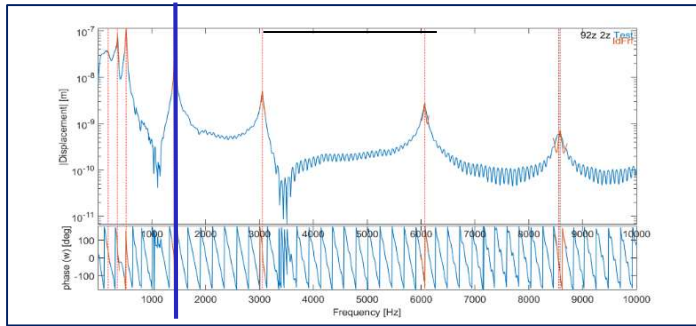
- Standard least-squares form



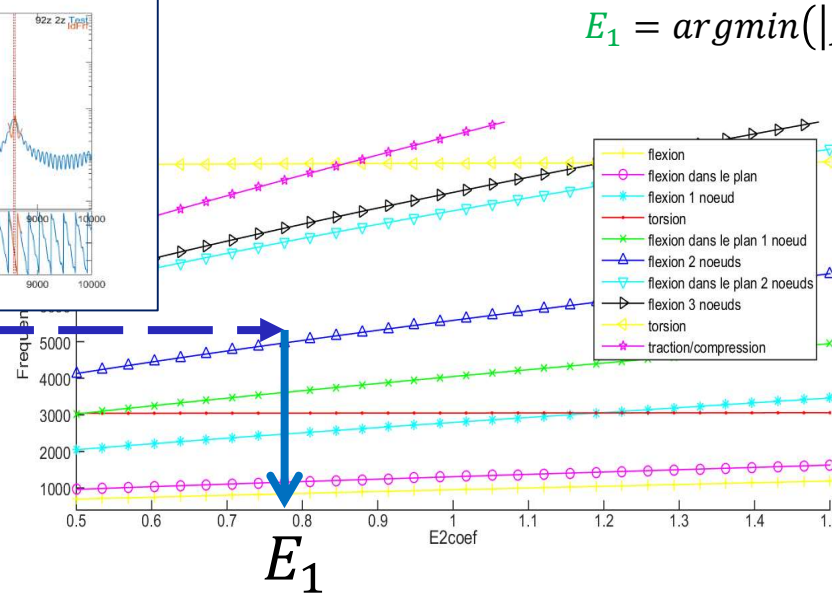
Parameter equivalence / conditioning



Parameter equivalence : material example



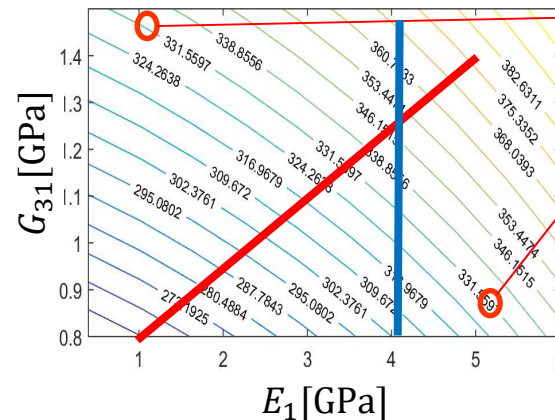
Vertical bending



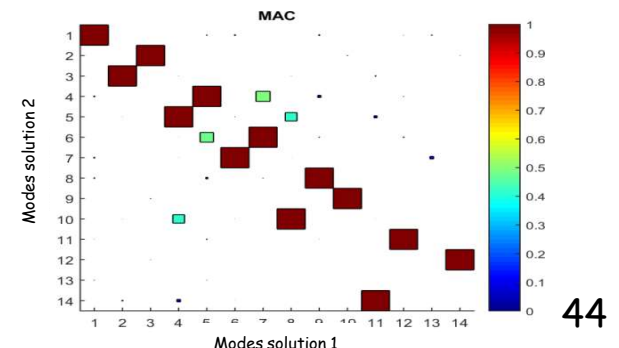
$$E_1 = \operatorname{argmin}(|f_{flexion}(E_1, G_{31}) - f_{flexion,obj}|)$$

Assumption must be made to improve conditioning

- G_{31}/E_1 constant ?
- E_1 traction ?



Solution is not unique
Mode-shape very similar



Least squares conditioning

- One solves LS

$$\min_{\{x\}} \|[A] \{x\} - [b]\|_2^2$$

- But errors

$$([A] + [\delta A]) (\{x\} + \{\delta x\}) = (\{b\} + \{\delta b\})$$

- Problem is well conditioned if :

$$\frac{\|\delta A\|}{\|A\|} \ll 1, \quad \frac{\|\delta b\|}{\|b\|} \ll 1 \quad \implies \quad \frac{\|\delta x\|}{\|x\|} \ll 1$$

- One can prove that (k condition number)

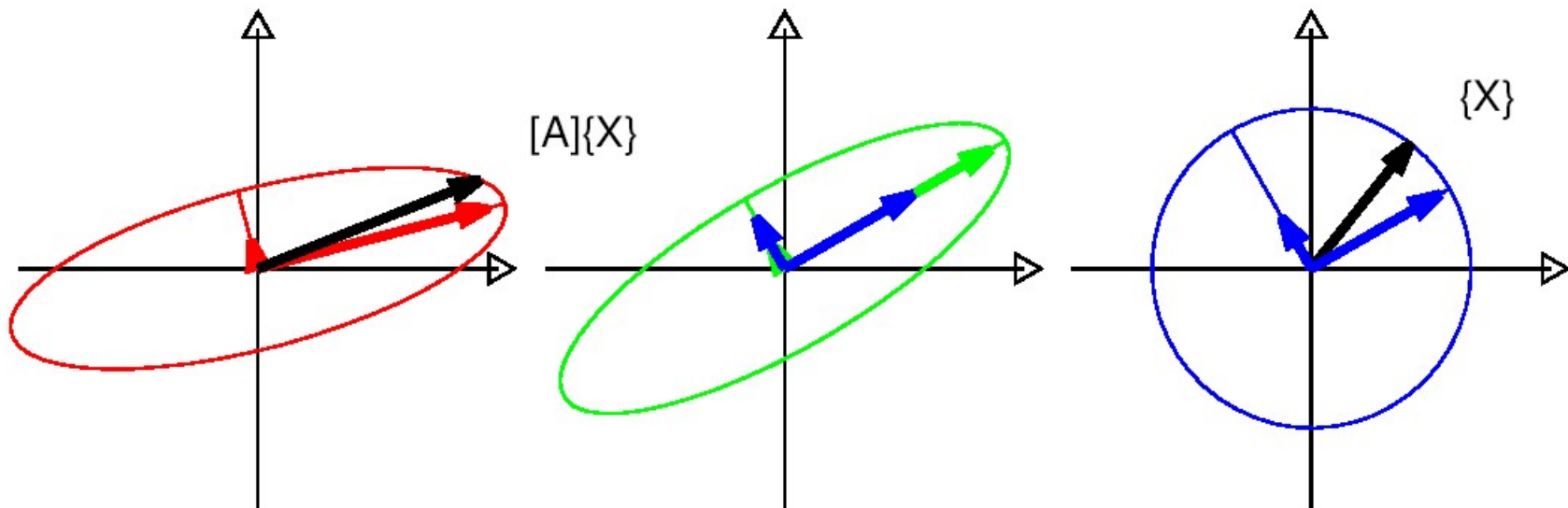
$$\frac{\|\delta x\|}{\|x\|} \leq \kappa(A) \left(\frac{\|\delta A\|}{\|A\|} + \frac{\|\delta b\|}{\|b\|} \right) \quad \kappa(A) = \|A\| \|A^{-1}\|$$

SVD (1/2)

$$A = U S V^H$$

Principal directions in the effect of operator A

$$[H H^H] \{U_j\} = \sigma_j^2 \{U_j\} \quad [H^H H] \{V_j\} = \sigma_j^2 \{V_j\}$$



SVD (2/2)

Matrix form $A = U S V^H$

Series form

$$[A] = \sum_{j=1}^{\min(n,m)} \sigma_j \{U_j\} \{V_j\}^T$$

Standard properties

$$\sigma_{\max}(A^{-1}) = \frac{1}{\sigma_{\min}(A)}$$

$$\sigma_{\min}(A^{-1}) = \frac{1}{\sigma_{\max}(A)}$$

$$\sigma_{\max}(A) - 1 \leq \sigma_{\max}(I + A) \leq \sigma_{\max}(A) + 1$$

$$\sigma_{\min}(A) - 1 \leq \sigma_{\min}(I + A) \leq \sigma_{\min}(A) + 1$$

$$\sigma_{\max}(A + B) \leq \sigma_{\max}(A) + \sigma_{\max}(B)$$

$$\sigma_{\max}(AB) \leq \sigma_{\max}(A)\sigma_{\max}(B)$$

Least squares and SVD

- SVD of A is of the form
- Least squares solution given by

$$[A] = \sum_{j=1}^{\min(n,m)} \sigma_j \{U_j\} \{V_j\}^T$$

- This is $\{x\} = \sum_{j=1}^{\min(n,m)} \sigma_j^{-1} \{V_j\} \{U_j\}^T [b] = [V] [\sigma_j^{-1}] [U] [b]$

$$\begin{bmatrix} A^+ & A & A^+ \end{bmatrix} = \begin{bmatrix} A^+ \end{bmatrix}$$

$$\begin{bmatrix} A & A^+ & A \end{bmatrix} = \begin{bmatrix} A \end{bmatrix}$$

$$\left(\begin{bmatrix} A & A^+ \end{bmatrix} \right)^T = \left(\begin{bmatrix} A & A^+ \end{bmatrix} \right)$$

$$\left(\begin{bmatrix} A^+ & A \end{bmatrix} \right)^T = \left(\begin{bmatrix} A^+ & A \end{bmatrix} \right)$$

SVD and conditioning

- For $||\delta A|| = \epsilon$ smaller eigenvalues need to be neglected (because not well known)

$$[\Sigma] = \begin{bmatrix} \sigma_1 & & & & \\ & \ddots & & & \\ & & \sigma_{NR} & & \\ & & & \epsilon & \\ & & & & \ddots \\ & & & & & \epsilon \end{bmatrix}$$

- Hence well conditioned pseudo-inverse

$$\{x\} = [A]^+ [b] = \left[\sum_{j=1}^{NR} \sigma_j^{-1} \{V_j\} \{U_j\}^T \right] [b]$$

Large scale and sensitivity saturation

- Structural parameters typically have **saturation ranges** for both low and high values
- Sensitive range needs to be determined first

