



Shapes & DOF : on the use of modal concepts in the context of parametric, non-linear studies

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ISMA / USD 2018 KU Leuven, september 17, 2018

Acknowledgements

Engineers @ SDTools

Guillaume Vermot des Roches, Jean Philippe Bianchi, Guillaume Martin, Jean -Michel Leclère

- Ph.D. students @ ECP & ENSAM
 M. Corus, A. Sternchuss C. Florens, T. Thénint, C. Hammami, E. Arlaud, O. Vo Van
- Companies & funding agencies for work shown







Modal analysis : the starting point







Direct (numerical) :

- CAD
- FEM
- Modes
- Observation
- Transfers
- Time simul.

Inverse (test):

- Expansion
- Correlation
- Modes @ sensors
- Transfers
- Time acquisition

10

city| [m/s/N]

[∞] 10⁻³

B 100³⁸⁵⁰

-100

€ o

3900

3900

3950

3950

4000

4000

4050 Frequency [Hz]



4100

4100

4150

4150

4200

4200



System modeling : the other key perspective



- Inputs u(t)
- Outputs y(t)
- Environment variables p
 - Dimensions, test piece (design point)
 - Temperature (value of constitutive law or state of thermoviscoelastic)
- State/DOF x(t)
 - Displacement & velocity field as function of time
 - $\{\dot{x}(t)\} = f(x(t), u(t), p, t)$ evolution $\{y(t)\} = g(x(t), u(t), p, t)$ observation



First bending = mass on a spring = 1 DOF

Strain energy prediction \Rightarrow FEM field many DOF

Linear modal analysis : assumptions

100

200

|H|² (dB)

Space/freq (time) decomposition



$$[H(s)] = \sum_{j=1}^{NM} \frac{\{c\psi_j\}_{NS} \{\psi_j^T b\}_{NA}}{s - \lambda_j}$$

Modeshapes

Modal amp.

Works perfectly if

• Linear, time invariant, single design

But we want

- Non-linear, stochastic
- Parametric design & environment
- Error control, ...



Learning shapes in squeal event

Time measurement during squeal







Sample time/frequency responses

Variability - influence of wheel angle

Reproductibility - Multiple events



What is expected from theory?





PhD. Vermot, 2011

Shapes, DOF, transition

- Start $\mu = 0$ small damping
- increase μ : coupling and transition towards instability





Participation of real shapes to complex modes



Subspace learning & basis selection



More details : G. Martin session SD1 Monday 17:10

Outline

System models, shapes & DOF

- Subspace learning = extract shapes & DOF
- Subspace for dynamics = modes & residuals
- Parameter loads & subspace, error control
- Basis generation (DOF selection) & objectives (complexity)
- Sparsity

Analyzing parametric models

- Coupling = mechanical root locus, S shape frequency evolution
- The parametric NL equivalence
- Modal amplitudes & shape tracking
- Local modes & damping

Component modes, interfaces & design

Measured transfers : what subspace is needed?



• Lower residual (rigid body inertia, ...)

Nearby modes = poor representation of static

FEM subspace : CMS modes+static



Craig-Bampton : applied disp. + fixed interface modes $T = \begin{bmatrix} I & 0 \\ -K_{cc}^{-1}K_{cb} & \phi_{c,1;NM} \end{bmatrix}$

Any « new » combination : static + trace of system modes

 $T = \begin{bmatrix} I \\ -K_{cc}^{-1}K_{cb} & \phi_{System} \end{bmatrix}_{c} \phi_{component}$

Issues :

- What are "applied loads" b
- Set of vectors ≠ basis
 - Cost of generating vector set (component/system)
 - Conditioning $K^{-1}b$ (two very close loads generate identical shape)



Parametric loads

cgcon

2005.00 Hz 0.00 % cgconi n2.33572e-05-cgconi t2.33572e

- Nominal model $[Ms^2 + Cs + K(p)]{q(s)} = [b_{ext}]{u_{ext}}$
- Variable contact surface, contact, sliding



- Choose nominal elastic model & rewrite using parametric load f_p $[M_0s^2 + K_0]\{q\} = [b_{ext}]\{u_{ext}\} + \{f_p(q(s), p)\}$
- Space time decomposition parametric load : $[b_{Contact}]{p(t)}$

LTI + parametric/NL loads



Parametric loads & reduction

Space/time decomposition of load $[b_{contact}]_{N \times Ng} \{p(t)\}$

- Know nothing about ${p(t)}_{Ng}$ too large
- {p(t)} associated with initial modes = $[[c_{NOR}][\phi_{1:NM}]]_{N \times NM}$ { $q_r(t)$ } Static correction for pressure load of elastic normal modes $T = [\phi(p_0) \ K^{-1}[b_c c_{NOR}\phi(p_0)]_{N \times NM}]_{\perp}$
- Multi-model learning $T = [\phi(p_1) \phi(p_2)_{N \times NM}]_{\perp}$

Error control (residue iteration)

$$R_{d} = K_{0}^{-1} \left\{ [M_{0}s^{2} + K_{0}] \{Tq_{R}\} - [b_{ext}] \{u_{ext}\} + \left\{ f_{p}(Tq_{R}, p) \right\} \right\}$$



Vector set => basis => reduced model

1. Initial (off-line) subspace learning phase

- Test : modes in band, shape due to input, shape change due to NL
- Computation : CMS= static, eigenvalue (series of static+orthog.), parametric learning/error control, Snapshot (NL transient)
- 2. Basis generation. Objectives
- Size (truncation)
- Sparsity



3. Model reduction/modal synthesis/Ritz-Galerkin $\{q(x,t)\} = [T(x)]\{q_R(t)\}$

$$Z_R(\omega, p) = T^T [Z(\omega, p)] T$$

T independent of p = parametric reduction



SVD & variants



SVD

- {X} on sphere in input space transformed in {Y}=[A]{X} ellipsoid
- Series of rank one contributions

Mode

 {φ} on unit strain energy sphere output is kinetic energy

• Singular value
$$\frac{1}{\omega_j^2} = \frac{\phi_j^T M \phi_j}{\phi_j^T K \phi_j} = 1/\text{Rayleigh quotient}$$



SVD, variants, related

Random fields Karhunen-Loeve :

- input-norm I for all DOFs
- output norm spatial correlation $C = \exp[-(|x_1 - x_2| + |y_1 - y_2|)]$

PCA Principal Component Analysis POD based on snapshot-reduction :

- input-norm I on snapshot vectors
- output norm I

Junction modes

- input-norm I for modes or contact stiffness
- output norm local stiffness

Non-linear dimensionality reduction (manifold)

More complex relation between parameters



Chung, Gutiérrez, & all, "stochastic finite element models," IJME, 2005. Kershen & al. "POD", Nonlinear dynamics, 2005 Balmes, Vermot, "Colloque assemblages 2015", + Bendhia 1-epsilon compatibility EJCM 2010 Ph.D. Olivier Vo Van 2016



Interface reduction / model size / sparsity

Craig-Bampton often sub-performant because of interfaces

 Unit motion can be redefined : interface modes Fourier, analytic polynomials, local eigenvalue 5000 -> 500 interface DOFs.

Disjoint internal DOF subsets

 $2^{e}6$ rest x 5000 Int = 74GB



Separate requirements for learning shapes :

bandwidth, inputs external & parameter truncation, sparsity

Interface reduction : wave/cyclic

Best interface reduction = learn from full system modes

- 1. Learn using wave (Floquet)/cyclic solutions
- 2. Build basis with left/right compatibility
- 3. Assemble reduced model















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Analyzing parametric models in vibration applications

- Coupling = mechanical root locus, S shape frequency evolution
- The parametric NL equivalence
- Modal amplitudes & shape tracking
- Local modes & damping
- Component modes, interfaces & design

Energy coupling in & NL system



Damping occurs in the transition





FEMTo-ST Orion Testbed



Transition

- Contact with broadband excitation level
- Friction : amplitude/normal force dependence
- Viscoelasticity : temperature/frequency dependence



ENSAM/PhD. Hammami 2014 Hammami/Balmes/Guskov, MSSP 2015



A mechanical root locus (coupling)



Related ideas : electro-mechanical coupling coefficient, pole/zero distance, modal strain energy

Following modes

Horizontal lines : global modes S curve : local cable guide mode Assembly 2400 5400 2200 5500 5000 Erednency evolution [Hz] Ledneuch [Hs] 1000 1800 .gcon 1787.37Hz 0.00 % cgconi, n0.143845-cgcc t0.143845 1400 1500 1000 00000 1597 45 Hz 0.00 1200 1000 10⁻⁸ 10⁻⁶ 10⁻² 10⁻⁴ 10⁰ Stiffness ratio

Sensors can help

KnKt of of se

104

40-3

5500

800

000

1500

But can we track modes better? ٠

٠

٠

Coupling & modal DOF

Squeal example

- Shapes are stable
- We looked at amplitudes





- Physical domain param. load: $[M_0]{\ddot{q}} + [K_0]{q} = {f(q, \dot{q}, t, p)}$
- Modal domain:
- mass / stiff orthog. condition $\phi^T M \phi = I$, $\phi_j^T K \phi_j = \omega_j^2$
- Modal equation

$$I]\{\ddot{q}_R\} + [C]\{\dot{q}_R\} + \left[\begin{smallmatrix} \ & \omega_j^2 \\ & \ddots \\ & \end{pmatrix} \{q_R\} = \{f\}$$

• Modal amplitudes computed for nominal model

 $\{q_R\} = \begin{bmatrix} \boldsymbol{\phi_0^{-1}} \end{bmatrix} \{q\} = \begin{bmatrix} \boldsymbol{\phi_0^T} \boldsymbol{M_0} \end{bmatrix} \{q\}$

Associated concepts : force appropriation, modal filter

• Modal energies $e_j = \frac{1}{2} (\dot{q}_{jR}^2 + \omega_j^2 q_{jR}^2)$

Modal participations in ODS



- Extract shape = SVD around « resonance »
- Obtain modal amplitudes of ٠ nominal modes $[\phi_0^T M_0] \{q\}$

10⁰

F (N RMS)

10¹

10⁶

¹⁰ (۳/۲) ۲

10²

10[°]

10⁻²

ε(λ_{id})

- 1-MAC-M(ψ_{id}(F) , ψ₁(k,c))

• 1-MAC(S_{NI} (F)/F² , S, (k,c))

10



Modal energy computations

• Energy in components, energy in linear shapes



• Does the shape change in NL behavior





Local component mode in an assembly



Problems with local/global modes

Track horizontal wave & understand origin of bandgap (cell wall mode)



Damping due to local modes/non structural mass









Soize "structural fuzzy" 1987 Arnoux, A. Batou, C. Soize, L. Gagliardini. JSV 2013 Loukota, Passieux, Michon. Journal of Aircraft 2017

Coupling & design process

Squeal applications

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

Design exploration 1000 points

• Full 80 days CPU, 22 TB

Pole history

Frequency [Hz]

• CMT a few hours off-line learning, <1h exploration







A few key challenges

 Modes & DOFs = fully separate geometric complexity (N of FEM DOF) and system complexity (N reduced model DOF)

In test & FEM

- define complexity of junctions with distributed load
- exploit NL/parameter equivalence
- track using generalized coordinates (DOF with meaning)
- Navigate multi-parameter design space (offline/online, memory/CPU trade-off)
- Software trade-off
 - GUI (© train/navigate) / Script (© extend/repeat)
 - Optimized/generic, extensive/easy to manipulate











Outlook



VIBRATION SOFTWARE & CONSULTING

