

Calcul des modes non linéaires avec Manlab-4.0

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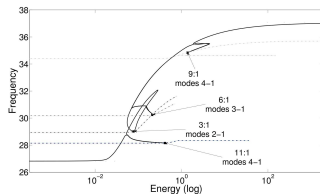
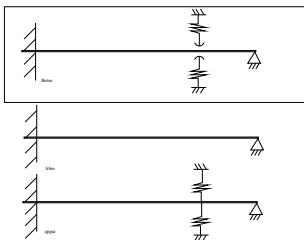
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Outline

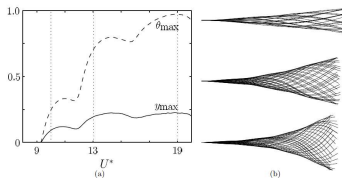
- 1 Overview
- 2 Continuation of algebraic system with Manlab-4.0
 - User point of view
 - More on the method used in Manlab-4.0

Nonlinear mode : 1-D continuum of periodic solutions

conservative systems



auto-oscillations



Aim of the talk :

Describe a method and a software to

- compute periodic solutions of dynamical systems using Harmonic Balance Method ([HBM](#)). Fourier series representation.
- do their continuation using Taylor series expansions, the so-called Asymptotic Numerical Method. ([ANM](#))
- using the interactive software MANLAB

Introduction to Harmonic Balance Method (HBM)

- We look for periodic solution to $\ddot{u} + 2\mu\dot{u} + u + u^3 = f \cos(\omega t)$ (μ and f cst)
- HBM using **one harmonic** : $u(t) = u_C \cos(\omega t) + u_S \sin(\omega t)$

$$\cos^3(x) = \frac{3}{4} \cos(x) + \frac{1}{4} \cos(3x)$$

$$\begin{aligned} & \{(1 - \omega^2)u_C + 2\mu\omega u_S + \frac{3}{4}u_C^3 + \frac{3}{4}u_C u_S^2\} \cos(\omega t) \\ \rightarrow & + \{(1 - \omega^2)u_S - 2\mu\omega u_C + \frac{3}{4}u_S^3 + \frac{3}{4}u_S u_C^2\} \sin(\omega t) \\ & + \{\dots\} \cos(3\omega t) \\ & + \{\dots\} \sin(3\omega t) = f \cos(\omega t) \end{aligned}$$

- balancing of the harmonics** yields an algebraic system with two equations

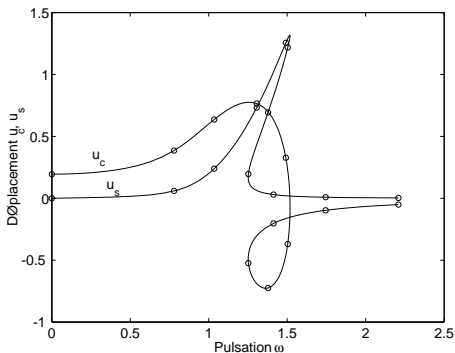
$$\begin{aligned} r_1(u_C, u_S, \omega) & := (1 - \omega^2)u_C + 2\mu\omega u_S + \frac{3}{4}u_C^3 + \frac{3}{4}u_C u_S^2 - f = 0 \\ r_2(u_C, u_S, \omega) & := (1 - \omega^2)u_S - 2\mu\omega u_C + \frac{3}{4}u_S^3 + \frac{3}{4}u_S u_C^2 = 0 \end{aligned}$$

for the two unknowns u_C, u_S and one parameters, ω .

Introduction to Taylor series continuation

- The branches of solution of an algebraic system $\mathbf{R}(\mathbf{U}) = 0$ with $\mathbf{R} \in \mathbb{R}^n$ and $\mathbf{U} = [u, \lambda] \in \mathbb{R}^{n+1}$ is determined as a succession of local Taylor series expansions with respect to a path parameter a .

$$\mathbf{U}(a) = \mathbf{U}_0 + a\mathbf{U}_1 + a^2\mathbf{U}_2 + \cdots + a^N\mathbf{U}_N$$



Summary of the method :

- Find the periodic solution of an ODE system

$$\dot{Y} = f(Y, \lambda) \quad Y(t) \in \mathbb{R}^n$$

by using a high order Fourier series expansion (HBM)

$$Y(t) = Y_0 + \sum_{k=1}^H Y_{c,k} \cos(k\omega t) + \sum_{k=1}^H Y_{s,k} \sin(k\omega t)$$

- The resulting (nonlinear) algebraic system on the Fourier coefficients reads :

$$R(U) = 0 \quad \text{with} \quad U = [Y_0, Y_{c,k}, Y_{s,k}, \omega, \lambda]$$

- Pathfollowing of the branches using high order Taylor series expansions with respect to a path parameter a .

$$U(a) = U_0 + a U_1 + a^2 + \dots + a^N U_N$$

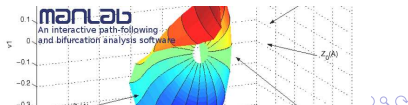
- The resulting linear algebraic system on the U_i reads :

$$\frac{\partial R}{\partial U} U_p = F_p^{nl}(U_1, \dots, U_{p-1})$$

Manlab software History

- Very first version : FORTRAN95 (2000), Matlab(2004) , no GUI !
- Manlab-1.0 on line April 2009 (written by R. Arquier).
 - first implementation of HBM in MANLAB (2009).
- Manlab-2.0 on line November 2010.
 - improved GUI, stability analysis by A. Lazarus and O. Thomas
- Diamanlab-1.0 2012
 - Automatic Differentiation with I. Charpentier
- Manlab-3.0 on line october 2014 (S. Karkar)
 - HBM,orthogonal collocation,Stability, tensor approach initiation.
- Manlab-4.0 on line may 2018 (L. Guillot)
 - Explicit definition of **auxiliary variables**
 - New sparse tensor formalism, condensation -> **efficiency (CPU/50 to /1000)**
 - treatment of elementary **function** more transparent

<http://manlab.lma.cnrs-mrs.fr>



The conerstone

- In HBM : insert $\theta(t) = \theta_0 + \sum_{k=1}^H \theta_{c,k} \cos(k\omega t) + \sum_{k=1}^H \theta_{s,k} \sin(k\omega t)$ into, for example,

$$R(\theta(t), \lambda) := \ddot{\theta} + \lambda \dot{\theta} + \theta^3 + \sin(\theta) = 0$$

and "balance" the harmonic !

- in ANM : insert $u(a) = u_0 + a u_1 + a^2 + \dots + a^N u_N$ into

$$R(u, \lambda) := u + u^2 + \frac{\tan(u)}{1 + u} - \lambda = 0$$

and collect term with the same powers !

How to :

- use Automatic Differentiation to do the job : nice for the user but poor efficiency.
- do a **quadratic recast** of the equation : then the job become easy and efficient

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Quadratic recast of algebraic systems

Exemple 1 : basique, sans fonction

Equations principales :

$$r(u, \lambda) := u + u^3 - \lambda = 0$$

Definition des variables auxiliaires :

$$v = u^2$$

Ecriture quadratique des équations principales

$$R_1 := u + u * v - \lambda = 0$$

Ecriture quadratique des équations définissant les variables auxiliaires

$$R_{aux1} := v - u^2 = 0$$

Finalement, le vecteur d'inconnues et le vecteur d'équation sont :

$$U_{tot} = [u \ \lambda \ v]^T$$

$$R_{tot} = [R_1 \ R_{aux1}]^T$$

Quadratic recast of algebraic systems

Exemple 2 : simple, sans fonction

Equations principales ($\mu = cst$) :

$$r_1(u_1, u_2, \lambda) := 2u_1 - u_2 + 100 \frac{u_1}{1+u_1+u_1^2} - \lambda = 0$$

$$r_2(u_1, u_2, \lambda) := 2u_2 - u_1 + 100 \frac{u_2}{1+u_2+u_2^2} - \lambda - \mu = 0$$

Definition des variables auxiliaires :

$$v_1 =$$

$$v_2 =$$

$$v_3 =$$

$$v_4 =$$

Quadratic recast of algebraic systems

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Definition des variables auxiliaires :

$$v_1 = 1 + u_1 + u_1^2$$

$$v_2 = 1 + u_2 + u_2^2$$

$$v_3 = \frac{1}{v_1}$$

$$v_4 = \frac{1}{v_2}$$

- All these expression are quadratic, or easily made quadratic
- "linear declaration rule" : an auxiliary variable v_i cannot appear on the left hand side before it has been explicitly defined as $v_i = f(\mathbf{U}, v_1, v_2, \dots, v_{i-1})$.

Quadratic recast of algebraic systems

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$$v_3 = \frac{1}{v_1}$$

$$v_4 = \frac{1}{v_2}$$

Ecriture quadratique des équations principales

$$R_1 := 2u_1 - u_2 + 100 u_1 * v_3 - \lambda = 0$$

$$R_2 := 2u_2 - u_1 + 100; u_2 * v_4 - \lambda - \mu = 0$$

Quadratic recast of algebraic systems

Exemple 2 : simple, sans fonction

Definition des variables auxiliaires :

$$v_1 = 1 + u_1 + u_1^2$$

$$v_2 = 1 + u_2 + u_2^2$$

$$v_3 = \frac{1}{v_1}$$

$$v_4 = \frac{1}{v_2}$$

Ecriture quadratique des équations définissant les variables auxiliaires

$$R_{aux1} := v_1 - 1 + u_1 + u_1^2 = 0$$

$$R_{aux2} := v_2 - 1 + u_2 + u_2^2 = 0$$

$$R_{aux3} := v_3 * v_1 - 1 = 0$$

$$R_{aux4} := v_4 * v_2 - 1 = 0$$

Quadratic recast of algebraic systems

Exemple 3 : un exemple mécanique simple

Equations principales :

$$r(u, \lambda) := \varepsilon * \frac{\partial \varepsilon}{\partial u} - \lambda = 0$$

Definition des variables auxiliaires :

$$eps = -\frac{u}{2} + \frac{u^2}{4}$$

Ecriture quadratique des équations principales

$$R_1 := eps * \left(-\frac{1}{2} + \frac{u}{2}\right) - \lambda = 0$$

Ecriture quadratique des équations définissant les variables auxiliaires

$$R_{aux1} := eps + \frac{u}{2} - \frac{u^2}{4} = 0$$

Finalement, le vecteur d'inconnues et le vecteur d'équation sont :

$$U_{tot} = [u \ \lambda \ eps]^T$$

$$R_{tot} = [R_1 \ R_{aux1}]^T$$

Quadratic recast with elementary function

let u and v related by

$$v = \exp(u)$$

Assume we know

$$u(a) = u_0 + a u_1 + a^2 u_2 + \cdots + a^N u_N$$

and want to compute

$$v(a) = v_0 + a v_1 + a^2 v_2 + \cdots + a^N v_N$$

this can be done easily by considering the differential

$$dv = \exp(u)du \quad \rightarrow \quad dv = v * du \quad (\text{quadratic})$$

Hence, **elementary function should appears isolated** in the list of auxiliary variable, and a **quadratic differential form** should be given also. See example 4.

Quadratic recast of algebraic systems

Exemple 6 : pour récapituler

$$r_1(u_1, u_2, \lambda) = u_1 + \lambda \frac{\exp(u_2)}{1+u_1}$$

$$r_2(u_1, u_2, \lambda) = u_2 + u_1 \tanh\left(\frac{-5u_1}{1+u_1 u_2}\right)$$

Definition des variables auxiliaires :

$$v_1 = \exp(u_2)$$

$$v_2 = \frac{v_1}{1+u_1}$$

$$v_3 = 1 + u_1 u_2$$

$$v_4 = -\frac{5u_1}{v_3}$$

$$v_5 = \tanh(v_4)$$

$$v_6 = 1 + v_5^2$$

Ecriture quadratique des équations principales

$$r_1 = u_1 + \lambda v_2$$

$$r_2 = u_2 + u_1 v_5$$

$$R_{aux1} := v_1 - \exp(u_2) = 0 \quad dR_{aux1} = dv_1 - v_1 dv_2 = 0$$

$$R_{aux2} := v_1 - v_2 - v_2 u_1 = 0$$

$$R_{aux3} := 1 + u_1 u_2 - v_3 - 3 = 0$$

$$R_{aux4} := -5u_1 - v_4 v_3 - 3 = 0$$

$$R_{aux5} := v_5 - \tanh(v_4) = 0 \quad dR_{aux5} = dv_5 - v_6 dv_4 = 0$$

$$R_{aux6} := 1 + v_5^2 - v_6$$

Demo Manlab-4.0 on algebraic system

- ECrema
- Elasticite 2D nlgeom

TP Manlab-4.0 on algebraic system

It's your turn

- exemple 3 : jambière
- exemple 5 : instabilité d'une barre en rotation

Taylor serie based continuation

Goal : determine solution branches of

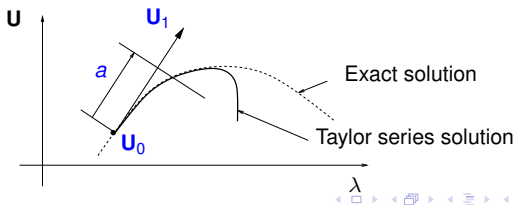
$$\mathbf{R}(\mathbf{U}) = 0 \quad \text{with} \quad \mathbf{R} \in \mathbb{R}^n \quad \text{and} \quad \mathbf{U} = [u, \lambda] \in \mathbb{R}^{n+1}$$

- let \mathbf{U}_0 be a regular point with $\|\mathbf{R}(\mathbf{U}_0)\| < \varepsilon_R$ (tolerance).
- let \mathbf{U}_1 be the tangent at \mathbf{U}_0 . ($\frac{\partial \mathbf{R}}{\partial \mathbf{U}} \mathbf{U}_1 = 0$ size : $n \times (n+1)$).
- let a be the pseudo arc length parameter $a = \mathbf{U}_1^T \cdot (\mathbf{U} - \mathbf{U}_0)$.

IFT theorem : The solution branch passing through \mathbf{U}_0 may be represented as a (truncated) Taylor series with respect to the pseudo-arclength parameter a .

$$\mathbf{U}(a) = \mathbf{U}_0 + a\mathbf{U}_1 + a^2\mathbf{U}_2 + \dots + a^N\mathbf{U}_N \quad \text{with} \quad \mathbf{U} = [U, \lambda] \quad \text{and} \quad N = 20 \text{ or } 30$$

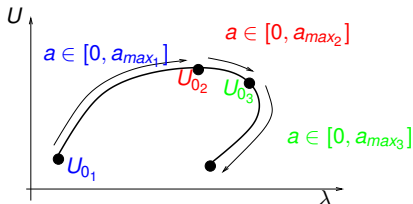
Series computation : solve a succession of linear systems that share the same stiffness matrix $\frac{\partial \mathbf{R}}{\partial \mathbf{U}} \mathbf{U}_p = F_p^{nl}(\mathbf{U}_1, \dots, \mathbf{U}_{p-1})$



The **Domain of utility** of the series is the interval $[0, a_{max}]$ for which $\|\mathbf{R}(\mathbf{U}(a))\| < \varepsilon_R$

A good approximation : $\mathbf{R}(\mathbf{U}_0 + \dots + a^N \mathbf{U}_N) \simeq \mathbf{R}(\mathbf{U}_0) + a^{N+1} \mathbf{R}^{N+1}$,

So, if one requires $\|a^{N+1} \mathbf{R}^{N+1}\| < \varepsilon_R$ then $a_{max} = \left(\frac{\varepsilon_R}{\|\mathbf{R}^{N+1}\|} \right)^{\frac{1}{N+1}}$



The complete solution branch is obtained as a succession of local Taylor series

$$\mathbf{U}(a) = \mathbf{U}_0 + a\mathbf{U}_1 + a^2\mathbf{U}_2 + \dots + a^N\mathbf{U}_N \quad \text{avec} \quad a \in [0, a_{max}]$$

- piece-wise continuous representation
- **auto-adaptative** step length \rightarrow Robustness (but small steps accumulation drawback near bifurcation)
- no algorithmic parameter

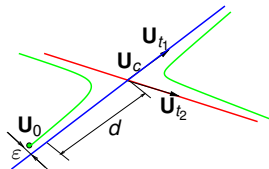
Bifurcation detection using series analysis

Numerical evidence : near a simple bifurcation, a **geometric series** emerge in the Taylor series.

$$\mathbf{U}(a) = \mathbf{U}_0 + a\mathbf{U}_1 + a^2\mathbf{U}_2 + a^3\mathbf{U}_3 + \dots$$

First order analysis : expression of a perturbed branches near a simple bifurcation [Cochelin & Médale, 2013]

$$\mathbf{U}(a) = \mathbf{U}_0 + a\mathbf{U}_{t_1} - \varepsilon \frac{\frac{a}{d}}{(1 - \frac{a}{d})} \mathbf{U}_{t_2}$$



- $\frac{\frac{a}{d}}{1 - \frac{a}{d}} = \frac{a}{d} + (\frac{a}{d})^2 + (\frac{a}{d})^3 + \dots$, a geometric serie with common ratio $\frac{1}{d}$
- After each Taylor series computation, we look for an emerging geometric series. When detected , it is extracted, completed to infinity and replaced by a fraction

$$\mathbf{U}(a) = \underbrace{\mathbf{U}_0 + a\hat{\mathbf{U}}_1 + a^2\hat{\mathbf{U}}_2 + \dots + a^{n-1}\hat{\mathbf{U}}_{n-1}}_{\hat{\mathbf{U}}(a) \text{ cleaned series}} + \frac{a}{d} \left(\frac{1}{1 - \frac{a}{d}} \right) \mathbf{U}_{scale}$$

We get d , \mathbf{U}_{t_2} and can go further the bifurcation thanks to the cleaned series.

Manlab 4.0

Let $R(\mathbf{U}) = 0$ be

$$\begin{aligned} r_1(u_1, u_2, \lambda) &= 2u_1 - u_2 + 100 \frac{u_1}{1+u_1+u_1^2} - \lambda &= 0 \\ r_2(u_1, u_2, \lambda) &= 2u_2 - u_1 + 100 \frac{u_2}{1+u_2+u_2^2} - (\lambda + \mu) &= 0 \end{aligned}$$

Definition of the auxiliary variables

$$\begin{aligned} v_1 &= 1 + u_1 + u_1 u_1 \\ v_2 &= 1 + u_2 + u_2 u_2 \\ v_3 &= 1/v_1 \\ v_4 &= 1/v_2 \end{aligned}$$

- All these expression are quadratic, or easily made quadratic
- "linear declaration rule" : an auxiliary variable v_i cannot appear on the left hand side before it has been explicitly defined as $v_i = f(\mathbf{U}, v_1, v_2, \dots, v_{i-1})$.

Let $\mathbf{U}_{aux} = [v_1, v_2, v_3, v_4]$ be the vector of auxiliary variables

Let $\mathbf{U}_{tot} = [\mathbf{U}, \mathbf{U}_{aux}]$

Manlab 4.0

The original system $R(\mathbf{U}) = 0$ is replaced by the equivalent quadratic one $\mathbf{R}(\mathbf{U}_{tot})$

$$\begin{aligned}
 r_{aux1} &:= v_1 - 1 + u_1 + u_1 * u_1 & = 0 \\
 r_{aux2} &:= v_2 - 1 + u_2 + u_2 * u_2 & = 0 \\
 r_{aux3} &:= v_3 * v_1 - 1 & = 0 \\
 r_{aux4} &:= v_4 * v_2 - 1 & = 0 \\
 r_1 &:= 2u_1 - u_2 + 100u_1v_3 - \lambda & = 0 \\
 r_2 &:= 2u_2 - u_1 + 100u_2v_3 - (\lambda + \mu) & = 0
 \end{aligned}$$

Tensor formalism : this quadratic system may be written

$$R_i = C_i + L_{ij}U_j + Q_{ijk}U_jU_k \quad i, j, k = 1, 2, \dots, n$$

with C, L, Q being tensors of order 1, 2 and 3

Here, we have 7 components C_i , 49 components L_{ij} and 343 components Q_{ijk} .

But most of them are zero !

Manlab 4.0 : Sparse tensor formalism

The sparse tensor C , L and Q are defined by the following lists (as in Matlab for a sparse matrix)

- order 1 tensor C

```
iC= [ 2  5  6 ]
vC= [-μ -1 -1 ]
```

- order 2 tensor L

```
iL= [1  1  1  2  2  2  3  3  4  4  5  6 ]
jL= [1  2  7  1  2  7  1  3  2  4  5  6 ]
vL= [2 -1 -1 -1  2  -1 -1  1 -1  1  1  1 ]
```

- order 3 tensor Q

```
iQ= [ 1  2  3  4  5  6 ]
jQ= [ 1  2  1  2  3  4 ]
kQ= [ 5  6  1  2  5  6 ]
vQ= [100 100 -1 -1  1  1 ]
```

In Manlab 4.0, these lists are **automatically generated** from the quadratic system.

Manlab 4.0 : Sparse tensor formalism

How to get the lists defining the sparse tensor, from the quadratic expression $R(X) := 0$?

Polarization formula :

$$C = R(0)$$

$$L(X) = \frac{1}{2} (R(X) - R(-X))$$

$$Q(X, Y) = \frac{1}{4} (R(X + Y) - R(X - Y) - R(Y) + R(-Y))$$

Manlab 4.0 : Sparse tensor formalism

Using these lists, the computation of the residual vector $R(\mathbf{U}) = C + L(\mathbf{U}) + Q(\mathbf{U}, \mathbf{U})$ stand in one (Matlab) line.

```
R =sparse(iC,ones(1,size(iC,2)),vC',neq,1)
+ sparse(iL,ones(1,size(iL,2)),vL'.*U(jL),neq,1)
+ sparse(iQ,ones(1,size(iQ,2)),vQ'.*(U(jQ).*U(kQ)),neq,1)
```

For the jacobian matrix $dRdU = L(.) + Q(\mathbf{U},.) + Q(.,\mathbf{U})$

```
dRdU = sparse(iL,jL,sys.vL,neq,ninc)
+ sparse(iQ,kQ,vQ'.*U(jQ),neq,ninc)
+ sparse(iQ,jQ,vQ'.*U(kQ),neq,ninc)
```

Manlab 4.0 : condensation

The linear problem to be solved at each order p reads :

$$\begin{bmatrix} B & A_{aux} \\ A & C \end{bmatrix} \begin{bmatrix} \mathbf{U} \\ \mathbf{U}_{aux} \end{bmatrix} = \begin{bmatrix} F_{aux\ p} \\ F_p \end{bmatrix} \quad \begin{array}{l} \leftarrow R_{aux} \\ \leftarrow R \end{array}$$

Thanks to the "linear declaration rule", the matrix A_{aux} is triangular which allows an easy and cheap block solving

We first solve

$$\left[A - C A_{aux}^{-1} B \right] [U] = \left[F_p - C A_{aux}^{-1} F_{aux\ p} \right]$$

where $\left[A - C A_{aux}^{-1} B \right]$ is the jacobian matrix of the original (non quadratic) system.