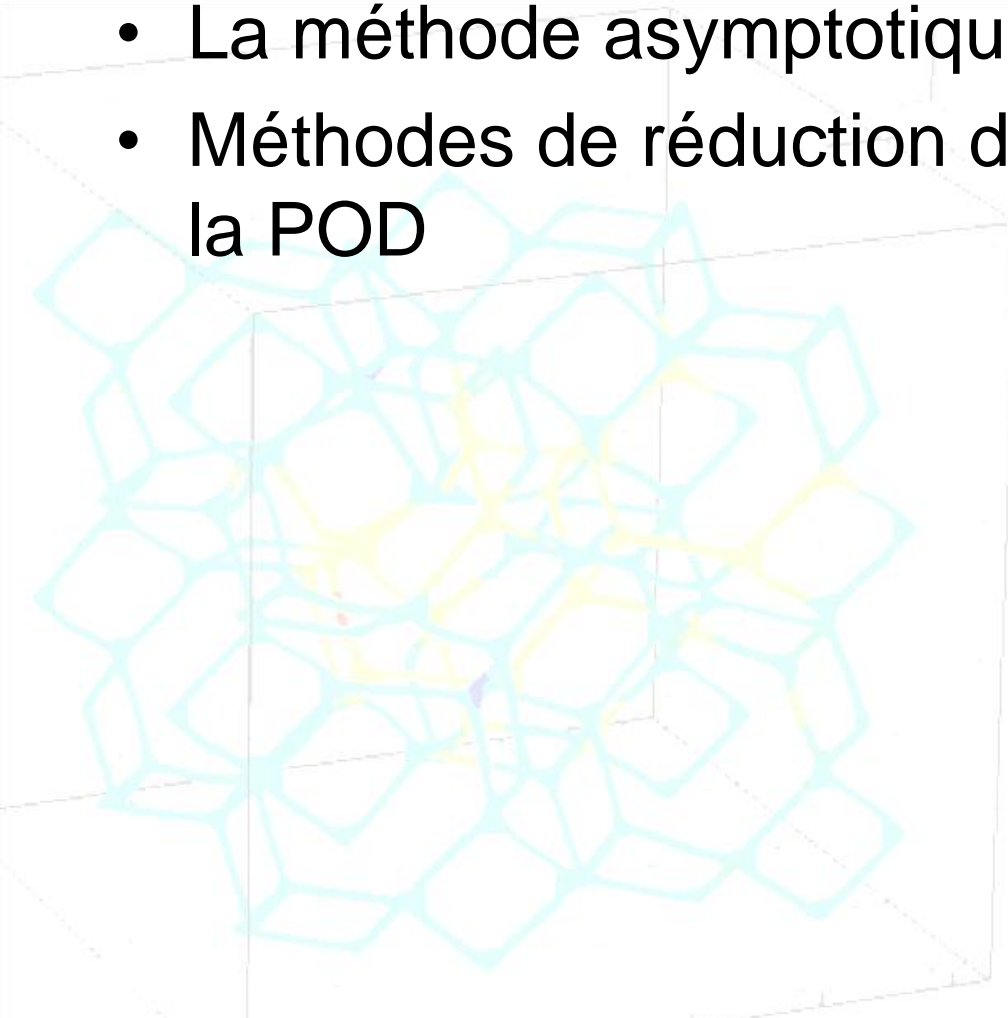




# Méthodes de résolution alternatives des problèmes non linéaires en mécanique

# Plan

- La méthode asymptotique numérique
- Méthodes de réduction de modèle basées sur la POD





# The Asymptotic Numerical Method

## Classical approaches for solving nonlinear problems in mechanics

- Finite Element method + Newton-Raphson iterative procedure

Weak form of PDE:

Find  $u \in \mathcal{G}$ ,  $\mathcal{G} = \{u / u = \bar{u} \text{ on } \partial\Omega_u, u \in W^{1,4}(\Omega)\}$  such that:

$$\int_{\Omega} S_{ij}(u) \delta E_{ij}(u, \delta u) d\Omega = \lambda \int_{\partial\Omega_F} F_i \delta u_i d\Gamma$$

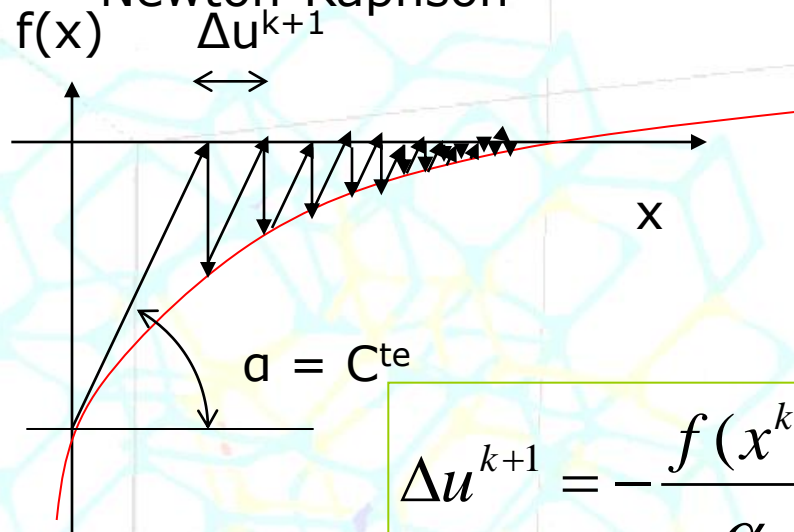
$$\forall \delta u \in \mathcal{G}_0, \mathcal{G}_0 = \{u / u = 0 \text{ on } \partial\Omega_u, u \in W^{1,4}(\Omega)\}$$

$$\delta E_{ij}(u, \delta u) = \frac{1}{2} \left( \frac{\partial(\delta u_i)}{\partial X_j} + \frac{\partial(\delta u_j)}{\partial X_i} + \frac{\partial(\delta u_j)}{\partial X_i} \frac{\partial(u_i)}{\partial X_j} + \frac{\partial(\delta u_i)}{\partial X_j} \frac{\partial(u_j)}{\partial X_i} \right)$$

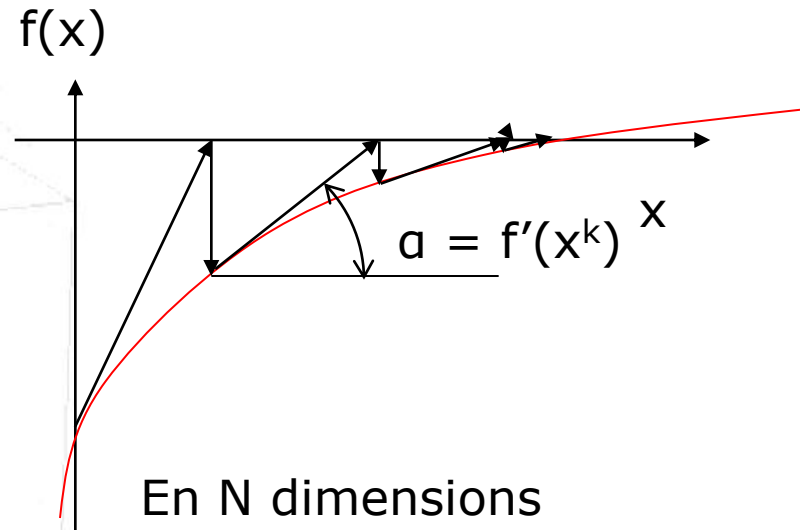
# Méthodes classiques pour la résolution itérative des équations non-linéaires

- Méthode de point fixe

- Newton-Raphson



$$\Delta u^{k+1} = -\frac{f(x^k)}{\alpha}$$



En N dimensions

$$\mathbf{K}_t(\mathbf{u}^k) \Delta \mathbf{u}^{k+1} = -\mathbf{R}(\mathbf{u}^k)$$

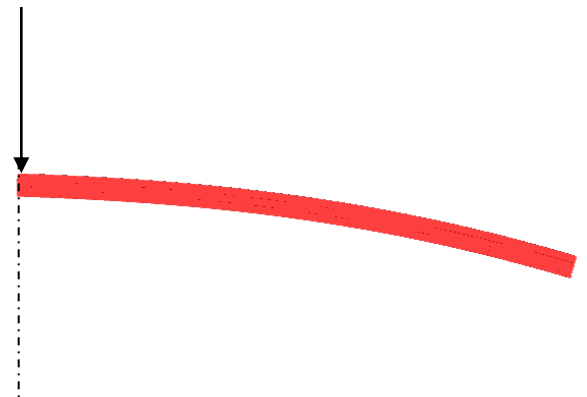
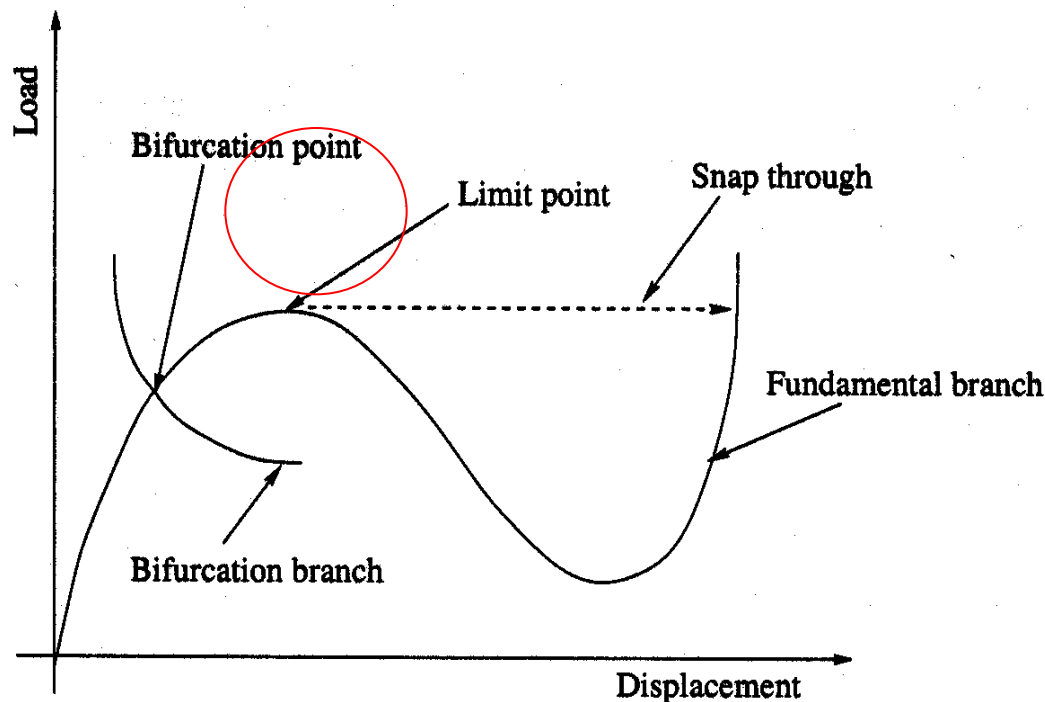
$$\mathbf{u}^{k+1} = \mathbf{u}^k + \Delta \mathbf{u}^{k+1}$$

# Issues with problems involving instabilities and bifurcation solutions

Divergence of Newton-Raphson algorithms near limit points

Solutions : arc-length control

Difficulties : choice of numerical control parameters, small iterations, detection of bifurcation points...



6

A typical (one scale) mechanical nonlinear problem: elasticity with geometric nonlinearities: nonlinear partial differential equation

$$\frac{\partial P_{ij}(u(X))}{\partial X_j} = 0 \quad \text{in } \Omega$$

P: 1st  
Piola  
Kirchhoff  
stress  
tensor

$$P_{ij} N_j = \lambda F_i \quad \text{on } \partial\Omega_F$$

$$u_i = \bar{u}_i \quad \text{on } \partial\Omega_u$$

Boundary conditions

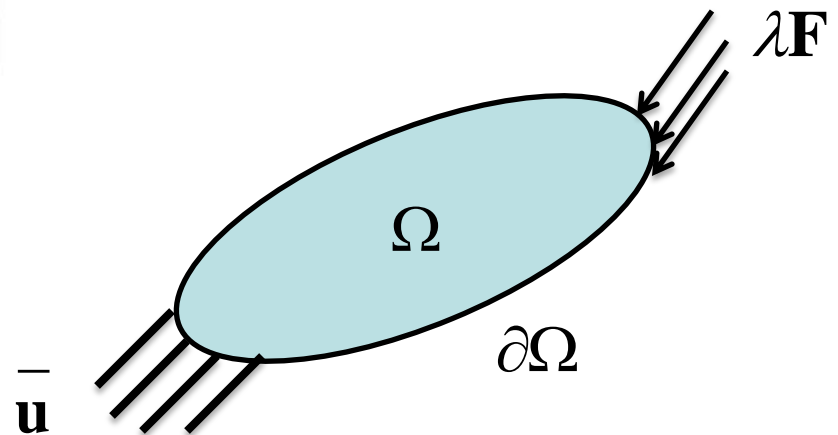
$$\partial\Omega = \partial\Omega_F \cup \partial\Omega_u, \quad \partial\Omega_F \cap \partial\Omega_u = \emptyset$$

$$P_{ij}(u) = F_{ik}(u) S_{kj}(u)$$

$$F_{ij}(u) = \delta_{ij} + \frac{\partial u_i}{\partial X_j}$$

$$S_{ij}(u) = C_{ijkl} E_{kl}(u)$$

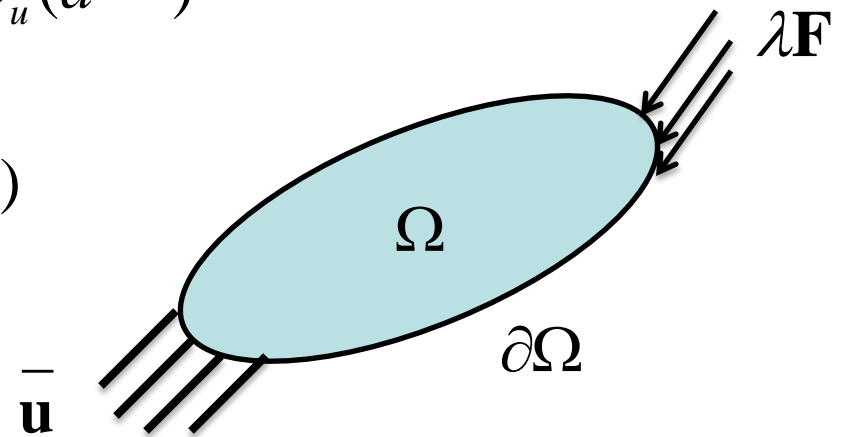
$$E_{ij}(u) = \frac{1}{2} (F_{ki}(u) F_{kj}(u) - \delta_{ij})$$



## The Asymptotic Numerical Method [Damil, Potier-Ferry, Cochelin 90-94....]

- Perturbation method (Taylor expansion) around a known solution  $u_0, \lambda_0$
- Development parameter  $a$ : related to the system evolution (in a mechanical problem: amplitude of applied force)
- Continuation procedure

$$\begin{cases} u(x, a) = u_0(x) + \sum_{i=1}^N \frac{a^i}{i!} \frac{\partial^i u(x, a)}{\partial a^i} + O_u(a^{N+1}) \\ \lambda(a) = \lambda_0 + \sum_{i=1}^N \frac{a^i}{i!} \frac{\partial^i \lambda(a)}{\partial a^i} + O_\lambda(a^{N+1}) \end{cases}$$





High order terms: not known explicitly in general

$$\begin{cases} u(x, a) = u_0(x) + \sum_{i=1}^N \frac{a^i}{i!} \frac{\partial^i u(x, a)}{\partial a^i} \\ \lambda(a) = \lambda_0 + \sum_{i=1}^N \frac{a^i}{i!} \frac{\partial^i \lambda(a)}{\partial a^i} \end{cases}$$

Defined as unknown fields  $u_i$  and  $\lambda_i$  and computed numerically (via the Finite Element method)

$$\begin{cases} u(x) = u_0(x) + \sum_{i=1}^N a^i u_i(x) \\ \lambda = \lambda_0 + \sum_{i=1}^N a^i \lambda_i \end{cases}$$

# A simple illustration

Example: Nonlinear PDE

$$\frac{\partial^2 u}{\partial x^2} + u^2 = \lambda \quad \text{+b.c}$$

$$\lambda = a$$

$$u(a) = u_0 + au_1 + a^2u_2 + \dots + a^N u_N$$

$$\lambda(a) = \lambda_0 + a\lambda_1 + a^2\lambda_2 + \dots + a^N \lambda_N$$

$$\Rightarrow \frac{\partial^2 u}{\partial x^2} = \frac{\partial^2 u_0}{\partial x^2} + a \frac{\partial^2 u_1}{\partial x^2} + a^2 \frac{\partial^2 u_1}{\partial x^2} + \dots + a^N \frac{\partial^2 u_N}{\partial x^2}$$

Exemple: EDP non linéaire

$$\left\{ \begin{array}{l} \frac{\partial^2 u}{\partial x^2} + u^2 = \lambda \\ \lambda = a \end{array} \right. \left\{ \begin{array}{l} \frac{\partial^2 u_0}{\partial x^2} + a \frac{\partial^2 u_1}{\partial x^2} + a^2 \frac{\partial^2 u_2}{\partial x^2} + \dots + a^N \frac{\partial^2 u_N}{\partial x^2} \\ + u_0^2 + a u_0 u_1 + a^2 u_0 u_2 + a^3 u_0 u_3 + \dots + a^N u_0 u_N \\ + a u_1 u_0 + a^2 u_1 u_1 + a^3 u_1 u_2 + \dots + \\ + a^2 u_2 u_0 + a^3 u_2 u_1 + \dots \\ + a^3 u_3 u_0 + \dots \\ + \dots \\ = \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N \\ \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N = a \end{array} \right.$$

Identification des même puissances  
de «  $a$  »:

$$\begin{aligned}
 & \left\{ \begin{aligned}
 & \frac{\partial^2 u_0}{\partial x^2} + a \frac{\partial^2 u_1}{\partial x^2} + a^2 \frac{\partial^2 u_2}{\partial x^2} + \dots + a^N \frac{\partial^2 u_N}{\partial x^2} \\
 & + u_0^2 + a u_0 u_1 + a^2 u_0 u_2 + a^3 u_0 u_3 + \dots + a^N u_0 u_N \\
 & \quad + a u_1 u_0 + a^2 u_1 u_1 + a^3 u_1 u_2 + \dots + \\
 & \quad \quad + a^2 u_2 u_0 + a^3 u_2 u_1 + \dots \\
 & \quad \quad \quad + a^3 u_3 u_0 + \dots \\
 & \quad \quad \quad \quad + \dots \\
 & = \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N \\
 & \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N = a
 \end{aligned} \right.
 \end{aligned}$$

Ordre 0

$$\begin{aligned}
 & \frac{\partial^2 u_0}{\partial x^2} + u_0^2 = \lambda_0 \\
 & \lambda_0 = 0
 \end{aligned}$$

Ordre 0

$$\frac{\partial^2 u_0}{\partial x^2} + u_0^2 = \lambda_0$$
$$\lambda_0 = 0$$

On suppose  $u_0$  connue, exemple:  
 $u_0 = 0$

$$\lambda_0 = 0$$

Identification des même puissances  
de a:

$$\left\{ \begin{aligned}
 & \frac{\partial^2 u_0}{\partial x^2} + a \frac{\partial^2 u_1}{\partial x^2} + a^2 \frac{\partial^2 u_2}{\partial x^2} + \dots + a^N \frac{\partial^2 u_N}{\partial x^2} \\
 & + u_0^2 + a u_0 u_1 + a^2 u_0 u_2 + a^3 u_0 u_3 + \dots + a^N u_0 u_N \\
 & + a u_1 u_0 + a^2 u_1 u_1 + a^3 u_1 u_2 + \dots + \\
 & + a^2 u_2 u_0 + a^3 u_2 u_1 + \dots \\
 & + a^3 u_3 u_0 + \dots \\
 & + \dots \\
 & = \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N \\
 & \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N = a \cdot 1
 \end{aligned} \right.$$

Ordre 1

$$\begin{aligned}
 & \frac{\partial^2 u_1}{\partial x^2} + 2u_0 u_1 = \lambda_1 \\
 & \lambda_1 = 1
 \end{aligned}$$

Ordre 1

Ordre 1

$$\frac{\partial^2 u_1}{\partial x^2} + 2u_0 u_1 = \lambda_1$$

$$\lambda_1 = 1$$

$$\frac{\partial^2 u_1}{\partial x^2} = \lambda_1$$

$$\lambda_1 = 1$$

On trouve

$u_1$

$$\lambda_1 = 1$$

Identification des même puissances  
de a:

$$\left\{ \begin{aligned}
 & \frac{\partial^2 u_0}{\partial x^2} + a \frac{\partial^2 u_1}{\partial x^2} + a^2 \frac{\partial^2 u_2}{\partial x^2} + \dots + a^N \frac{\partial^2 u_N}{\partial x^2} \\
 & + u_0^2 + a u_0 u_1 + a^2 u_0 u_2 + a^3 u_0 u_3 + \dots + a^N u_0 u_N \\
 & + a u_1 u_0 + a^2 u_1 u_1 + a^3 u_1 u_2 + \dots + \\
 & + a^2 u_2 u_0 + a^3 u_2 u_1 + \dots \\
 & + a^3 u_3 u_0 + \dots \\
 & + \dots \\
 & = \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N \\
 & \lambda_0 + a \lambda_1 + a^2 \lambda_2 + \dots + a^N \lambda_N = a.1
 \end{aligned} \right.$$

Ordre 2

$$\begin{aligned}
 & \frac{\partial^2 u_2}{\partial x^2} + 2u_0 u_2 + u_1^2 = \lambda_2 \\
 & \lambda_2 = 0
 \end{aligned}$$



Ordre 2

$$\frac{\partial^2 u_2}{\partial x^2} + 2u_0 u_2 + u_1^2 = \lambda_2$$
$$\lambda_2 = 0$$

$$\frac{\partial^2 u_2}{\partial x^2} = -u_1^2$$

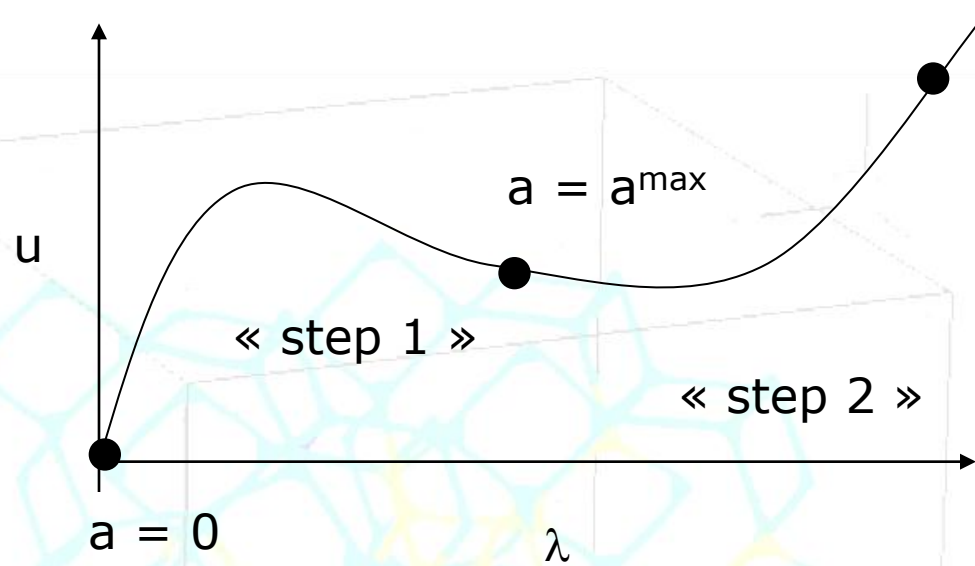
$$\lambda_2 = 0$$

On trouve

$$u_2$$
$$\lambda_1 = 1$$

Etc...

The series has a radius of convergence: the solution is defined along « steps » through a continuation method



Continuation method

Given  $N$ , a tolerance  $\epsilon$  is chosen such that

$$\epsilon = \frac{\|\mathbf{u}^{n+1}(N, a_{max}) - \mathbf{u}^{n+1}(N-1, a_{max})\|}{\|\mathbf{u}^{n+1}(N, a_{max}) - \mathbf{u}^n\|}$$

## A more involved example: elasticity with geometrical nonlinearity

$$\int_{\Omega} {}^t\mathbf{P} : \delta \mathbf{F} d\Omega = \lambda \mathcal{P}_{ext}(\delta \mathbf{u})$$

Constitutive law:

$$\mathbf{S} = \mathbb{C} : \gamma$$

Kinematic relations:

$$\mathbf{P} = \mathbf{F} \cdot \mathbf{S}$$

$$\gamma = \frac{1}{2} ({}^t\mathbf{F} \cdot \mathbf{F} - \mathbf{I})$$

$$\mathbf{F} = \mathbf{I} + \nabla \mathbf{u}$$

$$\begin{Bmatrix} \mathbf{U}(a) \\ \lambda(a) \end{Bmatrix} = \begin{Bmatrix} \mathbf{U}_0 \\ \lambda_0 \end{Bmatrix} + \sum_{i=1}^N a^i \begin{Bmatrix} \mathbf{U}_i \\ \lambda_i \end{Bmatrix}$$

$$\mathbf{U} = \{\mathbf{P}, \mathbf{F}, \mathbf{S}, \gamma, \mathbf{u}\}$$



$$\left\{ \begin{array}{l} \int_{\Omega} \left( \sum_{i=0}^N a^i {}^t\mathbf{P}_i : \delta \mathbf{F} \right) d\Omega = \sum_{i=0}^N a^i \lambda_i \mathcal{P}_{ext}(\delta \mathbf{u}) \\ \sum_{i=0}^N a^i \mathbf{S}_i = \mathbb{C} : \sum_{i=0}^N a^i \gamma_i \\ \sum_{i=0}^N a^i \mathbf{P}_i = \left( \sum_{i=0}^N a^i \mathbf{F}_i \right) \cdot \left( \sum_{i=0}^N a^i \mathbf{S}_i \right) \\ \sum_{i=0}^N a^i \gamma_i = \frac{1}{2} \left[ \left( \sum_{i=0}^N a^i {}^t\mathbf{F}_i \right) \cdot \left( \sum_{i=0}^N a^i \mathbf{F}_i \right) - \mathbf{I} \right] \\ \sum_{i=0}^N a^i \mathbf{F}_i = \mathbf{I} + \nabla \left( \sum_{i=0}^N a^i \mathbf{u}_i \right) \end{array} \right.$$

Grouping terms with same exponent produces at order k

$$\left\{ \begin{array}{l} \int_{\Omega} {}^t\mathbf{P}_k : \delta \mathbf{F} d\Omega = \lambda_k \mathcal{P}_{ext}(\delta \mathbf{u}) \\ \mathbf{S}_k = \mathbb{C} : \gamma_k \\ \mathbf{P}_k = \mathbf{F}_0 \cdot \mathbf{S}_k + \mathbf{F}_k \cdot \mathbf{S}_0 + \sum_{r=1}^{k-1} \mathbf{F}_{k-r} \cdot \mathbf{S}_r \\ \gamma_k = \frac{1}{2} \left( {}^t\mathbf{F}_0 \cdot \mathbf{F}_k + {}^t\mathbf{F}_k \cdot \mathbf{F}_0 + \sum_{r=1}^{k-1} {}^t\mathbf{F}_{k-r} \cdot \mathbf{F}_r \right) \\ \mathbf{F}_k = \nabla \mathbf{u}_k \end{array} \right.$$

Sequence of linear problems with same linear operator solved with FEM

$$\mathcal{L}(\mathbf{u}_k, \delta \mathbf{u}) = \lambda_k \mathcal{P}_{ext}(\delta \mathbf{u}) + \mathcal{F}_k^{nl}(\delta \mathbf{u})$$

Order k

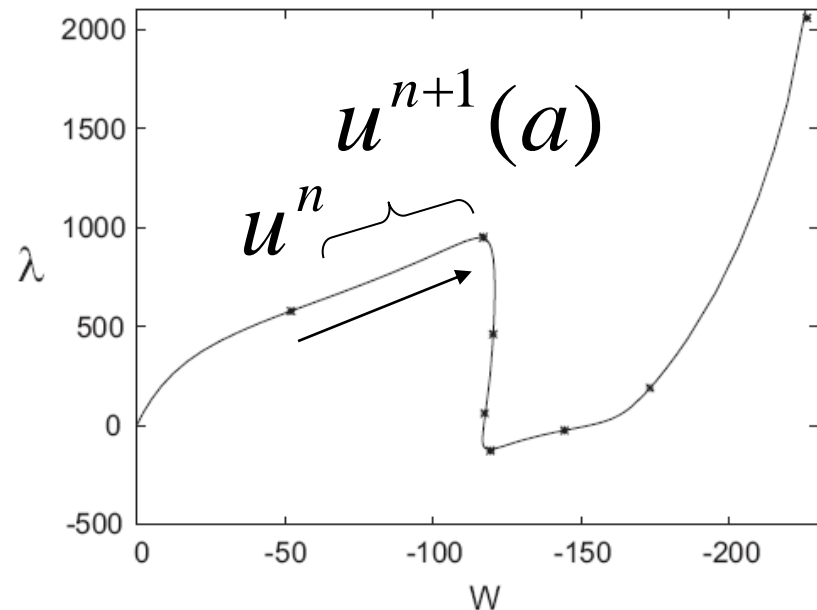
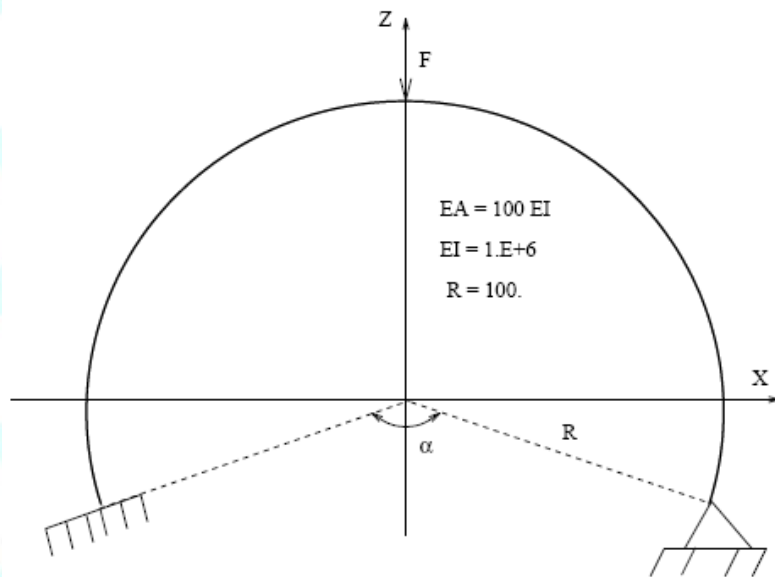
$$\mathcal{L}(\mathbf{u}_k, \delta \mathbf{u}) = \int_{\Omega} {}^t\nabla \mathbf{u}_k : \mathbb{H} : \nabla \delta \mathbf{u} d\Omega ,$$

$$\mathcal{F}_k^{nl}(\delta \mathbf{u}) = - \int_{\Omega} {}^t\mathbf{P}_k^{nl} : \nabla \delta \mathbf{u} d\Omega \quad 1 < k \leq N$$

# Application to elastic buckling

$$u^{n+1}(a) = u^n + au_1 + a^2u_2 + \dots + a^p u_p$$

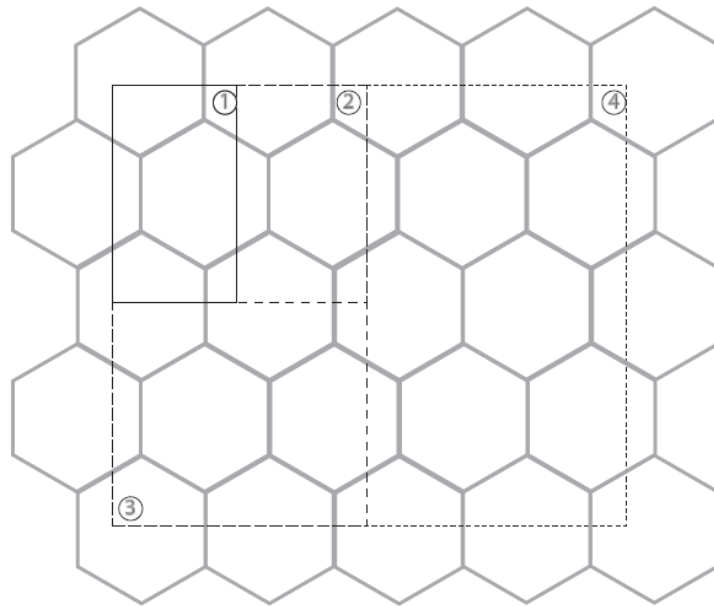
$$\lambda^{n+1}(a) = \lambda^n + a\lambda_1 + a^2\lambda_2 + \dots + a^p \lambda_p$$



- Piece-wise continuous approximation of the solution with respect to the development parameter  $a$
- remove difficulties related to limit points

# Méthode asymptotique numérique (MAN)

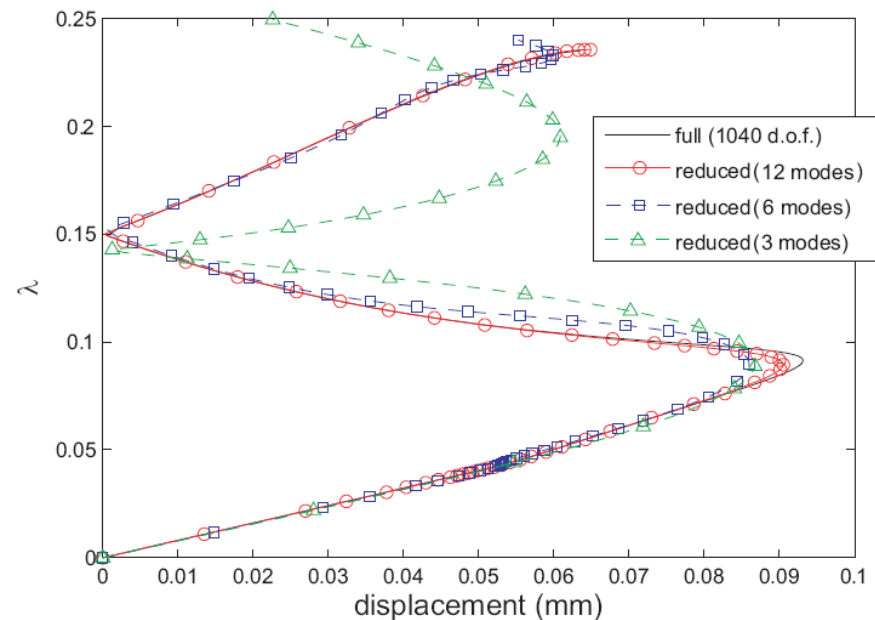
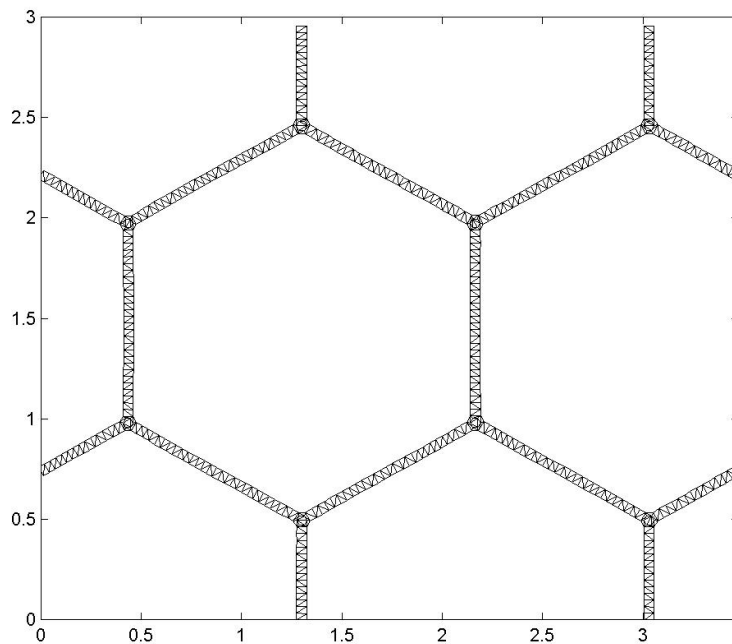
Exemples : flambement de microstructures cellulaires en compression



[Yvonnet, Zahrouni, Potier-Ferry, Computer Methods in Applied Mechanics and Engineering, 2008]

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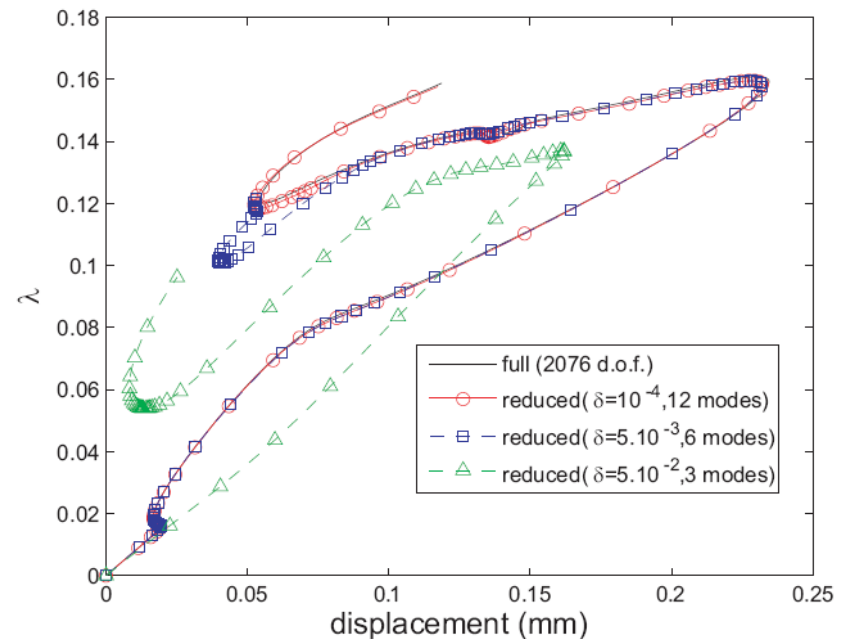
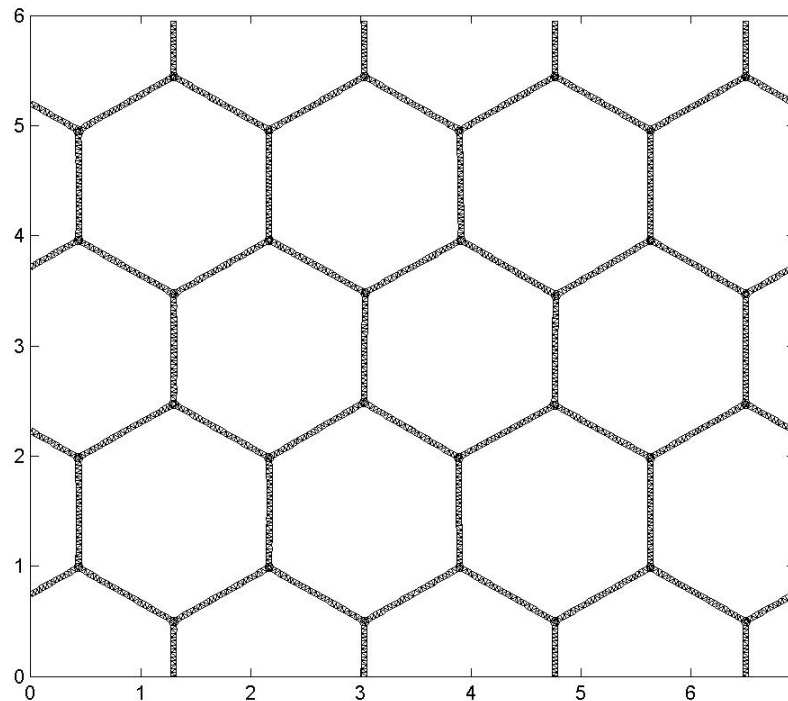
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Exemples : flambement de microstructures cellulaires

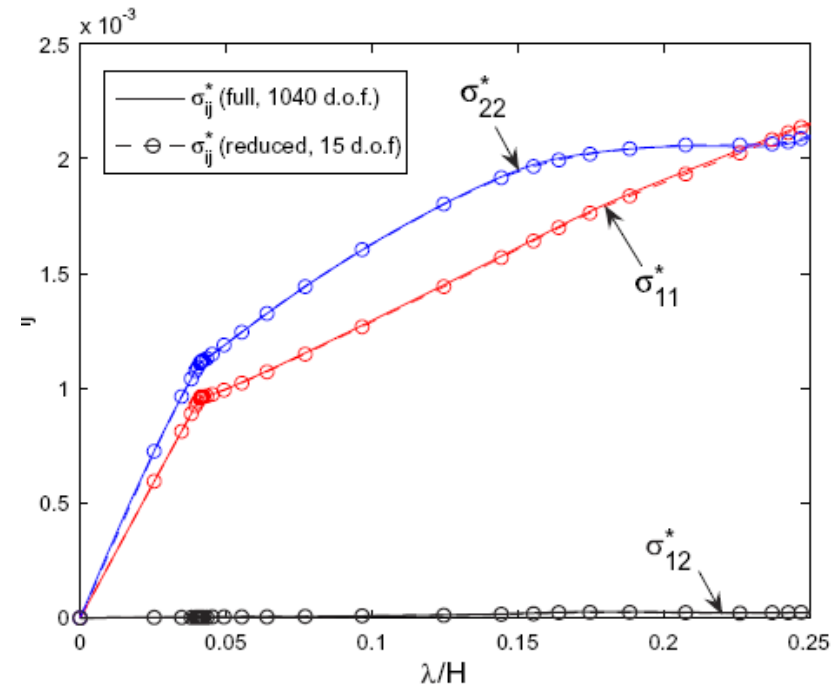
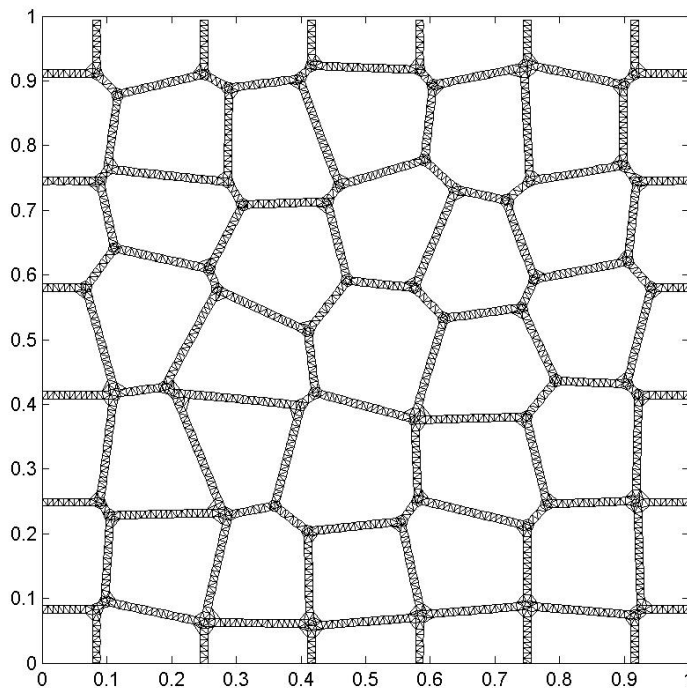


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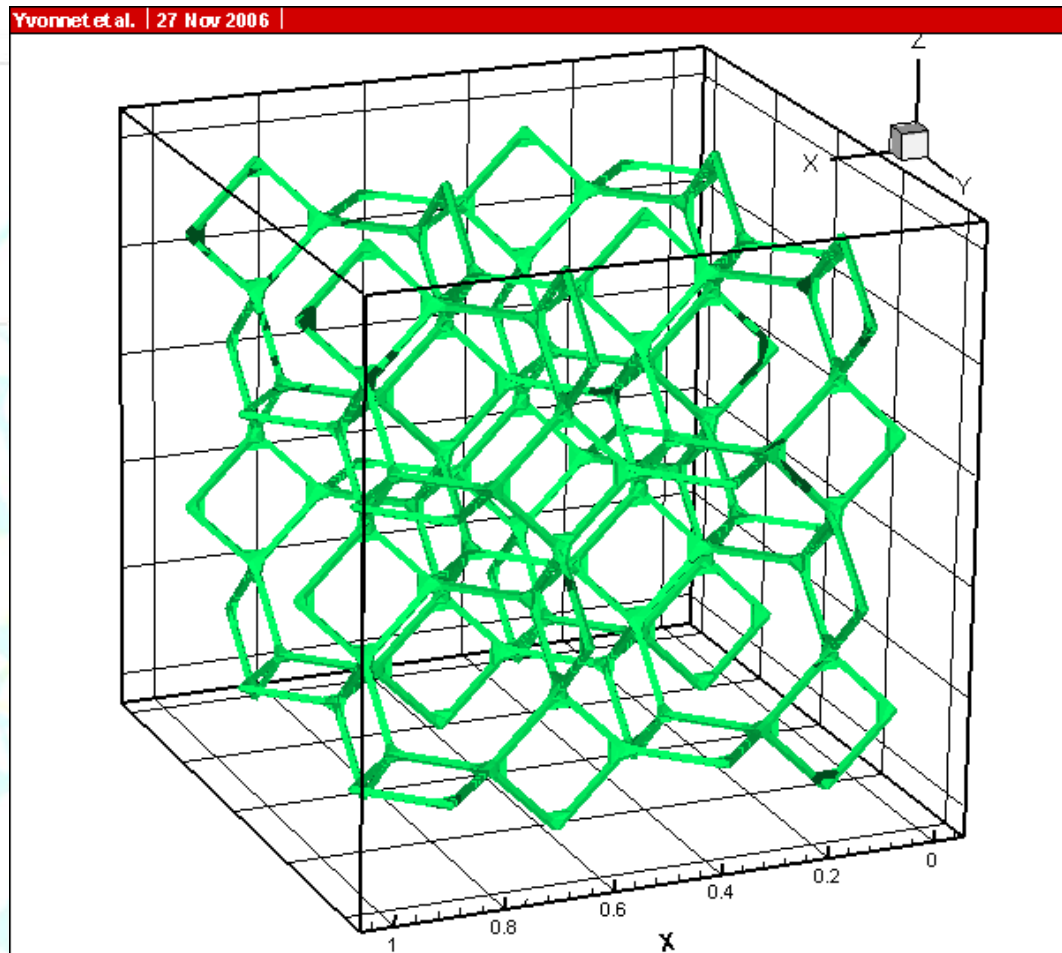
# Méthode asymptotique numérique (MAN)

Exemples : flambement de microstructures cellulaires

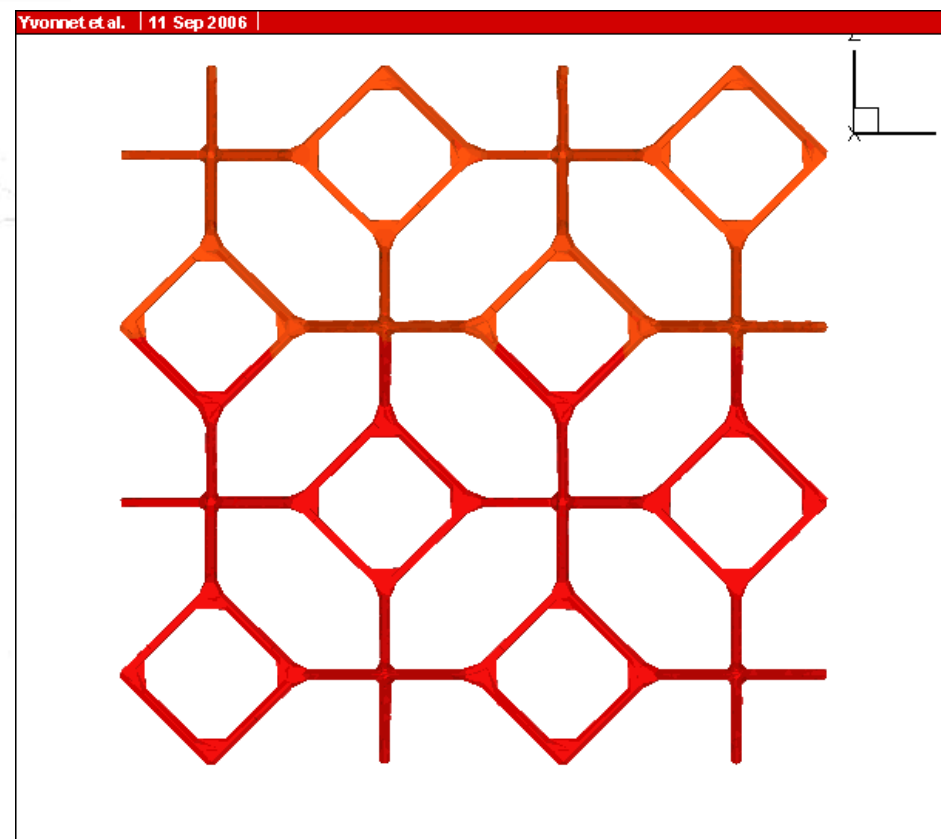
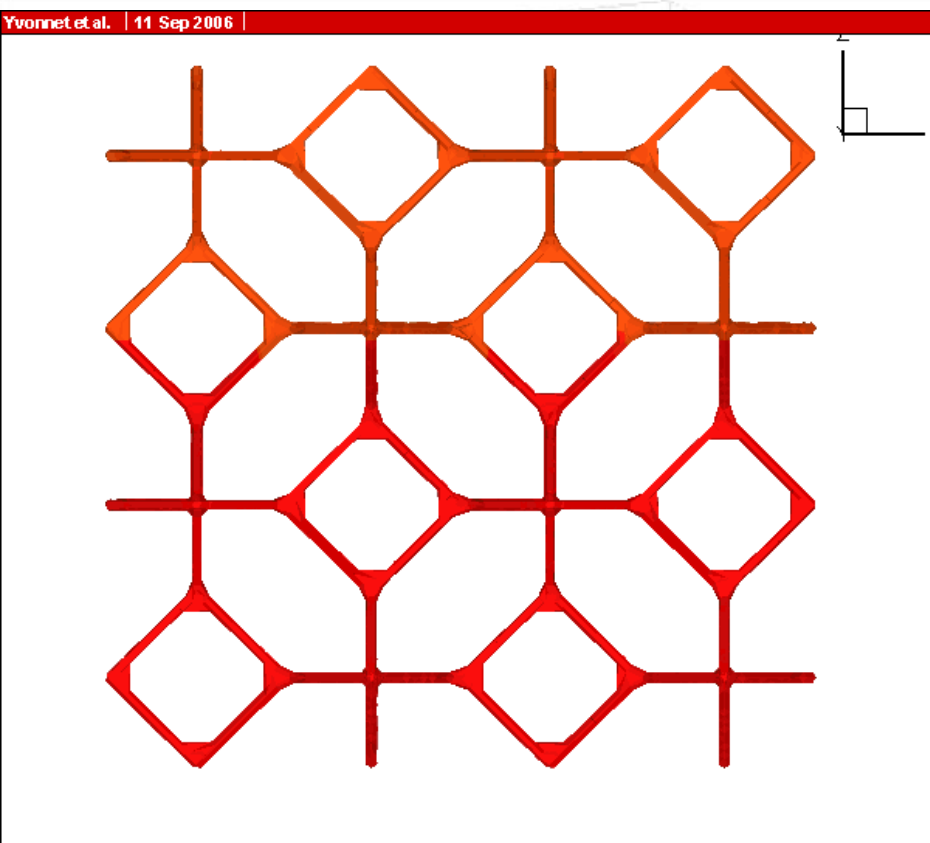


[Yvonnet, Zahrouni, Potier-Ferry, Computer Methods in Applied Mechanics and Engineering, 2008]

# Méthode asymptotique numérique (MAN)



# Méthode asymptotique numérique (MAN)





# Méthodes de réduction de modèles de type POD

# Méthodes de réduction de modèles

Systèmes de grandes dimensions (nombre important de d.d.l.) :

- Matrice de grandes tailles à décomposer (inverser)
- Assemblage des matrices coûteux, place mémoire importante

Enjeux des méthodes de réduction de modèle :  
transformer un système d'équations de taille  $N \times N$  en  
un système de « taille réduite »  $M \times M$  avec  $M \ll N$

Applications : séries de problèmes non-linéaires  
« similaires » de taille importante (Optimisation, calcul  
multi-échelles...)

# Méthode de réduction de modèles de type POD

## POD (Proper orthogonal decomposition)

Approximation de la solution Éléments Finis par projection dans un sous-espace de taille  $M$

$$\mathbf{q}^R(t) = \phi_0 + \sum_{m=1}^M \phi_m \xi_m(t) \longrightarrow \text{Variables arbitraires}$$

↓  
Fonctions de base globales

Soit  $H$  un espace de Hilbert d'ordre  $N$ ;  $\mathbb{R}^N$  muni du produit scalaire  $\langle \cdot, \cdot \rangle$  et de la norme  $\|\psi\| = \sqrt{\langle \psi, \psi \rangle}$ .

Fonctions de base définies par le problème de minimisation

$$\min_{\phi} \int_0^T \left\| \mathbf{u}(t) - \phi \phi^T \mathbf{u}(t) \right\|^2 dt$$

Sous la contrainte  $\langle \phi_i, \phi_j \rangle = \delta_{ij}$

# Méthode de réduction de modèles de type POD

La minimisation conduit, après discrétisation temporelle du problème aux valeurs propres [Liang 2002,etc...]

$$\mathbf{Q}\phi_i = \lambda_i \phi_i$$

$$\mathbf{Q} = \frac{1}{S} \mathbf{U} \mathbf{U}^T$$

$$\phi_0 = \bar{\mathbf{q}} = \frac{1}{S} \sum_{i=1}^S \mathbf{q}(t_i)$$

$$\mathbf{U} = \{\mathbf{q}(t_1) - \bar{\mathbf{q}}, \mathbf{q}(t_2) - \bar{\mathbf{q}}, \dots, \mathbf{q}(t_S) - \bar{\mathbf{q}}\}$$

On cherche à obtenir un sous-espace de dimension  $M \ll N$

Erreur liée à la troncature de la base  $\lambda_1 \geq \lambda_2 \geq \dots \lambda_M \geq \dots \geq \lambda_{DN} \geq 0$ .

$$\epsilon(M) = \sum_{i=1}^S \|\mathbf{q}(\mathbf{x}, t_i) - \mathbf{q}^R(\mathbf{x}, t_i)\| = \left( \sum_{i=M+1}^{DN} \lambda_i \right)^{1/2}$$

$$\text{Critère : } \frac{\left( \sum_{i=M+1}^{DN} \lambda_i \right)^{1/2}}{\left( \sum_{i=1}^{DN} \lambda_i \right)^{1/2}} < \delta$$



# Méthode de réduction de modèles de type POD

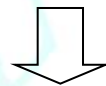
$$\mathbf{K}_\mu^k \Delta \mathbf{q}^{k+1} = \mathbf{f}_{ext(\mu)} - \mathbf{f}_{int}^k(\mathbf{q}_\mu^k)$$

Incréments de déplacements projetés dans la base réduite

$$\Delta \mathbf{q}^{k+1} = \sum_{m=1}^M \phi_m \Delta \xi_m^{k+1}$$

$$\Phi = \{\phi_1, \phi_2, \dots, \phi_M\}$$

$$\xi = \{\xi_1, \xi_2, \dots, \xi_M\}$$



Problème réduit linéarisé

$$\Phi^T \mathbf{K}_\mu^k \Phi \Delta \xi^{k+1} = \Phi^T [\mathbf{f}_{ext(\mu)} - \mathbf{f}_{int}^k(\mathbf{q}_\mu^k)]$$

$[M \times M]$

Actualisation des variables réduites

$$\xi^{k+1} = \xi^k + \Delta \xi^{k+1}$$

$$M \ll N$$



# Méthode de réduction de modèles de type POD

Exemple d'application : hyperélasticité non-linéaire en grandes transformations

$$\nabla \cdot \bar{\mathbf{P}} + \bar{\mathbf{B}} = 0 \quad \text{and} \quad \bar{\mathbf{P}} \bar{\mathbf{F}}^T = (\bar{\mathbf{P}} \bar{\mathbf{F}}^T)^T \quad \text{in } \Omega_0$$

$$\mathbf{P} = \frac{\partial \Psi(\mathbf{F})}{\partial \mathbf{F}} \quad \Psi = c(J - 1)^2 - d \log(J) + c_1(I_1 - 3) + c_2(I_2 - 3)$$

Matériau hyperélastique de type Mooney-Rivlin compressible

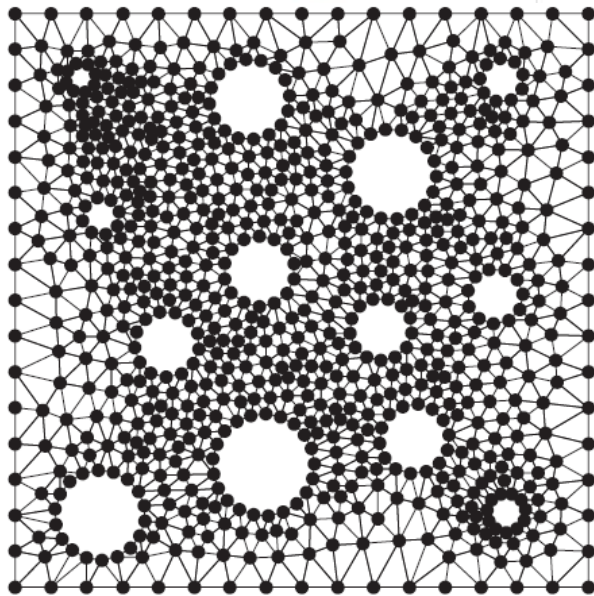
$$D_{\Delta \mathbf{u}} \delta W_{int}(\mathbf{u}, \delta \mathbf{u}) = \int_{\Omega_0} \left[ \nabla_X(\delta \mathbf{u}) : \nabla_X(\Delta \mathbf{u}) \bar{\mathbf{S}} + \bar{\mathbf{F}}^T \nabla_X(\delta \mathbf{u}) : \bar{\mathbf{C}}^e : \bar{\mathbf{F}}^T \nabla_X(\Delta \mathbf{u}) \right] d\Omega$$

Problème tangent

# Méthode de réduction de modèles de type POD

Exemple d'application : hyperélasticité non-linéaire en grandes transformations

Conditions aux limites :  $\mathbf{u} = [\mathbf{F} - \mathbf{1}]\mathbf{x}$



Réponse d'un VER sous sollicitations complexes

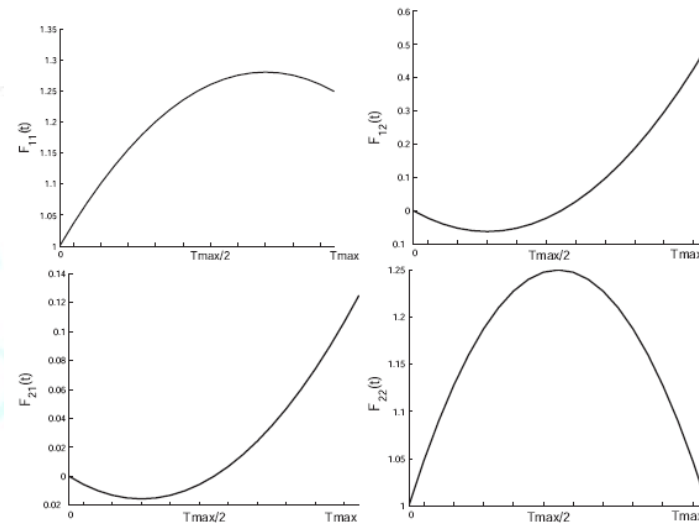
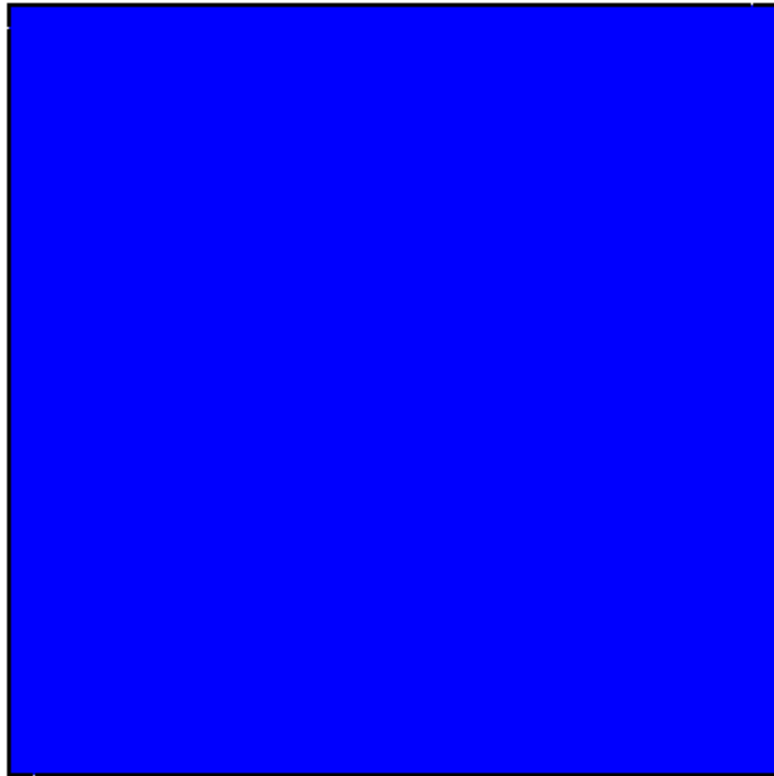


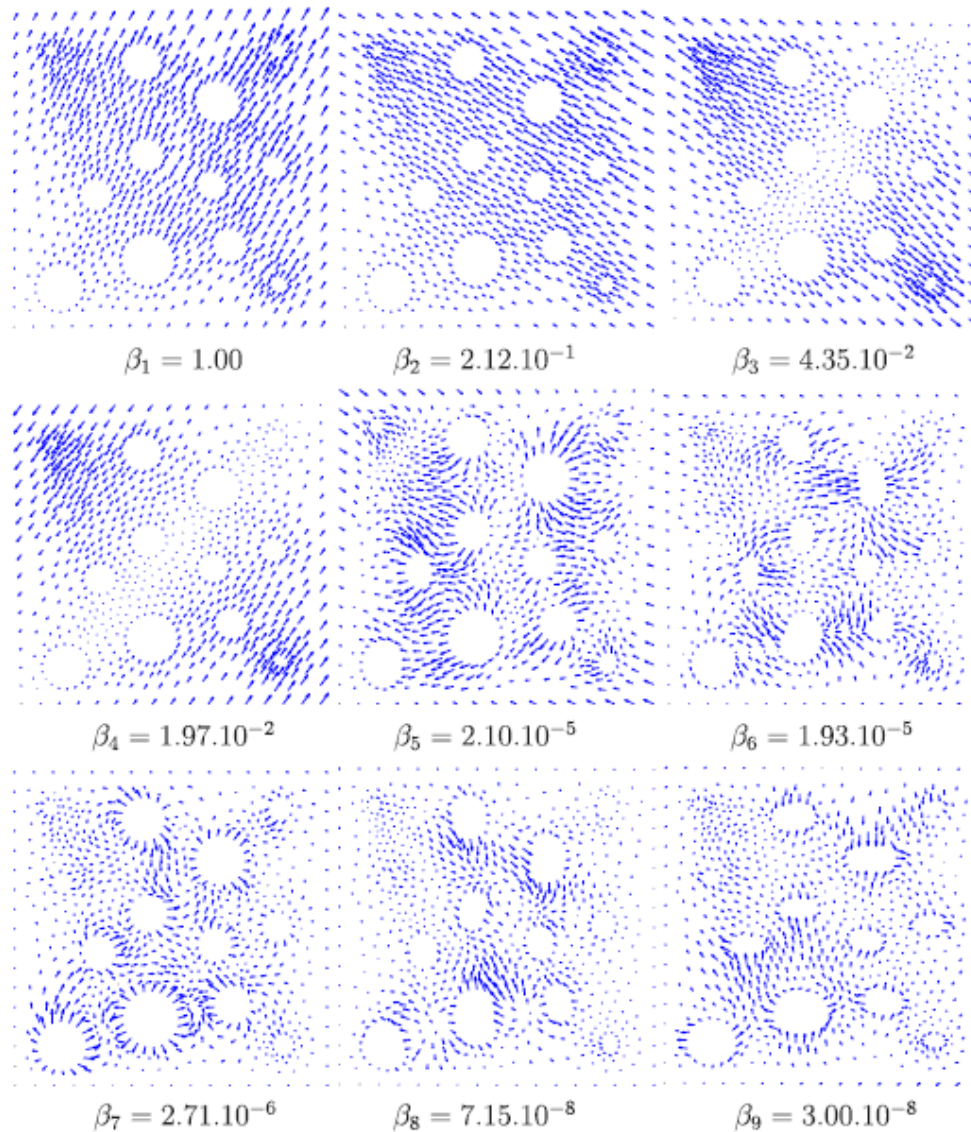
Fig. 5. Values of  $F_{11}$ ,  $F_{12}$ ,  $F_{21}$  and  $F_{22}$  along the simulation.

Évolution des composantes de  $\mathbf{F}$

[Yvonnet & He 2007]



# Méthode de réduction de modèles de

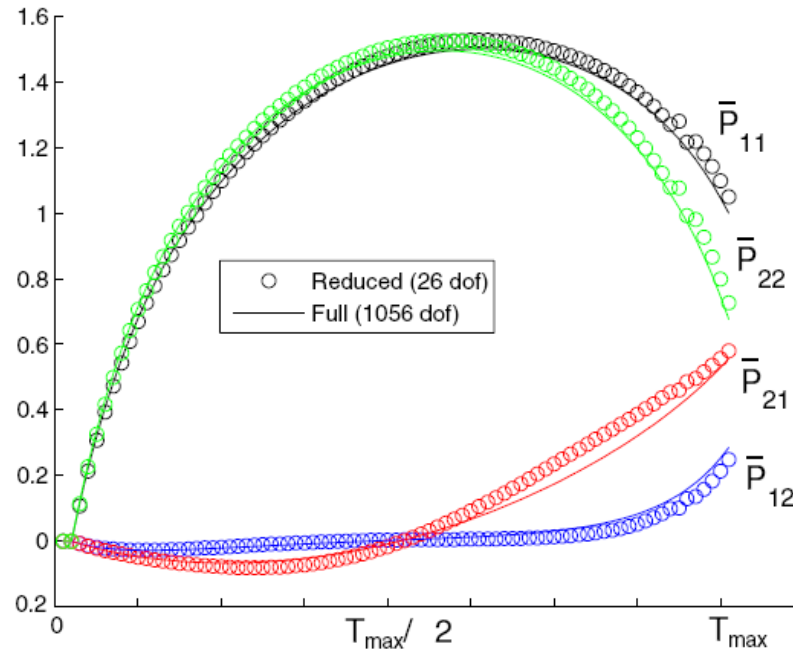


9 premiers modes

26 modes  
retenus pour  
1056 ddl

[Yvonnet & He,  
Computational  
Physics (2007)]

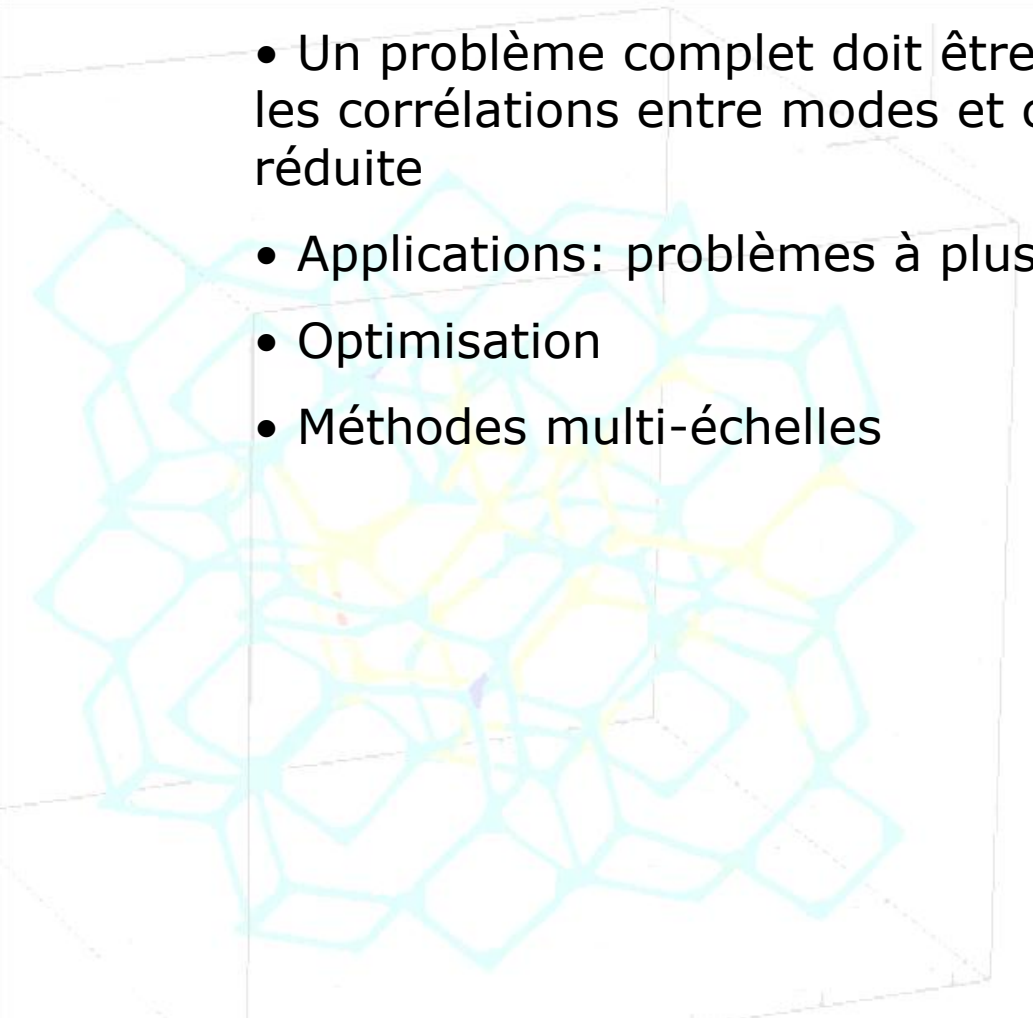
# Méthode de réduction de modèles de type POD



Comparaison entre solution complète (1056 ddl) et solution par modèle réduit (26 ddl)  
[Yvonnet & He, Computational Physics (2007)]

# Méthode de réduction de modèle

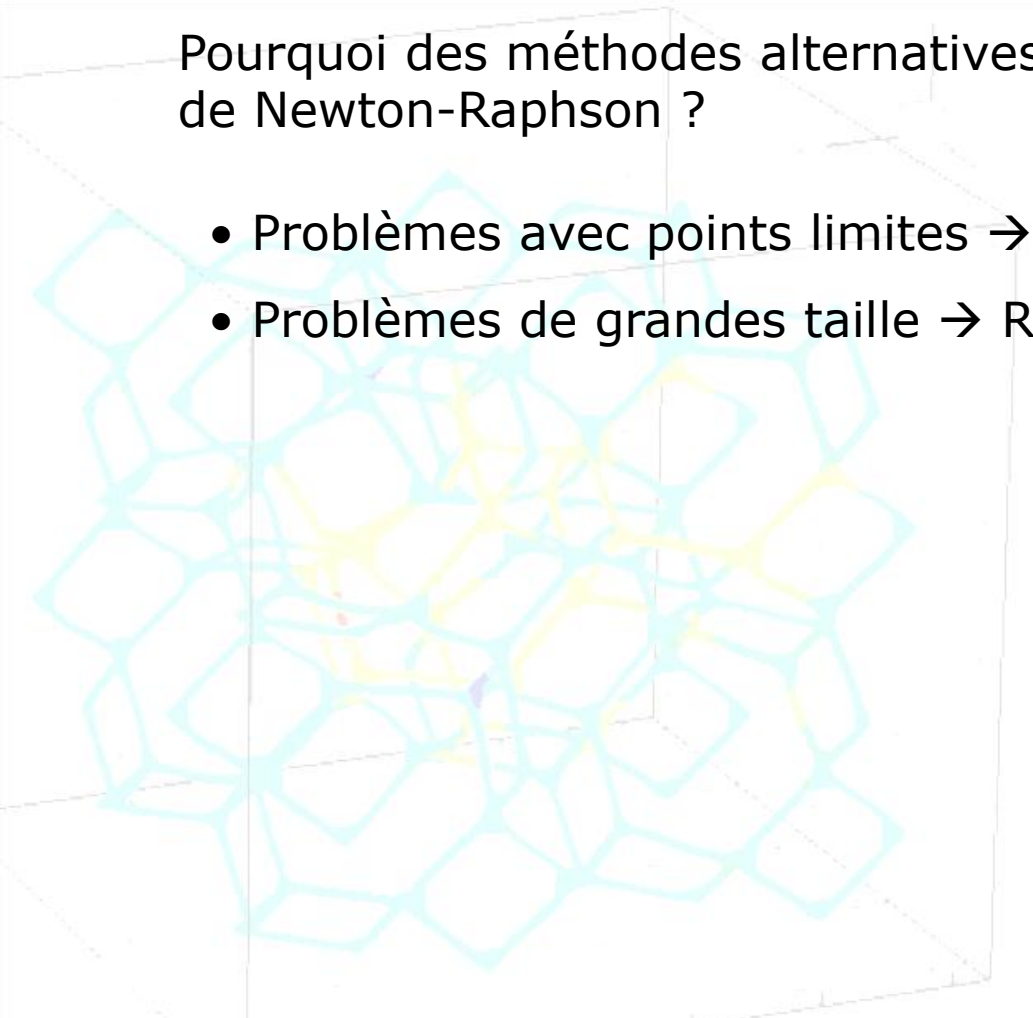
- Un problème complet doit être résolu pour détecter les corrélations entre modes et construire la base réduite
- Applications: problèmes à plusieurs paramètres
- Optimisation
- Méthodes multi-échelles



# Conclusion

Pourquoi des méthodes alternatives à la méthode de Newton-Raphson ?

- Problèmes avec points limites → MAN
- Problèmes de grandes taille → Réduction de modèle







# Homogenization of nonlinear problems using Asymptotic Numerical Method



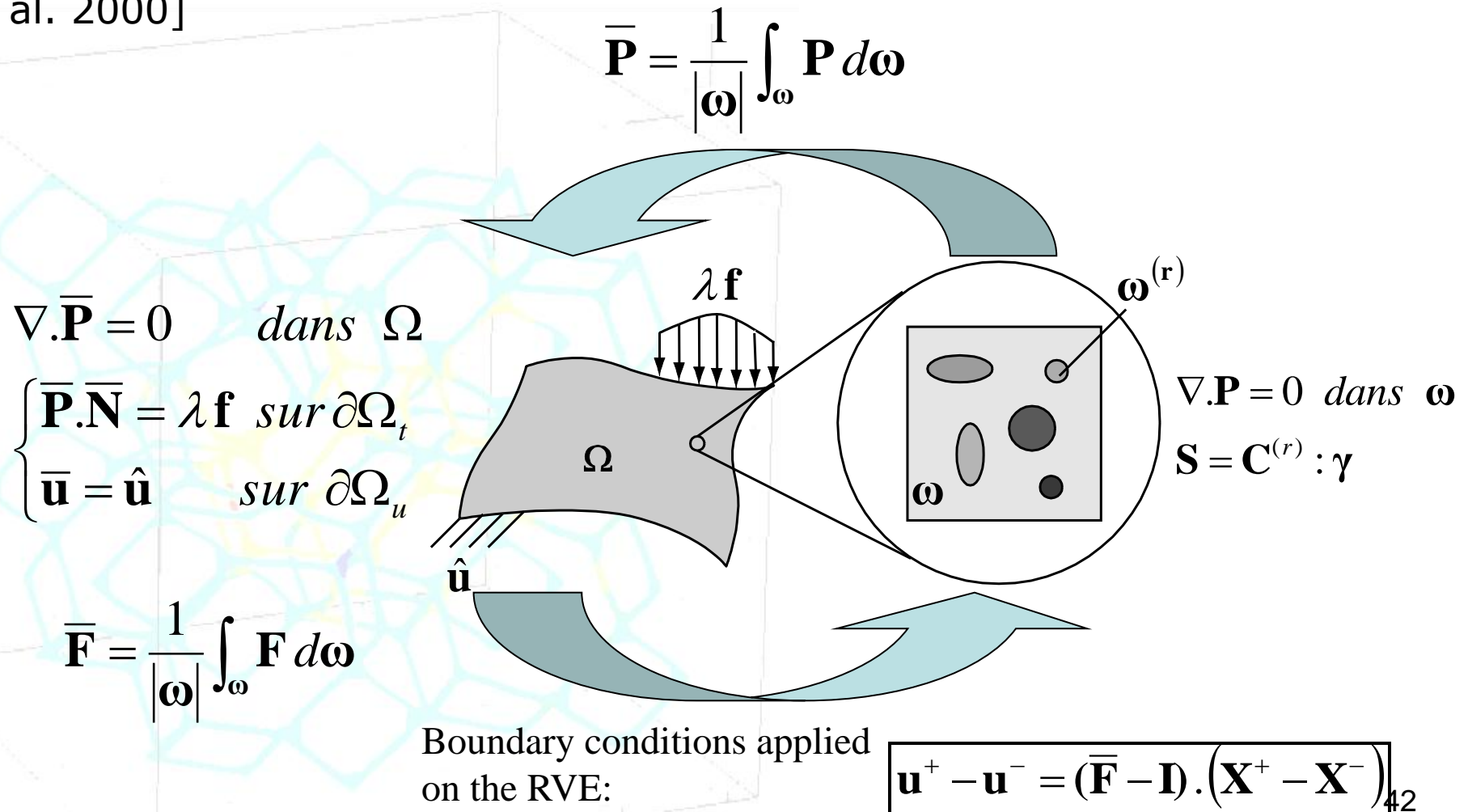
# A non linear homogenization method with possible instabilities: the multiscale MAN

The problem is formulated at two scales: both microscopic and macroscopic variables are expanded in finite series (ANM)

- Nonlinear problems (micro and macro) result in sequences of linear problems
- Superposition principle is applied on each linear problem;
- A localization tensor can be constructed for each step of the asymptotic expansion

# Multiscale elasticity problem with geometrical/material nonlinearities

Multiscale FEM [Renard et al. 1987, Feyel 1998, Smit 1998, Geers and Kouznetsova], Asymptotic Homogenization [Hou and Wu 1997, Terada et al. 2000]



## Multiscale ANM

$$\begin{Bmatrix} \boldsymbol{\Lambda}(\boldsymbol{a}) \\ \lambda(\boldsymbol{a}) \end{Bmatrix} = \begin{Bmatrix} \boldsymbol{\Lambda}_0 \\ \lambda_0 \end{Bmatrix} + \sum_{p=1}^N a^p \begin{Bmatrix} \boldsymbol{\Lambda}_p \\ \lambda_p \end{Bmatrix}$$

Series expansions of both microscopic and macroscopic variables

$$\boldsymbol{\Lambda} = (\bar{\mathbf{u}}, \bar{\mathbf{P}}, \mathbf{u}, \mathbf{P}, \dots)$$

# Macroscopic problem

Weak form related to the macroscopic problem

$$\int_{\Omega} {}^t \bar{\mathbf{P}} : \delta \bar{\mathbf{F}} d\Omega = \lambda \int_{\partial\Omega_t} \mathbf{f} \cdot \delta \bar{\mathbf{u}} d\Gamma$$

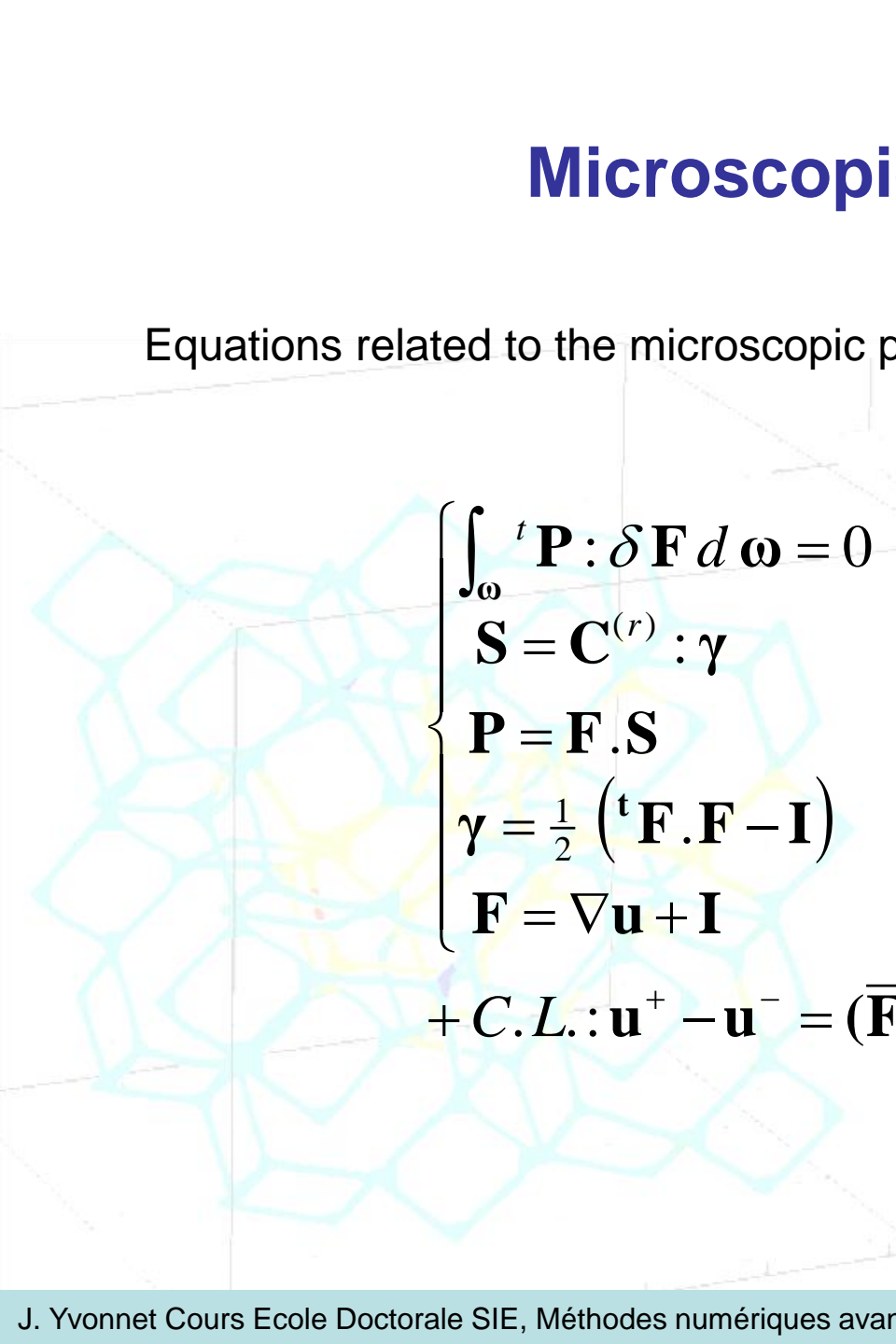
MAN expansion: problem at order p

$$\int_{\Omega} {}^t \bar{\mathbf{P}}_p : \delta \bar{\mathbf{F}} d\Omega = \lambda_p \int_{\partial\Omega_t} \mathbf{f} \cdot \delta \bar{\mathbf{u}} d\Gamma \quad \forall p = 1 \dots N$$

The constitutive law is not known explicitly at this scale, but known implicitly via the microscopic scale

# Microscopic problem

Equations related to the microscopic problem


$$\left\{ \begin{array}{l} \int_{\omega} {}^t \mathbf{P} : \delta \mathbf{F} d\omega = 0 \\ \mathbf{S} = \mathbf{C}^{(r)} : \boldsymbol{\gamma} \\ \mathbf{P} = \mathbf{F} . \mathbf{S} \\ \boldsymbol{\gamma} = \frac{1}{2} \left( {}^t \mathbf{F} . \mathbf{F} - \mathbf{I} \right) \\ \mathbf{F} = \nabla \mathbf{u} + \mathbf{I} \end{array} \right. \quad \text{in } \omega$$
$$+ C.L. : \mathbf{u}^+ - \mathbf{u}^- = (\bar{\mathbf{F}} - \mathbf{I}) . (\mathbf{X}^+ - \mathbf{X}^-) \quad \text{on } \partial\omega$$

# Microscopic problem

Introducing asymptotic expansion

Order 1:

$$\left\{ \begin{array}{l} \int_{\omega} {}^t\mathbf{P}_1 : \delta \mathbf{F} d\omega = 0 \\ \mathbf{S}_1 = \mathbf{C}^{(r)} : \gamma_1 \\ \mathbf{P}_1 = \mathbf{F}_0 \cdot \mathbf{S}_1 + \mathbf{F}_1 \cdot \mathbf{S}_0 \\ \gamma_1 = \frac{1}{2} ({}^t\mathbf{F}_0 \cdot \mathbf{F}_1 + {}^t\mathbf{F}_1 \cdot \mathbf{F}_0) \\ \mathbf{F}_1 = \nabla \mathbf{u}_1 \end{array} \right. \quad \text{in } \omega$$

$$+ C.L.: \mathbf{u}_1^+ - \mathbf{u}_1^- = \bar{\mathbf{F}}_1 \cdot (\mathbf{X}^+ - \mathbf{X}^-) \quad \text{on } \partial \omega$$

Order  $p$ :

$$\left\{ \begin{array}{l} \int_{\omega} {}^t\mathbf{P}_p : \delta \mathbf{F} d\omega = 0 \\ \mathbf{S}_p = \mathbf{C}^{(r)} : \gamma_p \\ \mathbf{P}_p = \mathbf{F}_0 \cdot \mathbf{S}_p + \mathbf{F}_p \cdot \mathbf{S}_0 + \sum_{r=1}^p \mathbf{F}_r \cdot \mathbf{S}_{p-r} \\ \gamma_p = \frac{1}{2} \left( {}^t\mathbf{F}_0 \cdot \mathbf{F}_p + {}^t\mathbf{F}_p \cdot \mathbf{F}_0 + \sum_{r=1}^p {}^t\mathbf{F}_r \cdot \mathbf{F}_{p-r} \right) \\ \mathbf{F}_p = \nabla \mathbf{u}_p \end{array} \right. \quad \text{in } \omega$$

$$+ C.L.: \mathbf{u}_p^+ - \mathbf{u}_p^- = \bar{\mathbf{F}}_p \cdot (\mathbf{X}^+ - \mathbf{X}^-) \quad \text{on } \partial \omega$$

# Constructing the macroscopic constitutive law at order p

Order 1:  $\mathbf{u}_1 = \mathbf{A} : \bar{\mathbf{F}}_1 \quad A_{ijk} = \tilde{u}_i^{jk}$

A is a third order tensor. Asymptotic development of gradient tensor F yields: :

$$\nabla \mathbf{u}_1 = \mathbf{u}_{1,X} = \mathbf{F}_1 = \mathbf{A}_{,X} : \bar{\mathbf{F}}_1$$

Localization tensor

$\mathbf{A}_{,X}$  is a fourth-order tensor.

At order p we obtain

$$\mathbf{u}_p = \mathbf{A} : \bar{\mathbf{F}}_p + \mathbf{u}_p^{nl}$$

The vector  $\mathbf{u}_p^{nl}$  is obtained by

$$\mathcal{L}_t(\mathbf{u}_p^{nl}, \delta \mathbf{u}) = \mathcal{F}_p^{nl}(\delta \mathbf{u}) \quad \text{in } \omega$$

Then

$$\mathbf{F}_p = \mathbf{A}_{,X} : \bar{\mathbf{F}}_p + \mathbf{u}_{p,X}^{nl}$$

**[Nezamabadi et al. 2009,  
Comput. Meth. Appl.  
Mech. Engng.]**

# Constructing the macroscopic constitutive law at order p

At order p we have:

$$\mathbf{P}_p = \mathbf{L} : \bar{\mathbf{F}}_p + \mathbf{P}_p^{nl}$$

The stress is obtained through:

$$\bar{\mathbf{P}}_p = \langle \mathbf{L} \rangle : \bar{\mathbf{F}}_p + \langle \mathbf{P}_p^{nl} \rangle$$

Introduction into the p-th order macroscopic problem

$$\int_{\Omega} {}^t \bar{\mathbf{P}}_p : \delta \bar{\mathbf{F}} d\Omega = \lambda_p \int_{\partial\Omega_t} \mathbf{f} \cdot \delta \bar{\mathbf{u}} d\Gamma$$



$$\int_{\Omega} {}^t \bar{\mathbf{F}}_p : \langle \mathbf{L} \rangle : \delta \bar{\mathbf{F}} d\Omega = \lambda_p \int_{\partial\Omega_t} \mathbf{f} \cdot \delta \bar{\mathbf{u}} d\Gamma - \int_{\Omega} {}^t \langle \mathbf{P}_p^{nl} \rangle : \delta \bar{\mathbf{F}} d\Omega$$

**Macroscopic problem at order p**



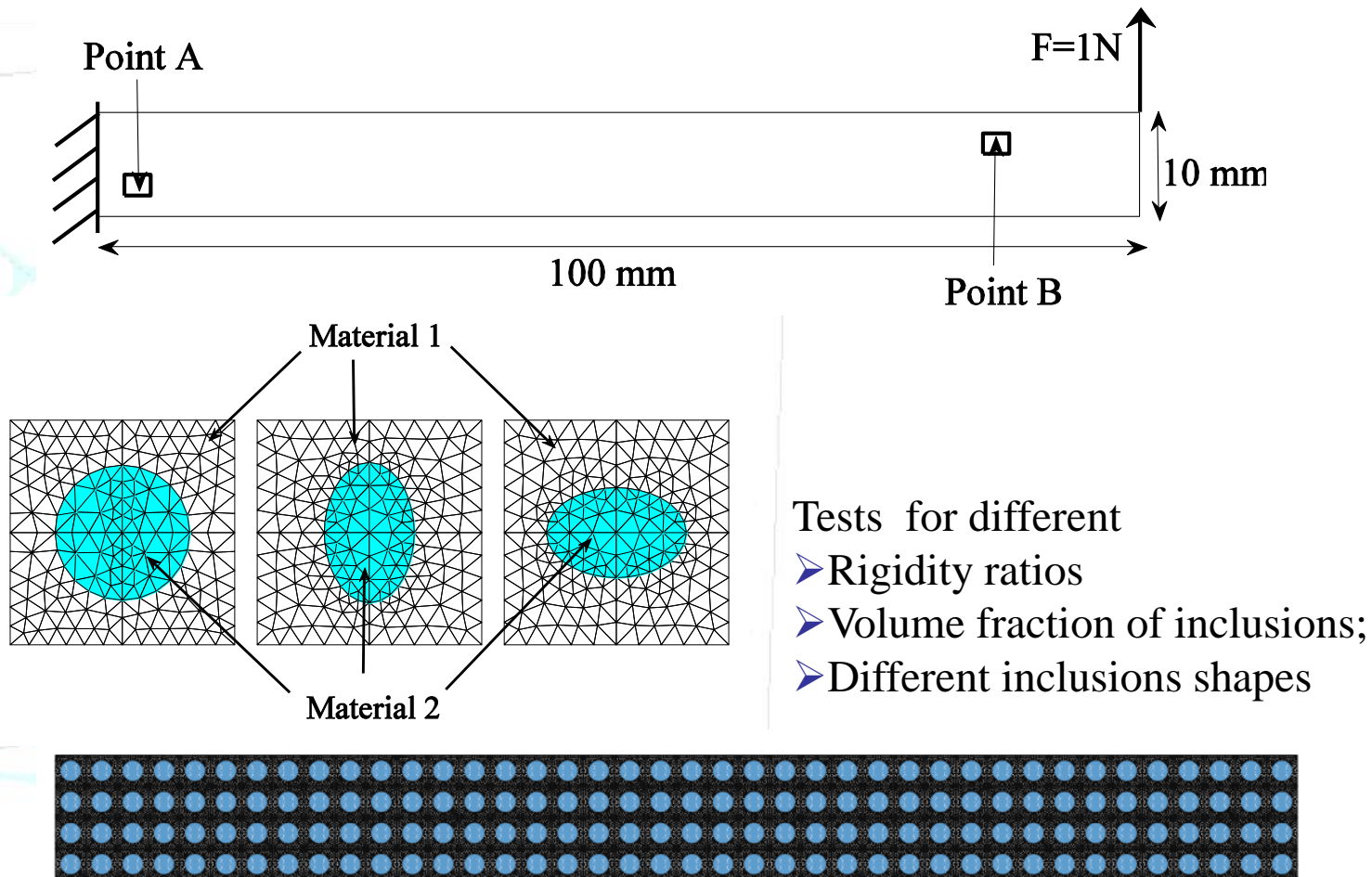
# Numerical examples

*In all examples:*

*ANM parameters:  $N = 15$        $\delta = 10^{-6}$*

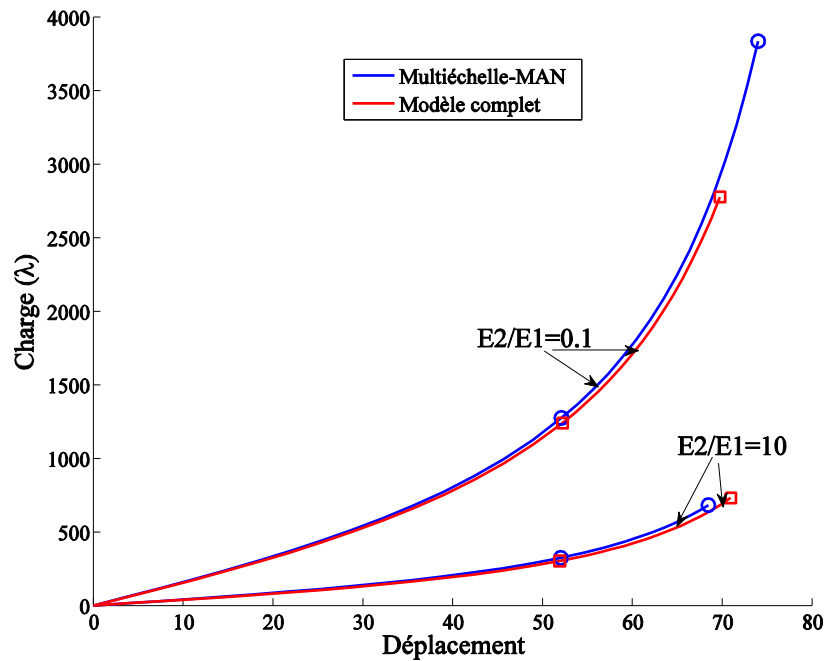
*Poisson's coefficient  $\nu = 0,3$*

# validation of the Multiscale ANM procedure: heterogeneous nonlinear elastic beam

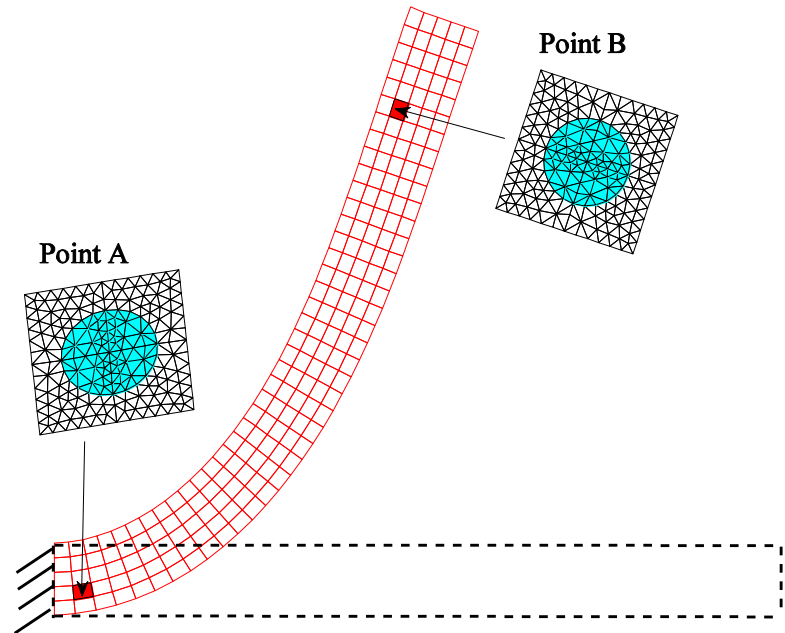


Complete mesh for comparison with the homogenized model

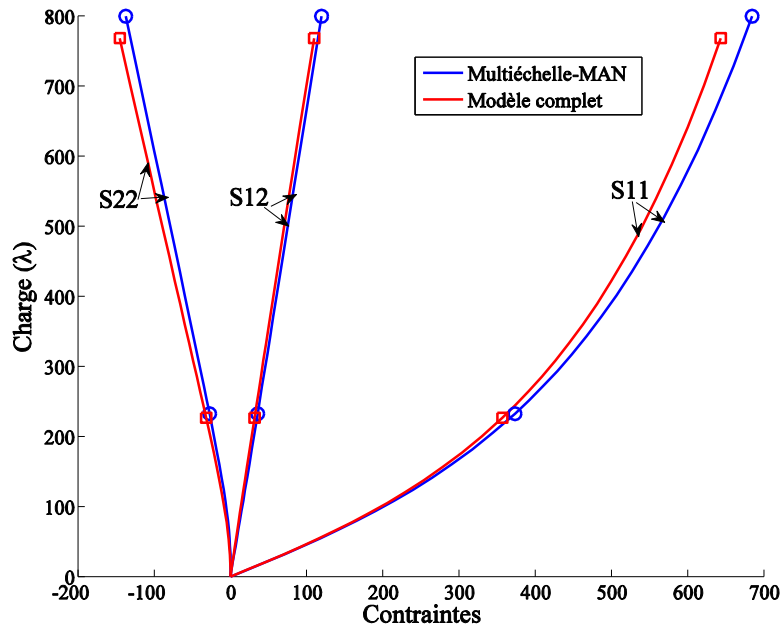
# heterogeneous nonlinear elastic beam



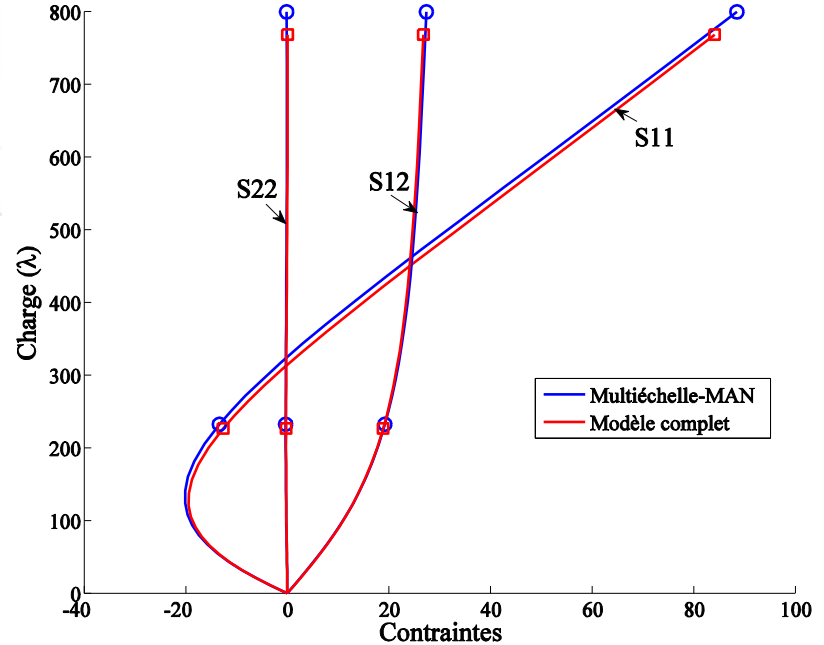
*Circular inclusion*



# heterogeneous nonlinear elastic beam



*Point A*



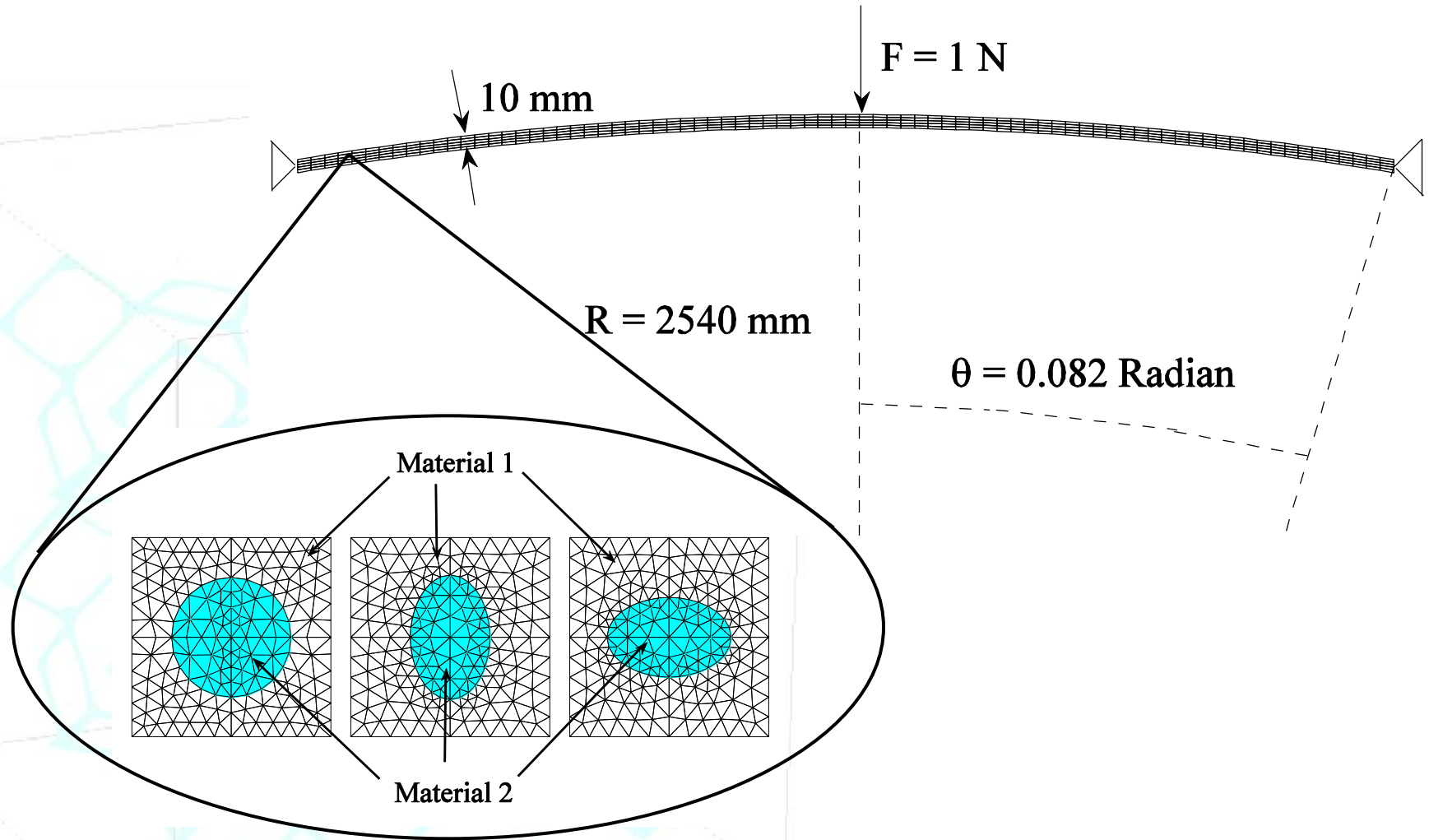
*Point B*

*Circular inclusion, rigidity ratio  $E_2/E_1 = 10$*

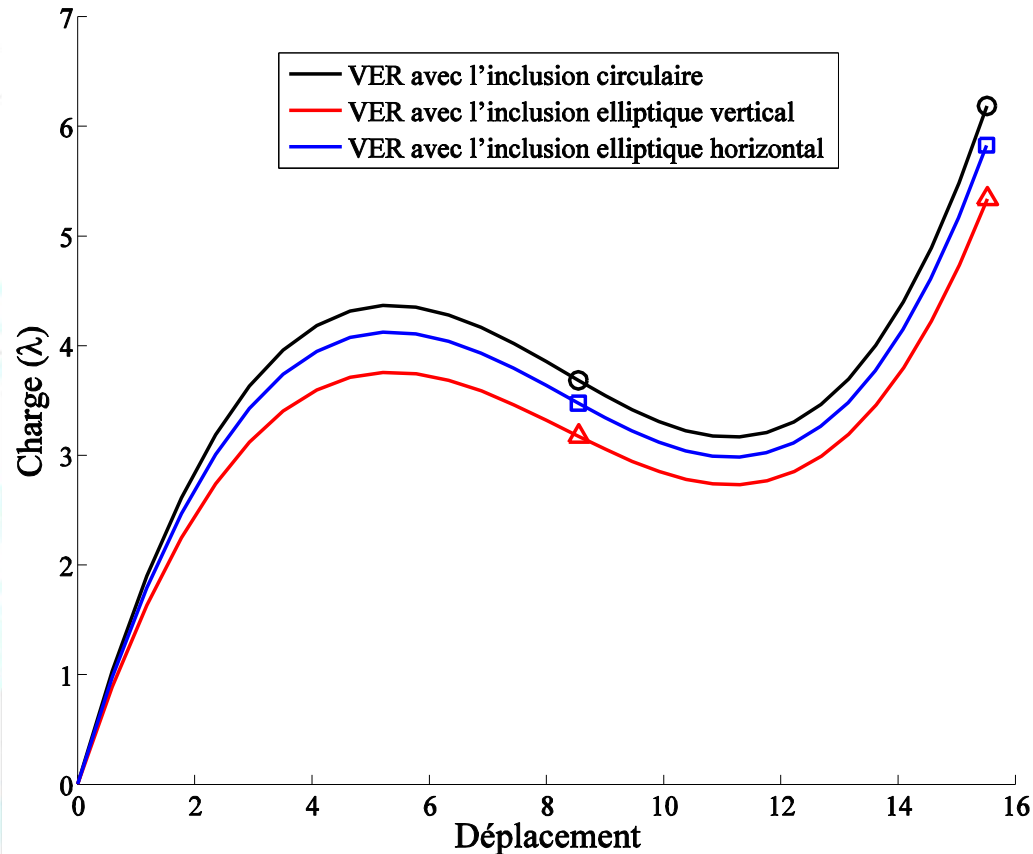
# Heterogeneous nonlinear problems with buckling

- **Macroscopic buckling: curved roof**
- **Microscopic buckling: cellular microstructure**

# A curved roof with heterogeneous microstructure

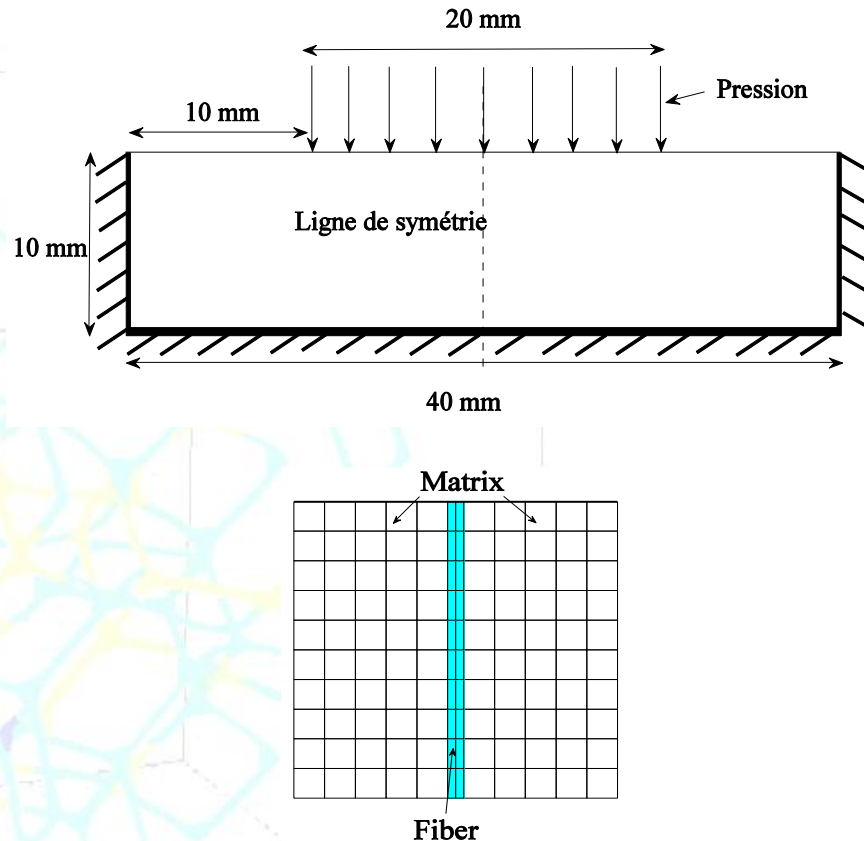


# A curved roof with heterogeneous microstructure



$$E_2/E_1=10$$

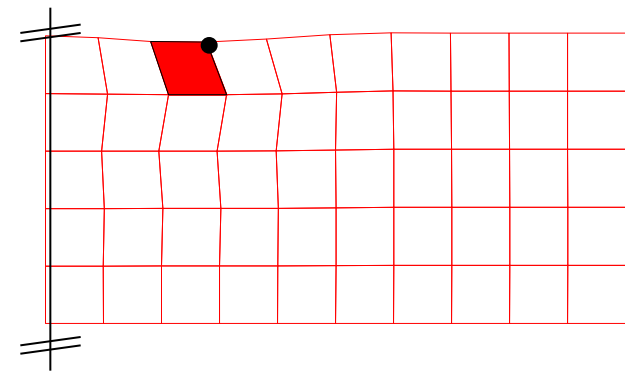
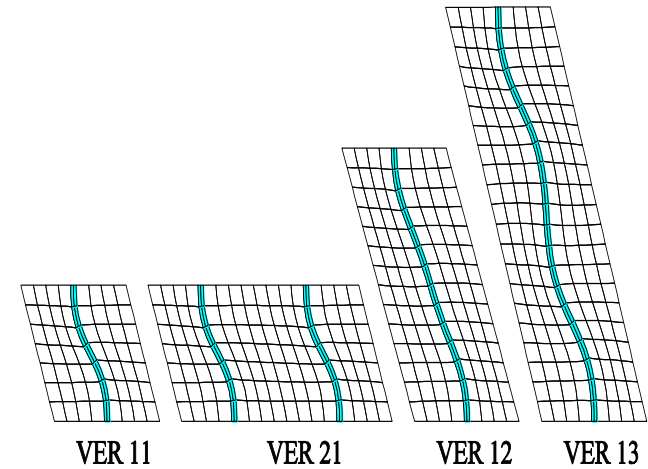
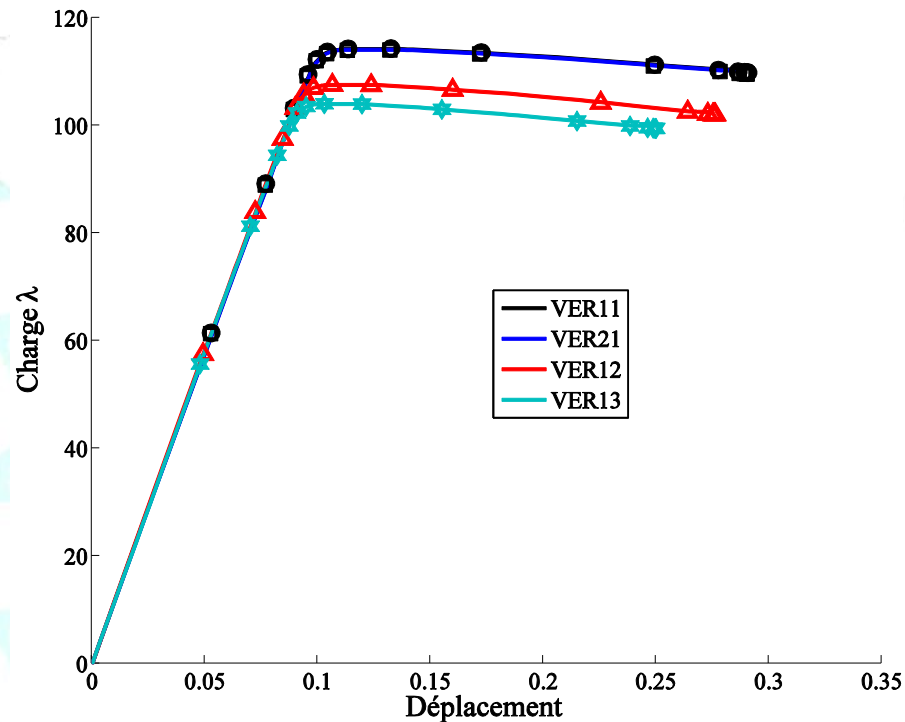
# Composite structure with long fibers in compression



*Rigidity ratio  $E_{Fiber}/E_{Matrix}=1000$ )*

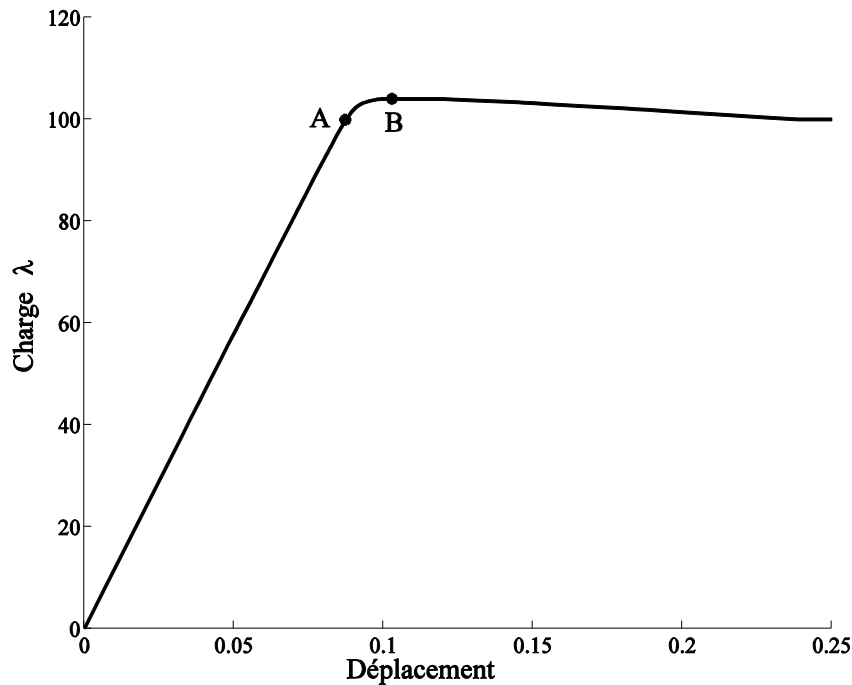


# Composite structure with long fibers in compression

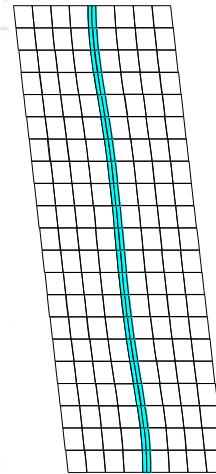


**Choice of the RVE: numerical convergence tests**

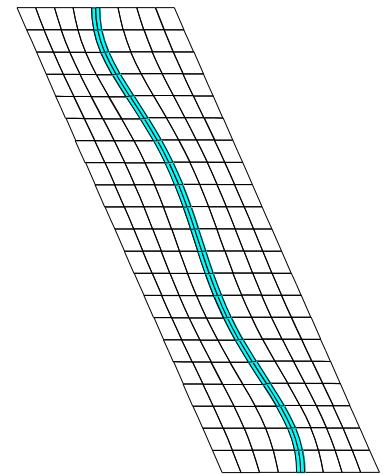
# Composite structure with long fibers in compression



Macroscopic  
response



Point A

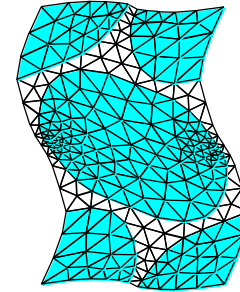
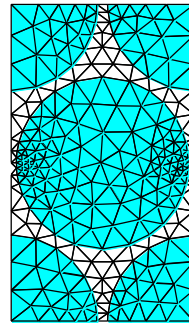
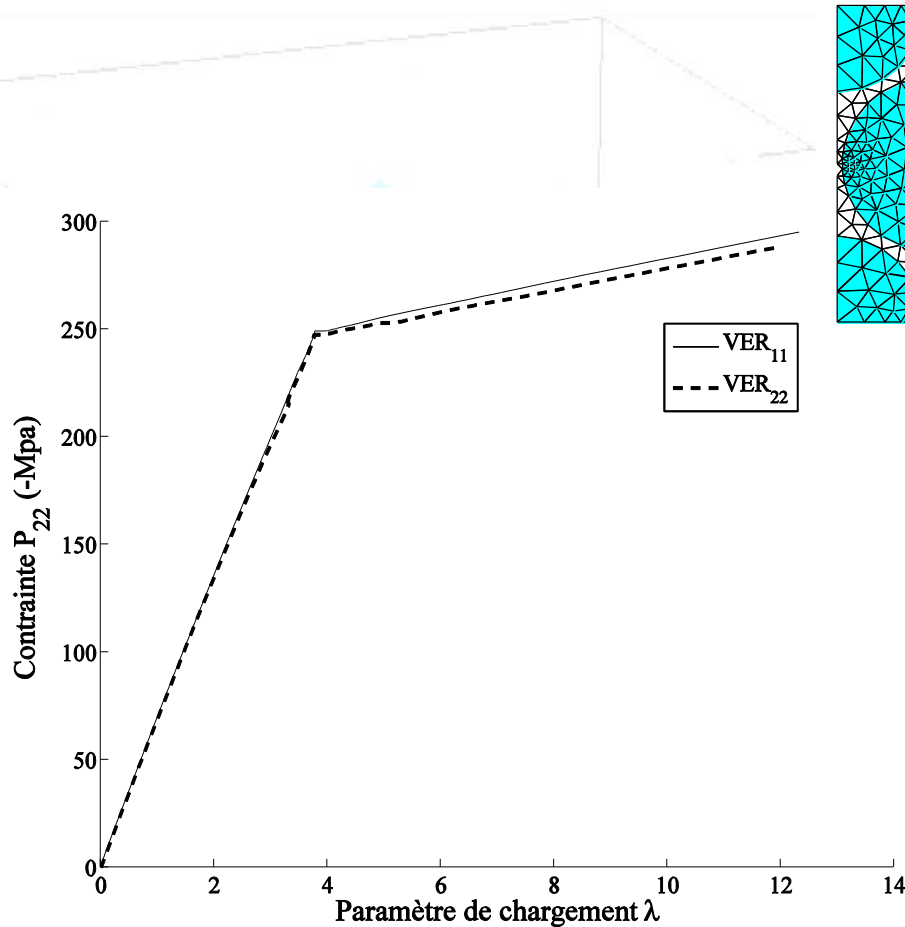


Point B

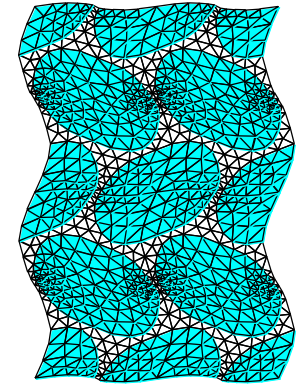
RVE deformation for  
macroscopic loads A and B

# Composite structure with cellular microstructure in compression

## Cellular microstructure



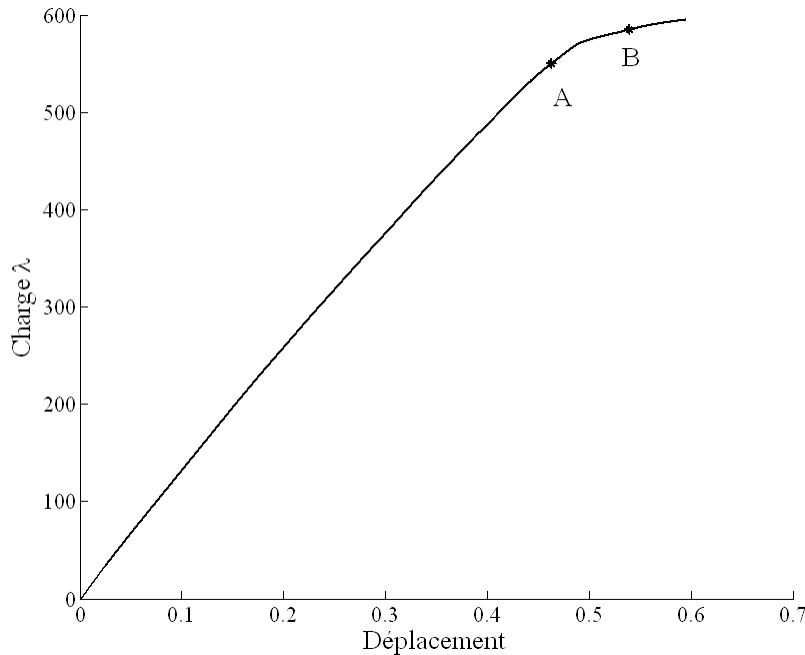
$VER_{11}$



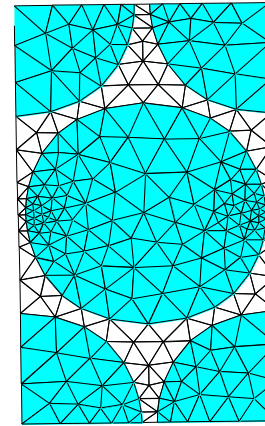
$VER_{22}$

Microscopic response  
(convergente tests)

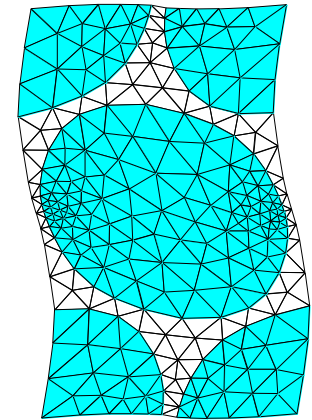
# Composite structure with cellular microstructure in compression



Macroscopic response  
(structure)



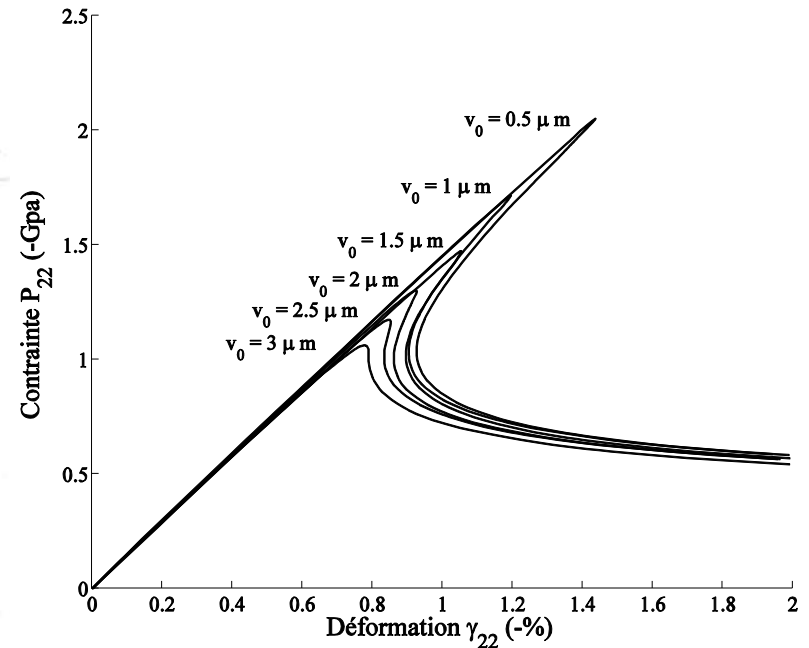
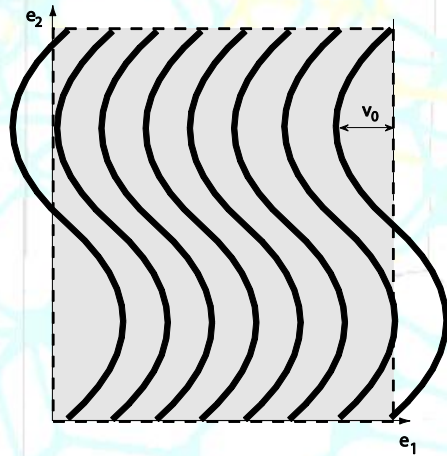
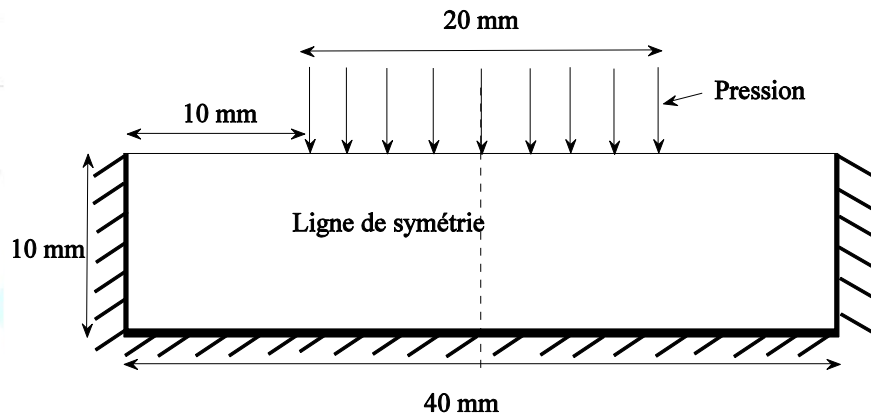
Point A



Point B

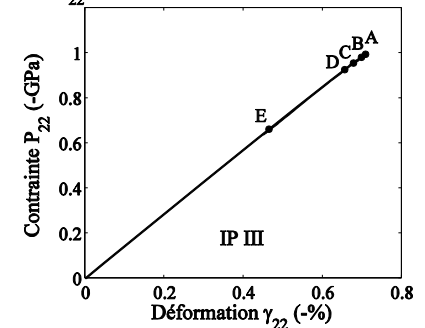
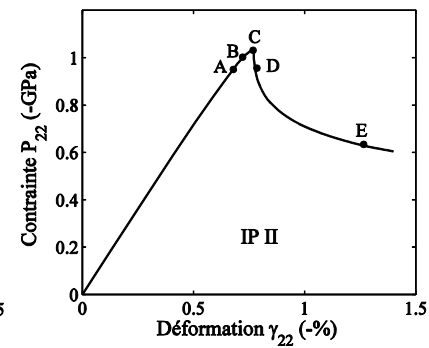
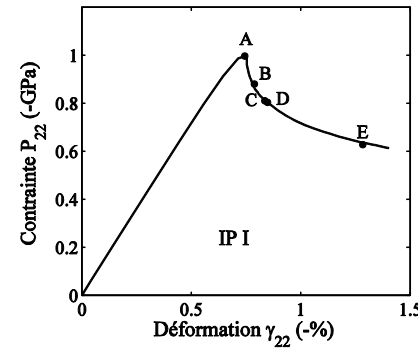
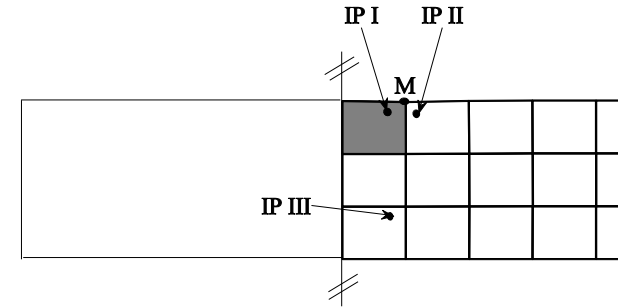
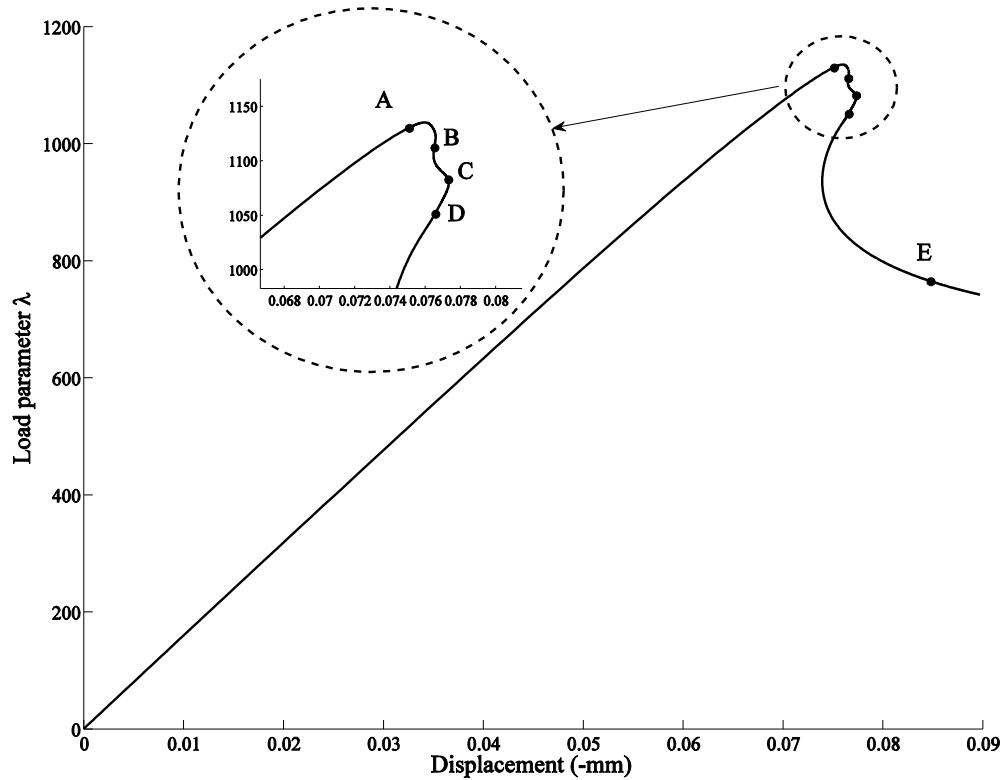
RVE deformation for  
macroscopic loads A and B

# Local buckling of composite with elastoplastic long fibers



[ Nezamabadi et al. In press,  
International Journal for Multiscale  
Computational Engineering 2010]

# Local buckling of composite with elastoplastic long fibers



# Conclusion

The ANM is an alternative to classical FEM/Newton-Raphon algorithms for solving nonlinear problems

**Removes issues of NR when bifurcation and limit points occur**

In some cases significantly **reduces computational costs**

(other advantages – not presented here – construction of bifurcation indicators)

**Nonlinear problems** are reformulated into a **sequence of linear problems**: advantages for homogenization problem: the superposition principle can be applied – well-known procedures for **linear homogenization can be applied at each order**

**Instabilities at several scales** can be handled



**Thank you for your attention !**



# References

*Nezamabadi S., Zahrouni H., Yvonnet J., Potier-Ferry M., A multiscale finite element approach for buckling analysis of elastoplastic long fibre composites, accepted in International Journal for Multiscale Computational Engineering, April 2009.*

*Nezamabadi S., Yvonnet J., Zahrouni H., Potier-Ferry M., A multilevel computational strategy for handling microscopic and macroscopic instabilities, Computer Methods in Applied Mechanics and Engineering, 198:2099-2110 (2009).*

*Yvonnet J., Zahrouni H., Potier-Ferry M., A model reduction method for the post-buckling analysis of cellular microstructures, Computer Methods in Applied Mechanics and Engineering, 197, 265-280 (2007).*

More references about previous papers on ANM can be found in these references