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# Simulations numériques d'instabilités thermo-convectives de fluides à seuil (modèle de Bingham) par Méthode Asymptotique Numérique

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# Key issues to be overcome

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- Visco-plastic (yield stress) modelling:
  - Augmented lagrangian techniques (Dean, Glowinski, Guidoboni, 2007)
  - Regularisation techniques (Bercovier & Engelman, 1980), (Frigaard & Nouar, 2005), etc.
- A reliable and computationally efficient coupled velocity-pressure formulation:
  - Discontinuous strain: only local approximations
  - Highly non-linear behaviour: challenging implementation of non-linear algorithm (tricky jacobian, etc.)
  - High computational costs in 3D ( $\text{CPU} \approx \text{Neq} \cdot L_b^2$ )
    - $\text{Cost\_Augmented Lagrangian} \approx 16 \times \text{Cost\_Regularization}$

# Governing equations

- Conservation equations

$$\vec{\nabla} \cdot \vec{V} = 0$$

$$\vec{V} \cdot \nabla \vec{V} = -\vec{\nabla} p + \frac{1}{Re} \vec{\nabla} \cdot \bar{\bar{\tau}} + \frac{Ra}{Re^2 Pr} \theta \vec{e}_z$$

$$\vec{V} \cdot \vec{\nabla} \theta = \frac{1}{Re Pr} \vec{\nabla} \cdot (\vec{\nabla} \theta)$$

- Constitutive equations: Bingham model

$$\begin{cases} \tau(\vec{V}) \leq Bn \iff \dot{\gamma}(\vec{V}) = 0 \\ \tau(\vec{V}) > Bn \iff \bar{\bar{\tau}}(\vec{V}) = \left[ 1 + \frac{Bn}{\dot{\gamma}(\vec{V})} \right] \bar{\bar{\dot{\gamma}}}(\vec{V}) \end{cases}$$

$$\tau(\vec{V}) = \left[ \frac{1}{2} (\bar{\bar{\tau}} : \bar{\bar{\tau}}) \right]^{\frac{1}{2}} \quad ; \quad \dot{\gamma}(\vec{V}) = \left[ \frac{1}{2} (\bar{\bar{\dot{\gamma}}} : \bar{\bar{\dot{\gamma}}}) \right]^{\frac{1}{2}}$$

$$Re = \frac{\rho V_{ref} L_{ref}}{\mu_0} \quad ; \quad Ra = \frac{g \beta \Delta T_{ref} L_{ref}^3}{\nu \alpha} \quad ; \quad Pr = \frac{\nu}{\alpha} \quad ; \quad Bn = \frac{\tau_0 L_{ref}}{\mu_0 V_{ref}} \quad ; \quad \theta = \frac{T - T_{ref}}{\Delta T_{ref}} \quad 3$$

# Regularization techniques

$$\bar{\tau}(\vec{V}) = \left[ 1 + \frac{Bn}{\dot{\gamma}_\eta(\vec{V})} \right] \bar{\dot{\gamma}}(\vec{V})$$

- Simple regularization:

$$\dot{\gamma}_\eta(\vec{V}) = \left[ \frac{1}{2} (\bar{\dot{\gamma}} : \bar{\dot{\gamma}}) \right]^{\frac{1}{2}} + \eta$$

- Bercovier-Engelmann regularization:

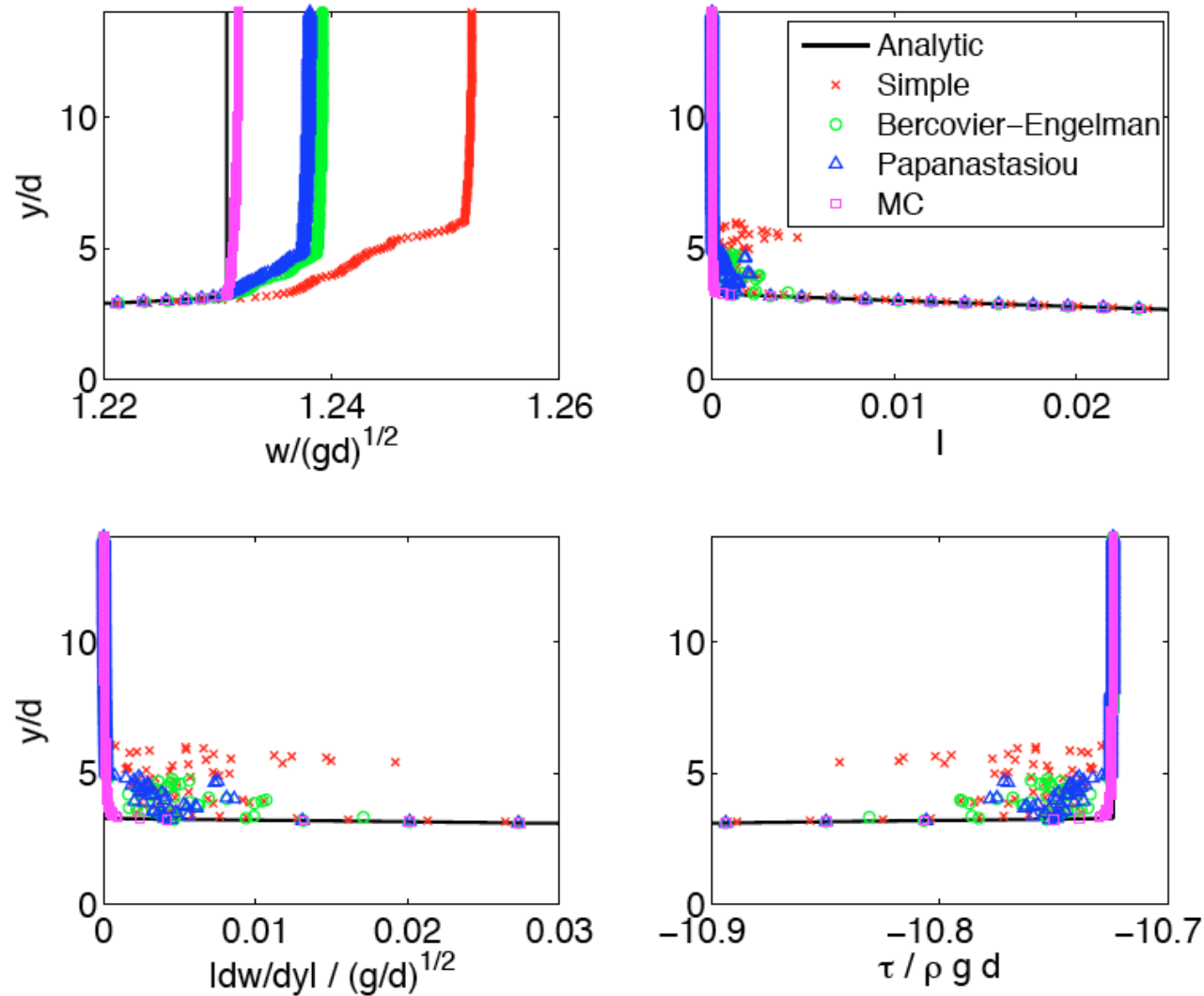
$$\dot{\gamma}_\eta(\vec{V}) = \left[ \frac{1}{2} (\bar{\dot{\gamma}} : \bar{\dot{\gamma}}) + \eta^2 \right]^{\frac{1}{2}}$$

- Papanastasiou regularization:

$$\frac{Bn}{\dot{\gamma}} = \frac{Bn(1 - e^{-\frac{\dot{\gamma}}{\eta}})}{\dot{\gamma}}$$

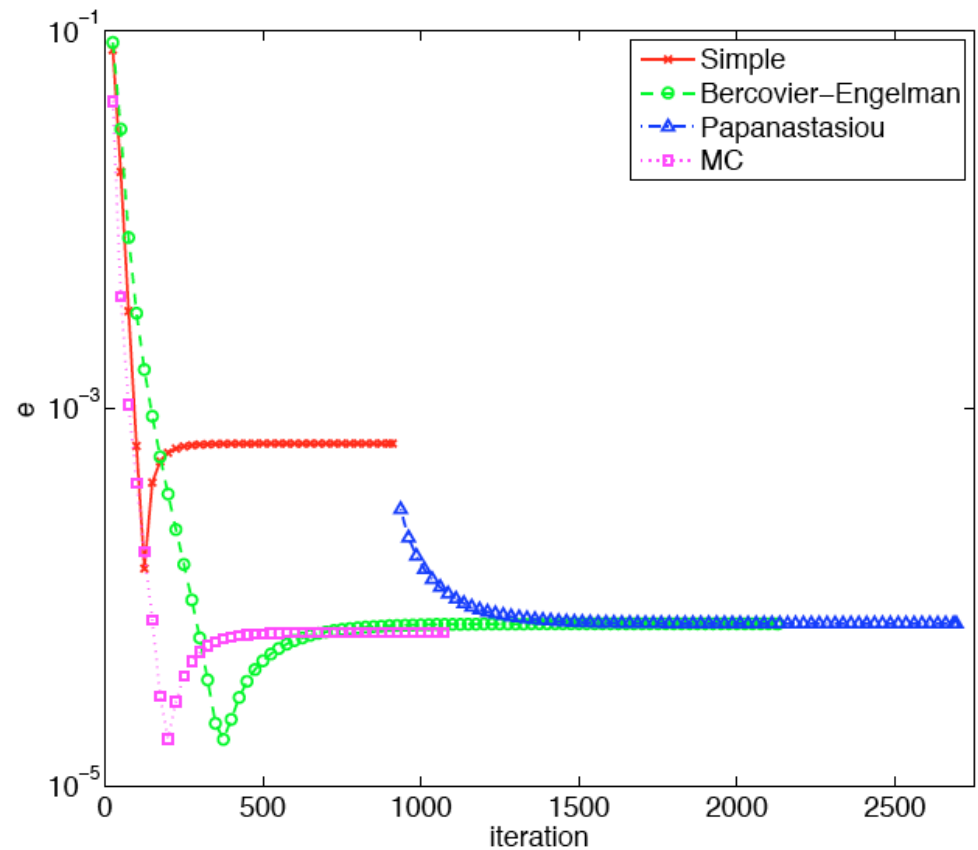
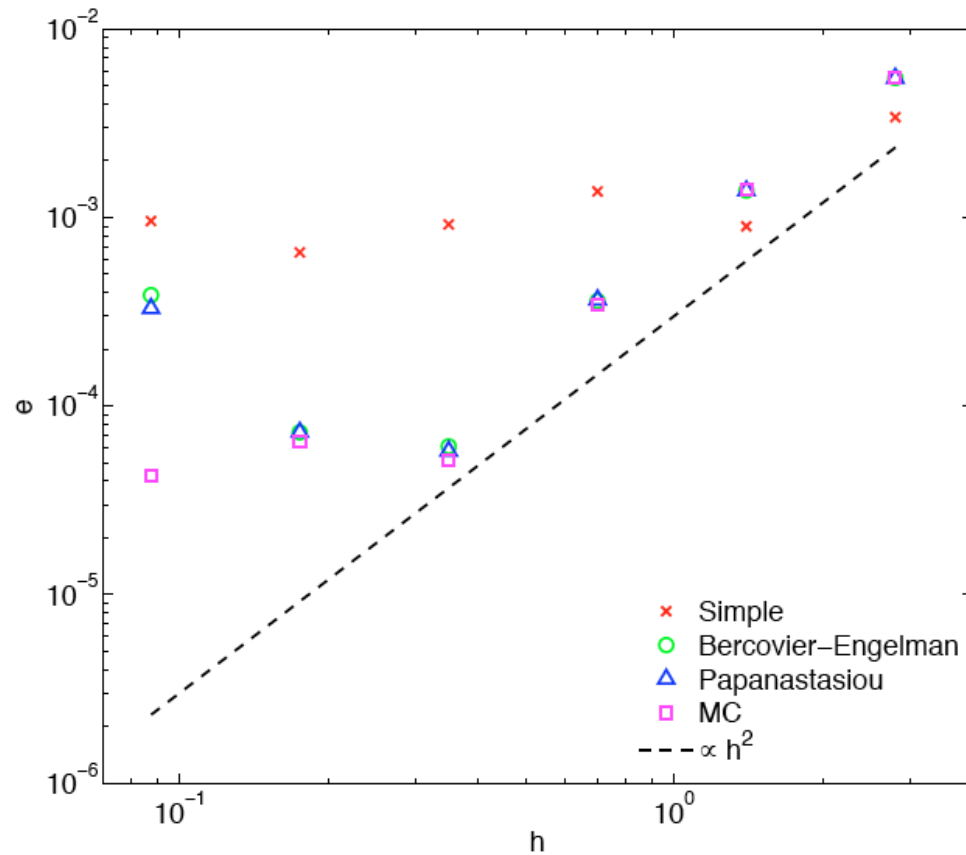
# Regularization results (1/2)

Vertical chute flow: Error localization (4/5)



→ All regularizations except the proposed one present spurious oscillations on  $\frac{dw}{dy}$  and  $\tau$ .

# Regularization accuracy & perf



- Simple regularization is the worst one in terms of error
- The other regularizations have qualitatively the same behavior
- Our method exhibits a better behavior for the finest mesh ( $e / 10$ )
- Simple and proposed regularizations converge faster than BE and Papanastasiou



# ANM continuation algorithm

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- Bifurcation diagram usually consists in performing a sequence of two steps:
  - Compute base state associated to a given value of control parameters;
  - Compute linear stability of the given base state
- Continuation is performed with our steady state ANM continuation algorithm<sup>1</sup>:
  - Fully coupled velocity-pressure (Q2-Q-1 approx., H27 FE)
  - No stabilization

<sup>1</sup> B. Cochelin and M. Medale. Power series analysis as a major breakthrough to improve the efficiency of Asymptotic Numerical Method in the vicinity of bifurcations. J. Comput. Phys. Vol. 236, 594-607, 2013.



# Asymptotic Numerical method (1)

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General non-linear quadratic form:

$$R(U, \lambda) = L(U) + Q(U, U) - \lambda F = 0 \quad (1)$$

Expand unknowns  $(U, \lambda)$  with respect to path parameter  $s$ :

$$\left\{ \begin{array}{l} U(s) = U_0 + s U_1 + s^2 U_2 + s^3 U_3 + s^4 U_4 + \dots + s^n U_n \quad (2) \\ \lambda(s) = \lambda_0 + s \lambda_1 + s^2 \lambda_2 + s^3 \lambda_3 + s^4 \lambda_4 + \dots + s^n \lambda_n \quad (3) \\ (U_0, \lambda_0) \text{ been known.} \end{array} \right.$$

Parameterizations:

- Bifurcation type:  $s = (U - U_0) \cdot U_1 + (\lambda - \lambda_0) \lambda_1$  (Cochelin, 1994)
- Minimal residual:  $\lambda_n = \text{Min}(\text{Res}(n+1))$  (Lopez, 2000) (4)



# Asymptotic Numerical Method (2)

Inserting (2-3) in (1, 4) and equating like powers of s:

$$\begin{aligned} \text{Order 1} & \begin{cases} L_t(U_1) = \lambda_1 F \\ \langle U_1, U_1 \rangle + \lambda_1^2 = 1 \end{cases} \\ \text{Order 2} & \begin{cases} L_t(U_2) = \lambda_2 F - Q(U_1, U_1) \\ \langle U_1, U_2 \rangle + \lambda_1 \lambda_2 = 0 \end{cases} \\ \vdots & \\ \text{Order n} & \begin{cases} L_t(U_n) = \lambda_n F - \sum_{r=1}^{n-1} Q(U_r, U_{n-r}) \\ \langle U_1, U_n \rangle + \lambda_1 \lambda_n = 0 \end{cases} \end{aligned}$$

FEM Spatial Discretization:

$$\begin{aligned} \text{Order 1} & \begin{cases} [K_t(U_0)] \{U_1\} = \lambda_1 \{F\} \\ \{U_1\}^t \{U_1\} + \lambda_1^2 = 1 \end{cases} \\ \text{Order 2} & \begin{cases} [K_t(U_0)] \{U_2\} = \lambda_2 \{F\} - \{FQ(1)\} \\ \{U_1\}^t \{U_2\} + \lambda_1 \lambda_2 = 0 \end{cases} \\ \vdots & \\ \text{Order n} & \begin{cases} [K_t(U_0)] \{U_n\} = \lambda_n \{F\} - \{FQ(n-1)\} \\ \{U_1\}^t \{U_n\} + \lambda_1 \lambda_n = 0 \end{cases} \end{aligned}$$

# Asymptotic Numerical Method (3)

Asymptotic expansion coefficients:

$$I_\tau = \int_{\Omega} \delta \bar{\bar{\varepsilon}} : \left[ (1 + D_0) \bar{\bar{\gamma}}_p + D_p \bar{\bar{\gamma}}_0 + \sum_{r=1}^{p-1} D_r \bar{\bar{\gamma}}_{(p-r)} \right] d\Omega$$

$$D_0 = \frac{Bn}{\dot{\gamma}_0} ; \quad D_1 = -\frac{D_0 \dot{\gamma}_1}{\dot{\gamma}_0} ; \quad D_p = -\frac{1}{\dot{\gamma}_0} \left[ D_0 \dot{\gamma}_p + \sum_{r=1}^{p-1} D_r \dot{\gamma}_{(p-r)} \right]$$

$$\dot{\gamma}_0 = \left[ \frac{1}{2} (\bar{\bar{\gamma}}_0 : \bar{\bar{\gamma}}_0) + \eta^2 \right]^{\frac{1}{2}} ; \quad \dot{\gamma}_1 = \frac{\bar{\bar{\gamma}}_0 : \bar{\bar{\gamma}}_1}{2\dot{\gamma}_0} ; \quad \dot{\gamma}_p = \frac{1}{2\dot{\gamma}_0} \left[ \bar{\bar{\gamma}}_0 : \bar{\bar{\gamma}}_p + \frac{1}{2} \sum_{r=1}^{p-1} \bar{\bar{\gamma}}_r : \bar{\bar{\gamma}}_{(p-r)} - \sum_{r=1}^{p-1} \dot{\gamma}_r \dot{\gamma}_{(p-r)} \right]$$

$$I_\tau = \int_{\Omega} \delta \bar{\bar{\varepsilon}} : \left[ (1 + D_0) \bar{\bar{\gamma}}_p - \frac{\bar{\bar{\gamma}}_0}{\dot{\gamma}_0} \left\{ \frac{D_0}{2\dot{\gamma}_0} \left( \bar{\bar{\gamma}}_0 : \bar{\bar{\gamma}}_p + \frac{1}{2} \sum_{r=1}^{p-1} \bar{\bar{\gamma}}_r : \bar{\bar{\gamma}}_{(p-r)} - \sum_{r=1}^{p-1} \dot{\gamma}_r \dot{\gamma}_{(p-r)} \right) + \sum_{r=1}^{p-1} D_r \dot{\gamma}_{(p-r)} \right\} + \sum_{r=1}^{p-1} D_r \bar{\bar{\gamma}}_{(p-r)} \right] d\Omega$$

# Bifurcation point detection

- In the ANM geometrical power series arises close to BP
- Geometric progression detection algorithm:

for  $n - 3 \leq p \leq n - 1$

$$\alpha_p = (U_p \cdot U_n) / (U_n \cdot U_n)$$

$$U_p^\perp = U_p - \alpha_p U_n$$

$$\text{if } \sum_{p=n-3}^{n-2} \left( (\alpha_p^{1/(n-p)} - \alpha_{n-1}) / \alpha_{n-1} \right)^2 < \varepsilon_{gp1}$$

$$\text{and } \sum_{p=n-3}^{n-1} \|U_p^\perp\| / \|U_p\| < \varepsilon_{gp2}$$

- If test satisfied (colinearity, proportionality), one computes
  - its common ratio:  $\alpha = \alpha_{n-1} \approx 1/r$
  - its scale factor:  $U_{t_1}^\perp = U_n / \alpha^n$

# Bifurcation point computation branch switching

- Solution at bifurcation point:

$$U_{BP} = U(s = \frac{1}{\alpha}) = U_0 + \frac{1}{\alpha} \hat{U}_1 + \frac{1}{\alpha^2} \hat{U}_2 + \cdots + \frac{1}{\alpha^{n-1}} \hat{U}_{n-1}$$

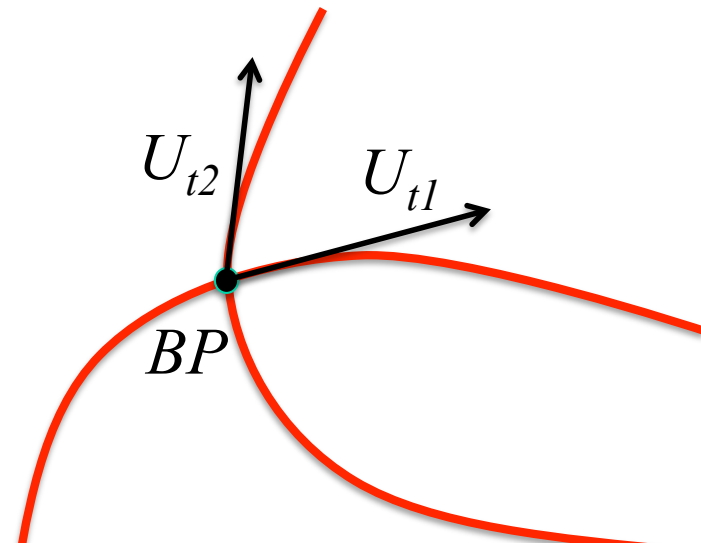
with  $\hat{U}_p = U_p - \alpha^p U_{t_1}^\perp$

- Tangent vector to branch 1:

$$U_{t_1} = \frac{\partial U}{\partial s}(s = \frac{1}{\alpha}) = \hat{U}_1 + \frac{2}{\alpha} \hat{U}_2 + \cdots + \frac{n-1}{\alpha^{n-2}} \hat{U}_{n-1}$$

- Tangent vector to branch 2:

$$U_{t_2} = \beta U_{t_1} + \gamma U_{t_1}^\perp$$
$$\psi^T R_{,UU}^c U_{t_2} U_{t_2} = 0$$





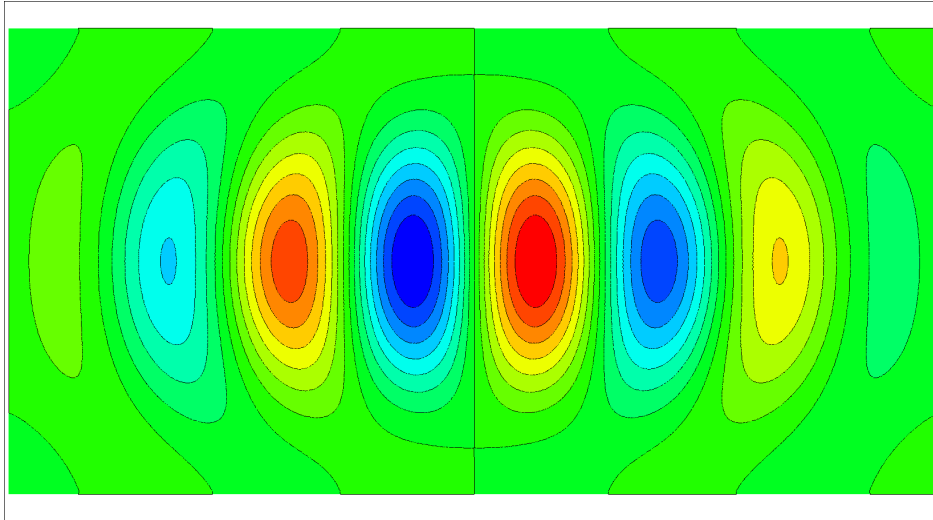
# Confined Rayleigh-Benard

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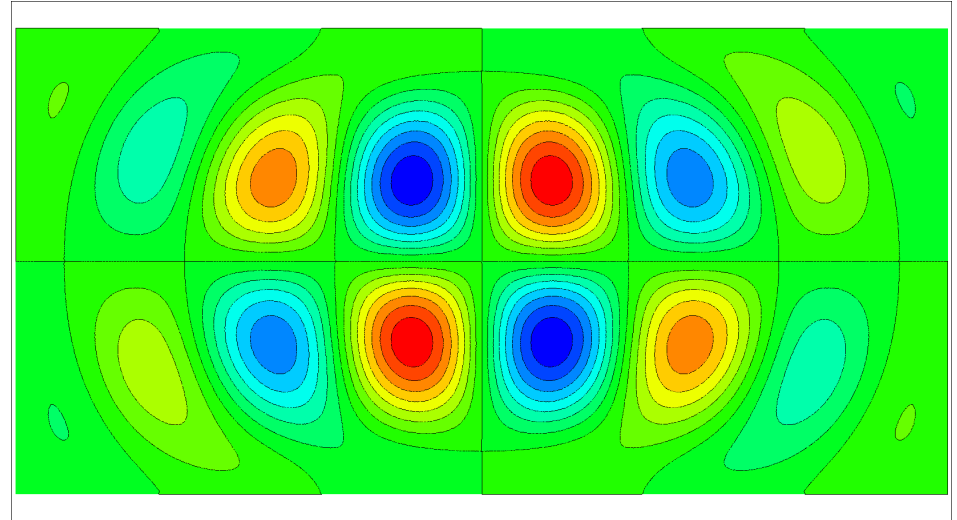
- Parallelepipedic computational domain:  $L/h=10$ ;  $l/h=4$
- Spatial discretization:  $180 \times 72 \times 18$  H27 Q2 FE (7 747 060 dof)
- ANM parameters:  $10 \leq n \leq 50$ ;  $\delta = 10^{-9}$ ;  $\varepsilon_{gp1} = 10^{-3}$  and  $\varepsilon_{gp2} = 10^{-6}$ ;  
1 NR end-of-step correction if  $\text{Res} > 10^{-6}$
- HPC: Petsc + MUMPS + BULL Bullx S6010 supercomputer  
(128 cores, 2 To RAM)
  - fact. time: 45 mn (on 64 cores)
  - av. cont. step ( $n \approx 50$ ): 50 mn (on 64 cores)

# Rayleigh-Benard in a box

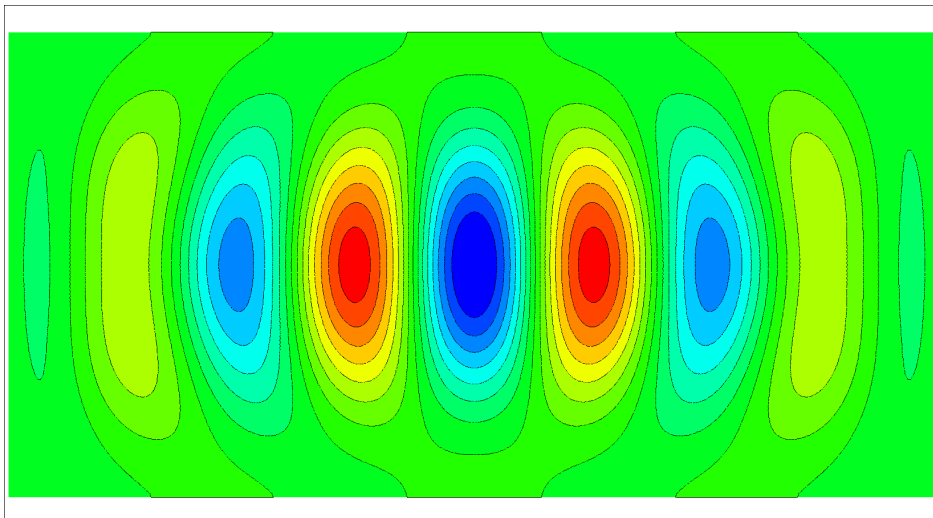
Newtonian fluid:  $Pr=9$



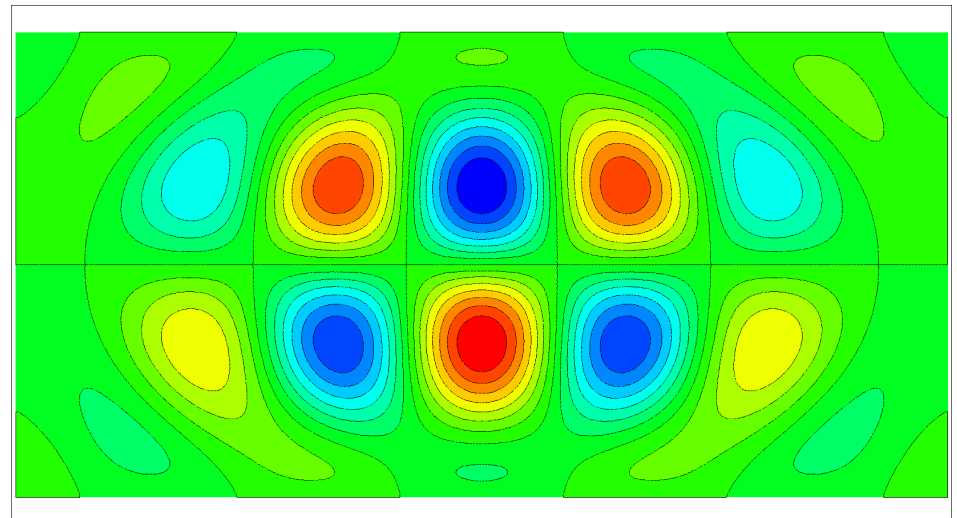
1st bifurcation mode:  $Ra_{c1}=1747.4$



3rd bifurcation mode:  $Ra_{c3}=1773.9$



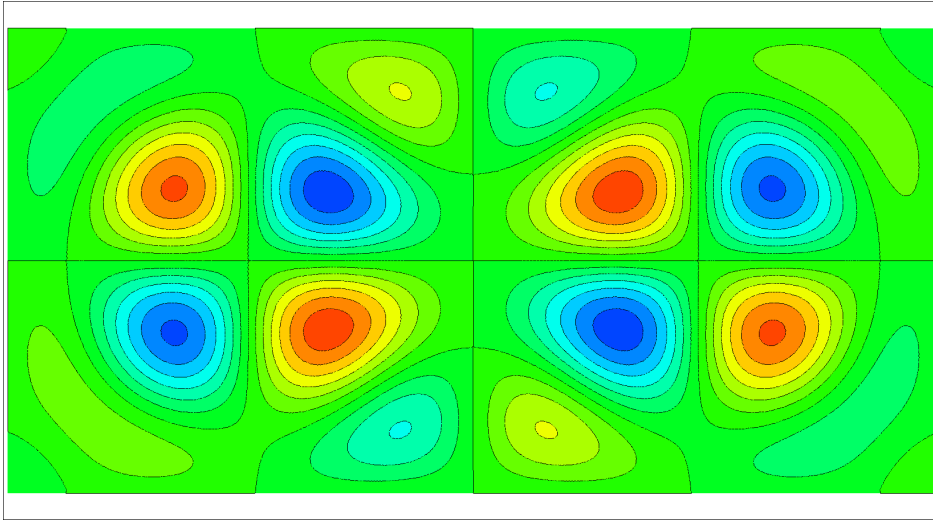
2nd bifurcation mode:  $Ra_{c2}=1748.8$



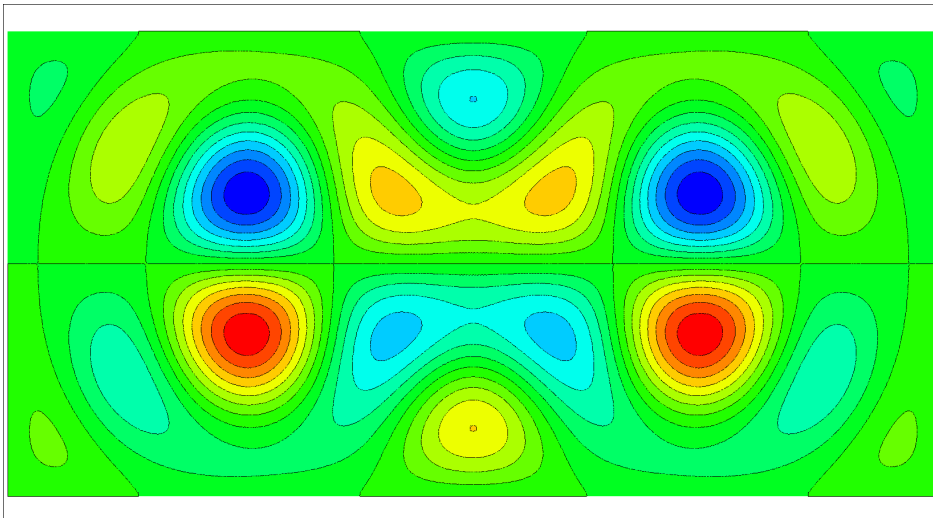
4th bifurcation mode:  $Ra_{c4}=1805.7$

# Rayleigh-Benard in a box

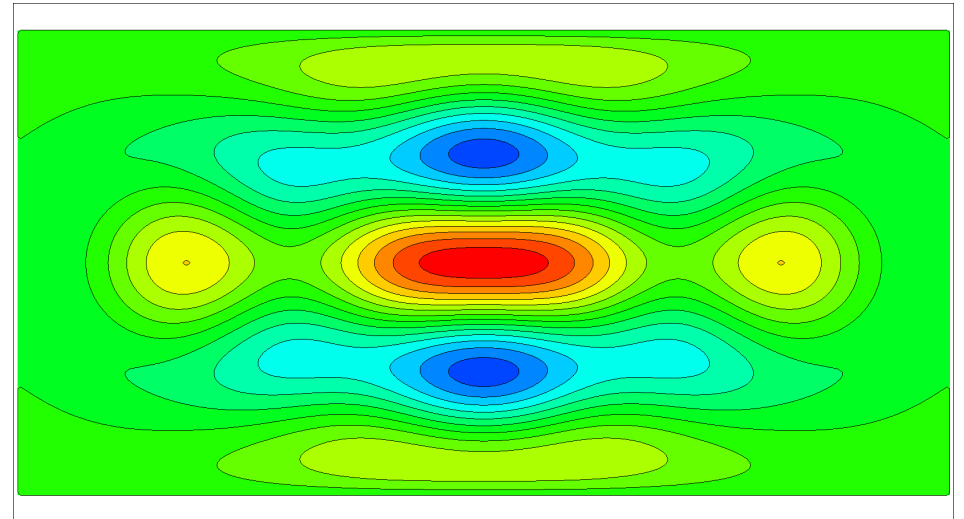
Newtonian fluid:  $Pr=9$



5th bifurcation mode:  $Ra_{c5}=1810.9$



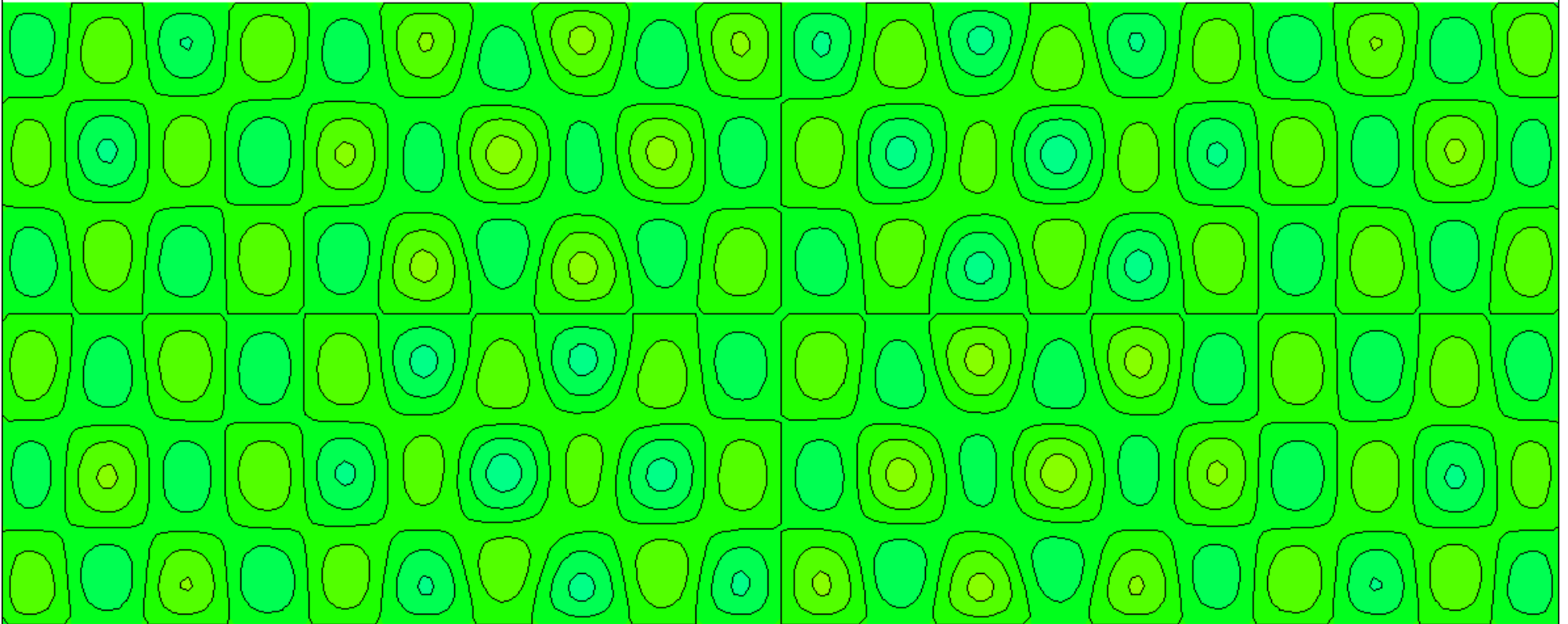
6th bifurcation mode:  $Ra_{c6}=1812.9$



7th bifurcation mode:  $Ra_{c7}=1817.2$

# Rayleigh-Benard in a box

Bingham fluid:  $Pr=9$ ,  $Bn=1$







# Summary

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- 3D Steady-State Solver for incompressible fluid flows
- Continuation method (ANM predictor - NR corrector) to perform bifurcation diagram and accurately locate bifurcation points
- Efficient and scalable parallel implementation for problems up to several millions of dof
- Rayleigh-Benard in a box ( $\Gamma_x=10$ ,  $\Gamma_y=4$ )
  - Detailed results for Newtonian fluids ( $Pr=9$ )
  - Preliminary results for a Bingham fluid ( $Pr=9$ ,  $Bn=1$ )
- Future works:
  - Nature of bifurcations, linear stability analysis
  - High multiplicity bifurcation points, Hopf bifurcation



# Asymptotic Numerical Method (6)

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ANM : Perturbation method + F.E.M.

- Optimal step length (Cochelin, 1994, 2003) :

$$s_{\text{opt}} = (\varepsilon \|F_1\| / \|F_{\text{nl}}(n+1)\|)^{(1/n+1)}$$

- Enables us to describe one part of the branch
- Loop with new starting point ( $U(s=s_{\text{opt}}), \lambda(s=s_{\text{opt}})$ )

High order predictor method, Newton based corrector

Main features:

- Parameter free continuation technique ( $\varepsilon, n$ )
- Analytical representation of quadratic non-linearities
- Power series contain substantial informations