

# Residual Distribution

basics,  
recent developments,  
relations with other techniques

## PART II



Centre d'Été de Mathématiques et de Recherche Avancée  
en Calcul Scientifique

### CEMRACS 2012

## Numerical Methods and Algorithms for High Performance Computing

**Summer School Lectures (July 16th - 20th):**

**Martin Gander (Univ. de Genève)**  
Extrapolation and Krylov Subspace Methods for Solving Linear Equations.

**Jean-Luc Guéroux (Inria AAM Univ.)**  
Massively Parallel Splitting Algorithms for the Incompressible and Slightly Compressible Navier-Stokes equations.

**Geon Hwang (CERN/EP)**  
The Discontinuous Galerkin Method: Discretization, Efficient Implementation, Application to Turbulent Flows.

**Quintus Jacquot (Stanford Univ.)**  
Quantification of Uncertainties in High-Fidelity Simulations of Turbulent Reactive Flows.

**Patrick Knopik (EPFL, ETH Zurich)**  
Particle Methods.

**Fabrice Nataf (Univ. Pierre et Marie Curie - Paris 6)**  
Two-Level Domain Decomposition Methods.

**Mario Ricchiuto (Inria Bordeaux SRI-Quest)**  
Residual Distribution: Basics, Recent Developments, and Relations with Other Techniques.

**Ulrich Rüde (Univ. Erlangen-Nuremberg)**  
Parallel Multigrid Methods, Simulating Complex Flows with the Lattice Boltzmann Method.

**Workshop HPC-Enterprises (August 20th - 21st):**  
Companies and researchers will share their experience, their results and questions about HPC. Program available on CEMRACS 2012 website.

**Research Projects (July 23rd - August 24th):**  
List of projects, partners and supports available on CEMRACS 2012 website.



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Luc Girard (Inria Bordeaux)  
Patrick Klumavitskaya (CSDE, Zurich)  
Pauline Laffitte (ICP)  
Stéphane Lantieri (Inria Sophia Antipolis)  
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Marc Masson (École Centrale Paris)  
Vincent Michel (Univ. Lyon 1)

**Informations and registration:**  
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<http://www.inria.fr/cemracs2012>










# PART II OUTLINE

Higher (than second) order

Navier-Stokes and viscous problems

Time dependent problems

RD for Shallow Water simulations

Summary and perspectives

# 3

## VERY HIGH ORDER

RD schemes

## Why higher (than second) order schemes

### Main motivation

Efficiency of  $k$ th order method

$$\eta_k = \frac{1}{\text{error} \times \text{CPU}} = \eta_k^{\text{scheme}} \frac{1}{n_{\text{DoF}-k} h^k}$$

with  $\eta_k^{\text{scheme}} = 1/\text{characteristic cost}$ . The bigger  $\eta_k^{\text{scheme}}$  the better the scheme.

Increasing  $\eta_k^{\text{scheme}}$  is hard.

But we can start by increasing  $k$ , thus boosting  $\eta_k$

trying to minimize the  $n_{\text{DoF}-k}$  required for a given error level

# How to get higher (than second) order schemes

How do we get high order mesh convergence rates...

1. Polynomial approximations of arbitrary degree
2. Discretizations verifying some conditions for some error estimate to hold

CAVEAT

To get any asymptotic convergence rates, we need convergence

STABILITY plays a role

## Higher (than second) order

1.  $\forall K \in \Omega_h$  compute :  $\phi^K = \int_K \nabla \cdot \mathcal{F}_h(u_h)$

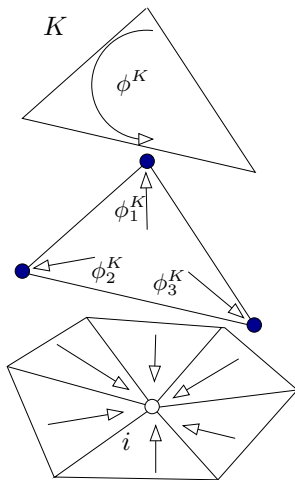
2. Distribution :  $\phi^K = \sum_{i \in K} \phi_i^K$

Distribution  
coeff.s :

$$\phi_i^K = \beta_i^K \phi^K$$

3. Compute nodal values :  
solve algebraic system

$$\sum_{T|i \in T} \phi_i^K = 0, \quad \forall i \in \Omega_h \quad (1)$$



## Higher (than second) order

### Accuracy condition

For a polynomial approximation of degree  $k$ ,  
a sufficient condition to have a  $\|\epsilon_h\| \leq C h^{k+1}$  is (in 2d)

$$\phi_i^K(w_h) = \mathcal{O}(h^{k+2}), \quad \forall K_h, \forall i \in K$$

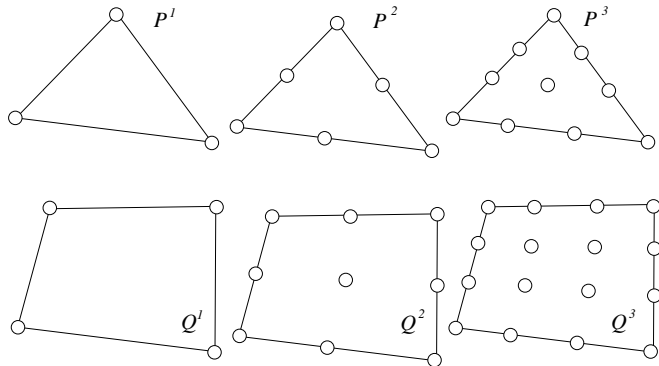
leading to the two high order prototypes

$$\phi_i^K = \int_K \omega_i^K \nabla \cdot \mathcal{F}_h(u_h), \quad \|\omega_i^K\| < C < \infty$$

$$\phi_i^K = \beta_i^K \phi^K, \quad \|\beta_i^K\| < C < \infty$$

## Higher (than second) order

### Continuous Lagrange elements





# Higher (than second) order

## RD on higher order elements

1.  $\forall K$  compute :  $\phi^K = \int_K \nabla \cdot \mathcal{F}_h(u_h)$

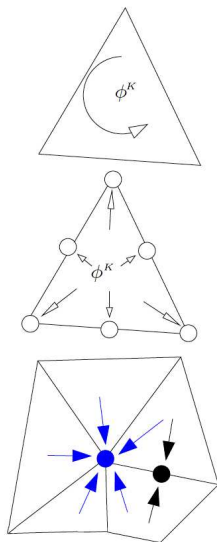
2. Distribution :  $\phi^K = \sum_{j \in K} \phi_j^K$

Distribution  
coeff.s :

$$\phi_i^K = \beta_i^K \phi^K$$

3. Evolution :  $\lim_{t \rightarrow \infty}$  of

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \phi_i^K = 0$$



## Higher (than second) order

### A naive approach

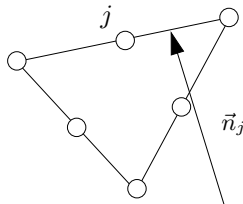
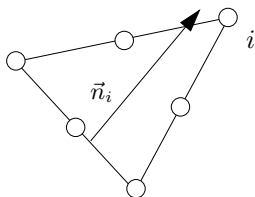
To solve the steady limit of

$$\partial_t u + \vec{a} \cdot \nabla u = 0$$

we write a scheme that “imitates” the  $P^1$  LDA scheme

$$\phi_i^K = \beta_i^K \phi^K, \quad \beta_i^K = k_i^+ \left( \sum_{j \in K} k_j^+ \right)^{-1}$$

where recall that  $k_i = \vec{a} \cdot \vec{n}_i$  defining the normals as



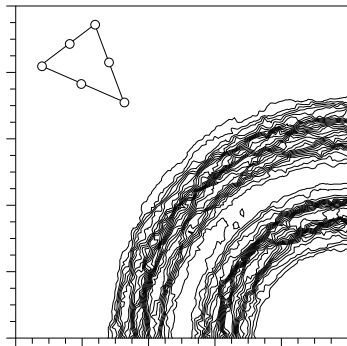
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We solve the scalar rotation problem we saw earlier and get



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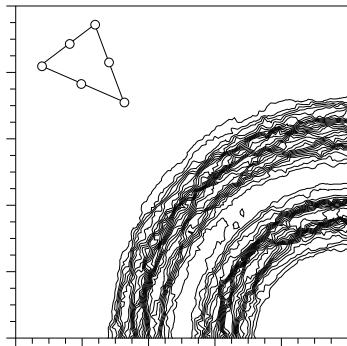
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Monotonicity ?



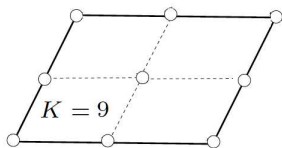
# Higher (than second) order

## A second naive approach

To solve the steady limit of

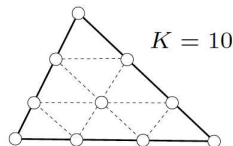
$$\partial_t u + \vec{a} \cdot \nabla u = 0$$

### 1. Generalize the LF distribution



Lax-Friedrich's (Rusanov) :

$$\phi_i^{\text{LF}} = \frac{1}{K} \phi^K + \alpha_{\text{LF}} \sum_{j \in T} (u_i - u_j)$$



LED scheme for

$$\alpha_{\text{LF}} \geq \frac{1}{2K} h \sup_{x \in K} \|\partial_u \mathcal{F}(u_h(x))\|$$

## Higher (than second) order

A second naive approach

To solve the steady limit of

$$\partial_t u + \vec{a} \cdot \nabla u = 0$$

1. Generalize the LF distribution
2. Apply the limiter as done in the  $P^1$  case

## Higher (than second) order

We proceed as in the  $P^1$  case

1. Evaluation of  $\phi^K = \oint_{\partial K} \mathcal{F}_h(u_h) \cdot \vec{n} \, dl$
2. Evaluation of  $\phi_i^{\text{LF}} = \frac{1}{K} \phi^K + \alpha_{\text{LF}} \sum_{j \in T} (u_i - u_j)$

3. Limiting :

$$\beta_i^{\text{LLF}} = \frac{\max(0, \beta_i^{\text{LF}})}{\sum_{j \in K} \max(0, \beta_j^{\text{LF}})}$$

4. Distribution :  $\phi_i^{\text{LLF}} = \beta_i^{\text{LLF}} \phi^K (= \gamma_i \phi_i^{\text{LF}}, \gamma_i \in [0, 1])$

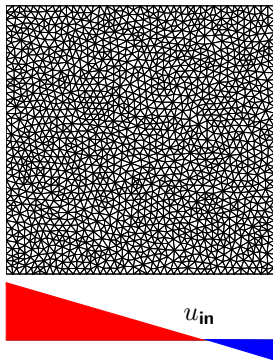
5. Evolve until steady state. Example :

$$u_i^{n+1} = u_i^n - \omega_i \sum_K \phi_i^{\text{LLF}} \xrightarrow{n \rightarrow \infty} \sum_K \phi_i^{\text{LLF}} = 0$$

## Higher (than second) order

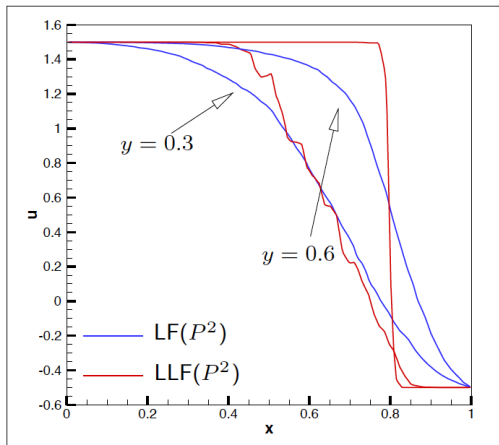
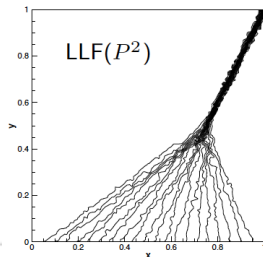
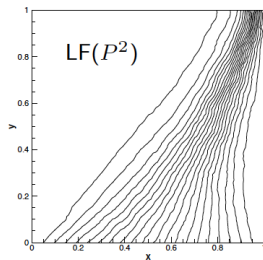
Scalar example :  $\nabla \cdot \mathcal{F}(u) = 0$  with  $\mathcal{F}(u) = (u, \frac{u^2}{2})$  and bcs

$$u(x, y = 0) = 1.5 - 2x$$





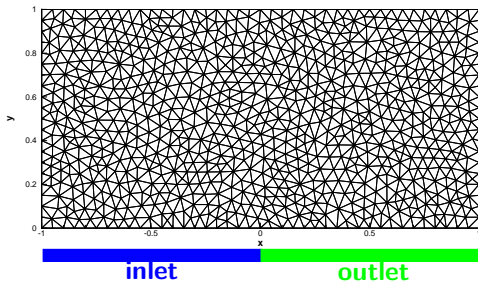
## Higher (than second) order



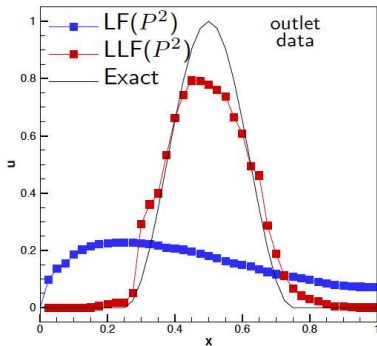
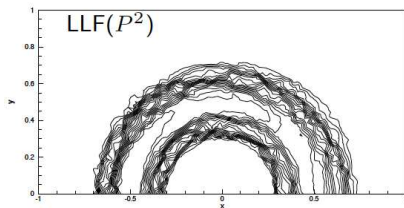
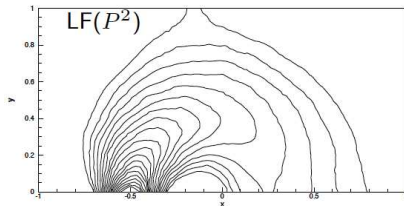
## Higher (than second) order

Scalar example :  $\vec{a} \cdot \nabla u = 0$  with  $\vec{a} = (y, 1 - x)$  and bcs

$$u_{\text{in}} = \begin{cases} \cos(2\pi(x + 0.5))^2 & \text{if } x \in [-0.75, -0.25] \\ 0 & \text{otherwise} \end{cases}$$



## Higher (than second) order



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### Structural problems

We observe two problems

1. The first is really structural
2. The second is related to stability and can be cured

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1. The first is really structural (Ricchiuto, 2011)

**Proposition.** (*Advection, spurious modes*) In 2d, any RD scheme for which  $\phi_i^K = \beta_i^K \phi^K$  applied to  $P^k$  triangles with  $k \geq 2$ , and  $Q^k$  quads with  $k \geq 1$  has spurious modes. These modes can be explicitly computed and are those for which  $\forall f \in K$

$$\int_f u_h = 0$$

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The paradigm of Roe and Deconinck  
has to be modified

## Higher (than second) order

### Structural problems

We observe two problems

1. The first is really structural
2. The second is related to stability and can be cured :  
the limited LF scheme is entirely built on the algebraic preservation of the LED condition. No upwind bias is introduced.

## Higher (than second) order

### Solution to spurious modes 1 : modify the stencil

Enlarge the stencil to compute  $\phi^K$

1. Reconstruction operator  $R_{1k}$  that maps the  $P^1$  approximation to a degree  $k$  edge continuous polynomial. This boils down to reconstructing gradients, Hessians etc. in the nodes. Explored in (Caraeni *Computers & Fluids*, 2005 ; Chou, Shu *J.Comput.Phys*, 2006 ; Hubbard *J.Comput.Phys*, 2007)
2. Conformal sub-triangulation of the element, writing the scheme by sub-cells while using the macro-cell to define the polynomial. Explored in (Abgrall, Roe *J.Sci.Comp.*, 2003 ; Ricchiuto et al. *J.Comput.Appl.Math*, 2008 ; Vymazal et al. *J.Comput.Phys*, 2011). similarities with the spectral volume of Z.J. Wang (*J.Comput.Phys*) 2002.

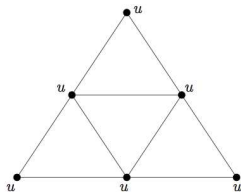


# Higher (than second) order

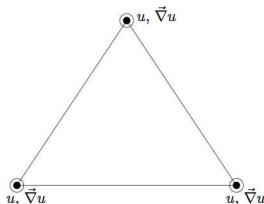
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Submesh reconstruction

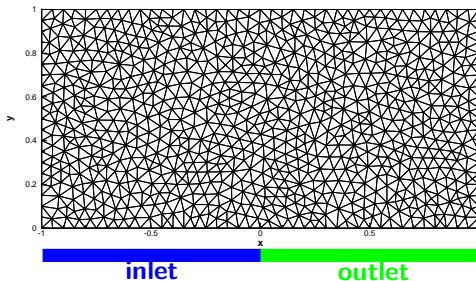


Gradient recovery

## Higher (than second) order

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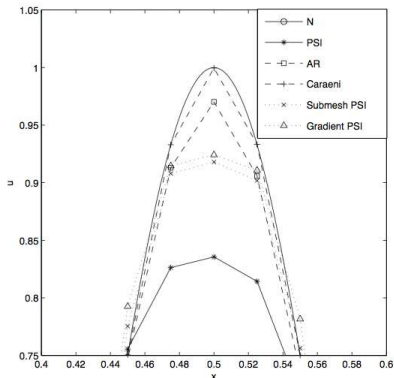
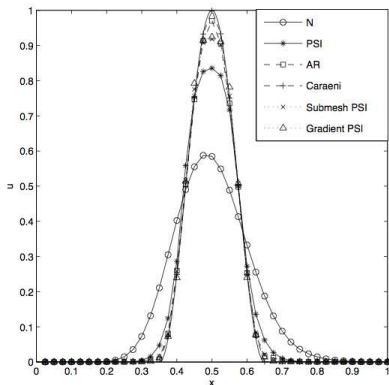
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Solution to spurious modes 1 : modify the stencil

From (Hubbard *J.Comput.Phys* 2007), outlet data

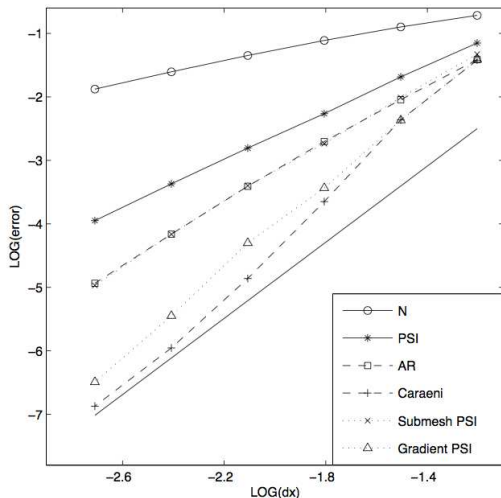


# Higher (than second) order

Solution to spurious modes 1 : modify the stencil

$L^1$  convergence

(Hubbard *J.Comput.Phys* 2007)



## Higher (than second) order

### Solution to spurious modes 2

Distribute using variable weights. The scheme basically becomes a Petrov-Galerkin FE-like method reading

$$\sum_{K \in \Omega_h} \int_K \omega_i^K(x, y, u_h) \nabla \cdot \mathcal{F}_h(u_h)$$

How to define  $\omega_i^K$  ?

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How to define  $\omega_i^K$  ?

One example is Hughes' SUPG scheme :

$$\omega_i^K = \psi_i + \vec{a}(u_h) \cdot \nabla \psi_i \tau$$

and all the limiters stuff ?

## Higher (than second) order

### Solution to spurious modes 2

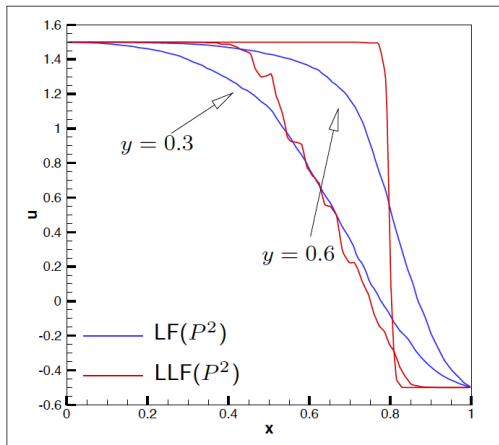
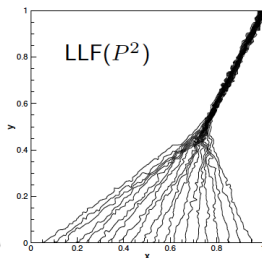
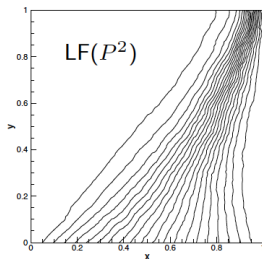
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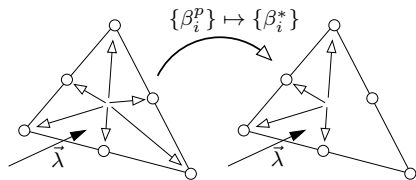
I will discuss one approach developed at Inria  
(Abgrall, Larat, Ricchiuto *J.Comput.Phys.*) 2011

## Higher (than second) order





## Smooth solutions and spurious modes

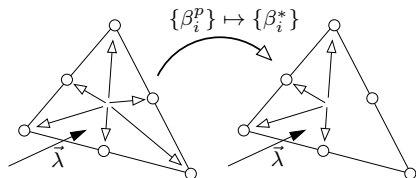


Think a little bit about it and ..

Things are OK in shocks. In smooth areas  $\phi^K = \mathcal{O}(h^{k+2}) \ll 1$   
(Abgrall, *J. Comput. Phys* 2006) :

- ▶ Linearize the nonlinear system  $\sum_{K|i \in K} \phi_i^K = 0$  :  $M_h^* = B_h^*$
- ▶  $M_h^*$  does not have full range : infinite solutions, hence spurious modes

## Smooth solutions and spurious modes



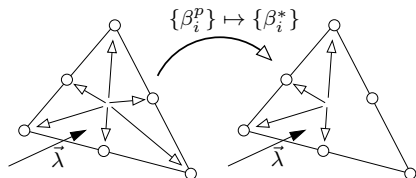
More simply (*Out the door, back through the window...*)

- ▶ The construction of the LLF scheme uses the algebraic constraint (for LED)

$$\phi_j^{\text{LF}} \times \beta_j^{\text{LLF}} \phi^K \geq 0 \rightarrow c_{ij}^{\text{LLF}} = \gamma_i^K c_{ij}^{\text{LF}} \geq 0$$

- ▶ Upwinding not included in the process
- ▶ Locally can have “down-winding” or zero entries in equation (as central scheme and advection)

## Smooth solutions and spurious modes



More simply (*Out the door, back through the window...*)

Upwinding not included in the process ..

we want to “put it back in” in smooth regions .. how ?

# Higher order nonlinear Lax Friedrich's scheme

The best we came up with so far

Add streamline diffusion (Abgrall *J.Comput.Phys* 2006 ; Abgrall, Larat, Ricchiuto *J.Comput.Phys* 2011)

$$\phi_i^{\text{LLFs}} = \beta_i^{\text{LLF}} \phi^K + \delta(u_h) \int_K \vec{a}(u_h) \cdot \nabla \psi_i \tau \vec{a}(u_h) \cdot \nabla u_h$$

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- ▶ As already seen
- ▶ To identify smooth regions :

$$\delta(u_h) = \mathcal{O}(h) \quad \text{in discontinuities}$$

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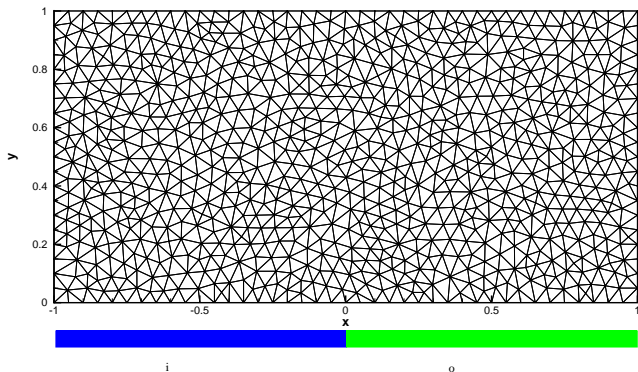
- ▶ As already seen
- ▶ To identify smooth regions :

$$\delta(u_h) = \mathcal{O}(h) \quad \text{in discontinuities}$$

- ▶ Exactly as in the SUPG scheme

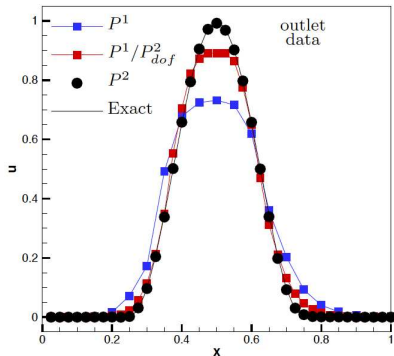
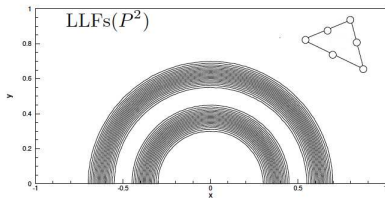
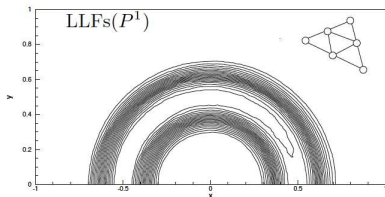
# Higher order nonlinear Lax Friedrich's scheme

$$\vec{a} \cdot \nabla u = 0 \quad \text{on} \quad [-1, 1] \times [0, 1]$$






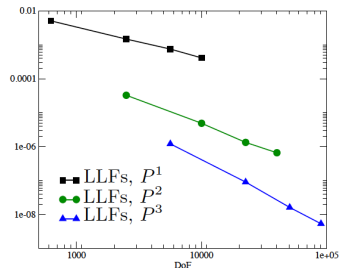


# Higher order nonlinear Lax Friedrich's scheme



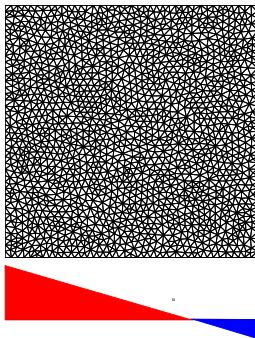
# Higher order nonlinear Lax Friedrich's scheme

			
$h$	$\epsilon_{L^2}(P^1)$	$\epsilon_{L^2}(P^2)$	$\epsilon_{L^2}(P^3)$
1/25	0.50493E-02	0.32612E-04	0.12071E-05
1/50	0.14684E-02	0.48741E-05	0.90642E-07
1/75	0.74684E-03	0.13334E-05	0.16245E-07
1/100	0.41019E-03	0.66019E-06	0.53860E-08
	$\mathcal{O}_{L^2}^{\text{ls}} = 1.790$	$\mathcal{O}_{L^2}^{\text{ls}} = 2.848$	$\mathcal{O}_{L^2}^{\text{ls}} = 3.920$

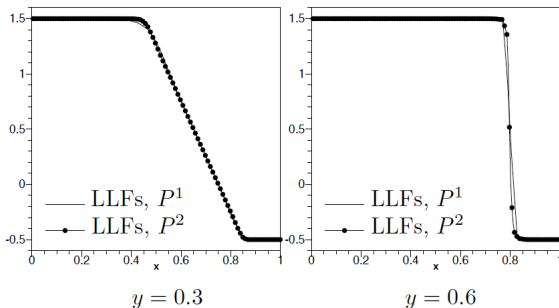
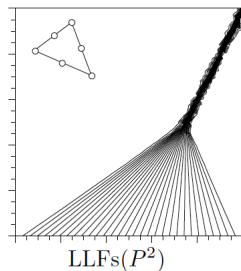
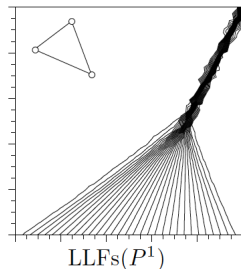


## Higher order nonlinear Lax Friedrich's scheme

$$\nabla \cdot \left( u, \frac{u^2}{2} \right) = 0 \quad \text{on} \quad [0, 1]^2$$



# Higher order nonlinear Lax Friedrich's scheme

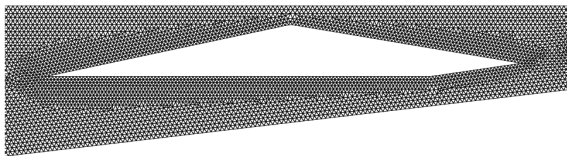


## Extension to systems

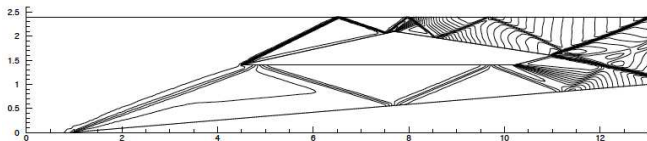
- ▶ All the steps extend formally
- ▶ Limiting step can either be done eq. by eq. or by a characteristic projection (as in FV schemes)

# Example 1 : Mach 3.6 scramjet inlet (Euler, perfect gas)

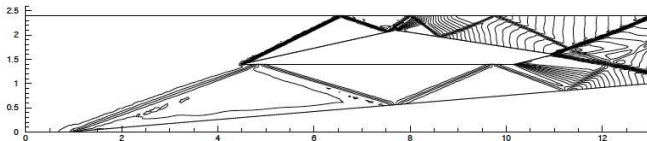
Mesh



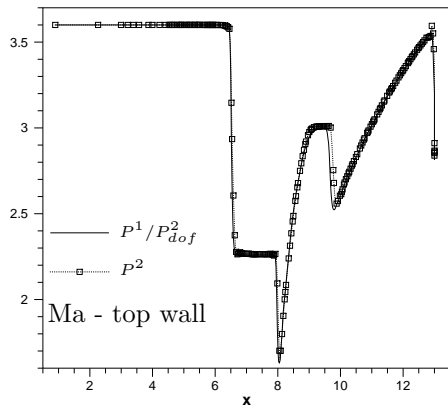
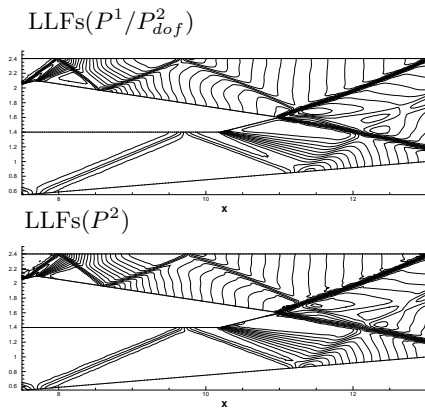
LLFs scheme :  $P^1$  on conformally refined mesh



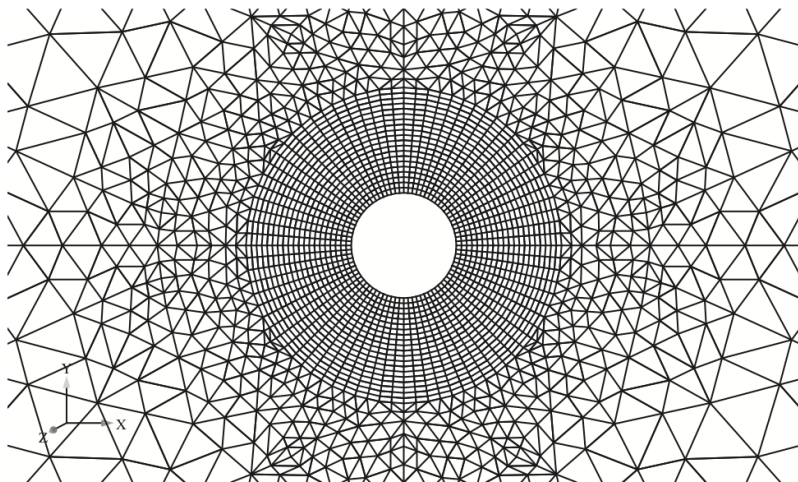
LLFs scheme :  $P^2$



# Scramjet inlet

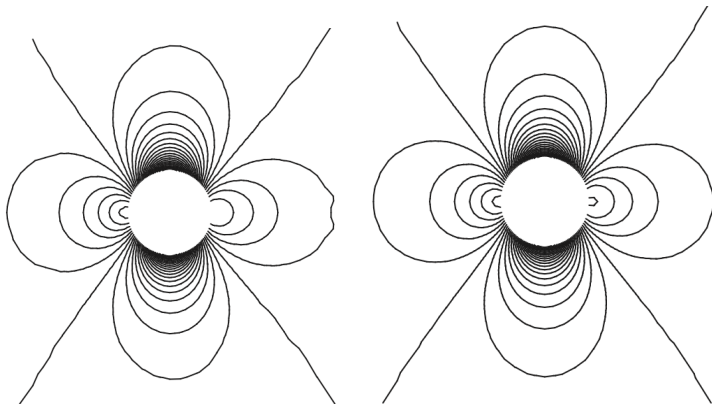


## Euler equations : subsonic cylinder



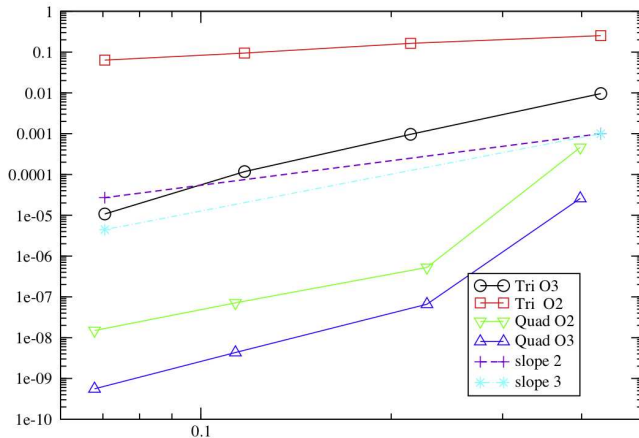


## Euler equations : subsonic cylinder



Conformally refined  $P^1 - Q^1$  (left) *vs*  $P^2 - Q^2$  (right)

# Grid convergence (entropy)



# 4

## RD BASED SCHEMES

for viscous problems

## Navier-Stokes and viscous problems

Consider now the problem

$$\nabla \cdot \mathcal{F}(u) = \nabla \cdot \mathcal{F}^\nu(u, \nabla u)$$

where most often

$$\mathcal{F}^\nu(u, \nabla u) = \mathcal{D}\nabla u \quad \text{with } \mathcal{D} \text{ a SPD matrix}$$

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Early work on the extension of RD to this problem used a Galerkin approximation for the viscous term (Paillere et al *Int.J.Num.Meth.Fluids* 1996). :

$$\sum_{K|i \in K} \beta_i^K \phi^K + \int_{\Omega_h} \mathcal{D} \nabla u_h \cdot \nabla \psi_i = 0$$

This “decoupling” introduces a loss of accuracy in the so-called Pe=1 region (Ricchiuto et al *J.Comput.Appl.Math* 2008).

## Navier-Stokes and viscous problems

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Attempts at improving this while keeping the viscous term in a variational form (Ricchiuto et al *J.Comput.Appl.Math* 2008 ; Villedieu et al *J.Comput.Phys* 2011) successful in the  $P^1$  case and hard to justify in general (despite some interesting numerical results).

# Navier-Stokes and viscous problems

$$\nabla \cdot \mathcal{F}(u) = \nabla \cdot \mathcal{F}^\nu(u, \nabla u), \quad \mathcal{F}^\nu(u, \nabla u) = \mathcal{D}\nabla u$$

A more sound approach : include the viscous flux in the element residual and carry it along together with the hyperbolic terms :

1. Evaluate

$$\phi^K(u_h) = \oint_{\partial K} (\mathcal{F}_h(u_h) - \mathcal{F}_h^\nu(u_h, \nabla u_h)) \cdot \hat{n}$$

2. Solve for  $t \rightarrow \infty$

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K(u_h) = 0$$

# Navier-Stokes and viscous problems

This approach introduces several issues which are only partially dealt with in

1. (Caraeni, Fuchs *Computes & Fluids* 2005)
2. (Chou, Shu *J.Comput.Phys* 2007)
3. (Nishikawa *J.Comput.Phys* 2007 ; Nishikawa *J.Comput.Phys* 2010 ; Nishikawa *Computers & Fluids* 2011)
4. (Abgrall et al *Int.J.Num.Meth.Fluids* 2012 ; Abgrall, De Santis, Ricchiuto *ICCFD7* 2012)

The objective is to give a little insight in these issues pointing out the common points with other numerical schemes



## Navier-Stokes and viscous problems

Let us stick to the steady limit of ( $\nu$  a constant viscosity)

$$\partial_t u + \vec{a} \cdot \nabla u = \nu \Delta u$$

Issue 1 :  $C^0$  continuity

## Navier-Stokes and viscous problems

Let us stick to the steady limit of ( $\nu$  a constant viscosity)

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### Issue 1 : $C^0$ continuity

The underpinning approximation of the solution in RD methods is only  $C^0$  continuous across the faces on which the hyperbolic flux is evaluated. But now (even in the  $P^1$  case) :

$$\phi^K = \oint_{\partial K} u_h \vec{a} \cdot \hat{n} - \oint_{\partial K} \nu \nabla u_h \cdot \hat{n}$$

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$$\phi^K = \oint_{\partial K} u_h \vec{a} \cdot \hat{n} - \oint_{\partial K} \hat{\mathcal{F}}_h^\nu(\nabla u_u|_K, \nabla u_u|_{K'}) \cdot \hat{n}$$

Conditions on  $\hat{\mathcal{F}}_h^\nu$

1. Consistency :  $\hat{\mathcal{F}}_h^\nu = \nu \nabla u$  if built using  $C^1$  continuous data
2. Accuracy :  $\hat{\mathcal{F}}_h^\nu - \nu \nabla u = \mathcal{O}(h^{k+1})$  on  $P^k$  if  $u$  is smooth

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## Solutions proposed

- ▶ Gradient reconstruction in the degrees of freedom
- ▶ First order formulation of the problem

# Navier-Stokes and viscous problems

## Issue 1 : $C^0$ continuity and gradient reconstruction

- ▶ Reconstruct in every degree of freedom  $i \in \Omega_h$  an accurate value of  $\nabla u_i$  using the set of values

$$\{u_j, \forall j \in K \text{ and } \forall K \in \Omega_h | i \in K\}$$

- ▶ Set  $\widehat{\nabla} u_h = \sum_j \psi_j \nabla u_j$
- ▶ Set  $\widehat{\mathcal{F}}_h^\nu = \nu \widehat{\nabla} u_h$

# Navier-Stokes and viscous problems

## Reconstruction procedures

- ▶ Green-Gauss reconstruction
- ▶ Least squares reconstruction
- ▶  $L^2$  projection

## Gradient reconstruction : pros & cons

- ▶ Well known procedures in the FV framework +
- ▶ Simple enough and can be coded efficiently +
- ▶ Non local -
- ▶ Accuracy condition :  $\nabla u_j - \nabla u^{\text{ex}}(x_j) = \mathcal{O}(h^{k+1})$  -

# Navier-Stokes and viscous problems

## Gradient Reconstruction procedures

- ▶ Simple and efficient approaches limited to second and third order accuracy
- ▶ More complex WENO reconstructions up to fourth order in (Chou, Shu *J.Comput.Phys* 2007) but limited to structured meshes



# Navier-Stokes and viscous problems

## Issue 1 : $C^0$ continuity and FOS

Recast the problem as the limit of the First Order System

$$\begin{aligned}\partial_t u + \vec{a} \cdot \nabla u - \nu \partial_x p - \nu \partial_y q &= 0 \\ T_R \partial_t p + p - \partial_x u &= 0 \\ T_R \partial_t q + q - \partial_y u &= 0\end{aligned}$$

Write a scheme for the coupled system until steady state.

# Navier-Stokes and viscous problems

## FOS pros & cons

- ▶ Works very well +
- ▶ Relaxation time  $T_R$  can be optimized to achieve fast convergence +
- ▶ Very memory demanding -
- ▶ Extension to Navier-Stokes difficult (see next) -
  
- ▶ A different way to look at the mixed problem
- ▶ Due to the coupling, equivalent at steady state to an (expensive) “implicit reconstruction” technique

$$\widehat{\nabla} u_h = F(u_h, \nabla u_h, \widehat{\nabla} u_h)$$

## Navier-Stokes and viscous problems

Let us stick to the steady limit of ( $\nu$  a constant viscosity)

$$\partial_t u + \vec{a} \cdot \nabla u = \nu \Delta u$$

**Issue 2 : what is the correct “distribution direction” ?**

How do we distribute the element residual

$$\phi^K = \oint_{\partial K} u_h \vec{a} \cdot \hat{n} - \oint_{\partial K} \hat{\mathcal{F}}_h^\nu \cdot \hat{n}$$

In 1d reduces to the a problem very similar to finding a viscous numerical flux with the right stability properties faced in the DG community

# Navier-Stokes and viscous problems

Issue 2 : what is the correct “distribution direction” ?

How do we distribute the element residual

$$\phi^K = \oint_{\partial K} u_h \vec{a} \cdot \hat{n} - \oint_{\partial K} \hat{\mathcal{F}}_h^\nu \cdot \hat{n}$$

## Three approaches

1. Blend the  $\nu = 0$  scheme with a central scheme. Blending parameter written as a function of the Reynolds number (Peclet)

$$\text{Re}_K = \frac{\|\vec{a}\|_K h_K}{\nu}$$

Stability problems (weak iterative convergence)

# Navier-Stokes and viscous problems

Issue 2 : what is the correct “distribution direction” ?

$$\begin{aligned}\partial_t u + \vec{a} \cdot \nabla u - \nu \partial_x p - \nu \partial_y q &= 0 \\ T_R \partial_t p + p - \partial_x u &= 0 \\ T_R \partial_t q + q - \partial_y u &= 0\end{aligned}$$

## Three approaches

### 2. Discretize the coupled First Order System

- ▶ The system is hyperbolic : use standard RD techniques
- ▶ Optimization of  $T_R$  : fast convergence with explicit time stepping.  $\Delta t = \mathcal{O}(h)$  offsets the cost of extra equations
- ▶ Convergence rates of  $\mathcal{O}(h^{k+1})$  in the  $H^1$  norm attainable
- ▶ **Memory demanding**
- ▶ **Navier-Stokes remains a challenge**

# Navier-Stokes and viscous problems

Issue 2 : what is the correct “distribution direction” ?

How do we distribute the element residual

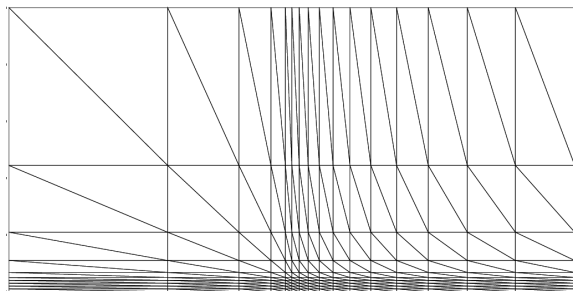
$$\phi^K = \oint_{\partial K} u_h \vec{a} \cdot \hat{n} - \oint_{\partial K} \hat{\mathcal{F}}_h^\nu \cdot \hat{n}$$

## Three approaches

3. Use the FOS to derive an equation for  $u$ . In the coupling terms containing the FOS gradients, replace these by reconstructed gradients.
  - ▶ Allows to define viscous numerical fluxes for DG, FV , etc. (Nishikawa *Computers & Fluids* 2011)
  - ▶ Works very well with the central (or nonlinear) + streamline dissipation developed at Inria
  - ▶ With simple schemes allows Navier-Stokes
  - ▶ Limited by the accuracy of the gradient reconstruction

# Laminar flat plate

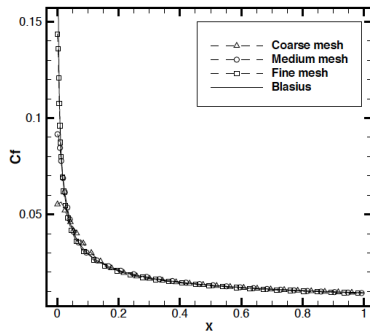
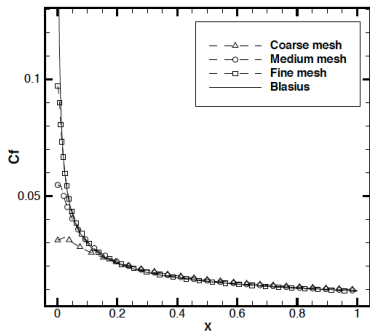
From (Abgrall, De Santis, Ricchiuto *ICCFD7* 2012)



$$\text{Ma}_\infty = 0.3, \quad \text{Re}_\infty = 5000$$

Coarse grid

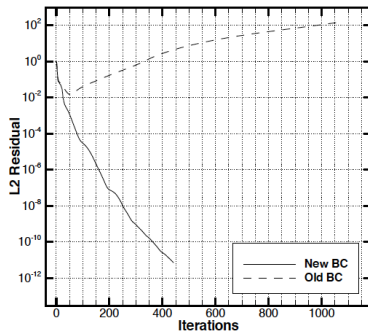
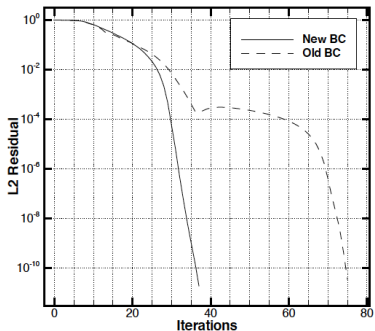
## Laminar flat plate



Skin friction coefficient *vs* Blasius' laminar boundary layer theory  
Left :  $P^1$ , Right :  $P^2$



# Laminar flat plate

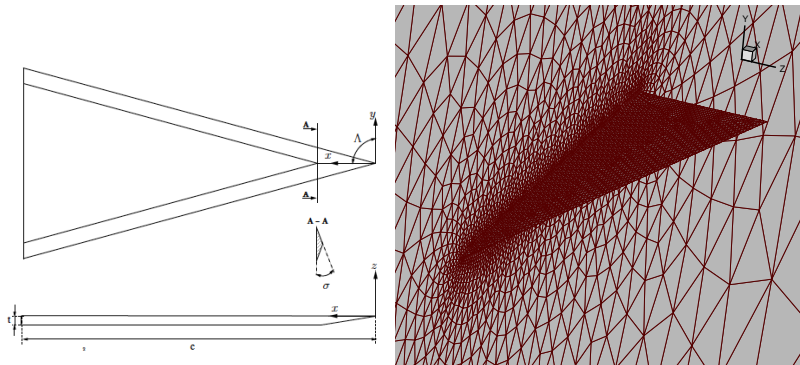


Typical iterative convergence (matrix free GMRES, LU-SGS prec.)

Left :  $P^1$ , Right :  $P^2$

# Laminar delta wing computations

From (Abgrall, De Santis, Ricchiuto *ICCFD7* 2012)



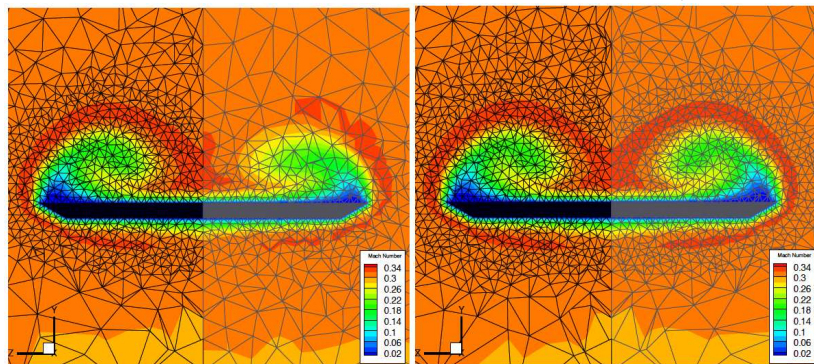
$$\text{Ma}_\infty = 0.3, \quad \text{Re}_\infty = 2000, \quad \text{AoA} = 12.5^\circ$$

# Laminar delta wing computations

2 cycles of refinement based on vorticity magnitude

(MMG3D generator by C.Dobrzynski available under GNU GPL license at

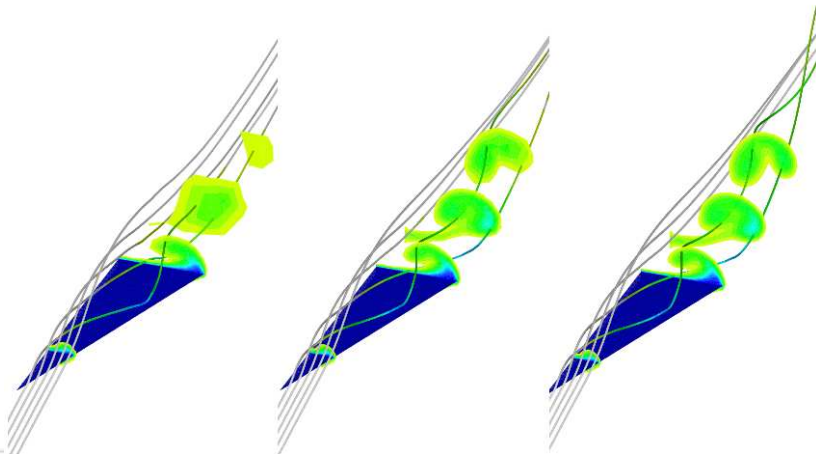
<http://www.math.u-bordeaux1.fr/~cdobrzyn/logiciels/mmg3d.php>)



Fine grid 600k tets

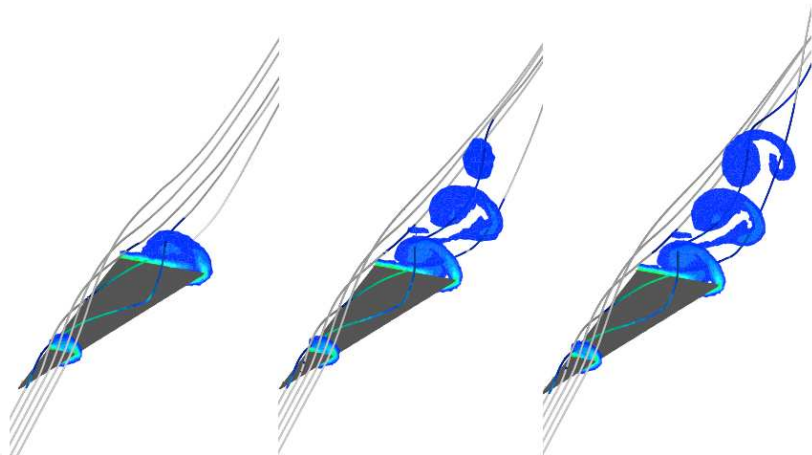
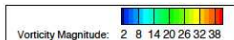
# Laminar delta wing computations

Flow separation and Mach number ( $P^1$  scheme)



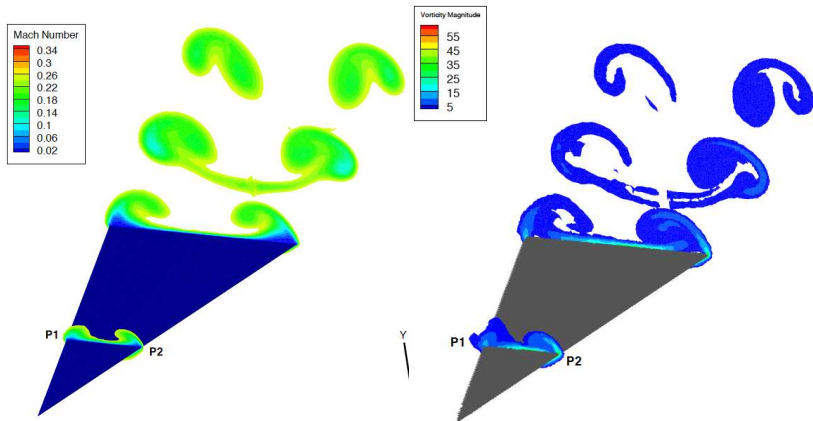
# Laminar delta wing computations

Flow separation and vorticity magnitude ( $P^1$  scheme)



# Laminar delta wing computations

Flow separation, Mach number and vorticity ( $P^1$  vs  $P^2$ )



## Laminar delta wing computations. Drag coefficient

Reference value from (Leicht, Hartmann J.Comput.Phys. 2010) using a second order DG scheme with adjoint based error estimation and grid adaptation (finest meshes 2.5M elements) :

$$C_D = 0.1658$$

Initial grid $P^1$	0.145
Adapted grid 1, $P^1$	0.146
Adapted grid 2, $P^1$	0.147
Adapted grid 2, $P^2$	0.162

# 2

## TIME DEPENDENT

... problems ...



## Time dependent problems

We now consider the time dependent advection equation

$$\partial_t u + \vec{a} \cdot \nabla u = 0$$

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$$\partial_t u + \vec{a} \cdot \nabla u = 0$$

### The accuracy problem

The prototype

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \phi_i^K = 0$$

with

$$\sum_{j \in K} \phi_j^K = \int_K \vec{a} \cdot \nabla u_h \quad \forall K \in \Omega_h$$

is in general first order accurate in space.

## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

$P^1$  triangles to fix ideas (Deconinck, Ricchiuto *Enc. Comput. Mech.* 2007)

## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

$P^1$  triangles to fix ideas (Deconinck, Ricchiuto *Enc. Comput. Mech.* 2007).

- (i) Let  $w$  be a smooth exact solution :  $\partial_t w + \vec{a} \cdot \nabla w = 0$
- (ii) Set  $w_i(t) = w(t, x_i, y_i)$
- (iii) Let  $\phi^K(w_h)$  the quantity obtained when formally replacing the nodal values of the numerical solution by the  $w_i$ s
- (iii)  $\psi \in C_0^1$  compactly supported smooth function,  $\psi_i = \psi(x_i, y_i)$
- (iv) define the integral truncation error

$$\epsilon(w, \psi) := \left| \sum_{i \in \Omega_h} \sum_{K|i \in K} \psi_i \left( |C_i| \frac{dw_i}{dt} + \beta_i^K \phi^K(w_h) \right) \right|$$

## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

Proceeding as in the steady case :

$$\epsilon(w, \psi) = \left| \sum_{K \in \Omega_h} \sum_{j \in K} \psi_j \left( |C_j| \frac{dw_j}{dt} + \beta_j^K \phi^K(w_h) \right) \right|$$

## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

Proceeding as in the steady case :

$$\begin{aligned} \epsilon(w, \psi) \leq & \left| \int_{\Omega} \psi_h (\partial_t(w_h - w) + \vec{a} \cdot \nabla(w_h - w)) \right| \\ & + \sum_{K \in \Omega_h} \sum_{i,j \in K} |\psi_j - \psi_i| \left( \left| |C_j| \frac{dw_j}{dt} + \beta_j^K \phi^K(w_h) \right| \right. \\ & \left. + \left| \int_K \varphi_i (\partial_t(w_h - w) + \vec{a} \cdot \nabla(w_h - w)) \right| \right) \end{aligned}$$

## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

Estimating terms (approximation theory on  $P^1$  triangles)

$$\epsilon(w, \psi) \leq C_1 h^2 + C_2 h^{-1} \sup_{\substack{K \in \Omega_h \\ j \in K}} \left| |C_j| \frac{dw_j}{dt} + \beta_j^K \phi^K(w_h) \right|$$

Second order local truncation error condition :

$$\sup_{\substack{K \in \Omega_h \\ j \in K}} \left| |C_j| \frac{dw_j}{dt} + \beta_j^K \phi^K(w_h) \right| = \sup_{\substack{K \in \Omega_h \\ j \in K}} \epsilon_j^K \leq Ch^3$$

## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

Pushing it a bit more :

$$\begin{aligned} \epsilon_j^K &= \left| C_j \frac{dw_j}{dt} + \beta_j^K \int_K \vec{a} \cdot \nabla w_h \right| \\ &= \left| C_j \frac{dw_j}{dt} - \beta_j^K \int_K \partial_t w_h + \beta_j^K \int_K (\partial_t(w_h - w) + \vec{a} \cdot \nabla(w_h - w)) \right| \\ &\leq \left| C_j \frac{dw_j}{dt} - \beta_j^K \int_K \partial_t w_h \right| + Ch^3 \end{aligned}$$



## Time dependent problems

$$|C_i| \frac{du_i}{dt} + \sum_{K|i \in K} \beta_i^K \phi^K = 0, \quad \phi^K = \int_K \vec{a} \cdot \nabla u_h$$

### Time continuous error analysis

Second order accuracy constraint :

$$\left| |C_j| \frac{dw_j}{dt} - \beta_j^K \int_K \partial_t w_h \right| \leq Ch^3$$

Only true for : centered scheme (mass lumping stuff)

# Time dependent problems

## High order schemes : time dependent case

There is a number of different ways to do it right discussed in

- ▶ (Hubbard, Roe *Int.J.Num.Meth.Fluids* 2000)
- ▶ (Csik, Ricchiuto, Deconinck *AIAA CP* 2001)
- ▶ (Abgrall, Mezine *J.Comput.Phys.* 2003)
- ▶ (Abgrall, Andrianov, Mezine *J.Comput.Phys.* 2003)
- ▶ (Caraeni, Fuchs *Computers & Fluids* 2005)
- ▶ (Ricchiuto, Csik, Deconinck *J.Comput.Phys.* 2005)
- ▶ (Dobes, Deconinck *J.Comput.Appl.Math.* 2008)
- ▶ (Ricchiuto, Bollermann *J.Comput.Phys.* 2009)
- ▶ (Ricchiuto, Abgrall *J.Comput.Phys.* 2010)
- ▶ (Hubbard, Ricchiuto *Computers & Fluids* 2011)

.. and references therein ...

# Time dependent problems

The main idea is to recover second (to fix ideas) order accuracy in space by modifying the (semi-)discrete equations as follows :

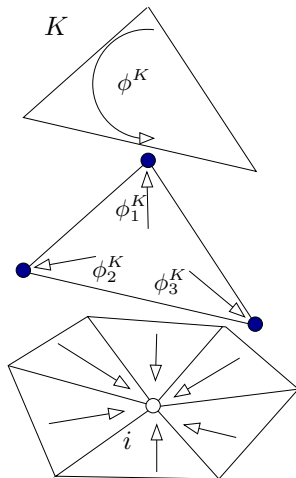
1.  $\forall K \in \Omega_h$  compute :

$$\phi^K = \int_K \left( \partial_t u_h + \nabla \cdot \mathcal{F}_h(u_h) \right)$$

2. Distribution :  $\phi_i^K = \beta_i^K \phi^K$

3. Integrate ODE system  $\sum_{T|i \in T} \beta_i^K \phi^K = 0$  :

$$\sum_{T|i \in T} \beta_i^K \int_K \partial_t u_h = - \sum_{T|i \in T} \beta_i^K \int_K \nabla \cdot \mathcal{F}_h(u_h)$$



# Time dependent problems

## Remarks

- ▶ After time discretization, independently on the explicit or implicit nature of the time stepping scheme, we end with a nonlinear system of equations of the type

$$M(\mathbf{u}^{n+1})\mathbf{u}^{n+1} + \Delta t F(\mathbf{u}^{n+1}) = \Delta t G(\mathbf{u}^n, t\mathbf{u}^{n-1}, \dots)$$

where  $M$  depends on  $\mathbf{u}^{n+1}$  via the  $\beta_i^K$

# Time dependent problems

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- ▶ As in all weighted residual methods we have a mass matrix
- ▶ Even for simple wave propagation problems, almost all of these schemes do not allow simple high order explicit (*e.g.* Runge-Kutta) time stepping
- ▶ A lot of effort has gone into understanding how and if to modify the distribution strategy w.r.t the steady state case

# Time dependent problems : genuinely explicit schemes

Explicit stabilization operators for stabilized FEM

Exception to the rule : (Ricchiuto, Abgrall *J.Comput.Phys* 2010).



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Idea explained for SUPG

# Time dependent problems : genuinely explicit schemes

## Explicit stabilization operators for stabilized FEM

Exception to the rule : (Ricchiuto, Abgrall *J.Comput.Phys* 2010).

Idea explained for SUPG

- ▶ Consider an explicit (single- or multi-step) high order time integration scheme

$$u' + f(u) = 0 \rightarrow u^{n+1} - u^n + \Delta t E(u^n, u^{n-1}, \dots) = 0$$

- ▶ Assume that, given a smooth exact solution  $w$ , the scheme has the local truncation error

$$\epsilon^n = |w^{n+1} - w^n + \Delta t E(w^n, w^{n-1}, \dots)| = C \Delta t^{p+1}$$

- ▶ Now set  $f(u) = \nabla \cdot \mathcal{F}(u)$  in our ODE

## Time dependent problems : genuinely explicit schemes

### Explicit stabilization operators for stabilized FEM

The SUPG finite element scheme reads

$$\int_{\Omega_h} \psi_i (u_h^{n+1} - u_h^n + \Delta t E_h(u_h^n, u_h^{n-1}, \dots)) \\ + \sum_{K|i \in K} \int_K \vec{a}(u_h^n) \cdot \nabla \psi_i \tau (u_h^{n+1} - u_h^n + \Delta t E_h(u_h^n, u_h^{n-1}, \dots)) = 0$$

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To remain consistent, the streamline dissipation term contains the whole unsteady residual leading to a skew algebraic system despite of the explicit time integration

## Time dependent problems : genuinely explicit schemes

### Explicit stabilization operators for stabilized FEM

In (Ricchiuto, Abgrall *J.Comput.Phys* 2010) : without any formal loss of accuracy, the time dependent residual in the stabilization term can be replaced by a weakly consistent shifted one

$$\sum_{l=0}^L \alpha_l u_h^{n-l} + \Delta t E_h(u_h^n, u_h^{n-1}, \dots)$$

that for a smooth solution  $w$  satisfies a lower order consistency estimate

$$\left| \sum_{l=0}^L \alpha_l w^{n-l} + \Delta t E(w^n, w^{n-1}, \dots) \right| = C' \Delta t^p$$

## Time dependent problems : genuinely explicit schemes

### Explicit stabilization operators for stabilized FEM

The SUPG finite element scheme is modified as

$$\int_{\Omega_h} \psi_i (u_h^{n+1} - u_h^n + \Delta t E_h(u_h^n, u_h^{n-1}, \dots)) \\ + \sum_{K|i \in K} \int_K \vec{a}(u_h^n) \cdot \nabla \psi_i \tau \left( \sum_{l=0}^L \alpha_l u_h^{n-l} + \Delta t E_h(u_h^n, u_h^{n-1}, \dots) \right) = 0$$

It remains to invert the SPD Galerkin mass matrix  
or further simplify via mass lumping

## Time dependent problems : genuinely explicit schemes

### Explicit stabilization operators for stabilized FEM

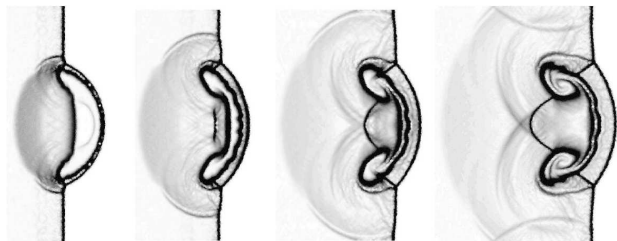
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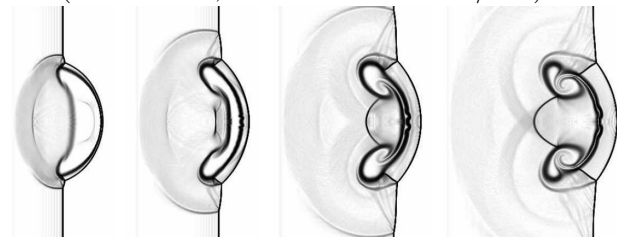
It remains to invert the SPD Galerkin mass matrix  
or further simplify via mass lumping

Same construction allows to obtain genuinely explicit second  
order nonlinear RD  
Higher order in the works

## Time dependent problems : shock “bubble” interaction



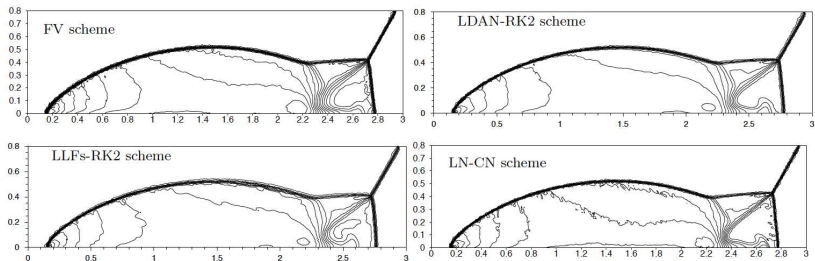
RD (CN in time, unstructured  $h = 1/200$ )



(Holden et al *J.Comput.Phys* 1999, cartesian  $h = 1/400$ )



## Time dependent problems : double Mach reflection



Reflection of a  $Ma=10$  moving shock on a 30 ramp  
Comparison on the same grid with cell centered FV + limiter of  
Barth and Jespersen + RK2

# 5

## SHALLOW WATER SIMULATIONS

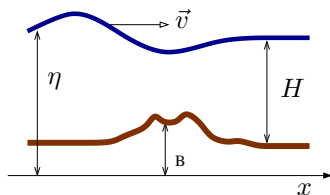
with RD based schemes

## RD for Shallow Water simulations

$$\partial_t H + \nabla \cdot (H\vec{v}) = 0$$

$$\partial_t(H\vec{v}) + \nabla \cdot (H\vec{v} \otimes \vec{v} + g\frac{H^2}{2}\mathbf{I}) + gH(\nabla b + c_f\vec{v}) = 0$$

- ▶ good in deep water
- ▶ not very good in the surf region and before wave break up (need non-hydrostatic corrections)
- ▶ quite good at predicting runup on sloping shores and flooding



## RD for Shallow Water simulations

$$\partial_t H + \nabla \cdot (H\vec{v}) = 0$$

$$\partial_t(H\vec{v}) + \nabla \cdot (H\vec{v} \otimes \vec{v} + g\frac{H^2}{2}\mathbf{I}) + gH(\nabla b + c_f\vec{v}) = 0$$

### Numerical challenges

- ▶ Std stuff of hyperbolic conservation laws (shocks, contacts, expansions, etc) ;
- ▶ Dry are areas ( $H = 0$ ) way more common than zero density in gas dynamics ;
- ▶ Source terms dominated flows ;
- ▶ A large number of simple and non-trivial equilibria flux div-source term ;

## RD for Shallow Water simulations

$$\partial_t H + \nabla \cdot (H\vec{v}) = 0$$

$$\partial_t (H\vec{v}) + \nabla \cdot (H\vec{v} \otimes \vec{v} + g\frac{H^2}{2}\mathbf{I}) + gH(\nabla b + c_f\vec{v}) = 0$$

### Interesting topics

- ▶ Preservation of equilibria with RD (well balancedness or C-property)
- ▶ Construction of well balanced FV fluxes using RD
- ▶ Long wave run up on complex bathymetries

## RD for Shallow Water simulations

### Equilibria with invariants I : homoenergetic frictionless flows

Consider the following set of derived quantities :

$$\begin{aligned}\mathcal{E} &= g\eta + \frac{\|\vec{v}\|^2}{2} && \text{(total energy)} \\ \vec{q} &= H\vec{v} && \text{(discharge)}\end{aligned}$$

Under the compatibility condition  $\vec{v}^\perp \cdot \nabla b = 0$ , the shallow water equations admit the family of steady solutions

$$\begin{aligned}\mathcal{E} &= g\eta + \frac{\|\vec{v}\|^2}{2} = \mathcal{E}_0 \\ \vec{q} &= H\vec{v} = \vec{q}_0\end{aligned}$$

The condition  $\vec{v}^\perp \cdot \nabla b = 0$  only allows pseudo-one dimensional flows, with no cross-wind bathymetry variations.

## RD for Shallow Water simulations

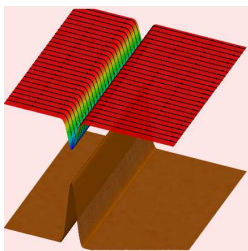
### Equilibria with invariants I & 1/2 : lake at rest

For  $\vec{v} = 0$  we recover the well known lake at rest state

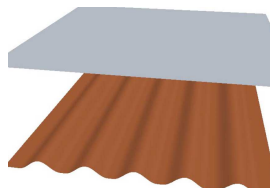
$$\eta = \eta_0$$

$$\vec{v} = 0$$

The velocity being null, the bathymetry can be arbitrary without violating the compatibility condition.



Constant energy



Lake at rest

## RD for Shallow Water simulations

### Equilibria with invariants II : sloping channels with friction and transverse bed variations

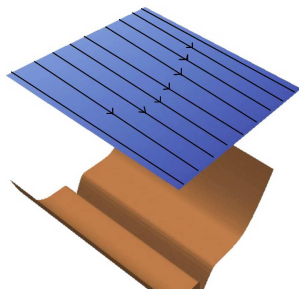
$$b(x, y) = b_0 - \xi_0 x + \beta(y)$$

$$\eta = \eta_0 - \xi_0 x$$

$$H = H_0 - \beta(y)$$

$$c_f(u(y), H(y))u(y) = \xi_0 = -\partial_x b$$

Pseudo 1d flow with transverse bathymetry, depth, velocity variations.





## RD for Shallow Water simulations

### Equilibria with invariants III : C-property

A scheme is said to verify the C-property (Conservation) for a certain steady equilibrium if it is able to preserve it exactly and indefinitely (Bermudez, Vazquez *Computers & Fluids*)

One speaks of approximate C-property, if the equilibrium is preserved within a certain error, smaller than the truncation of the scheme. In this case, here we say that the discretization is super-consistent with the given equilibrium.

Schemes verifying the C-property are often referred to as well-balanced (Greenberg, Leroux *SISC* 1996)

## RD for Shallow Water simulations

$$\partial_t u + \nabla \cdot \mathcal{F}(u) + \mathcal{S}(u, x, y) = 0 \quad \text{on } \Omega \times [0, T_f] \subset \mathbb{R}^2 \times \mathbb{R}^+$$

Super consistency results for RD schemes on  $P^1$  triangles

Consider now the RD schemes obtained as some form of

$$\sum_{K|i \in K} \beta_i^K \int_K (\partial_t u_h + \nabla \cdot \mathcal{F}_h + \mathcal{S}_h) = 0$$

## RD for Shallow Water simulations

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$$\sum_{K|i \in K} \beta_i^K \int_K (\partial_t u_h + \nabla \cdot \mathcal{F}_h + \mathcal{S}_h) = 0$$

Since the whole equation is “carried along” in the distribution it is natural to expect that equilibria between different terms should be resolved accurately

## RD for Shallow Water simulations

### Main result (Ricchiuto, 2011)

On  $P^1$  meshes, high order RD schemes preserve *exactly* steady equilibria with a set of invariants  $v$  provided that

1. exact integration is used
2. the approximation of the flux and of the source term is written as  $\mathcal{F}_h = \mathcal{F}(v_h)$ ,  $\mathcal{S}_h = \mathcal{S}(v_h)$

For approximate quadrature and for a smooth enough bathymetry the super consistency estimate holds

$$|\epsilon_h| \leq C h^l, \quad l = \min(p_f + 1, p_v + 2)$$

with  $p_f$  and  $p_v$  the degrees of the polynomials exactly integrated by the quadrature formulae used.

# RD for Shallow Water simulations

## Meaning and remarks

- ▶ Equilibria described by  $v = \text{const}$  :
  1. homoenergetic flow :  $v = [\mathcal{E}, \vec{q}]$
  2. lake at rest :  $v = [\eta, \vec{q}]$
  3. channel flows with friction (no transverse  $b$ ) :  $v = [H, \vec{v}]$
- ▶  $v$  is interpolated and everything else derived from its values
- ▶ perturbation = quadrature error in computing  $\phi^K(v_h)$
- ▶ Similar to FV (Gallouet, Hérard, Seguin *Computers & Fluids* 2003 ; Noelle, Xing, Shu *J.Comput.Phys.* 2007) but here unstructured triangulations

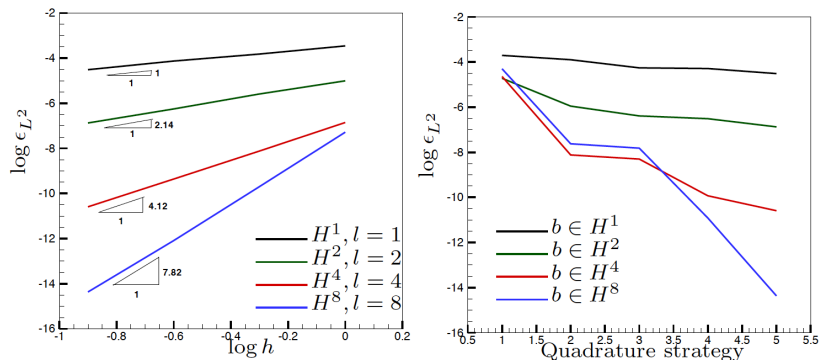
## RD for Shallow Water simulations

Example : homoenergetic flow, unstructured triangular grids

... video ...

# RD for Shallow Water simulations

Example : homoenergetic flow, unstructured triangular grids



The scheme is second order accurate

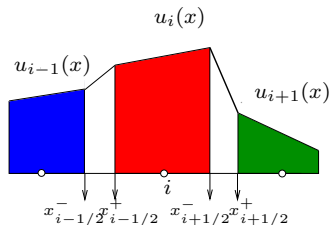
Dependence by data regularity well described by theory

## RD for Shallow Water : construction of FV fluxes

$$\partial_t \begin{bmatrix} h \\ hu \end{bmatrix} + \partial_x \begin{bmatrix} hu \\ hu^2 + g \frac{h^2}{2} \end{bmatrix} + gh \partial_x \begin{bmatrix} 0 \\ b(x) \end{bmatrix} = 0$$

Integrate over cell  $i$  plus RD scheme on ghost cells

$$\begin{aligned} & \Delta x_i \frac{u_i^{n+1} - u_i^n}{\Delta t} + f(u_i(x_{i+1/2}^-)) - f(u_i(x_{i-1/2}^+)) + \mathcal{S}_i + \\ & \beta_i^{i-1/2} \left( f(u_i(x_{i-1/2}^+)) - f(u_{i-1}(x_{i-1/2}^-)) + \mathcal{S}_{i-1/2} \right) + \\ & \beta_i^{i+1/2} \left( f(u_{i+1}(x_{i+1/2}^+)) - f(u_i(x_{i+1/2}^-)) + \mathcal{S}_{i+1/2} \right) = 0 \end{aligned}$$





## RD for Shallow Water : construction of FV fluxes

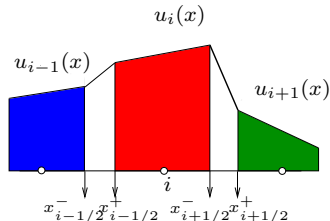
For a linear approximation of  $b(x)$

$$S_i = g\tilde{h}_i(b_i(x_{i+1/2}^-) - b_i(x_{i-1/2}^+))$$

$$S_{i-1/2} = g\tilde{h}_{i-1/2}(b_i(x_{i-1/2}^+) - b_{i-1}(x_{i-1/2}^-))$$

$$S_{i+1/2} = g\tilde{h}_{i+1/2}(b_{i+1}(x_{i+1/2}^+) - b_i(x_{i+1/2}^-))$$

The actual value of the average height  $\tilde{h}$  only depends on the interpolation within the cells.



## RD for Shallow Water : construction of FV fluxes

Using the well balanced RD approach we get in general

$$\tilde{h}_{i-1/2} = \frac{1}{x_{i-1/2}^+ - x_{i-1/2}^-} \int_{x_{i-1/2}^-}^{x_{i-1/2}^+} h \left( \begin{bmatrix} h + b + u^2/2g \\ hu \end{bmatrix} \right) = \sum_{gp} \omega_{gp} h \left( \begin{bmatrix} \mathcal{E}_{gp} \\ q_{gp} \end{bmatrix} \right)$$

in each quadrature point need to solve the nonlinear system

$$\begin{cases} h_{gp} + u_{gp}^2/2g & = \mathcal{E}_{gp} - b_{gp} \\ h_{gp}u_{gp} & = q_{gp} \end{cases}$$

Where the total energy  $\mathcal{E}$  and the flux  $q = hu$  are interpolated linearly.

## RD for Shallow Water : construction of FV fluxes

The final well balanced FV discretization reads :

$$\Delta x_i \frac{u_i^{n+1} - u_i^n}{\Delta t} + S_i + f_{i+1/2} - f_{i-1/2} + S_{i+1/2} + S_{i-1/2} = 0$$

$$S_i = \begin{bmatrix} 0 \\ g\tilde{h}_i(b_L^{i+1/2} - b_R^{i-1/2}) \end{bmatrix}$$

$$f_{i+1/2}(u_L, u_R) = f(u_L) + \beta_i^{i+1/2} (f(u_R) - f(u_L))$$

$$S_{i+1/2} = \beta_i^{i+1/2} \begin{bmatrix} 0 \\ g\tilde{h}_{i+1/2}(b_R - b_L) \end{bmatrix}$$

Generalized form of  
well balanced quadrature of (Noelle, Xing, Shu, *JCP* 226, 2007)

Extra degree of freedom in the actual splitting  
*viz* the definition of the  $\beta_i$  coefficients

## Run up on complex bathymetries

Studied in (Ricchiuto, Bollermann *JCP* 2009 ; Ricchiuto AIP Proc. 1389 2011)

1. Adapted nonlinear variants of the Lax-Friedrich's distribution guaranteeing in some form

$$H_i^{n+1} \geq 0 \text{ whenever } H_h^n \geq 0$$

2. Several time-stepping strategies allow the preservation of this constraint :
  - ▶ Implicit Crank-Nicholson. Positivity preserved under a CFL=2 constraint ;
  - ▶ Space-time schemes (discontinuous in time). Unconditional positivity ;
  - ▶ Genuinely explicit RK-RD schemes. Positivity preserved under a CFL=1 constraint.
3. A strategy to maintain the lake at rest near dry regions

see examples

# 6

## SUMMARY AND PERSPECTIVES

## Summary and perspectives

Acknowledge contributions of my collaborators/friends

- ▶ R. Abgrall, G. Baurin, P. Congedo, D. De Santis, C. Dobrzynski, D. Genet, P. Jacq +++ (Inria)
- ▶ H. Deconinck, S. D'angelo, N. Villedieu, M. Vymazal ++ (VKI)
- ▶ M. Hubbard, D. Sarmany, A. Warzynski ++ (Leeds University)
- ▶ A. Larat (Ecole Central Paris)
- ▶ E. Vazquez-Cendon (University of Santiago de Compostela)
- ▶ and all those I have inevitably forgotten

## Summary and perspectives

- ▶ RD as a general framework to study non-oscillatory higher order schemes
- ▶ On one hand “true RD schemes”
- ▶ On the other a means of improving other techniques via several “bridges” allowing to recast one as the other

## Current work related to RD

- ▶ Turbulence modeling (PhD D. De Santis)
- ▶ GPU implementation (PhD D. Genet)
- ▶ Adjoint error estimation for RD (PhD S. D'angelo)
- ▶ Higher order time dependent (with R. Abgrall, A. Larat and M. Hubbard - PhD A. Warzynski, Leeds)
- ▶ Other polynomial approximations (bridge with DG, Bezier)
- ▶ Non-hydrostatic free surface modeling with UQ (with R. Abgrall, P. Congedo, A.I. Delis, F. Marche)
- ▶ Local adaptation for unsteady problems (with R. Abgrall, C. Dobrzynski)
- ▶ Mass consistent coupling with transport equations (with E. Vazquez)
- ▶ etc. etc.



THANK YOU



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BACCHUS Inria team :

<http://bacchus.bordeaux.inria.fr>