

A multiscale method applied to shallow water flow

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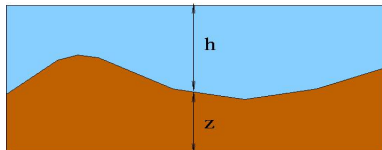
Benasque
September, 2007

Outline

- 1 Shallow water equations
 - The numerical scheme
 - The semi-discrete formulation
- 2 The multilevel algorithm
 - General framework
 - Main steps of the Algorithm
- 3 Numerical Experiments
 - Evaluation of the algorithm: Quality and Efficiency
 - The C-property
 - 2-D Test

Shallow water equations

- hyperbolic system of conservation laws
- source terms are due to topography
- not considering wind effects nor Coriolis force.



$$U_t + F(U)_x + E(U)_y = S$$

$$\begin{pmatrix} h \\ q_1 \\ q_2 \end{pmatrix}_t + \begin{pmatrix} q_1 \\ \frac{q_1^2}{h} + \frac{1}{2}gh^2 \\ \frac{q_1 q_2}{h} \end{pmatrix}_x + \begin{pmatrix} q_2 \\ \frac{q_1 q_2}{h} \\ \frac{q_2^2}{h} + \frac{1}{2}gh^2 \end{pmatrix}_y = \begin{pmatrix} 0 \\ -ghz_x \\ -ghz_y \end{pmatrix}$$

Numerical Treatment

- Fractional step methods do not respect steady/quasy-steady states.

[Leveque]

- Source term upwinding.

[Roe, Bermúdez-Vázquez]

- Follow approach of

[Caselles-Donat-Haro, Donat-Marquina, Gascón-Corberán]

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The numerical scheme

$$U_t + (F + B)_x + (E + C)_y = 0$$

FORMAL FLUXES = physical fluxes + $\left\{ \begin{array}{l} B(x, y, t) = \left(0, \int_{\bar{x}}^x ghz_x ds, 0 \right)^T \\ C(x, y, t) = \left(0, 0, \int_{\bar{y}}^y ghz_y ds \right)^T \end{array} \right.$

NUMERICAL METHOD (Shu-Osher, Finite Difference framework)

- TVD-Runge-Kutta method (time integration).
- dimension by dimension discretization.
- ENO reconstruction of formal fluxes:
 - characteristic speeds of physical fluxes.
 - trapezoidal rule (integral approximation).

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The semi-discrete formulation 1D

[CaDoHa] V. Caselles, R. Donat, G. Haro, *Flux Gradient and Source Term Balancing for certain high resolution shock capturing schemes*. Submitted.

$$\frac{dU}{dt} + \text{Div}(U) = 0$$

$$\frac{G_{i+\frac{1}{2}}^+ - G_{i-\frac{1}{2}}^-}{\Delta x}$$

$G_{i+\frac{1}{2}}^\pm$ involve quantities: $B_{i,i+1} = \int_{x_i}^{x_{i+1}} ghz_x dx$.

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C-PROPERTY.

quiescent flow: $h = \text{constant} - z \quad q_1 = q_2 = 0$

- **exact** C-property \implies numerical scheme exact.
- **approximate** C-property \implies numerical scheme $\mathcal{O}(\Delta x^2)$.

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$$G_{i+\frac{1}{2}}^{\pm} = \sum_{p=1}^2 (\tilde{G}_{i+\frac{1}{2}}^{p,\pm})^L R^p(U^L) + (\tilde{G}_{i+\frac{1}{2}}^{p,\pm})^R R^p(U^R)$$

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- If $\lambda^p(U_{i+\frac{1}{2}}^L) > 0$ and $\lambda^p(U_{i+\frac{1}{2}}^R) > 0$: **upwind from the left** $(\tilde{G}_{i+\frac{1}{2}}^{p,\pm})^R = 0$

$$(\tilde{G}_{i+\frac{1}{2}}^{p,+})^L = L^p(U^L) \cdot F_i + \text{HOT}_{i+\frac{1}{2}}^L$$

$$(\tilde{G}_{i+\frac{1}{2}}^{p,-})^L = L^p(U^L) \cdot (F_i - B_{i,i+1}) + \text{HOT}_{i+\frac{1}{2}}^L$$

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- If $\lambda^p(U_{i+\frac{1}{2}}^L) < 0$ and $\lambda^p(U_{i+\frac{1}{2}}^R) < 0$: **upwind from the right** $(\tilde{G}_{i+\frac{1}{2}}^{p,\pm})^L = 0$

$$(\tilde{G}_{i+\frac{1}{2}}^{p,+})^R = L^p(U^R) \cdot (F_{i+1} + B_{i,i+1}) + HOT_{i+\frac{1}{2}}^R$$

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- If $\lambda^p(U_{i+\frac{1}{2}}^L) * \lambda^p(U_{i+\frac{1}{2}}^R) < 0$: **sonic point nearby** $\alpha = \max(|\lambda^p(U_{i+\frac{1}{2}}^L)|, |\lambda^p(U_{i+\frac{1}{2}}^R)|)$

$$(\tilde{G}_{i+\frac{1}{2}}^{p,+})^L = \frac{1}{2} L^p(U^L) \cdot (F_i + \alpha U_i) + HOT_{i+\frac{1}{2}}^L$$

$$(\tilde{G}_{i+\frac{1}{2}}^{p,-})^L = \frac{1}{2} L^p(U^L) \cdot (F_i + \alpha U_i - B_{i,i+1}) + HOT_{i+\frac{1}{2}}^L$$

$$(\tilde{G}_{i+\frac{1}{2}}^{p,+})^R = \frac{1}{2} L^p(U^R) \cdot (F_{i+1} - \alpha U_{i+1} + B_{i,i+1}) + HOT_{i+\frac{1}{2}}^R$$

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- 1J scheme $\implies U^* = (U^L + U^R)/2 \implies$ exact C-property.
- 2J scheme $\implies U^L \neq U^R \implies$ approximate C-property ($r \geq 2$).
- 1J-2J scheme \implies get the benefits of both (our choice).

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General framework $2D$

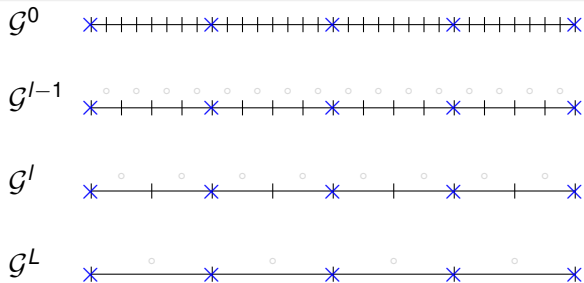
Goal

Reduce the CPU time

Means

Analyze smoothness using Harten's interpolatory. Multiresolution transform. [Chiavassa-Donat, SISC01]

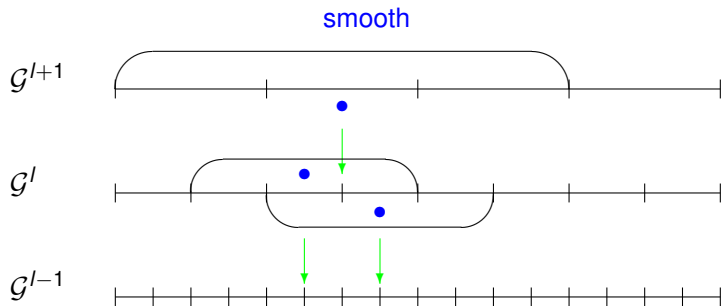
Algorithm



Interpolatory multiresolution

- $\{\mathcal{G}^l, l = 0, \dots, L\}: (x_i, y_j) \in \mathcal{G}^l \iff (x_{2^l i}, y_{2^l j}) \in \mathcal{G}^0$.
- $(d'_{ij})_{l,i,j}$ (wavelet coefficients) USED TO DETERMINE $(b'_{ij})_{l,i,j}$ (marker).
 - $(u'_{i,j})_{l,i,j}$ on $\mathcal{G}^0 \implies u'_{i,j} = u'_{2^l i, 2^l j}$ on \mathcal{G}^l
 - $d'_{i,j} = u'_{i,j} - I[(x_i, y_j); u']$ $(x_i, y_j) \in \mathcal{G}^{l-1} \setminus \mathcal{G}^l$

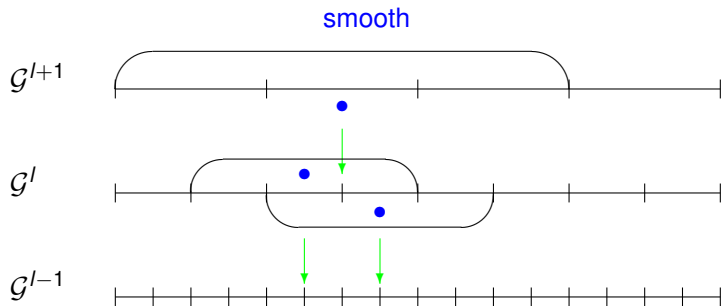
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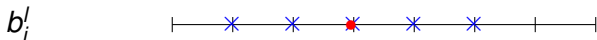
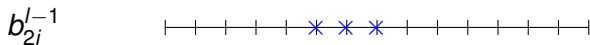
Algorithm

A thresholding algorithm

$l = L, \dots, 1$

$$|d_{ij}^l| \geq \varepsilon \implies b_{i-k, j-m}^l = 1 \quad k, m = -2, \dots, 2$$

$$|d_{ij}^l| \geq 2^r \varepsilon \quad \text{and} \quad l > 1 \implies b_{2i-k, 2j-m}^{l-1} = 1 \quad k, m = -1, 0, 1$$



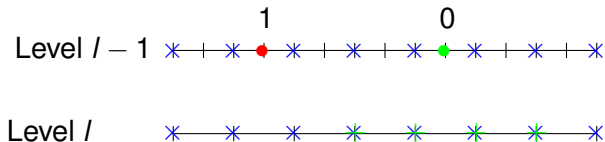
Algorithm

Numerical divergence evaluate on the coarsest grid \mathcal{G}^L

Multilevel evaluation

$l = L, \dots, 1$

$b_{ij}^l = 1$, compute $Div^{l-1}(U)_{ij}$ directly with the scheme
 $b_{ij}^l = 0$, compute $Div^{l-1}(U)_{ij} = \mathcal{I}[(x_i, y_j); Div^l(U)]$



Evaluation of the algorithm: Quality and Efficiency

Quality

$$\frac{\|h_{mr}^n - h_{ref}^n\|_{\ell_1}}{\|h_{ref}^n\|_{\ell_1}}$$

Efficiency

- %f percentage of numerical divergence computed.
- θ cpu gain

$$\theta = \frac{\text{CPU time for reference computation}}{\text{CPU time for multilevel computation}}$$

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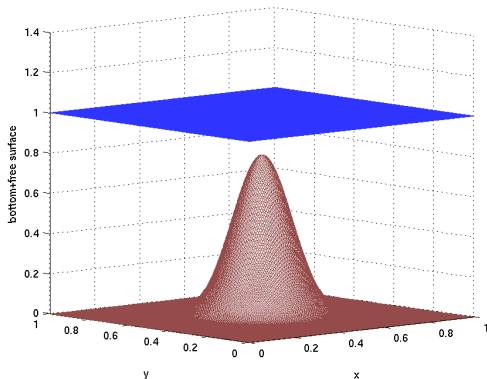
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The C-property

Grid size \mathcal{G}^0	$\%f_{min}$	-	$\%f_{max}$	l_1 -error
257×257	6,5784	-	6,5784	$5,4674 \cdot 10^{-15}$
513×513	1,6510	-	1,6510	$1,1376 \cdot 10^{-14}$



2-D Test

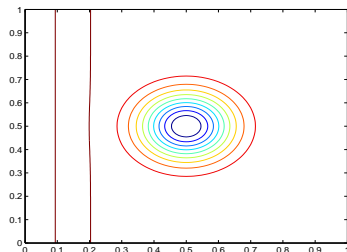
Initial Data

$$z(x, y) = 0,5e^{-50((x-0,5)^2+(y-0,5)^2)}$$

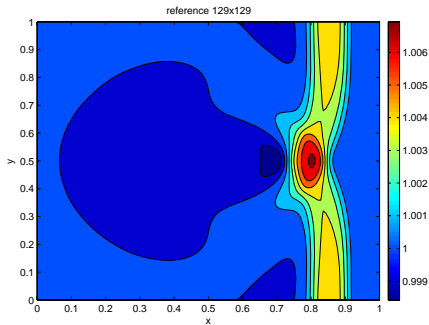
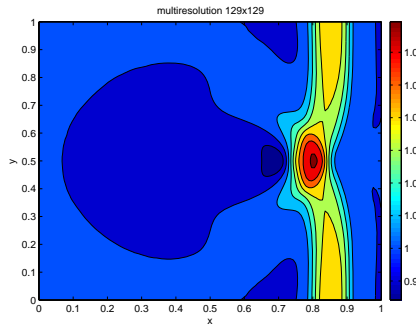
$$q_1(x, y) = 0$$

$$h(x, y) = \begin{cases} 1,01 - z(x, y), & 0,1 < x < 0,2; \\ 1 - z(x, y), & \textit{otherwise.} \end{cases}$$

$$q_2(x, y) = 0$$



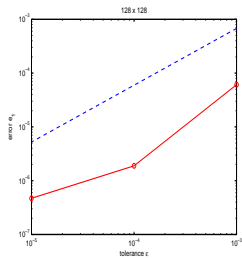
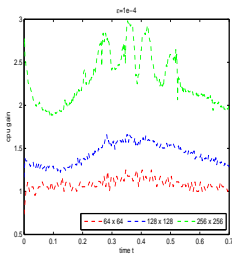
2-D Test

REFERENCE SIMULATION ($t = 0,7$)MULTILEVEL SIMULATION ($\varepsilon = 10^{-4}$)

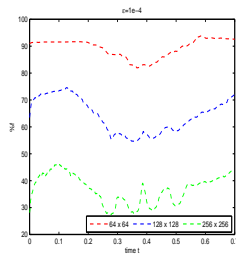
Evaluation of the algorithm: Efficiency

Grid size \mathcal{G}^0	$\%f_{min}$	-	$\%f_{max}$	cpu gain θ
64×64	81.92	-	93.80	1.0778
128×128	54.55	-	74.64	1.4220
256×256	27.49	-	46.52	2.2592

Tolerance versus error

Time evolution of θ 

Time evolution of %



Future work

- Simulation of water avalanches or dam-breaks over dry beds with variable topography.
- Simulations on real topographies.
- Consider source terms due to wind effects and/or Coriolis force.