Well-Balanced and Positivity Preserving DG Schemes for Shallow Water Flows with Shock Capturing by Adaptive Filtering Procedures

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The 2D Shallow Water Equations



- H water height
- b bottom elevation

$$\frac{\partial \mathbf{u}}{\partial t} + \nabla \cdot \mathcal{F}(\mathbf{u}(\mathbf{x},t)) = \mathbf{s}(\mathbf{u},\mathbf{x},t)$$

Where
$$\mathbf{u} = (H, Hv_1, Hv_2)^T$$
, $\nabla \cdot \mathcal{F} = \frac{\partial}{\partial x} \mathbf{f}_1 + \frac{\partial}{\partial y} \mathbf{f}_2$,
 $\mathbf{f}_1 = \begin{pmatrix} Hv_1 \\ Hv_1^2 + \frac{1}{2}gH^2 \\ Hv_1v_2 \end{pmatrix}$, $\mathbf{f}_2 = \begin{pmatrix} Hv_2 \\ Hv_1v_2 \\ Hv_2^2 + \frac{1}{2}gH^2 \end{pmatrix}$,
 $\mathbf{s} = (0, -gHb_x, -gHb_y)^T$

SWE – Challenges for a numerical scheme

Shock capturing, Well-balancedness, Positivity preservation

Xing, Zhang, Shu '10:

3rd order DG scheme on cartesian grids, WB + PP, Limiters Bryson, Epshteyn, Kurganov, Petrova '10:

2nd order central-upwind scheme on triangular grids, WB + PP, Limiters

In this talk:

- 3rd order WB and PP DG scheme on triangular grids
- Shock capturing by (low cost) filtering procedures
- Positivity preservation for implicit time integration *

Joint work with:

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* Andreas Meister (Univ. Kassel)
+ Thomas Sonar (TU Braunschweig)
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The DG Scheme for SWE





- Triangulation of computational domain
- Piecewise polynomial approximation $\mathbf{u}_h(\mathbf{x}, t)$ of degree N
- Variational formulation in space
- DG advantages: conservative method, highly local, unstructured grids easy to implement

Variational formulation

$$\frac{d}{dt} \int_{\tau_i} \underbrace{\mathbf{u}}_{h} \Phi \, d\mathbf{x} + \int_{\partial \tau_i} \underbrace{\mathcal{F}(\mathbf{u})}_{h} \cdot \mathbf{n} \Phi \, d\sigma - \int_{\tau_i} \underbrace{\mathcal{F}(\mathbf{u})}_{h} \cdot \nabla \Phi \, d\mathbf{x} = \int_{\tau_i} \underbrace{\mathbf{s}}_{h} \Phi \, d\mathbf{x}$$
$$= \int_{\tau_i} \underbrace{\mathbf{s}}_{h} \Phi \, d\mathbf{x}$$
$$= \int_{\tau_i} \underbrace{\mathbf{s}}_{h} \Phi \, d\mathbf{x}$$

Variational formulation

$$\frac{d}{dt}\int_{\tau_i}\mathbf{u}_h\,\Phi\,d\mathbf{x} + \int_{\partial\tau_i}\mathbf{F}^{num}(\mathbf{u}_h^-,\mathbf{u}_h^+,\mathbf{n})\,\Phi\,d\sigma - \int_{\tau_i}\mathcal{F}(\mathbf{u}_h)\cdot\nabla\Phi\,d\mathbf{x} = \int_{\tau_i}\mathbf{s}_h\,\Phi\,d\mathbf{x}$$

Use orthogonal polynomials

$$\mathbf{u}_h|_{ au_i}(\mathbf{x},t) = \sum_{k=1}^{q(N)} \hat{\mathbf{u}}_k^i(t) \Phi_k(\mathbf{x}), \qquad q(N) = (N+1)(N+2)/2$$

Semidiscrete equation for coefficients

$$\frac{d}{dt} \hat{\mathbf{u}}_{k}^{i} = \left(-\int_{\partial \tau_{i}} \mathbf{F}^{num}(\mathbf{u}_{h}^{-}, \mathbf{u}_{h}^{+}, \mathbf{n}) \Phi_{k} d\sigma + \int_{\tau_{i}} \mathcal{F}(\mathbf{u}_{h}) \cdot \nabla \Phi_{k} d\mathbf{x} \right) / \|\Phi_{k}\|_{L^{2}}^{2}$$

$$+ \int_{\tau_{i}} \mathbf{s}_{h} \Phi_{k} d\mathbf{x} / \|\Phi_{k}\|_{L^{2}}^{2}$$

Quadrature rules

Variational formulation

$$\frac{d}{dt}\int_{\tau_i}\mathbf{u}_h\,\Phi\,d\mathbf{x} + \int_{\partial\tau_i}\mathbf{F}^{num}(\mathbf{u}_h^-,\mathbf{u}_h^+,\mathbf{n})\,\Phi\,d\sigma - \int_{\tau_i}\mathcal{F}(\mathbf{u}_h)\cdot\nabla\Phi\,d\mathbf{x} = \int_{\tau_i}\mathbf{s}_h\,\Phi\,d\mathbf{x}$$

Use orthogonal polynomials

$$\mathbf{u}_h|_{\tau_i}(\mathbf{x},t) = \sum_{k=1}^{q(N)} \hat{\mathbf{u}}_k^i(t) \Phi_k(\mathbf{x}), \qquad q(N) = (N+1)(N+2)/2$$

Semidiscrete equation \rightarrow Time integration: RK method

$$\frac{d}{dt}\hat{\mathbf{u}}_{k}^{i} = \left(-\int_{\partial\tau_{i}}\mathbf{F}^{num}(\mathbf{u}_{h}^{-},\mathbf{u}_{h}^{+},\mathbf{n}) \Phi_{k}d\sigma + \int_{\tau_{i}}\mathcal{F}(\mathbf{u}_{h})\cdot\nabla\Phi_{k}d\mathbf{x}\right)/\|\Phi_{k}\|_{L^{2}}^{2} + \int_{\tau_{i}}\mathbf{s}_{h}\Phi_{k}\,d\mathbf{x}/\|\Phi_{k}\|_{L^{2}}^{2}$$

Quadrature rules

Orthogonal Proriol-Koornwinder-Dubiner Basis

$$\Phi_{lm}(r,s) = P_l^{0,0} \left(2\frac{1+r}{1-s} - 1 \right) \left(\frac{1-s}{2} \right)^l P_m^{2l+1,0}(s), \quad l+m \le N$$

 $P_n^{\alpha,\beta}$ Jacobi polynomials

Defined on reference element $\ensuremath{\mathbb{T}}$



 $\mathbb{T} = \{(r, s) \in \mathbb{R}^2 \mid -1 \le r, s; r + s \le 0\}$

Proriol (1957), Koornwinder (1975), Dubiner (1991)

New Modal Filtering Approach

• Damping step: Carried out after each time step

$$\hat{\mathbf{u}}_{lm}^{i,mod} = \exp\left(-\frac{const\cdot N\Delta t}{h_i} \left(\frac{l+m}{N}\right)^{2p}\right) \hat{\mathbf{u}}_{lm}^i$$

 \rightsquigarrow Multiplication by filter function $\sigma(\eta) = \exp(-\alpha_i \cdot \eta^{2p})$

 \bullet Resulting from SV modified equation on ${\mathbb T}$

Meister, Ortleb, Sonar:

- On Spectral Filtering for Discontinuous Galerkin Methods on Unstructured Triangular Grids, NMPDE, '11

- New Adaptive Modal and DTV Filtering Routines for the DG Method on Triangular Grids applied to the Euler Equations, GEM, '12

Shock Indication and Adaptivity

$$\hat{\mathbf{u}}_{lm}^{i,mod} = \sigma(\eta) \cdot \hat{\mathbf{u}}_{lm}^{i}$$

$$\sigma(\eta) = \exp\left(-\frac{\mathbf{s}_{i}}{\sigma_{i}} \alpha_{i} \eta^{2p}\right) \quad \text{Modified filter function}$$

 $\mathbf{s}_i = \begin{cases} \tilde{\mathbf{s}}_i & \text{if } \tilde{\mathbf{s}}_i > tol \\ 0 & \text{otherwise} \end{cases}$

tol = 0.01

Resolution indicator
$$\omega^{res}$$

$$\omega_i^{res} = \frac{\sum_{l+m=N} (\hat{u}_{lm}^i)^2}{\sum_{l+m
$$\tilde{s}_i = \min \left\{ 1000(5N^4 + 1)\omega_i^{res}, 1 \right\}$$$$

Persson, Peraire '06; Barter, Darmofal '07

- Computational domain: $[0, 40] \times [0, 40]$
- Initial conditions:

$$H(x, y, 0) = \begin{cases} 2.5 & \text{if } (x - 20)^2 + (y - 20)^2 \le 2.5^2 \\ 0.5 & \text{else} \end{cases}$$

Boundaries: outflow conditions





Modally filtered DG solution: Water height

N = 6, K = 2200. $\alpha_i = 2 \frac{N \Delta t}{h_i}, p = 1.$



Modally filtered DG solution: Water height

$$N = 6, K = 2200.$$

 $\alpha_i = 2 \frac{N \Delta t}{h_i}, p = 1.$



DTV Filtering on Non-Cartesian Graphs



$$\mathbf{U}^{(0)} = [u_h(\mathbf{x}_j)]_{j \in DTV \text{ nodes}}$$

for k = 0,1,2,...
$$\mathbf{U}^{(k+1)} = DTV(\mathbf{U}^{(k)})$$

Meister, Ortleb, Sonar '11

t = 0.4:



before DTV filtering



after DTV filtering

DTV filtered solutions: Water height

N = 6, K = 2200.49 subtris, $\lambda = 1.$



DTV filtered solutions: Water height

N = 6, K = 2200.49 subtris, $\lambda = 1.$



$$\frac{d}{dt}\int_{\tau_i}\mathbf{u}_h\,\Phi\,d\mathbf{x} + \int_{\partial\tau_i}\mathbf{F}^{WB}(\mathbf{u}^-_*,\mathbf{u}^+_*,H^-,\mathbf{n})\,\Phi\,d\sigma - \int_{\tau_i}\mathcal{F}(\mathbf{u}_h)\cdot\nabla\Phi\,d\mathbf{x} = \int_{\tau_i}\mathbf{s}_h\,\Phi\,d\mathbf{x}$$

•
$$\mathbf{F}^{WB}(\mathbf{u}_{*}^{-},\mathbf{u}_{*}^{+},H^{-},\mathbf{n}) = \mathbf{F}^{num}(\mathbf{u}_{*}^{-},\mathbf{u}_{*}^{+},\mathbf{n}) + \begin{pmatrix} 0\\ \frac{n_{1}g}{2}\left((H^{-})^{2}-(H^{-}_{*})^{2}\right)\\ \frac{n_{2}g}{2}\left((H^{-})^{2}-(H^{-}_{*})^{2}\right) \end{pmatrix}$$

Hydrostatic reconstruction Audusse et al. '04

$$\begin{aligned} \mathbf{u}_{*}^{\pm} &= (H_{*}^{\pm}, H_{*}^{\pm} \cdot (v_{1})^{\pm}, H_{*}^{\pm} \cdot (v_{2})^{\pm})^{T}, \\ H_{*}^{\pm} &= \max\left\{0, H^{\pm} + b^{\pm} - \max\left\{b^{-}, b^{+}\right\}\right\} \end{aligned}$$

$$\frac{d}{dt}\int_{\tau_i}\mathbf{u}_h\,\Phi\,d\mathbf{x} + \int_{\partial\tau_i}\mathbf{F}^{WB}(\mathbf{u}^-_*,\mathbf{u}^+_*,H^-,\mathbf{n})\,\Phi\,d\sigma - \int_{\tau_i}\mathcal{F}(\mathbf{u}_h)\cdot\nabla\Phi\,d\mathbf{x} = \int_{\tau_i}\mathbf{s}_h\,\Phi\,d\mathbf{x}$$

•
$$\mathbf{F}^{WB}(\mathbf{u}_{*}^{-},\mathbf{u}_{*}^{+},H^{-},\mathbf{n}) = \mathbf{F}^{num}(\mathbf{u}_{*}^{-},\mathbf{u}_{*}^{+},\mathbf{n}) + \begin{pmatrix} 0\\ \frac{n_{1}g}{2}\left((H^{-})^{2}-(H^{-}_{*})^{2}\right)\\ \frac{n_{2}g}{2}\left((H^{-})^{2}-(H^{-}_{*})^{2}\right) \end{pmatrix}$$

Hydrostatic reconstruction Audusse et al. '04

- Higher order quadrature necessary
 - $\int_{\tau} \left(\frac{1}{2} g H^2 \right)_x \Phi d\mathbf{x} = \int_{\tau} g H H_x \Phi d\mathbf{x} = \int_{\tau} g H b_x \Phi d\mathbf{x}$

• WB filter: Indicator
$$\omega_i^{res} = \sum_{l+m=N} (\hat{u}_{lm}^i)^2 \cdot \left(\sum_{l+m$$

now depending on $\hat{u}^i_{lm} = \hat{H}^i_{lm} + \hat{b}^i_{lm}$

Small Perturbation Test

Smooth bottom elevation on $\Omega = [0,2] \times [0,1]$:

$$b(\mathbf{x}) = 0.8e^{-5(x-0.9)^2 - 50(y-0.5)^2}$$



Initial water height

$$H(\mathbf{x},0) = \begin{cases} 1 - b(\mathbf{x}) + 0.01, & \text{if } 0.05 \le x \le 0.15, \\ 1 - b(\mathbf{x}), & \text{else.} \end{cases}$$

Initial velocity: $\mathbf{v} = \mathbf{0}$



Small Perturbation Test

Water surface w = H + b (for N = 2, K = 46360, $\alpha_i = 5 \frac{N \Delta t}{h_i}$, p = 1)



(30 contour levels from 0.99 to 1.01)

Problem: boundary conditions

Positivity Treatment

Xing/Zhang/Shu-approach assumes

- Non-negative cell means \bar{H}_i^n at time t^n
- Positivity preserving flux (e.g. HLL, Lax-Friedrichs)
- Non-negative values of H at certain quadrature nodes



Zhang, Xia, Shu '11: Maximum-principle-satisfying and PP DG schemes on triangular grids (scalar & Euler equations)

 \Rightarrow Non-negative cell means at next time level t^{n+1}

$$\bar{H}_i^{n+1} = \bar{H}_i^n - \frac{\Delta t}{|\tau_i|} \int_{\partial \tau_i} F_1^{WB}(\mathbf{u}_*^{n,-}, \mathbf{u}_*^{n,+}, \mathbf{n}) \, d\sigma$$

Under suitable CFL-like time step restriction

Positivity Treatment

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Under suitable CFL-like time step restriction

• Comp. velocities via
$$\mathbf{v} = \begin{cases} \mathbf{0} & \text{if } H < 10^{-6} \\ \frac{2H \cdot (H\mathbf{v})}{H^2 + \max\{H^2, \epsilon\}} & \textit{else} (Bryson \text{ et al.}) \end{cases}$$

• Filter modification: $\hat{\mathbf{u}}_{lm}^{i,mod} = \exp\left(-\mathbf{s}_{i} \alpha_{i} \eta^{2p}\right) \cdot \hat{\mathbf{u}}_{lm}^{i}$ $\mathbf{s}_{i} = (\bar{H}_{i})^{-1} \cdot \min\left\{1000(5N^{4}+1)\omega_{i}^{res}, 0\right\}$ Paraboloidal vessel on $\Omega = [-2, 2]^2$:

$$b(\mathbf{x}) = 0.1(x^2 + y^2).$$



2D cut at y = 0

Periodic analytical solution known

$$\begin{array}{lll} H({\bf x},t) &=& \max\{0,0.05(2x\cos(\omega t)+2y\sin(\omega t))+0.075-b({\bf x}))\},\\ v_1({\bf x},t) &=& -0.1\omega\sin(\omega t),\\ v_2({\bf x},t) &=& 0.1\omega\cos(\omega t), \qquad \omega = \sqrt{0.2g} \end{array}$$

Oscillating Lake

Water surface w = H + b (for N = 2, K = 23138, $\alpha_i = 10 \frac{N \Delta t}{h_i}$, p = 1)



1.6

 $t = T_{per}/2^{x}$





Positivity Treatment – Implicit Euler Case

Extension of Xing/Zhang/Shu-approach

- Non-negative cell means \bar{H}_i^n at time t^n
- Positivity preserving flux (e.g. HLL, Lax-Friedrichs)
- Non-negative values of H at certain quadrature nodes

 \Rightarrow Non-negative cell means at next time level t^{n+1}

$$\bar{H}_i^{n+1} = \bar{H}_i^n - \frac{\Delta t}{|\tau_i|} \int_{\partial \tau_i} F_1^{WB}(\mathbf{u}_*^{n+1,-}, \mathbf{u}_*^{n+1,+}, \mathbf{n}) \, d\sigma$$

No time step restriction! Implicit Euler is unconditionally SSP

Higueras '05

Why implicit?

N = 0 (first order): explicit vs. implicit Euler time integration output time $t = T_{per}/6 \approx 0.7475$





stiff grid, $S = \frac{max_area}{min_area} = 413.7$

Stiffness	Avg. Δt	Avg. Δt	CPU_{EX} Avg. iter. per Δt		
	(EX)	(IM)	CPU	Newton	GMRES
6.5	9.0e-4	1.1e-2	1.10	4.6	6.6
25.9	4.6е-4	5.5e-3	1.31	4.0	5.1
103.4	2.3e-4	2.7e-3	1.32	4.0	5.0
413.7	1.1e-4	1.4e-3	2.07	2.9	3.0

Unconditionally PP schemes: implicit and higher order?

Implicit SSP RK schemesNo unconditionally SSP scheme of
order ≥ 2

Gottlieb, Shu, Tadmor '01

Hence no unconditionally PP implicit RK scheme of order ≥ 2

L-stable SDIRK2

Semidiscrete DG scheme

$$\frac{d}{dt}\mathbf{u}(t) = \mathcal{L}_{h}(\mathbf{u}(t)) \qquad \qquad \frac{\alpha}{1} \frac{\alpha}{1-\alpha} \frac{\alpha}{\alpha} \qquad \alpha = 1 - \frac{\sqrt{2}}{2}$$

(boundary terms neglected)

$$\begin{array}{c|cccc} 1 & 1-\alpha & \alpha \\ \hline & 1-\alpha & \alpha \end{array} \qquad \alpha = 1-\alpha$$

$$\mathbf{u}^{(1)} = \mathbf{u}^{n} + \alpha \Delta t \mathcal{L}_{h} \left(\mathbf{u}^{(1)} \right) \quad \text{OK}$$
(cut-off timestep with implicit Euler)
$$\mathbf{u}^{(2)} = \underbrace{\mathbf{u}^{n} + (1 - \alpha) \Delta t \mathcal{L}_{h} \left(\mathbf{u}^{(1)} \right)}_{\text{NOT positivity preserving}} + \alpha \Delta t \mathcal{L}_{h} \left(\mathbf{u}^{(2)} \right)$$

$$\rightarrow \text{ modification necessary}$$

 $u^{n+1} = u^{(2)}$

The DG scheme ...

... for cell means of water height $ar{H}_i(t) = rac{1}{| au_i|}\int_{ au_i} H_h(\mathbf{x},t)\,d\mathbf{x}$

$$\frac{d}{dt}(|\tau_i|\bar{H}_i) = -\sum_{j\in N(\tau_i)} \int_{\Gamma_{ij}} F_1^{WB}(\mathbf{u}_*^-, \mathbf{u}_*^+, \mathbf{n}) d\sigma$$
$$= \sum_{j\in N(\tau_i)} p_{ij} - \sum_{j\in N(\tau_i)} d_{ij}$$

$$p_{ij} = \max\left\{0, -\int_{\Gamma_{ij}} F_1^{WB}(\mathbf{u}^-_*, \mathbf{u}^+_*, \mathbf{n}) \, d\sigma\right\} = d_{ji} \ge 0$$

$$d_{ij} = \max\left\{0, +\int_{\Gamma_{ij}} F_1^{WB}(\mathbf{u}^-_*, \mathbf{u}^+_*, \mathbf{n}) \, d\sigma\right\} = p_{ji} \ge 0$$

$$\Rightarrow \begin{cases} p_{ij} - d_{ij} = -\int_{\Gamma_{ij}} F_1^{WB}(\mathbf{u}_*^-, \mathbf{u}_*^+, \mathbf{n}) \, d\sigma \\ p_{ij} = d_{ji} \end{cases}$$

What's the use of this reformulation?

What's the use of this reformulation?

The Patankar Trick for production-destruction equations:

$$\frac{d}{dt}c_i = \sum_{i=1}^{l} p_{ij}(\mathbf{c}) - \sum_{i=1}^{l} d_{ij}(\mathbf{c}), \quad p_{ij} = d_{ji} \quad (i, j = 1, \dots, l)$$

positive quantities c_i , e.g. concentrations

explicit Euler

$$c_i^{n+1} = c_i^n + \Delta t \left(\sum_{i=1}^l p_{ij}(\mathbf{c}^n) - \sum_{i=1}^l d_{ij}(\mathbf{c}^n) \right)$$
 NO positive scheme

Patankar-Euler (PE)

$$c_i^{n+1} = c_i^n + \Delta t \left(\sum_{i=1}^l p_{ij}(\mathbf{c}^n) - \sum_{i=1}^l d_{ij}(\mathbf{c}^n) \frac{c_i^{n+1}}{c_i^n} \right)$$

$$\Rightarrow \underbrace{(1 + \Delta t \sum_{i=1}^l d_{ij}(\mathbf{c}^n)(c_i^n)^{-1})}_{>0} c_i^{n+1} = \underbrace{c_i^n + \Delta t \sum_{i=1}^l p_{ij}(\mathbf{c}^n)}_{>0}$$

positive but NOT conservative

What's the use of this reformulation?

The Patankar Trick for production-destruction equations:

$$rac{d}{dt}c_i = \sum_{i=1}^l p_{ij}(\mathbf{c}) - \sum_{i=1}^l d_{ij}(\mathbf{c}), \qquad p_{ij} = d_{ji}$$

positive quantities c_i , e.g. concentrations

explicit Euler

$$c_i^{n+1} = c_i^n + \Delta t \left(\sum_{i=1}^{l} p_{ij}(\mathbf{c}^n) - \sum_{i=1}^{l} d_{ij}(\mathbf{c}^n) \right)$$
 NO positive scheme

Modified PE scheme Burchard, Deleersnijder, Meister '03

$$c_i^{n+1} = c_i^n + \Delta t \left(\sum_{i=1}^l p_{ij}(\mathbf{c}^n) \frac{c_j^{n+1}}{c_j^n} - \sum_{i=1}^l d_{ij}(\mathbf{c}^n) \frac{c_i^{n+1}}{c_i^n} \right)$$

positive and conservative

L-stable SDIRK2

$$\frac{d}{dt} \mathbf{u}(t) = \mathcal{L}_{h}(\mathbf{u}(t))$$

$$\mathbf{u}^{(1)} = \mathbf{u}^{n} + \alpha \Delta t \mathcal{L}_{h}(\mathbf{u}^{(1)}) \quad \text{OK}$$
(cut-off timestep with implicit Euler)
$$\mathbf{u}^{(2)} = \underbrace{\mathbf{u}^{n} + (1 - \alpha) \Delta t \mathcal{L}_{h}(\mathbf{u}^{(1)})}_{\text{replace by } \mathbf{u}^{(2)}_{*} \geq 0} + \alpha \Delta t \mathcal{L}_{h}(\mathbf{u}^{(2)}) \quad \text{OK}$$

$$u^{(2)}_{*,i} = u^{n}_{i} + (1 - \alpha) \Delta t \sum_{j \in \mathcal{N}(\tau_{i})} \left(p_{ij} \frac{u^{(2)}_{*,j}}{\tilde{u}_{j}} - d_{ij} \frac{u^{(2)}_{*,i}}{\tilde{u}_{i}} \right)$$

$$\tilde{u}_{i} = \max \left\{ \epsilon, u^{n}_{i} + (1 - \alpha) \Delta t \left[\mathcal{L}_{h}(\mathbf{u}^{(1)}) \right]_{i} \right\}, \quad \epsilon = 10^{-10}$$

 $u^{n+1} = u^{(2)}$

Accuracy Study for Production-Destruction ODE

Simple nonlinear model c(t) positive constituents

$$c_1' = -\frac{c_2c_1}{c_1+1} = -d_{12}$$

$$c_2' = \frac{c_2c_1}{c_1+1} - ac_2 = p_{21} - d_{23}$$

$$c_3' = ac_2 = p_{32}$$

$$a = 0.3$$

 $\mathbf{c}^0 = (9.98, 0.01, 0.01)^7$



Truncation errors:

$$E(\Delta t) = \frac{\sqrt{\frac{1}{N_{\Delta t}}\sum_{n=1}^{N_{\Delta t}} (c_1(n\Delta t) - c_1^n)^2}}{\frac{1}{N_{\Delta t}}\sum_{n=1}^{N_{\Delta t}} c_1(n\Delta t)}$$

MPSDIRK2 for Oscillating Lake Test



Output time $t = T_{per}/6$, N = 2, avg. $\Delta t = 0.0054$ Stiffness S = 25.9 (K = 23284), Filter parameters $\alpha_i = 10 \frac{N\Delta t}{h_i}$, p = 1

No.	Avg.	CPI Lin c	Patankar	Avg. iter. per Δt	
Δt	Δt		steps	Newton	GMRES
139	0.0054	599	0	9	16
28	0.0267	1552	1	26	229
17	0.044	16071	8	60	4027

Conclusion

- Well-balanced and positivity preserving DG scheme for SWE
- Shock capturing by filtering procedures
- Unconditionally positiv implicit time integration

Outlook

- Improvement of iterative solver
- Higher order implicit PP schemes
- $\bullet~$ Inclusion of bottom friction \rightarrow stiff source terms
- IMEX time integration