

CWENOZ and 2D

Course: High order reconstructions in hyperbolic
conservation and balance laws

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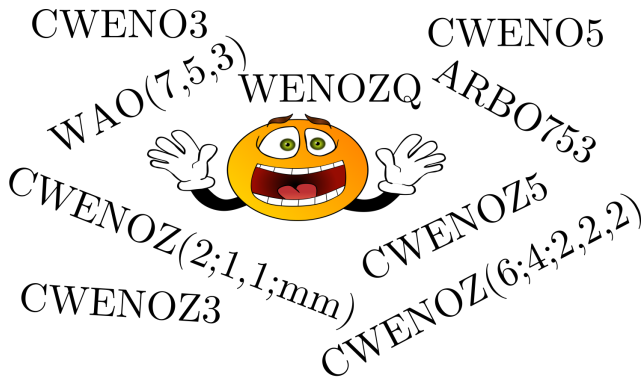
Shenzhen, March 2024





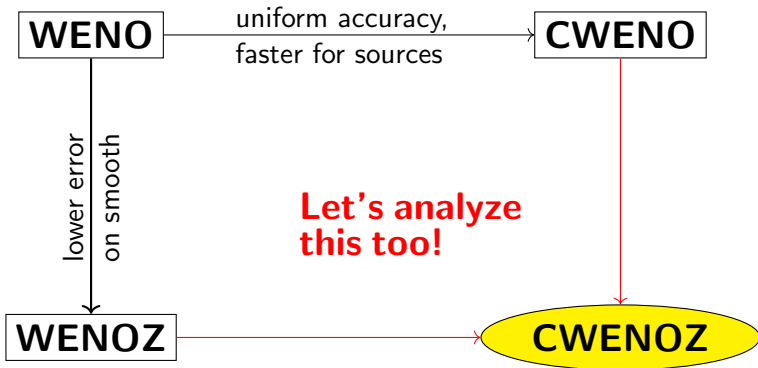
Disclaimer

You might feel like this:



but we'll try to fix this!

Why CWENOZ?



Also, some CWENOZ constructions were already employed under names like "generalized WENO" and WENO-AO.

The CWENO(Z) master equation

Freely choose $d_0, \dots, d_n \in (0, 1)$ such that $\sum_0^n d_k = 1$

$$\forall x \in \text{cell} : P_{\text{opt}}(x) = d_0 P_0(x) + \sum_{k=1}^M d_k P_k(x) \quad (\text{linear})$$

$$P_0(x) = \frac{P_{\text{opt}} - \sum_{k=1}^n d_k P_k}{d_0}$$

$$\alpha_i = d_i \left(1 + \left(\frac{\tau}{\text{OSC}[P_i] + \epsilon} \right)^t \right)$$

$$\forall x \in \text{cell} : R(x) = \omega_0 P_0(x) + \sum_{k=1}^n \omega_k P_k(x) \quad (\text{nonlinear})$$

Taylor expansions of multidimensional smoothness indicators

Using the multi-index notation $\beta = (\beta_1, \dots, \beta_d)$, when $\mathbf{x} \in \mathbb{R}^d$,

$$OSC[q] := \sum_{|\beta| \geq 1} \Delta \mathbf{x}^{2\beta-1} \int_{\Omega_0} (\partial_\beta q(\mathbf{x}))^2 d\mathbf{x}.$$

Proposition

Let \mathcal{S} be a stencil around the reconstruction cell and let $q(\mathbf{x})$ be a polynomial approximating a regular function $u(\mathbf{x})$ with accuracy $\geq g$,

$$\Rightarrow OSC[q] = B_g + R[q]$$

- B_g depends on g , but not on \mathcal{S} .
- $R[q] = o(B_g)$ and depends on \mathcal{S} .

B_g can be given an explicit expression in terms of the derivatives of u at the reconstruction cell center.

Example: CWENO(Z)3 indicators

For CWENOZ3 we have

$$l_1 = \text{OSC}[P_L] \quad l_2 = \text{OSC}[P_R] \quad l_0 = \text{OSC}[P_{\text{opt}}]$$

$$\begin{aligned} l_1 &= B_1 - u'(0)u''(0)\Delta x^3 + \left(\frac{5}{12}u'(0)u'''(0) + \frac{1}{4}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5), \\ l_2 &= B_1 + u'(0)u''(0)\Delta x^3 + \left(\frac{5}{12}u'(0)u'''(0) + \frac{1}{4}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5), \\ l_0 &= B_1 + \left(\frac{5}{12}u'(0)u'''(0) + \frac{13}{12}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5). \end{aligned}$$

for

$$B_1 = (u'(0))^2 \Delta x^2$$

Example: CWENO(Z)5 indicators

$$l_1 = u'(0)^2 \Delta x^2 + \frac{13}{12} u''(0)^2 \Delta x^4 - \frac{7}{12} u'(0) u'''(0) \Delta x^4 + \mathcal{O}(\Delta x^5)$$

$$l_2 = u'(0)^2 \Delta x^2 + \frac{13}{12} u''(0)^2 \Delta x^4 + \frac{5}{12} u'(0) u'''(0) \Delta x^4 + \mathcal{O}(\Delta x^6)$$

$$l_3 = u'(0)^2 \Delta x^2 + \frac{13}{12} u''(0)^2 \Delta x^4 - \frac{7}{12} u'(0) u'''(0) \Delta x^4 + \mathcal{O}(\Delta x^5)$$

$$I[P_{\text{opt}}] = \underbrace{u'(0)^2 \Delta x^2 + \frac{13}{12} u''(0)^2 \Delta x^4}_{B_2} + \frac{1}{12} u'(0) u'''(0) \Delta x^4 + \mathcal{O}(\Delta x^6)$$

Remark: $R = o(B_2)$ also at critical points (i.e. $u'(0) = 0$)

For more terms in the expansions and for the 7th and 9th order case, see the Supplementary Material of the paper.



Main theorem

$$\alpha_i = d_i \left(1 + \left(\frac{\tau}{\text{OSC}[P_i] + \Delta x^{\hat{m}}} \right)^{\ell} \right)$$

Theorem

Assume that

- $P_1(\mathbf{x}), \dots, P_m(\mathbf{x}) \in \mathbb{P}_g$ and $P_{\text{opt}}(\mathbf{x}) \in \mathbb{P}_G$ in the CWENOZ scheme have accuracy $\geq M$
- M, \hat{m}, ℓ satisfy

$$\hat{m} \leq 2M + 1 \quad (1a)$$

$$\ell(2M + 2 - \hat{m}) \geq G - g - 1 \quad (1b)$$

Then, on smooth data, the CWENOZ scheme achieves the optimal order $G + 1$ as $\Delta x \rightarrow 0$. $\ell[\theta(\tau)] \min(\hat{m}, 2M) \geq G - g - 1$ (1c)

$$\theta(f) = n \Leftrightarrow f = C \Delta x^n + o(\Delta x^n)$$



In practice...

$$\alpha_i = d_i \left(1 + \left(\frac{\tau}{\text{OSC}[P_i] + \Delta \mathbf{x}^{\hat{m}}} \right)^{\ell} \right)$$

- if $\theta(\tau) > 2M$,
there exist ℓ for optimal convergence for any $\hat{m} \leq 2M + 1$
- else,
there exist ℓ for optimal convergence for all \hat{m} s.t. $\hat{m} < \theta(\tau) < 2M + 1$
- in any case, the larger is $\theta(\tau)$, the smaller ϵ and smaller ℓ are needed to achieve optimal convergence.

Take home message:

Choose λ_k s.t.

$$\tau = \sum_{k=0}^m \lambda_k l_k$$

is as small as possible!

The “squeeze τ ” game

Since

$$I_k = B_M + R$$

with $R = o(B_M)$ then

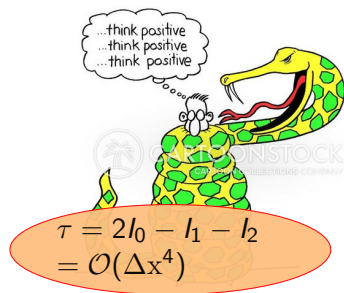
For free, any combination

$$\tau = \sum_k \lambda_k I_k$$

with $\sum_k \lambda_k = 0$ will grant that $\tau = o(B_M)$.

But look closely at the CWENOZ3 case:

$$\begin{aligned} I_1 &= B_1 - u'(0)u''(0)\Delta x^3 + \left(\frac{5}{12}u'(0)u'''(0) + \frac{1}{4}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5), \\ I_2 &= B_1 + u'(0)u''(0)\Delta x^3 + \left(\frac{5}{12}u'(0)u'''(0) + \frac{1}{4}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5), \\ I_0 &= B_1 + \left(\frac{5}{12}u'(0)u'''(0) + \frac{13}{12}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5). \end{aligned}$$



CWENOZ optimal τ in 1D (uniform grids)

- CWENOZ 3

$$\hat{\tau}_3 = \left| OSC[P_L^1] + OSC[P_R^1] - 2 OSC[P_{opt}] \right| = \mathcal{O}(\Delta x^4)$$

instead of $\tau_3 = \mathcal{O}(\Delta x^3)$ without using $OSC[P_{opt}]$.

- CWENOZ 5

$$\hat{\tau}_5 = \left| OSC[P_L^2] + 4 OSC[P_C^2] + OSC[P_R^2] - 6 OSC[P_{opt}] \right| = \mathcal{O}(\Delta x^6)$$

instead of $\tau_5 = \mathcal{O}(\Delta x^5)$ without using $OSC[P_{opt}]$.

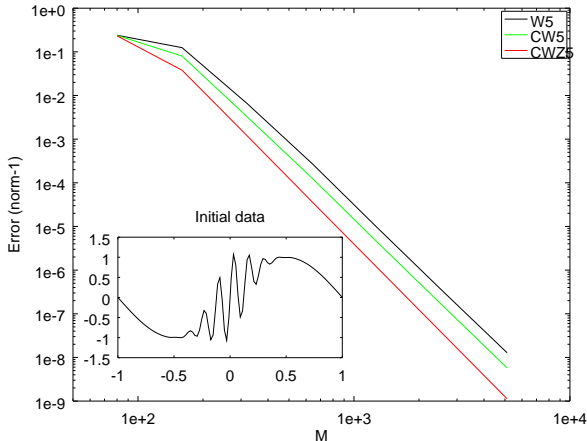
- for **higher orders** the optimal definition for WENOZ is also optimal for CWENOZ, but you may still define τ involving also $OSC[P_{opt}]$

Is CWENOZ really better? (1)

- more versatile than WENO
- little extra effort than CWENO
- but yields **better errors**...

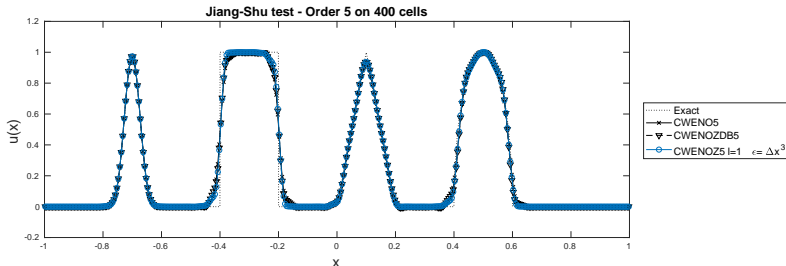
$$u_t + u_x = 0$$

schemes of order 5

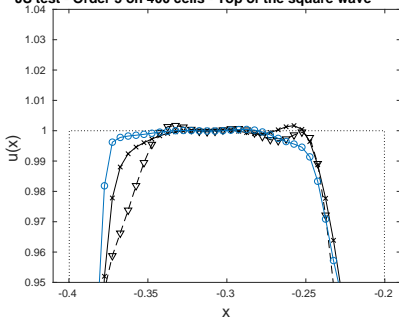


Jiang-Shu linear transport test with CWENOZ5

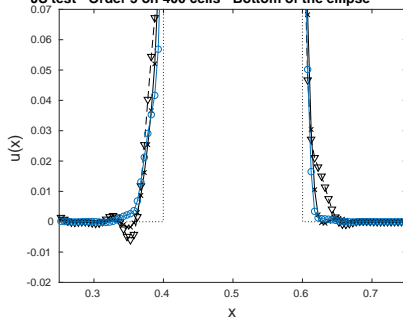
- ... and often even **less oscillations!**

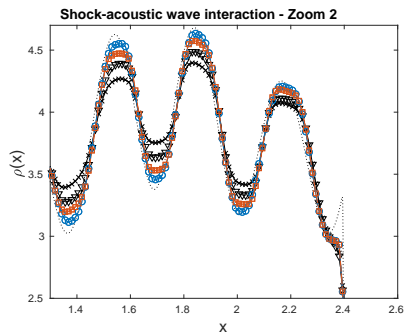
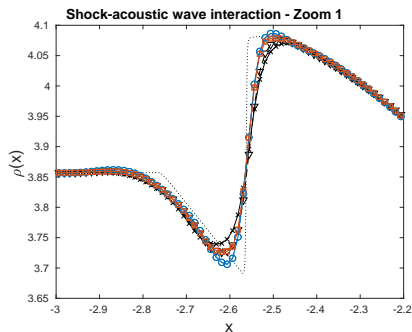
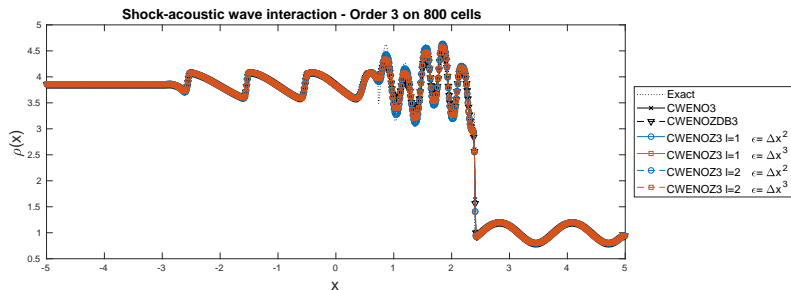


JS test - Order 5 on 400 cells - Top of the square wave



JS test - Order 5 on 400 cells - Bottom of the ellipse





Open-source implementation

claw1dArena

- Downloadable from [zenodo.org](https://zenodo.org/doi/10.5281/zenodo.2641724)
DOI 10.5281/zenodo.2641724
- Developed with numerical experimentation in mind:
 - C++ implementation with very few required libraries
 - choose conservation law at compile time
 - choose any combination of reconstruction, timestepper, numerical flux, well-balancing, discretization parameters, etc at run-time
 - GPL licence: just give credit, if you use it

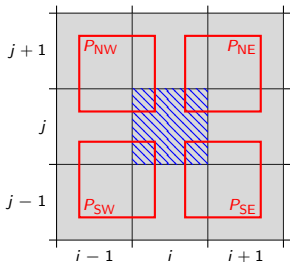
Let's you experiment with all my 1d reconstructions: CWENO, CWENOZ, CWENO-AO, CWENO-boundary

Currently: linear advection, Burgers, SWE (w.b.), Euler, easily extendible to other conservation laws



CWENOZ3 in two space dimensions

$$\text{CWENOZ}(P_{\text{opt}}; P_{\text{NE}}, P_{\text{SE}}, P_{\text{SW}}, P_{\text{NW}}) = \sum_{k=0}^4 \omega_k P_k$$



- $P_{\text{opt}}(x) \in \mathbb{P}_2(x, y)$ on the central 3×3 stencil
- $P_{\text{NE}}, P_{\text{SE}}, P_{\text{SW}}, P_{\text{NW}} \in \mathbb{P}_1(x, y)$ on the 2×2 sub-stencil
- $d_0 = \frac{3}{4}$ and $d_{1,\dots,4} = \frac{1}{16}$

All polynomial are defined in a least square sense:

$$P(x) = \bar{U}_j + \sum_k c_k \phi_k(x)$$

where the basis is such that $\int_{\Omega_j} \phi_k = 0$ and solve least squares for \mathbf{c} .

CWENOZ3 in two space dimensions

Recall that the theorem $\text{OSC}[P] = B_M + R$ holds also in \mathbb{R}^d .

On a 2d uniform Cartesian grid with $\Delta x = \Delta y = h$,

$$\text{OSC}[P_{NE}^{(1)}] = (u_x^2 + u_y^2)h^2 + (+u_x u_{xx} + \frac{2}{3}u_x u_{xy} + \frac{2}{3}u_y u_{xy} + u_y u_{yy})h^3 + \mathcal{O}(h^4)$$

$$\text{OSC}[P_{NW}^{(1)}] = (u_x^2 + u_y^2)h^2 + (-u_x u_{xx} + \frac{2}{3}u_x u_{xy} - \frac{2}{3}u_y u_{xy} + u_y u_{yy})h^3 + \mathcal{O}(h^4)$$

$$\text{OSC}[P_{SE}^{(1)}] = (u_x^2 + u_y^2)h^2 + (+u_x u_{xx} - \frac{2}{3}u_x u_{xy} + \frac{2}{3}u_y u_{xy} - u_y u_{yy})h^3 + \mathcal{O}(h^4)$$

$$\text{OSC}[P_{SW}^{(1)}] = (u_x^2 + u_y^2)h^2 + (-u_x u_{xx} - \frac{2}{3}u_x u_{xy} - \frac{2}{3}u_y u_{xy} - u_y u_{yy})h^3 + \mathcal{O}(h^4)$$

$$\text{OSC}[P_{\text{opt}}] = (u_x^2 + u_y^2)h^2 + \mathcal{O}(h^4)$$

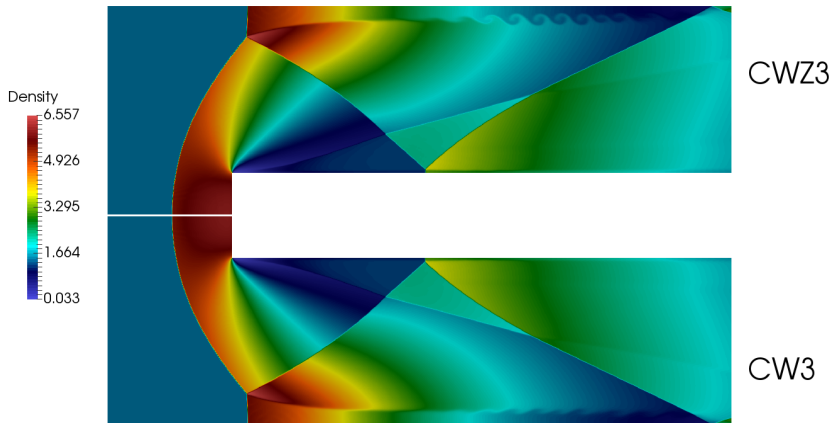
$$\text{In general, } \sum_{k=0}^m \lambda_k = 0, \quad \Rightarrow \quad \boxed{\tau = \lambda_0 \text{OSC}[P_{\text{opt}}] + \sum_{k=1}^4 \lambda_k \text{OSC}[P_k^{(1)}] = \mathcal{O}(h^3)}$$

but the symmetries allow an even better definition of τ :

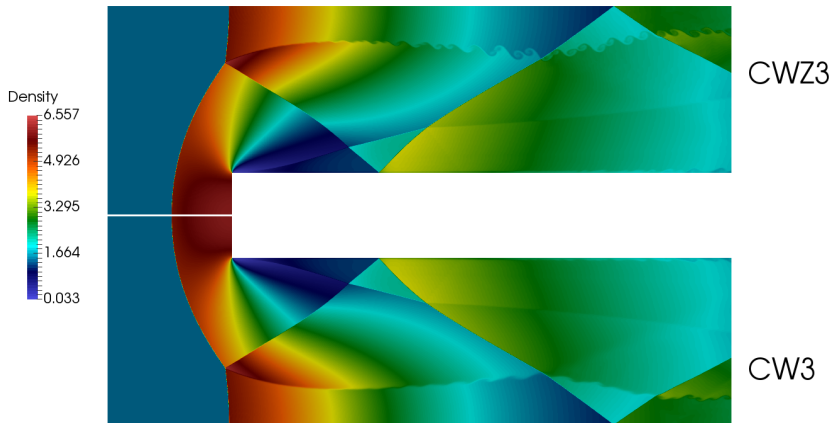
$$\tau = \text{OSC}[P_{NE}^{(1)}] + \text{OSC}[P_{NW}^{(1)}] + \text{OSC}[P_{SE}^{(1)}] + \text{OSC}[P_{SW}^{(1)}] - 4\text{OSC}[P_{\text{opt}}] = \mathcal{O}(h^4)$$



Forward-facing step at $t = 2.4$ with 1M dofs



Forward-facing step at $t = 4.0$ with 4M dofs



Thank you for your kind attention!



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