

# CWENOZ and 2D

Course: High order reconstructions in hyperbolic conservation and balance laws

**Matteo Semplice**

**Dipartimento di Scienza e Alta Tecnologia  
Università dell'Insubria**

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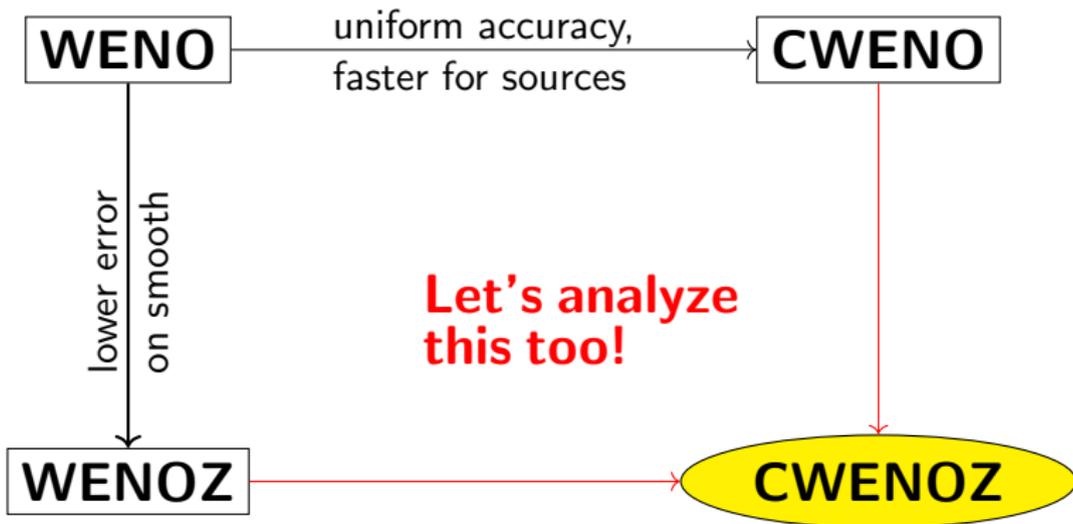
# Disclaimer

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  $\sum_0^n d_k = 1$

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

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

$$\omega_i = \frac{\alpha_i}{\sum \alpha_k}$$

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

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

# Taylor expansions of multidimensional smoothness indicators

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

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

## 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

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

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

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

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

**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  $\geq M$
- $M, \hat{m}, \ell$  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  $\ell[\theta(\tau)] = \min(\hat{m}, 2M) \geq G - g - 1$  (1c) as  $\Delta x \rightarrow 0$ .

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



## In practice...

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

- if  $\theta(\tau) > 2M$ ,  
there exist  $\ell$  for optimal convergence for any  $\hat{m} \leq 2M + 1$
- else,  
there exist  $\ell$  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  $\epsilon$  and smaller  $\ell$  are needed to achieve optimal convergence.

### Take home message:

Choose  $\lambda_k$  s.t.

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

is as small as possible!

# The “squeeze $\tau$ ” game

Since

$$I_k = B_M + R$$

with  $R = o(B_M)$  then

For free, any combination

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

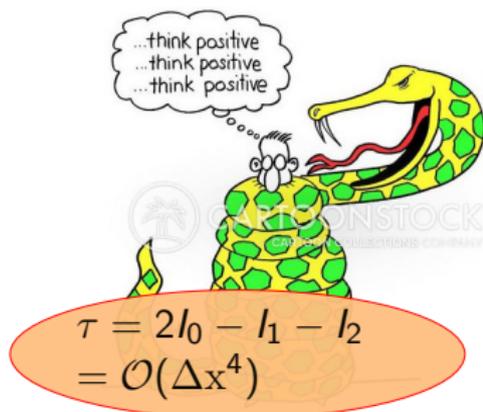
with  $\sum_k \lambda_k = 0$  will grant that  $\tau = o(B_M)$ .

But look closely at the CWENOZ3 case:

$$I_1 = B_1 - u'(0)u''(0)\Delta x^3 + \left( \frac{5}{12}u'(0)u'''(0) + \frac{1}{4}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5),$$

$$I_2 = B_1 + u'(0)u''(0)\Delta x^3 + \left( \frac{5}{12}u'(0)u'''(0) + \frac{1}{4}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5),$$

$$I_0 = B_1 + \left( \frac{5}{12}u'(0)u'''(0) + \frac{13}{12}u''(0)^2 \right) \Delta x^4 + \mathcal{O}(\Delta x^5).$$



# CWENOZ optimal $\tau$ in 1D (uniform grids)

- CWENOZ3

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

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

- CWENOZ5

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

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

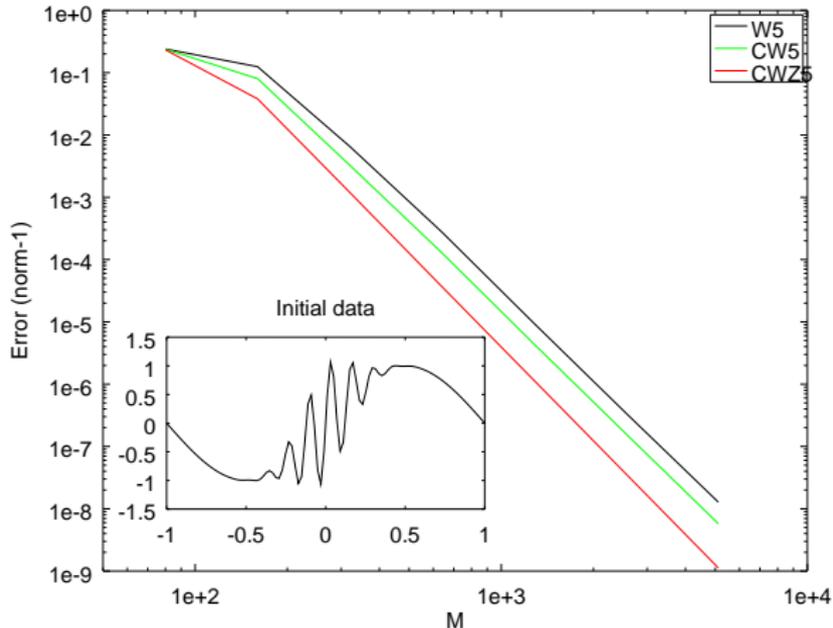
- for **higher orders** the optimal definition for WENOZ is also optimal for CWENOZ, but you may still define  $\tau$  involving also  $OSC[P_{opt}]$

# Is CWENOZ really better? (1)

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

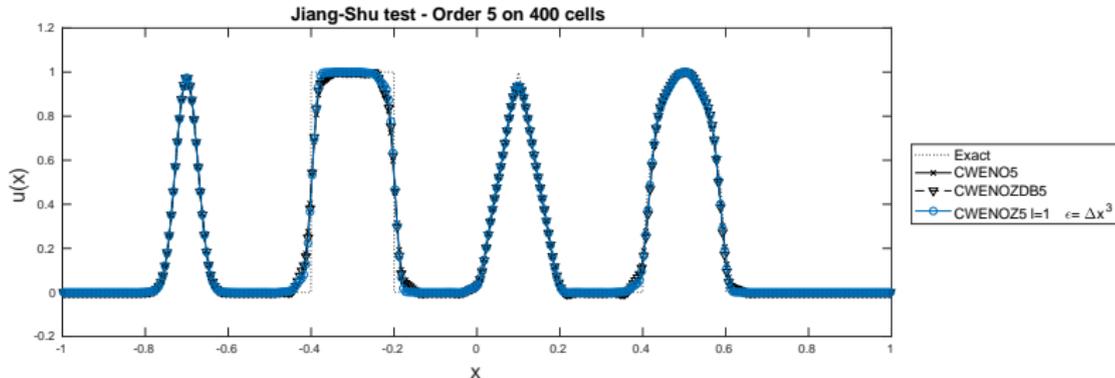
$$u_t + u_x = 0$$

schemes of order 5

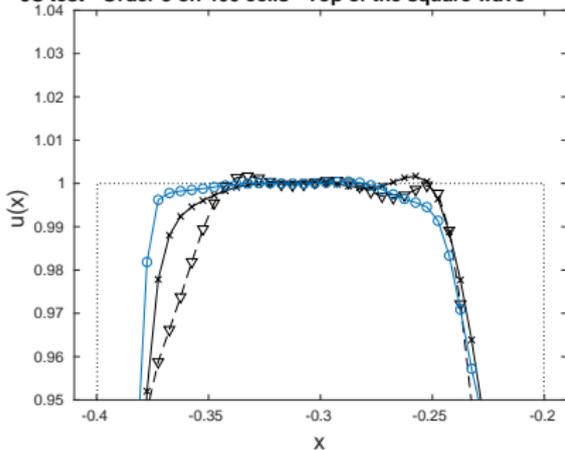


# Jiang-Shu linear transport test with CWENOZ5

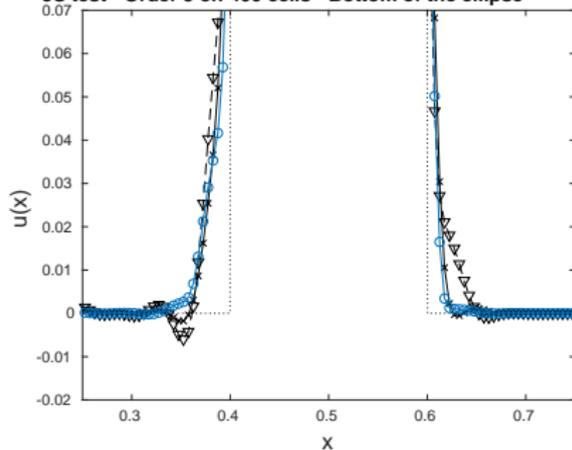
- ... and often even **less oscillations!**

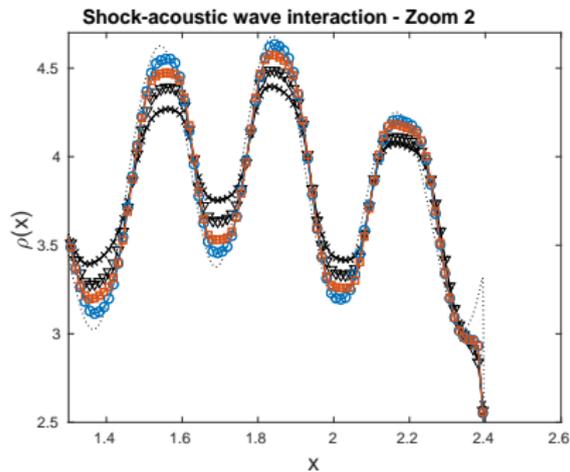
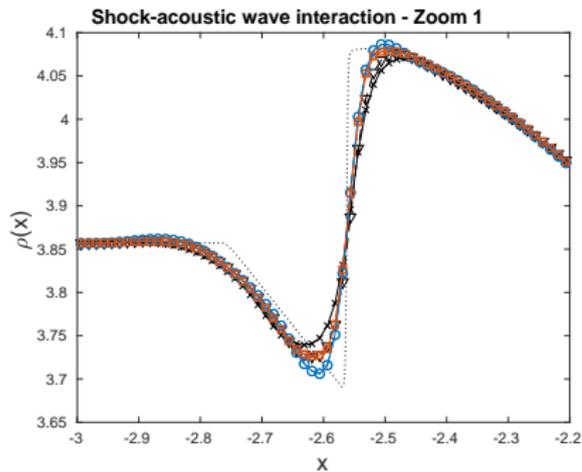
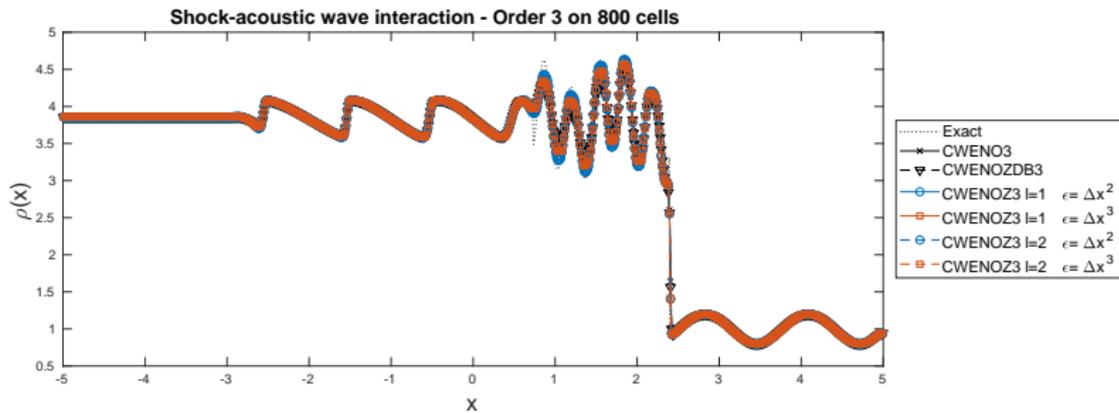


JS test - Order 5 on 400 cells - Top of the square wave



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
- 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

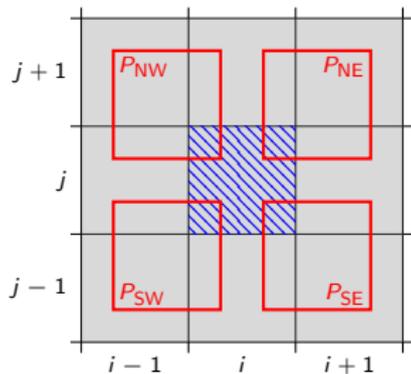
Let's you experiment with all my 1d reconstructions: CWENO, CWENOZ, CWENO-AO, CWENO-boundary

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



## CWENO3 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  $3 \times 3$  stencil
- $P_{\text{NE}}, P_{\text{SE}}, P_{\text{SW}}, P_{\text{NW}} \in \mathbb{P}_1(x, y)$  on the  $2 \times 2$  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  $\mathbf{c}$ .

## CWENO3 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  $\Delta x = \Delta y = h$ ,

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

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

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

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

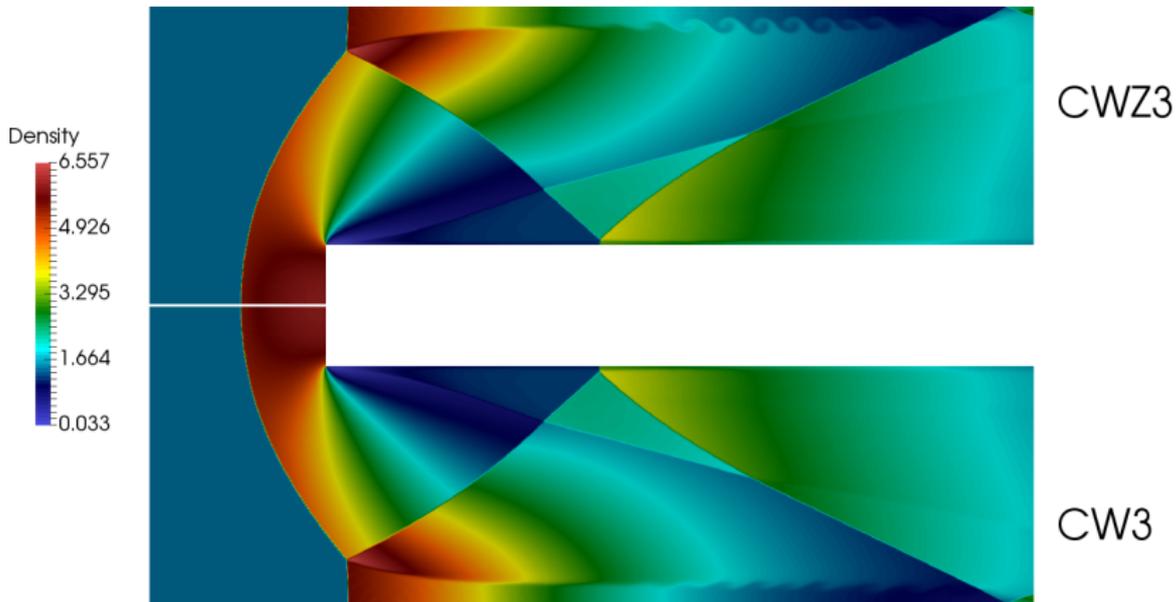
$$\text{OSC}[P_{\text{opt}}] = (u_x^2 + u_y^2)h^2 + \mathcal{O}(h^4)$$

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

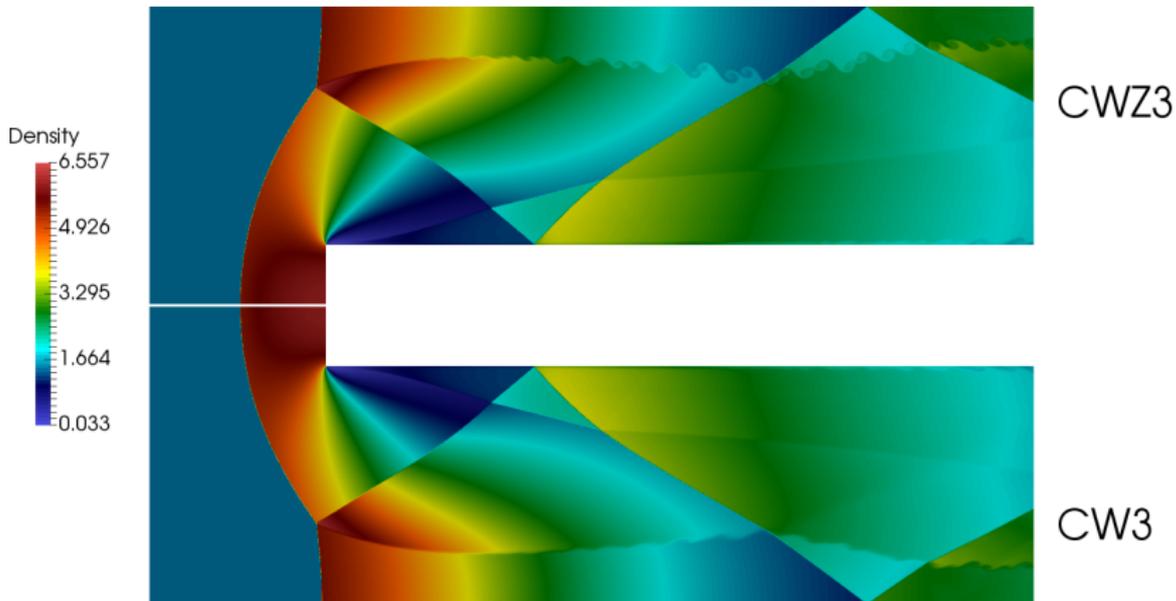
but the symmetries allow an even better definition of  $\tau$ :

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

# 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!



**Matteo Semplice**

[matteo.semplice@uninsubria.it](mailto:matteo.semplice@uninsubria.it)

Dipartimento di Scienza e Alta Tecnologia  
Università dell'Insubria  
Via Valleggio, 11  
Como