

# **Spectral Methods for Discontinuous Problems**

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## Plan of the Talk

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- Spectral Galerkin and Collocation Schemes.
- Approximation Theory: The Gibbs phenomenon
  - The resolution of the Gibbs Phenomenon.
  - Application: Splicing of pictures.
  - Application: The Direct Fourier Method in Computed Tomography.
- Linear Hyperbolic equations with discontinuous solutions.
  - Stability and filtering.
  - Recovering spectral accuracy.
- Non linear Hyperbolic Equations:
  - The Spectral Viscosity Method
  - The Super Viscosity Methods and its relationship to filtering.

## Plan of the Talk (Cont.)

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- Applications:
  - The Leblanc Problem.
  - Interactions of shock waves and hydrogen jets in reactive flows.
  - Recessed cavity flameholders.
  - Richtmyer-Meshkov instabilities.

# Spectral Methods - Introduction

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The solution of an equation is assumed to be in a space  $\mathcal{B}_N$  spanned by smooth

- For periodic problems

$$\phi_k = e^{ikx}$$

- for non-periodic problems in finite intervals

- Either

$$\phi_k = T_k(x)$$

where  $T_k$  are the Chebyshev polynomials.

- Or

$$\phi_k = L_k(x)$$

where  $L_k$  are the Legendre polynomials.



## Spectral Methods - Introduction (Cont.)

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The approximation is obtained in one of two ways:

- Galerkin projection:

$$\mathcal{P}_{\mathcal{N}}F(x) = \sum_{k=0}^N (F, \phi_k) \phi_k(x)$$

where

$$(F, \phi_k) = \int \omega(x) F(x) \phi_k(x) dx$$

## Spectral Methods - Introduction (Cont.)

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- The Pseudo-spectral approximation

$$\mathcal{I}_N F(x) = \sum_{k=0}^N (F, \phi_k)_N \phi_k(x)$$

where

$$(F, \phi_k)_N = \sum_{i=0}^N \omega_i F(x_i) \phi_k(x_i)$$

- $x_i$  and  $\omega_i$  are the Gauss-Lobatto Quadrature nodes and weights respectively
- Thus

$$\int f(x) \omega(x) dx = \sum_{i=0}^N \omega_i f(x_i) \phi_k(x_i)$$

when  $f(x)$  is a polynomial or trigonometrical polynomial of degree  $2N$

## Spectral Methods - Introduction (Cont.)

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- Alternatively

$$\mathcal{I}_N F(x) = \sum_{i=0}^N F(x_i) g_i(x)$$

where

–  $g_i(x)$  is a polynomial of degree  $N$  such that  $g_i(x_k) = \delta_{i,k}$ .

This leads naturally to the formula:

$$\frac{d}{dx} \mathcal{I}_N F(x_k) = \sum_{i=0}^N F(x_i) g'_i(x_k)$$

and defines the *Differentiation Matrix*

$$\mathcal{D} = g'_i(x_k)$$

# Approximation Results

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$$\|F - \mathcal{P}_N F\| \leq K \frac{\|F\|_s}{N^s}$$
$$\|F - \mathcal{I}_N F\| \leq K \frac{\|F\|_s}{N^s}$$

# Approximation Theory

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- We assume that

- $f(x)$  is in  $L^2[-1, 1]$ ;
- there is a subinterval  $[a, b] \subset [-1, 1]$  in which  $f(x)$  is analytic;
- there exists an orthonormal family  $\{\Psi_k(x)\}$ , under a scalar product (.

- Denote

$$f_N(x) = \sum_{k=0}^N (f, \Psi_k) \Psi_k(x)$$

- 

$$\lim_{N \rightarrow \infty} |f(x) - f_N(x)| = 0$$

almost everywhere in  $x \in [-1, 1]$ .

- Denote  $\xi = -1 + 2\frac{x-a}{b-a}$  such that if  $a \leq x \leq b$  then  $-1 \leq \xi \leq 1$ .

## Approximation (Cont.)

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### **Definition:**

The two parameters family  $\{\Phi_k^\lambda(\xi)\}$  is called a **Gibbs complementary** to  $t$

#### (a) **Orthogonality**

$$\langle \Phi_k^\lambda(\xi), \Phi_l^\lambda(\xi) \rangle_\lambda = \delta_{kl}.$$

#### (b) **Spectral Convergence**

The expansion of an analytic function  $g(\xi)$  in the basis  $\Phi_k^\lambda(\xi)$  converges ex

$$\max_{-1 \leq \xi \leq 1} \left| g(\xi) - \sum_{k=0}^{\lambda} \langle g, \Phi_k^\lambda \rangle_\lambda \Phi_k^\lambda(\xi) \right| \leq e^{-q_1 \lambda}, \quad q_1 > 0$$

#### (c) **The Gibbs Condition**

There exists a number  $\beta < 1$  such that if  $\lambda = \beta N$  then

$$\left| \langle \Phi_l^\lambda(\xi), \Psi_k(x(\xi)) \rangle_\lambda \right| \max_{-1 \leq \xi \leq 1} |\Phi_l^\lambda(\xi)| \leq \left( \frac{\alpha N}{k} \right)^\lambda, \quad k > N, l \leq$$

## Approximation (Cont.)

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### Comments:

- Condition (b) implies that the expansion of a function  $g$  in the basis  $\{\Phi_l^\lambda\}$  converges exponentially fast if  $g$  is analytic in  $-1 \leq \xi \leq 1$  (corresponding to  $a \leq x \leq b$ ).
- Condition (c) states that the projection of  $\{\Psi_k\}$  for large  $k$  on the low modes (small  $l$ ) is exponentially small *in the interval*  $-1 \leq \xi \leq 1$ .

## Approximation (Cont.)

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### Theorem

- $f(x) \in L^2[-1, 1]$  and analytic in  $[a, b] \subset [-1, 1]$ .
- $\{\Psi_k(x)\}$  is an orthonormal family with the inner product  $(\cdot, \cdot)$ .
- $\{\Phi_k^\lambda(\xi)\}$  is a Gibbs complementary to the family  $\{\Psi_k(x)\}$  as defined in (a).

Then

$$\max_{a \leq x \leq b} \left| f(x) - \sum_{l=0}^{\lambda} \langle f_N, \Phi_l^\lambda \rangle_\lambda \Phi_l^\lambda(\xi(x)) \right| \leq e^{-qN}, \quad q > 0.$$



## Approximation (Cont.)

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### Comment:

- Even if we have a slowly converging series

$$\sum_{k=0}^N (f, \Psi_k) \Psi_k(x)$$

it is still possible to get a rapidly converging approximation to  $f(x)$  if one can find a function that yields a rapidly converging series to  $f$  as long as *the projection of  $f$  in the basis  $\{\Psi\}$  on the low modes in the new basis is exponentially small*

## Approximation (Cont.)

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### Examples

- In all the following examples, we choose

$$\Phi_k^\lambda(\xi) = \frac{1}{\sqrt{h_k^\lambda}} C_k^\lambda(\xi)$$

where  $C_k^\lambda(\xi)$  is the Gegenbauer polynomial and  $h_k^\lambda$  is the normalization factor. The inner product is defined by

$$\langle f, g \rangle_\lambda = \int_{-1}^1 (1 - \xi^2)^{\lambda - \frac{1}{2}} f(\xi) g(\xi) d\xi$$

We have to check only if:

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$$\left| \int_{-1}^1 (1 - \xi^2)^{\lambda - \frac{1}{2}} C_l^\lambda(\xi) \Psi_k(x(\xi)) d\xi \right| \leq \left( \frac{\alpha N}{k} \right)^\lambda,$$

for  $k > N, l \leq \lambda = \beta N, 0 < \alpha < 1$ .

## Approximation (Cont.)

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### Example 1: Fourier Case

$$\Psi_k(x) = \frac{1}{2} e^{ik\pi x}, |k| \leq \infty.$$

We would need to verify that

$$\left| \int_{-1}^1 (1 - \xi^2)^{\lambda - \frac{1}{2}} e^{ik\pi x(\xi)} C_l^\lambda(\xi) d\xi \right| \leq \left( \frac{\alpha N}{k} \right)^\lambda,$$

for  $k > N, l \leq \lambda = \beta N, 0 < \alpha < 1$ .

There is an explicit formula for the integral

$$\int_{-1}^1 (1 - \xi^2)^{\lambda - \frac{1}{2}} e^{ik\pi x(\xi)} C_l^\lambda(\xi) d\xi = h_l^\lambda \Gamma(\lambda) \left( \frac{2}{\pi k \epsilon} \right)^\lambda i^l (l + \lambda) J_{l+\lambda}(\pi$$

where  $\epsilon = b - a$  and  $J_\nu(x)$  is the Bessel function.

The Gibbs condition is satisfied when

$$\beta = \frac{2\pi\epsilon}{27}.$$

## Approximation (Cont.)

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### Example 2: Legendre Case

$$\Psi_k(x) = \frac{1}{\sqrt{h_k^{\frac{1}{2}}}} C_k^{\frac{1}{2}}(x).$$

We need to verify that

$$\frac{1}{\sqrt{h_k^{\frac{1}{2}}}} \left| \int_{-1}^1 (1 - \xi^2)^{\lambda - \frac{1}{2}} C_k^{\frac{1}{2}}(x(\xi)) C_l^\lambda(\xi) d\xi \right| \leq \left( \frac{\alpha N}{k} \right)^\lambda,$$

for  $k > N, l \leq \lambda = \beta N, 0 < \alpha < 1$ .

This had been verified.

## Approximation (Cont.)

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### Example 3: Gegenbauer Case

$$\Psi_k(x) = \frac{1}{\sqrt{h_k^\mu}} C_k^\mu(x).$$

We need to verify that

$$\frac{1}{\sqrt{h_k^\mu}} \left| \int_{-1}^1 (1 - \xi^2)^{\lambda - \frac{1}{2}} C_k^\mu(x(\xi)) C_l^\lambda(\xi) d\xi \right| \leq \left( \frac{\alpha N}{k} \right)^\lambda,$$

for  $k > N, l \leq \lambda = \beta N, 0 < \alpha < 1$ .

## Computer Tomography

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We have to recover a density function  $f(x, y)$  from its Radon Transform  $p(r, \theta)$

$$p(r, \theta) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \delta(x \cos(\theta) + y \sin(\theta) - r) dx dy$$

**The Slice Theorem:**

$$\hat{p}(\rho, \theta) = \hat{f}(\rho \cos \theta, \rho \sin \theta)$$

where

- $\hat{p}$  is the Fourier Transform (in  $r$ ) of  $p$ ,
- $\hat{f}$  is the two dimensional Fourier Transform of  $f$ .

The DFM

$$p \rightarrow \hat{p} \rightarrow \hat{f} \rightarrow f$$

# Linear Hyperbolic Equations

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Consider first approximating of

$$\frac{\partial U}{\partial t} = \frac{\partial U}{\partial x}$$

by the Fourier method.

**Denote:**

- $u_N(x)$  is the PS solution satisfying

$$\frac{\partial u_N}{\partial t}(x_l) = \sum_{i=0}^N g'_i(x_l) u_N(x_i)$$

- $\vec{u}_i = u_N(x_i)$ .

The PS can be written as

$$\frac{d}{dt} \vec{u} = \mathcal{D} \vec{u} \quad .$$

Since  $\mathcal{D}$  is anti-symmetric, stability is assured.

## The variable coefficient case

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Consider now

$$\frac{\partial U}{\partial t} = a(x) \frac{\partial U}{\partial x}$$

- The matrix  $\mathcal{AD}$  is not skew symmetric.

How to stabilize?

- Rewrite the equation as

$$\frac{\partial U}{\partial t} = \frac{1}{2}a(x) \frac{\partial U}{\partial x} + \frac{1}{2} \frac{\partial a(x)U}{\partial x} - \frac{1}{2}a'(x) \frac{\partial U}{\partial x}$$

Doubles the amount of work. Loses conservation.



## Linear-Analysis

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### Analysis:

Write the equation as follows

$$\begin{aligned}\frac{\partial u_N}{\partial t} &= \mathcal{I}_N a(x) \frac{\partial u_N}{\partial x} \\ &= N_1 + N_2 + N_3\end{aligned}$$

where

$$\begin{aligned}2N_1 &= \frac{\partial \mathcal{I}_N a u_N}{\partial x} + \mathcal{I}_N \left( a(x) \frac{\partial u_N}{\partial x} \right) \\ 2N_2 &= \mathcal{I}_N \left( a(x) \frac{\partial u_N}{\partial x} \right) - \mathcal{I}_N \frac{\partial a u_N}{\partial x} \\ 2N_3 &= \mathcal{I}_N \frac{\partial a u_N}{\partial x} - \frac{\partial \mathcal{I}_N a u_N}{\partial x}\end{aligned}$$

## Linear Analysis (Cont.)

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Now

- 

$$(u_N, N_1 u_N) = 0.$$

Since  $\mathcal{A}\mathcal{D} + \mathcal{D}\mathcal{A}$  is skew symmetric.

- 

$$|(u_N, N_2 u_N)| \leq K \max_{0 \leq x \leq 2\pi} |a'| (u_N, u_N)$$

- 

$$|(u_N, N_3 u_N)| \leq (u_N, (-1)^{p+1} \epsilon_N \frac{\partial^{2p} u_N}{\partial x^{2p}})$$

where  $\epsilon_N \sim N^{1-2p}$

## Stability and Filters

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### Conclusion:

To stabilize the PS method we can write

$$\frac{\partial u_N}{\partial t} = \mathcal{I}_N a(x) \frac{\partial u_N}{\partial x} + (-1)^{p+1} \epsilon_N \frac{\partial^{2p} u_N}{\partial x^{2p}}$$

Consider

$$\frac{\partial u_N}{\partial t} = (-1)^{p+1} \epsilon_N \frac{\partial^{2p} u_N}{\partial x^{2p}}$$

Let

$$u_N = \sum_{k=-N}^N a_k(t) e^{ikx}$$

Then we have

$$\frac{da_k(t)}{dt} = (-1)^{p+1} \epsilon_N (ik)^{2p} a_k(t)$$

Thus

$$a_k(t + \Delta t) = e^{-\epsilon_N k^{2p} \Delta t} a_k(t) \quad (\text{This is a low pass filter!!!})$$

## Stability and Filters (Cont.)

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### Important:

The dissipation in the Legendre case will be therefore

$$(-1)^{p+1} \left( \frac{\partial}{\partial x} (1 - x^2) \frac{\partial}{\partial x} \right)^p$$

# Linear Hyperbolic Equations

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Consider the hyperbolic system of the form

$$\frac{\partial U}{\partial t} = \mathcal{L}U$$

with initial conditions

$$U(t = 0) = U_0$$

- Example: The symmetric hyperbolic system,

$$\frac{\partial U}{\partial t} = \sum_{i=1}^d A_i(x_1, \dots, x_d) \frac{\partial U}{\partial x_i}$$

## Linear Hyperbolic Equations (Cont.)

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Let  $u$  be the Fourier Galerkin approximation:

$$\begin{aligned}(U - u, e^{ikx}) &= 0 & -N \leq k \leq N \\ (U_0 - u_0, e^{ikx}) &= 0 & -N \leq k \leq N\end{aligned}$$

### Theorem

$$\|U - u\| \leq K \|U_0\|_s \frac{1}{N^{s-1}}$$

This fails for piecewise smooth  $U_0$ . This is not surprising since  $U$  is not a smooth

## Linear Hyperbolic Equations (Cont.)

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### **Theorem**

Let  $\phi$  be a smooth function, then

$$|(U(T) - u(T), \phi)| \leq K \|\phi\|_s \frac{1}{N^s}$$

### **Proof**

We consider the adjoint problem:

$$V_t = \mathcal{L}^* V$$

with the smooth initial condition:

$$V(t = 0) = \phi$$

## Linear Hyperbolic Equations (Cont.)

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The Green's identity is derived from

$$\begin{aligned}\frac{d(U(t), V(T-t))}{dt} &= (\mathcal{L}U(t), V(T-t)) - (U(t), \mathcal{L}^*V(T-t)) \\ &= 0\end{aligned}$$

So that

$$(U(T), \phi) = (U_0, V(T))$$

We can also show that

$$(u(T), \phi_N) = (u_0, v(T))$$

where  $v$  is the Galerkin approximation to  $V$ .



## Linear Hyperbolic Equations (Cont.)

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- Since the problem for  $V$  is smooth then

$$\|V(T) - v(T)\| \leq K \frac{\|\phi\|_s}{N^{s-1}}$$

- 

$$(u_0, v(T)) = (U_0, v(T))$$

- 

$$(u(T), \phi_N) = (u(T), \phi)$$

Thus

$$(U(T) - u(T), \phi) = (U_0, V(T) - v(T))$$

## Linear Hyperbolic Equations (Cont.)

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### Comments

- Similar Results are obtained for the collocation (Pseudospectral) method. Here the initial condition  $(u_0)$  has to be the Galerkin approximation to the initial condition.
- The Fourier coefficients of  $u$  approximate those of  $U$  with spectral accuracy. It is possible to postprocess to get spectral accuracy for the point values.
- Alternatively, define

$$\phi = \sum (1 - \eta^2)^{\lambda - \frac{1}{2}} C_k^\lambda(\eta) C_k^\lambda(\xi)$$

for any interval of smoothness, and  $\phi$  vanishes outside.

Then

$$(u, \phi) = u(\xi) \quad \text{with } \textit{spectral} \text{ accuracy.}$$

# Nonlinear Equations

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Consider

$$\frac{\partial U}{\partial t} + \frac{\partial f(U)}{\partial x} = 0$$

The spectral method is

$$\frac{\partial u_N}{\partial t} + \frac{\partial \mathcal{P}_N f(u_N)}{\partial x} = 0$$

The method is **unstable** (as in the linear case!)

In the **Spectral Viscosity Method (SV)**

$$\frac{\partial u_N}{\partial t} + \frac{\partial \mathcal{P}_N f(u_N)}{\partial x} = \epsilon_N (-1)^{s+1} \frac{\partial^s}{\partial x^s} \left[ Q_m(x, t) * \frac{\partial^s u_N}{\partial x^s} \right]$$

## Nonlinear Equations (Cont.)

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$$\frac{\partial u_N}{\partial t} + \frac{\partial \mathcal{P}_N f(u_N)}{\partial x} = \epsilon_N (-1)^{s+1} \frac{\partial^s}{\partial x^s} \left[ Q_m(x, t) * \frac{\partial^s u_N}{\partial x^s} \right]$$

The operator  $Q_m$  can be expressed in the Fourier space:

$$\epsilon_N (-1)^{s+1} \frac{\partial^s}{\partial x^s} \left[ Q_m(x, t) * \frac{\partial^s u_N}{\partial x^s} \right] \sim \epsilon \sum_{m < |k| < N} (ik)^{2s} \hat{Q}_k \hat{u}_k e^{ikx}$$

with

$$\epsilon \sim CN^{2s-1}; \quad m \sim N^\theta, \quad \theta < \frac{2s-1}{2s}; \quad 1 - \left( \frac{m}{|k|} \right)^{\frac{2s-1}{\theta}} \leq \hat{Q}_k$$

In the **Super viscosity**

$$\epsilon (-1)^{s+1} \frac{\partial^{2s}}{\partial x^{2s}}$$

## Nonlinear Equations (Cont.)

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### **Theorem:**

Consider the Fourier Superviscosity method, subject to  $L^\infty$  initial data. Then  $u_N$  converges to the entropy solution.

### **Theorem:**

Let  $u_N$  be the solution of the SV approximation, subject to bounded initial conditions

$$\|u_N(\cdot, 0)\|_{L^\infty} + \epsilon^s \|\partial^s u_N(\cdot, 0)\|_{L^2} \leq \text{Const.}$$

then  $u_N$  converges strongly to the unique entropy solution.

# LeBlanc Problem

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The one-dimensional Euler equation

$$\begin{aligned}\rho_t + (\rho u)_x &= 0 \\ (\rho u)_t + (\rho u u)_x &= 0 \\ E_t + ((E + P)u)_x &= 0\end{aligned}$$

with

$$E = \frac{P}{\gamma - 1} + \frac{1}{2}\rho u^2$$

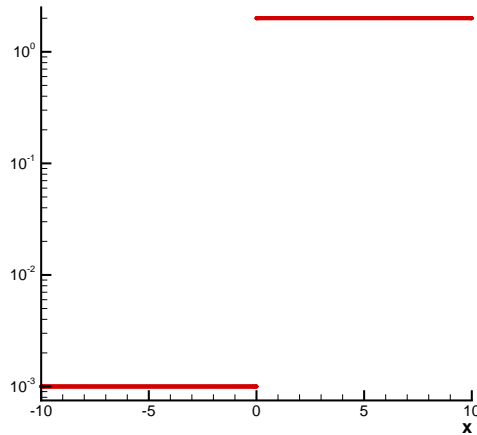
and  $\gamma = 1.4$ .

## LeBlanc Problem (Cont.)

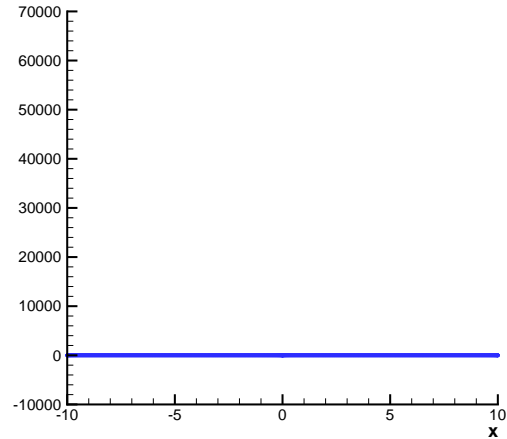
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The initial condition is the Riemann data :

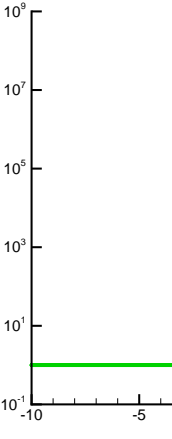
**Density**



**Velocity**



**Pressure**



$$\begin{aligned} \rho = 2, \quad u = 0, \quad P = 10^9 & \quad -10 \leq x \leq 0 \\ \rho = 10^{-3}, \quad u = 0, \quad P = 1 & \quad 0 < x \leq 10 \end{aligned}$$

and  $t_{final} = 10^{-4}$ .

## LeBlanc Problem (Cont.)

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### Algorithms :

- Spatial Algorithm :
  1. Multi-Domain Chebyshev collocation method (Spectral),
    - Differentiation and Smoothing operations are done via an optimized (Costa & Don);
    - a 9'th order exponential filter used to stabilize the scheme.
    - Riemann Solver applied at the domain interface.
    - The non-oscillatory solutions are obtained by post-processing technique (Gottlieb).
  2. WENO fifth order finite difference scheme (WENO) with Lax-Friedrichs flux limiter.
- Temporal Algorithm :

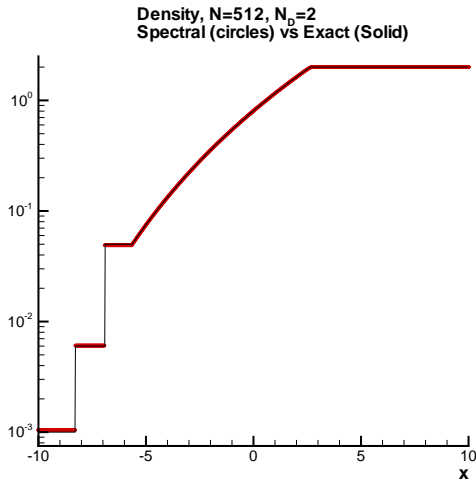
Third order TVD Runge Kutta method (Shu and Osher).



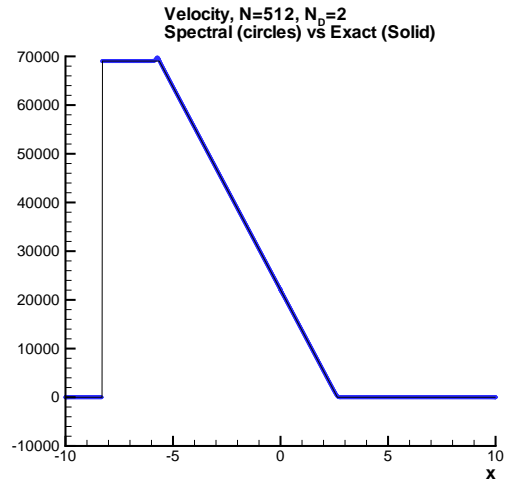
# LeBlanc Problem (Cont.)

## Spectral Scheme

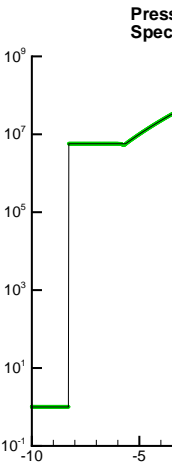
### Density



### Velocity



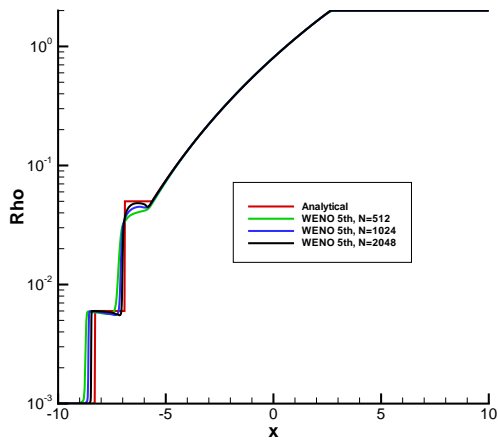
### Pressure



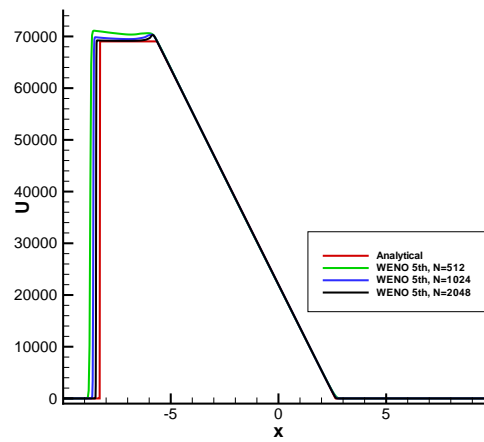
# LeBlanc Problem (Cont.)

## WENO Scheme

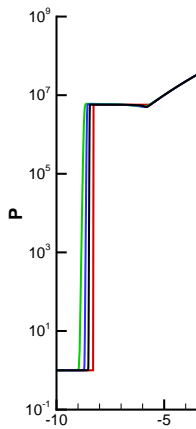
### Density



### Velocity



### Pressure



# Richtmyer-Meshkov Instability

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- Shock induced instability of the interface between fluids of different densities (first discovered by Richtmyer and experimented by Meshkov.)
- Growth of the interface amplitude and secondary shear instability promote mixing.
- Applications included but not limited to mixing enhancement and inertial

## Richtmyer-Meshkov Instability (Cont.)

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The two-dimensional Euler equations,

$$\mathbf{Q}_t + \mathbf{F}_x + \mathbf{G}_y = 0.$$

The state vector  $\mathbf{Q}$  is

$$\mathbf{Q} = (\rho, \rho u, \rho v, E)^\top.$$

The inviscid fluxes  $\mathbf{F}$  and  $\mathbf{G}$  are given by

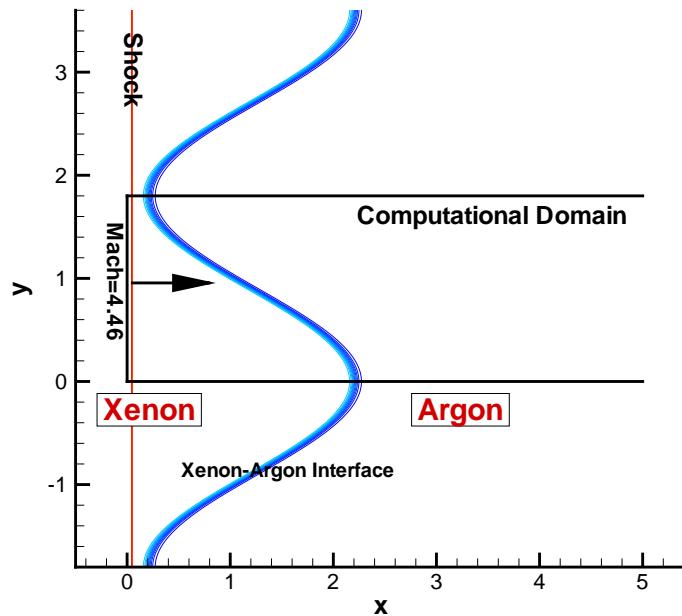
$$\begin{aligned}\mathbf{F} &= (\rho u, \rho u^2 + P, \rho uv, (E + P)u)^\top, \\ \mathbf{G} &= (\rho v, \rho uv, \rho v^2 + P, (E + P)v)^\top.\end{aligned}$$

where

$$P = (\gamma - 1)\left(E - \frac{1}{2}\rho\mathbf{U} \cdot \mathbf{U}\right).$$

## Richtmyer-Meshkov Instability (Cont.)

Initial Condition :



- Hugoniot-Rankine condition for shock
- Pre-Shock Temperature  $T = 2000 \text{ K}$
- Pre-Shock Pressure  $P = 0.5 \text{ atm}$
- Xenon and Argon density are  $\rho_{Xe} = 5.5 \times 10^{-3} \frac{g}{cm^3}$  and  $\rho_{Ar} = 0.89 \times 10^{-3} \frac{g}{cm^3}$  at half of the normal atmospheric pressure
- Specific heat ratio  $\gamma = \frac{5}{3}$
- Atwood number  $At = 5.4$
- Mach number  $M = 4.46$
- Wave Length  $\lambda = 3.6 \text{ cm}$
- Amplitude  $a = 1.0 \text{ cm}$

## Richtmyer-Meshkov Instability (Cont.)

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Xenon-Argon interface definition :

$$S(x, y) = \exp(-\alpha|d|^\beta) \quad 0 < d < 1$$

where

- 

$$d = \frac{(x_i + a \cos(2\pi y/\lambda) + \delta) - x}{2\delta}$$

- $\delta > 0$  is the interface thickness,
- $\beta = 8$  is the interface order,
- $\alpha = -\ln \epsilon$  with  $\epsilon$  being the machine zero.

## Richtmyer-Meshkov Instability (Cont.)

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### Algorithms :

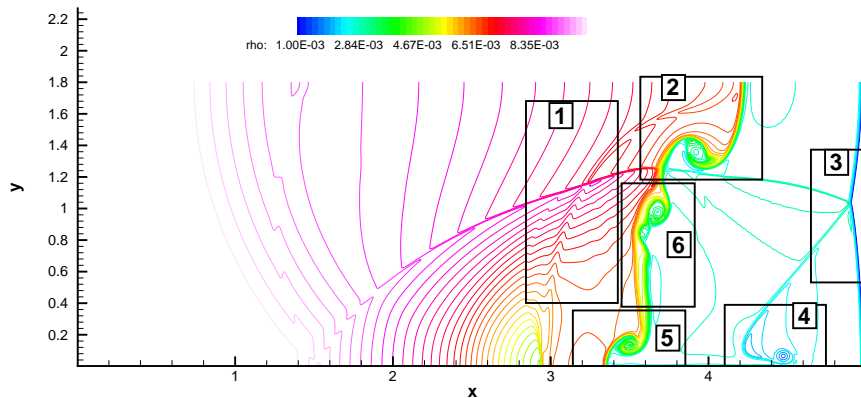
- Spatial Algorithm :
  1. Combined Chebyshev and Fourier collocation method (Spectral),
    - Differentiation and Smoothing operations are done via an optimized (Costa & Don);
    - a 10'th and 9'th order exponential filter used for the differentiation and respectively.
  2. WENO fifth order finite difference scheme (WENO) with Lax-Frederick
  3. Symmetry property in  $y$  is utilized to reduce the cost of computation.
- Temporal Algorithm :

Third order TVD Runge Kutta method (Shu and Osher).

## Richtmyer-Meshkov Instability (Cont.)

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### Regions of Interest :



1. Reflected shock generation and its effect on the mass fraction;
2. The penetration of the lighter (Ar) fluid forms the jet;
3. Triple point on the trailing edge of the jet;
4. A small jet and its vortex formation due to Kelvin-Helmholtz instability along the contact surface during long time simulation;
5. The penetration of the heavier (Xe) fluid forms the bubble;
6. Vortical rollups of the jet.



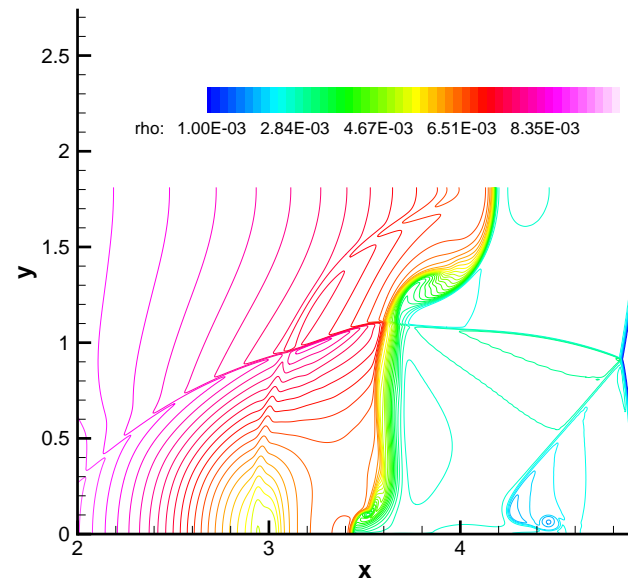
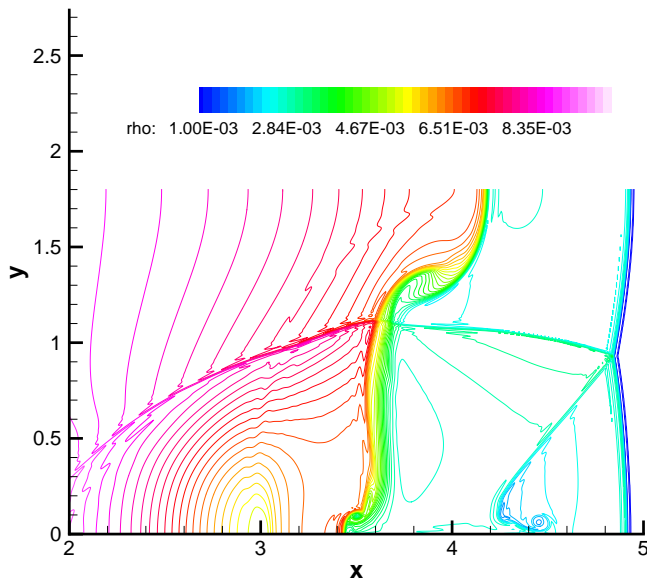
## Richtmyer-Meshkov Instability (Cont.)

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Convergence Study ( $M = 4.46, \delta = 0.6 \text{ cm}, t = 50 \mu\text{s}$ ) : Density

Spectral

WENO



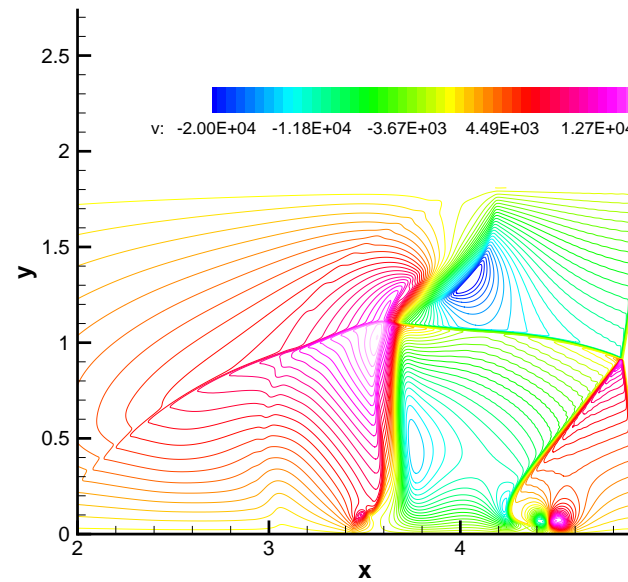
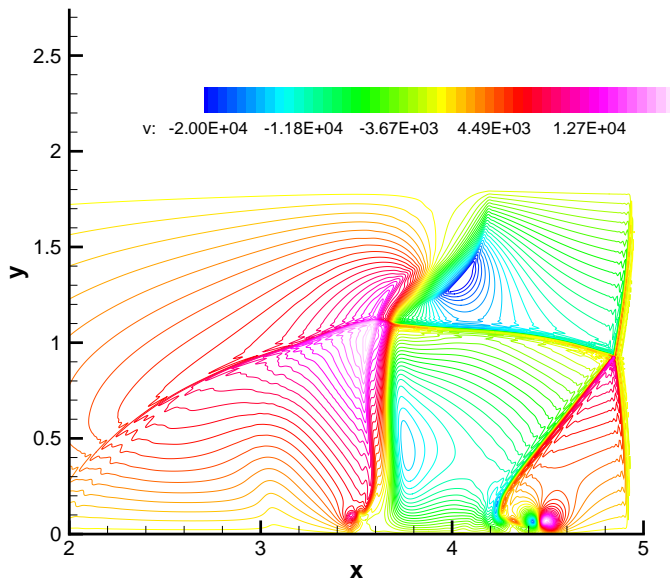
- Grid size for the Spectral and WENO schemes are  $1024 \times 512$ .

## Richtmyer-Meshkov Instability (Cont.)

Convergence Study ( $M = 4.46, \delta = 0.6 \text{ cm}, t = 50 \mu\text{s}$ ) : V-Velocity

Spectral

WENO

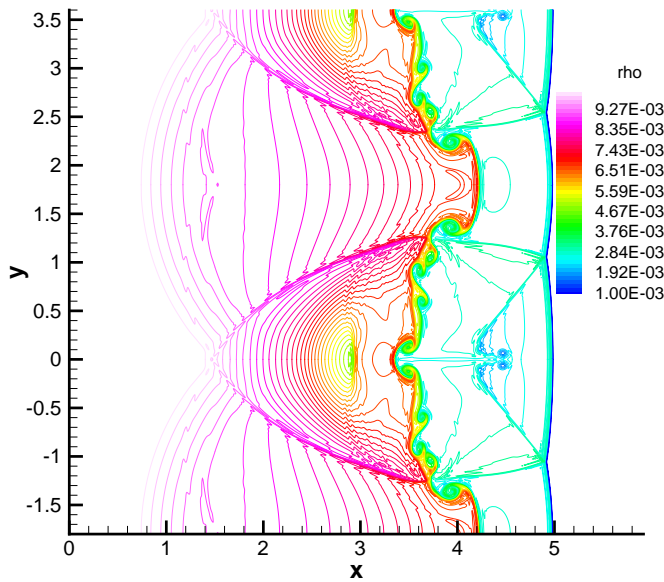


- Grid size for the Spectral and WENO schemes are  $1024 \times 512$ .

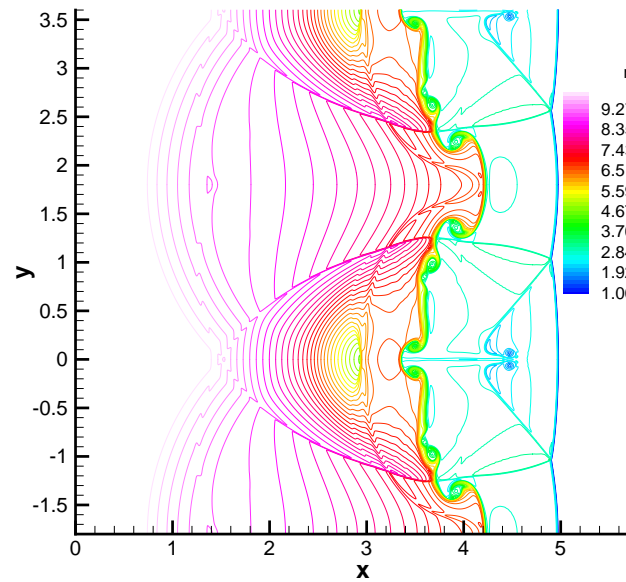
## Richtmyer-Meshkov Instability (Cont.)

Convergence Study ( $M = 4.46, \delta = 0.2 \text{ cm}, t = 50 \mu\text{s}$ ) : Density

Spectral



WENO

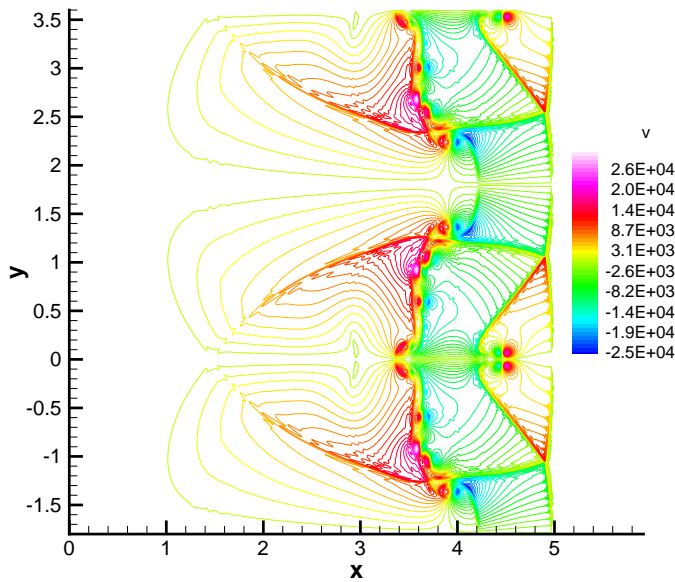


- Grid size for the Spectral and WENO schemes are 1024x256 and 1024x512

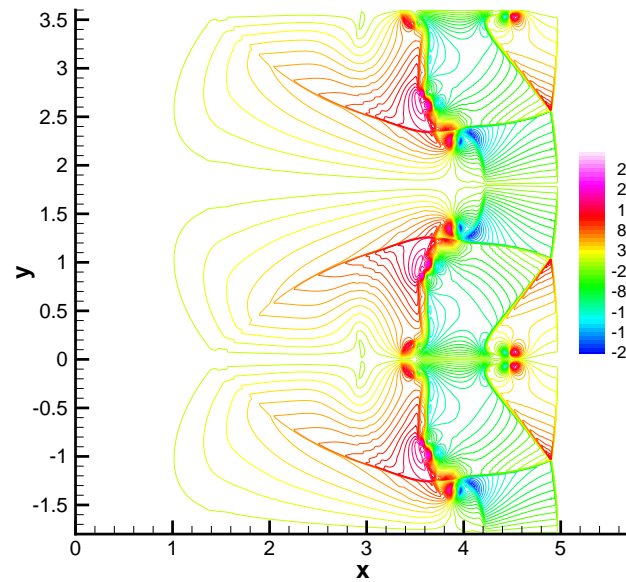
## Richtmyer-Meshkov Instability (Cont.)

Convergence Study ( $M = 4.46, \delta = 0.2 \text{ cm}, t = 50 \mu\text{s}$ ) : V-Velocity

Spectral



WENO

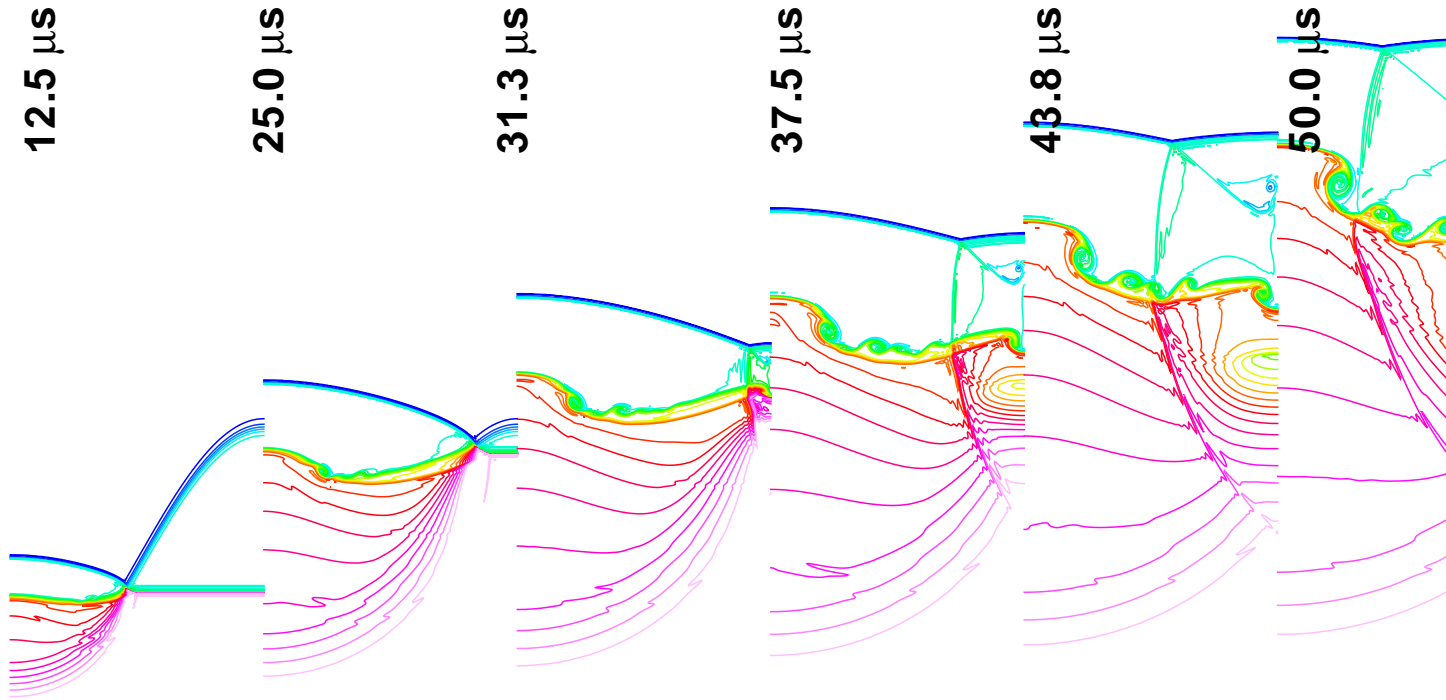


- Grid size for the Spectral and WENO schemes are  $1024 \times 256$  and  $1024 \times 512$

## Richtmyer-Meshkov Instability (Cont.)

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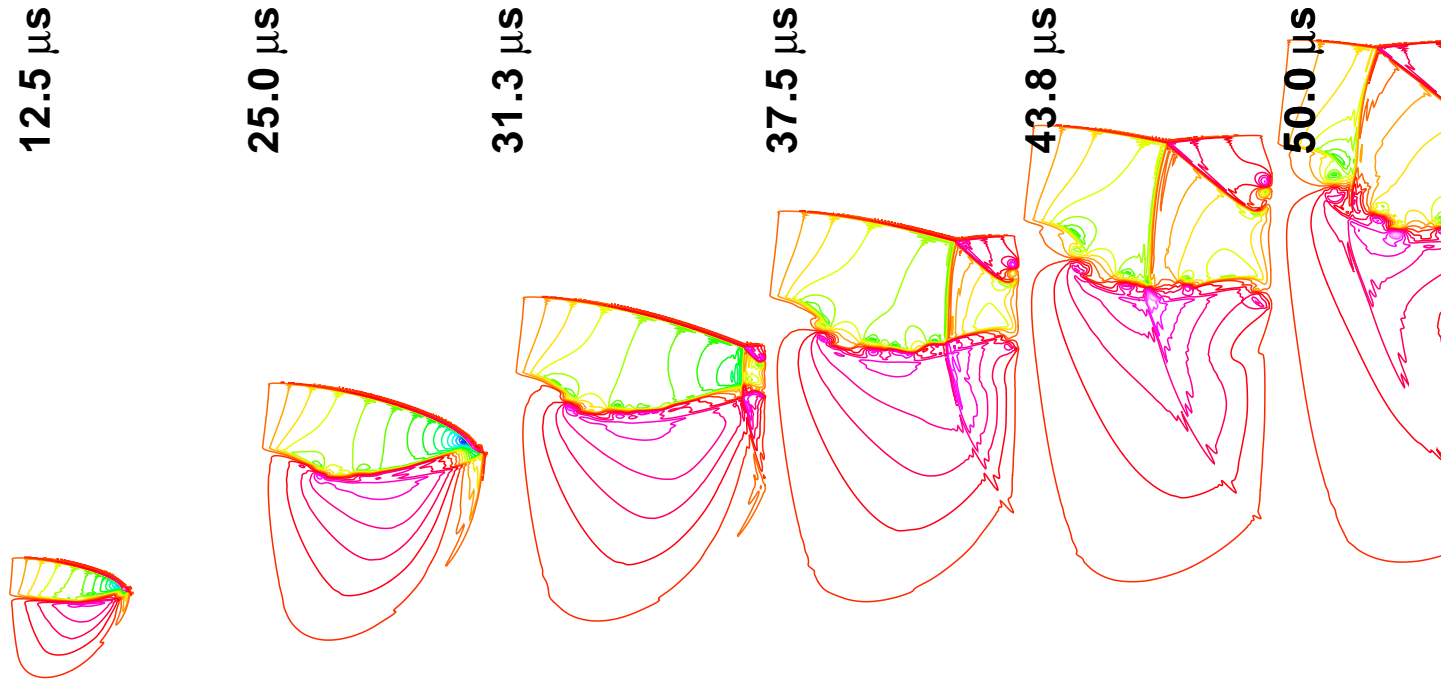
Snapshot of the Density ( $M = 4.46, \delta = 0.2 \text{ cm}$ ) : Spectral Scheme



## Richtmyer-Meshkov Instability (Cont.)

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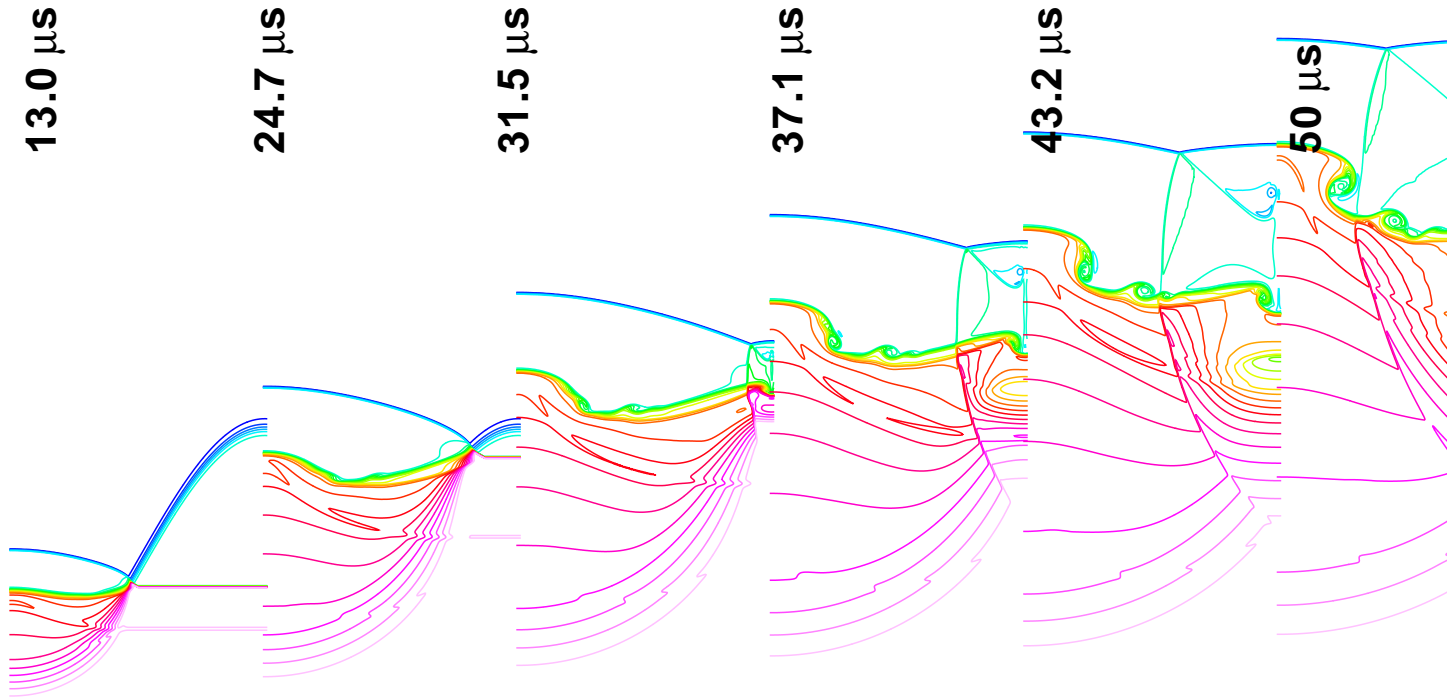
Snapshot of the V-Velocity ( $M = 4.46, \delta = 0.2 \text{ cm}$ ) : Spectral Scheme



## Richtmyer-Meshkov Instability (Cont.)

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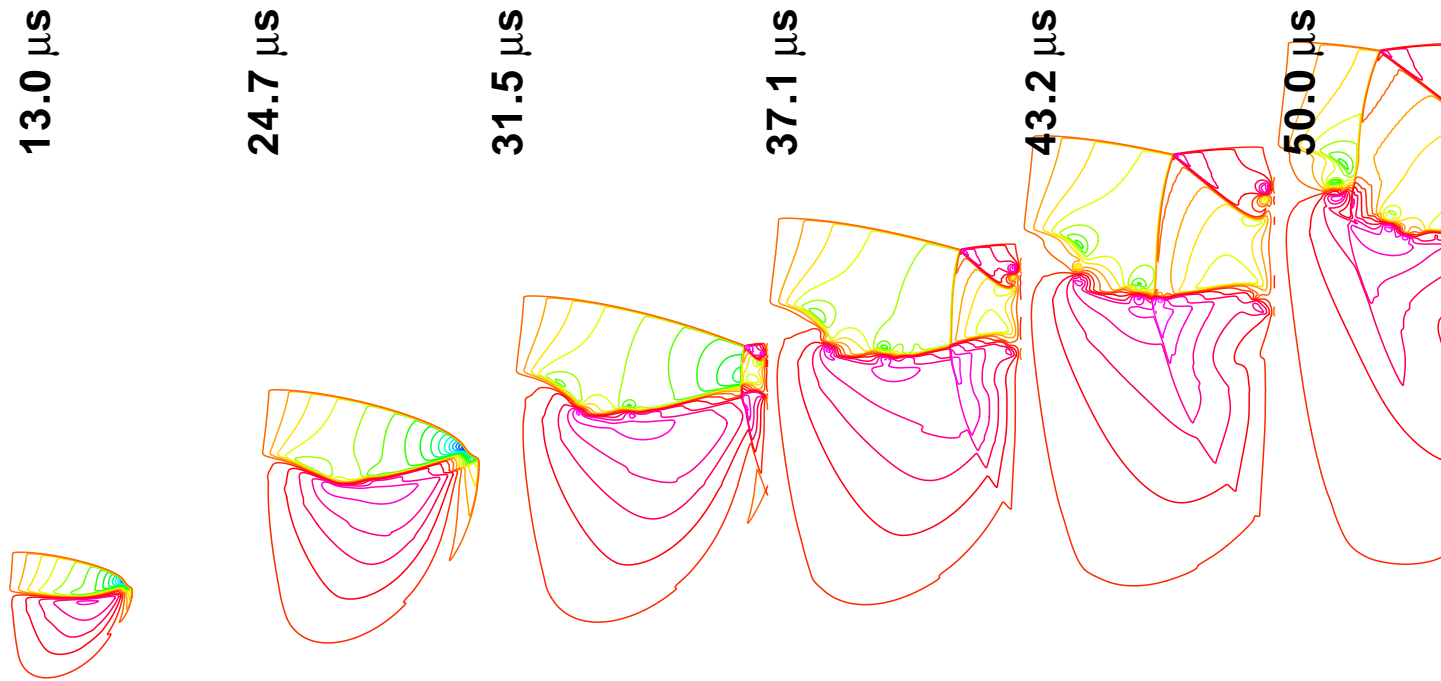
Snapshot of the Density ( $M = 4.46, \delta = 0.2 \text{ cm}$ ) : WENO Scheme



## Richtmyer-Meshkov Instability (Cont.)

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Snapshot of the V-Velocity ( $M = 4.46, \delta = 0.2 \text{ cm}$ ) : WENO Scheme





## Richtmyer-Meshkov Instability (Cont.)

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### Observations :

- Good agreement of the Global large and medium features between the Spectral and WENO scheme.
- Some discrepancy of the fine scale vortical structures along the gaseous interface in the two simulations of this sensitive nature to small perturbation (physically and/or numerically).

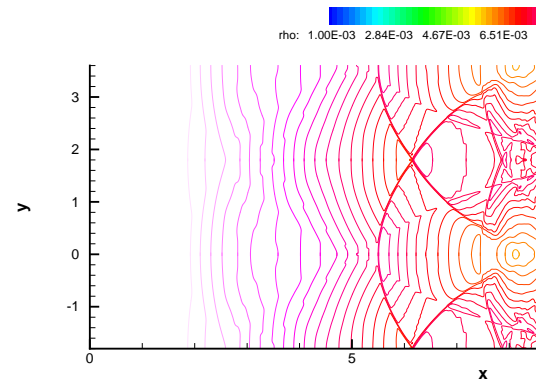
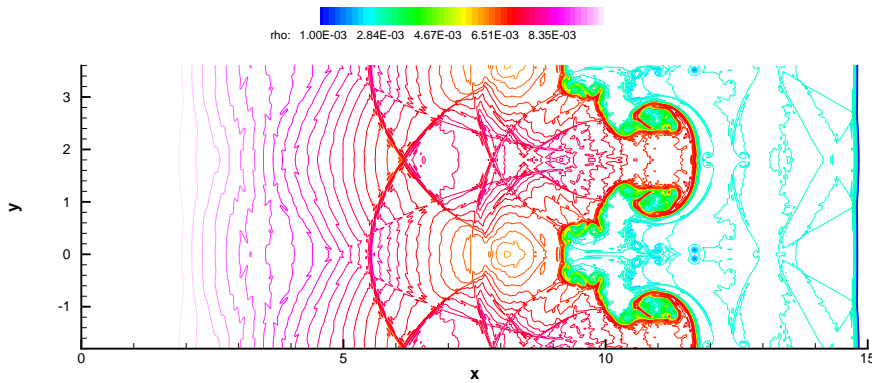
# Richtmyer-Meshkov Instability (Cont.)

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Long time Case ( $M = 4.46, \delta = 0.2 \text{ cm}, t = 124 \mu\text{s}$ ) : Density

Spectral

WE



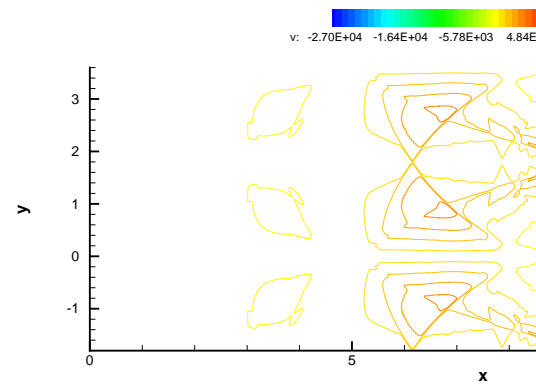
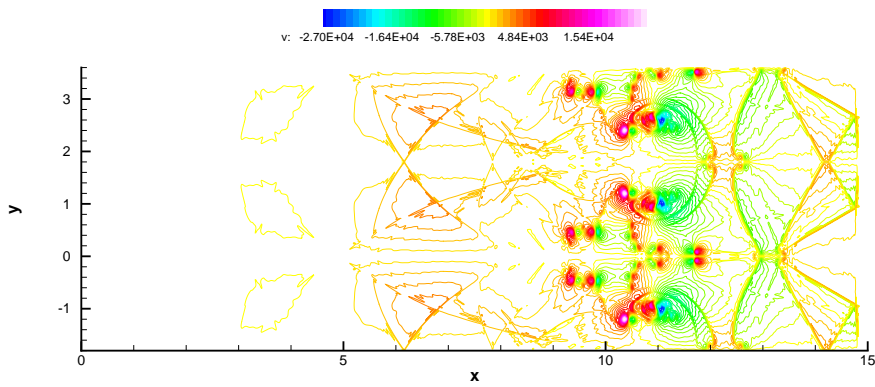
# Richtmyer-Meshkov Instability (Cont.)

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Long time Case ( $M = 4.46, \delta = 0.2 \text{ cm}, t = 124 \mu\text{s}$ ) : V-Velocity

Spectral

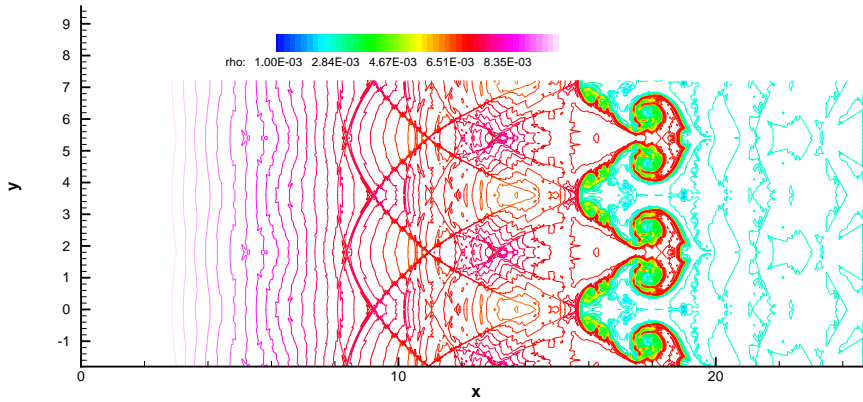
WE



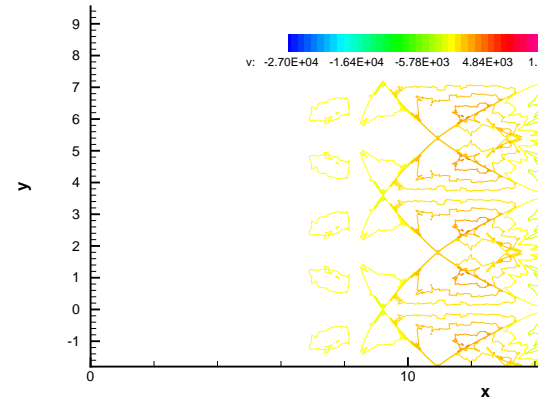
# Richtmyer-Meshkov Instability (Cont.)

Large Domain Case ( $M = 4.46, \delta = 0.2 \text{ cm}, t = 237 \mu\text{s}$ ) : Spectral sc

Density



V-V

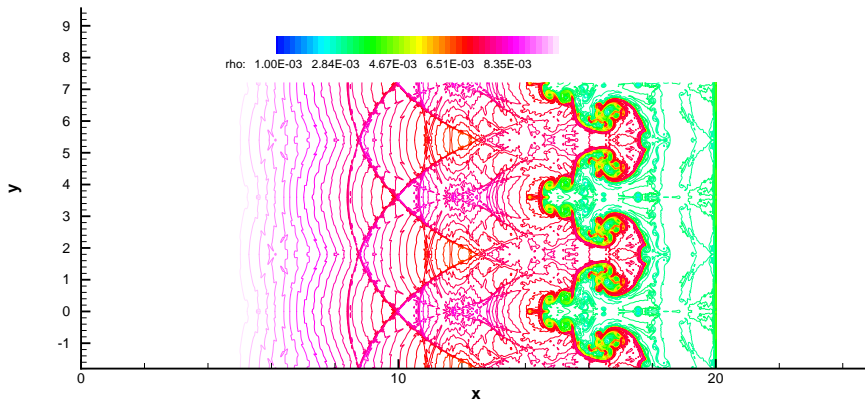


## Richtmyer-Meshkov Instability (Cont.)

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High Mach Number ( $M = 8, \delta = 0.2 \text{ cm}, t = 200 \text{ } \mu\text{s}$ ) : Spectral scheme

Density



V-V

