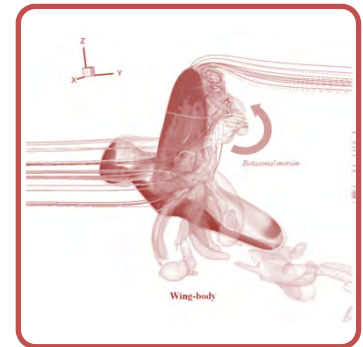
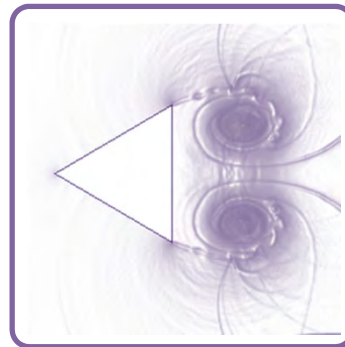
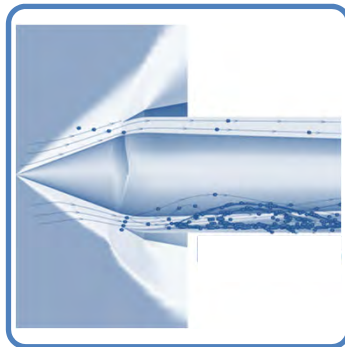


# Progress in Limiting Strategy: From One-dimensional Concepts to Multi-dimensional Limiting Process



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*11, November, 2013*

- **A Discipline to Solve Conservation Laws of Fluid Dynamics Numerically.**

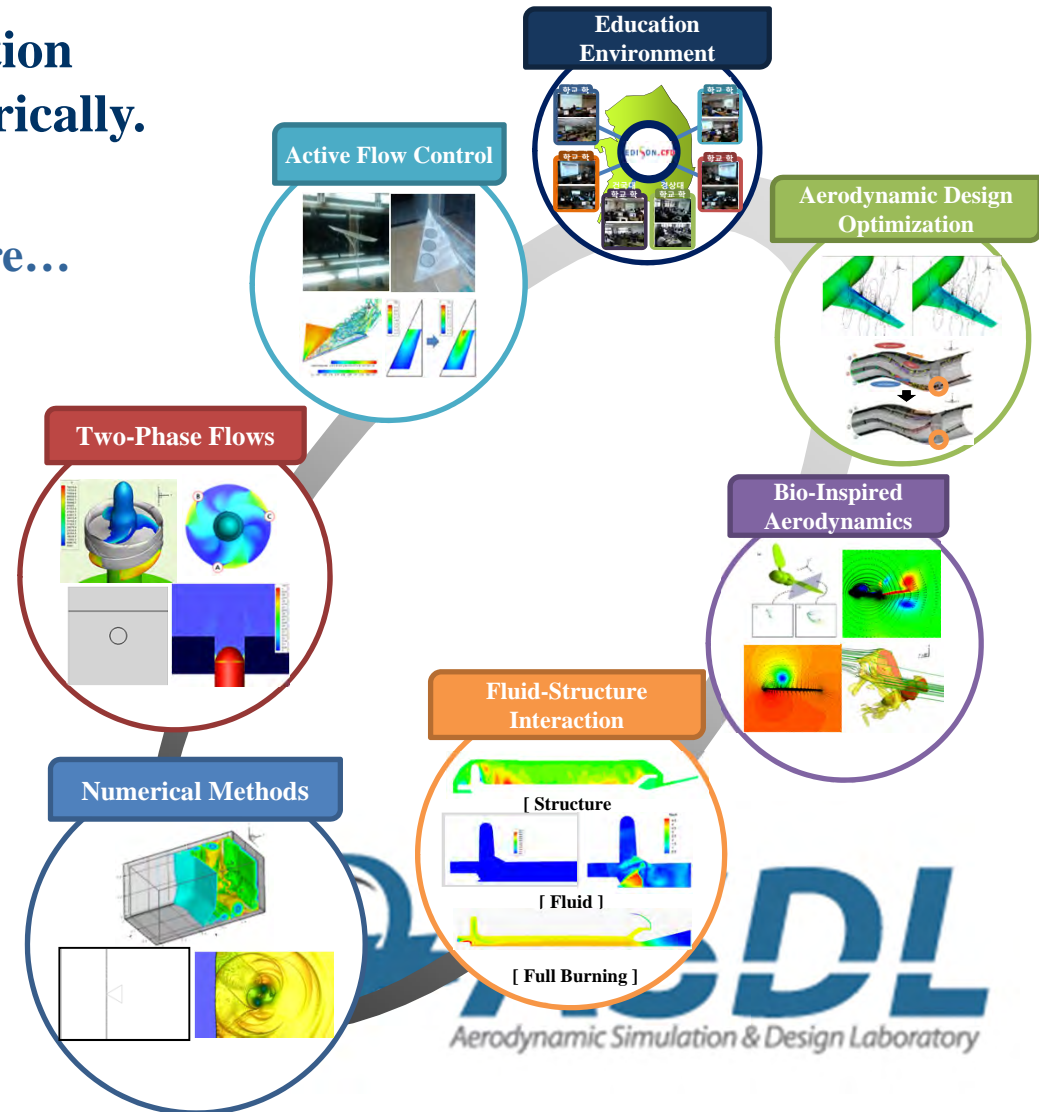
- **Flow information**

- pressure, velocity, temperature...
- force (lift, drag) & moment or heat flux distribution

- **Multidisciplinary design optimization and Flow control**
- performance improvement

- **Input data**

- Structural design
- Engine performance
- S&C, and so on



- Numerical Discretization
  - FDM, FEM, and FVM
- Finite Volume Discretization
  - Directly apply the integral form of the conservation laws to each computational cell in physical domain.
  - One-step conservative finite volume discretization for 1-D Euler eqns.

$$\mathbf{U}_t + \mathbf{F}(\mathbf{U})_x = \mathbf{0} \text{ with } \mathbf{U} = \begin{bmatrix} \rho \\ \rho u \\ \rho E \end{bmatrix}, \mathbf{F}(\mathbf{U}) = \begin{bmatrix} \rho u \\ \rho u^2 + p \\ \rho u H \end{bmatrix}, \begin{cases} E = e + \frac{u^2}{2} \\ H = h + \frac{u^2}{2} \end{cases}, p = (\gamma - 1)\rho \left[ E - \frac{u^2}{2} \right], \gamma = 1.4$$

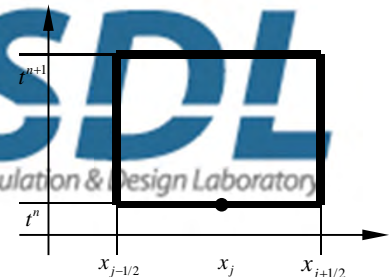
From  $\int_{t^n}^{t^{n+1}} \int_{x_{j-1/2}}^{x_{j+1/2}} (\mathbf{U}_t + \mathbf{F}(\mathbf{U})_x) dx dt = \mathbf{0}$ , we have

$$\int_{x_{j-1/2}}^{x_{j+1/2}} \mathbf{U}(x, t^{n+1}) dx - \int_{x_{j-1/2}}^{x_{j+1/2}} \mathbf{U}(x, t^n) dx = - \left( \int_{t^n}^{t^{n+1}} \mathbf{F}[\mathbf{U}(x_{j+1/2}, t)] dt - \int_{t^n}^{t^{n+1}} \mathbf{F}[\mathbf{U}(x_{j-1/2}, t)] dt \right)$$

Seeking a solution in an average sense over  $\Delta x = [x_{j-1/2}, x_{j+1/2}]$ ,  $\Delta t = [t^n, t^{n+1}]$

by introducing  $\mathbf{U}_j^n = \frac{1}{\Delta x} \int_{x_{j-1/2}}^{x_{j+1/2}} \mathbf{U}(x, t^n) dx$ ,  $\hat{\mathbf{F}}_{j+1/2}^n = \frac{1}{\Delta t} \int_{t^n}^{t^{n+1}} \mathbf{F}[\mathbf{U}(x_{j+1/2}, t)] dx$ ,

We have  $\mathbf{U}_j^{n+1} = \mathbf{U}_j^n - \frac{\Delta t}{\Delta x} (\hat{\mathbf{F}}_{j+1/2}^n - \hat{\mathbf{F}}_{j-1/2}^n)$  with  $\hat{\mathbf{F}}_{j+1/2}^n = \hat{\mathbf{F}}_{j+1/2}^n(\mathbf{U}_j^n, \mathbf{U}_{j+1}^n)$ .



- **Some conditions for an entropy-satisfying weak solution**

- **Consistency**

- $\hat{f}(u, u) = f(u)$

- Lipschitz continuity

$$|\hat{f}(u_{j-1}, u_j) - f(u)| \leq L \max(|u_{j-1} - u|, |u_j - u|)$$

- **Conservation**

- $\hat{f}_{j+1/2,L} = \hat{f}_{j+1/2,R}$

- **Monotonicity**

- Monotone scheme (Harten-Hyman-Lax, 1987)

$$u_j^{n+1} = u_j^n - \frac{\Delta t}{\Delta x} (\hat{f}_{j+1/2} - \hat{f}_{j-1/2}) = H_j(u_{j+k}, \dots, u_j, \dots, u_{j-k}),$$

$$\frac{\partial H_j}{\partial u_{j+l}} \geq 0, \quad \forall |l| \leq k.$$

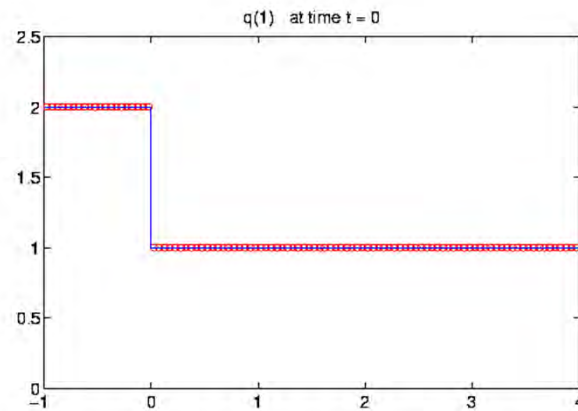
- Consider 3-point scheme

$$u_j^{n+1} = u_j^n - \frac{\Delta t}{\Delta x} (\hat{f}_{j+1/2} - \hat{f}_{j-1/2}) = H_j(u_{j-1}, u_j, u_{j+1}) \text{ with } \hat{f}_{j+1/2} = \hat{f}_{j+1/2}(u_j, u_{j+1})$$

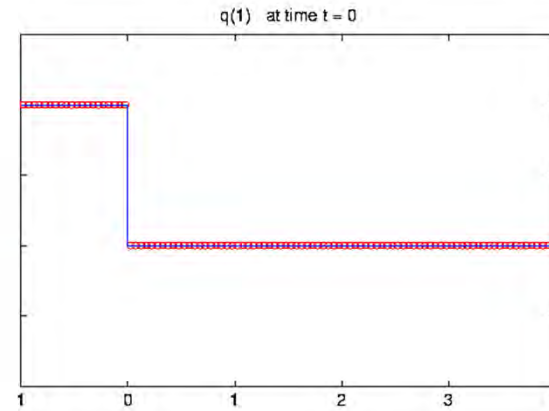
$$\frac{\partial H_j}{\partial u_{j+l}} \geq 0 \text{ gives } \hat{f}_{j+1/2}(u, v) \text{ is monotonically increasing/decreasing with respect to } u / v.$$

- **Computed results**

- Conservative and non-conservative scheme

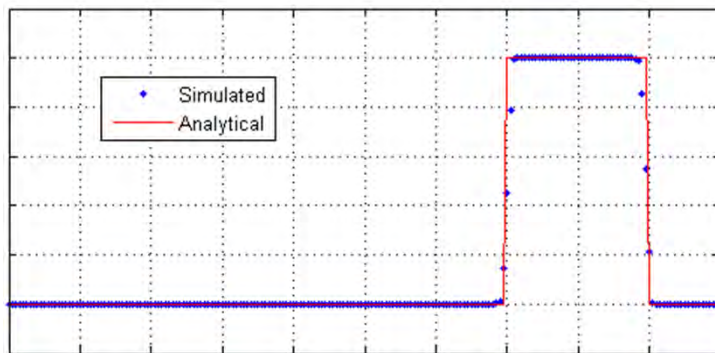


< *Conservative scheme* >

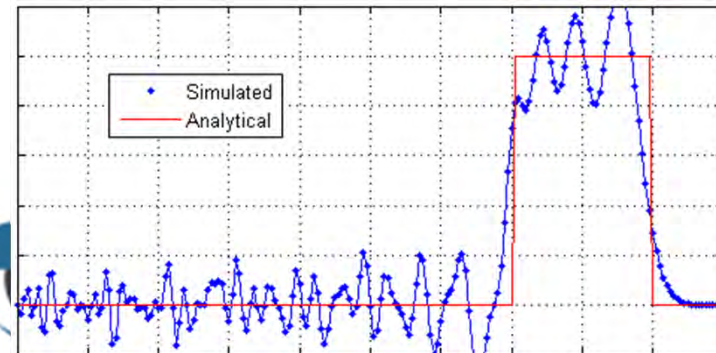


< *Non-conservative scheme* >

- Monotone and non-monotone scheme



< *Monotone scheme* >



< *Non-monotone scheme* >

## **Non-linear Stability and Hyperbolic PDE**

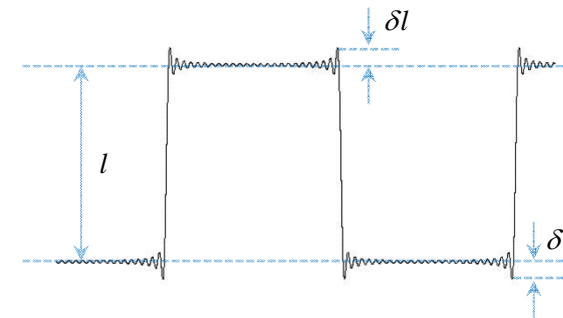
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- **Gibb's Phenomenon and Monotonicity**
- **Non-linear Stability**
- **Oscillation Control Strategies**



- **Gibb's Phenomenon (1899)**
  - **Approximation of a profile including discontinuity by Fourier Series (or interpolating techniques based on global basis functions)**
    - Numerical oscillations across discontinuity with  $O(1) \rightarrow$  It never dies out even if the number of basis function is increasing.

harmonics: 1



< Animation of the Gibbs phenomenon >

- **Magnitude of overshoot/undershoot**  $(\delta l / l) \sim \pm 14\%$
- **Locally converge (or  $L_1, L_2$  convergence) but not uniformly ( $L_\infty$  convergence)**
  - $\rightarrow$  warning to naive capturing of discontinuities by increasing the number of interpolating function or mesh point

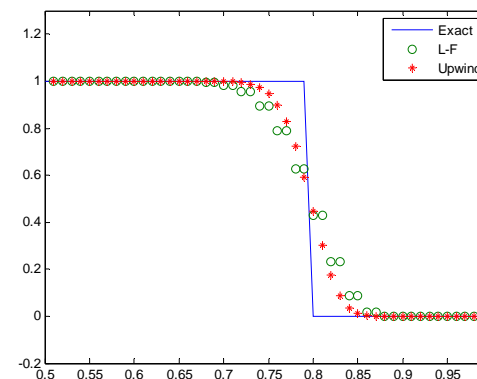
- **First-order Scheme and Numerical Diffusion**

- For  $u_t + au_x = 0$  with Upwind or L-F scheme

- Modified equation

$$u_t + au_x = \begin{cases} \frac{\Delta x^2}{2\Delta t}(1 - \sigma^2)u_{xx} & \text{for L-F} \\ \frac{a\Delta x}{2}(1 - \sigma^2)u_{xx} & \text{for upwind} \end{cases}$$

$$\Rightarrow u_t + au_x = c_1(\Delta x, \Delta t)u_{xx} \text{ with } \sigma = a \frac{\Delta t}{\Delta x}$$



- **Leading error term is numerically dissipative → a smooth transition without oscillations**

- Excessive numerical dissipation
  - Unacceptable loss of accuracy → Too many grid points
  - Viscous computation and resolution of boundary layer



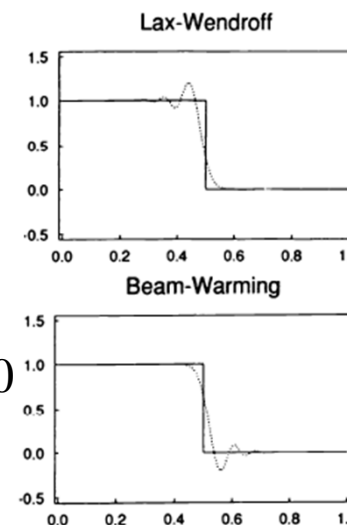
- **Second-order Scheme and Numerical Dispersion**

- **Modified equation of L-W and B-W scheme**

$$u_t + au_x = \begin{cases} -\frac{a\Delta x^2}{6}(1-\sigma^2)u_{xxx} & \text{for L-W} \\ \frac{a\Delta x^2}{6}(2-3\sigma+\sigma^2)u_{xxx} & \text{for B-W} \end{cases}$$

$$\Rightarrow u_t + au_x = c_1 u_{xxx} \xrightarrow[u=\hat{u}(\omega)e^{i\omega x}]{FT} \hat{u}(\omega, t)_t + i(a\omega + c_1\omega^3)\hat{u}(\omega, t) = 0$$

$$\Rightarrow \hat{u}(\omega, t) \sim e^{-i(a\omega + c_1\omega^3)t} \quad \text{vs.} \quad \hat{u}_{ex}(\omega, t) \sim e^{-ia\omega t}$$



- **Numerical dispersion relation:  $a(\omega) = a\omega + c_1\omega^3$**

**For each Fourier component with  $\omega$ , group velocity  $a_g(\omega) \equiv da(\omega)/d\omega$**

$a_g(\omega) = a + 3c_1\omega^2 \approx a$  for small wave number (long wave length)

$\neq a$  for large wave number (short wave length)

if  $a > 0, c_1 < 0$  for L-W  $\Rightarrow$  lagging error

$c_1 > 0$  for B-W  $\Rightarrow$  leading error

- **Numerical oscillations across discontinuity regardless of central differencing or upwinding once the order of accuracy is greater than one.**

- **Godunov Barrier Theorem on Monotonicity**

- (Godunov, 1959) *For the fully discretized 'linear' scheme of  $u_j^{n+1} = Lu_j^n \equiv \sum_q c_{jq} u_{j+q}^n$ ,*

*it can not be better than first-order accurate if the scheme is maximum-norm bounded.*

- (Positivity condition) If  $L$  is stable in maximum-norm,  $c_{jq}$  should be non-negative (or  $\frac{du_j^{n+1}}{du_{j+q}^n} \geq 0$ ).
- The monotonicity condition by HHL can be recovered.

- **Observation**

- **To obtain more than 2<sup>nd</sup>-order monotone scheme, the scheme should be non-linear even for linear equation.**

- $c_{jq}(\Delta x, \Delta t) \Rightarrow c_{jq}(\Delta x, \Delta t, u_{j+q}^n)$

- **Limiting strategy to realize higher-order accuracy and oscillation-free profile is essential.**

- **Non-linear Stability Criteria**

- **Stability using  $L_1$  norm**

- Total Variation Diminishing (TVD) and Total Variation Bounded (TVB)

- **Stability using  $L_\infty$  norm**

- Discrete maximum principle
- Local Extremum Diminishing (LED)

## ● Total Variation Stability

- A way to connect non-linear stability with the convergence of a computed solution

$$TV(u) \equiv \int_{-\infty}^{\infty} |u'(x)| dx \quad \text{or} \quad TV(u) \equiv \sum_{j=-\infty}^{\infty} |u_j - u_{j-1}| = 2(\sum \text{maxima} - \sum \text{minima})$$

- A useful tool to measure local oscillation
- A monotone scheme is TVD, and a TVD scheme is monotonicity-preserving.
  - 3-point TVD formulation of  $TV(u^{n+1}) \leq TV(u^n)$  (Harten, 1983)
- TVB allows oscillation only if it does not grow unboundedly.
  - $TV(u^{n+1}) \leq (1 + \alpha \Delta t) TV(u^n)$
  - TVB may avoid clipping at extrema but badly influences convergence.

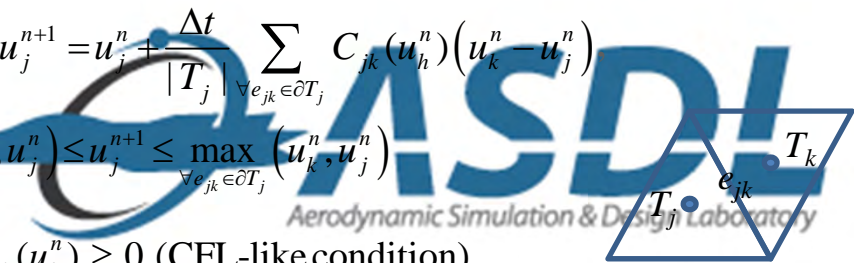
## ● Discrete Maximum Principle

- Edge-based local space-time discrete maximum principle

- (Barth, 2003) For a fully-discrete scheme of  $u_j^{n+1} = u_j^n + \frac{\Delta t}{|T_j|} \sum_{\forall e_{jk} \in \partial T_j} C_{jk}(u_h^n) (u_k^n - u_j^n)$

$$\min_{\forall e_{jk} \in \partial T_j} (u_k^n, u_j^n) \leq u_j^n \leq \max_{\forall e_{jk} \in \partial T_j} (u_k^n, u_j^n) \Rightarrow \min_{\forall e_{jk} \in \partial T_j} (u_k^n, u_j^n) \leq u_j^{n+1} \leq \max_{\forall e_{jk} \in \partial T_j} (u_k^n, u_j^n)$$

$$\text{If } C_{jk}(u_h^n) \geq 0 \text{ for } \forall e_{jk} \in \partial T_j, \text{ and } 1 - \frac{\Delta t}{|T_j|} \sum_{\forall e_{jk} \in \partial T_j} C_{jk}(u_h^n) \geq 0 \text{ (CFL-like condition)}$$



- **Remarkable Oscillation Control Strategies since 1970s**
  - **Flux Corrected Transport (FCT)**
    - Boris (1973), Zalesak (1984)
  - **MUSCL and Geometric Subcell Reconstruction**
    - Van Leer (1977, 1979)
  - **Total Variation Diminishing (TVD) / Total Variation Bounded (TVB)**
    - Harten (1983), Sweby (1984), Shu (1987)
  - **Essentially Non-Oscillatory (ENO) / Weighted ENO (WENO)**
    - Harten and Shu et al. (1987), Liu et al. (1994)
  - **Spekreijse's Monotonic Concept**
    - Spekreijse (1989)
  - **Multi-dimensional Reconstruction with Slope Limiter**
    - Barth (1989, 1990)
  - **Local Extremum Diminishing (LED)**
    - A. Jameson (1993)
  - **Adaptive Stencil Reconstruction (ENO/WENO)**
    - Abgrall (1993), Shu (1999)

## One-dimensional Limiting Strategies

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- Flux Corrected Transport
- TVD Scheme using Flux Limiter
- MUSCL and Slope Limiter
- ENO/WENO Schemes



- **Flux Corrected Transport (FCT) Method and Flux Limiter**

- The first algorithm that recognized the consequence of Godunov's theorem and presented a non-linear way of limiting the cell-interface flux

- For  $u_t + au_x = 0$  with  $u_j^{n+1} = u_j^n - \frac{\Delta t}{\Delta x} (F_{j+1/2} - F_{j-1/2})$

Design  $F_{j+1/2}$  s.t.  $F_{j+1/2} = \begin{cases} 2\text{nd-order in smooth region} \\ 1\text{st-order across local extrema} \end{cases}$

$\Rightarrow$  Let  $\begin{cases} F_{j+1/2}^H : \text{a 2nd-order 'non-monotonic' flux} \\ F_{j+1/2}^L : \text{a 1st-order 'monotonic' flux} \end{cases}$

$$F_{j+1/2} = F_{j+1/2}^L + \alpha_{j+1/2} (F_{j+1/2}^H - F_{j+1/2}^L) \text{ with } \alpha_{j+1/2} \begin{cases} \approx 1 \text{ for smooth region} \\ \approx 0 \text{ near local extrema} \end{cases}$$

- **Procedure**

- S1) Compute  $F_{j+1/2}^L, F_{j+1/2}^H$  from  $u_j^n$
- S2) Define 'anti-diffusive' flux as  $\tilde{F}_{j+1/2} = F_{j+1/2}^H - F_{j+1/2}^L = d_{j+1/2}^L - d_{j+1/2}^H = \varepsilon \Delta u_{j+1/2}^n$  ( $\varepsilon = \varepsilon^L - \varepsilon^H > 0$ )
- S3) Obtain the intermediate lower-order (or 1st-order) 'monotonic' solution

$$\bar{u}_j = u_j^n - \frac{\Delta t}{\Delta x} (F_{j+1/2}^L - F_{j-1/2}^L)$$

- S4) Correct  $\tilde{F}_{j+1/2}$  s.t. the final updated solution ( $u_j^{n+1}$ ) is free of extrema not found in  $\bar{u}_j^n$  or  $u_j^n$

$$F_{j+1/2}^c \equiv \alpha_{j+1/2} \tilde{F}_{j+1/2} = \alpha_{j+1/2} (d_{j+1/2}^L - d_{j+1/2}^H) \text{ with } 0 \leq \alpha_{j+1/2} \leq 1$$

- S5) Update the final solution with the corrected flux  $F_{j+1/2}^c$

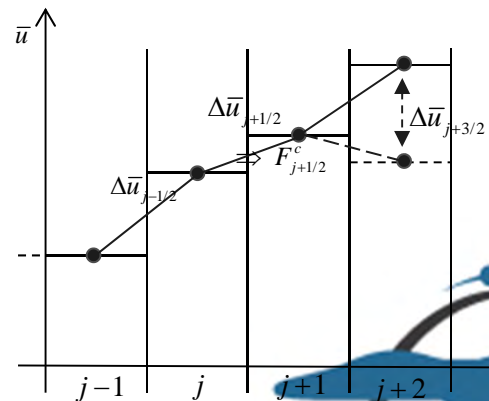
$$u_j^{n+1} = \bar{u}_j - \frac{\Delta t}{\Delta x} (F_{j+1/2}^c - F_{j-1/2}^c) = u_j^n - \frac{\Delta t}{\Delta x} \left\{ \left[ F_{j+1/2}^L + \alpha_{j+1/2} (F_{j+1/2}^H - F_{j+1/2}^L) \right]_{j+1/2} - [\dots]_{j-1/2} \right\}$$

- S6) The corrected flux  $F_{j+1/2}^c$  is designed as

$$F_{j+1/2}^c = \min \text{ mod} \left( \frac{\Delta x}{\Delta t} \Delta \bar{u}_{j-1/2}, \tilde{F}_{j+1/2}, \frac{\Delta x}{\Delta t} \Delta \bar{u}_{j+3/2} \right)$$

- The anti-diffusive flux is controlled such that it does not create new local extrema.
- Updated soln. satisfies the monotonic constraint in terms of the intermediate distribution

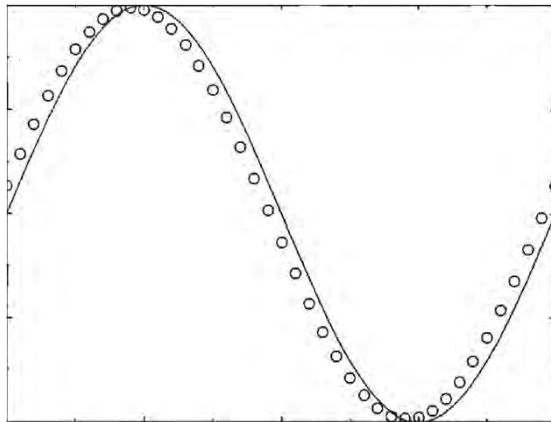
$$\min(\bar{u}_{j-1}, \bar{u}_j, \bar{u}_{j+1}) \leq u_j^{n+1} \leq \max(\bar{u}_{j-1}, \bar{u}_j, \bar{u}_{j+1})$$



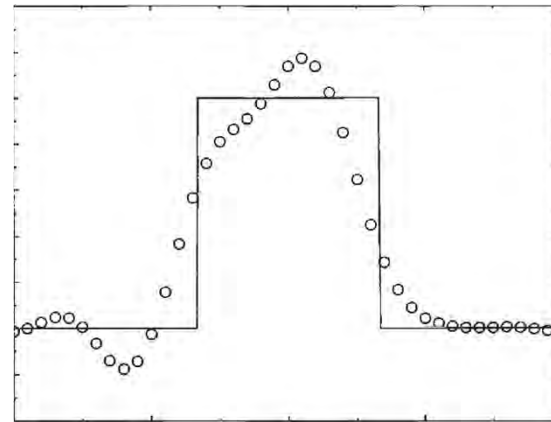
< Determination of the corrected flux >

- **Example**

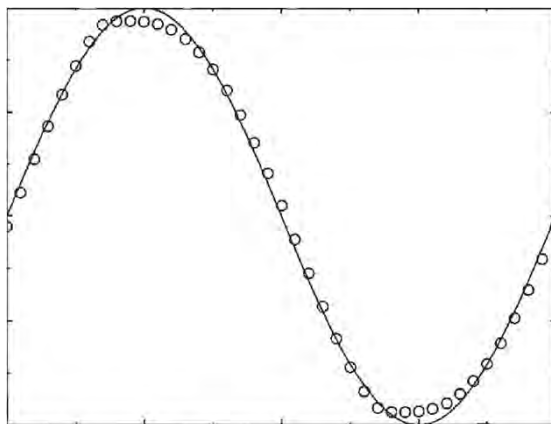
- Linear advection problem with smooth and discontinuous profiles



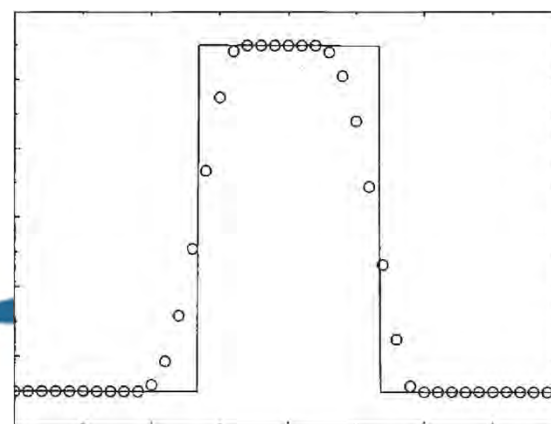
Lax-Wendroff method for test case 1



Lax-Wendroff method for test case 2



FCT method for test case 1



FCT method for test case 2



- **TVD Schemes and One-step Flux Limiting Function**
- **A Class of one-step scheme using refined form of flux limiters and satisfying TVD property**

- The limited flux form is assumed as  $F_{j+1/2} = F_{j+1/2}^L + \phi_j (F_{j+1/2}^H - F_{j+1/2}^L)$ .

$\phi_j$  is a limiter function monitoring the behavior of local solution  $u_j^n$ .

- For  $u_t + au_x = 0$  with  $a > 0$ , consider  $F_{j+1/2}^H$  as the L-W flux, and  $F_{j+1/2}^L$  as the upwind flux

$$\begin{aligned} u_j^{n+1} &= u_j^n - \frac{\sigma}{2}(u_{j+1}^n - u_{j-1}^n) + \frac{\sigma^2}{2}(u_{j+1}^n - 2u_j^n + u_{j-1}^n) \\ &= u_j^n - \sigma(u_j^n - u_{j-1}^n) - \frac{\sigma(1-\sigma)}{2}(u_{j+1}^n - 2u_j^n + u_{j-1}^n) \\ &= u_j^n - \frac{\Delta t}{\Delta x}(F_{j+1/2} - F_{j-1/2}) \end{aligned}$$

$$\Rightarrow F_{j+1/2} = \underbrace{au_j^n}_{\text{upwind}} + \underbrace{\frac{a(1-\sigma)}{2}\Delta u_{j+1/2}^n}_{\text{Lax-Wendroff correction}}$$

Thus, the limited flux form is  $F_{j+1/2} = au_j + \frac{a(1-\sigma)}{2}\Delta u_{j+1/2}\phi_j$  with  $\phi_j \geq 0$ .

- Define  $r_j = \Delta u_{j-1/2} / \Delta u_{j+1/2}$  to measure the change of local slope (or total variation) and design  $\phi_j = \phi(r_j)$  with the TVD principle

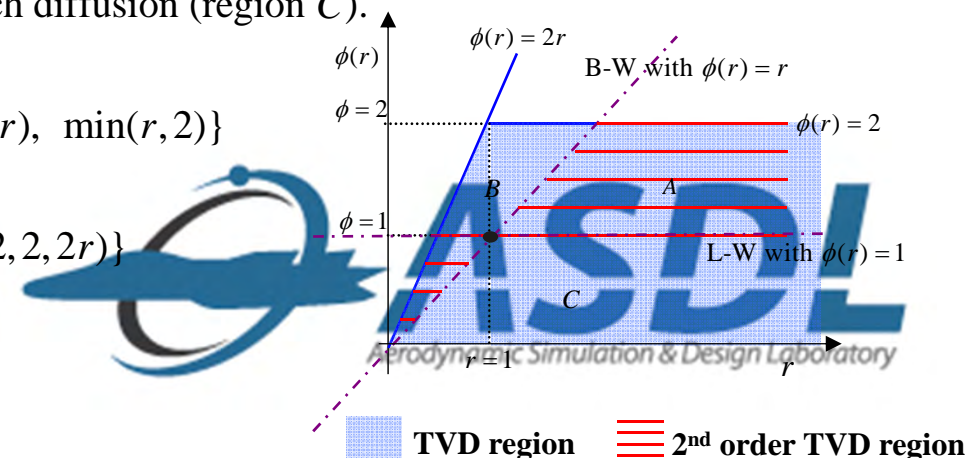
## ● Three-pint TVD condition

- The three-point scheme of the form  $u_j^{n+1} = u_j^n - C_{j-1/2} \Delta u_{j-1/2}^n + D_{j+1/2} \Delta u_{j+1/2}^n$  is TVD if  $C_{j-1/2}, D_{j+1/2} \geq 0$ , and  $C_{j+1/2} + D_{j+1/2} \leq 1$  for  $\forall j$ .
- The flux limited from of L-W scheme satisfies the TVD condition is satisfied if

we have  $\phi(r)$  s.t. 
$$\begin{cases} 0 \leq \frac{\phi(r)}{r} \leq 2 \text{ and } 0 \leq \phi(r) \leq 2, \text{ if } r \geq 0 \\ \phi(r) = 0, \text{ if } r < 0 \text{ (to prevent the accentuation of local extrema)} \end{cases}$$

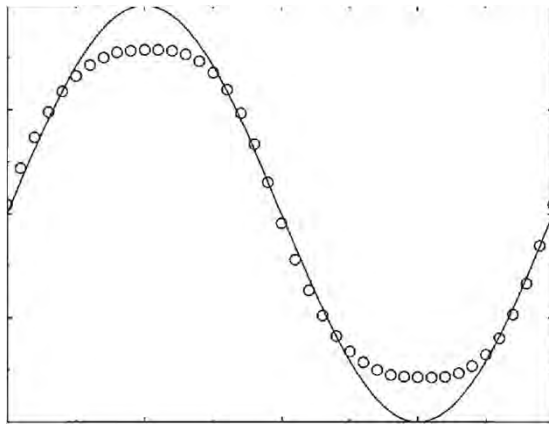
## ● TVD region and limiters

- $\phi(r)_{r=1} = 1$  to get a smooth transition with second-order accuracy
- Convex combination of L-W and B-W is desirable to avoid too much compression (region B) or too much diffusion (region C).
- Some TVD limiters
  - superbee limiter:  $\phi(r) = \max\{0, \min(1, 2r), \min(r, 2)\}$
  - van Leer limiter:  $\phi(r) = (|r|+r)/(1+|r|)$
  - MC limiter:  $\phi(r) = \max\{0, \min((1+r)/2, 2, 2r)\}$
  - minmod limiter:  $\phi(r) = \min(1, r)$

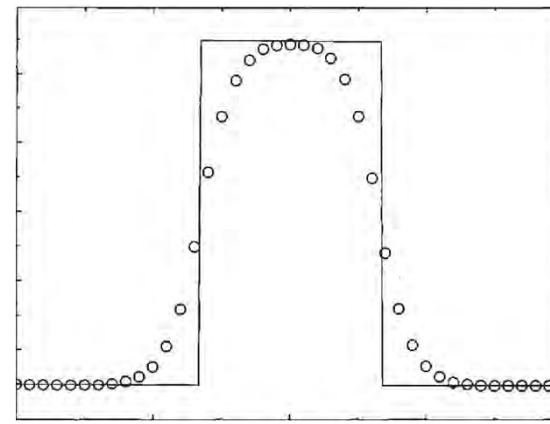


- **Example**

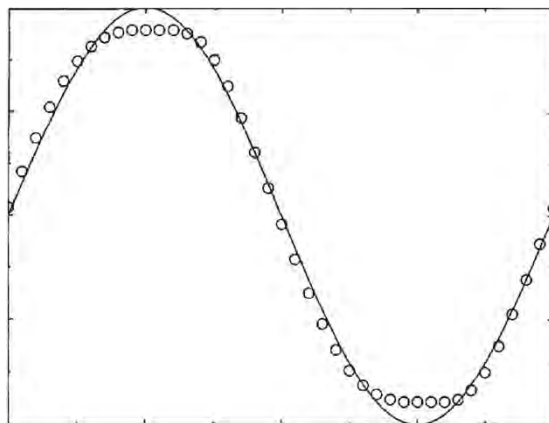
- Linear advection problem with smooth and discontinuous profiles



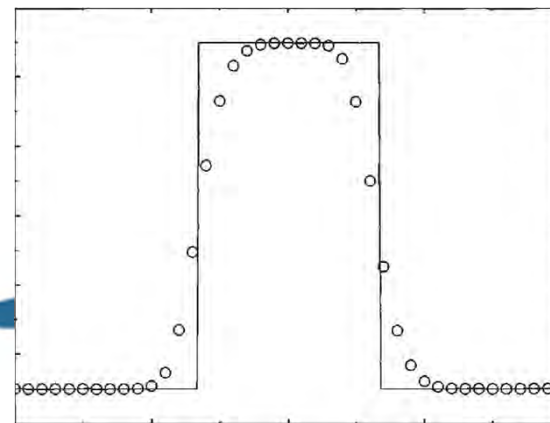
Flux-limited method with minmod limiter



Flux-limited method with minmod limiter



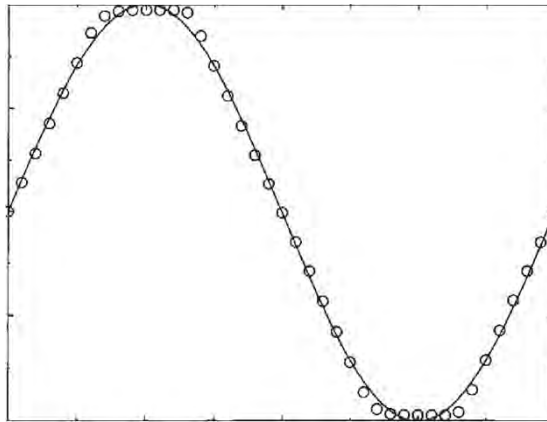
Flux-limited method with van Leer limiter



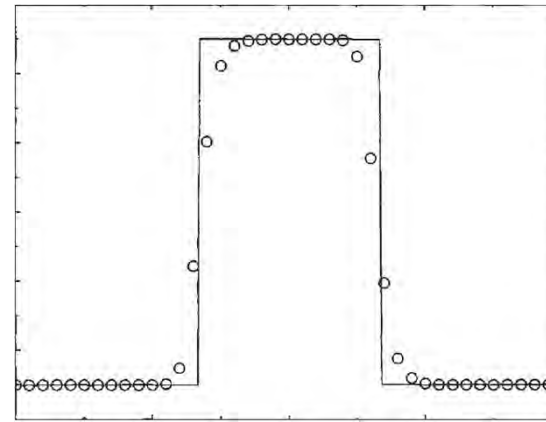
Flux-limited method with van Leer limiter

- **Example**

- Linear advection problem with smooth and discontinuous profiles



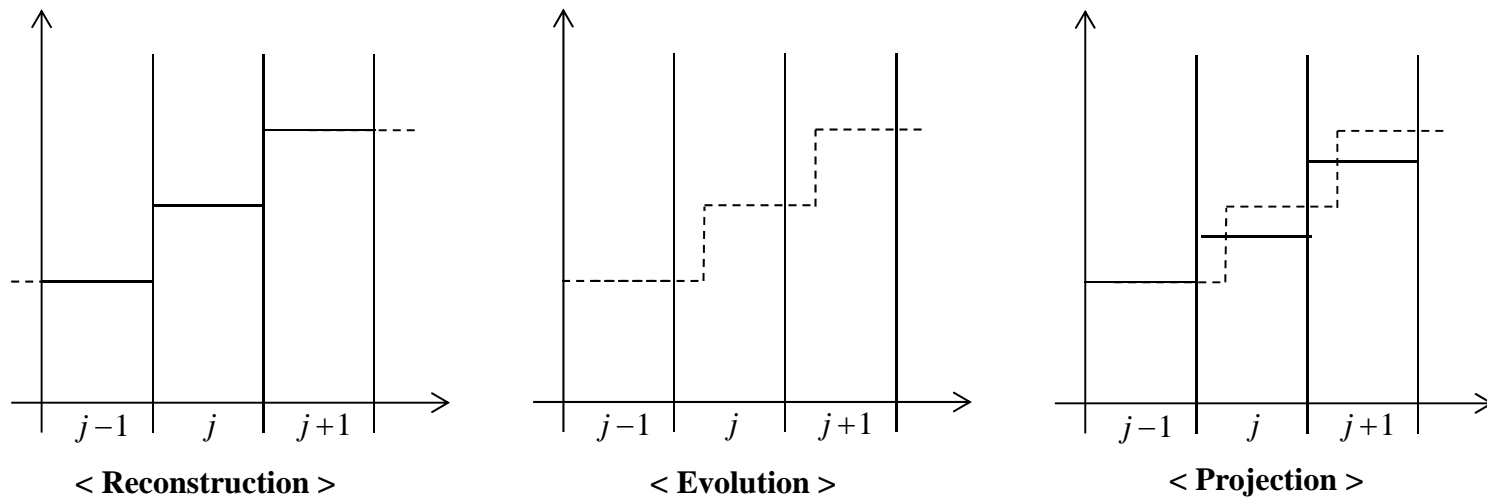
Flux-limited method with superbee limiter



Flux-limited method with superbee limiter

- Accuracy preservation in both smooth and discontinuous region

- **Geometric Approach for Monotonic Schemes**
  - Numerical scheme is analyzed in terms of reconstruction, evolution and projection stage.
    - Ex) Flow physics of first-order upwind scheme for  $u_t + au_x = 0, a > 0$



- Reconstruction stage: approximation of exact initial distribution from cell-averaged values
- Evolution stage: reconstructed initial profile is convected  $\Delta s_x = a \cdot \Delta t$  by  $u_t + au_x = 0$ .
- Projection stage: projection of the convected profile to update a new cell-averaged solution at  $t^{n+1} = t^n + \Delta t$

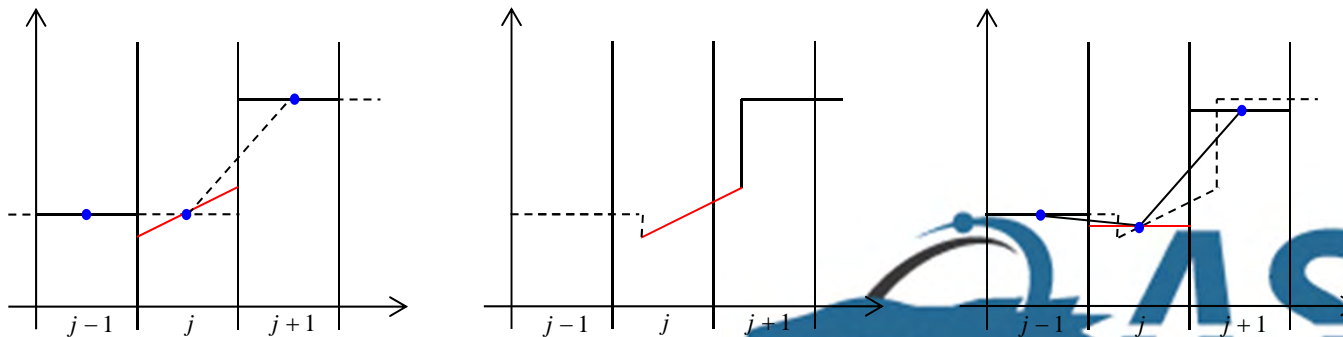
- Evolution and projection stages are not handled numerically.
- A higher-order monotonic scheme via a higher-order reconstruction of the initial data with the monotonic constraint
- Sub-cell distribution for second-order schemes

$$u(x) = u_j + \left(\frac{\partial u}{\partial x}\right)_j (x - x_j) + \left(\frac{\partial^2 u}{\partial x^2}\right)_j \frac{(x - x_j)^2}{2} + O(\Delta x^3)$$

$$\cong u_j + \frac{\Delta u_j}{\Delta x} (x - x_j), \quad x_{j-1/2} \leq x_j \leq x_{j+1/2} \quad \text{with } u_j = \int_{x_{j-1/2}}^{x_{j+1/2}} u(x) dx$$

- Estimation of the local slope  $\Delta u_j$

## ● Second-order upwind scheme with monotonicity



- Non-monotonic estimation of the local slope ( $\Delta u_j$ ) creates a new (artificial) local maxima/minima.

- If  $TV(u(x)) \leq TV(u^n)$  in reconstruction stage, the whole stages are TVD, since evolution and projection stages are intrinsically total variation not increasing.
- Slope limiters to satisfy the monotonic subcell distribution
  - $\Delta u_j|_{\min} = \min(\Delta u_{j+1/2}, \Delta u_{j-1/2})$     •  $\Delta u_j|_{sb} = \max(\Delta u_{i1}, \Delta u_{i2})$   
 with  $\Delta u_{i1} = \min(\Delta u_{j+1/2}, 2\Delta u_{j-1/2})$ ,  $\Delta u_{i2} = \min(\Delta u_{j-1/2}, 2\Delta u_{j+1/2})$
  - $\Delta u_j|_{MC} = \min\left(\frac{\Delta u_{j+1/2} + \Delta u_{j-1/2}}{2}, 2\Delta u_{j+1/2}, 2\Delta u_{j-1/2}\right)$
- **MUSCL schemes with slope limiters**
  - First-order upwinding:  $\hat{f}_{j+1/2} = \frac{1}{2}(f_j + f_{j+1}) - \frac{|a_{j+1/2}|}{2} \Delta u_{j+1/2}$
  - Sub-cell linear distribution using slope limiters:  $u(x) = u_j + \frac{\Delta u_j}{\Delta x}(x - x_j)$ ,  $x_{j-1/2} \leq x \leq x_{j+1/2}$   
 $\rightarrow u_{jR} = u(x_{j+1/2}) = u_j + \frac{\Delta u_j}{2}$  and  $u_{jL} = u(x_{j-1/2}) = u_j - \frac{\Delta u_j}{2}$  to replace  $u_j$  and  $u_{j+1}$
  - Interface flux using the interpolated values:  $\hat{f}_{j+1/2} = \frac{1}{2}(f_L + f_R) - \frac{|a_{j+1/2, L/R}|}{2} \Delta u_{j+1/2, L/R}$

- **Essentially Non-Oscillatory Higher-order Interpolation Procedure**

- **TVD vs. ENO**

- (TVD scheme) Locally first-order accurate across all extrema to strictly enforce monotonicity → accuracy loss across smooth extrema with excessive diffusion or clipping
- (TVD scheme) A fixed stencil (3-point TVD scheme with  $u_j, u_{j\pm 1}$ ) → difficult to obtain a higher-order TVD scheme
- (ENO scheme) Allow the increase of local extrema up to the order of truncation error to achieve higher-order accuracy across smooth extrema
- (ENO scheme) Locally adaptive smooth stencil for higher-order interpolation

- **Procedure to construct ENO stencil (or locally adaptive stencil)**

- Cell-averaged value or flux function

- **ENO polynomial reconstruction using cell-averaged value,  $u_i$**

- Step 1. Introduce Newton's divided difference based on cell-averaged values

$$u[x_i] \equiv u_i, u[x_i, x_{i+1}] \equiv \{u[x_{i+1}] - u[x_i]\} / \Delta x = (u_{i+1} - u_i) / \Delta x, \dots,$$

$$u[x_i, x_{i+1}, \dots, x_{i+k}] \equiv \{u[x_{i+1}, x_{i+2}, \dots, x_{i+k}] - u[x_i, x_{i+1}, \dots, x_{i+k-1}]\} / (k\Delta x)$$

If flow field is sufficiently smooth,  $u[x_i, x_{i+1}, \dots, x_{i+k}] = \frac{1}{k!} \frac{d^k u}{dx^k} \Big|_{\xi}, x_i \leq \xi \leq x_{i+k}$



- Step 1. Introduce Newton's divided difference based on cell-averaged values

And, if  $u(x)$  has a jump discontinuity of the  $l$ -th derivative at  $x = x_p$ ,

$$u[x_i, x_{i+1}, \dots, x_{i+k}] = O\left(\frac{1}{\Delta x^{k-l}} \left( \frac{d^l u(x_p^+)}{dx^l} - \frac{d^l u(x_p^-)}{dx^l} \right)\right), \quad x_i \leq \xi \leq x_{i+k}$$

- Step 2. Compare each divided difference to determine ENO stencil successively

Starting from the cell  $i$ , take  $\begin{cases} i+1 & \text{if } |u[x_i, x_{i+1}]| \leq |u[x_{i-1}, x_i]| \\ i-1 & \text{if } |u[x_i, x_{i+1}]| > |u[x_{i-1}, x_i]| \end{cases}$

$\Rightarrow u(x) = u_i + u[x_j, x_{j+1}](x - x_i), \quad x_{j-1/2} \leq x \leq x_{j+1/2}$  (linear reconstruction)

with  $j = \begin{cases} i & \text{if } |u[x_i, x_{i+1}]| \leq |u[x_{i-1}, x_i]| \\ i-1 & \text{if } |u[x_i, x_{i+1}]| > |u[x_{i-1}, x_i]| \end{cases}$

- Step 3. ENO stencil in a recursive manner

For  $m = 0, 1, 2, \dots, n-1$  with  $l_0(i) = i$ ,

$$l_{m+1}(i) = \begin{cases} l_m(i) & \text{if } |u[x_{l_m(i)}, x_{l_m(i)+1}, \dots, x_{l_m(i)+m+1}]| \leq |u[x_{l_m(i)-1}, x_{l_m(i)}, \dots, x_{l_m(i)+m}]| \\ l_m(i)-1 & \text{if } |u[x_{l_m(i)}, x_{l_m(i)+1}, \dots, x_{l_m(i)+m+1}]| > |u[x_{l_m(i)-1}, x_{l_m(i)}, \dots, x_{l_m(i)+m}]| \end{cases}$$

- Step 3. ENO stencil in a recursive manner

From  $[x_{l_m(i)}, x_{l_m(i)+1}, \dots, x_{l_m(i)+m}]$  with  $0 \leq m \leq n$ , construct a  $n$ -th order ENO polynomial as

$$u_n(x) = u[x_{l_n(i)}] + u[x_{l_n(i)}, x_{l_n(i)+1}](x - x_{l_n(i)}) + u[x_{l_n(i)}, x_{l_n(i)+1}, x_{l_n(i)+2}](x - x_{l_n(i)})(x - x_{l_n(i)+1}) + \dots$$

$$+ u[x_{l_n(i)}, x_{l_n(i)+1}, \dots, x_{l_n(i)+n}](x - x_{l_n(i)})(x - x_{l_n(i)+1}) \dots (x - x_{l_n(i)+n-1})$$

$$\Rightarrow u(x) = \sum_{j=0}^n u[x_{l_n(i)}, x_{l_n(i)+1}, \dots, x_{l_n(i)+j}] \prod_{k=0}^{j-1} (x - x_{l_n(i)+k}), \quad x_{j-1/2} \leq x \leq x_{j+1/2}$$

- With the conservation constraint of  $\int_{x_{i-1/2}}^{x_{i+1/2}} u(x) = u_i$ , it can be shown that

$$TV(u^{n+1}) \leq TV(u^n) + O(\Delta x^r) \text{ for } r\text{-th order ENO interpolation,}$$

if there are at least  $(r+1)$  smooth points between local smooth extrema.

## ENO interpolation using flux function, $f_i$

- Step 1. Start from first-order flux (mostly, Lax-Friedrich flux)

$$f_i = f_i^+ + f_i^- \text{ with } f_i^+ \equiv \frac{1}{2}(f_i + \alpha u_i), f_i^- \equiv \frac{1}{2}(f_i - \alpha u_i), \alpha \geq \max_{u \in [u_i, u_{i+1}]} |\partial f / \partial u|$$

the cell-interface flux is obtained by  $\hat{f}_{i+1/2} = f_i^+ + f_{i+1}^- = [(f_i + f_{i+1}) - \alpha \Delta u_{i+1/2}] / 2$ .

- Step 2. Carry out the same piecewise polynomial reconstruction using  $f_i^\pm$

$$u[x_i, x_{i+1}, \dots, x_{i+k}] \rightarrow f^\pm[x_i, x_{i+1}, \dots, x_{i+k}] \Rightarrow \hat{f}_{i+1/2} = (f_{iL}^+ + f_{(i+1)R}^-) / 2$$

- **Characteristics of ENO scheme**

- Pure interpolation not limiting technique → accuracy preservation across smooth extrema
- Adaptive smoothest stencil of ENO scheme could be changed by a small perturbation at round-off level → convergence and accuracy problem

- **Weighted ENO (WENO) Scheme**

- **Smoothness indicator is introduced for non-linear weighting coefficients.**

- **WENO polynomial reconstruction using cell-averaged value,  $u_j$**

- Step 1. Interpolation on global/local stencil

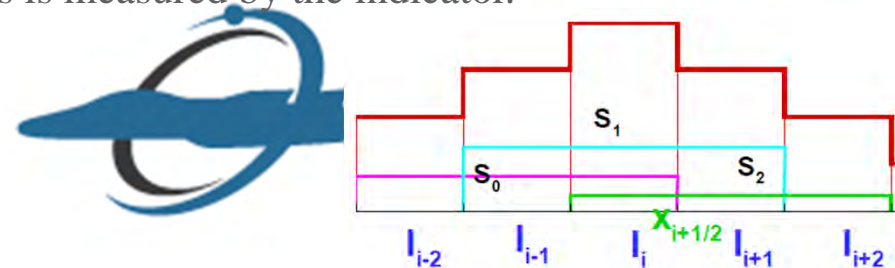
Reconstruct  $k - th$  degree polynomial  $p_j(x)$  on  $S_j$  and  $2k - th$  degree polynomial  $Q(x)$  on  $T = \sum_j S_j$  as

$$\bar{u}_{i+l} = \int_{I_{i+l}} p_j(x) dx / \Delta x_{i+l}, \quad (l = -k + j, \dots, j) \text{ and } \bar{u}_{i+l} = \int_{I_{i+l}} Q(x) dx / \Delta x_{i+l}, \quad (l = -k, \dots, k)$$

Find linear weights such that  $Q(x_{i+1/2}) = \sum_{j=0}^k \gamma_j p_j(x_{i+1/2})$

- Step 2. For each cell  $S_j$ , local smoothness is measured by the indicator.

$$\beta_j = \sum_{l=1}^k \int_{x_{i-1/2}}^{x_{i+1/2}} \Delta x^{2l-1} \left( \frac{\partial^l}{\partial x^l} p_j(x) \right)^2 dx$$



- Step 3. Non-linear weights to satisfy the ENO property
  - If the stencil  $S_j$  is in smooth region:  $\omega_j = O(1)$
  - If the stencil  $S_j$  is in non-smooth region:  $\omega_j \leq O(h^k)$

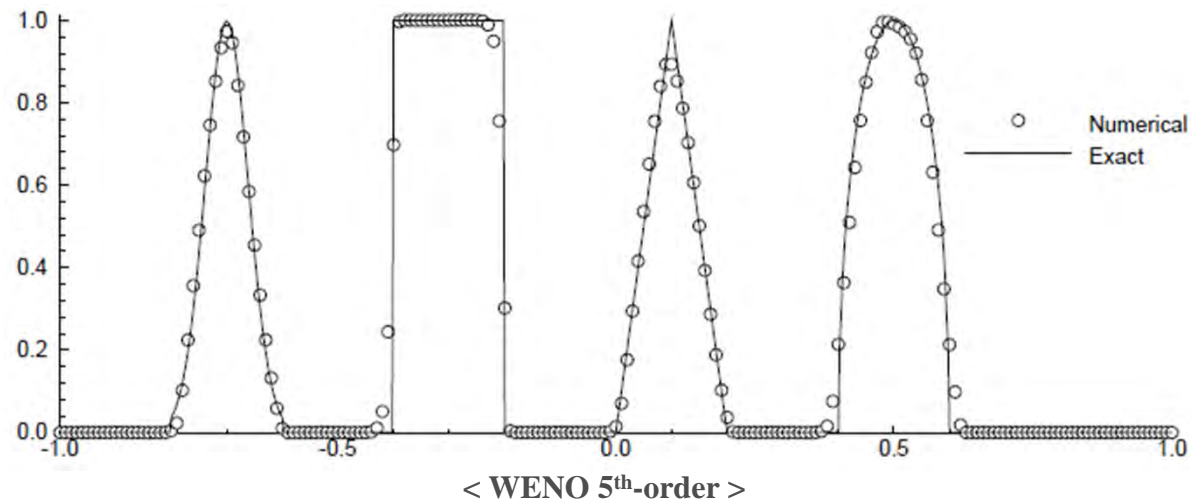
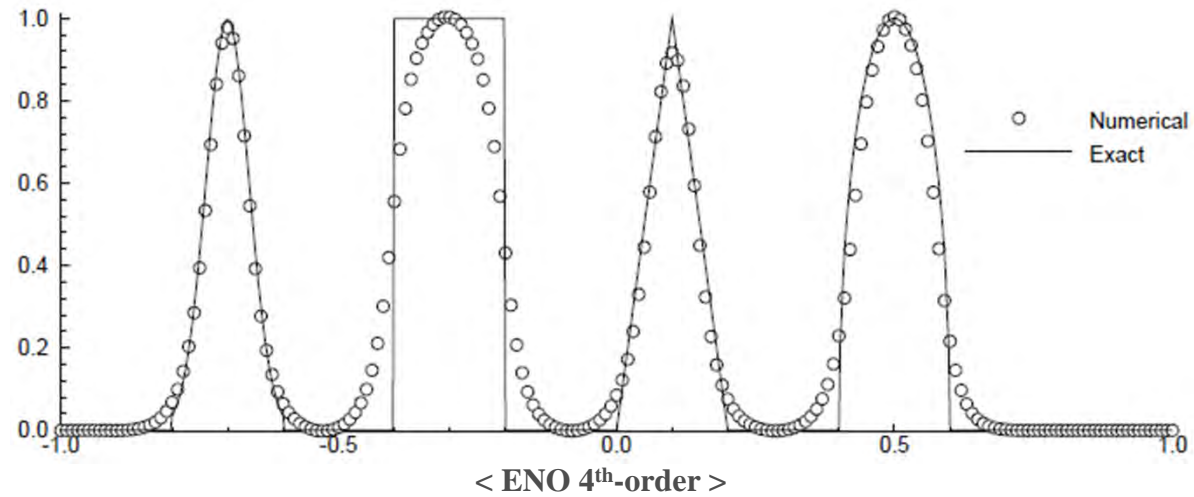
Compute the nonlinear weights based on the smoothness indicator

$$\bar{\omega}_j = \frac{\gamma_j}{(\varepsilon + \beta_j)^2} \quad \text{and} \quad \omega_j = \frac{\bar{\omega}_j}{\sum_l \bar{\omega}_l}$$

- Step 4.  $(2k+1)$ -th degree WENO reconstruction via convex combination with the non-linear weights

$$R_i(x) = \sum_{j=0}^k \omega_j p_j(x) \Rightarrow u_{iR} = R_i(x_{i+1/2}), \quad u_{iL} = R_i(x_{i-1/2})$$

- Example



## Multi-dimensional Limiting Process

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- **Problems in Multi-dimensional Extension**
- **Limiting Condition on Multiple Dimensions**
- **Formulation of MLP on Structured Grids**
- **Maximum Principle and MLP Condition**
- **Numerical Results**



## ● Critical Survey

### ● Analyses based on one-dimensional flow physics

- 1-D SCL of  $\frac{\partial u}{\partial t} + \frac{\partial f(u)}{\partial x} = 0$  with  $f(u) = au, \frac{u^2}{2} \Rightarrow \frac{d\bar{u}_j}{dt} = -\frac{1}{\Delta x}(\hat{f}_{j+1/2} - \hat{f}_{j-1/2})$
- Monotonicity constraint on  $u_{j+1/2}$ :  $\min(\bar{u}_j, \bar{u}_{j+1}) \leq u_{j+1/2} \leq \max(\bar{u}_j, \bar{u}_{j+1})$

### ● Multi-dimensional extension

- 2-D SCL of  $\frac{\partial u}{\partial t} + \frac{\partial f(u)}{\partial x} + \frac{\partial g(u)}{\partial y} = 0 \Rightarrow \frac{d\bar{u}_{i,j}}{dt} = -\frac{1}{\Delta x}(\hat{f}_{i+1/2,j} - \hat{f}_{i-1/2,j}) - \frac{1}{\Delta y}(\hat{g}_{i,j+1/2} - \hat{g}_{i,j-1/2})$

- Monotonicity constraint by dimensional splitting:

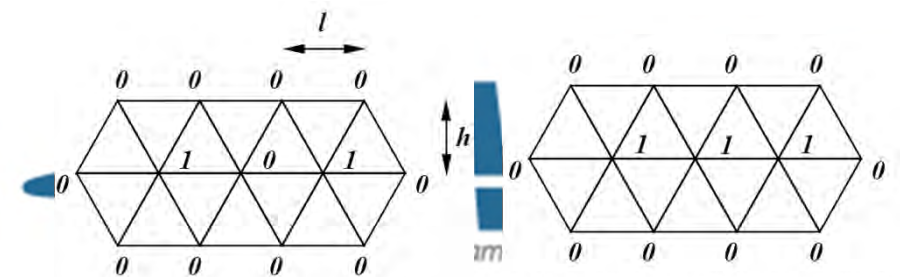
$$\left. \begin{array}{l} (x - dir) \min(\bar{u}_{i,j}, \bar{u}_{i+1,j}) \leq u_{i+1/2,j} \leq \max(\bar{u}_{i,j}, \bar{u}_{i+1,j}) \\ (y - dir) \min(\bar{u}_{i,j}, \bar{u}_{i,j+1}) \leq u_{i,j+1/2} \leq \max(\bar{u}_{i,j}, \bar{u}_{i,j+1}) \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} \text{Is it good enough to handle multi-} \\ \text{dimensional flow situation?} \end{array} \right.$$

### ● Two-peaks and one-ridge problem (Jameson, 1995)

- For  $TV(\bar{u}) = \int_V \|\nabla \bar{u}\|_p ds$  with  $p = 1, 2, \infty$

$$TV_{two-peaks} < TV_{one-ridge}$$

- 2-D TVD scheme is at most first-order accurate (Goodman and LeVeque, 1985)

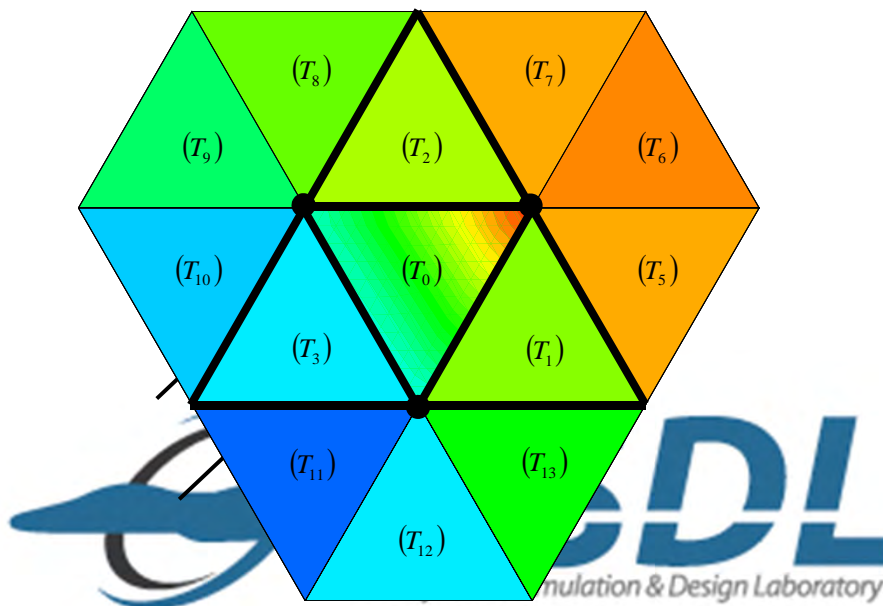
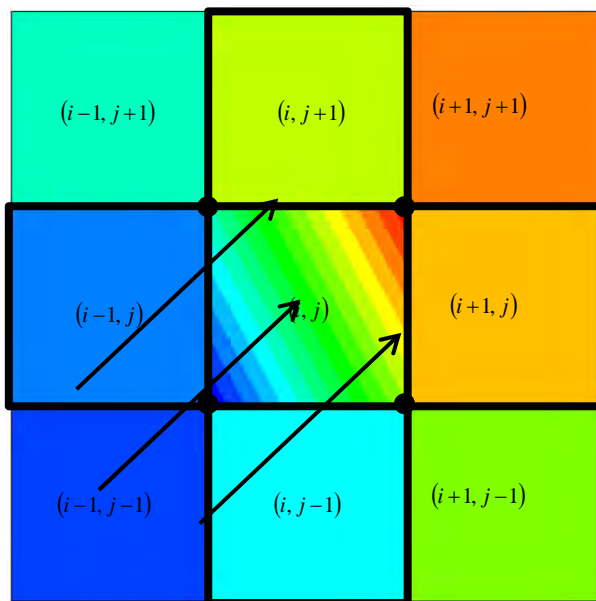


Two Peaks:  $TV = 4 + 2\sqrt{3}$  ( $L_1$ ),  $6$  ( $L_2$ ),  
or  $2 + 2\sqrt{3}$  ( $L_\infty$ ).

One Ridge:  $TV = 6 + \sqrt{3}$  ( $L_1$ ),  $7$  ( $L_2$ ),  
or  $5 + 3\sqrt{3}$  ( $L_\infty$ ).



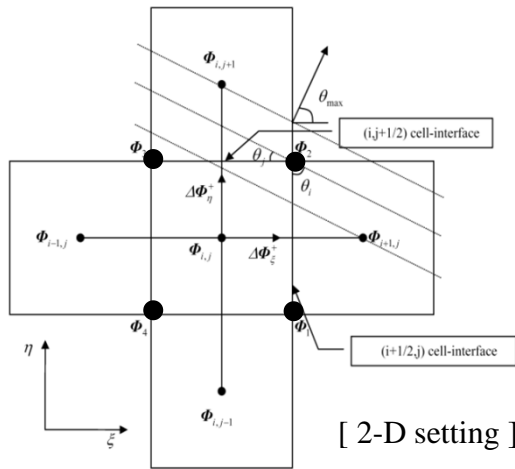
- Dimensional splitting extension is not sufficient to control numerical oscillations in multiple dimensions.
- Dimensional splitting TVD schemes do not guarantee monotonicity at vertex (Kim & Kim, 2005 / Yoon & Kim, 2008)
- Limiting strategy in Non Grid-aligned Distributions
  - Dimensional splitting approach cannot handle local extrema at vertex



< Rectilinear vs. Curvilinear Grids & Limiting? >



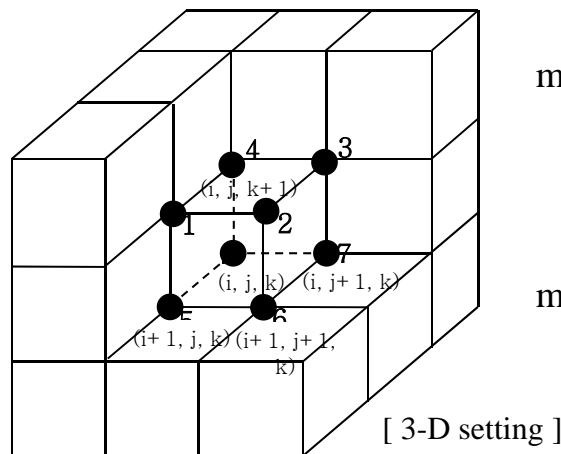
- **Multiple Dimensional Limiting Condition:**  $\bar{u}_{neighbor}^{\min} \leq u \leq \bar{u}_{neighbor}^{\max}$
- **MLP Condition on 2-D and 3-D Structured Grids**



$$\min(\bar{u}_{i,j}, \bar{u}_{i+1,j}, \bar{u}_{i,j-1}, \bar{u}_{i+1,j-1}) \leq u_1 \leq \max(\bar{u}_{i,j}, \bar{u}_{i+1,j}, \bar{u}_{i,j-1}, \bar{u}_{i+1,j-1})$$

...

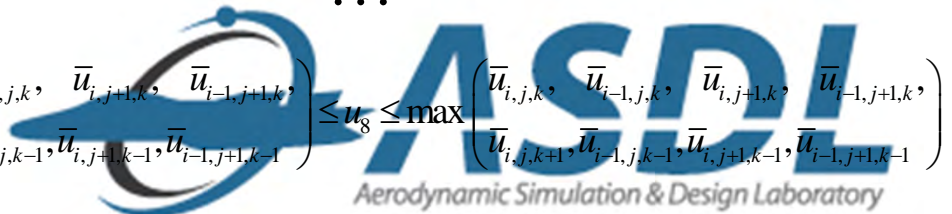
$$\min(\bar{u}_{i,j}, \bar{u}_{i-1,j}, \bar{u}_{i,j+1}, \bar{u}_{i-1,j+1}) \leq u_4 \leq \max(\bar{u}_{i,j}, \bar{u}_{i-1,j}, \bar{u}_{i,j+1}, \bar{u}_{i-1,j+1})$$



$$\min \begin{pmatrix} \bar{u}_{i,j,k}, \bar{u}_{i+1,j,k}, \bar{u}_{i,j-1,k}, \bar{u}_{i+1,j-1,k} \\ \bar{u}_{i,j,k+1}, \bar{u}_{i+1,j,k+1}, \bar{u}_{i,j-1,k+1}, \bar{u}_{i+1,j-1,k+1} \end{pmatrix} \leq u_1 \leq \max \begin{pmatrix} \bar{u}_{i,j,k}, \bar{u}_{i+1,j,k}, \bar{u}_{i,j-1,k}, \bar{u}_{i+1,j-1,k} \\ \bar{u}_{i,j,k+1}, \bar{u}_{i+1,j,k+1}, \bar{u}_{i,j-1,k+1}, \bar{u}_{i+1,j-1,k+1} \end{pmatrix}$$

...

$$\min \begin{pmatrix} \bar{u}_{i,j,k}, \bar{u}_{i-1,j,k}, \bar{u}_{i,j+1,k}, \bar{u}_{i-1,j+1,k} \\ \bar{u}_{i,j,k+1}, \bar{u}_{i-1,j,k+1}, \bar{u}_{i,j+1,k+1}, \bar{u}_{i-1,j+1,k+1} \end{pmatrix} \leq u_8 \leq \max \begin{pmatrix} \bar{u}_{i,j,k}, \bar{u}_{i-1,j,k}, \bar{u}_{i,j+1,k}, \bar{u}_{i-1,j+1,k} \\ \bar{u}_{i,j,k+1}, \bar{u}_{i-1,j,k+1}, \bar{u}_{i,j+1,k+1}, \bar{u}_{i-1,j+1,k+1} \end{pmatrix}$$

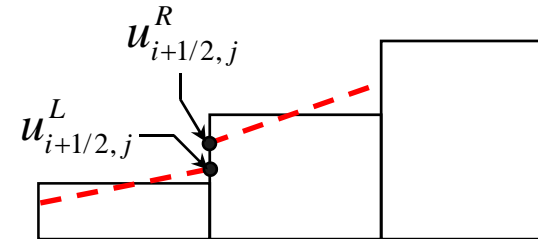


## ● TVD-MUSCL vs. MLP on Structured Mesh

$$u_{i+1/2,j}^L = \bar{u}_{i,j} + 0.5\phi(r_L)\Delta\bar{u}_{i-1/2,j}$$

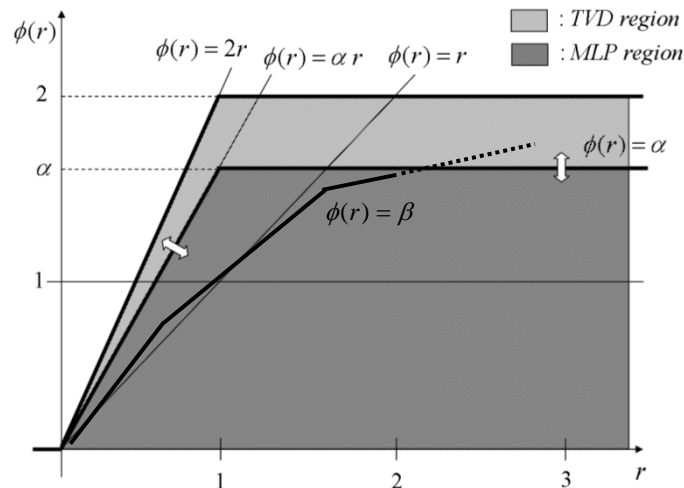
$$u_{i+1/2,j}^R = \bar{u}_{i+1,j,k} - 0.5\phi(r_R)\Delta\bar{u}_{i+3/2,j}$$

$$\text{with } r_L = \frac{\Delta\bar{u}_{i+1/2,j}}{\Delta\bar{u}_{i-1/2,j}}, r_R = \frac{\Delta\bar{u}_{i+1/2,j}}{\Delta\bar{u}_{i+3/2,j}}$$



## ● Limiting region

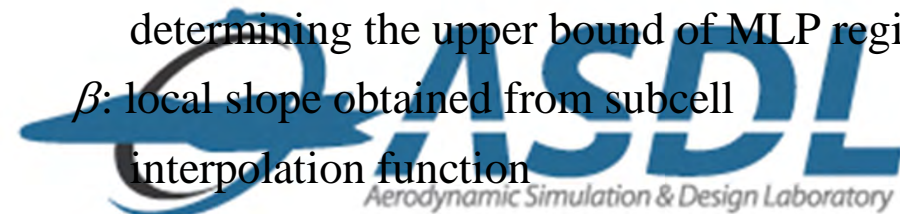
- $\phi(r) = \max(0, \min(2, 2r))$ : TVD limiting region based on 1-D analysis
- $\phi(r) = \max(0, \min(\alpha, \alpha r))$ : MLP limiting region for multi-dimensional flows



[ TVD vs. MLP region ]

## ● MLP limiting function

- $\phi(r) = \max(0, \min(\alpha, \alpha r, \beta))$  with
  - $\alpha$ : multi-dimensional restriction coefficient determining the upper bound of MLP region
  - $\beta$ : local slope obtained from subcell interpolation function





- **Determination of Multi-dimensional Restriction Coefficient,  $\alpha$**

- **Step 1. Assume a cell-vertex value estimated by**

$$u_{i+1/2,j+1/2} = \bar{u}_{i,j} + \Delta u_{i+1/2,j}^x + \Delta u_{i,j+1/2}^y = \bar{u}_{i,j} + (1 + r_{xy}) \Delta u_{i+1/2,j}^x \quad \text{with } r_{xy} = \Delta u_{i,j+1/2}^y / \Delta u_{i+1/2,j}^x$$

- Note that we only need to check the upper bound of the maximum vertex value and the lower bound of the minimum vertex values. Thus,  $r_{xy}$  is always positive.

- **Step 2. Obtain the neighboring minimum and maximum values by checking all cell-averaged values sharing the same vertex point ( $i + 1/2, j + 1/2$ )**

$$u_{i+1/2,j+1/2}^{\min} = \min(\bar{u}_{i,j}, \bar{u}_{i+1,j}, \bar{u}_{i,j+1}, \bar{u}_{i+1,j+1}), \quad u_{i+1/2,j+1/2}^{\max} = \max(\bar{u}_{i,j}, \bar{u}_{i+1,j}, \bar{u}_{i,j+1}, \bar{u}_{i+1,j+1})$$

- **Step 3. Enforce the MLP limiting condition  $\phi_{MLP}(r) = \min(\alpha, \alpha r)$  onto  $u_{i+1/2,j+1/2}$**

$$\bar{u}_{i+1/2,j+1/2}^{\min} \leq u_{i+1/2,j+1/2} \leq \bar{u}_{i+1/2,j+1/2}^{\max} \quad \text{or} \quad \bar{u}_{i+1/2,j+1/2}^{\min} \leq \bar{u}_{i,j} + (1 + r_{xy}) \Delta u_{i+1/2,j}^x \leq \bar{u}_{i+1/2,j+1/2}^{\max}$$

$$\frac{\bar{u}_{i+1/2,j+1/2}^{\min} - \bar{u}_{i,j}}{1 + r_{xy}} \leq \Delta u_{i+1/2,j}^x \leq \frac{\bar{u}_{i+1/2,j+1/2}^{\max} - \bar{u}_{i,j}}{1 + r_{xy}} \quad \text{or}$$

$$\frac{\bar{u}_{i+1/2,j+1/2}^{\min} - \bar{u}_{i,j}}{1 + r_{xy}} \leq 0.5 \phi_{MLP}(r_x) \Delta \bar{u}_{i-1/2,j} \leq \frac{\bar{u}_{i+1/2,j+1/2}^{\max} - \bar{u}_{i,j}}{1 + r_{xy}} \quad \text{with } r_x = \Delta \bar{u}_{i+1/2,j}^x / \Delta \bar{u}_{i-1/2,j}$$



- **Determination of Multi-dimensional Restriction Coefficient,  $\alpha$  (cont'd)**

- **(S4) Obtain the range of  $\alpha$**

If  $\Delta u_{i+1/2,j}^x > 0$  (local maximum),

$$0 \leq 0.5\phi(r_x)\Delta\bar{u}_{i-1/2,j} \leq \frac{\bar{u}_{i+1/2,j+1/2}^{\max} - \bar{u}_{i,j}}{1+r_{xy}} \Rightarrow 0 \leq \alpha_{i+1/2,j+1/2} \leq \frac{2\max(1,r_x)}{(1+r_{xy})\Delta\bar{u}_{i+1/2,j}} (\bar{u}_{i+1/2,j+1/2}^{\max} - \bar{u}_{i,j})$$

If  $\Delta u_{i+1/2,j}^x < 0$  (local minimum),

$$\frac{\bar{u}_{i+1/2,j+1/2}^{\min} - \bar{u}_{i,j}}{1+r_{xy}} \leq 0.5\phi(r_x)\Delta\bar{u}_{i-1/2,j} \leq 0 \Rightarrow 0 \leq \alpha_{i+1/2,j+1/2} \leq \frac{2\max(1,r_x)}{(1+r_{xy})\Delta\bar{u}_{i+1/2,j}} (\bar{u}_{i+1/2,j+1/2}^{\min} - \bar{u}_{i,j})$$

Thus, we have

$$0 \leq \alpha_{i+1/2,j+1/2} \leq \left| \frac{2\max(1,r_x)}{(1+r_{xy})\Delta\bar{u}_{i+1/2,j}} \right| \min \left( \left| \bar{u}_{i+1/2,j+1/2}^{\max} - \bar{u}_{i,j} \right|, \left| \bar{u}_{i+1/2,j+1/2}^{\min} - \bar{u}_{i,j} \right| \right)$$

By checking four vertex points  $(i \pm 1/2, j \pm 1/2)$ , we finally have

$$0 \leq \alpha \leq \left| \frac{2\max(1,r_x)}{(1+r_{xy})\Delta\bar{u}_{i+1/2,j}} \right| \min \left( \left| \bar{u}_{i+1/2,j+1/2}^{\max} - \bar{u}_{i,j} \right|, \left| \bar{u}_{i+1/2,j+1/2}^{\min} - \bar{u}_{i,j} \right| \right) \text{ with } r_{xy} = \Delta u_{i,j+1/2}^y / \Delta u_{i+1/2,j}^x$$

- Note that if  $r_{xy} = 0$ ,  $\alpha$  recovers the 1-D TVD form.

- **MLP Slope Limiters**

- **Cell-interface values in the  $x$ -direction**

$$u_{i+1/2,j}^L = \bar{u}_{i,j} + 0.5\phi(r_L)\Delta\bar{u}_{i-1/2,j} = \bar{u}_{i,j} + 0.5 \max(0, \min(\alpha_L, \alpha_L r_L, \beta_L)) \Delta\bar{u}_{i-1/2,j}$$

$$u_{i+1/2,j}^R = \bar{u}_{i+1,j} - 0.5\phi(r_R)\Delta\bar{u}_{i+3/2,j} = \bar{u}_{i+1,j} - 0.5 \max(0, \min(\alpha_R, \alpha_R r_R, \beta_R)) \Delta\bar{u}_{i+3/2,j}$$

$$\alpha_L = \left| \frac{2 \max(1, r_L)}{(1 + r_{xy}) \Delta\bar{u}_{i+1/2,j}} \right| \min \left( \left| \bar{u}_{i\pm 1/2, j\pm 1/2}^{\max} - \bar{u}_{i,j} \right|, \left| \bar{u}_{i\pm 1/2, j\pm 1/2}^{\min} - \bar{u}_{i,j} \right| \right), \quad \alpha_R = \alpha_L \Big|_{i \rightarrow i+1}$$

with  $r_L = \Delta\bar{u}_{i+1/2,j} / \Delta\bar{u}_{i-1/2,j}$  and  $r_R = \Delta\bar{u}_{i+1/2,j} / \Delta\bar{u}_{i+3/2,j}$

- **Choice of local slopes ( $\beta$ )**

- MLP-van Leer:  $\beta_{L \text{ or } R} = \frac{2r_{L \text{ or } R}}{1 + |r_{L \text{ or } R}|}$

- MLP-superbee:  $\beta_{L \text{ or } R} = \begin{cases} \min(2r_{L \text{ or } R(i-1)}, 1), & \text{if } 0 < r_{L \text{ or } R(i-1)} < 1 \\ \min(r_{L \text{ or } R(i-1)}, 2), & \text{if } r_{L \text{ or } R(i-1)} > 1 \end{cases}$  with  $r_{R(i-1)} = \frac{\Delta\bar{u}_{i-1/2,j}}{\Delta\bar{u}_{i+1/2,j}}$

- MLP3 (MLP slope limiting with 3<sup>rd</sup>-order polynomial):  $\beta_{L \text{ or } R} = \frac{1 + 2r_{L \text{ or } R}}{3}$

- MLP5 (MLP slope limiting with 5<sup>th</sup>-order polynomial):

$$\beta_L = \frac{-2/r_{L(i-1)} + 11 + 24r_L - 3r_L r_{L(i+1)}}{30}, \quad \beta_R = \frac{-2/r_{R(i+2)} + 11 + 24r_{R(i+1)} - 3r_{R(i+1)} r_R}{30}$$

- **Maximum Principle**

- Maximum principle is well-established in parabolic and elliptic PDEs as a tool for  $L_\infty$  stability.
- TVD condition, which is a sufficient condition for the maximum principle, is not available in multi-dimensional situation.
- A condition ensuring multi-dimensional monotonicity
  - Cockburn *et al.* (1990), Liu (1993), Barth (2003), Kim *et al.* (2008, 2010, 2012)

- **$L_\infty$  Stability of MLP Limiting**

- *For a multi-dimensional hyperbolic scalar conservation law of*

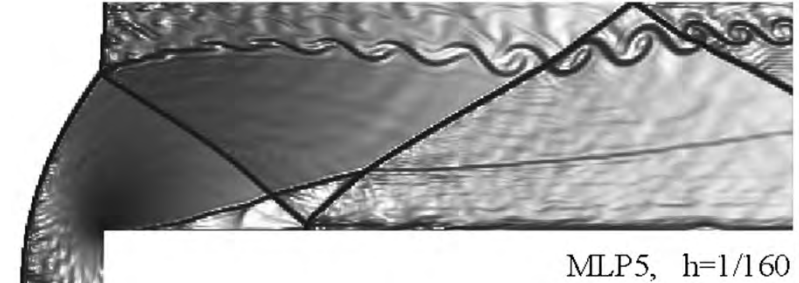
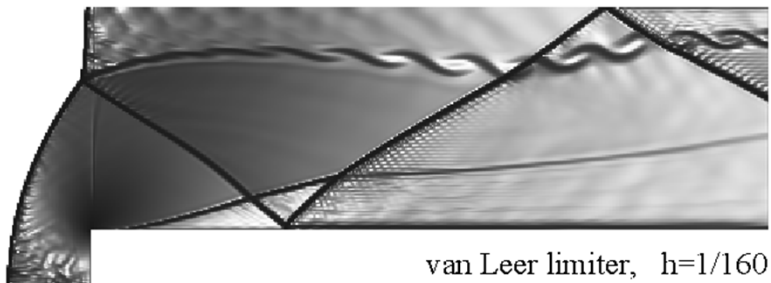
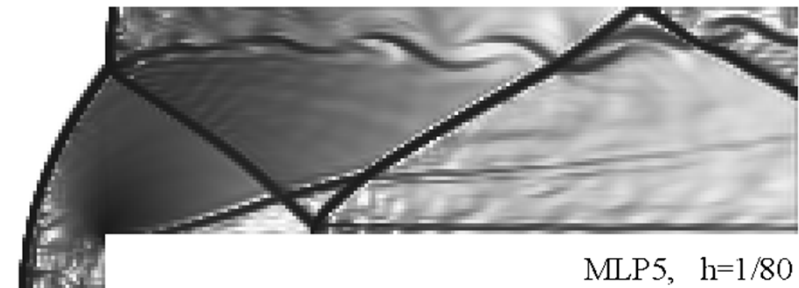
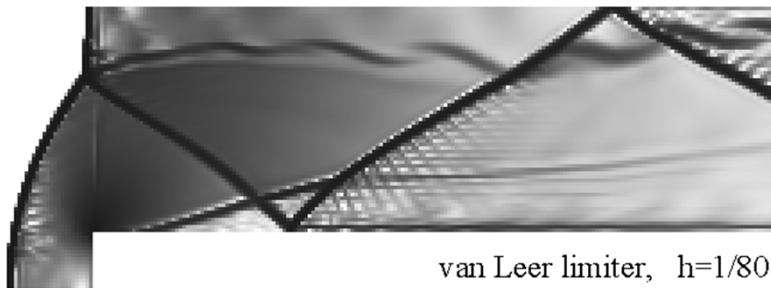
$$\frac{\partial u}{\partial t} + \frac{\partial f(u)}{\partial x} + \frac{\partial g(u)}{\partial y} = 0,$$

*the fully discrete scheme using the MLP limiting satisfies the local maximum principle under a suitable CFL condition.*

If  $\bar{u}_{neighbor}^{\min,n} \leq \bar{u}_{i,j}^n \leq \bar{u}_{neighbor}^{\max,n}$ , then  $\bar{u}_{neighbor}^{\min,n} \leq \bar{u}_{i,j}^{n+1} \leq \bar{u}_{neighbor}^{\max,n}$ .



- A Mach 3 Wind Tunnel with a Step
  - A standard test case for high resolution schemes
  - Advantage in resolving the slip lines

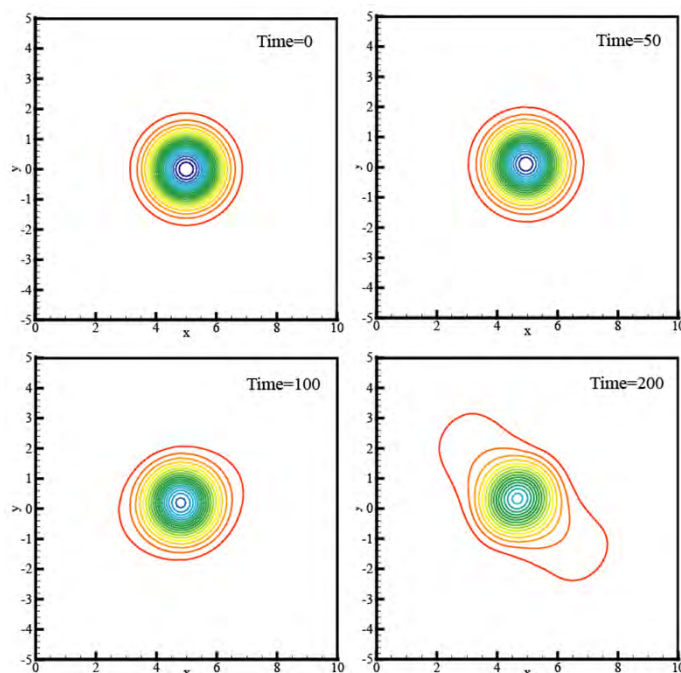


< Numerical Schlieren of a Mach 3 wind tunnel with a step problem >

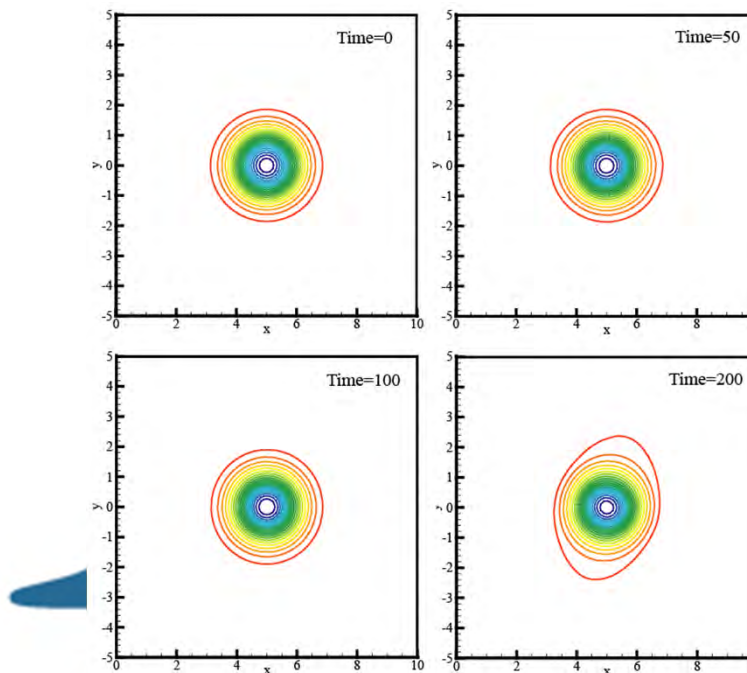
- **Isentropic Vortex Advection**
- **Initial mean flow and perturbation values**

$u_\infty = 1, v_\infty = 1, p_\infty = \rho_\infty = T_\infty = 1$  with

$$(\delta u, \delta v) = \frac{\beta}{2\pi} e^{(1-r^2)/2} (-\bar{y}, \bar{x}), \delta T = -\frac{(\gamma-1)\beta^2}{8\gamma\pi^2} e^{1-r^2}.$$



< *van Leer limiter* >



< *MLP5* >

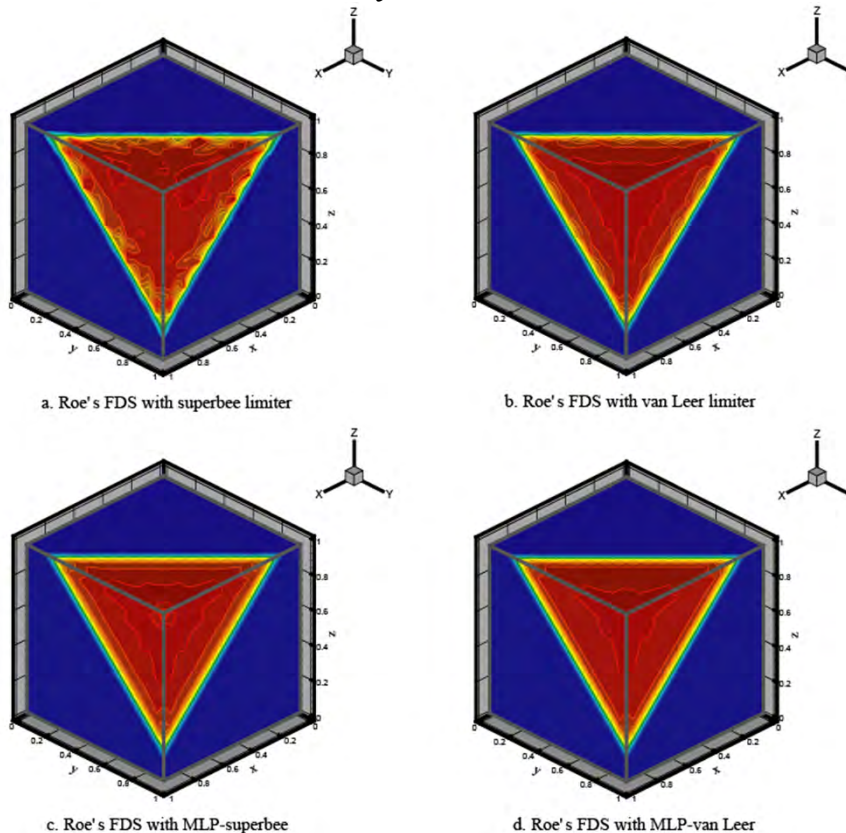


Scheme	Size	$L_1$ error	$L_1$ order	$L_\infty$ error	$L_\infty$ order
van Leer limiter	50 × 50	4.6165E-03	–	8.9412E-02	–
	100 × 100	1.0489E-03	2.14	2.2579E-02	1.99
	150 × 150	4.5167E-04	2.08	1.0169E-02	1.97
	200 × 200	2.5121E-04	2.04	5.6879E-03	2.02
MLP-van Leer limiter	50 × 50	3.7699E-03	–	8.0859E-02	–
	100 × 100	7.1064E-04	2.41	2.0432E-02	1.98
	150 × 150	2.9575E-04	2.16	9.2804E-03	1.95
	200 × 200	1.5657E-04	2.21	5.1476E-03	2.05
MLP3	50 × 50	2.0713E-03	–	3.9473E-02	–
	100 × 100	2.9620E-04	2.81	7.2746E-03	2.44
	150 × 150	9.8557E-05	2.71	2.9668E-03	2.21
	200 × 200	4.3257E-05	2.86	1.4262E-03	2.55
MLP5	50 × 50	1.1441E-03	–	2.4834E-02	–
	100 × 100	2.2427E-04	2.35	5.5389E-03	2.16
	150 × 150	7.6479E-05	2.65	1.9767E-03	2.54
	200 × 200	4.0756E-05	2.19	1.0291E-03	2.27
Third order polynomial	50 × 50	1.5736E-03	–	4.1150E-02	–
	100 × 100	2.6245E-04	2.58	6.3480E-03	2.70
	150 × 150	9.6728E-05	2.46	2.0090E-03	2.84
	200 × 200	4.9086E-05	2.36	8.9188E-04	2.82

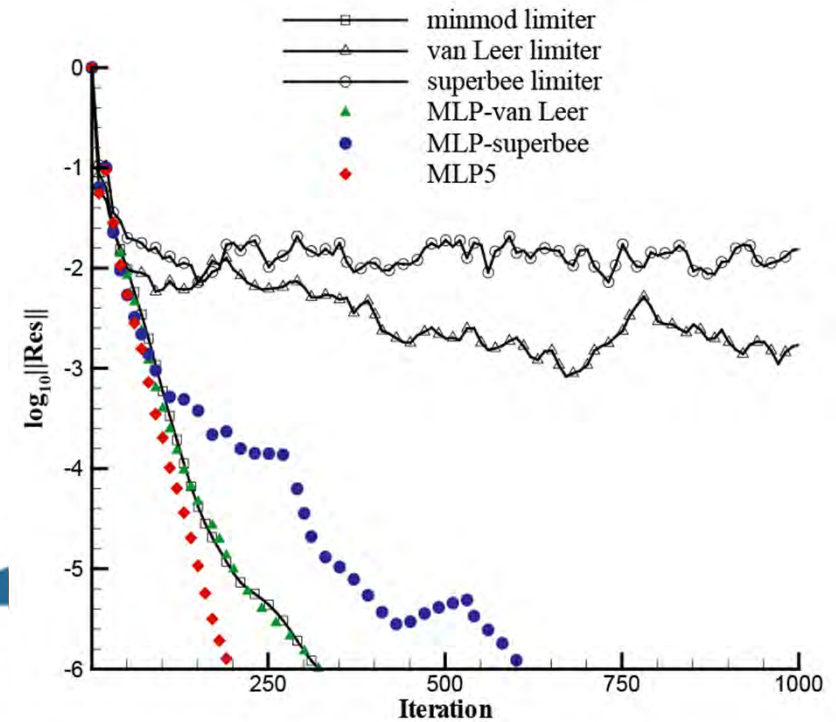
< Grid refinement test >

- **Three-dimensional Normal Shock Discontinuity**
  - Examine the shock-capturing characteristics of TVD and MLP
  - Normal shock discontinuity inclined by 45 degree angle to each cell-surface

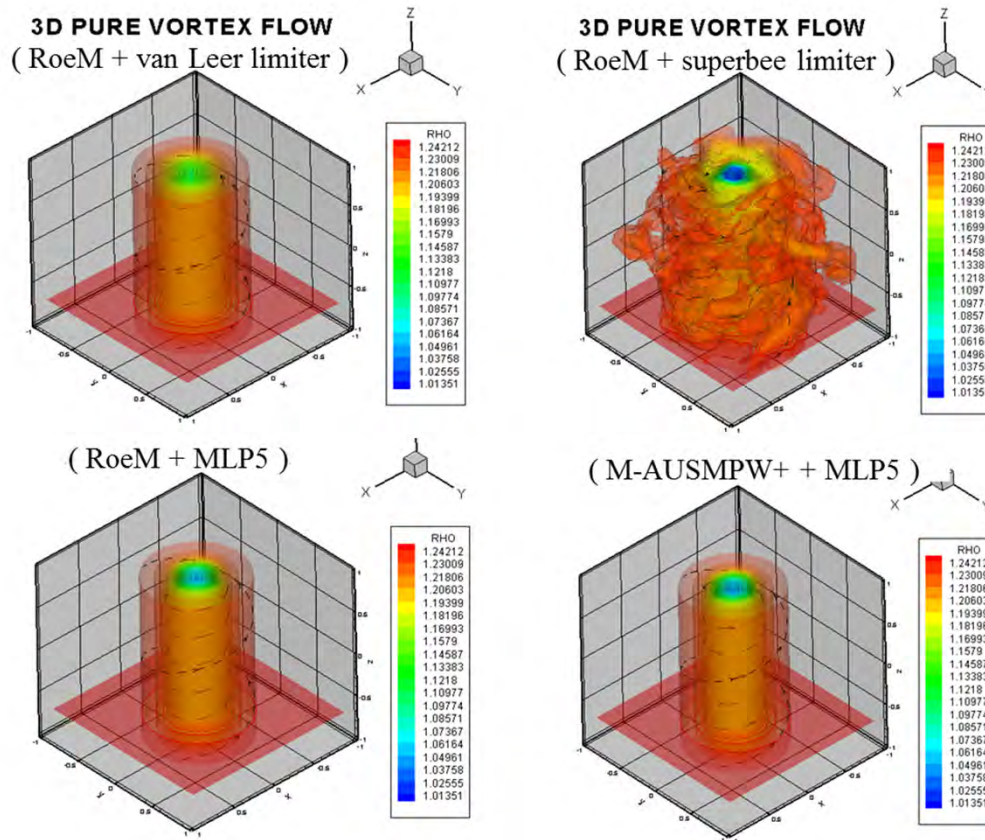
< Density contour >



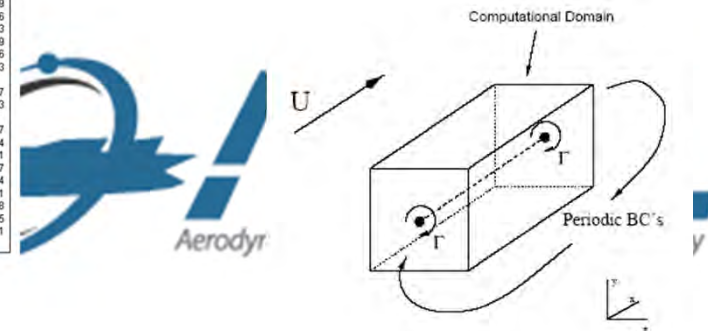
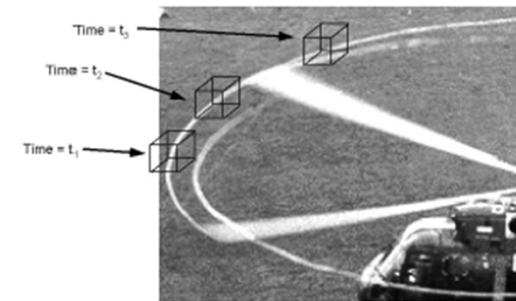
< Error history >



- **Three-dimensional Pure Vortex Flow**
  - A simple model of a tip vortex element released from a rotor blade.
    - Thomson-Rankine vortex model with 50 x 50 x 50 grids

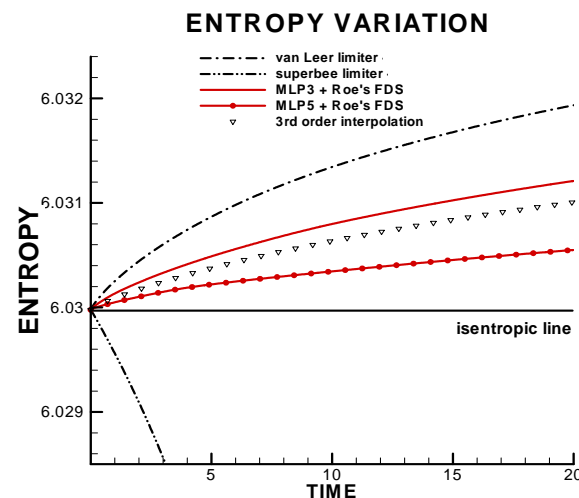
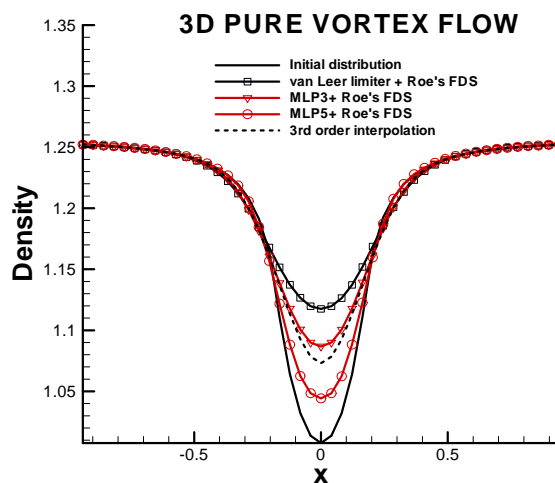


$$v_{\theta} = \begin{cases} \omega r, & r < a \\ \frac{\omega a^2}{r}, & r > a \end{cases} \text{ with } \omega = \text{const}$$

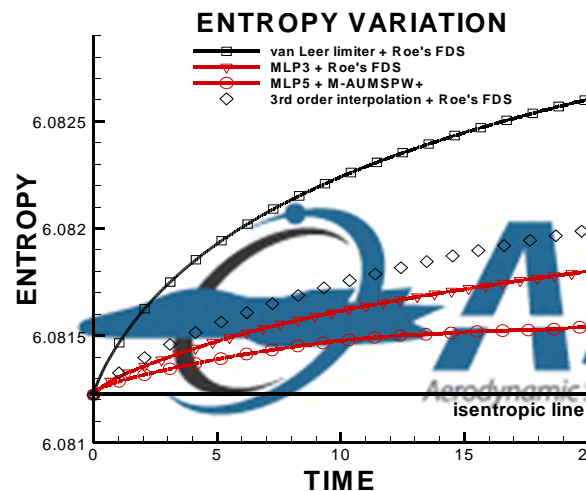
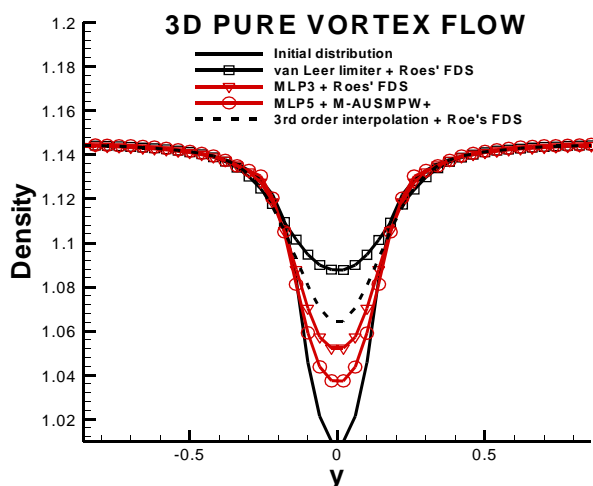


< Density contours at non-dimensional time 20.0 >

## ● The Case of Vortex Core Aligned with z-axis

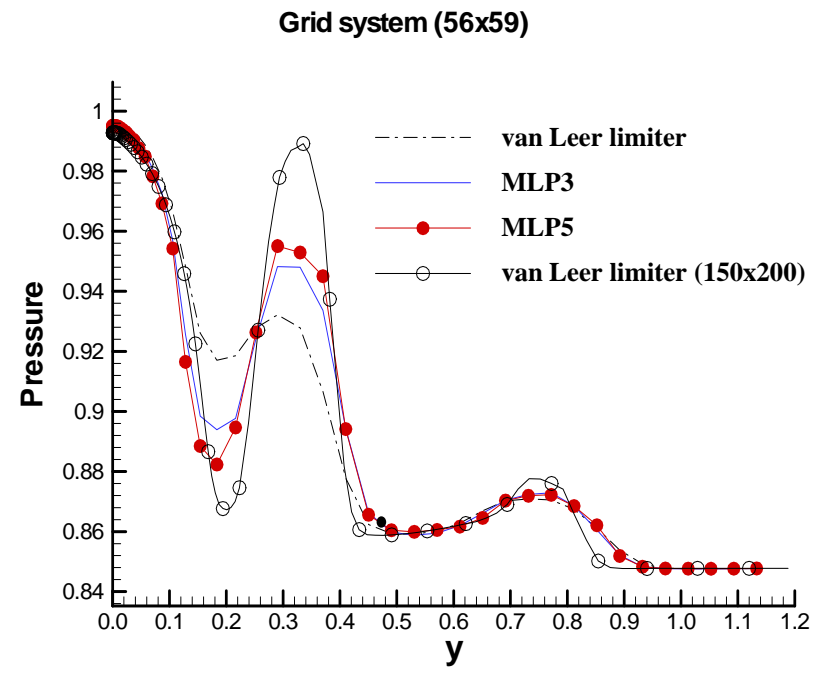
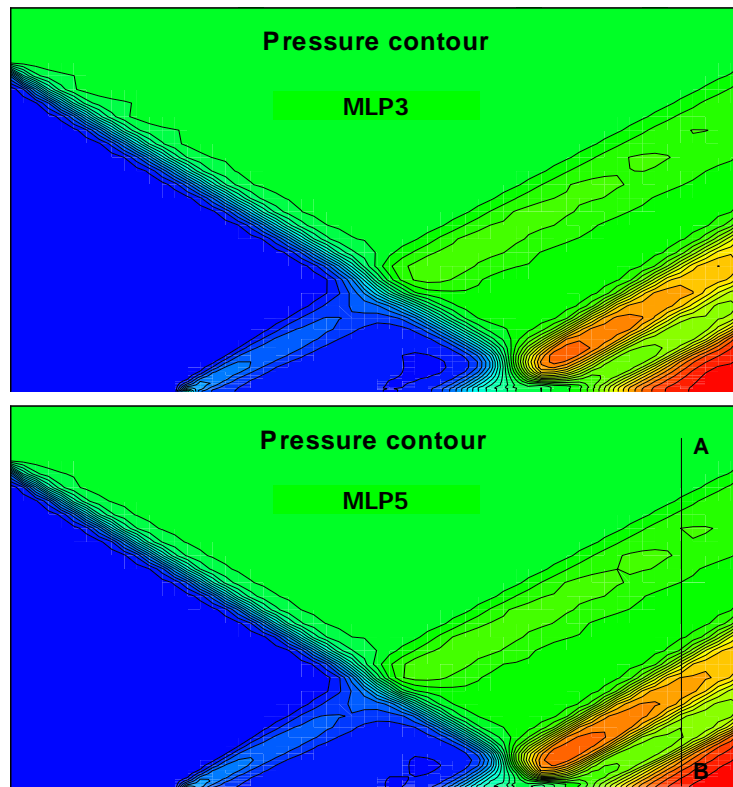


## ● The Case of Vortex Core Non-aligned with z-axis

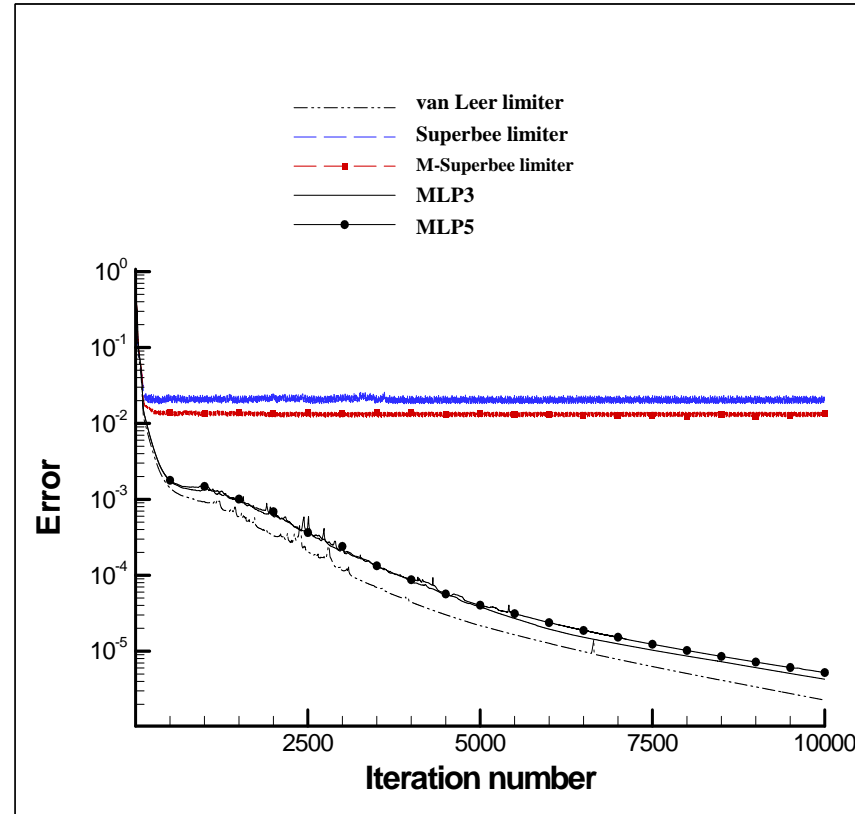
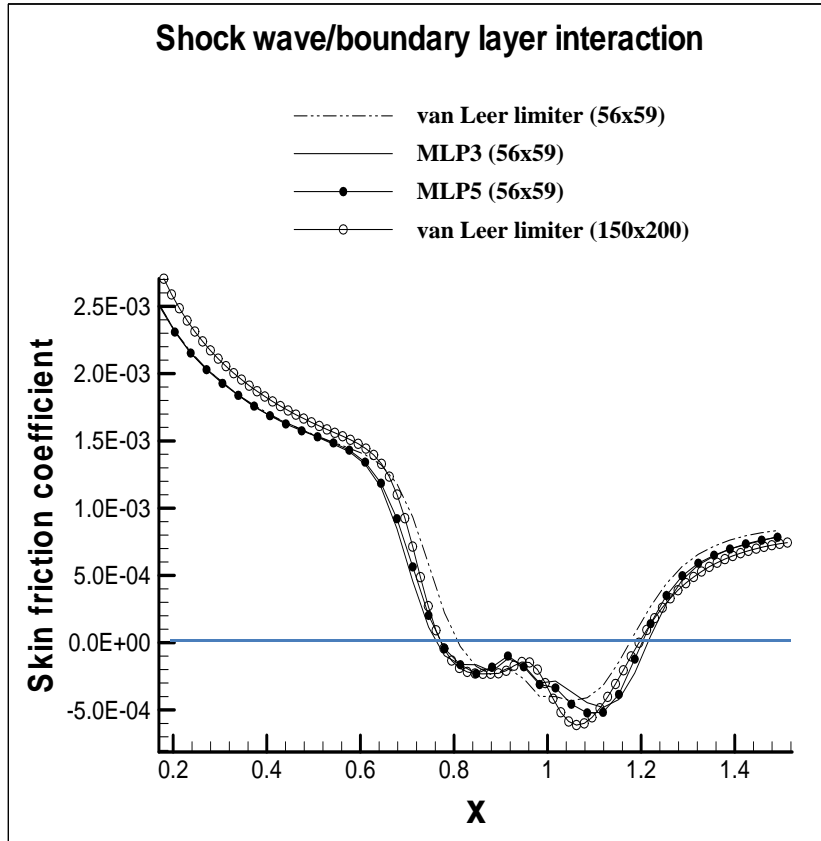


- **Laminar Shock-Boundary Layer Interaction**

- Free stream Mach number :  $M_\infty = 2.0$ , shock impinging angle :  $\theta = 32.585$  degree
- Reynolds number :  $Re = 2.96 \times 10^5$
- Grid system :  $56 \times 59, 150 \times 200$



< Comparison of pressure distribution along line AB >



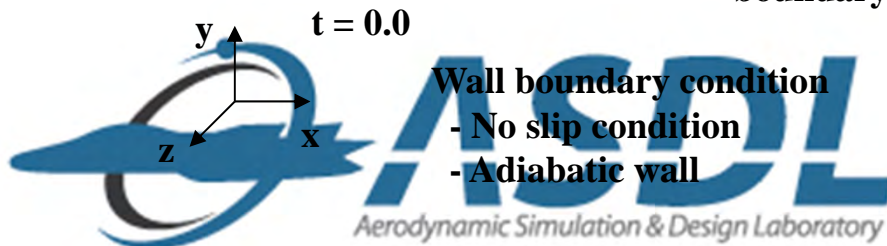
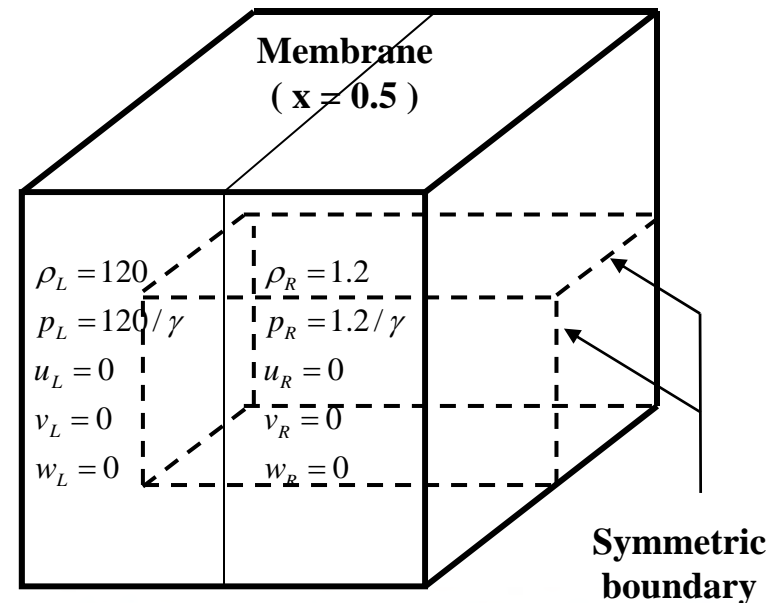
< Skin friction coefficient and error history >



- **3-D Viscous Shock Tube Problem**

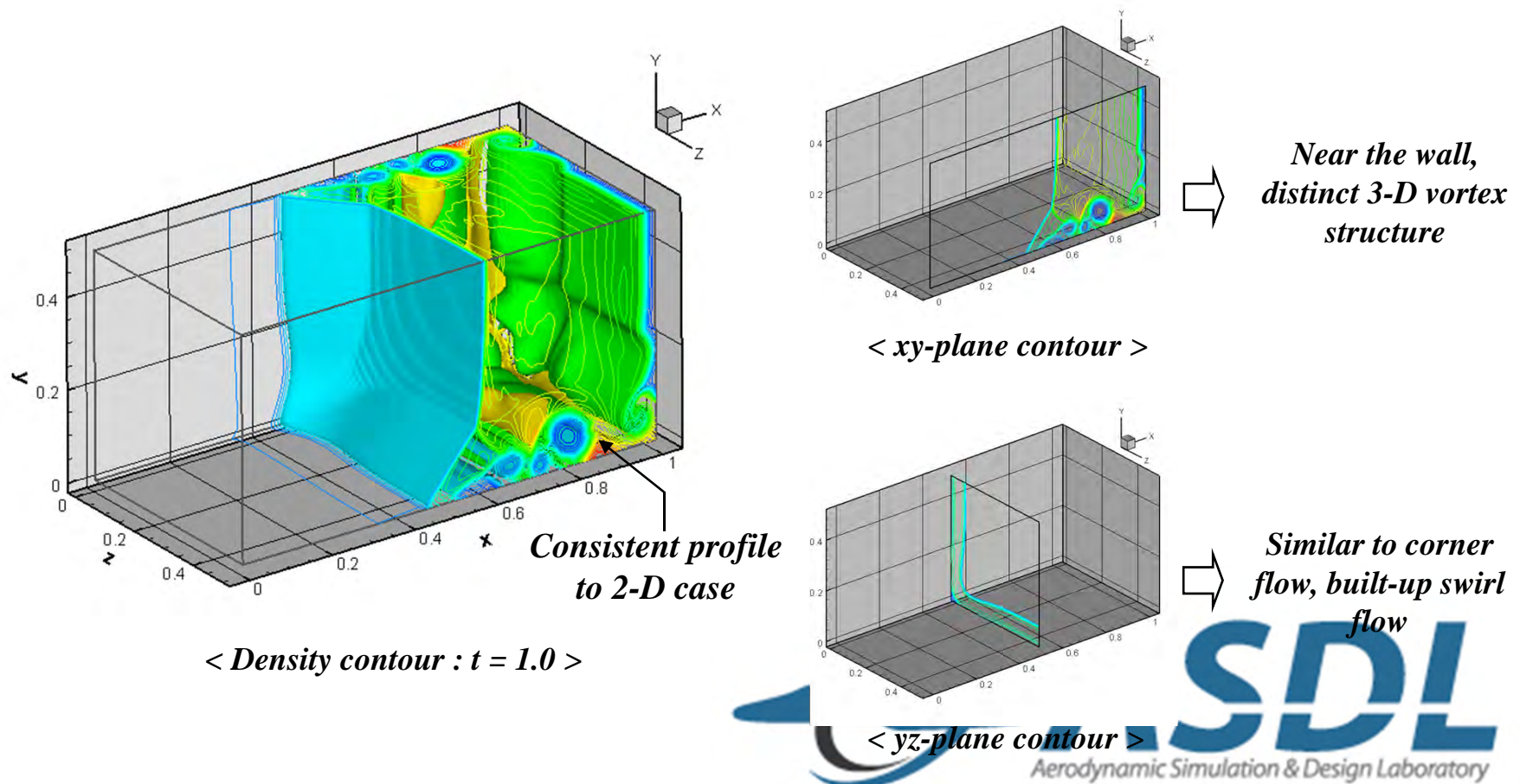
- **Computed conditions**

- Spatial discretization
  - AUSMPW+, M-AUSMPW+
- Time integration
  - 3<sup>rd</sup>-order TVD Runge-Kutta
- Grid system
  - 250x125x125, 350x175x175, 500x250x250
- Reynolds number: 200 with constant viscosity
- Boundary values
  - No slip B.C ( $x=0, x=1, y=0, z=0$ )
  - Symmetric B.C ( $y=0.5, z=0.5$ )



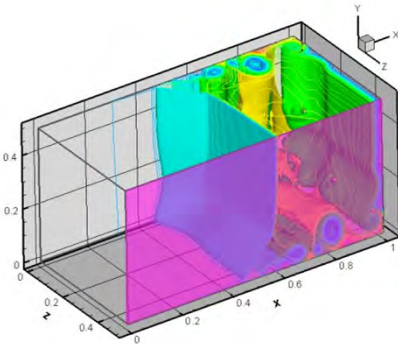
- Profiles of 3-D Viscous Shock Tube Problem

- Complicated shock-boundary interaction, and vortical structure



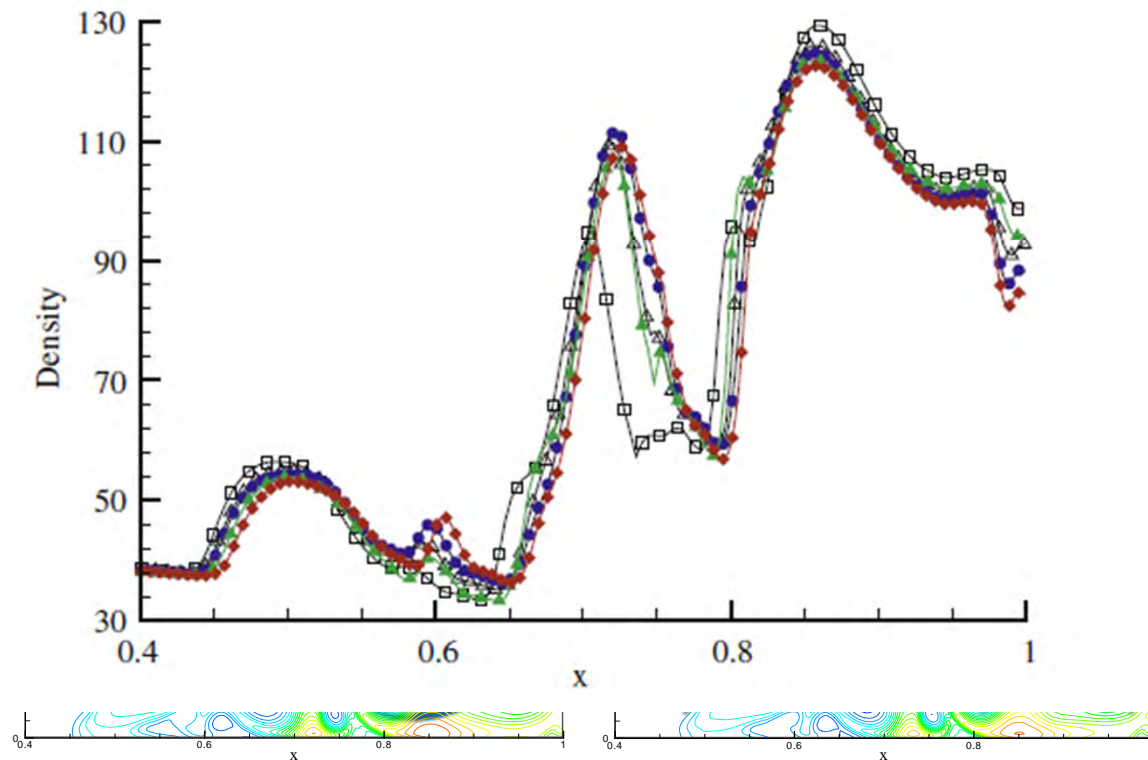


*Comparison :  
profile of  
separated vortex*



	No. of grid pts.
Coarse mesh	3,906,250 (250x125x125)
Medium mesh	10,718,750 (350x175x175)
Fine mesh	31,250,000 (500x250x250)

- van Leer limiter with AUSMPW+ (250x125x125)
- △— van Leer limiter with AUSMPW+ (350x175x175)
- van Leer limiter with AUSMPW+ (500x250x250)
- ▲— MLP5 with M-AUSMPW+ (250x125x125)
- ◆— MLP5 with M-AUSMPW+ (500x250x250)



- **In order to obtain high-resolution flow structure, higher-order numerical scheme is essential.**
  - Treatment of discontinuous and smooth region in multiple dimensions
- **From one- to multi-dimensional flows, a limiting process must include to the treatment of vertex to efficiently include a missing flow physics at 1-D based limiting strategy.**
  - Overcome the weakness of dimensional splitting approach, and eliminate spurious oscillations caused by multi-dimensional discontinuity.
- **Characteristics of MLP**
  - Multi-dimensional monotonicity guided by the MLP condition and the discrete maximum principle  $\rightarrow L_\infty$  stability
  - Slopes based on higher-order polynomial reconstruction
  - Computational efficiency comparable to conventional TVD limiters
  - Enhanced accuracy and convergence characteristics