

Instructional workshop on OpenFOAM  
programming  
LECTURE # 7

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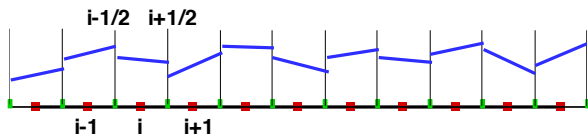
# Outline

*gaussLaplacianScheme* - walk through

Introduction to flux limiters

Limiters in OpenFOAM

# Conservation Laws <sup>1</sup>



$$Q_i^{n+1} = Q_i^n + \frac{\Delta t}{\Delta x} \left( F_{i+\frac{1}{2}}^n - F_{i-\frac{1}{2}}^n \right) \quad (1)$$

- ▶  $Q$  is the cell average value
- ▶  $F_{i+\frac{1}{2}}^n$  is approximation to average flux along  $x = x_{i+\frac{1}{2}}$

$$F_{i+\frac{1}{2}}^n = \mathcal{F} \left( Q_i^n, Q_{i+1}^n \right) \quad (2)$$

## Method of Godunov

- ▶ Reconstruct a piecewise polynomial function  $\tilde{q}^n(x, t_n)$  defined for all  $x$ , from the cell averages  $Q_i^n$ .
- ▶ In the simplest case this is a piecewise constant function that takes the value  $Q_i^n$  in the  $i^{\text{th}}$  grid cell, i.e.,

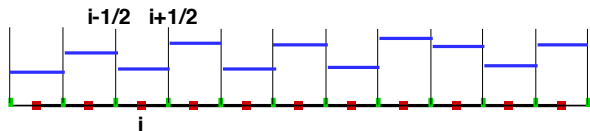
$$\tilde{q}^n(x, t_n) = Q_i^n \quad \text{for all } x \in \mathcal{C}_i \quad (3)$$

- ▶ Evolve the hyperbolic equation exactly (or approximately) with this initial data to obtain  $\tilde{q}^n(x, t_{n+1})$  a time  $\Delta t$  later.
- ▶ Average this function over each grid cell to obtain new cell averages

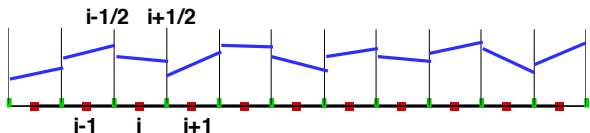
$$Q_i^{n+1} = \frac{1}{\Delta x} \int_{\mathcal{C}_i} \tilde{q}^n(x, t_{n+1}) dx \quad (4)$$

# High resolution finite volume schemes

- ▶ Assuming constant cell variation leads to  $O(\Delta x)$  error



- ▶ Linear variation of values in cell leads to  $O(\Delta x^2)$  error



- ▶ Reconstruct slopes from cell centroid values
- ▶ Extrapolate to faces to get  $L/R$  states

## Piecewise linear reconstruction

- ▶ To achieve better than first-order accuracy need better than piecewise constant function
- ▶ Construct a piecewise linear function using  $Q_i^n$

$$\tilde{q}^n(x, t_{n+1}) = Q_i^n + \sigma_i^n(x - x_i) \quad \text{for } x_{i-\frac{1}{2}} \leq x \leq x_{i+\frac{1}{2}} \quad (5)$$

and

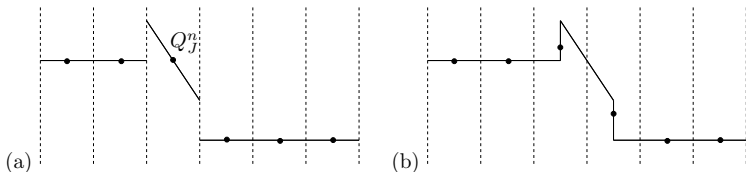
$$x_i = x_{i-\frac{1}{2}} + \frac{1}{2}\Delta x \quad (6)$$

- ▶  $\sigma_i^n$  is the function slope of the  $i^{\text{th}}$  cell

# Problems with linear reconstruction

- ▶ Consider a distribution of  $Q_i^n$  as shown below ( $J$  is some arbitrary cell)

$$Q_i^n = \begin{cases} 1 & \text{if } i \leq J \\ 0 & \text{if } i > J \end{cases} \quad (7)$$



- (a) Linear reconstruction for  $f(x)$  using cell averages  $Q_i^n$
- (b) Simple linear advection of reconstructed values to  $t_{n+1}$

Under/Overshoot in solution

## The remedy

- ▶ Introduce a function  $\phi(\theta_i^n)$  to limit the slope of the function as shown below,

$$\tilde{q}^n(x, t_{n+1}) = Q_i^n + \underbrace{\phi(\theta_i^n)}_{\text{limiter}} \underbrace{\sigma_i^n}_{\text{slope}} (x - x_i) \quad \text{for } x_{i-\frac{1}{2}} \leq x \leq x_{i+\frac{1}{2}} \quad (8)$$

- ▶  $\theta_i^n$  is a measure of the variation of the function  $Q_i^n$  in cell  $i$
- ▶  $\phi = 0$  is piecewise constant reconstruction
- ▶  $\phi = 1$  is piecewise linear reconstruction
- ▶  $0 < \phi < 1$  is piecewise linear reconstruction (with loss of accuracy)



## Variation measure $\theta_i^n$

- ▶ Many ways to obtain  $\theta$
- ▶ Implementation dependent function
- ▶ An example upwind version

$$\theta_i^n = \frac{\Delta Q_{l-\frac{1}{2}}^n}{\Delta Q_{i-\frac{1}{2}}^n} \quad (9)$$

where,

$$\Delta Q_{i-\frac{1}{2}}^n = Q_i^n - Q_{i-1}^n \quad (10)$$

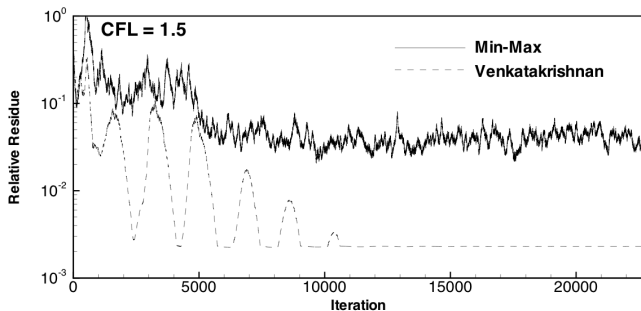
and

$$l = \begin{cases} i-1 & \text{if } a > 0 \\ i+1 & \text{if } a < 0 \end{cases} \quad (11)$$

where,  $a$  is the wave speed (advection)

## Limiter function $\phi^2$

- ▶ Many choices available and no-universal choice
- ▶ Can be differentiable or non-differentiable
- ▶ Differentiable limiter = smoother convergence



- ▶ Min-Max is non-differentiable
- ▶ Venkatakrishnan is differentiable

## Higher than 2<sup>nd</sup> order

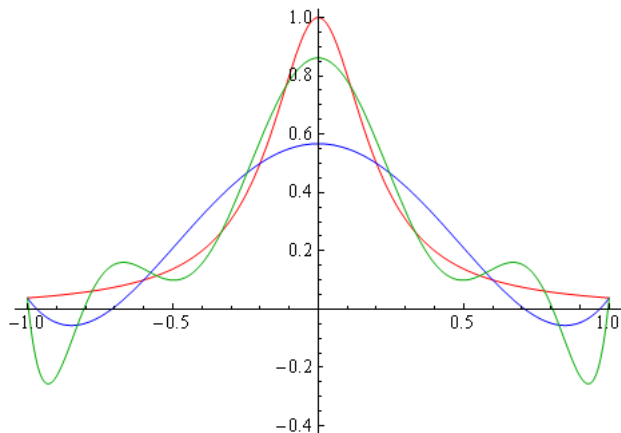
- ▶ Is it possible to go higher than second order ?
- ▶ Can we use higher-order ( $> 1$ ) polynomials to reconstruct ?
- ▶ There are two main problems discussed in the next two slides

# Curse of polynomial interpolation

Fit a polynomial over functions

- ▶ Runge phenomena

$$f(x) = \frac{1}{1+x^2} \text{ for } -1 \leq x \leq 1 \quad (12)$$

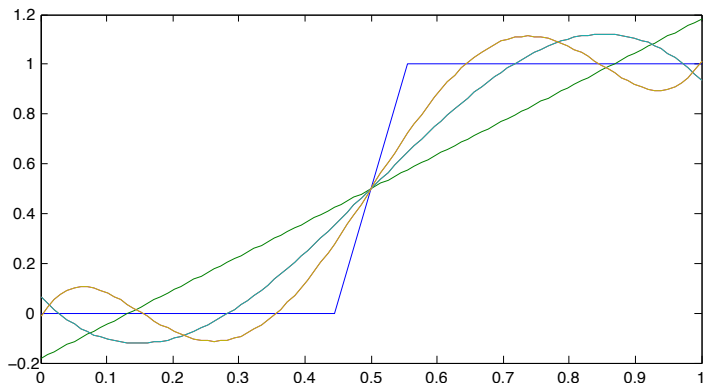


# Curse of polynomial interpolation

Fit a polynomial over functions

- ▶ Gibbs oscillation

$$f(x) = \begin{cases} 0 & \text{if } x \leq 0.5 \\ 1 & \text{if } x > 0.5 \end{cases} \quad \text{for } 0 \leq x \leq 1 \quad (13)$$



# Higher order polynomial

At steep gradients

- ▶ Suffer from under-shoot and over-shoot
- ▶ Violation of bounds
- ▶ Monotonic solution can become non-monotonic

Remedy

- ▶ Use orthogonal polynomial like Chebyshev, Legendre, etc
- ▶ ENO or WENO type limited polynomials

Beyond the scope of OpenFOAM

# OpenFOAM: Limited gradient schemes

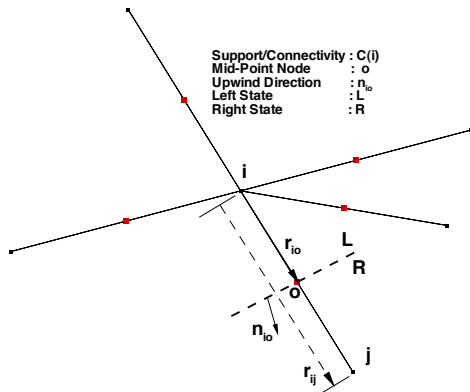
- ▶ cellLimited
- ▶ cellMDLimited
- ▶ faceLimited
- ▶ faceMDLimited

Source files located in

```
$FOAM_SRC
|__ /finiteVolume
  |__ /finiteVolume
    |__ /gradSchemes
      |__ /limitedGradSchemes
        |__ cellLimitedGrad
        |__ cellMDLimitedGrad
        |__ faceLimitedGrad
        |__ faceMDLimitedGrad
```

## cellLimited gradient scheme - Some notations

- ▶ Let the face neighbouring cells of cell  $i$  be  $j \in C_{ij}$
- ▶ Let the faces of cell  $i$  be  $o \in C_{io}$
- ▶  $r_{io}$  is the line joining cell  $i$ 's centroid to its face centroids
- ▶  $r_{ij}$  is the line joining cell  $i$ 's centroid to its face neighbouring cell  $j$ 's centroids





## cellLimited gradient scheme

- ▶ Find maximum and minimum values of all cell (face) neighbours ( $C_{ij}$ )

$$q_i^{max} = \max_{j \in C_{ij}} q_{ij} \quad (14)$$

$$q_i^{min} = \min_{j \in C_{ij}} q_{ij} \quad (15)$$

- ▶  $ij$  is the cell  $i$ 's  $j^{th}$  face-cell neighbour
- ▶  $io$  is the cell  $i$ 's  $o^{th}$  face

## cellLimited gradient scheme

- ▶ Subtract the max/min values from the actual cell values

$$\Delta q_i^{max} = q_i - q_i^{max} \text{ for } i \in \mathcal{C}_i \quad (16)$$

$$\Delta q_i^{min} = q_i - q_i^{min} \text{ for } i \in \mathcal{C}_i \quad (17)$$

- ▶ An input parameter  $\kappa$  is used to adjust the  $\Delta q$  as follows,

$$\Delta \tilde{q}_i^{max} = \Delta q_i^{max} + \left( \frac{1 - \kappa}{\kappa} \right) (\Delta q_i^{max} - \Delta q_i^{min}) \quad (18)$$

$$\Delta \tilde{q}_i^{min} = \Delta q_i^{min} - \left( \frac{1 - \kappa}{\kappa} \right) (\Delta q_i^{max} - \Delta q_i^{min}) \quad (19)$$

$$(20)$$

- ▶  $\kappa = 0$  is no-limiting
- ▶  $\kappa = 1$  is full limiting

## cellLimited gradient scheme

- ▶ Calculate gradient  $\nabla q_i$  using a suitable *gradScheme*
- ▶ Set the cell limiter values  $\phi_i$  as follows, (loop over all faces of cell)

$$\phi_i = \begin{cases} \min \left[ 1, \min_{o \in C_{io}} \left( \frac{\Delta \tilde{q}_i^{max}}{\nabla q_i \cdot r_{io}} \right) \right] & \text{if } \Delta \tilde{q}_i^{max} < \nabla q_i \cdot r_{io} \\ \min \left[ 1, \min_{o \in C_{io}} \left( \frac{\Delta \tilde{q}_i^{min}}{\nabla q_i \cdot r_{io}} \right) \right] & \text{if } \Delta \tilde{q}_i^{min} > \nabla q_i \cdot r_{io} \end{cases} \quad (21)$$

- ▶ Now the limited gradient value is  $\phi_i \nabla q_i$

This is the min/max limiter and remember the same  $\phi_i$  value is used for all components of  $\nabla q_i$

## cellMDLimited gradient scheme

- ▶ Exactly similar to the *cellLimited* version with the exception that we now calculate the limiter value for each row of a gradient tensor

$$+ \sum_{o \in C_{io}} \frac{\mathbf{r}_{io}}{|\mathbf{r}_{io}|} \left[ \begin{array}{l} \Delta \tilde{q}_i^{max} - \nabla q_i[n] \cdot r_{io} \quad \text{if } \Delta \tilde{q}_i^{max} < \nabla q_i[n] \cdot r_{io} \\ \Delta \tilde{q}_i^{min} - \nabla q_i[n] \cdot r_{io} \quad \text{if } \Delta \tilde{q}_i^{min} > \nabla q_i[n] \cdot r_{io} \end{array} \right] \quad (22)$$

for rows  $n = 1, 2, 3$

- ▶ Now the limited gradient value is calculated in-place in  $\nabla q_i$

## faceLimited gradient scheme

- ▶ Max/min values calculated using local face owner/neighbour cell values

$$q_{io}^{max} = \max(q_{io}^{own}, q_{io}^{nei}) \quad (23)$$

$$q_{io}^{min} = \min(q_{io}^{own}, q_{io}^{nei}) \quad (24)$$

- ▶ Correct max/min face values using input parameter  $\kappa$

$$\tilde{q}_{io}^{max} = q_{io}^{max} + \left( \frac{1 - \kappa}{\kappa} \right) (q_{io}^{max} - q_{io}^{min}) \quad (25)$$

$$\tilde{q}_{io}^{min} = q_{io}^{min} - \left( \frac{1 - \kappa}{\kappa} \right) (q_{io}^{max} - q_{io}^{min}) \quad (26)$$

## faceLimited gradient scheme

- ▶ Let  $\nabla q_i$  be the gradient obtained from *gradScheme*
- ▶ Limiter function  $\phi_i$  is calculated as follows,

$$\phi_i = \begin{cases} \min \left[ 1, \min_{o \in C_{io}} \left( \frac{\Delta q_{io}^{max}}{\nabla q_i \cdot r_{io}} \right) \right] & \text{if } \Delta q_{io}^{max} < \nabla q_i \cdot r_{io} \\ \min \left[ 1, \min_{o \in C_{io}} \left( \frac{\Delta q_{io}^{min}}{\nabla q_i \cdot r_{io}} \right) \right] & \text{if } \Delta q_{io}^{min} > \nabla q_i \cdot r_{io} \end{cases} \quad (27)$$

where,  $\Delta q_{io}^{max/min} = q_i - \tilde{q}_{io}^{max/min}$

- ▶ Now the limited gradient value is  $\phi_i \nabla q_i$

More dissipative than cellLimited version but cheaper computationally (2X faster)

## faceMDLimited gradient scheme

- ▶ Multidimensional version limiting is for gradient tensors

$$+ \sum_{o \in C_{io}} \frac{\mathbf{r}_{io}}{|\mathbf{r}_{io}|} \left[ \begin{array}{l} \Delta q_{io}^{max} - \nabla q_i[n] \cdot r_{io} \quad \text{if } \Delta q_{io}^{max} < \nabla q_i[n] \cdot r_{io} \\ \Delta q_{io}^{min} - \nabla q_i[n] \cdot r_{io} \quad \text{if } \Delta q_{io}^{min} > \nabla q_i[n] \cdot r_{io} \end{array} \right] \quad (28)$$

*for rows  $n = 1, 2, 3$*

- ▶ Limited gradient value calculated in-place  $\nabla q_i$

End of Week 3 Day 2