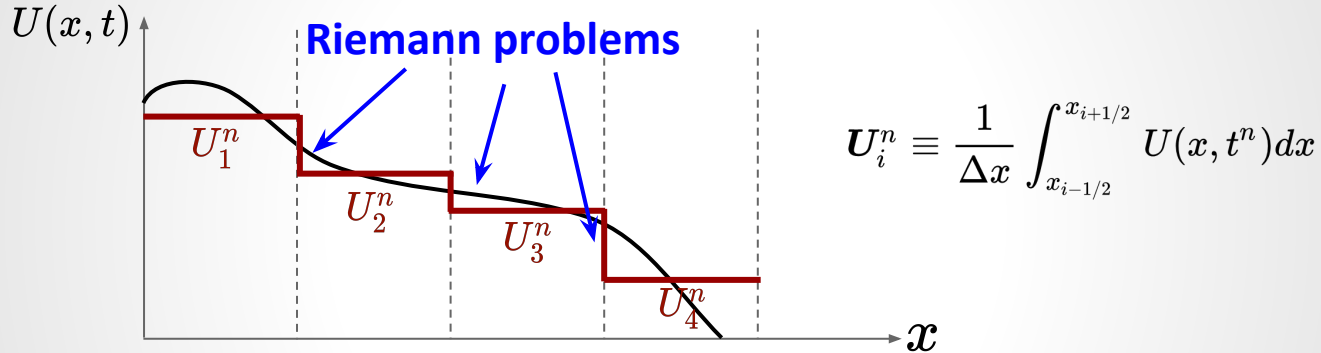


# Hydrodynamics-II

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# High-Resolution Shock-Capturing Methods

- Godunov method
  - Approximate data with a piecewise constant distribution

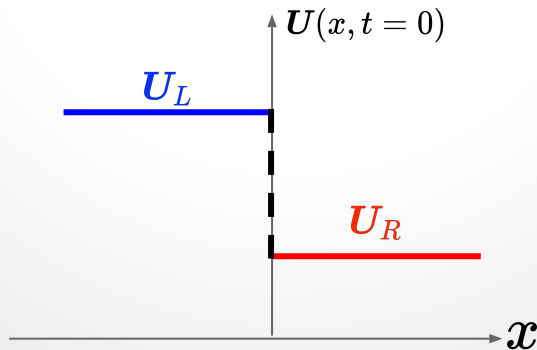


- Solve the local Riemann problems
  - Piecewise constant data with a single discontinuity
  - Apply either exact or approximate solutions
- Update data by averaging the Riemann problem solution over each cell
  - Equivalently, we can solve the intercell fluxes
  - Avoid wave interaction within each cell

# Riemann Problem in 1D Hydro

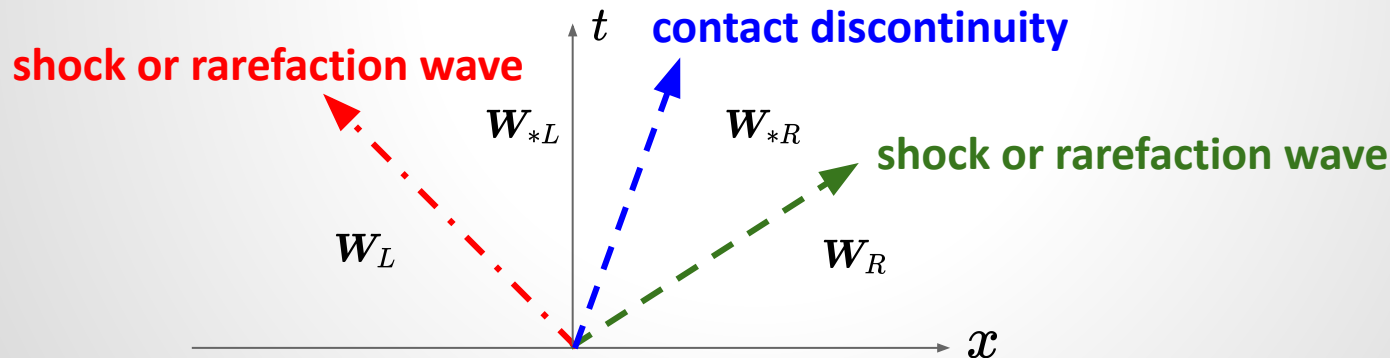
- Euler eqs. in 1D:  $\frac{\partial U}{\partial t} + \frac{\partial F_x(U)}{\partial x} = 0$ ,  $U = \begin{bmatrix} \rho \\ \rho v_x \\ E \end{bmatrix}$ ,  $F_x = \begin{bmatrix} \rho v_x \\ \rho v_x^2 + P \\ (E + P)v_x \end{bmatrix}$

- Riemann problem:  $U(x, t = 0) = \begin{cases} U_L = \begin{bmatrix} \rho_L \\ \rho_L v_{xL} \\ E_L \end{bmatrix}, & x \leq 0 \\ U_R = \begin{bmatrix} \rho_R \\ \rho_R v_{xR} \\ E_R \end{bmatrix}, & x > 0 \end{cases}$  left state  
right state



# Riemann Problem in 1D Hydro

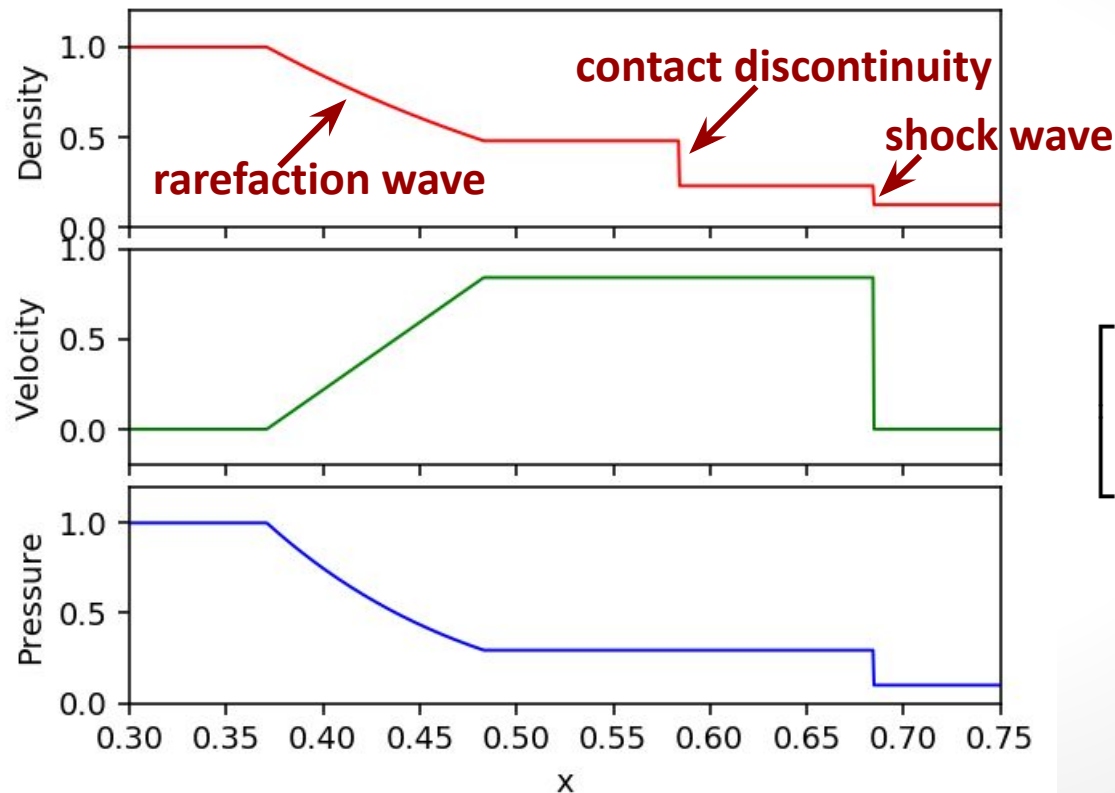
- Exact solution of the Riemann problem involves three waves
  - Contact discontinuity
  - Shock wave
  - Rarefaction wave
- Decompose the entire domain into four regions  $W_L, W_{*L}, W_{*R}, W_R$



# Riemann Problem in 1D Hydro

- Riemann problem can be solved analytically
  - Known:  $W_L, W_R$
  - Unknowns:  $W_{*L}, W_{*R}$ 
    - In fact, we always have  $P_{*L} = P_{*R}$  and  $v_{x,*L} = v_{x,*R}$  (because the middle wave is always a contact discontinuity)
    - So only 4 unknown variables:  $\rho_{*L}, \rho_{*R}, P_*, v_{x*}$
- However, exact Riemann solver is very computationally expensive
  - Approximate Riemann solvers are usually accurate enough
    - All we need is the interface fluxes
    - Examples
      - Roe solver
      - HLLE solver
      - HLLC solver

# Example: Sod Shock Tube Problem



Initial condition

Left state

Right state

$$\begin{bmatrix} \rho_L \\ v_L \\ P_L \end{bmatrix} = \begin{bmatrix} 1.0 \\ 0.0 \\ 0.125 \end{bmatrix}, \quad \begin{bmatrix} \rho_R \\ v_R \\ P_R \end{bmatrix} = \begin{bmatrix} 0.125 \\ 0.0 \\ 0.1 \end{bmatrix}$$

# Diagonalization of a 1D Linear System

$$\frac{\partial U}{\partial t} + A \frac{\partial U}{\partial x} = 0, \text{ where } A \text{ is a constant}$$

Diagonalize  $A$ :  $\lambda = K^{-1} A K$

$$\lambda = \begin{bmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ 0 & 0 & \lambda_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_N \end{bmatrix}, \quad A K = K \lambda, \quad K = \begin{bmatrix} K^{(1)} & \dots & K^{(N)} \end{bmatrix}$$

right eigenvectors (constant)

eigenvalues (constant)

# Diagonalization of a 1D Linear System

Introduce the characteristic variables:  $C \equiv K^{-1}U$

$$K \frac{\partial C}{\partial t} + AK \frac{\partial C}{\partial x} = 0 \rightarrow \frac{\partial C}{\partial t} + \lambda \frac{\partial C}{\partial x} = 0$$
$$\rightarrow \boxed{\frac{\partial C_i}{\partial t} + \lambda_i \frac{\partial C_i}{\partial x} = 0, \quad i = 1, \dots, N}$$

**N decoupled linear advection equations  
with characteristic speed  $\lambda_i$**

$$U(x, t) = KC = \sum_{i=1}^N C_i(x, t) K^{(i)} \rightarrow C_i(x, t) \text{ is the coefficient in the eigenvector expansion of } U(x, t)$$



# Riemann Problem for Linearized Hydro

$$\begin{cases} \frac{\partial \rho}{\partial t} + \rho_0 \frac{\partial v_x}{\partial x} = 0 \\ \frac{\partial v_x}{\partial t} + \frac{C_s^2}{\rho_0} \frac{\partial \rho}{\partial x} = 0 \end{cases}, \begin{bmatrix} \rho \\ v_x \end{bmatrix} = \begin{bmatrix} \rho_L \\ v_{xL} \end{bmatrix} \text{ for } x \leq 0, \begin{bmatrix} \rho \\ v_x \end{bmatrix} = \begin{bmatrix} \rho_R \\ v_{xR} \end{bmatrix} \text{ for } x > 0$$

$$\Rightarrow \frac{\partial \mathbf{U}}{\partial t} + \mathbf{A} \frac{\partial \mathbf{U}}{\partial x} = 0, \mathbf{U} = \begin{bmatrix} \rho \\ v_x \end{bmatrix}, \mathbf{A} = \begin{bmatrix} 0 & \rho_0 \\ C_s^2/\rho_0 & 0 \end{bmatrix}$$

$$\Rightarrow \boldsymbol{\lambda} = \begin{bmatrix} -C_s & 0 \\ 0 & C_s \end{bmatrix}, \mathbf{K} = \begin{bmatrix} \rho_0 & \rho_0 \\ -C_s & C_s \end{bmatrix}, \mathbf{K}^{-1} = \frac{1}{2C_s\rho_0} \begin{bmatrix} C_s & -\rho_0 \\ C_s & \rho_0 \end{bmatrix}$$

# Riemann Problem for Linearized Hydro

left-moving component with  $-C_s$



$$\mathbf{C} = \mathbf{K}^{-1}\mathbf{U} = \frac{1}{2C_s\rho_0} \begin{bmatrix} C_s\rho - \rho_0 v_x \\ C_s\rho + \rho_0 v_x \end{bmatrix} \equiv \begin{bmatrix} C_1 \\ C_2 \end{bmatrix}$$



right-moving component with  $C_s$

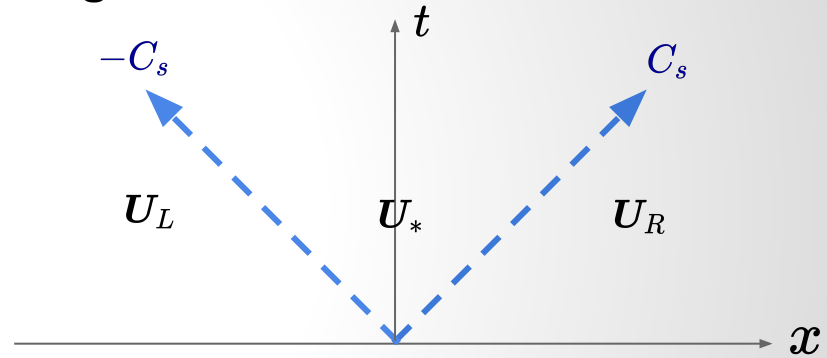
➔

$$\begin{cases} \mathbf{U}_L = \begin{bmatrix} \rho_L \\ v_{xL} \end{bmatrix} = C_{1,L}\mathbf{K}^{(1)} + \underline{C_{2,L}\mathbf{K}^{(2)}} \\ \mathbf{U}_R = \begin{bmatrix} \rho_R \\ v_{xR} \end{bmatrix} = \underline{C_{1,R}\mathbf{K}^{(1)}} + C_{2,R}\mathbf{K}^{(2)} \end{cases}$$

# Riemann Problem for Linearized Hydro

- The two characteristic waves (propagating with speeds  $\pm C_s$ ) decompose the solution into three regions

➔ 
$$U(x, t) = \begin{cases} U_L, & x < -C_s t \\ U_R, & x > C_s t \\ U_*, & |x| < C_s t \end{cases}$$



$$\begin{aligned} U_* &= C_{1,R} \mathbf{K}^{(1)} + C_{2,L} \mathbf{K}^{(2)} \\ &= \begin{bmatrix} \frac{1}{2}(\rho_L + \rho_R) - \frac{\rho_0}{2C_s}(v_{xR} - v_{xL}) \\ \frac{1}{2}(v_{xL} + v_{xR}) - \frac{C_s}{2\rho_0}(\rho_R - \rho_L) \end{bmatrix} \end{aligned}$$

**solution in the star region in between  
two characteristic waves**

# Roe's Riemann Solver

- Roe, P. L., 1981. *JCP*, 43, 357
- Rewrite the 1D Euler eqs. into a matrix form

$$\frac{\partial U}{\partial t} + \frac{\partial F_x(U)}{\partial x} = 0, \quad A(U) \equiv \frac{\partial F_x}{\partial U}$$
$$\rightarrow \frac{\partial U}{\partial t} + A(U) \frac{\partial U}{\partial x} = 0$$

$$U = \begin{bmatrix} \rho \\ \rho v_x \\ \rho v_y \\ \rho v_z \\ E \end{bmatrix}, \quad A(U) = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ -v_x^2 + \hat{\gamma} v^2 / 2 & (2 - \hat{\gamma}) v_x & -\hat{\gamma} v_y & -\hat{\gamma} v_z & \hat{\gamma} \\ -v_x v_y & v_y & v_x & 0 & 0 \\ -v_x v_z & v_z & 0 & v_x & 0 \\ -v_x H + \hat{\gamma} v_x v^2 / 2 & H - \hat{\gamma} v_x^2 & -\hat{\gamma} v_x v_y & \hat{\gamma} v_x v_z & (\hat{\gamma} + 1) v_x \end{bmatrix}$$

# Roe's Riemann Solver

- **Diagonalize**  $A(U)$

$$\lambda = [v_x - C_s, v_x, v_x, v_x, v_x + C_s]$$

$$K = \begin{bmatrix} 1 & 0 & 0 & 1 & 1 \\ v_x - C_s & 0 & 0 & v_x & v_x + C_s \\ v_y & 1 & 0 & v_y & v_y \\ v_z & 0 & 1 & v_z & v_z \\ H - v_x C_s & v_y & v_z & v^2/2 & H + v_x C_s \end{bmatrix}$$

$$H = (E + P)/\rho, \quad \hat{\gamma} = \gamma - 1, \quad v^2 = v_x^2 + v_y^2 + v_z^2, \quad C_s^2 = \gamma P/\rho$$

# Roe's Riemann Solver

- Approximate  $A(U)$  by a **constant** Jacobian matrix  $A(\bar{U})$ , where  $\bar{U} = \bar{U}(U_L, U_R)$  is a constant mean state between the left and right states

$$\frac{\partial U}{\partial t} + \tilde{A}(U_L, U_R) \frac{\partial U}{\partial x} = 0 \leftarrow \text{approximate (linearized) equations}$$

- Finding an appropriate form of  $\bar{U}(U_L, U_R)$  is non-trivial
  - Roe proposed the following linearization

$$\begin{cases} \bar{\rho} = \sqrt{\rho_L} \sqrt{\rho_R} \\ \bar{\mathbf{v}} = \frac{\sqrt{\rho_L} \mathbf{v}_L + \sqrt{\rho_R} \mathbf{v}_R}{\sqrt{\rho_L} + \sqrt{\rho_R}} \\ \bar{H} = \frac{\sqrt{\rho_L} H_L + \sqrt{\rho_R} H_R}{\sqrt{\rho_L} + \sqrt{\rho_R}} \end{cases}$$



- Ensure conservation:  $F(U_R) - F(U_L) = \tilde{A} \cdot (U_R - U_L)$
- Next, compute the averaged eigenvalues  $\bar{\lambda}$  and the averaged right eigenvectors  $\bar{K}$

# Roe's Riemann Solver

- Compute the Roe averaged fluxes  $F_{\text{Roe}}$

$$\bar{C} = \bar{K}^{-1}(U_R - U_L)$$

$$F_{\text{Roe}} = F_L + \sum_{\bar{\lambda}_i \leq 0} \bar{C}_i \bar{\lambda}_i \bar{K}^{(i)}$$

$$= F_R - \sum_{\bar{\lambda}_i \geq 0} \bar{C}_i \bar{\lambda}_i \bar{K}^{(i)}$$

$$= \frac{1}{2}(F_L + F_R) - \frac{1}{2} \sum_{\bar{\lambda}_i} \bar{C}_i |\bar{\lambda}_i| \bar{K}^{(i)}$$

- Procedure summary:
  - Compute the Roe average values of primitive variables:  $\bar{\rho}, \bar{v}, \bar{H}$
  - Compute the averaged eigenvalues and eigenvectors:  $\bar{\lambda}, \bar{K}$
  - Compute the coefficients in the eigenvector expansion:  $\bar{C}$
  - Compute the Roe flux:  $\bar{F}_{\text{Roe}}$

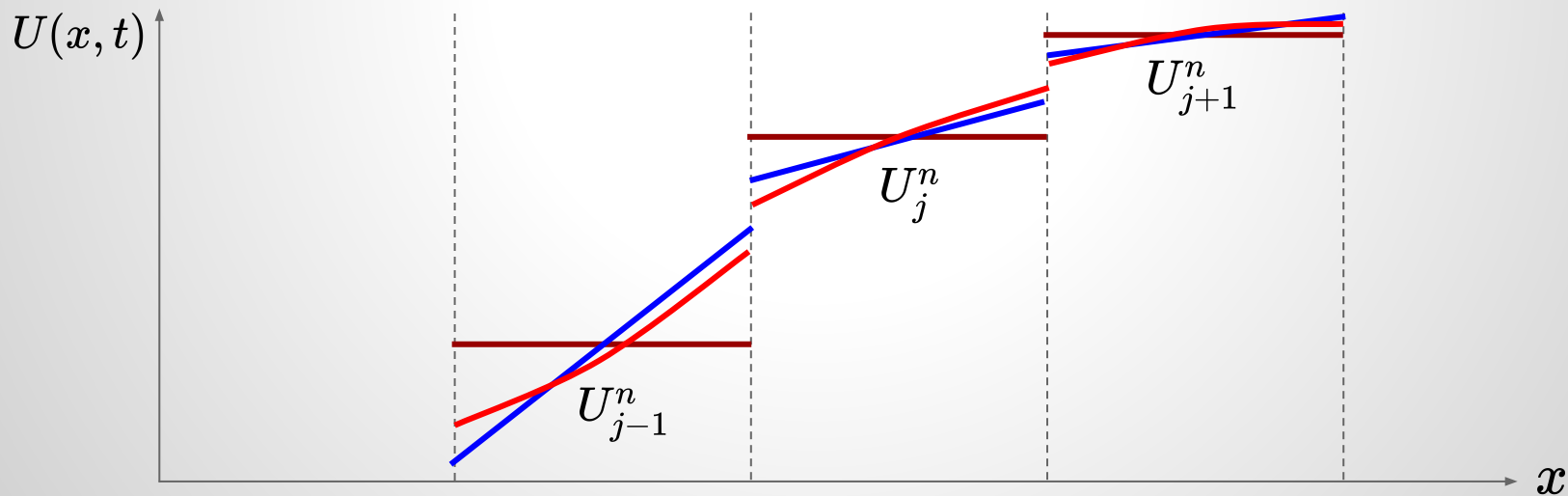
# Other Riemann Solvers

- HLL-like solvers solve approximate solutions to the original nonlinear equations
  - HLLE: *Einfeldt, et al., 1991. Comp. Phys., 92, 273*
  - HLLC: *Toro, et al., 1994. Shock Waves, 4:25–34*
- In comparison, Roe solver solves exact solutions to the approximate (linearized) equations



# Higher-Order Godunov Methods

- **MUSCL** (*Monotone Upstream-centred Scheme for Conservation Laws*)
- Data reconstruction within each cell
  - Original Godunov's scheme: piecewise constant method (**PCM**)
  - Piecewise linear method (**PLM**)
  - Piecewise parabolic method (**PPM**)



# Higher-Order Godunov Methods

- **Avoid introducing new local extrema during data reconstruction**
  - Reduce spurious (i.e., unphysical) oscillations
  - Avoid unphysical values such as negative density/pressure

- **Slope limiters**

- $U_j(x) = U_j + \frac{(x - x_j)}{\Delta x} \bar{\delta}_i, \quad |x - x_j| \leq \Delta x/2$

where  $\bar{\delta}_i = \bar{\delta}_i(\delta_{i-1/2}, \delta_{i+1/2}), \quad \delta_{i-1/2} \equiv U_i - U_{i-1}$

 **limited slope satisfying the TVD (Total Variation Diminishing) condition**

- **Examples:** van Leer:  $\bar{\delta}_i = \begin{cases} \frac{2\delta_{i-1/2}\delta_{i+1/2}}{\delta_{i-1/2} + \delta_{i+1/2}}, & \delta_{i-1/2}\delta_{i+1/2} \geq 0 \\ 0, & \delta_{i-1/2}\delta_{i+1/2} < 0 \end{cases}$

MinMod:  $\bar{\delta}_i = \begin{cases} \text{sign}(\delta_{i-1/2}) \min(|\delta_{i-1/2}|, |\delta_{i+1/2}|), & \delta_{i-1/2}\delta_{i+1/2} \geq 0 \\ 0, & \delta_{i-1/2}\delta_{i+1/2} < 0 \end{cases}$

# Higher-Order Godunov Methods

- **Effects of various slope limiters**
  - Resolution (diffusiveness) vs. robustness
- **Left and right states are not equal unless the flow is smooth**
  - Define Riemann problems
- **Data reconstruction on the primitive variables usually results in better results (less oscillatory) than on the conserved variables**
  - It may be even better to reconstruct the characteristic variables
    - Diagonalize the linearized eqs. of motion in the primitive variables
    - Determine eigenvectors
    - Perform eigen-decomposition on  $\delta_{i-1/2}$  and  $\delta_{i+1/2}$  to get the characteristic variables
    - Compute limited slopes on these characteristic variables

# Second-Order Accuracy in Time

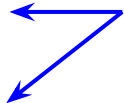
Example: **MUSCL-Hancock** scheme

1. Data reconstruction  $\rightarrow$  obtain the face-centered data (i.e., data on the left and right edges of each cell) at  $t^n$

$$U_{i,L}^n = U_i^n - \frac{1}{2}\bar{\delta}_i, \quad U_{i,R}^n = U_i^n + \frac{1}{2}\bar{\delta}_i$$

2. Evolve the face-centered data by  $\Delta t/2$  using

$$U_{i,L}^{n+1/2} = U_{i,L}^n - \frac{\Delta t}{2\Delta x} \left[ F_x(U_{i,R}^n) - F_x(U_{i,L}^n) \right]$$
$$U_{i,R}^{n+1/2} = U_{i,R}^n - \frac{\Delta t}{2\Delta x} \left[ F_x(U_{i,R}^n) - F_x(U_{i,L}^n) \right]$$

 exactly the same fluxes;  
no ghost zones are required

3. Riemann solver  $\rightarrow$  compute the inter-cell fluxes

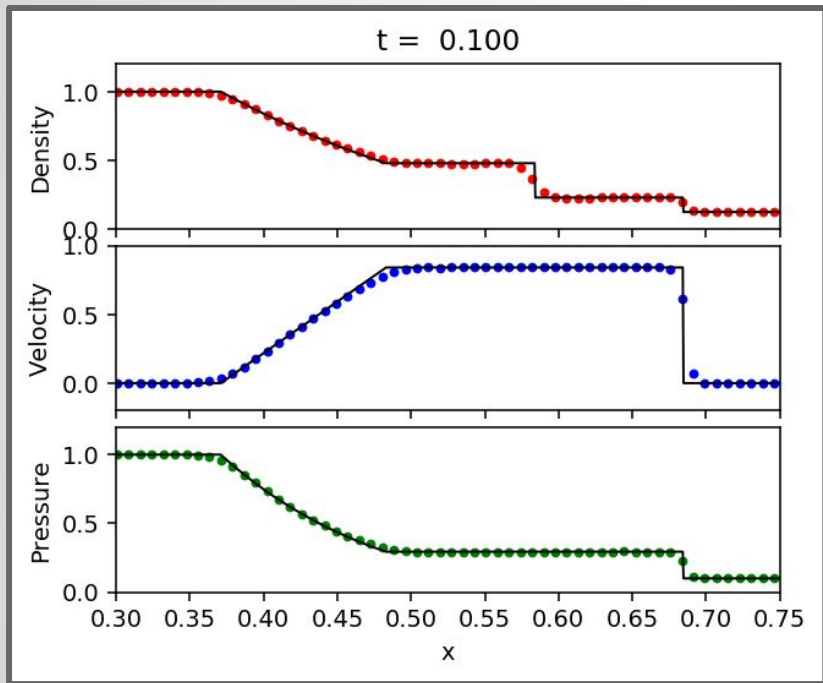
$$F_{x,i-1/2}^{n+1/2} = \text{Riemann}(U_L, U_R), \text{ where } U_L = U_{i-1,R}^{n+1/2} \text{ and } U_R = U_{i,L}^{n+1/2}$$

4. Evolve the volume-averaged data by  $\Delta t$

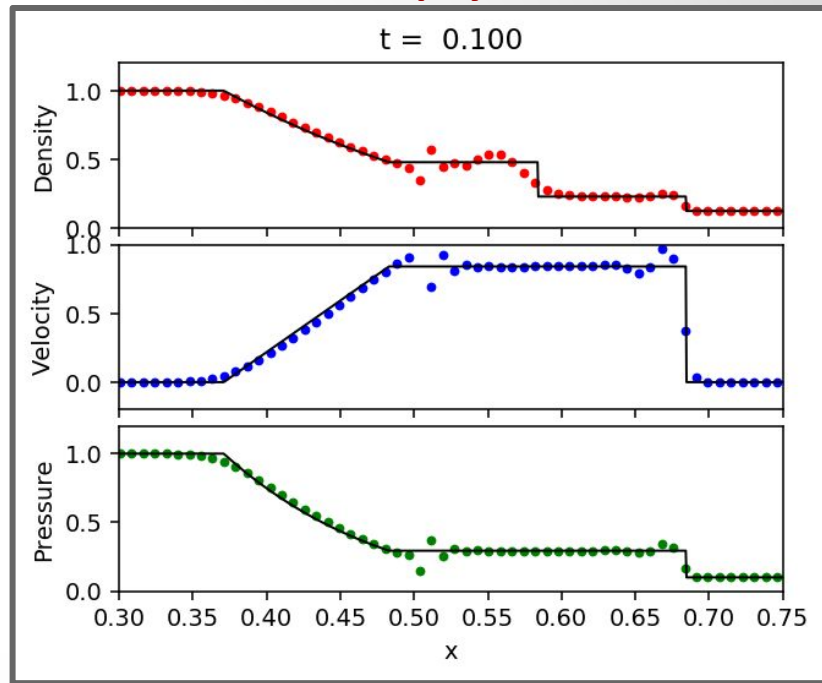
$$U_i^{n+1} = U_i^n - \frac{\Delta t}{\Delta x} \left[ F_{x,i+1/2}^{n+1/2} - F_{x,i-1/2}^{n+1/2} \right]$$

# Sod Shock Tube with MUSCL-Hancock

MUSCL-Hancock → **much better!**



Lax-Wendroff → **unphysical oscillations...**



# Demo: MUSCL-Hancock Scheme

```
def ComputePressure( d, px, py, pz, e ):
```

```
def ComputeTimestep( U ):
```

```
def ComputeLimitedSlope( L, C, R ):
```

```
def Conserved2Primitive( U ):
```

```
def Primitive2Conserved( W ):
```

```
def DataReconstruction_PLM( U ):
```

```
def Conserved2Flux( U ):
```

```
def Roe( L, R ):
```

# Demo: MUSCL-Hancock Scheme

```
def update( frame ):
    # set the boundary conditions
    BoundaryCondition( U )

    # estimate time-step from the CFL condition
    dt = ComputeTimestep( U )

    # data reconstruction
    L, R = DataReconstruction_PLM( U )

    # update the face-centered variables by  $0.5*dt$ 
    for j in range( 1, N-1 ):
        flux_L, flux_R = Conserved2Flux( L[j] ), Conserved2Flux( R[j] )
        dflux =  $0.5*dt/dx$ *( flux_R - flux_L )
        L[j] -= dflux
        R[j] -= dflux
```

# Demo: MUSCL-Hancock Scheme

```
def update( frame ):
    ...

# compute fluxes
flux = np.empty( (N,5) )
for j in range( nghost, N-nghost+1 ):
    flux[j] = Roe( R[j-1], L[j] )

# update the cell-centered input variables by dt
U[nghost:N-nghost] -= dt/dx*( flux[nghost+1:N-nghost+1] - \
                                flux[nghost:N-nghost] )
```



Run **lec03-demo03/04.py**

- Compare with demo01 & 02
- Find a Riemann problem that crashes the code!