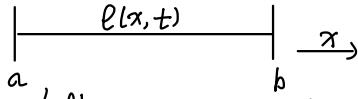


Lecture 01

Conservation laws



$$\frac{d}{dt} \int_a^b l(x, t) dx = f_b - f_a$$

assumption: $f_b = f(l(b, t))$ (There are also other assumptions).

$f_b = f(l(b, t))$: conservation laws, inviscid

$f_b = f(l(b, t), l_x(b, t))$: NS, viscous

$$\frac{d}{dt} \int_a^b l(x, t) dx + f(l(b, t)) - f(l(a, t)) = 0$$

Assume l is smooth (C').

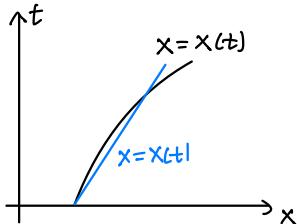
$$\int_a^b l_t(x, t) dx + \int_a^b f(l)_x dx = 0 \Rightarrow \int_a^b l_t + f(l)_x dx = 0$$

f is usually smooth $\Rightarrow l_t + f(l)_x = 0$ p.t. nice

$$u_t + f(u)_x = 0 \quad f(u) = u \Rightarrow u_t + u_x = 0$$

$$f(u) = \frac{1}{2}u^2 \Rightarrow u_t + (\frac{u^2}{2})_x = 0 \quad \text{Burgers' equation}$$

Characteristics



$x = x(t)$ is the solution to the characteristic ODE:

$$\begin{cases} x'(t) = f'(u(x(t), t)) \\ x(0) = x^0 \end{cases}$$

The solution along characteristics is const.

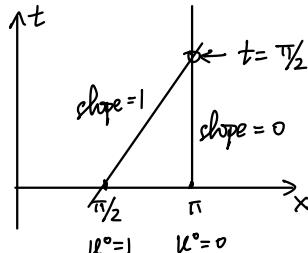
$$\frac{dx}{dt} u(x(t), t) = u_x \cdot x'(t) + u_t = u_x f'(u(x(t), t)) + u_t = u_t + f(u(x(t), t))_x = 0.$$

$u(x(t), t) = u(x(0), 0)$ is a constant. And $x'(t) = f'(u(x(0), 0)) = \text{const.}$

So $x'(t)$ is in fact a straight line.

Burgers equation

$$\begin{cases} u_t + (\frac{u^2}{2})_x = 0 & 0 \leq x \leq 2\pi \\ u(x, 0) = \sin x. \text{ periodic} \end{cases}$$

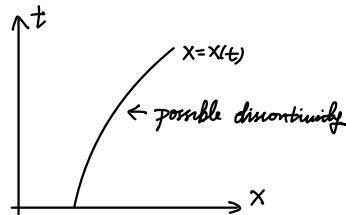


We assumed that u is differentiable.

But this assumption is wrong.

Weak solutions

$$\begin{aligned}
 0 &= \frac{d}{dt} \int_a^b u(x, t) dx + f(u(b, t)) - f(u(a, t)) \\
 &= \frac{d}{dt} \left(\int_a^{x(t)} u(x, t) dx + \int_{x(t)}^b u(x, t) dx \right) + f(u(b, t)) - f(u(a, t)) \\
 &= u(x(t)^-, t) x'(t) + \int_a^{x(t)} u_x(x, t) dx \\
 &\quad - u(x(t^+, t)) x'(t) + \int_{x(t)}^b u_x(x, t) dx \\
 &\quad + f(u(x(t)^-, t)) - f(u(a, t)) \quad \text{Claim: } u_t + f(u)_x = 0 \text{ on either left or right side of } x(t) \\
 &= x'(t) [u(x(t)^-, t) - u(x(t)^+, t)] - \int_a^{x(t)} f(u(x, t), t) dx - \int_{x(t)}^b f(u(x, t), t) dx \\
 &\quad + f(u(x(t)^-, t)) - f(u(a, t)) \\
 &= x'(t) [u(x(t)^-, t) - u(x(t)^+, t)] - (f(u(x(t)^-, t)) - f(u(a, t))) - (f(u(x(t)^+, t)) - f(u(x(t)^+, t))) \\
 &= x'(t) [u(x(t)^-, t) - u(x(t)^+, t)] - f(u(x(t)^-, t)) + f(u(x(t)^+, t)).
 \end{aligned}$$



Rankine-Hugoniot jump condition. $f(u^+) - f(u^-) = x'(t)(u^+ - u^-)$.

Def 1: A p.w. smooth function $u(x, t)$ is a weak solution iff it is ⁽¹⁾ smooth and satisfies the PDE p.t. wise except a few smooth curves, at which the R-H jump condition is satisfied. Practical but not good mathematically

Def 2: $u(x, t)$ is a weak solution iff $\forall t_1, t_2$,

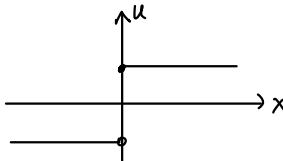
$$\int_a^b u(x, t_2) dx - \int_a^b u(x, t_1) dx + \int_{t_1}^{t_2} f(u(b, t)) dt - \int_{t_1}^{t_2} f(u(a, t)) dt = 0$$

$$\begin{aligned}
 \int_0^\infty \int_{-\infty}^{t+\infty} \varphi \cdot [u_t + f(u)_x] dx dt &= 0 \quad \text{assume } \varphi \in C_0^1(\mathbb{R}^2), \\
 -\int_0^\infty \int_{-\infty}^{t+\infty} \varphi_t u + \varphi_x \cdot f(u) dx dt + \int_{-\infty}^{t+\infty} \varphi u dx \Big|_0^\infty + \int_0^\infty \varphi f(u) dt \Big|_{-\infty}^\infty &= 0 \\
 -\int_0^\infty \int_{-\infty}^{t+\infty} \varphi_t u + \varphi_x \cdot f(u) dx dt - \int_{-\infty}^{t+\infty} \varphi(x, 0) u(x, 0) dx &= 0. \quad (\star)
 \end{aligned}$$

Def 3: $u(x, t)$ is a weak solution iff $\forall \varphi \in C_0^1(\mathbb{R}^2)$, (\star) holds.

Examples:

$$\begin{cases} u_t + \left(\sum_i u_i^i\right)_x = 0 \\ u(x, 0) = \begin{cases} -1 & x < 0 \\ 1 & x \geq 0 \end{cases} \end{cases}$$

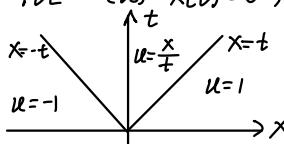


Riemann problem

$$(1) \quad u(x, t) = \begin{cases} -1 & x < 0 \\ 1 & x \geq 0 \end{cases} \quad \text{is a weak solution}$$

check (i) const satisfies PDE $(ii) \quad x(t) = 0, x'(t) = 0, f(u^+) - f(u^-) = \frac{1}{2} \cdot 1^2 - \frac{1}{2}(-1)^2 = 0$.

$$(2) \quad u(x, t) = \begin{cases} -1 & x < -t \\ \frac{x-t}{t} & -t \leq x < t \\ 1 & x \geq t \end{cases}$$



$$(i): u_t = -\frac{x}{t^2}, \quad (\frac{u}{x})_x = \frac{\partial}{\partial x} \left(\frac{x^2}{2x^2} \right) = \frac{x}{t^2} \Rightarrow u_t + f(u)_x = 0.$$

(ii): In fact $u(x,t)$ is cts for $t > 0$.

This is also a weak solution.

We want uniqueness, which leads to the entropy solution.

HW #1.

(1) (i) Find the smallest time t^* for which two characteristic lines of Burgers equation,

$$\begin{cases} u_t + \left(\frac{u}{x}\right)_x = 0 & 0 \leq x < 2\pi \\ u(x,0) = \sin x & \text{periodic} \end{cases}$$

will intersect

(ii) Generalise the result to $\begin{cases} u_t + f(u)_x = 0 \\ u(x,0) = u^0(x) \end{cases} ?$

(2) (i) How does $u_x(x(t), t)$ behave along the characteristic line $x = x(t)$?

(ii) If $u_x = 0$ at $t=0$ (foot of characteristic line), what is the answer to (i)?

(iii) If $u_x = 0$ at $t=0$, how does u_{xx} behave along the characteristic line?

Lecture 02

Entropy Solution

Consider the viscous Burgers equation

$$\begin{cases} u_t + f(u)_x = \varepsilon u_{xx} \\ u^\varepsilon(x, 0) = u^0(x) \end{cases} \quad \varepsilon > 0, \varepsilon \ll 1.$$

The transition region is $O(\varepsilon)$, so impractical to solve numerically

$u(x, t) = \lim_{\varepsilon \rightarrow 0} u^\varepsilon(x, t)$ exists, so we define it as the entropy solution of conservation law.

Exercise: Prove $u(x, t)$ is a weak solution. (Prove existence).

This definition is not practical.

Other versions of entropy conditions

1. $u'(u) \geq 0$ convex
2. entropy function

$$u'(u^\varepsilon)(u_t^\varepsilon + f(u^\varepsilon)_x) = u'(u^\varepsilon) \varepsilon u_{xx}^\varepsilon$$

$$(u(u^\varepsilon))_t + F(u^\varepsilon)_x = \varepsilon ((u'(u^\varepsilon)u_x^\varepsilon)_x - (u''(u^\varepsilon))(u_x^\varepsilon)^2)$$

where $F(u) \triangleq \int^u u'(u) f'(u) du$ (entropy flux).

$$\text{RHS} = \varepsilon u(u^\varepsilon)_{xx} - \varepsilon u''(u^\varepsilon)(u_x^\varepsilon)^2 \leq \varepsilon u(u^\varepsilon)_{xx}. \quad u_x^\varepsilon \rightarrow \delta \text{ function} \rightarrow \delta^2 \& L'.$$

$\forall \varphi \in C_0^2(\mathbb{R} \times \mathbb{R}^+), \varphi \geq 0,$

$$\int_0^{+\infty} \int_{-\infty}^{+\infty} (u(u^\varepsilon)_t + F(u^\varepsilon)_x) \varphi dt dx \leq \int_0^{+\infty} \int_{-\infty}^{+\infty} \varphi \varepsilon u(u^\varepsilon)_{xx} dx dt$$

$$- \int_0^{+\infty} \int_{-\infty}^{+\infty} u(u^\varepsilon) \varphi_t + F(u^\varepsilon) \varphi_x dx dt \leq \varepsilon \int_0^{+\infty} \int_{-\infty}^{+\infty} u(u^\varepsilon) \varphi_{xx} dx dt$$

u^ε is bdd uniformly, then by DCT, let $\varepsilon \rightarrow 0^+$, ($u^\varepsilon \rightarrow u$ a.e.).

$$- \int_0^{+\infty} \int_{-\infty}^{+\infty} u(u) \varphi_t + F(u) \varphi_x dx dt \leq 0 \quad \text{Integral entropy condition}$$

2. If u is a generic weak solution

In smooth region it holds immediately

$$u'(u) \cdot (u_t + f(u)_x) = 0.$$

Now take a region intersects $x(t)$, large enough.

$$- \iint_{S_2} u(u) \varphi_t + F(u) \varphi_x dx dt - \int_{S_2} u(u) \varphi_t + F(u) \varphi_x dx dt \leq 0.$$

Apply IBP and Stokes theorem (Check this!).

$$\iint_{S_2} u(u)_t \varphi + F(u)_x \varphi dx dt - \int_T (F(u) \varphi, u(u) \varphi) \vec{n} ds$$

$$+ \iint_{S_2} u(u)_t \varphi + F(u)_x \varphi dx dt + \int_T (F(u^+) \varphi, u(u^+) \varphi) \cdot \vec{n} ds \leq 0 \quad \vec{n} = \frac{(1, -x'(t))}{\sqrt{1+x'^2(t)}}$$

$$- \int_T (F(u^-) - x'(t) u(u^-)) \frac{\varphi}{\sqrt{1+x'^2(t)}} ds + \int_T F(u^+) - x'(t) u(u^+) \frac{\varphi}{\sqrt{1+x'^2(t)}} ds \leq 0.$$

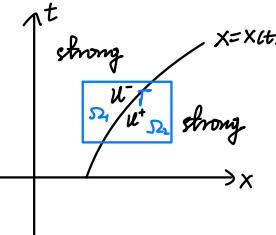
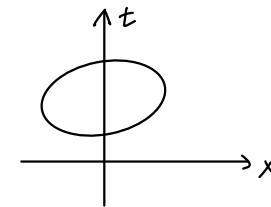
$$\Rightarrow -F(u^-) + x'(t) u(u^-) + F(u^+) - x'(t) u(u^+) \leq 0.$$

$$F(u^+) - F(u^-) \leq x'(t) (u(u^+) - u(u^-)).$$

Like RH condition

Example: Burgers equation

$$\text{D } u(x, t) = \begin{cases} 1 & x \geq 0 \\ -1 & x < 0 \end{cases}$$



$$dx = \frac{1}{\sqrt{1+x'^2(t)}} ds$$

$$\text{Take } U(u) = \frac{u^2}{2}, \quad F(u) = \int U'(u) f'(u) du = \int u \cdot u du = \frac{1}{3} u^3$$

$$U^+ = 1, \quad U^- = 1 \Rightarrow U^+ = \frac{1}{2}, \quad U^- = \frac{1}{2}. \quad F^+ = \frac{1}{3}, \quad F^- = -\frac{1}{3}.$$

$$F^+ - F^- = \frac{2}{3} \neq 0 = X'(t) \cdot (U^+ - U^-).$$

So this is not an entropy solution.

3. Oleinik entropy condition (How to derive)

Along possible discontinuity curves, $S(U^-, Z) \geq S(U^-, U^+) \geq S(Z, U^+)$, where $S(a, b) = \frac{f(a) - f(b)}{a - b}$, $\forall Z$ within U^- & U^+ . This still works for system.

4. If the conservation law is convex, i.e. $f'' > 0$, then Lax entropy condition is enough.

$$\frac{f(u^+) - f(z)}{u^- - z} \geq X''(t) \geq \frac{f(u^+) - f(z)}{u^+ - z}. \quad \text{let } Z \rightarrow U^\pm, \quad f'(U^-) \geq X'(t) \geq f'(U^+) \Rightarrow U^- \geq U^+. \quad \checkmark$$

HW #2

Write a computer program to obtain the exact entropy solution of

$$\begin{cases} Ut + (\frac{v^2}{2})_x = 0, & -\pi \leq U < \pi. \\ \end{cases}$$

$$\begin{cases} U(x, 0) = \alpha + \beta \sin x, & \text{periodic.} \\ \end{cases}$$

α, β are constants

(1) Try to transfer the problem to

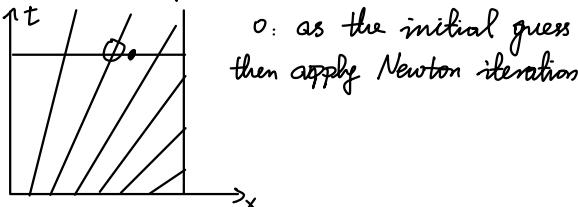
$$\begin{cases} vt + (\frac{v^2}{2})_x = 0, & -\pi \leq v < \pi \\ v(x, 0) = \sin x, & \text{periodic} \end{cases}$$

Consider $x' = ax + bt, \quad t' = cx + dt, \quad U = ev + f$.

(2) Antisymmetric w.r.t. $x=0$.

(3) Shock speed is 0 by (2), so the computational area is confined

(4) Draw 200 forward characteristics



Lecture 03

Entropy solution:

1. $u(x, t) = \lim_{\varepsilon \rightarrow 0} u^\varepsilon(x, t)$. $u_t^\varepsilon + f(u^\varepsilon)_x = \varepsilon u_{xx}^\varepsilon$
2. Entropy function: $U(u)$, $U'' \geq 0$. $F(u)$: $F'(u) = U'(u)f'(u)$, $F(u) = \int_0^{+\infty} U'(u)f'(u)du$.
 $\int_0^{+\infty} \int_{-\infty}^{+\infty} U(u)\varphi_t + F(u)\varphi_x dx dt \leq 0 \quad \forall \varphi \in C_0^1(\mathbb{R}^+ \times \mathbb{R}), \varphi \geq 0$.
3. For piecewise smooth weak solutions, $(F(u^+)) - F(u^-)) \leq x'(t)(U(u^+) - U(u^-))$.
4. Oleinik: $S(u^-, u) \geq S(u^-, u^+) \geq S(u, u^+) \quad \forall u$ between u^- and u^+ along T .
 $S(a, b) \triangleq \frac{f(b) - f(a)}{b - a}$
5. Lax: $u^- \geq u^+$ (sufficient only if $f''(u) \geq 0$)

Entropy solution for Burgers equation

$$\begin{cases} u_t + \left(\frac{u^2}{2}\right)_x = 0 \\ u(x, 0) = \begin{cases} -1, & x < 0 \\ 1, & x \geq 0 \end{cases} \end{cases} \text{ rarefaction waves}$$

$$u(x, t) = \begin{cases} -1, & x < 0 \\ 1, & x \geq 0 \end{cases} \text{ is not an entropy solution}$$

$$u(x, t) = \begin{cases} -1 & x < -t \\ x/t & -t < x < t \\ 1 & x \geq t \end{cases} \text{ is an entropy solution}$$

In fact, entropy solution is unique.

Theorem: The entropy solution evolution operator is L^1 -contractive, i.e. if

$$\begin{cases} u_t + f(u)_x = 0 \\ u(x, 0) = u^0(x) \end{cases} \quad \begin{cases} v_t + f(v)_x = 0 \\ v(x, 0) = v^0(x) \end{cases}$$

Then $\int_{-\infty}^{+\infty} |u(x, t) - v(x, t)| dx \leq \int_{-\infty}^{+\infty} |u(x, 0) - v(x, 0)| dx$.

Pf: We use the viscosity solution as the def of entropy solution.

For fixed $\varepsilon > 0$, $u_t^\varepsilon + f(u^\varepsilon)_x = \varepsilon u_{xx}^\varepsilon$. $v_t^\varepsilon + f(v^\varepsilon)_x = \varepsilon v_{xx}^\varepsilon$.

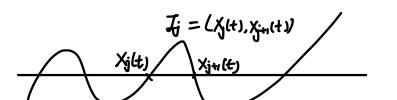
$w^\varepsilon = u^\varepsilon - v^\varepsilon$. (Assume w^ε has countably many 0).

$$\begin{aligned} \frac{d}{dt} \int_{-\infty}^{+\infty} |w(x, t)| dx &= \frac{d}{dt} \sum_j \int_{x_{j(t)}}^{x_{j(t)+1}} |w(x, t)| dx = \frac{d}{dt} \sum_j \int_{x_{j(t)}}^{x_{j(t)+1}} s_j(t) w(x, t) dx \\ &= \sum_j \frac{d}{dt} \int_{x_{j(t)}}^{x_{j(t)+1}} s_j(t) w(x, t) dx \\ &= \sum_j \left[x_{j(t)+1} s_j(t) w(x_{j(t)+1}, t) - x_{j(t)} s_j(t) w(x_{j(t)}, t) + \int_{x_{j(t)}}^{x_{j(t)+1}} s_j(t) w_x(x, t) dx \right] \\ &= \sum_j \int_{x_{j(t)}}^{x_{j(t)+1}} s_j(t) (u^\varepsilon(x, t) - v^\varepsilon(x, t)) dx \quad \text{Plug in the PDEs.} \\ &= \sum_j \int_{x_{j(t)}}^{x_{j(t)+1}} s_j(t) (-f(u^\varepsilon)_x + \varepsilon u_{xx}^\varepsilon + f(v^\varepsilon)_x - \varepsilon v_{xx}^\varepsilon) dx \\ &= \sum_j s_j(t) \left[-f(u^\varepsilon(x_{j(t)}, t)) + f(u^\varepsilon(x_{j(t)+1}, t)) + f(v^\varepsilon(x_{j(t)+1}, t)) - f(v^\varepsilon(x_{j(t)}, t)) \right. \\ &\quad \left. + \varepsilon (u^\varepsilon(x_{j(t)+1}, t) - u^\varepsilon(x_{j(t)}, t) - v^\varepsilon(x_{j(t)+1}, t) + v^\varepsilon(x_{j(t)}, t)) \right] \end{aligned}$$

$$= \sum_j s_j(t) (u_x(x_{j(t)+1}, t) - u_x(x_{j(t)}, t)) \leq 0$$

as $s_j(t) \geq 0 \Rightarrow u_x|_{x_{j(t)+1}} \leq 0$, $u_x|_{x_{j(t)}} \geq 0$. and vice versa.

So $\frac{d}{dt} \int_{-\infty}^{+\infty} |u^\varepsilon(x, t) - v^\varepsilon(x, t)| dx \leq 0$. let $\varepsilon \rightarrow 0$ (by DCT).



$$s_j(t) = w(x, t) \text{ on } (x_j(t), x_{j+1}(t)).$$

$$w = u^\varepsilon - v^\varepsilon$$

Generic space for entropy solutions

$BV \cap L^1$. $\|\cdot\|_{BV}$ is a semi norm.

L' -contraction \Rightarrow order preserving: $u(x, 0) \geq v(x, 0) \Rightarrow u(x, t) \geq v(x, t)$.

Pf: $\|u(x, t) - v(x, t)\|_1 \leq \|u(x, 0) - v(x, 0)\|_1 = \int_{-\infty}^{+\infty} |u(x, 0) - v(x, 0)| dx$.

$$u_t + f(u)_x = 0 \Rightarrow \frac{d}{dt} \int_{-\infty}^{+\infty} u(x, t) dx = 0 \text{ as } f(u)|_{t=0} = 0.$$

$$\int_{-\infty}^{+\infty} |u(x, 0) - v(x, 0)| dx = \int_{-\infty}^{+\infty} |u(x, t) - v(x, t)| dx \leq \|u(x, t) - v(x, t)\|_1.$$

All " \leq " becomes " $=$ ", thus $u(x, t) \geq v(x, t)$.

Total Variation Diminishing (TVD).

Pf: $\begin{cases} v_t + f(v)_x = 0 \\ v(x, 0) = u^*(x + \Delta x) \end{cases} \quad \begin{cases} u_t + f(u)_x = 0 \\ u(x, 0) = u^*(x). \end{cases} \quad v(x) = u(x + \Delta x).$

$$\|v(x, t) - u(x, t)\|_1 \leq \|v(x, 0) - u(x, 0)\|_1,$$

$$\int_{-\infty}^{+\infty} \frac{|v(x + \Delta x, t) - u(x, t)|}{\Delta x} dx \leq \int_{-\infty}^{+\infty} \frac{|u(x + \Delta x, 0) - u(x, 0)|}{\Delta x} dx \text{ let } \Delta x \rightarrow 0. \quad TV(u(x, t)) \leq TV(u(x, 0)).$$

TVD doesn't guarantee uniqueness.

Maximum Principle:

If $m \leq u(x, 0) \leq M$, then $m \leq u(x, t) \leq M$.

Pf: $m(M)$ is a solution to $u_t + f(u)_x = 0$.

$$m = v(x, 0) \leq u(x, t) \Rightarrow m = v(x, t) \leq u(x, t).$$

Riemann Problem

$$\begin{cases} u_t + f(u)_x = 0 \\ u(x, 0) = \begin{cases} l_e, & \text{if } x < 0 \\ l_r, & \text{if } x \geq 0 \end{cases} \end{cases}$$

(1) $f''(u) > 0$

(1a) $l_e > l_r$, then the entropy solution is a shock $u(x, t) = \begin{cases} l_e, & x < st \\ l_r, & x \geq st \end{cases}$ where $s = \frac{f(l_r) - f(l_e)}{l_r - l_e}$ is shock speed

(1b) $l_e < l_r$, then the entropy solution is a rarefaction wave $u(x, t) = \begin{cases} l_e & x < f'(l_e)t \\ f'(l_e) & f'(l_e)t \leq x < f'(l_r)t \\ l_r & x \geq f'(l_r)t \end{cases}$
check it's a weak solution ($\text{cts} \Rightarrow \text{RH, Lax hold}$).

HW #3

(1) Prove that $u(x, t) = (f'')^{-1}\left(\frac{x}{t}\right)$ is a solution to $u_t + f(u)_x = 0$. Here $f''(u) > 0$.

Rarefaction solution

Lecture 04

Inversibility of entropy solution

$$\begin{cases} u_t + \left(\frac{u^2}{2}\right)_x = 0 \\ u(x, 0) = u_0(x) \end{cases}$$

$$(i) u_0(x) = \begin{cases} 1 & x < -1 \\ 0 & x \geq -1 \end{cases} \quad (ii) u_0(x) = \begin{cases} 1 & x < -1 \\ -x & -1 \leq x < 0 \\ 0 & x \geq 0 \end{cases}$$

$$\text{But } u(x, 1) = \begin{cases} 1 & x < 0 \\ 0 & x \geq 0 \end{cases} \text{ for (i) \& (ii).}$$

Riemann Problem

(1b) PDE is scalar-scaling invariant: $x' = ax$, $t' = at$ $\forall a > 0$.

$$u(1, 1) = u(2, 2) \dots \text{Self-similarity: } u(x, t) = v\left(\frac{x}{t}\right)$$

(2) f is not convex: f'' changes sign. Osher

$$\text{Let } \zeta = \frac{x}{t}, \quad u(x, t) = \begin{cases} -\frac{d}{d\zeta} \left(\min_{v \in \text{values}} (f(v) - \zeta v) \right) & \text{if } u_r > u_l \\ -\frac{d}{d\zeta} \left(\max_{v \in \text{values}} (f(v) - \zeta v) \right) & \text{if } u_l > u_r \end{cases}$$

But if we only care about $f(u_0, t)$, then

$$f(u(0, t)) = \begin{cases} \min_{u \in \text{values}} f(u), & u_l < u_r \\ \max_{u \in \text{values}} f(u), & u_l > u_r. \end{cases}$$

Lax equivalence theorem:

Strong. Stability + dissipative

Example. Riemann Problem for Burgers equation

$$\begin{cases} u_t + \left(\frac{u^2}{2}\right)_x = 0 \\ u(x, 0) = \begin{cases} 1, & x < 0 \\ 0, & x \geq 0 \end{cases} \end{cases} \Rightarrow u(x, t) = \begin{cases} 1 & x < \frac{1}{2} \\ 0 & x \geq \frac{1}{2} \end{cases}$$

$$u_t + uu_x = 0. \text{ Consider } u_j^{**} = u_j^n - u_j^n \frac{\Delta t}{\Delta x} (u_j^n - u_{j-1}^n) \quad (\text{Resemble } u_j^{**} = u_j^n - \alpha \frac{\Delta t}{\Delta x} (u_{j+1}^n - u_j^n), \frac{\Delta t}{\Delta x} < 0).$$

The scheme is good if the solution of the PDE is smooth.

$$u_j^* = u_j^n - u_j^n \frac{\Delta t}{\Delta x} (u_j^n - u_{j-1}^n) = \begin{cases} 1, & j < 0 \\ 0, & j = 0 \\ 0, & j > 0 \end{cases} \text{ which is the initial condition. } u_j^n \rightarrow \begin{cases} 1, & x < 0 \\ 0, & x \geq 0 \end{cases} \text{ is not a weak solution}$$

Conservative scheme

$$u_j^{**} = u_j^n - \lambda (\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}}) \text{ where } \hat{f}_{j+\frac{1}{2}} = \hat{f}(u_{j-p}, \dots, u_{j+q}) \text{ satisfying}$$

(1) Consistency: $\hat{f}(u, \dots, u) = f(u)$ for constant u

(2) Lipschitz continuity

\hat{f} is numerical flux. $\lambda \equiv \frac{\Delta t}{\Delta x}$

Lax-Wendroff Theorem

If a conservative scheme converges, then the limit is a weak solution

$$P.P.: \sum_{j,n} \left(\frac{U_j^{n+1} - U_j^n}{\Delta t} - \frac{f_{j+\frac{1}{2}}^n - f_{j-\frac{1}{2}}^n}{\Delta x} \right) \psi_j^n = 0 \quad \psi \in C^1(\mathbb{R} \times \mathbb{R}^+), \quad \psi_j^n = \psi(x_j, t^n).$$

Sum by parts, $\sum_j (a_{j+1} - a_j) b_j = \sum_j a_j (b_j - b_{j-1})$

$$0 = \dots = - \sum_{j,n} \left(U_j^n \frac{\psi_j^n - \psi_{j-1}^n}{\Delta x} + f_{j+\frac{1}{2}}^n \frac{\psi_{j+\frac{1}{2}}^n - \psi_{j-\frac{1}{2}}^n}{\Delta x} \right) \Delta t \Delta x$$

$U_j^n \rightarrow u$, $f_{j+\frac{1}{2}}^n \rightarrow f(u)$ by consistency & Lipschitz continuity.

$$\text{DCT} \int_0^\infty \int_{-\infty}^\infty u \psi_t + f(u) \psi_x \, dx dt.$$

Examples of conservative schemes

(1) If $f'(u) > 0$, then $\hat{f}_{j+\frac{1}{2}} = f(U_j)$. Less useful. (upwind)

(2) Lax-Friedrichs scheme: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_j, U_{j+1}) = \frac{1}{2} (f(U_j) + f(U_{j+1}) - \alpha(U_{j+1} - U_j))$ $\alpha = \max_u |f'(u)|$. (LF)

(3) Local Lax-Friedrichs: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_j, U_{j+1}) = \frac{1}{2} (f(U_j) + f(U_{j+1}) - \alpha_{j+\frac{1}{2}} (U_{j+1} - U_j))$ $\alpha_{j+\frac{1}{2}} = \max_{u \in [U_j, U_{j+1}]} |f'(u)|$. (LLF)

(4) Roe scheme: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_j, U_{j+1}) = \begin{cases} f(U_j), & \text{if } s_{j+\frac{1}{2}} \geq 0 \\ f(U_{j+1}), & \text{if } s_{j+\frac{1}{2}} < 0 \end{cases}$ (can do multiplication to determine the sign when coding).

(5) Godunov scheme: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_j, U_{j+1}) = f(V(0, t))$ where $v(0, t)$ is the solution to the Riemann problem

$$\begin{cases} v_t + f(v)_x = 0 \\ v(x, 0) = \begin{cases} U_j & x < 0 \\ U_{j+1} & x \geq 0 \end{cases} \end{cases} \quad \hat{f}(U_j, U_{j+1}) = \frac{U_j + U_{j+1}}{2} \hat{f}'(U_{j+\frac{1}{2}}) (\hat{f}(U_{j+1}) - \hat{f}(U_j))$$

(6) Lax-Wendroff: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_j, U_{j+1}) = \frac{1}{2} (f(U_j) + f(U_{j+1}) - \lambda f(\frac{U_j + U_{j+1}}{2}) (f(U_{j+1}) - f(U_j)))$

(7) MacCormack: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_j, U_{j+1}) = \frac{1}{2} [f(U_{j+1}) + f(U_j - \lambda (f(U_{j+1}) - f(U_j)))]$

HW #4

Code up (1) Roe (2) Godunov (3) LF (4) LW for

$$\begin{cases} u_t + (\frac{u^2}{2})_x = 0 \\ u(x, 0) = \frac{1}{3} + \frac{2}{3} \sin x \quad [0, 2\pi] \end{cases}$$

Get the result for $t=0.3$, $t=2$.

Take $N=20, 40, 80, \dots, 640$, error table:

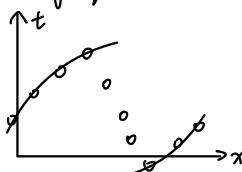
$\frac{\Delta x}{N}$	L^1 , order: L^1 , order: L^∞ , order	$E_n = Ch^r$, $e_{2n} = c(2h)^r$
$\frac{2\pi/20}{140}$	$\frac{1}{N} \sum e_n \left(\frac{1}{N} \sum e_n ^r \right)^{\frac{1}{r}}$ $2.41 \times 10^{-4} \quad 1.18 \quad (\text{three digits})$	$\frac{e_{2n}}{e_n} = 2^r$, $r = \log_2 \frac{e_{2n}}{e_n}$

$$|f'(cw)| \cdot \frac{\Delta t}{\Delta x} \approx 0.6$$

$$\frac{e_{2n}}{e_n} = 2^r, \quad r = \log_2 \frac{e_{2n}}{e_n}.$$

For $t=2$, another error table for points $|x_j - x_{\text{shock}}| \geq 1$.

Finally, plot $t=2$ for $N=40$



Lecture 05

A non-conservative scheme may converge to a function which is not a weak solution.

General upwind scheme

A 3-point scheme (ω -point flux) is called an upwind scheme if

$$\hat{f}(u_j, u_{j+1}) = \begin{cases} f(u_j) & \text{if } f'(u) \geq 0 \text{ in } [u_j, u_{j+1}] \\ f(u_{j+1}) & \text{if } f'(u) \leq 0 \text{ in } [u_j, u_{j+1}] \end{cases}$$

Exercise: Roe, Godunov are upwind. LF, Local LF, LW are not upwind.

Consider $U_t + (\frac{u^2}{2})_x = 0$. The correct solution is $u(x,t) = \begin{cases} -1 & x < -t \\ x/t & -t \leq x \leq t \\ 1 & x > t \end{cases}$

For Roe scheme, $u_j^o = \begin{cases} -1 & j < 0 \\ 1 & j \geq 0 \end{cases}$, $\hat{f}_{j+\frac{1}{2}} = \frac{1}{2} \forall j$, $u_j = u_j^o$, so it doesn't converge to entropy solution.

Monotone schemes

If $u_j^{n+1} = G(u_{j-p}, \dots, u_{j+q}) = u_j^n - \lambda(\hat{f}(u_{j-p}, \dots, u_{j+q}) - \hat{f}(u_{j-p-1}, \dots, u_{j+q-1}))$ and G is increasing (non-decreasing) w.r.t. every argument ($G(1, 1, \dots, 1)$), then the scheme is called a monotone scheme.

For example, LF is monotone where $\alpha \frac{\partial f}{\partial x} \leq 1$.

$$u_j^{n+1} = u_j^n - \lambda(\hat{f}(u_j, u_{j+1}) - \hat{f}(u_{j+1}, u_j)) \text{ with}$$

$$\hat{f}(u_j, u_{j+1}) = \frac{1}{2}(f(u_j) + f(u_{j+1}) - \alpha(u_{j+1} - u_j)), f(u_j, u_{j+1}) = \frac{1}{2}(f(u_j) + f(u_{j+1}) - \alpha(u_j - u_{j+1})).$$

$$\frac{\partial G}{\partial u_j} = \lambda(\frac{1}{2}\alpha + \frac{1}{2}f'(u_{j+1})) \geq 0$$

$$\frac{\partial G}{\partial u_{j+1}} = 1 - \lambda(\frac{1}{2}f'(u_j) + \frac{\alpha}{2} - \frac{1}{2}f'(u_{j+1}) + \frac{\alpha}{2}) = 1 - \lambda\alpha$$

$$\frac{\partial G}{\partial u_{j+1}} = \lambda(\frac{1}{2}\alpha - \frac{1}{2}f'(u_{j+1})) \geq 0.$$

We consider only 3-point (ω -point flux) monotone schemes from now on.

Suppose $\hat{f}(1, 1)$ and $\lambda(\hat{f}_1 - \hat{f}_2) \leq 1 \Rightarrow$ monotone scheme

$$u_j^{n+1} = u_j^n - \lambda(\hat{f}(u_j^n, u_{j+1}^n) - \hat{f}(u_{j+1}^n, u_j^n)) = G(u_j^n, u_{j+1}^n, u_{j+1}^n)$$

$$G_1(\frac{\partial G}{\partial u_j}) = \lambda\hat{f}_1 \geq 0, G_2(\frac{\partial G}{\partial u_{j+1}}) = 1 - \lambda(\hat{f}_1 - \hat{f}_2) \geq 0, G_3(\frac{\partial G}{\partial u_{j+1}}) = -\lambda\hat{f}_2 \geq 0.$$

Properties of monotone scheme

(1) Order preserving: If $u_j \leq v_j \forall j \Rightarrow G(u)_j \leq G(v)_j \Rightarrow G(u_{j-p}, \dots, u_{j+q}) \leq \dots \leq G(v_{j-p}, \dots, v_{j+q})$.

(2) Local maximum principle: $\min_{j-p \leq i \leq j+q} u_i \leq G(u)_j \leq \max_{j-p \leq i \leq j+q} u_i$ ($u_j \in G(u)_j \in M_j$)
 Let $v_i = \begin{cases} u_i & \text{aw.} \\ u_i & \text{if } u_i < v_i, \text{ by (1). } G(u)_i \leq G(v)_i. \text{ In particular, } G(u)_j \leq G(v)_j = v_j \in M_j. \end{cases}$ \rightarrow consistency

(3) ℓ^1 contraction: $\|G(u) - G(v)\|_{\ell^1} \leq \|u - v\|_{\ell^1}$.

$$\text{Let } w_j = u_j \vee v_j = v_j + (u_j - v_j)^+$$

$$G(u)_j \leq G(w)_j, G(v)_j \leq G(w)_j. G(w)_j - G(v)_j \geq \begin{cases} G(u)_j - G(v)_j & \text{(both)} \\ 0 & \end{cases}$$

$$G(w)_j - G(w)_j \geq G(u)_j - G(v)_j = (G(u)_j - G(v)_j) \vee 0 = (G(u)_j - G(v)_j)^+.$$

$$\sum_j (G(u)_j - G(v)_j)^+ = \sum_j G(w)_j - G(v)_j = \sum_j w_j - v_j = \sum_j (u_j - v_j)^+$$

conservative (periodic / compact supports).

$$\sum_j |G(u)_j - G(v)_j| = \sum_j (G(u)_j - G(v)_j)^+ + \sum_j (G(v)_j - G(u)_j)^+ \leq \sum_j (u_j - v_j)^+ + \sum_j (v_j - u_j)^+ = \sum_j |u_j - v_j|.$$

(Crandall-Tartar lemma).

$$(4) TVD: TV(G(u)) \leq TV(u), TV(u) \triangleq \sum_j |u_{j+1} - u_j|.$$

Take $v_j = u_{j+1}$, then $G(v)_j = G(u)_{j+1}$, and by (3).

(5) Entropy inequality:

$$U(u) = |u - c|, F(u) = \text{sgn}(u - c)(f(u) - f(c)) \quad \forall c.$$

$$U'(u) = \begin{cases} 1, & u > c \\ -1, & u < c \end{cases}, \quad U'' = 2f'(x - c) \geq 0.$$

$$\text{WTS: } \frac{u_j^{**} - u_j^*}{\Delta t} + \frac{\hat{F}_{j+\frac{1}{2}} - \hat{F}_{j-\frac{1}{2}}}{\Delta x} \leq 0 \quad \forall j \quad \text{Cell-entropy inequality}$$

$$\hat{F} = \hat{f}(cVu) - \hat{f}(c \wedge u). \quad \text{Verify } \hat{F} \text{ is consistent with } F.$$

$$|u_j^{**} - c| - \lambda(\hat{F}_{j+\frac{1}{2}} - \hat{F}_{j-\frac{1}{2}}) = G(cVu)_j - G(c \wedge u)_j. \quad \text{check by def.}$$

$$c = G(c, \dots, c) \leq G(cVu^n), u^{**} = G(u^n) \leq G(cVu^n), CVu^{**} \leq G(cVu^n).$$

$$c \wedge u^{n+1} = G(c \wedge u^n), CVu^{**} - c \wedge u^{n+1} \leq |u_j^{**} - c| - \lambda(\hat{F}_{j+\frac{1}{2}} - \hat{F}_{j-\frac{1}{2}})$$

$$CVu^{**} - c \wedge u^{n+1} = |u^{**} - c|. \quad \text{So } \frac{|u^{**} - c| + |u_j^{**} - c|}{\Delta t} + \frac{\hat{F}_{j+\frac{1}{2}} - \hat{F}_{j-\frac{1}{2}}}{\Delta x} \leq 0.$$

(6) Rate of convergence: $\|u_h - u\|_L \leq C\sqrt{\Delta x}$ (general). Expect $\|u_h - u\|_L \leq C\Delta x$.

HW #5

Prove that L \bar{F} , LL \bar{F} and Godunov schemes are monotone, but Roe and LW are not.

$$\text{e.g. } \hat{f}(2,5) < \hat{f}(1,5)$$

Flow chart

$$N=40, \Delta x = \frac{2\pi}{N}, x(i) = i \Delta x, i = 0, 1, \dots, N+1 \quad \begin{matrix} 0 & 1 & \dots & N & N+1 \end{matrix}$$

$$u(i) = \alpha + \beta \sin(x(i)), i = 1, \dots, N.$$

$$t=0, t_end=0.3, ic=0.$$

loop:

$$\text{Boundary condition: } u(0) = u(N), u(N+1) = u(1).$$

$$\alpha_m = \max |f'(u)|.$$

$$\Delta t = CFL \times \Delta x / \alpha_m, \quad CFL = 0.6.$$

if $(t + \Delta t > t_{end})$ then

$$\Delta t = t_{end} - t$$

$$ic = 1.$$

endif

$$f(i) = u(i)^2 / 2, \quad i = 0, 1, \dots, N+1.$$

$$\text{flux}(i+\frac{1}{2}) = \frac{1}{2}(f(i) + f(i+1)) - \alpha_m(u(i+1) - u(i)), \quad i = 0, 1, \dots, N$$

$$u(i) = u(i) - \lambda(\text{flux}(i+\frac{1}{2}) - \text{flux}(i-\frac{1}{2})).$$

$t = t + \Delta t$. if $ic = 0$, go back (loop).

Lecture 06

Monotone scheme convergence rate:

$c\sqrt{\Delta x}$, (sharp for linear flux: $f(u) = u$)

Δx Some cases like burgers.

Theorem: Monotone schemes are at most first order accurate.

Theorem: TVD \Rightarrow monotonicity preserving MP. If $U_j^n \geq U_{j+1}^n$, then $U_{j+1}^{n+1} \geq U_j^{n+1}$

Pf: $TV(U^{n+1}) = \sum_j |U_{j+1}^{n+1} - U_j^{n+1}| \leq \sum_j |U_{j+1}^n - U_j^n| = \sum_j U_j^n - U_{j+1}^n \quad \text{②: } U_{j+1}^{n+1} - U_j^{n+1} \geq 0 \quad \text{③: conservation.}$

This implies $U_{j+1}^{n+1} - U_j^{n+1} \geq 0$.

Theorem (Godunov): MP + linear scheme \Rightarrow monotone

Def: A scheme is called a linear scheme if it is linear when applied to linear PDEs. ($U_t + AU_x = 0$).

In fact, Godunov = Roe = Lax-Friedrichs for linear PDEs: $U_j^{n+1} = U_j^n - \lambda(U_j^n - U_{j+1}^n)$.

LW is a linear scheme

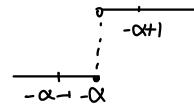
Pf: $U_t + AU_x = 0$. $U_j^{n+1} = \sum_{l=-k}^k C_l(\lambda) U_{j+l}^n$. Consider $U_m^n = \begin{cases} 0 & i \leq -\alpha \\ 1 & i > -\alpha \end{cases}$. Then $U_{m+1}^{n+1} \geq U_m^n$.

$$U_{j+1}^{n+1} = \sum_l C_l(\lambda) U_{j+1+l}^n, \quad U_j^{n+1} = \sum_l C_l(\lambda) U_{j+l}^n. \quad \text{Let } \Delta t U_j = U_{j+1} - U_j$$

$$\Delta t U_{j+1}^{n+1} = \sum_l C_l(\lambda) \Delta t U_{j+1+l}^n, \quad \Delta t U_j^{n+1} = \sum_l C_l(\lambda) \Delta t U_{j+l}^n.$$

$$\Delta t U_{j+1}^{n+1} = \sum_l C_l(\lambda) \Delta t U_{j+1+l}^n = C_0(\lambda), \quad \Delta t U_j^{n+1} = \sum_l C_l(\lambda) \Delta t U_{j+l}^n = C_0(\lambda) \quad \text{if } m = -\alpha.$$

$\Delta t U_{j+1}^{n+1} \geq 0$ by MP. $C_0(\lambda) \geq 0$ $\forall l$. Then it's a monotone scheme.



Also called positive schemes. Linear schemes solving linear PDEs.

Finite volume scheme (FV)

$$\bar{U}_j = \frac{1}{\Delta x} \int_{x_{j-\frac{1}{2}}}^{x_{j+\frac{1}{2}}} u(x) dx$$

cell averages

For the conservation law,

$$J_j = (x_{j-\frac{1}{2}}, x_{j+\frac{1}{2}})$$

$U_t + f(u)_x = 0$. Integrate, $\frac{d}{dt} \bar{U}_j + \frac{1}{\Delta x} (f(u(x_{j+\frac{1}{2}}, t)) - f(u(x_{j-\frac{1}{2}}, t))) = 0$.

Initially, $\bar{U}_j^0 = \frac{1}{\Delta x} \int_{x_{j-\frac{1}{2}}}^{x_{j+\frac{1}{2}}} u^0(x) dx$, which is available.

$\bar{U}_j = \bar{U}_j^0 - \lambda (f(u(x_{j+\frac{1}{2}}, t)) - f(u(x_{j-\frac{1}{2}}, t))) = 0$. (Euler Forward for \bar{U}_j). $u(x_{j+\frac{1}{2}}, t)$, $u(x_{j-\frac{1}{2}}, t)$ are unknown.

Need a reconstruction procedure to get, from $\{\bar{U}_j\}$, approximations to $U_{j+\frac{1}{2}}$ ($\hat{U}_{j+\frac{1}{2}}$).

Relation with monotone scheme:

Monotone scheme: $\frac{du}{dt} + (\hat{f}(\bar{U}_j, \bar{U}_{j+1}) - \hat{f}(\bar{U}_{j-1}, \bar{U}_j)) = 0$.

~~$U_{j+\frac{1}{2}} = \bar{U}_j$~~ FV can use numerical flux after reconstruction

~~$U_{j+\frac{1}{2}} = \bar{U}_j$~~

FV:

① reconstruction to get $U_{j+\frac{1}{2}}$ & $U_{j-\frac{1}{2}}$ at $x_{j+\frac{1}{2}}$.

② monotone flux: $\hat{f}_{j+\frac{1}{2}} = \hat{f}(U_{j+\frac{1}{2}}, U_{j-\frac{1}{2}})$

$$\textcircled{3} \text{ discretize } \frac{d\bar{u}_j}{dt} + \frac{1}{\Delta x} (\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}}) = 0 \text{ in time}$$

Reconstruction

$P(x)$ is a reconstruction polynomial in the stencil $\{I_{j-r}, \dots, I_{j+q}\}$ if $\int_{x_{j-r-\frac{1}{2}}}^{x_{j+\frac{1}{2}}} P(x) dx = \bar{u}_i$. $i = j-r, \dots, j+q$. ($*$) $r+q+1$ conditions. $\deg P \leq r+q$

(i) Chain of the stencil:

$$S_0 = \{\bar{f}_j\}, \quad P(x) = \bar{f}_j$$

$S_2 = \{I_{j-1}, I_j, I_{j+1}\}$, $P(x) = a + bx + cx^2$. ($*$) for $i=j-1, j, j+1$. Give 3 linear equations for a, b, c .

$$S_4 = \{I_{j-2}, I_{j-1}, I_j, I_{j+1}, I_{j+2}\}$$

Assume $\bar{u}_i = \frac{1}{\Delta x} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} u(x) dx$ is known for $i=j-r, \dots, j+q$.

$$\text{Define } U(x) = \int_{x_{j-r-\frac{1}{2}}}^x u(\zeta) d\zeta. \quad U(x_{i+\frac{1}{2}}) = \int_{x_{j-r-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} u(\zeta) d\zeta = \sum_{\ell=j-r}^{i} \int_{x_{\ell-\frac{1}{2}}}^{x_{\ell+\frac{1}{2}}} u(\zeta) d\zeta = \sum_{\ell=j-r}^i \bar{u}_\ell \Delta x$$

So we know the values of $U(x)$ at half grids: $i = j-r-1, j-r, \dots, j+q$. ($U(x_{j-r-1+\frac{1}{2}}) = 0$).

From $U(x)$, we can build $P(x)$ is a polynomial s.t. $P(x_{i+\frac{1}{2}}) = U(x_{i+\frac{1}{2}})$ (interpolation polynomial)

$\deg P(x) = r+q+1$ ($r+q+2$ values)

Claim: $P(x) = P'(x)$.

\rightarrow must be in $j-r-1, \dots, j+q$

Check: $\int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} P(x) dx = \frac{1}{\Delta x} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} P'(x) dx = \frac{1}{\Delta x} (P(x_{i+\frac{1}{2}}) - P(x_{i-\frac{1}{2}})) \stackrel{?}{=} \frac{1}{\Delta x} (U(x_{i+\frac{1}{2}}) - U(x_{i-\frac{1}{2}})) = \bar{u}_i$ by def of $U(x)$.

HW#6

Prove the Godunov theorem: A positive linear scheme for solving $u_t + u_x = 0$ is at most 1st order accurate.

Lecture 07

Finite volume schemes:

$$\frac{d\bar{u}_j}{dt} + \frac{1}{\Delta x} (\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}}) = 0$$

$$\bar{u}_j = \frac{1}{\Delta x} \int_{x_{j-\frac{1}{2}}}^{x_{j+\frac{1}{2}}} u(x, t) dx, \quad \hat{f}_{j+\frac{1}{2}} = \hat{f}(u_{j+\frac{1}{2}}, u_{j+\frac{1}{2}}^+), \quad \hat{f} \text{ is a monotone flux.}$$

$\{\bar{u}_j\}$ reconstruction $\rightarrow \{u_j^\pm\}$: build $p(x)$: $u \rightarrow u \rightarrow p \rightarrow p$. (P can be in Lagrange form / Newton form)

But in fact, $\{u_j^\pm\} \rightarrow u_{j+\frac{1}{2}}$ is indeed linear, that is $u_{j+\frac{1}{2}} = a\bar{u}_j + b\bar{u}_{j+1} + c\bar{u}_{j+2}$

$$u_{j+\frac{1}{2}} = -\frac{1}{6}\bar{u}_{j-1} + \frac{5}{6}\bar{u}_j + \frac{1}{3}\bar{u}_{j+1}, \quad u_{j+\frac{1}{2}}^+ = \frac{1}{3}\bar{u}_j + \frac{5}{6}\bar{u}_{j+1} - \frac{1}{6}\bar{u}_{j+2}$$

If we use Euler forward for time marching, then the scheme is unconditionally unstable.

Runge-Kutta: RK3 as spatial order is 3. That is SSP-RK3:

$$u_j^{(0)} = \bar{u}_j - \lambda(\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}}) = L(\bar{u}_j^n)$$

$$u_j^{(1)} = \frac{3}{4}\bar{u}_j^n + \frac{1}{4}L(\bar{u}_j^{(0)}) = \frac{3}{4}\bar{u}_j^n + \frac{1}{4}[\bar{u}_j^{(0)} - \lambda(\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}})]$$

$$u_j^{(2)} = \frac{1}{3}\bar{u}_j^n + \frac{2}{3}L(\bar{u}_j^{(1)}) = \frac{1}{3}\bar{u}_j^n + \frac{2}{3}[\bar{u}_j^{(1)} - \lambda(\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}})]$$

Not monotone, monotone preserving, TVD, ... Must have oscillations.

Is it possible to make the scheme TVD? (Preserve good properties like accuracy)

$TV(u) = \sum_j |u_{j+1} - u_j|$. $TV(u^{n+1}) \leq TV(u^n) \Rightarrow$ no oscillation will be created

Lemma (Harten): A scheme written in the following form is TVD: $\bar{u}_j^n = \bar{u}_j^n + G_{j+\frac{1}{2}}(\bar{u}_{j+1}^n - \bar{u}_j^n) - D_{j-\frac{1}{2}}(\bar{u}_j^n - \bar{u}_{j-1}^n)$

with $G_{j+\frac{1}{2}} \geq 0$, $D_{j+\frac{1}{2}} \geq 0$, $G_{j+\frac{1}{2}} + D_{j+\frac{1}{2}} \leq 1$.

Pf: $\bar{u}_{j+1}^{n+1} = \bar{u}_j^n + G_{j+\frac{1}{2}} \Delta + \bar{u}_{j+1}^n - D_{j+\frac{1}{2}} \Delta + \bar{u}_j^n$.

$$\begin{aligned} \Delta + \bar{u}_{j+1}^{n+1} &= \Delta + \bar{u}_j^n + G_{j+\frac{1}{2}} \Delta + \bar{u}_{j+1}^n - D_{j+\frac{1}{2}} \Delta + \bar{u}_j^n - G_{j+\frac{1}{2}} \Delta + \bar{u}_{j-1}^n \\ &= (1 - G_{j+\frac{1}{2}} - D_{j+\frac{1}{2}}) \Delta + \bar{u}_j^n + G_{j+\frac{1}{2}} \Delta + \bar{u}_{j+1}^n + D_{j-\frac{1}{2}} \Delta + \bar{u}_{j-1}^n \end{aligned}$$

$$\begin{aligned} TV(u^{n+1}) &= \sum_j |\Delta + \bar{u}_{j+1}^{n+1}| \leq \sum_j (1 - G_{j+\frac{1}{2}} - D_{j+\frac{1}{2}}) |\Delta + \bar{u}_j^n| + G_{j+\frac{1}{2}} |\Delta + \bar{u}_{j+1}^n| + D_{j-\frac{1}{2}} |\Delta + \bar{u}_{j-1}^n| \\ &= \sum_j |\Delta + \bar{u}_j^n| = TV(u^n). \end{aligned}$$

$$\bar{u}_j^{n+1} = \bar{u}_j^n - \lambda(\hat{f}(u_{j-\frac{1}{2}}, u_{j+\frac{1}{2}}) - \hat{f}(u_{j-\frac{1}{2}}, u_{j+\frac{1}{2}}^+)).$$



Idea: using limiter

$$\text{Let } \tilde{u}_j = u_{j+\frac{1}{2}} - \bar{u}_j, \quad \tilde{u}_j = \bar{u}_j - u_{j-\frac{1}{2}} \quad m: \text{minmod}$$

$$\bar{u}_j^{(\text{mod})} = m(\tilde{u}_j, \bar{u}_{j+1} - \bar{u}_j, \bar{u}_j - \bar{u}_{j-1}), \quad \tilde{u}_j = m(\tilde{u}_j, \bar{u}_{j+1} - \bar{u}_j, \bar{u}_j - \bar{u}_{j-1}).$$

$$m(a_1, \dots, a_n) = \begin{cases} S \cdot \min |a_i| & \text{if } \operatorname{sgn}(a_i) = \dots = \operatorname{sgn}(a_n) = S, \\ 0 & \text{otherwise.} \end{cases}$$

Then $u_{j+\frac{1}{2}}^{(\text{mod})} = \bar{u}_j + \tilde{u}_j^{(\text{mod})}$, $u_{j-\frac{1}{2}}^{(\text{mod})} = \bar{u}_j - \tilde{u}_j^{(\text{mod})}$. The scheme becomes

$$\begin{aligned} \bar{u}_j^{n+1} &= \bar{u}_j^n - \lambda(\hat{f}(u_{j-\frac{1}{2}}^{(\text{mod})}, u_{j+\frac{1}{2}}^{(\text{mod})}) - \hat{f}(u_{j-\frac{1}{2}}^{(\text{mod})}, u_{j+\frac{1}{2}}^{(\text{mod})})) \\ &= \bar{u}_j^n - \lambda(\hat{f}(u_{j-\frac{1}{2}}^{(\text{mod})}, u_{j+\frac{1}{2}}^{(\text{mod})}) - \hat{f}(u_{j-\frac{1}{2}}^{(\text{mod})}, u_{j+\frac{1}{2}}^{(\text{mod})})) \longrightarrow G_{j+\frac{1}{2}}(\bar{u}_{j+1}^n - \bar{u}_j^n) \\ &\quad + \hat{f}(u_{j+\frac{1}{2}}^{(\text{mod})}, u_{j-\frac{1}{2}}^{(\text{mod})}) - \hat{f}(u_{j+\frac{1}{2}}^{(\text{mod})}, u_{j-\frac{1}{2}}^{(\text{mod})})). \longrightarrow D_{j-\frac{1}{2}}(\bar{u}_j^n - \bar{u}_{j-1}^n) \end{aligned}$$

$$G_{j+\frac{1}{2}} = -\lambda(\hat{f}(u_{j-\frac{1}{2}}^{(\text{mod})}, u_{j+\frac{1}{2}}^{(\text{mod})}) - \hat{f}(u_{j-\frac{1}{2}}^{(\text{mod})}, u_{j+\frac{1}{2}}^{(\text{mod})})) / (\bar{u}_{j+1}^n - \bar{u}_j^n)$$

$$= -\lambda \hat{f}_x(u_{j+\frac{1}{2}}^{(\text{mod})}, \xi)(u_{j+\frac{1}{2}}^{(\text{mod})} - u_{j-\frac{1}{2}}^{(\text{mod})}) / (\bar{u}_{j+1}^n - \bar{u}_j^n) \quad (\text{Mean value theorem}).$$

$$= -\lambda \hat{f}_x(\xi, \xi)(\bar{u}_{j+1}^n - \bar{u}_{j+\frac{1}{2}}^{(\text{mod})} - \bar{u}_j^n + \bar{u}_{j-\frac{1}{2}}^{(\text{mod})}) / (\bar{u}_{j+1}^n - \bar{u}_j^n)$$

$$= -\lambda \hat{f}_2(\xi) \left[1 - \frac{\tilde{U}_{j+1}^{(mon)}}{\tilde{U}_{j+1}^{(mon)} - \tilde{U}_j^{(mon)}} + \frac{\tilde{U}_j^{(mon)}}{\tilde{U}_{j+1}^{(mon)} - \tilde{U}_j^{(mon)}} \right], \quad [\dots] \in [0, 2].$$

So $\tilde{f}_{j+\frac{1}{2}} \geq 0$. So as $\tilde{D}_{j+\frac{1}{2}} \geq 0$.

Exercise: Show $\tilde{g}_{j+\frac{1}{2}} + \tilde{D}_{j+\frac{1}{2}} \leq 1$ for $\lambda(\hat{f}_1 - \hat{f}_2) \leq \frac{1}{2}$.

This modification keeps accuracy to large extent.

Generalized MUSCL scheme.

HW #7

- (1) Repeat HW #3 with LF-flux for third-order FVM scheme with SSP-RK3
- (2) Repeat (1) with generalized MUSCL limiter.

Lecture 8

SSP-RK3 does not change the TVD property.

$$\text{Pf: } \bar{U}_j^{(n)} = \bar{U}_j^n - \lambda(\hat{f}_{j+\frac{1}{2}}^{(n)} - \hat{f}_{j-\frac{1}{2}}^{(n)}). \quad TV(\bar{u}^{(n)}) \leq TV(\bar{u}^n)$$

$$\bar{U}_j^{(n)} = \frac{3}{4} U_j^n + \frac{1}{4} (\bar{U}_j^n - \lambda(\hat{f}_{j+\frac{1}{2}}^{(n)} - \hat{f}_{j-\frac{1}{2}}^{(n)})) = \frac{3}{4} \bar{U}_j^n + \frac{1}{4} \bar{w}$$

$$TV(\bar{u}^{(n)}) \leq \frac{3}{4} TV(\bar{u}^n) + \frac{1}{4} TV(\bar{w}) \leq \frac{3}{4} TV(\bar{u}^n) + \frac{1}{4} TV(\bar{u}^n) = TV(\bar{u}^n).$$

So $TV(\bar{u}^{(n)}) \leq TV(\bar{u}^n)$.

Accuracy

$$\bar{U}_j = U_{j+\frac{1}{2}} - \bar{U}_j = U_j + U_{j+\frac{1}{2}} \frac{\Delta x}{2} + O(\Delta x^2) - (U_j + O(\Delta x^2)) = U_{j+\frac{1}{2}} \frac{\Delta x}{2} + O(\Delta x^2).$$

$$\bar{U}_{j+\frac{1}{2}} - \bar{U}_j = U_{j+\frac{1}{2}} - U_j + O(\Delta x^2) = U_{j+\frac{1}{2}} \Delta x + O(\Delta x^2), \quad \bar{U}_j - \bar{U}_{j-\frac{1}{2}} = U_{j-\frac{1}{2}} \Delta x + O(\Delta x^2).$$

$$m(\bar{U}_j, \bar{U}_{j+\frac{1}{2}} - \bar{U}_j, \bar{U}_j - \bar{U}_{j-\frac{1}{2}}) = \bar{U}_j \text{ if } O(\Delta x^2) \text{ is small enough}$$

So if $U_x \sim O(1)$, the scheme is of the same accuracy. (Smooth monotone region).

But when x is close to the extrema, then the order cannot be preserved. In fact, we have

Theorem: All TVD schemes degenerate to first order accuracy at smooth extrema. (Proof).

TVD schemes will be at most 2nd order accurate in L' & first order accurate in L[∞].

High resolution schemes (some TVD)

To deal with this problem, there are two approaches

(1) TVB: $TV(\bar{u}^{(n)}) \leq (1 + C\alpha t) TV(\bar{u}^n)$ or $\leq TV(\bar{u}^n) + C\alpha t$ for fixed $C > 0$.

This implies $TV(\bar{u}^n) \leq K(T)$ for $n \leq T$, K is a const w.r.t. T .

To make TVB from TVD, just need to change m (minmod) to \tilde{m} .

$$\tilde{m}(a_1, \dots, a_p) = \begin{cases} a_i & \text{if } |a_i| \leq \max^2 \\ m(a_1, \dots, a_p) & \text{o.w.} \end{cases}$$

"Pf" of curing degeneration: $\bar{U}_j = \frac{1}{2} U_x \Delta x + O(\Delta x^2)$, pick $M \geq \frac{2}{3} \max |U_{xx}|$

(Recall HW1, just choose M at $t=0$ as U_{xx} doesn't change along the characteristic lines).

(2) ENO: Essentially Non-oscillatory scheme

Intuition: Choose other stencils if there is discontinuous

ENO

ENO interpolation.

Newton form interpolation

Newton divided differences:

0th: $f[x_i] = x_i$

1st: $f[x_i, x_{i+1}] = \frac{f[x_{i+1}] - f[x_i]}{x_{i+1} - x_i}$

2nd: $f[x_i, x_{i+1}, x_{i+2}] = \frac{f[x_{i+1}, x_{i+2}] - f[x_i, x_{i+1}]}{x_{i+2} - x_i}$

$P_0(x) = f[x_i]$, $P_1(x) = P_0(x) + f[x_i, x_{i+1}](x - x_i)$, $P_2(x) = P_1(x) + f[x_i, x_{i+1}, x_{i+2}](x - x_i)(x - x_{i+1})$, ...

Start from a small stencil of 1 or 2 points

Start from 2 points:

$$S_0 = \{x_0, x_1\}, P_0 = f[x_0] + f[x_0, x_1](x - x_0)$$

$$S_1^{(1)} = \{x_0, x_1, x_{-1}\}, P_1^{(1)}(x) = P_0(x) + f[x_0, x_1, x_{-1}](x - x_0)(x - x_1)$$

$$S_1^{(2)} = \{x_0, x_1, x_2\}, P_1^{(2)}(x) = P_0(x) + f[x_0, x_1, x_2](x - x_0)(x - x_1).$$

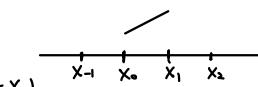
$P_1^{(1)}$ and $P_1^{(2)}$ have the same shape. So to minimize the oscillation.

$$P_1 = \begin{cases} P_1^{(1)} & \text{if } |f[x_0, x_1, x_{-1}]| \leq |f[x_0, x_1, x_2]| \\ P_1^{(2)} & \text{if } |f[x_0, x_1, x_2]| \leq |f[x_0, x_1, x_{-1}]|. \end{cases}$$

For the next step, if we choose $S_i = S_i^{(2)}$,

$$P_2^{(1)} = P_1(x) + f[x_0, x_1, x_2, x_{-1}](x - x_0)(x - x_1)(x - x_2)$$

$$P_2^{(2)} = P_1(x) + f[x_0, x_1, x_2, x_3](x - x_0)(x - x_1)(x - x_2)$$



HW#8

Repeat HW#7 with TVB: $M=0$ (TVD), $M=1$, $M=5$, $M=10$.

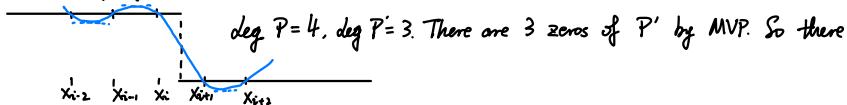
Lecture 9

ENO interpolation

Feature:

- (1) Always include the interval started
- (2) Uniformly accurate for p.w. smooth function (*Proof?*)
- (3) If there is a discontinuity between x_i & x_{i+1} , then $p(x)$ is monotone in (x_i, x_{i+1})

Partial proof:



Numerical observation: Even the discontinuities are close, still accurate (monotone).

Reconstruction:

$$\{ \bar{u}_j \} \rightarrow \bar{u}_{j+\frac{1}{2}} = \int_{x_j}^{x_{j+\frac{1}{2}}} u(\zeta) d\zeta.$$

$$S^1 = \{ \bar{u}_{j-\frac{1}{2}}, \bar{u}_{j+\frac{1}{2}} \} \rightarrow \text{ENO} \rightarrow p(x) = P'(x).$$

$$\bar{u}[x_{j-\frac{1}{2}}, x_{j+\frac{1}{2}}] = \frac{\bar{u}_{j+\frac{1}{2}} - \bar{u}_{j-\frac{1}{2}}}{x_{j+\frac{1}{2}} - x_{j-\frac{1}{2}}} = \bar{u}_j.$$

For uniform mesh, only need to compute the undivided differences.

$$\bar{u}[x_i] = \bar{u}_i$$

$$\bar{u}[x_i, x_{i+1}] = \bar{u}_{i+1} - \bar{u}_i$$

$$\bar{u}[x_i, x_{i+1}, x_{i+2}] = \bar{u}[x_{i+1}, x_{i+2}] - \bar{u}[x_i, x_{i+1}]$$

Hence we choose the stencil as following

$$S_1 = \{ \bar{u}_j \}$$

$$S_2 = \begin{cases} \{ \bar{u}_{j-1}, \bar{u}_j \} & \text{if } |\bar{u}[x_{j-1}, x_j]| < |\bar{u}[x_j, x_{j+1}]| \\ \{ \bar{u}_j, \bar{u}_{j+1} \} & \text{otherwise} \end{cases}$$

...

Then for r-th order, need $i = -r+1$ to $N+r$

Code:

$$f[i, 0] = u(x_i), \quad i = -r+1, N+r$$

do $k=1, r$

do $i=-r+1, N+r-k$

$$f[i, k] = f[i+1, k-1] - f[i, k-1]$$

enddo

enddo

$$is(i) = i \quad i=1, \dots, N$$

do $k=1, r$

do $i=1, N$

if $|f(is(i)-1, k)| \leq |f(is(i), k)| : is(i) = is(i)-1$

enddo

enddo.

For example: $\tau=2$

$$\begin{array}{lll} is(j)=j-2 & \bar{U}_{j-2}, \bar{U}_{j-1}, \bar{U}_j & U_{j+1/2}^{-} = \frac{1}{3}\bar{U}_{j-2} - \frac{7}{6}\bar{U}_{j-1} + \frac{11}{6}\bar{U}_j ; U_{j+1/2}^{+} = -\frac{1}{6}\bar{U}_{j-2} + \frac{5}{6}\bar{U}_{j-1} + \frac{1}{3}\bar{U}_j \\ is(j)=j-1 & \bar{U}_{j-1}, \bar{U}_j, \bar{U}_{j+1} & U_{j+1/2}^{-} = -\frac{1}{6}\bar{U}_{j-1} + \frac{5}{6}\bar{U}_j + \frac{1}{3}\bar{U}_{j+1} ; U_{j+1/2}^{+} = \frac{1}{3}\bar{U}_{j-1} + \frac{5}{6}\bar{U}_j - \frac{1}{6}\bar{U}_{j+1} \\ is(j)=j & \bar{U}_j, \bar{U}_{j+1}, \bar{U}_{j+2} & U_{j+1/2}^{-} = \frac{1}{3}\bar{U}_j + \frac{5}{6}\bar{U}_{j+1} - \frac{1}{6}\bar{U}_{j+2} ; U_{j+1/2}^{+} = \frac{11}{6}\bar{U}_j - \frac{7}{6}\bar{U}_{j+1} + \frac{1}{3}\bar{U}_{j+2}. \end{array}$$

Weighted ENO (WENO)

$$ENO: U_{j+1/2}^{(0)} = \frac{1}{3}\bar{U}_{j-2} - \frac{7}{6}\bar{U}_{j-1} + \frac{11}{6}\bar{U}_j , U_{j+1/2}^{(1)} = -\frac{1}{6}\bar{U}_{j-1} + \frac{5}{6}\bar{U}_j + \frac{1}{3}\bar{U}_{j+1} , U_{j+1/2}^{(2)} = \frac{1}{3}\bar{U}_j + \frac{5}{6}\bar{U}_{j+1} - \frac{1}{6}\bar{U}_{j+2}$$

ENO has the problem of losing accuracy at least 1 order.

$$u_t + u_x = 0 \text{ with } e^{x-t}(e^x), u_j = (-1)^j, u(x, 0) = \sin x$$

Not using all stencils touched.

This motivates WENO.

$$U_{j+1/2} = w_1 U_{j+1/2}^{(0)} + w_2 U_{j+1/2}^{(1)} + w_3 U_{j+1/2}^{(2)}, w_i \geq 0, \sum w_i = 1.$$

How to choose w_i ?

(1) "Easy" to compute

(2) If all 3 stencils are smooth (u is smooth on the 3 stencils), "higher" accuracy: 5th

(3) If one of the 3 stencils is not smooth, then that stencil gets a small w .

If we take $w_i = \gamma_i$ (linear weights), $\gamma_1 = \frac{1}{10}, \gamma_2 = \frac{3}{5}, \gamma_3 = \frac{3}{10}$, then (2) is achieved.

So we choose $w_i = \frac{\gamma_i}{(\rho_i + \varepsilon)^2}$. $\varepsilon = 10^{-6}$, ρ_i : smoothness indicators: smaller \rightarrow smoother

$$w_i = \frac{\gamma_i}{\sum w_i} \quad \rho_i = \sum_{k=1}^r \Delta X^{2k-1} \int_{I_j} \left(\frac{\partial^k p(x)}{\partial x^k} \right)^2 dx$$

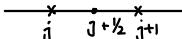
By Guangshan Jiang,

$$\rho_1 = \frac{13}{12} (\bar{U}_{j-2} - 2\bar{U}_{j-1} + \bar{U}_j)^2 + \frac{1}{4} (\bar{U}_{j-2} - 4\bar{U}_{j-1} + 3\bar{U}_j)^2$$

$$\rho_2 = \frac{13}{12} (\bar{U}_{j-1} - 2\bar{U}_j + \bar{U}_{j+1})^2 + \frac{1}{4} (\bar{U}_{j-1} - \bar{U}_{j+1})^2$$

$$\rho_3 = \frac{13}{12} (\bar{U}_j - 2\bar{U}_{j+1} + \bar{U}_{j+2})^2 + \frac{1}{4} (3\bar{U}_j - 4\bar{U}_{j+1} + \bar{U}_{j+2})^2$$

Mirror symmetric w.r.t. $j+1/2$ for $U_{j+1/2}^{-}$ & $U_{j+1/2}^{+}$



HW #9

(1) Repeat HW #8 with 3rd order ENO

(2) Repeat (1) with 5th order WENO

(i) The same st as before

(ii) Take st = $C \Delta x^{\frac{5}{3}}$

Lecture 10

Review: $u_t + f(u)_x = 0 \quad (1D)$

- weak solution, entropy condition
- L' contraction
- monotone schemes
- high-order FV schemes
 - reconstruction $\begin{cases} TVD/TVB \\ ENO/WENO \end{cases}$
 - monotone flux
 - time discretization: SSP-RK3

In 2D cases, FVM becomes expensive. Alternative: FDM.

FDM:

$$u_j \approx u(x_j, t). \frac{du}{dt} + f(u)_x = 0$$

How to approximate flux?

- Use u_{j-1}, u_j, u_{j+1} to interpolate a polynomial, then take derivative. NOT conservative
- $f(u) = \frac{\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}}}{\Delta x} + O(\Delta x^r)$. $f(u_{j+\frac{1}{2}})$ can be obtained by interpolation.

But even if $f(u_{j+\frac{1}{2}})$ is exact, r can be at most 2.

Suppose there is a function $h(x)$ which satisfies

$$f(u) = \frac{1}{\Delta x} \int_{x-\Delta x/2}^{x+\Delta x/2} h(\xi) d\xi.$$

Then $\frac{d}{dx} f(u(x)) = \frac{1}{\Delta x} (h(x + \frac{\Delta x}{2}) - h(x - \frac{\Delta x}{2}))$. $(f(u(x)))_x|_{x=x_j} = \frac{1}{\Delta x} (h(x_j + \frac{\Delta x}{2}) - h(x_j - \frac{\Delta x}{2}))$.

This is Shu-Osher Lemma. So just need to get h .

We know $f(u_i) = \frac{1}{\Delta x} \int_{x_j-\frac{\Delta x}{2}}^{x_j+\frac{\Delta x}{2}} h(\xi) d\xi = \bar{f}_j$. Then $\bar{f}_j \Rightarrow h(x_j \pm \frac{\Delta x}{2})$ is the reconstruction.

This allows higher accuracy?

Stability:

Separable flux: $\hat{f}(u^-_i, u^+_i) = \hat{f}^+(u^-_i) + \hat{f}^-(u^+_i)$. $\frac{\partial \hat{f}^+(u)}{\partial u} \geq 0$, $\frac{\partial \hat{f}^-(u)}{\partial u} \leq 0$.

Lax-Friedrichs is separable.

Example:

$$u_i \rightarrow f(u_i)$$

WENO5:

use $\hat{f}^+(u_{j-2}), \hat{f}^+(u_{j-1}), \hat{f}^+(u_j), \hat{f}^+(u_{j+1}), \hat{f}^+(u_{j+2})$ reconstruct $\hat{f}_{j+\frac{1}{2}}^-$

$\hat{f}^-(u_{j-1}), \hat{f}^-(u_{j-2}), \hat{f}^-(u_j), \hat{f}^-(u_{j+1}), \hat{f}^-(u_{j+2})$ reconstruct $\hat{f}_{j-\frac{1}{2}}^+$

$$\hat{f}_{j+\frac{1}{2}} = \hat{f}_{j+\frac{1}{2}}^- + \hat{f}_{j+\frac{1}{2}}^+$$

Flux-splitting: $f = f^+ + f^-$.

2D Case: $u_t + f(u)_x + g(u)_y = 0$.

- PDE properties: same as in 1D.
- monotone schemes: same as in 1D.

Notation: \tilde{u}_{ij} : - x-integration, \bar{u} : y-integration

$$\tilde{u}_{ij} := \frac{1}{\Delta x} \frac{1}{\Delta y} \int_{y_{j-\frac{1}{2}}}^{y_{j+\frac{1}{2}}} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} u(x, y, t) dx dy \quad \tilde{u}_{ij} : \text{operator on } u.$$

Apply on both sides of PDE.

$$\tilde{u}_{ij} + \frac{\tilde{f}_{i+\frac{1}{2}, j} - \tilde{f}_{i-\frac{1}{2}, j}}{\Delta x} + \frac{\tilde{g}_{i, j+\frac{1}{2}} - \tilde{g}_{i, j-\frac{1}{2}}}{\Delta y} = 0.$$

where

$$\tilde{f}_{i+\frac{1}{2}, j} = \frac{1}{\Delta y} \int_{y_{j-\frac{1}{2}}}^{y_{j+\frac{1}{2}}} f(u(x_{i+\frac{1}{2}}, y, t)) dy = \tilde{f}_{i+\frac{1}{2}, j}$$

$$\tilde{g}_{i, j+\frac{1}{2}} = \frac{1}{\Delta x} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} g(u(x, y_{j+\frac{1}{2}}, t)) dx = \tilde{g}_{i, j+\frac{1}{2}}.$$

Linear case:

If $f(u) = u$, $g(u) = u$ (constant u is also good), then $\tilde{f}_{i+\frac{1}{2}, j} = \tilde{u}_{i+\frac{1}{2}, j}$, $\tilde{g}_{i, j+\frac{1}{2}} = \tilde{u}_{i, j+\frac{1}{2}}$.

$\tilde{u}_{ij} \rightarrow \tilde{u}_{i+\frac{1}{2}, j}$ is like 1-dim reconstruction (line by line).

The cost is 1D reconstruction.

Nonlinear case:

In general, $\tilde{f}_{i+\frac{1}{2}, j} = \frac{1}{\Delta y} \int_{y_{j-\frac{1}{2}}}^{y_{j+\frac{1}{2}}} f(u(x_{i+\frac{1}{2}}, y, t)) dy \overset{\text{cannot pull out } f}{\not\equiv} f\left(\frac{1}{\Delta y} \int_{y_{j-\frac{1}{2}}}^{y_{j+\frac{1}{2}}} u(x_{i+\frac{1}{2}}, y, t) dy\right) \rightarrow \bar{u}, \text{ comes from 1D reconstruction}$

So need to use two reconstruction and quadrature: $\bar{u} \xrightarrow{Y} \bar{u} \xrightarrow{R} u \rightarrow f(u) \xrightarrow{Q} \tilde{f}$

In fact, $\tilde{u}_{ij} = u_{ij} + O(\Delta x^2)$ ($\Delta x = \Delta y$). So we can identify \tilde{u}_{ij} and u_{ij} up to 2nd order.

H.W. #10

Use 5-th FD WENO with L-F flux-splitting to repeat HW#9.

$$u_t + f(u)_x + g(u)_y = 0$$

- FV: hard to program and expensive
- FD: Easy to generalize from 1D.

$$u_{ij} \approx u(x_i, y_j, t),$$

$$\frac{du_{ij}}{dt} + \frac{1}{\Delta x} (\hat{f}_{i+\frac{1}{2}, j} - \hat{f}_{i-\frac{1}{2}, j}) + \frac{1}{\Delta y} (\hat{g}_{i, j+\frac{1}{2}} - \hat{g}_{i, j-\frac{1}{2}}) = 0.$$

1D algorithm with fixed j

...

Lecture 11

$U_t + f(U)_x = 0$. U is a vector, $U = \begin{pmatrix} U_1 \\ \vdots \\ U_m \end{pmatrix}$. $f(\cdot)$ is a vector function. $f(U) = \begin{pmatrix} f_1(U_1, \dots, U_m) \\ \vdots \\ f_m(U_1, \dots, U_m) \end{pmatrix}$

$Df(f') = \begin{pmatrix} \frac{\partial f_1}{\partial U_1} & \cdots & \frac{\partial f_1}{\partial U_m} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial U_1} & \cdots & \frac{\partial f_m}{\partial U_m} \end{pmatrix}$ should have real eigenvalues and a complete set of eigenvectors.

This is called hyperbolic system

Linear case: $f(U) = AU$. A has eigenvalues $\lambda_1 \leq \dots \leq \lambda_m$ and a complete set of eigenvectors v_1, \dots, v_m .

$(Av_1, \dots, Av_m) = (U_1v_1, \dots, U_mv_m)$. $AR = RA$. ($\Lambda = \text{diag}(\lambda_1, \dots, \lambda_m)$).

$A = R\Lambda R^{-1}$. $U_t + AU_x = 0$. $U_t + R\Lambda R^{-1}U_x = 0 \Rightarrow (R^{-1}U)_t + \Lambda(R^{-1}U)_x = 0$.

Let $V = R^{-1}U$. Then $\begin{cases} (V_1)_t + \lambda_1(V_1)_x = 0 \\ \vdots \\ (V_m)_t + \lambda_m(V_m)_x = 0 \end{cases}$. $U = RV$.

General case:

f is smooth w.r.t. U . $U_t + f'(U)U_x = 0$. $f'(U) = A(U)$. $R^{-1}(U)U_t + \Lambda(U)R^{-1}(U)U_x = 0$.

① Assuming smoothness ② In general $(R^{-1}(U)U)_t \neq R^{-1}(U)U_t \dots$.

For the componentwise eqn, $\begin{cases} (U_1)_t + f_1(U_1, \dots, U_m)_x = 0 \\ \vdots \\ (U_m)_t + f_m(U_1, \dots, U_m)_x = 0 \end{cases}$

For FD, need flux splitting, $\frac{\partial f^+(U)}{\partial U} \geq 0$, $\frac{\partial f^-(U)}{\partial U} \leq 0$. (All eigenvalues positive & negative).

If use L-F flux, $f^\pm(U) = \frac{1}{2}(f(U) \pm \alpha(U))$. Choose α large enough, then $\frac{\partial f^+(U)}{\partial U} \geq 0$, $\frac{\partial f^-(U)}{\partial U} \leq 0$ are satisfied.

This is oscillatory b/c single shock is hard to happen (hard to follow R-H vector condition).

So discontinuity will propagate at different speed.

Local characteristic decomposition

$R^{-1}(U) = \begin{pmatrix} l_i(U) \\ \vdots \\ b_m(U) \end{pmatrix}$. l_i is left eigenvectors of $A(U)$. $b_m(U)A(U) = \lambda_m(U)b_m(U)$.

Given $U_{j+1/2}$, to compute $f_{j+1/2}$, we proceed as follows

(1) Compute an average of U_j & U_{j+1} , denote by $U_{j+1/2}$, e.g. $U_{j+1/2} = \frac{1}{2}(U_j + U_{j+1})$, or the Roe average (by mean value thm)

$$f(U_{j+1}) - f(U_j) = f'(U_{j+1/2})(U_{j+1/2} - U_j)$$

(2) $f'(U_{j+1/2})$, $R_{j+1/2} = R(U_{j+1/2})$, $R_{j+1/2}^{-1} = R^{-1}(U_{j+1/2})$, $\Lambda_{j+1/2} = \Lambda(U_{j+1/2})$

(3) Project $f(U_i)$ & U_i to the local characteristic fields. $V_i = R_{j+1/2}^{-1}U_i$. $i = j-2, \dots, j+3$ (depending on algorithm)
 $q_i = R_{j+1/2}^{-1}f(U_i)$.

(4) Compute the flux splitting in each field

let $g_i = \begin{pmatrix} g_1 \\ \vdots \\ g_m \end{pmatrix}$, then $(g_\ell)_i^\pm = \frac{1}{2}((g_\ell)_i \pm \alpha_\ell(v_\ell)_i)$. $\ell = 1, \dots, m$. $i = j-2, \dots, j+3$.

$$\alpha_\ell = \max_u |\lambda_\ell(u)|$$

Feed it into the scalar routines. $(\hat{g}_\ell)_{j+1/2}^\pm = (\hat{g}_\ell)_{j+1/2}^\pm + (\hat{g}_{\ell+1})_{j+1/2}^\pm$.

(5) Project back to $\hat{f}_{j+1/2} = R_{j+1/2}\hat{g}_{j+1/2}$.

Example: Euler Equations of gas dynamics

$\mathbf{U} = \begin{pmatrix} \rho \\ m \\ E \end{pmatrix}$ $m = \rho v$, momentum. $E = \frac{P}{\gamma-1} + \frac{1}{2} \rho v^2$. P : pressure. Here we assume ideal gas.
 $\gamma = 1.4$ for air at room temperature.

$$f(u) = \begin{pmatrix} m \\ \rho v^2 + p \\ v(E+p) \end{pmatrix} = \begin{pmatrix} \frac{m}{\rho} + (\gamma-1)(E - \frac{1}{2} \frac{m^2}{\rho}) \\ \frac{m}{\rho} (\gamma-1)(E + (\gamma-1)(E - \frac{1}{2} \frac{m^2}{\rho})) \end{pmatrix}$$

$$\Lambda(u) = \begin{pmatrix} v-c & v & v+c \end{pmatrix}, \quad c = \sqrt{\frac{\gamma P}{\rho}} \text{ is the sound speed. (CW Shu, Applied Numerical Math, vol 9, 1992, pp 45-71).}$$

$$R(u) = \begin{pmatrix} 1 & 1 & 1 \\ v-c & v & v+c \\ H-vc & \frac{1}{2}v^2 & H+vc \end{pmatrix} \quad H = \frac{E+P}{\rho} \text{ is the enthalpy.}$$

$$R^{-1}(u) = \frac{1}{2} \begin{pmatrix} b_2 + \frac{v}{c} & b_1 v + \frac{1}{c} & b_1 \\ 2(1-b_1) & 2b_1 v & -2b_1 \\ b_2 - \frac{v}{c} & -(b_1 v - \frac{1}{c}) & b_1 \end{pmatrix}, \quad b_1 = \frac{c-1}{c^2}, \quad b_2 = \frac{1}{2} v^2 b_1.$$

H.W. #11

Solve $\left\{ \begin{array}{l} u_t + \left(\frac{u^2}{2}\right)_x + \left(\frac{u^3}{2}\right)_y = 0 \\ u(x, y, 0) = \frac{1}{3} + \frac{2}{3} \sin\left(\frac{x+y}{2}\right). \end{array} \right. \quad \text{on } [0, 4\pi] \times [0, 4\pi].$

Error table & figures at $t=0.3$ & $t=2$. WENO5, FD.

Lecture 12

1D system: $\partial_t + f(u)x = 0$.

$$u = \begin{pmatrix} u_1 \\ \vdots \\ u_m \end{pmatrix} \quad f(u) = \begin{pmatrix} f_1(u) \\ \vdots \\ f_m(u) \end{pmatrix} \quad f'(u) = \begin{pmatrix} \frac{\partial f_1}{\partial u_1} & \cdots & \frac{\partial f_1}{\partial u_m} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial u_1} & \cdots & \frac{\partial f_m}{\partial u_m} \end{pmatrix}$$

real eigenvalues $\lambda_1(u) \leq \lambda_2(u) \leq \cdots \leq \lambda_m(u)$
complete set of eigenvectors: $\phi_1(u), \dots, \phi_m(u)$.

Euler equations:

$$u = \begin{pmatrix} \rho \\ m \\ E \end{pmatrix} \quad f(u) = \begin{pmatrix} m \\ \rho v^2 + p \\ v(E+p) \end{pmatrix}$$

ρ : density m : momentum E : total energy
 v : velocity p : pressure

 $\nu = \frac{m}{\rho} \quad E = \frac{P}{\gamma-1} + \frac{1}{2}\rho v^2 \quad \gamma = 1.4 \quad \text{for air}$

Then for the $f'(u)$

$$\lambda_1(u) = v - c, \quad \lambda_2(u) = v, \quad \lambda_3(u) = v + c. \quad c = \sqrt{\frac{\rho P}{\rho}}. \quad \text{sound speed. } P > 0, \rho > 0.$$

Algorithm flowchart of 5th FD WENO

Given $\{u_j\}$, $\{f(u_j)\}$

1. Compute eigenvalues $\lambda_1(u)$, $\lambda_2(u)$, $\lambda_3(u)$ for all $u = u_j$. $\alpha = \max_{k,u} |u_k(u)|$. Choose st. s.t. $\alpha \frac{\Delta t}{\Delta x} \leq \text{CFL}$.
2. Compute a "middle stage" $u_{j+\frac{1}{2}}$, an average of u_j & u_{j+1} , e.g. Roe average (P.L.Roe, JCP, 63, 1981, pp357-372).
 $t_0 = \frac{\sqrt{\rho_j}}{\sqrt{\rho_j} + \sqrt{\rho_{j+1}}}$, $t_1 = 1 - t_0 = \frac{\sqrt{\rho_{j+1}}}{\sqrt{\rho_j} + \sqrt{\rho_{j+1}}}$, $V_{xm} = t_0 V_j + t_1 V_{j+1}$, $H_i = \frac{\rho_i + E_i}{\rho_i}$, $i = j, j+1$. $H_{xm} = t_0 H_j + t_1 H_{j+1}$.
 $q_m = \frac{1}{2} V_{xm}^2$, $C_m = \sqrt{(\gamma-1)(H_{xm} - q_m)}$.

Let $t_2 = V_{xm} - C_m$. $R = \begin{pmatrix} 1 & 1 & 1 \\ V_{xm} - C_m & V_{xm} & V_{xm} + C_m \\ H_{xm} - t_2 & q_m & H_{xm} + t_2 \end{pmatrix}$

Let $\tau_{cm} = \frac{1}{C_m}$, $b_1 = (\gamma-1) \tau_{cm}^2$, $b_2 = q_m \cdot b_1$, $t_0 = V_{xm} \cdot \tau_{cm}$, $t_1 = b_1 \cdot V_{xm}$, $t_2 = 0.5 b_1$.

$$L = \begin{pmatrix} \frac{1}{2}(b_1 + t_0) & -\frac{1}{2}(t_1 + \tau_{cm}) & t_2 \\ 1 - b_2 & t_1 & -b_1 \\ \frac{1}{2}(b_2 - t_0) & -\frac{1}{2}(t_1 - \tau_{cm}) & t_2 \end{pmatrix} \quad (\text{or } u_{j+\frac{1}{2}} = u_j + u_{j+1})$$

3. Local characteristics decomposition:

$$w_i = L u_i, \quad g_i = L f(u_i), \quad i = j-2, j-1, j, j, j+1, j+2, j+3.$$

4. $g_i^* = \frac{1}{2} (g_i \pm (\alpha_1 \alpha_3) w_i)$: L-F flux, private α for each component. $\alpha = \max_u |u_k(u)|$.
 $\alpha_1 = \max_i |v_i - c_i|$, $\alpha_2 = \max_i |v_i|$, $\alpha_3 = \max_i |v_i + c_i|$.

5. Use the scalar WENO to compute $\hat{g}_{i+\frac{1}{2}}^\pm$. $\Rightarrow g_{i+\frac{1}{2}} = \hat{g}_{i+\frac{1}{2}}^+ + \hat{g}_{i+\frac{1}{2}}^-$

$$\hat{g}_{i+\frac{1}{2}} = R \hat{g}_{j+\frac{1}{2}}$$

H.W. #12

Use 5th FD WENO with SSP-RK3 to compute the Euler equation

(i) Initial condition: $\rho(x) = 1 + 0.2 \sin x$, $v(x, 0) = 1$, $p(x, 0) = 1$, $\gamma = 1.4$. (Periodic BC).

Exact solution: $\rho(x, t) = 1 + 0.2 \sin(x - vt)$, $v = 1$, $p = 1$.

$$(ii) y=3. \quad \ell(x,0)=1+\frac{1}{2}\sin x, \quad v(x,0)=\frac{1}{2}\sin x, \quad p(x,0)=(1+\frac{1}{2}\sin x)^2$$

Exact solution can be obtained by the fact that $w=v-\sqrt{3}\ell$ and $z=v+\sqrt{3}\ell$ both satisfy the standard Burgers equations

(iii) Shock tube problem: $\Omega = [-5, 5]$

$$\text{Sod: } t=2. \quad \begin{pmatrix} \ell(x,0) \\ v(x,0) \\ p(x,0) \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \quad x \leq 0, \quad \begin{pmatrix} 0.125 \\ 0 \\ 0.1 \end{pmatrix} \quad x > 0$$

$$\text{Lax: } t=1.3 \quad \dots = \begin{pmatrix} 0.445 \\ 0.698 \\ 3.528 \end{pmatrix} \quad \dots \quad \begin{pmatrix} 0.5 \\ 0 \\ 0.571 \end{pmatrix} \quad \dots$$

BC: Use ghost points. $u_{ij}^l = u(0) \cdot j = -1, -2, \dots$ or $u_{ij}^r = \text{init condition}$

2D Systems

$$u_t + f(u)_x + g(u)_y = 0$$

$$u = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_m \\ u_m \end{pmatrix}, \quad f(u) = \begin{pmatrix} f_1(u) \\ \vdots \\ f_m(u) \end{pmatrix}, \quad g(u) = \begin{pmatrix} g_1(u) \\ \vdots \\ g_m(u) \end{pmatrix}$$

Hyperbolic: $\lambda_1 f'(u) + \lambda_2 g'(u)$ has real eigenvalues and a complete set of eigenvectors $V \lambda_1, \lambda_2 \in \mathbb{R}$.

2D Euler:

$$u = \begin{pmatrix} \rho \\ m_1 \\ m_2 \\ E \end{pmatrix}, \quad f(u) = \begin{pmatrix} m_1 \\ \rho v_1 + p \\ \rho v_2 \\ V_1(E+p) \end{pmatrix}, \quad g(u) = \begin{pmatrix} m_2 \\ \rho v_2 \\ \rho v_1 + p \\ V_2(E+p) \end{pmatrix} \quad V_1 = \frac{m_1}{\rho}, \quad V_2 = \frac{m_2}{\rho}, \quad E = \frac{1}{\gamma-1} p + \frac{1}{2} \rho (v_1^2 + v_2^2)$$

Eigenvalues: $f'(u)$: $v_1 - C, v_1, v_1, v_1 + C$; $g'(u) = v_2 - C, v_2, v_2, v_2 + C$.

Eigenvectors: in the paper

FD WENO Scheme:

Given $\{u_{ij}\}$, hence also $\{f(u_{ij})\}, \{g(u_{ij})\}$.

For $j=1:N_y$,

fix j . $u_i = u_{ij}$, $f_{xi} = f(u_{ij})$. use the 1D algorithm based on $\{u_{ij}\}$ & $\{f_{xi}\}$.

we can get the approximation: $\frac{1}{2\Delta x} (\hat{f}_{xi+\frac{1}{2},j} - \hat{f}_{xi-\frac{1}{2},j})$ to $f(u)_x$ at (x_i, y_j) .

For $i=1:N_x$

fix i . $u_j = u_{ij}$, $g_{xj} = g(u_{ij})$. use the 1D algorithm based on $\{u_{ij}\}$ & $\{g_{xj}\}$.

we can get the approximation: $\frac{1}{2\Delta y} (\hat{g}_{y_i,j+\frac{1}{2}} - \hat{g}_{y_i,j-\frac{1}{2}})$ to $g(u)_y$ at (x_i, y_j) .

Theoretical properties of 1D system

$$\begin{cases} u_t + f(u)_x = 0 \\ u(x, 0) = u^0(x) \end{cases} \text{ is hyperbolic}$$

$\lambda(u)$: eigenvalue, $\eta(u)$ eigenvector $(\nabla \lambda(u)) \cdot \eta$ is the directional derivative along the eigenvector direction
 $(\nabla \lambda(u)) \cdot \eta = \begin{cases} > 0 \text{ (or } < 0\text{)} : \text{genuinely non-linear} \\ = 0 : \text{linearly degenerate} \end{cases}$

For the Euler equations in 1D,

1st & 3rd fields (with $v-c, v+c$) are genuinely nonlinear.

2nd field (with v) is linearly degenerate.

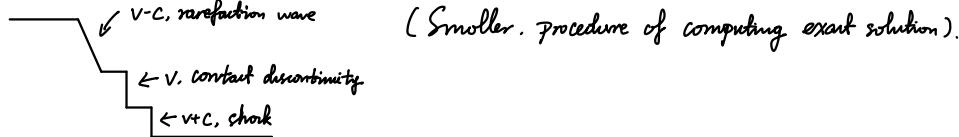
With these properties, we can solve the Riemann problems with entropy solutions for small data (in general) and big data for Euler.

Lax Entropy condition:



$$\begin{aligned} \lambda_1(u_l) &\leq \dots \leq \lambda_k(u_l) \leq \dots \leq \lambda_m(u_l) \\ \lambda_1(u_r) &\leq \dots \stackrel{s}{\leq} \lambda_k(u_r) \leq \dots \leq \lambda_m(u_r). \end{aligned}$$

Example: Sod



Lecture 13

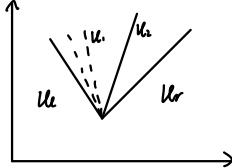
For the genuinely nonlinear characteristic field,

$$\lambda_1(u_r) \leq \lambda_2(u_r) \leq \cdots \leq \lambda_k(u_r) \leq \lambda_{k+1}(u_r) \leq \cdots \leq \lambda_m(u_r)$$

$$\lambda_1(u_e) \leq \lambda_2(u_e) \leq \cdots \leq \lambda_k(u_e) \leq \lambda_{k+1}(u_e) \leq \cdots \leq \lambda_m(u_e)$$

Lax condition: $\lambda_k(u_r) \leq s \leq \lambda_k(u_e)$.

Riemann problem:



Lax: Genuinely nonlinear or linearly degenerate, given small data, then $\exists u_r, u_e$ connecting u_r, u_e .

If we have the Riemann solver, then we can immediately write the Godunov scheme.

By divergence theorem, $\bar{u}_j^{nn} = \bar{u}_j + \frac{\Delta t}{\Delta x} (\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}})$. Riemann solver.

Glimm proves $\begin{cases} u_t + f(u)_x = 0 \\ u(x, 0) = u^0(x) \end{cases}$, $u^0(x)$ is small w.r.t TV-semi-norm has a solution.

The proof uses numerical schemes to construct a convergent sequence, which converges to a solution. Sample randomly from the true solution.

Boundary condition

$$\begin{cases} u_t + u_x = 0 \\ u(x, 0) = u^0(x) \\ u(0, t) = g(t). \end{cases}$$

For 1st order scheme, $\bar{u}_j^{nn} = \bar{u}_j - \frac{\Delta t}{\Delta x} (\hat{f}_{j+\frac{1}{2}} - \hat{f}_{j-\frac{1}{2}}) = \bar{u}_j - \lambda (\bar{u}_j^n - \bar{u}_{j-1}^n)$, where $\hat{f}_{j+\frac{1}{2}} = \bar{u}_j^n$.

But if we want to use high order schemes, we need 0, -1, -2, n+1, n+2, n+3.

Extrapolation: accuracy is fine. For stability, outflow extrapolation is fine, but not at inflow.

One order lower at boundary is usually fine. But alignment is a problem.

Inverse Lax-Wendroff procedure (ILW)

$u_t + u_x = 0$. RK3 use 7 points instead of 3. Not compact.

$$LW: \bar{u}_j^{nn} = \bar{u}_j^n + (u_{t,j}) \frac{1}{2} \Delta t + (u_{x,j}) \frac{1}{2} \Delta t^2 = \bar{u}_j^n - (u_{x,j}) \Delta t + (u_{xx,j}) \frac{1}{2} \Delta t^2.$$

$$\text{Now suppose } x_1 = x_0 + \alpha \Delta x. u_1 = u_0 + (u_x)_0 (\alpha \Delta x) + (u_{xx})_0 \frac{(\alpha \Delta x)^2}{2} = u_0 - (u_{x,0}) (\alpha \Delta x) + (u_{tt,0}) (\frac{\alpha \Delta x}{2})^2.$$

$$u_1 = u_0 - g'(t) \alpha \Delta x + g''(t) \frac{(\alpha \Delta x)^2}{2} + \dots$$

(Simplified ILW: do extrapolation for high-order derivatives).

For parabolic PDEs, like $u_t = u_{xx}$. $u_1 = u_0 + (u_x)_0 \Delta x + (u_{xx})_0 \frac{\Delta x^2}{2}$ u_x : extrapolation, u_{xx} : ILW.

For convection dominated PDEs, $u_x = \frac{1}{\alpha} (u_t - u_{xx}) \quad \alpha \gg 1$. the other term use extrapolation diffusion \dots $u_{xx} = u_t - \alpha u_x \quad \alpha \ll 1$.

Lecture 14

Bound-preserving

$$E = \frac{P}{\gamma-1} + \frac{1}{2} \rho v^2, \quad \frac{P}{\gamma-1}: \text{internal energy}, \quad \frac{1}{2} \rho v^2: \text{kinetic energy}$$

When $\text{Ma} \gg 1$, i.e. $v \gg c$, $P = (\gamma-1)(E - \frac{1}{2} \rho v^2)$. "--" can cause trouble.

$$\begin{cases} u_t + f(u)x = 0 \\ u(x, 0) = u_0(x). \end{cases}$$

$$m = \min_x u^\circ(x), \quad M = \max_x u^\circ(x). \quad \text{Then } m \leq u(x, t) \leq M. \quad \text{Maximum principle.}$$

Monotone schemes satisfies the maximum principle.

$$\text{Local maximal principle: } m_j = \min(u_{j-\frac{1}{2}}, \dots, u_{j+\frac{1}{2}}), \quad M_j = \max(u_{j-\frac{1}{2}}, \dots, u_{j+\frac{1}{2}}), \quad m_j \leq u_{j+\frac{1}{2}} \leq M_j$$

But this restricted the accuracy to 2nd order.

$$P_0: u_t + ux = 0$$

$$u(x, 0) = \sin x.$$

$$\text{Consider } x = \frac{\pi}{2}. \quad \text{Then } M_j = 1 - \Delta x^2. \quad u_{j+\frac{1}{2}}^{(0)} \approx 1 - \Delta x^2.$$

$$\text{Let } \Delta t = \frac{1}{2} \Delta x. \quad u_{j+\frac{1}{2}}^{(0, \text{exact})} = 1.$$

Boundary preserving high-order scheme

$$\bar{u}_j^{(m)} = \bar{u}_j^{(n)} - \lambda (\hat{f}(u_{j+\frac{1}{2}}, u_{j+\frac{1}{2}}) - \hat{f}(u_{j-\frac{1}{2}}, u_{j-\frac{1}{2}})) \\ = G(\bar{u}_j^{(n)}, u_{j+\frac{1}{2}}, u_{j+\frac{1}{2}}, u_{j-\frac{1}{2}}, u_{j-\frac{1}{2}}) = G(\uparrow, \downarrow, \uparrow, \uparrow, \downarrow).$$

$\bar{u}_j^{(n)}$, $u_{j+\frac{1}{2}}$, $u_{j-\frac{1}{2}}$ belongs to the same polynomial.

Idea: Use Gauss-Lobatto quadrature to compute \bar{u}_j .

For G-L quadrature, K-point is exact for polynomial degree $\leq 2K-3$.

$$\bar{u}_j^{(n)} = w_1 u_{j-\frac{1}{2}}^{+} + w_K u_{j+\frac{1}{2}}^{-} + \sum_{k=2}^{K-1} w_k P(X_j^k) - \lambda [\hat{f}(u_{j+\frac{1}{2}}, u_{j+\frac{1}{2}}) - \hat{f}(u_{j-\frac{1}{2}}, u_{j+\frac{1}{2}}) + \hat{f}(u_{j-\frac{1}{2}}, u_{j+\frac{1}{2}}) - \hat{f}(u_{j-\frac{1}{2}}, u_{j-\frac{1}{2}})]. \\ = w_1 [u_{j-\frac{1}{2}}^{+} - \frac{\lambda}{w_1} (\hat{f}(u_{j+\frac{1}{2}}, u_{j+\frac{1}{2}}) - \hat{f}(u_{j-\frac{1}{2}}, u_{j+\frac{1}{2}}))] + w_K [u_{j+\frac{1}{2}}^{-} - \frac{\lambda}{w_K} (\hat{f}(u_{j+\frac{1}{2}}, u_{j+\frac{1}{2}}) - \hat{f}(u_{j+\frac{1}{2}}, u_{j-\frac{1}{2}}))] + \sum_{k=2}^{K-1} w_k P(X_j^k).$$

Require: $\frac{\lambda}{w_1} \leq \lambda_0$, $\frac{\lambda}{w_K} \leq \lambda_0$, and $P(X_j^k) \in [m, M]$ for all G-L points.

So we have to apply limiter to make the $P(\cdot)$ satisfies boundary preserving

Limiter: ① $P(X_j^k) \in [m, M]$ ② Do not lose order.

$$p_j(x) = \bar{u}_j^{(n)} + (\bar{p}_j(x) - \bar{u}_j^{(n)}), \quad \tilde{p}_j(x) = \bar{u}_j^{(n)} + \theta(p_j(x) - \bar{u}_j^{(n)}). \quad \theta = \min_j \left\{ \frac{M - \bar{u}_j^{(n)}}{p_j(x) - \bar{u}_j^{(n)}}, 1 \right\} \text{ (and } \theta \text{ for min}).$$

Lemma: $\tilde{p}_j(x) - u(x) = O(\Delta x^r)$ if $p_j(x) = O(\Delta x^r)$.