

Chapter 10: Linear And Nonlinear Conservation Laws

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Mathematical Models, Analysis and Simulation, Part I

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Plan

- A simple upstream discretization of the transport equation
- Proof of convergence via Fourier transformations
- Nonlinear conservation laws
- Shock waves
- Unique solvability: viscosity solutions

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The problem

- Consider the pure initial value problem

$$u_t = cu_x, \quad x \in \mathbb{R}, \quad u(x, 0) = u^0(x)$$

with $c > 0$.

- The discretization is

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} = c \frac{u_{j+1}^n - u_j^n}{\Delta x}.$$

- Convergence:

Q: How to compare the discrete solution u_j^n and the continuous solution $u(x, t_n)$?

A: Use *interpolation* by trigonometric polynomials!

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Trigonometric Interpolation

- Let $\mathbf{f} = \{f_j | j \in \mathbb{Z}\}$ be an infinite sequence:

$$\tilde{f}(x) := \sum_{j=-\infty}^{+\infty} \operatorname{sinc} \frac{\pi}{h}(x - x_j) f_j, \quad x \in \mathbb{R}.$$

- Because of

$$\operatorname{sinc} \frac{\pi}{h}(x - x_j) = \frac{\sin \frac{\pi}{h}(x_k - x_j)}{\frac{\pi}{h}(x_k - x_j)} = \begin{cases} 0, & k \neq j \\ 1, & k = j \end{cases}$$

it holds

$$\tilde{f}(x_j) = f_j$$

\tilde{f} is a *trigonometric interpolant* of \mathbf{f} .

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Semi-Discrete Fourier Transform

- The *semi-discrete Fourier transform* is given by

$$\hat{f}(k) = h \sum_{j=-\infty}^{+\infty} f_j e^{-ikx_j}, k \in \mathbb{R}.$$

Notes:

- \hat{f} is defined on all of \mathbb{R} .
- Compared to the DFT, a different scaling is used.
- \hat{f} is $2\pi/h$ -periodic,

$$\hat{f}(k) = \hat{f}(k + 2\pi/h).$$

- The *band-limited* Fourier transform:

$$\check{f}(k) = \begin{cases} \hat{f}(k), & |k| < \pi/h, \\ 0, & \text{otherwise.} \end{cases}$$

Theorem.

$$\hat{\hat{f}}(k) \equiv \check{f}(k)$$

Proof of Theorem

- Define, for $a > 0$,

$$\chi_a(x) = \begin{cases} 1, & |x| < a, \\ 0, & \text{otherwise.} \end{cases}$$

- Fourier transform (Strang, p 310):

$$\widehat{\chi}_a(k) = 2a \frac{\sin ak}{ak}.$$

- Using the inversion theorem, one gets

$$\left(\frac{\sin bx}{bx}\right)^\wedge = \frac{\pi}{b} \chi_b(x)$$

Proof: Cont

$$\begin{aligned} \hat{\hat{f}}(k) &= \int_{-\infty}^{+\infty} \left(\sum_{j=-\infty}^{+\infty} \frac{\sin \frac{\pi}{h}(x-x_j)}{\frac{\pi}{h}(x-x_j)} f_j \right) e^{-ikx} dx \\ &= \sum_{j=-\infty}^{+\infty} f_j \int_{-\infty}^{+\infty} \frac{\sin \frac{\pi}{h}(x-x_j)}{\frac{\pi}{h}(x-x_j)} e^{-ikx} dx \\ &\stackrel{y=x-x_j}{=} \sum_{j=-\infty}^{+\infty} f_j e^{-ikx_j} \int_{-\infty}^{+\infty} \frac{\sin \frac{\pi}{h}y}{\frac{\pi}{h}y} e^{-iky} dy \\ &= \sum_{j=-\infty}^{+\infty} f_j e^{-ikx_j} h \chi_{\pi/h}(k) \\ &= \begin{cases} h \sum_{j=-\infty}^{+\infty} f_j e^{-ikx_j}, & |k| < \pi/h \\ 0, & \text{otherwise} \end{cases} \\ &= \check{f}(k) \end{aligned}$$

Convergence

$$\|u(\cdot, t_n) - \tilde{u}(\cdot, t_n)\| \xrightarrow{h, \Delta t \rightarrow 0} 0??$$

Continuous Fourier transform of $u_t = cu_x$, $u(\cdot, 0) = u^0$ yields

$$\frac{d\hat{u}}{dt} = ick\hat{u}, \quad \hat{u}(\cdot, 0) = \hat{u}^0.$$

Hence,

$$\hat{u}(k, t + \Delta t) = e^{ick\Delta t} \hat{u}(k, t).$$

Finally,

$$\hat{u}(\cdot, t) = H(\Delta t)^n \hat{u}(\cdot, 0) = H(k, \Delta t)^n \hat{u}^0$$

$$H(k, \Delta t) = e^{ick\Delta t}.$$

Convergence: Cont

Discrete Write the discretization like

$$u_j^{n+1} = u_j^n + r(u_{j+1}^n - u_j^n), \quad r = \frac{c\Delta t}{\Delta x}.$$

Semi-discrete Fourier transform

$$\hat{u}^{n+1} = \hat{u}^n + r(e^{ikh}\hat{u}^n - \hat{u}^n).$$

Finally,

$$\begin{aligned} \hat{u}^n &= G^n \hat{u}^0 \\ \check{u}^n &= G^n \check{u}^0, |kh| < \pi \end{aligned}$$

$$G = 1 + r(e^{-ikh} - 1).$$

The initial value can be chosen in different ways. We use the most obvious

$$u_j^0 = u^0(x_j).$$

Convergence: Cont

Remember Plancherel's and Parseval's identities:

$$\|\hat{u}\|^2 = 2\pi\|u\|^2, \quad 2\pi\|\check{u}^0\|^2 = \|\check{u}^0\|^2.$$

We obtain

$$\begin{aligned} \|u(\cdot, t_n) - \tilde{u}(\cdot, t_n)\| &= \frac{1}{\sqrt{2\pi}} \|\hat{u}(\cdot, t_n) - \check{u}^n\| \\ &= \frac{1}{\sqrt{2\pi}} \|H^n \hat{u}^0 - G^n \check{u}^0\| \\ &= \frac{1}{\sqrt{2\pi}} \|H^n \hat{u}^0 - H^n \check{u}^0 + H^n \check{u}^0 - G^n \check{u}^0\| \\ &\leq \frac{1}{\sqrt{2\pi}} (\|H^n(\hat{u}^0 - \check{u}^0)\| + \|(H^n - G^n)\check{u}^0\|) \end{aligned}$$

- The first term contains the propagation of the initial error (in frequency space). In the present case, $|H| = 1$.
- The first term becomes small if the approximation of the initial values is consistent:

$$\|\hat{u}^0 - \check{u}^0\| \rightarrow 0 \text{ for all } u^0 \in L^2(\mathbb{R}).$$

The Second Term

- Assume stability:

$$|G| < 1 \Leftrightarrow 0 < c\Delta t < h.$$

- Split the factors:

$$H^n - G^n = (H - G) \sum_{j=0}^{n-1} H^{n-j-1} G^j.$$

- Because of stability, it holds

$$\left| \sum_{j=0}^{n-1} H^{n-j-1} G^j \right| \leq \sum_{j=0}^{n-1} 1 \cdot |G^j| < n = \frac{t_n}{\Delta t}.$$

The Second Term: Cont

- Fix the wave number k now. Then

$$\begin{aligned} H - G &= e^{ick\Delta t} - (1 + r(e^{ikh} - 1)) \\ &= 1 + ick\Delta t + O((\Delta t)^2) - (1 + \frac{c\Delta t}{h}(1 + ikh + O(h^2)) - 1) \\ &= 1 + ick\Delta t + O((\Delta t)^2) - (1 + ick\Delta t + O((\Delta t)^2)) \\ &= O((\Delta t)^2) \end{aligned}$$

This is a consequence of first order consistency:

$$|H - G| = O((\Delta t)^{p+1})$$

- Putting everything together,

$$|H^n - G^n| = t_n \cdot O(\Delta t).$$

Conclusions

- We have just seen:

consistency + stability \Rightarrow convergence.

- Previous chapter:

no stability \Rightarrow no convergence.

- This is obvious:

convergence \Rightarrow consistency.

- Conclusion:

consistency + stability \Leftrightarrow convergence

This is one instance of the *Lax Equivalence Theorem*.

Nonlinear Conservation Laws: An Example

Read: Strang, Ch 6.6

Problem: Determine the flow of cars on a narrow street.

- Density ρ of cars: number of cars per unit length.
- Velocity of cars v depends on density: $v = v(\rho)$.

ρ small $\Rightarrow v$ large

ρ large $\Rightarrow v$ low

Normalization:

- Largest possible velocity: $v = 1$.
- Road full means $v = 0$. Let this be at $\rho = 1$.

- Simple velocity model: $v(\rho) = 1 - \rho$.
- Conservation: Flux is $\Phi = v\rho = (1 - \rho)\rho$,

$$\rho_t + \text{div}((1 - \rho)\rho) = 0.$$

- Initial data: $\rho(x, 0) = \rho^0(x)$.
- Solution by characteristics:

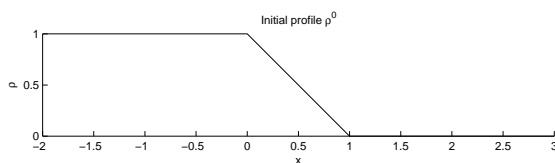
$$\frac{dx}{dt} = 1 - 2\rho, \quad \frac{d\rho(x(t), t)}{dt} = 0.$$

- ρ is constant along each characteristic.
- Since dx/dt is constant, the characteristics are *straight lines*.

Example: Smooth Solutions

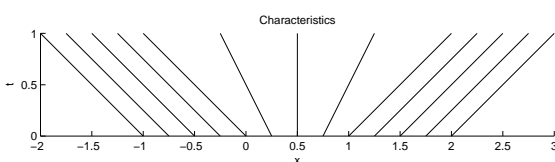
- Assume initial data:

$$\rho^0(x) = \begin{cases} 1, & x \leq 0 \\ 1 - x, & 0 < x < 1 \\ 0, & x \geq 1 \end{cases}$$



- Characteristics:

$$x(t) = \frac{dx}{dt} \Big|_{x_0} t + x_0 = (1 - 2\rho^0(x_0))t + x_0.$$



Example: Cont

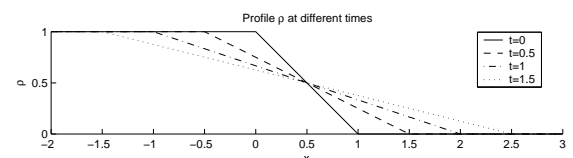
- The following possibilities arise:

case	density	solution
$x \geq t + 1$	$\rho^0 = 0$	$\rho = 0$
$x \leq -t$	$\rho^0 = 1$	$\rho = 1$
$-t < x < t + 1$	$\rho^0 = 1 - x_0$	

$$x = (1 - 2(1 - x_0))t + x_0$$

$$x_0 = \frac{x + t}{2t + 1}$$

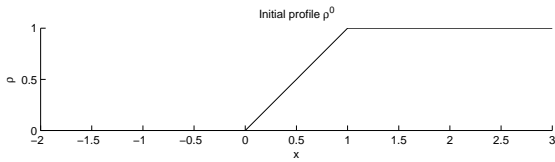
$$\rho = 1 - x_0 = \frac{1 + t - x}{2t + 1}$$



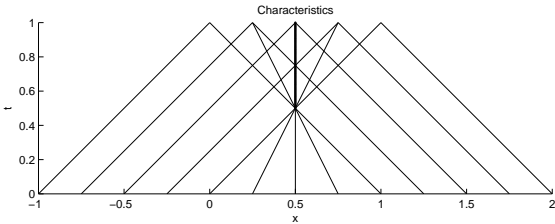
Example: Shocks

- Initial data:

$$\rho^0(x) = \begin{cases} 0, & x \leq 0 \\ x, & 0 < x < 1 \\ 1, & x \geq 1 \end{cases}$$



- Characteristics:



- What is the correct solution at crossing characteristics?
- A *shock* will form. **Where?**

Shocks: Cont

Compute the shock location $s(t)$: Integrate the equation on $[s - \varepsilon, s + \varepsilon]$:

$$\int_{s(t)-\varepsilon}^{s(t)+\varepsilon} \frac{\partial \rho}{\partial t} dx + \Phi(\rho(s(t) + \varepsilon)) - \Phi(\rho(s(t) - \varepsilon)) = 0$$

It holds

$$\frac{d}{dt} \int_{s(t)-\varepsilon}^{s(t)+\varepsilon} \rho dx = \int_{s(t)-\varepsilon}^{s(t)+\varepsilon} \frac{\partial \rho}{\partial t} dx + \rho(s(t) + \varepsilon) \frac{ds}{dt} - \rho(s(t) - \varepsilon) \frac{ds}{dt}$$

Taking the limit $\varepsilon \rightarrow 0$, this leads to

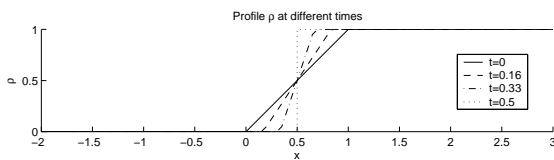
$$0 = \lim_{\varepsilon \rightarrow 0} \int_{s(t)-\varepsilon}^{s(t)+\varepsilon} \frac{\partial \rho}{\partial t} dx + \rho(s(t)+) \frac{ds}{dt} - \rho(s(t)-) \frac{ds}{dt}.$$

Rankine-Hugoniot Jump Condition

Result:

$$\frac{ds}{dt} = \frac{\Phi(\rho^+) - \Phi(\rho^-)}{\rho^+ - \rho^-}$$

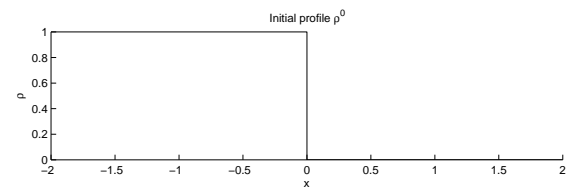
This is the famous *Rankine-Hugoniot jump condition*.



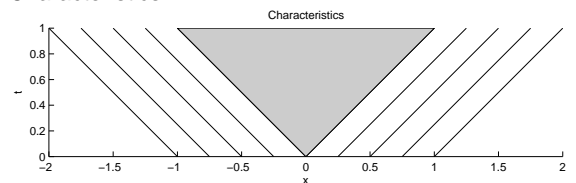
Entropy Solutions, Rarefaction Waves

- Consider initial data

$$\rho^0(x) = \begin{cases} 1, & x \leq 0 \\ 0, & x > 0 \end{cases}$$



- Characteristics



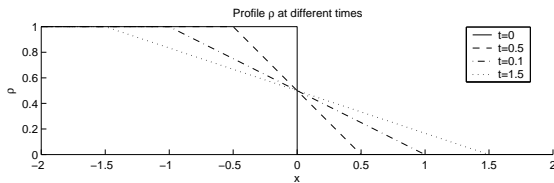
Entropy Solutions, Rarefaction Waves: Cont

There are (at least) two solutions which satisfy the Rankine-Hugoniot condition:

- $\rho(x, t) = \rho^0(x)$ for all $t > 0$.
- Rarefaction wave

$$\rho(x, t) = \begin{cases} 1, & x \leq -t \\ 0, & x \geq t \\ \frac{1}{2}\left(1 - \frac{x}{t}\right), & -t < x < t \end{cases}$$

Which one is the "correct" one?



In order to select one solution, we must use physical considerations.

Entropy Solutions, Rarefaction Waves: Cont

Idea: Modify the equation by adding small artificial diffusion:

$$v = 1 - \rho - \varepsilon \rho_x / \rho, 0 < \varepsilon \ll 1$$

(Say, drivers look ahead.)

Modified equation:

$$\rho_t + (\rho(1 - \rho))_x = \varepsilon \rho_{xx}$$

- *Entropy solution*: The limiting function $\rho = \lim_{\varepsilon \rightarrow 0} \rho_\varepsilon$
- All solutions are smooth. However, gradients may become large.
- The added diffusion is also necessary for numerical schemes.

Entropy Solutions, Rarefaction Waves: Cont

Numerical approximations: How to choose ε ?

1. $\varepsilon = O(h)$ gives nice, non-wiggly solutions, but only first-order accurate.
2. Choose $\varepsilon = \varepsilon(\rho_x)$, where

$$\varepsilon \sim \begin{cases} h, & \text{if } \rho_x \text{ is large} \\ 0, & \text{if } \rho_x \text{ is small} \end{cases}$$

This is called *switched artificial diffusion*.

- ON: near shocks,
- OFF: where the solution is smooth.