

5) The CFL (Courant-Lewy-Friedrichs) condition

Def. Let $\begin{cases} D_{\text{num}} = \text{numerical domain of dependence,} \\ D_{\text{exact}} = \text{analytical domain of dependence of the PDE.} \end{cases}$

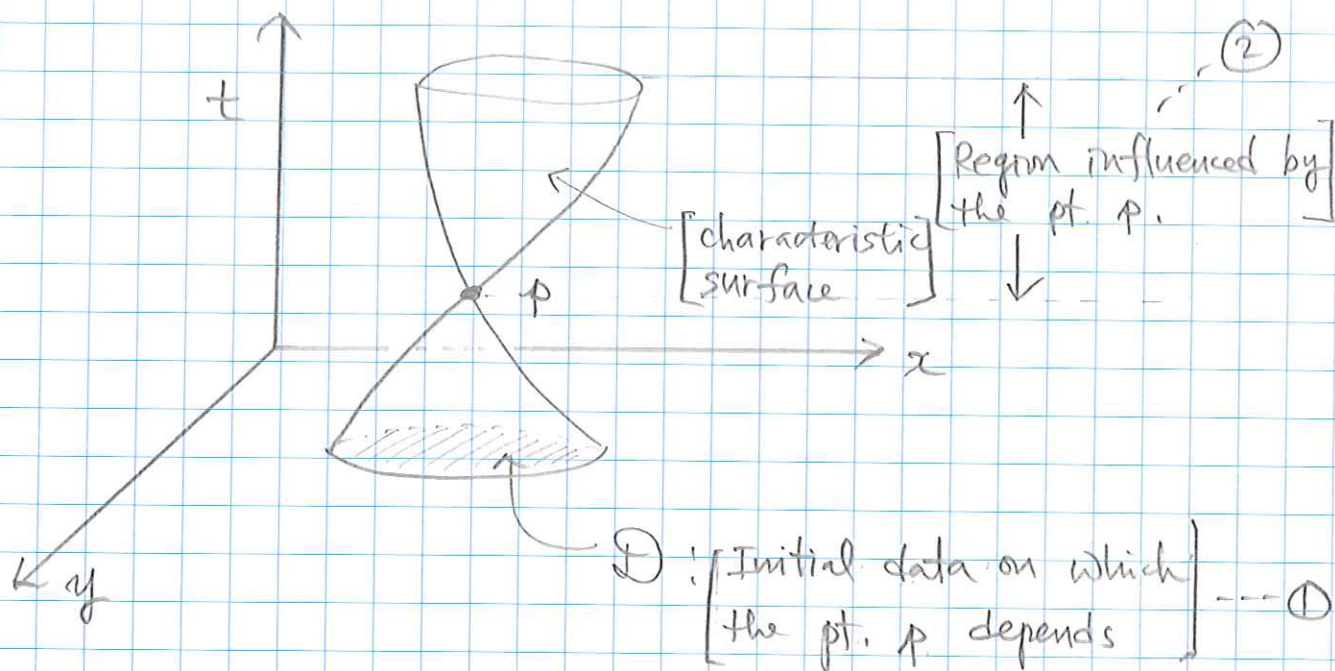
A numerical method can be convergent only if

$$D_{\text{num}} \supset D_{\text{exact}}.$$

This is called the CFL condition.

Hint. Recall that, in hyperbolic PDEs, we have:

- (i) domain of dependence, &
- (ii) range of influence.



\Rightarrow $\begin{cases} \textcircled{1} \text{ is the domain of dependence of } p, \\ \textcircled{2} \text{ is the region of influence of } p. \end{cases}$

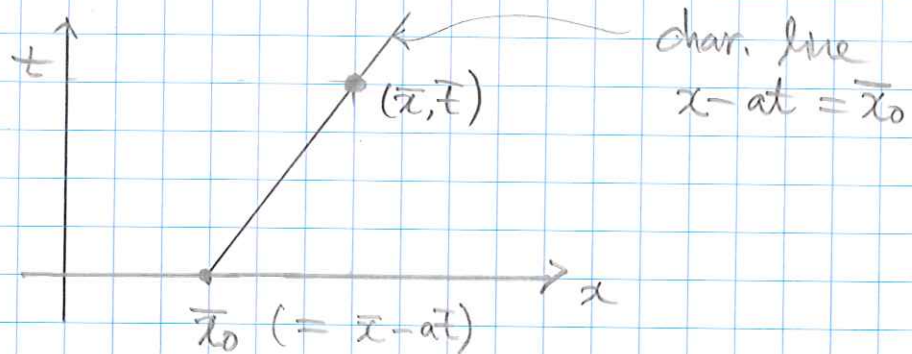
Rule. Mathematically, we state them as follows:

The soln $u(x,t)$ at any pt. (\bar{x}, \bar{t}) depends only on the initial data u^0 at a single pt., namely, \bar{x}_0 sit. (\bar{x}, \bar{t}) lies on the characteristic lines (or curves) through \bar{x}_0 .

Case 1: linear scalar constant coeff. ($u_t + au_x = 0$)

$\rightarrow \exists$ a single characteristic line, defined by

$$x - at = \bar{x}_0.$$



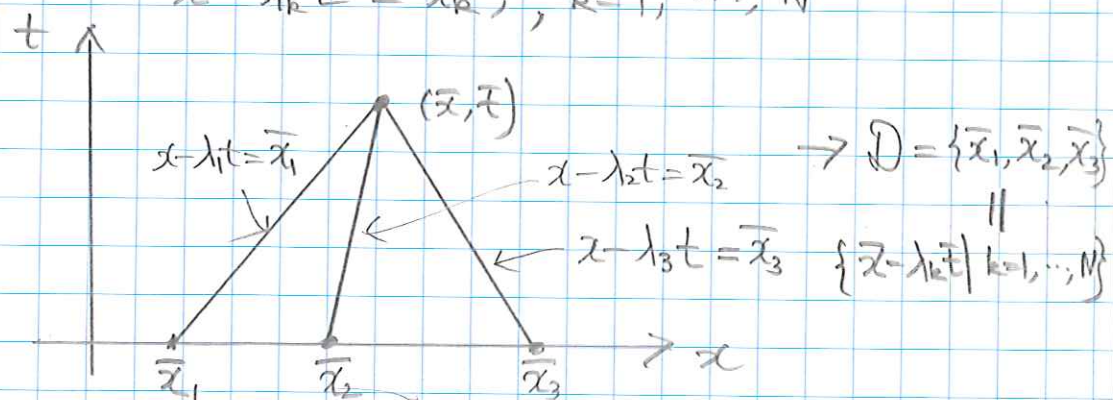
$\rightarrow \mathcal{D} = \{\bar{x}_0\}$; domain of dep. of (\bar{x}, \bar{t}) .

Case 2: system of N advection Eqns. ($\underbrace{u_t}_{N \times 1} + \underbrace{A}_{N \times N} \underbrace{u_x}_{N \times 1} = 0$)

$\rightarrow \exists \lambda_1, \dots, \lambda_N$; multiple eigenvalues of A

$\rightarrow \exists N$ many char. lines defined by

$$x - \lambda_k t = \bar{x}_k, \quad k=1, \dots, N$$

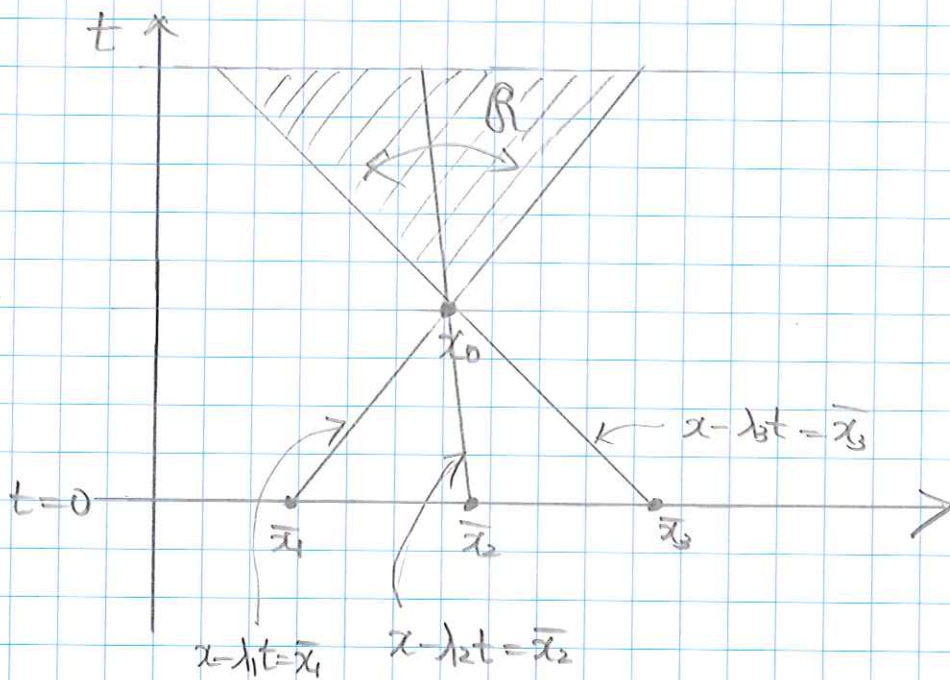


On the other hand, the range of influence is defined by

$$R = \{ (x,t) \mid \lambda_N t \leq (x-x_0) \leq \lambda_1 t \}, \text{ where}$$

$$\lambda_1 = \max_k \lambda_k,$$

$$\lambda_N = \min_k \lambda_k.$$



Remark, For $u_t = k u_{xx}$, $D = R = (-\infty, \infty)$

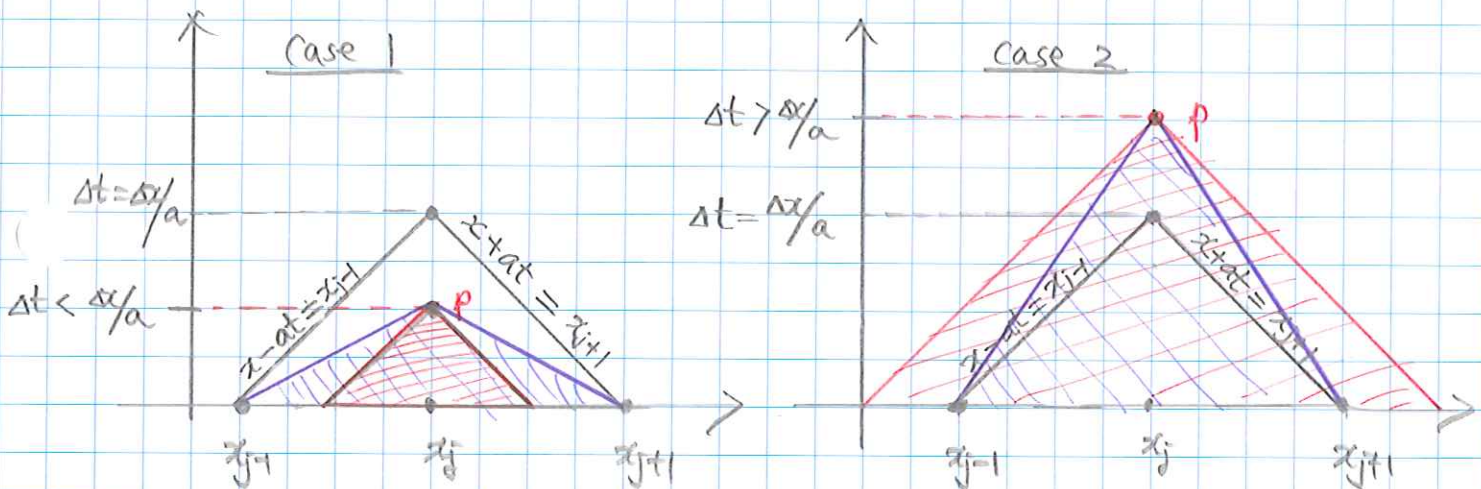
Ex. Let's consider $u_t + au_x = 0$, $a > 0$, solved by the upwind method:

$$U_j^{n+1} = U_j^n - \frac{a\Delta t}{\Delta x} [U_j^n - U_{j-1}^n].$$

Consider two cases:

Case 1: $\frac{a\Delta t}{\Delta x} < 1$, or $a\Delta t < \Delta x$

Case 2: $\frac{a\Delta t}{\Delta x} > 1$, or $a\Delta t > \Delta x$



→ The red triangles are the analytical domain of dependence of the pt. P.

→ The blue triangles are the numerical domain of dependence of the pt. P, using the upwind scheme.

→ Case 1:

Case 2:

$$D_{\text{num}} \supset D_{\text{exact}}$$

$$D_{\text{num}} \subset D_{\text{exact}}$$

∴ CFL condition ✓

∴ CFL condition ✗

Conclusion:

We want (i) $\frac{a \Delta t}{\Delta x} \leq 1$ for stability, &

(ii) $\frac{a \Delta t}{\Delta x} \approx 1$ for accuracy.

Point We often use $Ca = \begin{cases} \frac{a \Delta t}{\Delta x} & \text{for advection } (u_t + a u_x = 0) \\ & (a > 0) \\ \frac{k \Delta t}{\Delta x^2} & \text{for diffusion } (u_t = k u_{xx}) \\ & (k > 0) \end{cases}$

$$\Rightarrow \boxed{0 < Ca \leq 1}$$

Point $\Delta t = \begin{cases} \frac{Ca \Delta x}{|a|} & \text{for advection } (a > 0 \text{ or } a < 0) \\ & \textcircled{3} \\ \frac{Ca \Delta x^2}{k} & \text{for diffusion } (k > 0) \\ & \textcircled{4} \end{cases}$

$$\Rightarrow \begin{cases} \Delta t_{adv} \equiv \frac{Ca \Delta x}{|a|} \\ \Delta t_{diff} \equiv \frac{Ca \Delta x^2}{k} \end{cases}$$

\Rightarrow If we solve an advection-diffusion eqn:

$$u_t + a u_x = k u_{xx}, \text{ then}$$

$\Delta t = \min \{ \Delta t_{adv}, \Delta t_{diff} \}$ for explicit schemes.

Rank In general,

$$\Delta t_{adv} \gg \Delta t_{diff}, \text{ since } \begin{cases} \Delta t_{adv} \sim \mathcal{O}(\Delta x) \\ \Delta t_{diff} \sim \mathcal{O}(\Delta x^2), \end{cases}$$

In this case, we want

[explicit scheme for advection, Δ
[implicit scheme for diffusion,

in order to overcome the small Δt_{diff}
that makes computation very expensive.