

# Finite Difference methods for linear scalar advection.

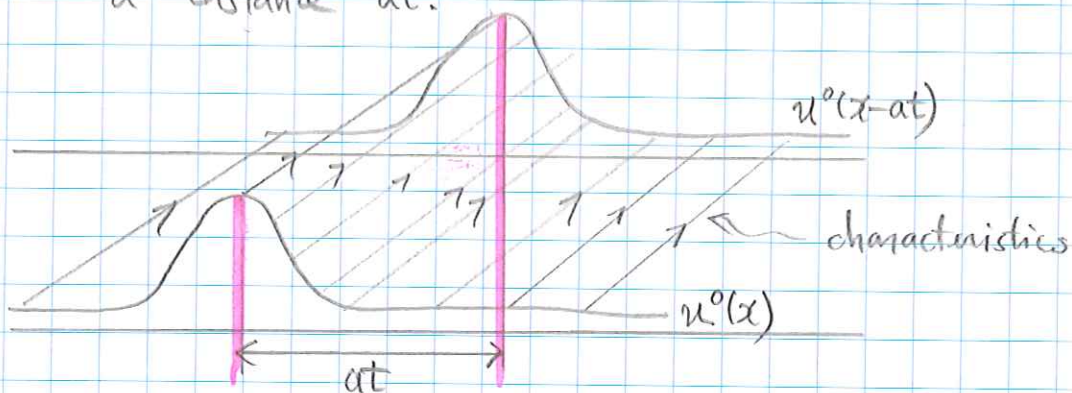
→ We are interested in numerically solving a linear scalar advection given by, on  $[x_a, x_b]$ ,

$$\boxed{u_t + a u_x = 0}, \quad a > 0 \text{ or } a < 0,$$

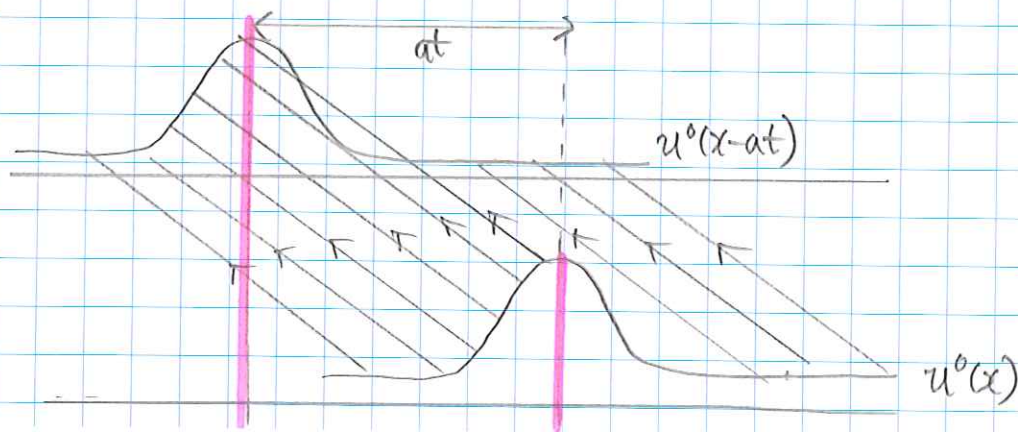
$$\begin{cases} \text{IC: } u(x, 0) = u^0(x) \\ \text{BC: } \begin{cases} u(x_a, t) = g_a(t), \\ u(x_b, t) = g_b(t) \end{cases} \end{cases}$$

→ Since the exact solution is  $u(x, t) = u^0(x - at)$ ,

(i) if  $a > 0$ ;  $u^0(x)$  is transported to the right by a distance "at".



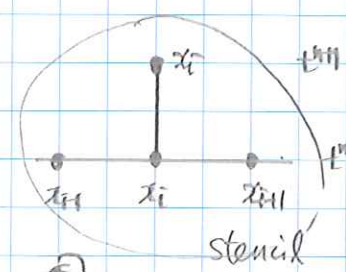
(ii) if  $a < 0$ ;  $u^0(x)$  is transported to the left by a distance "at".



Ex. If we use FTCS for Eqn. (1):

$$\frac{U_i^{n+1} - U_i^n}{\Delta t} = -\frac{a}{2\Delta x} [U_{i+1}^n - U_{i-1}^n], \quad \text{or}$$

$$\boxed{U_i^{n+1} = U_i^n - \frac{a\Delta t}{2\Delta x} [U_{i+1}^n - U_{i-1}^n]} \quad \dots (5)$$



→ The method is  $\mathcal{O}(\Delta t + \Delta x^2)$  (HW), similar to the FTCS for the heat eqn.

→ Contrary to the case with the heat eqn, FTCS for the advection eqn is NOT stable. which will be examined soon later. → two-level method

Ex. A minor modification of (5) is available to produce a bit more improved method, called "Lax-Friedrichs" (LF) method,

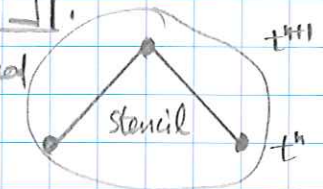
→ The modification is to replace, in (5),

$$U_i^n \quad \text{with} \quad \frac{U_{i+1}^n + U_{i-1}^n}{2}, \quad \text{giving}$$

$$\boxed{U_i^{n+1} = \frac{1}{2} [U_{i+1}^n + U_{i-1}^n] - \frac{a\Delta t}{2\Delta x} [U_{i+1}^n - U_{i-1}^n]}.$$

→ two-level method

→ LF is  $\mathcal{O}(\Delta t + \Delta x^2)$ . (HW)



→ Although improved, LF is NOT practically useful. The major importance of LF is to provide a guidance on some stability issues which will be examined later as well.

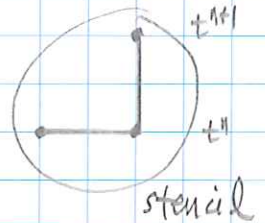
→ LF is stable & convergent if  $|\frac{a\Delta t}{\Delta x}| \leq 1$ .  
(shown later).



Ex. FTBS (Forward in time, Backward in space)  $\sim O(\Delta t + \Delta x)$

$$\frac{U_i^{n+1} - U_i^n}{\Delta t} + a \left[ \frac{U_i^n - U_{i-1}^n}{\Delta x} \right] = 0, \text{ or}$$

$$U_i^{n+1} = U_i^n - \frac{a\Delta t}{\Delta x} [U_i^n - U_{i-1}^n] = 0$$

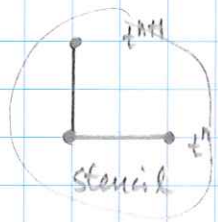


→ two-level method

Ex. FTFS (Forward in time, Forward in space)  $\sim O(\Delta t + \Delta x)$

$$\frac{U_i^{n+1} - U_i^n}{\Delta t} + a \left[ \frac{U_{i+1}^n - U_i^n}{\Delta x} \right] = 0, \text{ or}$$

$$U_i^{n+1} = U_i^n - \frac{a\Delta t}{\Delta x} [U_{i+1}^n - U_i^n] = 0$$

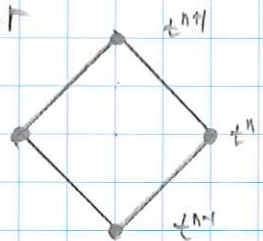


→ two-level method

Ex. Leapfrog  $\sim O(\Delta t^2 + \Delta x^2)$

$$\frac{U_i^{n+1} - U_i^{n-1}}{2\Delta t} + a \left[ \frac{U_{i+1}^n - U_{i-1}^n}{2\Delta x} \right] = 0, \text{ or}$$

$$U_i^{n+1} = U_i^{n-1} - \frac{a\Delta t}{\Delta x} [U_{i+1}^n - U_{i-1}^n] = 0$$



→ three-level method

Ex. Lax-Wendroff  $\sim O(\Delta t^2 + \Delta x^2)$

→ Consider a given the semi-discrete form with CS:

$$\frac{du(x_i, t)}{dt} = -a \left[ \frac{u(x_{i+1}, t) - u(x_{i-1}, t)}{2\Delta x} \right]$$

which is exact in time, &  $O(\Delta x^2)$  in space,





→ To simplify further discussion, let's suppose we're given a periodic BC:

$$\Rightarrow \begin{cases} U_0(t) = U_{N-1}(t) \\ U_{NH}(t) = U_1(t) \end{cases}$$

→ Then we can include  $g(t)$  into  $\hat{A}$ , giving a simpler form

$$\boxed{\frac{d\underline{U}}{dt} = \hat{A} \underline{U}(t)} \quad \text{--- (7), where}$$

$$\hat{A} = -\frac{a}{2\Delta x} \begin{pmatrix} 0 & 1 & & & \\ -1 & 0 & 1 & & \\ & -1 & 0 & 1 & \\ & & & & \\ & & & & -1 & 0 & 1 \\ & & & & & -1 & 0 \end{pmatrix}$$

(1) BC

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→ From (7),  $\underline{U}'(t) = \hat{A} \underline{U}(t) \Rightarrow \underline{U}''(t) = \hat{A} \underline{U}'(t) = \hat{A}^2 \underline{U}(t)$ .

→ Using this in the Taylor series expansion (5) up to  $O(\Delta t^2)$ , we get

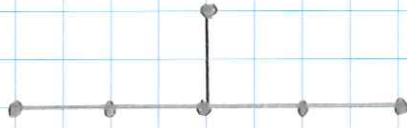
$$\boxed{\underline{U}(t^{n+1}) = \underline{U}(t^n) + \Delta t \hat{A} \underline{U}(t^n) + \frac{\Delta t^2}{2} \hat{A}^2 \underline{U}(t^n)} \quad \text{--- (8)}$$

→ (8) is the 2<sup>nd</sup> order method in both time & space, (Note the 2<sup>nd</sup> order in space is due to the 2<sup>nd</sup> order spatial discretization reflected in  $\hat{A}$ ).

→ Computing  $\hat{A}^2$  and writing the method at grid pts:

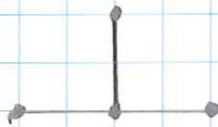
$$U_i^{n+1} = U_i^n + \Delta t \left( \frac{a}{-2\Delta x} \right) [U_{i+1}^n - U_{i-1}^n] + \frac{\Delta t^2}{2} \left( \frac{a}{-2\Delta x} \right)^2 [U_{i-2}^n - 2U_i^n + U_{i+2}^n]$$

→ Note that this method needs a large 5-pt stencil:



due to the last term which approximates  $\frac{a^2}{2} \Delta t^2 U_{xx}$  using a centered differencing with  $2\Delta x$ .

→ Instead, we can approximate it on a more compact 3-pt stencil:



obtaining the Lax-Wendroff method (LW)

$$U_i^{n+1} = U_i^n - \frac{a\Delta t}{2\Delta x} [U_{i+1}^n - U_{i-1}^n] + \frac{\Delta t^2}{2} \left( \frac{a}{-\Delta x} \right)^2 [U_{i-1}^n - 2U_i^n + U_{i+1}^n]$$

Prmk, One can derive the LW method in a simpler way:

Given  $u_t + au_x = 0$ ,

→ Apply Taylor series expansion directly on the PDE:

$$u(x, t + \Delta t) = u(x, t) + \Delta t u_t(x, t) + \frac{\Delta t^2}{2} u_{tt}(x, t) + O(\Delta t^3) \quad \text{--- (9)}$$

→ Use  $\begin{cases} u_t = -au_x, & \& \end{cases}$

$$u_{tt} = (-au_x)_t = (-a)(u_t)_x = (-a)(-au_x)_x = a^2 u_{xx}$$

→ (9) becomes

$$u(x, t + \Delta t) = u(x, t) - a\Delta t u_x(x, t) + \frac{\Delta t^2}{2} a^2 u_{xx}(x, t) + O(\Delta t^3) \quad \text{--- (10)}$$



→ LW is then easily achieved by using standard centered differencing approximations to  $u_x$  &  $u_{xx}$ :

$$\begin{cases} u_x \cong \frac{1}{\Delta x} [u(x+\Delta x, t) - u(x-\Delta x, t)], \\ u_{xx} \cong \frac{1}{\Delta x^2} [u(x+\Delta x, t) - 2u(x, t) + u(x-\Delta x, t)] \end{cases}$$

and dropping terms with  $\mathcal{O}(\Delta t^3)$  and higher.



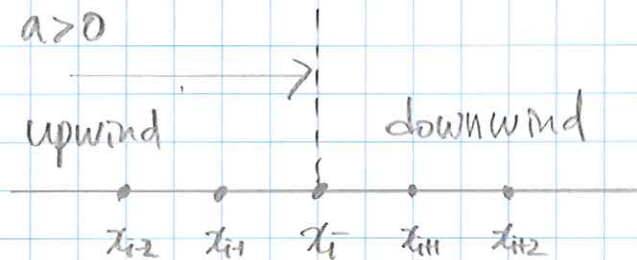
## Ex. Upwind method $\sim O(\Delta t + \Delta x)$

→ First we define the upwind direction.

Def. The upwind direction is the direction facing the wind. The opposite direction is called the downwind direction.

→ We can therefore have different discrete formulations depending on the wind direction.

→ Case 1:  $a > 0$



⇒ At  $x = x_i$ , the wind has already swept the upwind region  $\{x \mid x \leq x_i\}$ .

⇒ The first-order upwind method is to use the data in the upwind region only for the  $a \Delta x$  approximation.

$$\Rightarrow \frac{U_i^{n+1} - U_i^n}{\Delta t} + a \frac{U_i^n - U_{i-1}^n}{\Delta x} = 0, \text{ or}$$

upwind: 
$$U_i^{n+1} = U_i^n - \frac{a \Delta t}{\Delta x} [U_i^n - U_{i-1}^n] = 0 \quad \dots \quad (11)$$

⇒ On the other hand, the downwind method becomes

$$U_i^{n+1} - U_i^n + a \frac{U_{i+1}^n - U_i^n}{\Delta x} = 0, \text{ or}$$

downwind: 
$$U_i^{n+1} = U_i^n - \frac{a \Delta t}{\Delta x} [U_{i+1}^n - U_i^n] = 0 \quad \dots \quad (12)$$



→ Case 2:  $a < 0$  : the rolls are changed and hence

upwind: (12)

downwind: (11)

→ The crucial importance on the upwind method is that the method provides the proper stability depending on the direction of advection.

→ This stability gain in the upwind method is much more important than anything, although it seems, at first glance, to be less accurate than the standard centered spatial differencing.

→ We will see, using von Neumann stability analysis, that

upwind → stable  
downwind → unstable  
centered in space → unstable

→ In summary, for the advection eqn, a difference method that includes information in the downwind region, the method becomes unstable.