

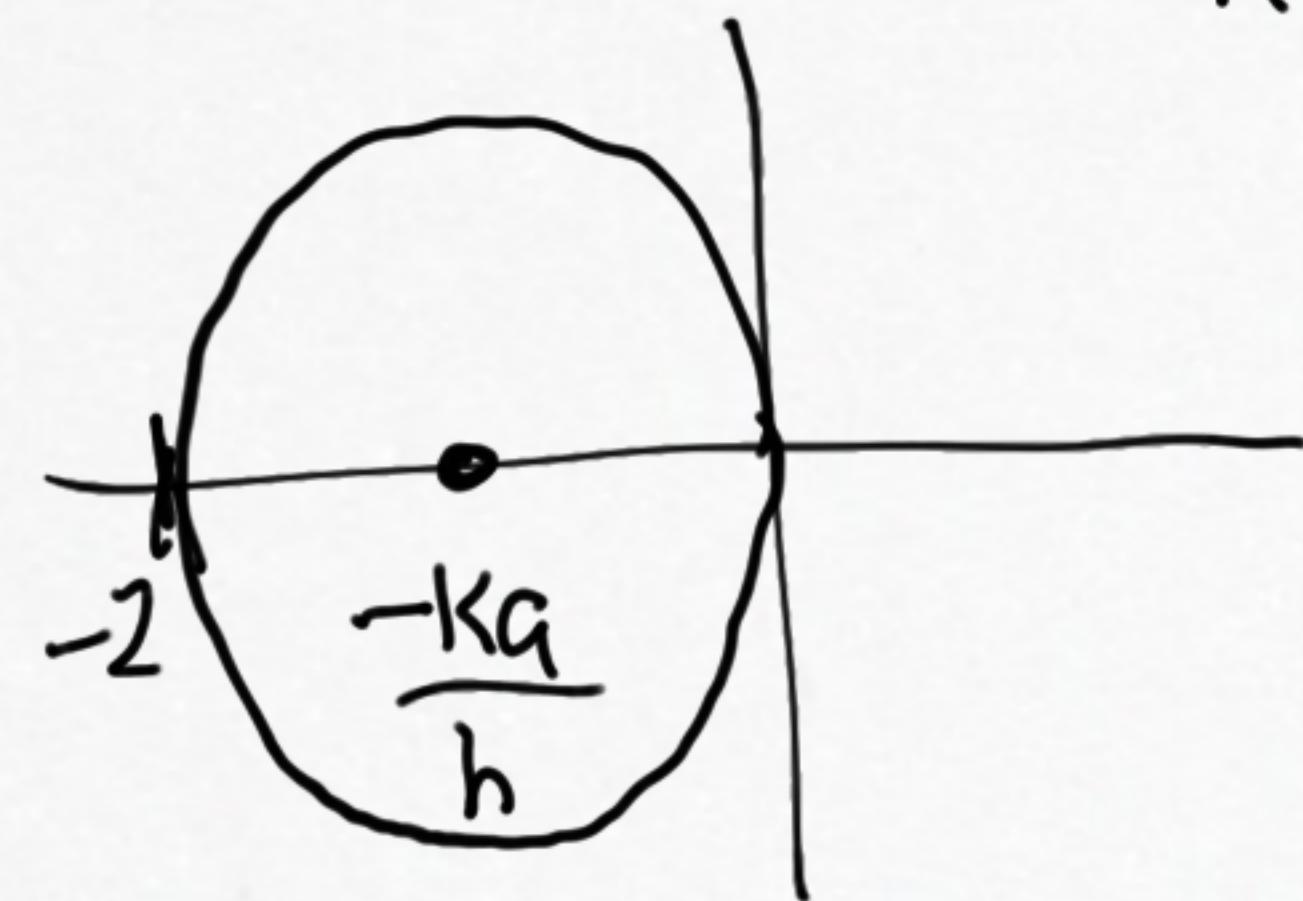
The upwind method

$$U_t + aU_x = 0 \quad a > 0$$

$$U_j^{n+1} = U_j^n - \frac{Ka}{h} (U_j^n - U_{j-1}^n)$$

$$\text{Semi-discrete: } U'(t) = -\frac{a}{h} \begin{bmatrix} 1 & & & \\ -1 & 2 & & \\ & -1 & 2 & \\ & & \ddots & \end{bmatrix} U(t)$$

$$\text{Eigenvalues of } A: -\frac{a}{h}$$



$$-\frac{Ka}{h} \geq -2 \Leftrightarrow \boxed{\frac{Ka}{h} \leq 2} \quad K \leq \frac{2h}{a}$$

Von Neumann analysis (Periodic BCs) $U_j^n \rightarrow g^n e^{ikj\theta}$

$$U_j^{n+1} = U_j^n - \frac{Ka}{h} (U_j^n - U_{j-1}^n)$$

$$g^{n+1} e^{ijk\theta} = g^n e^{ijk\theta} - \frac{Ka}{h} (g^n e^{ijk\theta} - g^{n-1} e^{i(j-1)k\theta}) \quad \text{Divide by } g^n e^{ijk\theta}$$

$$g = 1 - \frac{Ka}{h} (1 - e^{-ik\theta}) = 1 - \frac{Ka}{h} (1 - \cos(k\theta) + i\sin(k\theta)) \quad \frac{Ka}{h} = \gamma$$

$$g = 1 - \gamma + \gamma \cos(k\theta) - i\gamma \sin(k\theta)$$

$$\gamma^2$$

$$|g|^2 = 1 - 2\gamma + \gamma^2 + 2\gamma \cos(k\theta) - 2\gamma^2 \cos^2(k\theta) + \gamma^2 \cos^2(k\theta) + \gamma^2 \sin^2(k\theta)$$

$$= 1 - 2\gamma + 2\gamma^2 + 2\gamma \cos(k\theta) - 2\gamma^2 \cos(k\theta)$$

$$= 1 - 2\gamma(1 - \gamma) (1 - \cos(k\theta))$$

$$|g|^2 = (1-\lambda)(1-\lambda^*) (1-\cos(\lambda)) \quad \left(\text{If } 0 \leq \lambda \leq 1 \text{ then } |g|^2 \leq 1 \right)$$

Observe: $0 \leq \lambda(1-\lambda) \leq \frac{1}{4}$ $0 \leq 1-\cos(\lambda) \leq 2$

$$0 \leq \lambda(1-\lambda) \leq \frac{1}{4}$$

So $|g|^2 \leq 1$.

We need $0 \leq \frac{\lambda a}{h} \leq 1 \rightarrow \text{unstable if } a < 0$.

Let $A \in \mathbb{C}^{m \times m}$ be diagonalizable: $A = R \Lambda R^{-1}$
 $AR = R \Lambda$

$$A^n = R \Lambda R^{-1} R \Lambda R^{-1} \cdots R \Lambda R^{-1}$$

$$A^n = R \Lambda^n R^{-1}$$

$$\|A^n\|_2 \leq \|R\|_2 \| \Lambda^n \|_2 \|R^{-1}\|_2 = \|\Lambda\|_2^n K(R) = \rho(A)^n K(R)$$

↓
spectral radius ↓
condition #

If A is normal, then R is unitary $\rightarrow K(R) = 1 \rightarrow \|A^n\|_2 = \rho(A)^n$

Dfn.: Matrix $A \in \mathbb{C}^{m \times m}$ is normal if $A^*A = AA^*$.

Symm., skew-symm. and circulant matrices are all normal.

For a normal matrix with $\rho(A) < 1$: $\|A^n\| \leq \rho(A)^n < 1$

$$\lim_{n \rightarrow \infty} \|A^n\| = 0$$

Non-normal case: we can have $\|A^n\| > 1$ even though $\rho(A) < 1$.

but still $\lim_{n \rightarrow \infty} \|A^n\| = 0$



Moral of the Story: For non-normal matrices,
eigenvalues don't fully characterize the behavior.

Example: $A_1 = \begin{bmatrix} 0.8 & 0 \\ 0 & 0.9 \end{bmatrix}$ $\lambda = 0.8, 0.9$ $\|A^n\| \leq \|A\|^n$

$A_2 = \begin{bmatrix} 0.8 & 100 \\ 0 & 0.9 \end{bmatrix}$ $\lambda = 0.8, 0.9$ $\|A^n\| > \|A\|^n$ for $n=2, 3, \dots$

$R = \begin{bmatrix} 1 & 1 \\ 0 & 10^{-3} \end{bmatrix}$ $\kappa(R) = 2000$

Modified Equation Analysis

Goal: Find a PDE that our numerical discretization satisfies exactly.

$$V(x_j, t_n) = U_j^n$$

$$\text{Upwind: } \frac{U_j^{n+1} - U_j^n}{k} = -a \frac{U_j^n - U_{j-1}^n}{h}$$

Substitute V into the discretization:

$$\frac{V(x_j, t_{n+1}) - V(x_j, t_n)}{k} = -a \frac{V(x_j, t_n) - V(x_{j-1}, t_n)}{h}$$

$$V(x_j, t_{n+1}) = V(x_j, t_n) + k V_t(x_j, t_n) + \frac{k^2}{2} V_{tt}(x_j, t_n) + \mathcal{O}(k^3)$$

$$V(x_{j-1}, t_n) = V(x_j, t_n) - h V_x(x_j, t_n) + \frac{h^2}{2} V_{xx}(x_j, t_n) + \mathcal{O}(h^3)$$

$$V_t + \frac{k}{2} V_{tt} + \mathcal{O}(k^2) = -a V_x + a \frac{h}{2} V_{xx} + \mathcal{O}(h^2)$$

$$V_t + a V_x = \frac{1}{2} (a h V_{xx} - k V_{tt}) + \mathcal{O}(h^2, k^2)$$

$$V_t \approx -a V_x \Rightarrow V_{tt} \approx -a V_{xx} \quad \left. \begin{array}{l} V_{tt} = +a^2 V_{xx} + \mathcal{O}(h, k) \\ V_{tx} \approx -a V_{xx} \end{array} \right\}$$

$$V_t + a V_x = \frac{1}{2} (a h V_{xx} - a^2 k V_{xx}) + \mathcal{O}(h^2, k^2, kh)$$

$$V_t + a V_x = \frac{a}{2} V_{xx} / (h - ka) \Rightarrow V_t + a V_x = \frac{1}{2} ah \left(1 - \frac{ka}{h} \right) V_{xx} + \dots$$

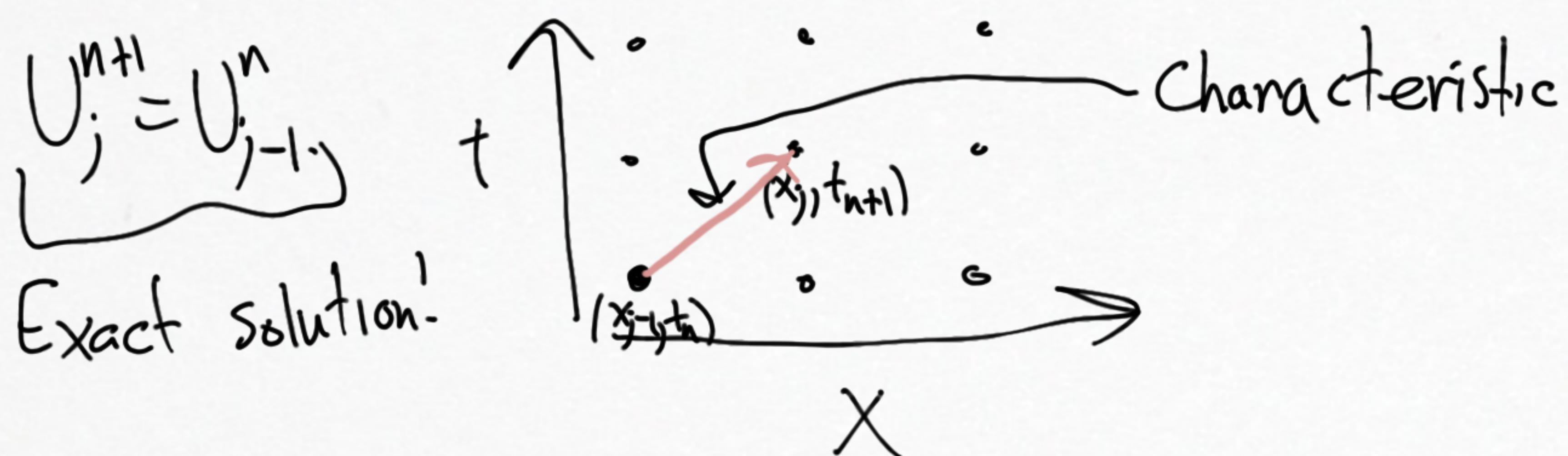
$$V_t + \alpha V_x = \underbrace{\frac{ah}{2}(1-\nu)V_{xx}}_{\text{diffusion term}} + O(k^2, h^2, kh)$$

V satisfies an advection-diffusion equation

The diffusion term is proportional to h

If $\nu=1$ (if $ka=h$) \rightarrow no diffusion.

$$U_j^{n+1} = U_j^n - \frac{ka}{h}(U_j^n - U_{j-1}^n) \quad \text{Let } \frac{ka}{h} = 1.$$

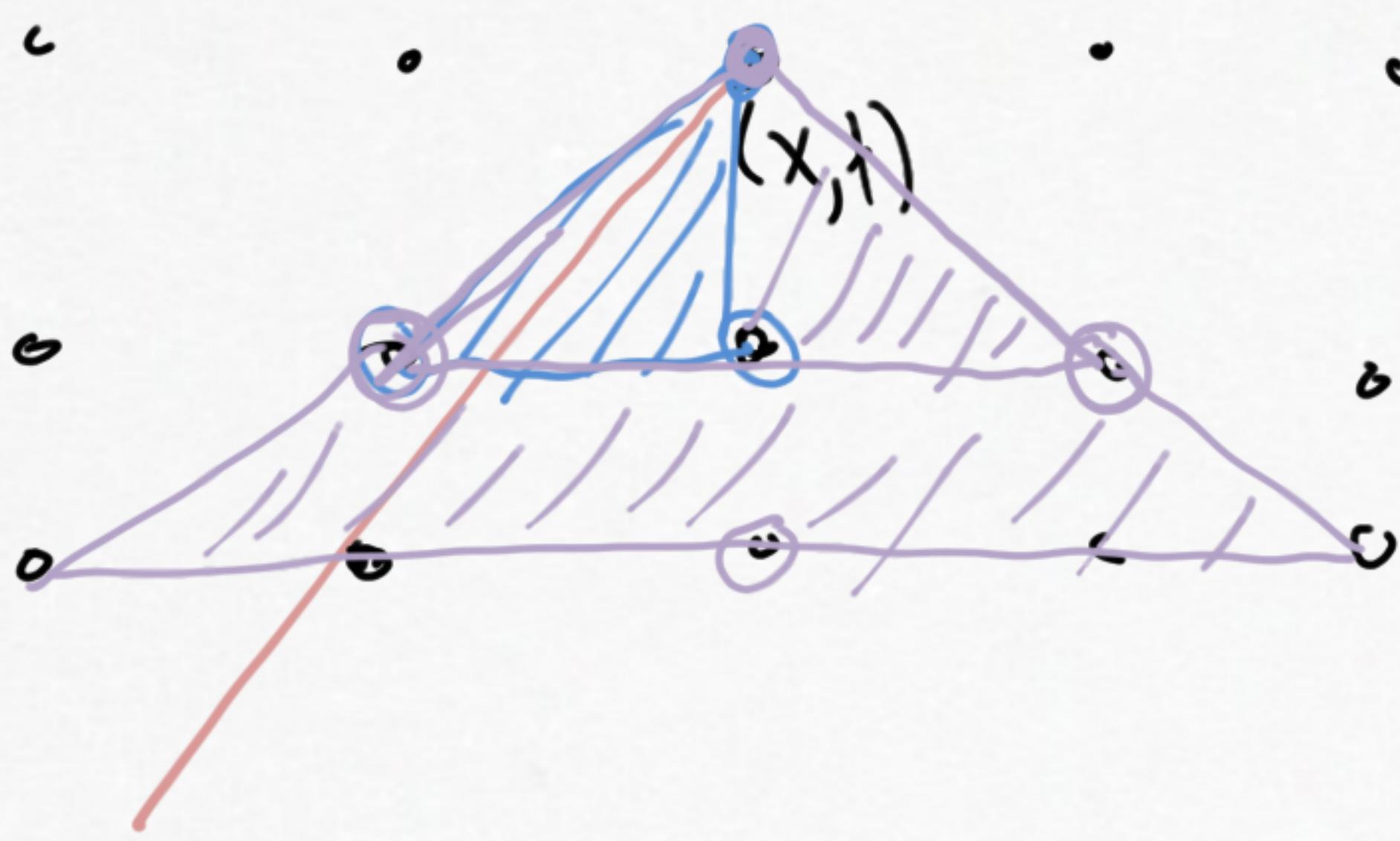


The CFL condition

The domain of dependence of $u(x, t)$ is the set of points at earlier times on which $u(x, t)$ depends.

The numerical domain of dependence of U_j^n is the set of points on the computational grid that affect U_j^n .

For any convergent method, the numerical d. of. d. must contain the (PDE d. of. d.) in the limit $h, k \rightarrow 0$.



Upwind
Leapfrog

$$-1 \leq \frac{K_a}{h} \leq 1$$

$$0 \leq \frac{K_a}{h} \leq 1$$

(upwind)

For hyp. PDEs: finite speed of propagation

For parabolic PDEs: Infinite speed of propagation.

Text sections:

10.4-107, 10.9, D.4