

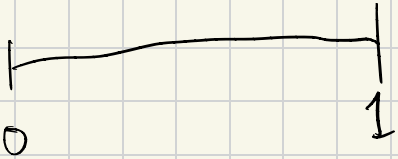
含参数的偏微分方程

$u(t, x)$ 偏微分方程的解

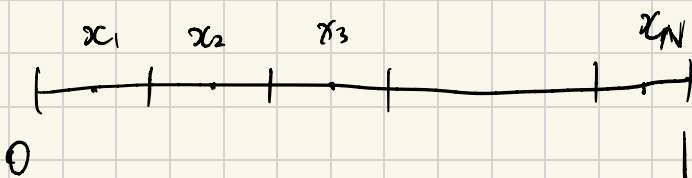
粗体 $u(t) = \begin{bmatrix} u(t, x_1) \\ u(t, x_2) \\ \vdots \\ u(t, x_N) \end{bmatrix} \in \mathbb{R}^N$ 是空间离散后的向量

① 对流扩散方程

$$\partial_t u + \partial_x (au) = \partial_x (v \partial_x u)$$


$$\partial_t \int_0^1 u \, dx + \underbrace{\int_0^1 \partial_x (au) \, dx}_0 = \underbrace{\int_0^1 \partial_x v \partial_x u \, dx}_0$$

周期性



$$u(t) = \begin{bmatrix} u_1(t) \\ u_2(t) \\ \vdots \\ u_N(t) \end{bmatrix} \quad u_i(t) = \frac{1}{\Delta x_i} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} u(t, x) dx$$

$$\approx u(t, x_i) + O(\Delta x_i^2)$$

守恒律:

$$\int_0^1 u dx = \sum \Delta x_i u_i(t)$$

$u_i(t)$ 如何随时间演化

$$\begin{aligned} \frac{d}{dt} u_i(t) &= \frac{1}{\Delta x_i} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \frac{\partial u(t, x)}{\partial t} dx \\ &= \frac{1}{\Delta x_i} \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \partial_x (v \partial_x u) - \partial_x (au) dx \\ &= \frac{[v \partial_x u - (au)](x_{i+\frac{1}{2}}) - [v \partial_x u - (au)](x_{i-\frac{1}{2}})}{\Delta x_i} \\ &:= \frac{f_{i+\frac{1}{2}} - f_{i-\frac{1}{2}}}{\Delta x_i} \end{aligned}$$

$$\Delta x_i \frac{d u_i(t)}{dt} = \sum f_{i+\frac{1}{2}} - f_{i-\frac{1}{2}} = 0$$

$$\dot{u}_i = \left[\frac{u_i + u_{i+1}}{2} - \frac{u_i + u_{i-1}}{2} \right] / \Delta x$$

$$\dot{u}_i = \frac{u_{i+1} - u_{i-1}}{2 \Delta x}$$

$$u_i^{n+1} = u_i^n + (u_{i+1}^n - u_{i-1}^n) \frac{\Delta t}{2 \Delta x}$$

$$u_i^n \sim \varepsilon_i^n = e^{kx_i I} \quad (\text{误差})$$

$$\varepsilon_i^{n+1} = \varepsilon_i^n + (\varepsilon_{i+1}^n - \varepsilon_{i-1}^n) \lambda$$

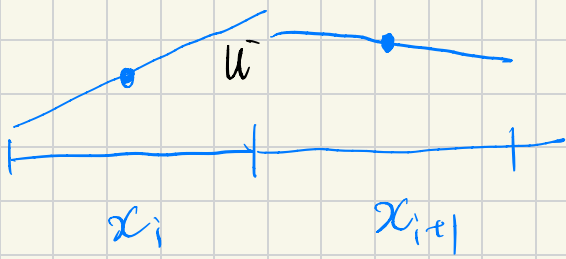
$$= \varepsilon_i^n \left(1 + (e^{k\Delta x I} - e^{-k\Delta x I}) \lambda \right)$$

$$= \varepsilon_i^n \left(1 + 2\lambda \sin(\Delta x k) i \right)$$

$$|\varepsilon_i^{n+1}|^2 \geq |\varepsilon_i^n|^2 \left(1 + [2\lambda \sin(\Delta x k)]^2 \right)$$

限制器 (重构)

$$u_{i-1}(t) \quad u_i(t) \quad u_{i+1}(t) \quad \rightarrow \quad u(t, x)$$



$$u^- = u_i + \frac{\Delta x_i}{2} S_i \quad S_i = \min \{ \partial u \}$$

$$u^+ = u_{i+1} - \frac{\Delta x_{i+1}}{2} S_{i+1} \quad S_{i+1} = \min \{ \partial u \}$$

时间离散:

$$\frac{du}{dt} = f(u, t, x)$$

$$u(t + \Delta t) = u(t) + \int_t^{t + \Delta t} f(u, t, x) dt$$

向前欧拉方法

截断误差 $O(\Delta t^2)N$

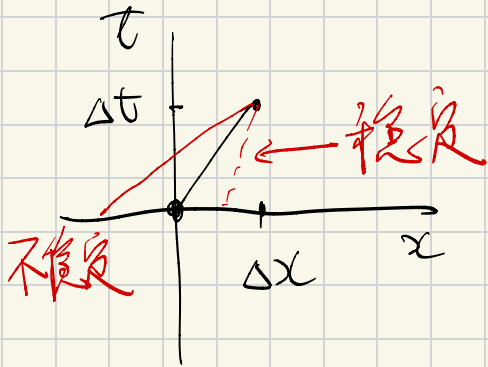
$$u(t + \Delta t) = u(t) + \Delta t f(u(t), t, x)$$

Runge Kutta 方法

截断误差 $N \sim \frac{1}{\Delta t}$
 $O(\Delta t^3)N$

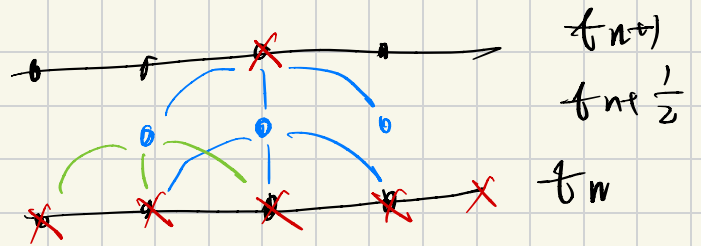
时间离散 CFL 条件:

信息传递

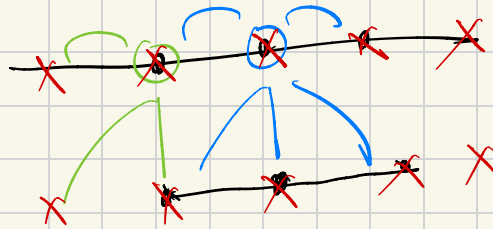


$$a \cdot \Delta t \leq \Delta x$$

2nd Runge-kutta 格式



隐式格式



扩散项

$$\ddot{u}_i = \nu \frac{u_{i+1} - 2u_i + u_{i-1}}{\Delta x^2}$$

$$u_i^{n+1} = u_i^n + \nu \frac{\Delta t}{\Delta x^2} (u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

$$\varepsilon_i^n \sim e^{kx_i \Gamma}$$

$$\begin{aligned} \varepsilon_i^{n+1} &= \varepsilon_i^n \left(1 + \nu \frac{\Delta t}{\Delta x^2} (e^{k\Delta x \Gamma} - 2 + e^{-k\Delta x \Gamma}) \right) \\ &= \varepsilon_i^n \left((1 - 2\lambda) + 2\lambda \cos(k\Delta x) \right) \end{aligned}$$

$$\lambda \leq 2 \quad \left(\lambda \leq \frac{1}{2} \right)$$

Darcy 方程 (有限元方法)

二阶可微 \Rightarrow 一阶可微 (数学性质)

$$\int \partial_x(a \partial_x u) + \partial_y(a \partial_y u) \phi \, d\Omega$$

$$= \int \partial_x(a \partial_x u \phi) + \partial_y(a \partial_y u \phi) - a \partial_x u \partial_x \phi - a \partial_y u \partial_y \phi$$

$$= - \int a \partial_x u \partial_x \phi + a \partial_y u \partial_y \phi \, d\Omega \quad (\phi = 0 \text{ on } \partial\Omega)$$

$$= \int f \phi \, d\Omega$$

$$u = \sum_{i=1}^N u_i \phi_i(x, y)$$

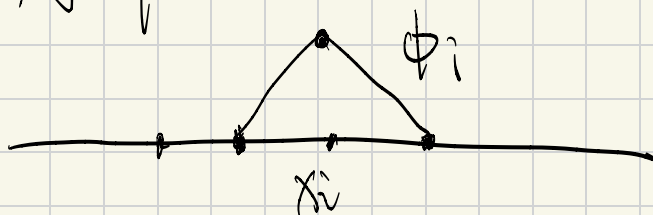
N 个未知量

N 个方程

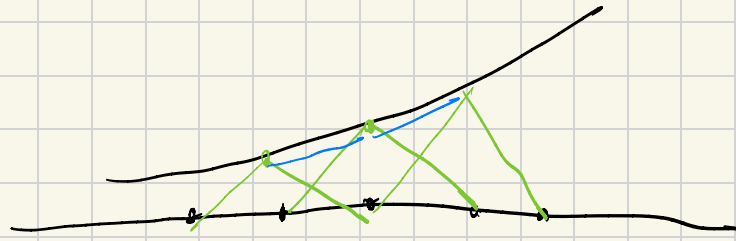
初期 $\Rightarrow \phi_i = \sin(ax + by) \dots$

ϕ_i 较为局部的函数

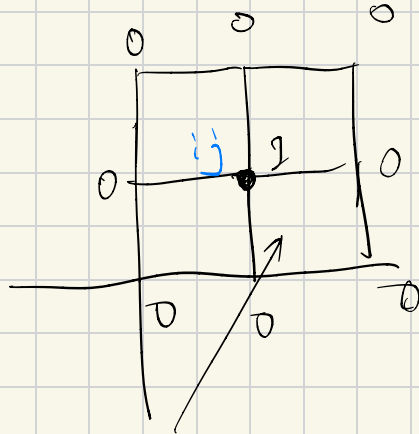
一维



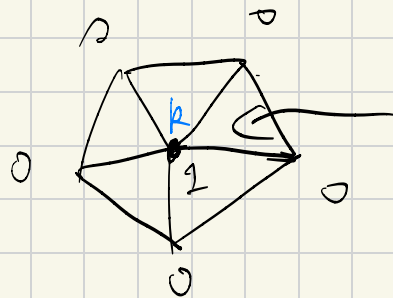
分片线性



二维



$$\phi|_e = axy + bx + cy + d$$



$$\phi|_e = ax + by + c$$