

Chapter 5

偏微分方程数值解

偏微分方程(PDE)分类 Let $u := u(x, t)$ defined on a domain Ω , where u satisfy

椭圆型

$$a(x)u_{xx} + b(x)u_x + c(x)u = f(x)$$

抛物型

$$u_t = (b(x, t)u_x)_x + c(x, t)u + d(x, t)$$

双曲型

$$u_{tt} - \nu^2 u_{xx} = f(x)$$

with certain 初值 and/or 边值 conditions.

5.1 椭圆边值问题

5.1.1 五点差分格式

记 $u_{i,j}$ 表示网格点 (i, j) 上待求函数的近似值, 那么

$$\begin{aligned}\frac{\partial^2 u}{\partial x^2} &= \frac{-u_{i-1,j} + 2u_{i,j} - u_{i+1,j}}{h_x^2} + O(h_x^2) \\ \frac{\partial^2 u}{\partial y^2} &= \frac{-u_{i,j-1} + 2u_{i,j} - u_{i,j+1}}{h_y^2} + O(h_y^2).\end{aligned}$$

因此在二维情形 (设 $h_x = h_y = h$, 舍去高阶无穷小项后) 可得

$$-\Delta u \approx \frac{1}{h^2} \begin{pmatrix} & -u_{i,j+1} & \\ -u_{i-1,j} & 4u_{i,j} & -u_{i+1,j} \\ & -u_{i,j-1} & \end{pmatrix} \quad (5.1)$$

稀疏Laplace算子-Matlab表示: 上述稀疏Laplace算子在Matlab中的实现通常有两中不同的做法

- kron — 基于内部未知点 $((n-1) \times (n-1))$ 个;
- spLaplacian5.m — 基于全部点 $((n+1) \times (n+1))$ 个。

推荐采用后一种方案, 代价是将边界条件也作为方程, 优点是算子离散和边界条件的处理相对独立, 因此数学上更直观。

例5.1.1. 考虑如下Poisson方程的Dirichlet边值问题

$$-\Delta u = f, \quad (x, y) \in [0, 1]^2. \quad (5.2)$$

取解析函数

$$u(x, y) = \sin(\pi x) \sin(\pi y), \quad (5.3)$$

作为测试解, 相应地取 $f = 2\pi^2 u(x, y)$ 。

实现细节参考testPoisson.m以及相关.m文件。

5.1.2 非线性问题

考虑如下形式的“半”线性问题

$$-\Delta u + f(\mathbf{x}, u) = g, \quad \mathbf{x} \in \Omega \subset \mathbf{R}^2 \quad (5.4)$$

仍取前例中的解析函数作为测试解, 此外令

$$f(\mathbf{x}, u) = u^3. \quad (5.5)$$

右端项 g 根据 u 和 f 的表达式可计算得到, 参考fung.m

离散格式

可在所有内点 $i, j = 1, \dots, n-1$ 上建立方程

$$\frac{1}{h^2} \begin{pmatrix} -u_{i,j+1} & 4u_{i,j} & -u_{i+1,j} \\ -u_{i-1,j} & & -u_{i,j-1} \end{pmatrix} + f(x_i, y_j, u_{i,j}) = g_{i,j}.$$

如果 f 中不包含一阶微分算子(反应-扩散问题), 那么 f (线性化后)只对 $u_{i,j}$ 有贡献, 若考虑带一阶微分算子项的 f , 称为对流-扩散问题。不妨将上述代数方程简记为

$$A\mathbf{u} + f(\mathbf{u}) = \mathbf{g} \quad (5.6)$$

Newton迭代格式

令 $F(\mathbf{u}) = A\mathbf{u} + f(\mathbf{u}) - \mathbf{g}$ 则有Newton迭代

$$\mathbf{u}^{new} = \mathbf{u}^{old} - F'(\mathbf{u}^{old})^{-1} F(\mathbf{u}^{old}) \quad (5.7)$$

其中 $F'(\mathbf{u}) = A + f'(\mathbf{u})$. Matlab实现的时候会引进一些额外的变量以及向量化, 请参考test_semi_newton.m

例5.1.2 (练习).

- 修改算例一(用不同 u), 记录 $h = 10, 20, 40, 80, 160, 320$ 时数值解的 L_2 或 L_∞ 误差, 计算收敛阶
- 修改半线性问题算例, 计算 $Newton$ 迭代法的收敛阶
- 尝试将上述方法扩展至 x -和 y -方向的区间和步长不一致的情形: $h_x \neq h_y$ 以及 $(x, y) \in [a, b] \times [c, d]$
- 尝试变更方程编号的顺序 (列优先或其他感兴趣的顺序)
- 进一步研究如何利用 $Newton$ 迭代法求逆矩阵, 以及其他可能加速求解线性方程足的方法

5.2 抛物型

一维(空间)模型 热传导过程的数学建模

$$u_t = \nu u_{xx}, \quad x \in (0, 1), \quad t > 0 \quad (5.8)$$

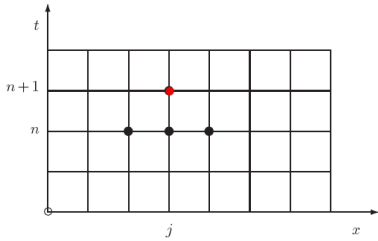
$$u(x, 0) = f(x), \quad x \in [0, 1] \quad (5.9)$$

$$u(0, t) = a(t), \quad t \geq 0 \quad (5.10)$$

$$u(1, t) = b(t), \quad t \geq 0 \quad (5.11)$$

where $f(0) = a(0)$ and $f(1) = b(1)$.

显格式(Explicit Schemes) Let h and τ be the spacing, then $x_j = jh, t_n = n\tau$. So that



$$\begin{aligned} \frac{\partial u}{\partial t}(x_j, t_n) &\approx \frac{u(x_j, t_{n+1}) - u(x_j, t_n)}{\tau} \Gamma := \frac{U_j^{n+1} - U_j^n}{\tau} \\ \frac{\partial^2 u}{\partial x^2}(x_j, t_n) &\approx \frac{u(x_{j+1}, t_n) - 2u(x_j, t_n) + u(x_{j-1}, t_n))}{(h)^2} \\ &:= \frac{U_{j+1}^n - 2U_j^n + U_{j-1}^n}{(h)^2} \end{aligned}$$

Then at (x_j, t_n) the 1D parabolic equation yields ($\mu = \nu \frac{\tau}{h^2}$)

$$U_j^{n+1} = U_j^n + \mu(U_{j+1}^n - 2U_j^n + U_{j-1}^n), \quad (5.12)$$

It is preferred as an **explicit Scheme**.

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1 Given  $\nu, f, [a, b]$  and  $N, T, \tau$ ;
2  $h = (b - a)/N$  and set  $x_j = j * h, \forall j = 0, 1, \dots, N$ ;
3  $u = \text{zeros}(N+1, T+1)$ ;
4 for  $n = 1, 2, \dots, T$  do
5    $u(0, n) = a(n\tau); u(N, n) = b(n\tau)$ ;
6   for  $j = 1, 2, \dots, N - 1$  do
7      $U_j^{n+1} = U_j^n + \mu(U_{j+1}^n - 2U_j^n + U_{j-1}^n)$ ;
8   end
9 end
```

• 算例参数: $\nu = 5, f(x, 0) = \cos \frac{\pi x}{2}, a(0, t) = 0, b(1, t) = 0$

• 计算参数: $N = 100, T = 1, \tau = 0.0015$

相容性: Does it do the right thing?

定理5.2.1 (Consistency). Let $L = \frac{\partial}{\partial t} - \nu \frac{\partial^2}{\partial x^2}, (\nu > 0)$ be the operator and $U_j^{n+1} = L_h U_j^n$ be the finite difference scheme, where L_h dependent on the time and space step τ and h . It is defined that the finite difference scheme is consistent with the original differential equation, if

$$T(x_j, t_n) = (L_h u(x_j, t_n) - u(x_j, t_{n+1})) \rightarrow 0, \quad \tau, h \rightarrow 0.$$

截断误差(Truncation Error):

$$\begin{aligned}
T(x, t) &= \frac{u(x, t + \tau) - u(x, t)}{\tau} - \nu \frac{(u(x + h, t) - 2u(x, t) + u(x - h, t)))}{h^2} \\
&= (u_t(x, t) + \frac{\tau}{2} u_{tt}(x, t) + \dots) - \nu(u_{xx} + \frac{h^2}{12} u_{xxxx} + \dots) \\
&\approx \frac{\tau}{2} u_{tt}(x, t) - \frac{\nu h^2}{12} u_{xxxx}
\end{aligned}$$

收敛性: Is $U_j^n \rightarrow u(x_j, t_n)$?

定理5.2.2 (Convergent). Using fixed initial and boundary values and $\mu = \tau/(h)^2$, and let $\tau \rightarrow 0, h \rightarrow 0$. If on any given position $(x^*, t^*) \in (0, 1) \times (0, T)$,

$$U_j^n \rightarrow u(x_j, t_n), \forall x_j \rightarrow x^*, t_n \rightarrow t^*.$$

• **Approximation Error:** $e_j = U_j^n - u(x_j, t_n)$

• Finite difference scheme - $T(x, t)$ yields

$$e_{j+1} = (1 - 2\mu)e_j^n + \mu e_{j+1}^n + \mu e_{j-1}^n - T_j^n \tau,$$

which yield $E^n \leq \frac{1}{2}\tau(M_{tt} + \frac{1}{6\mu}M_{xxxx})$ if define $E^n = \max\{|e_j|, j = 0, 1, \dots, n\}$ and M_{tt} and M_{xxxx} be the upper limit for u_{tt} and u_{xxxx} respectively.

- The previous explicit scheme convergent if $\mu := \frac{\tau}{h^2} \leq \frac{1}{2}$.

Fourier(误差)分析方法 Using **Fourier mode**

$$U_j^n = (\lambda)^n e^{ik(jh)}$$

as the solution of the finite difference scheme (5.12) it yields

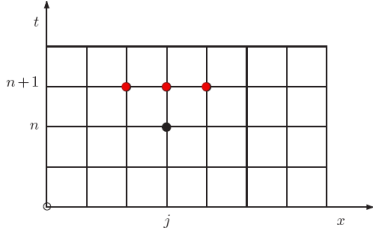
$$\begin{aligned} \lambda &:= \lambda(k) = 1 + \mu(e^{ikh} - 2 + e^{-ikh}) \\ &= 1 - 2\mu(1 - \cos(kh)) \\ &= 1 - 4\mu \sin^2 \frac{1}{2}kh \end{aligned}$$

- Since $U_j^{n+1} = \lambda U_j^n$, λ is referred as **amplification factor**
- 特殊频率 $k = m\pi$ 处, $\mu > \frac{1}{2}$ makes $\lambda > 1$, 导致发散
- **stable**: there exist a K independent of k , which makes

$$|[\lambda(k)]^n| \leq K, \quad \forall k, n\tau \leq T$$

隐格式(Implicit schemes) The stability condition $\mu = \frac{\tau}{h^2} \leq \frac{1}{2}$ is too strict, which means too small timestep $\tau \leq \frac{1}{2}h^2$ when the grid space $h \rightarrow 0$. The following scheme is another good choice

$$U_j^{n+1} = U_j^{n+1} + \mu(U_{j+1}^{n+1} - 2U_j^{n+1} + U_{j-1}^{n+1}) \quad (5.13)$$



The implicit scheme yields

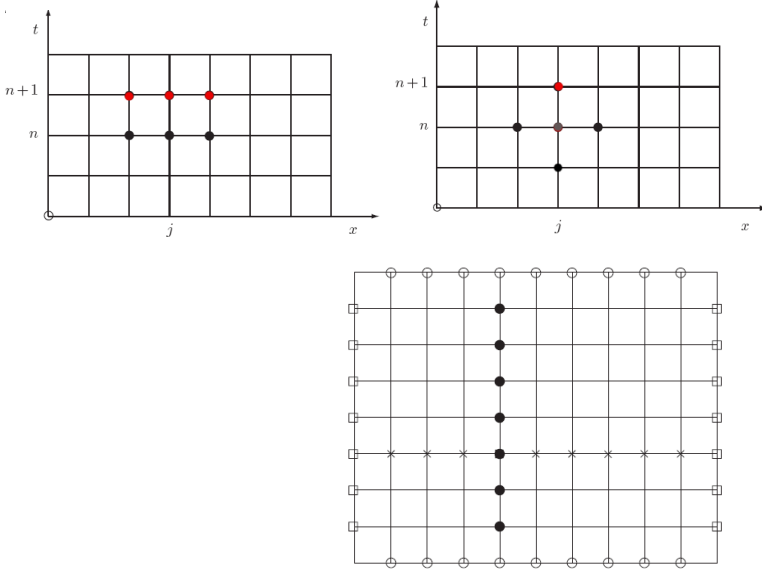
$$-\mu U_{j-1}^{n+1} + (1 + 2\mu)U_j^{n+1} - \mu U_{j+1}^{n+1} = U_j^n, \quad \forall j = 1, 2, \dots, (N-1).$$

U_0^{n+1} and U_N^{n+1} are known with the boundary condition.

- **Thomas algorithm** is most efficient for tri-diagonal system
- using Fourier mode $U_j^n = (\lambda)^n e^{ik(jh)}$ yields

$$\lambda = \frac{1}{1 + 4\mu \sin^2 \frac{1}{2}kh} < 1,$$

which says the implicit scheme is **unconditionally stable**



- However, the truncation error is same with the explicit one.

其他隐格式

- Crand-Nickson(Left, $\lambda < -1$): $\mu(1 - 2\theta) > \frac{1}{2}$
- Leap Frog(Right): $\lambda^2 + 8\lambda\mu \sin^2 \frac{1}{2}kh - 1 = 0$

更一般的边界条件

$$\frac{\partial u}{\partial x} = \alpha(t)u + g(t), \alpha(t) > 0, x = 0$$

- First order: $\frac{U_1^n - U_0^n}{h} = \alpha^n U_0^n + g^n$
- Second order: $\frac{2U_0^n - 3U_1^n + U_2^n}{h} = \alpha^n U_0^n + g^n$

非线性

$$u_t = b(u)u_{xx}, \forall x \in (0, 1)$$

The linearization is necessary at each time step

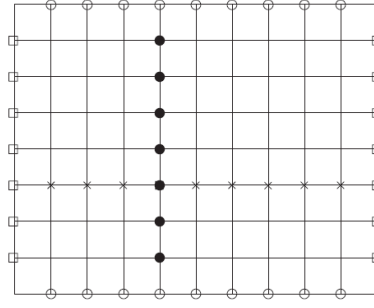
$$U_j^{n+1} = U_j^n + \mu b(U_j^n)(U_{j+1}^n - 2U_j^n + U_{j-1}^n)$$

- The error analysis at each step is similar with the linear case
- It is very hard to obtain a general global error analysis, which is dependent heaily on $b(u)$

空间变量多元情形 Let Ω be a rectangular domain $(0, X) \times (0, Y)$

Find a function $u(x, y, t)$ defined on Ω

$$\begin{aligned} u_t(x, y, t) &= b(u_{xx}(x, y, t) + u_{yy}(x, y, t)), \quad (b > 0) \\ &:= b\Delta u(x, y, t) := b\nabla^2 u(x, y, t), \end{aligned}$$



with proper Dirichlet boundary condition and initial value $u(x, y, 0)$

Explicit V.S. Implicit time step Δt , grid space Δx and Δy

$$U_{r,s}^n \approx u(x_r, y_s, t_n), \quad \forall r = 0, \dots, Nx, s = 0, \dots, Ny.$$

- Explicit scheme

$$\frac{U_{r,s}^{n+1} - U_{r,s}^n}{\Delta t} = b \left[\frac{U_{r+1,s}^n - 2U_{r,s}^n + U_{r-1,s}^n}{(\Delta x)^2} - \frac{U_{r,s+1}^n - 2U_{r,s}^n + U_{r,s-1}^n}{(\Delta y)^2} \right]$$

- Implicit scheme(**Jacobi** and **Gauss Siedel** solver)

$$\frac{U_{r,s}^{n+1} - U_{r,s}^n}{\Delta t} = b \left[\frac{U_{r+1,s}^{n+1} - 2U_{r,s}^{n+1} + U_{r-1,s}^{n+1}}{(\Delta x)^2} - \frac{U_{r,s+1}^{n+1} - 2U_{r,s}^{n+1} + U_{r,s-1}^{n+1}}{(\Delta y)^2} \right]$$

交替方向(隐) Alternative Direction Interaction(ADI) Two dimensional Crank-Nicolson scheme

$$(1 - \frac{1}{2}\mu_x\delta_x^2 - \frac{1}{2}\mu_y\delta_y^2)U^{n+1} = (1 + \frac{1}{2}\mu_x\delta_x^2 + \frac{1}{2}\mu_y\delta_y^2)U^n$$

with a slight modification

$$(1 - \frac{1}{2}\mu_x\delta_x^2)(1 - \frac{1}{2}\mu_y\delta_y^2)U^{n+1} = (1 + \frac{1}{2}\mu_x\delta_x^2)(1 + \frac{1}{2}\mu_y\delta_y^2)U^n$$

- Peaceman D.W. and Rachford H.H. Jr(1955), The numerical solution of parabolic and elliptic differential equations, J. Soc. Indust. Appl. Math. 3, 28-41.

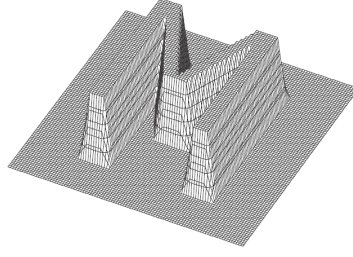
At last, split into two steps as

$$\begin{aligned} (1 - \frac{1}{2}\mu_x\delta_x^2)U^{n+\frac{1}{2}} &= (1 + \frac{1}{2}\mu_y\delta_y^2)U^n \\ (1 - \frac{1}{2}\mu_y\delta_y^2)U^{n+1} &= (1 + \frac{1}{2}\mu_x\delta_x^2)U^{n+\frac{1}{2}} \end{aligned}$$

延伸：算子分裂法
一个算例

$$u_t = u_{xx} + u_{yy}, \quad (x, y) \in (0, 1) \times (0, 1)$$

with given initial function $u(x, y, 0) = f(x, y)$ and fixed value 0 on all the four boundaries.



- set $f(x, y)$ as any function you like, for e.g.,
- try different Δx and Δy , for e.g. $\frac{1}{100}, \frac{1}{200}, \frac{1}{400}$
- try the implicit scheme and the ADI iterative method

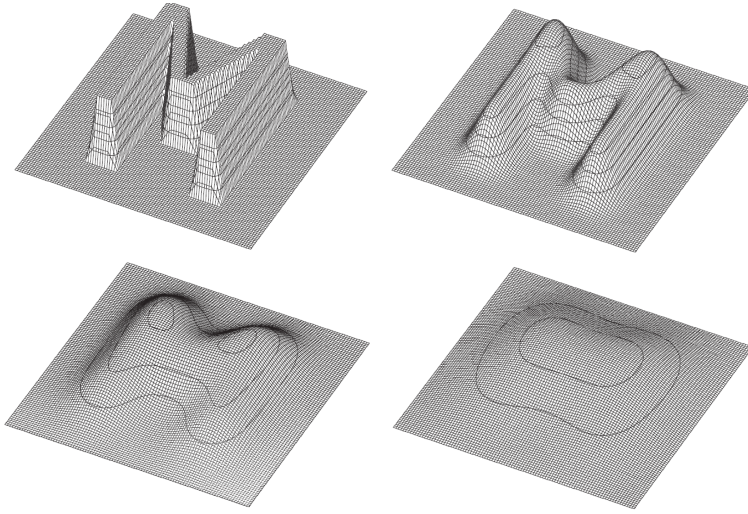


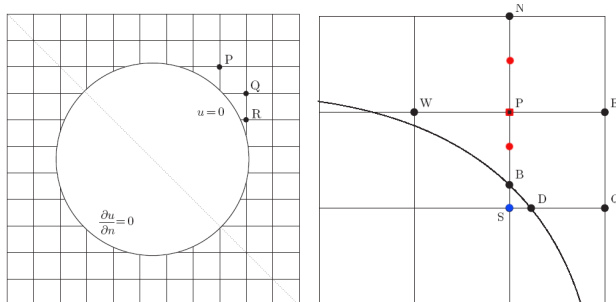
Figure 5.1: Numerical solution at $t = 0$ (Upper Left), $t = 0.001$ (Upper right), $t = 0.004$ (Lower Left) and $t = 0.01$ (Lower Right)

非规则区域的边界处理

1. Set up equation at P with non-uniform finite difference scheme
2. Firstly **Extrapolating**(外插) at S with B, N and P , and take S and the new boundary point, for e.g. with second order,

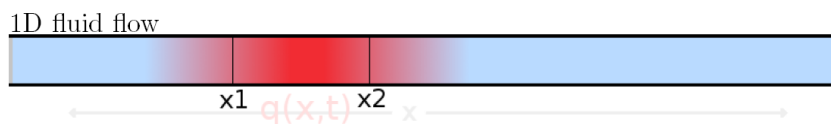
$$U_S = \frac{\alpha(1 - \alpha)u_N + 2u_B - 2(1 - \alpha^2)u_P}{\alpha(\alpha + 1)},$$

where $|PB| = \alpha|PS| := \alpha\Delta y$



5.3 双曲型

输运项(流通量)概念 以如下一维情形为例



$$\int_{x_1}^{x_2} q(x, t) dx = \text{mass of tracer between } x_1 \text{ and } x_2.$$

$$\frac{d}{dt} \int_{x_1}^{x_2} q(x, t) dx = F_1(t) - F_2(t),$$

where F_i is the flux of mass from right to left at x_i .

守恒(Conservation)的积分形式

微分形式: 守恒律

Shengtai Li, An HLLC Riemann solver for magneto-hydrodynamics, J. Comp. Phys., 203, 344-357, 2005.

1. 线性化 例: the advection equation

$$\begin{cases} \omega_t + \lambda \omega_x = 0, \\ \omega(x, 0) = \omega_0(x) \end{cases}$$

solved with the method of characteristics $\omega(x, t) = \omega_0(x - \lambda t)$.

Boundary condition for IBVP($a \leq x \leq b$)?

依赖域(Domain of dependence) 左图
影响域(Range of Influence) 右图

For general autonomous flux $F = f(q)$, we have

$$\frac{d}{dt} \int_{x_1}^{x_2} q(x, t) dx = f(q(x_1, t)) - f(q(x_2, t)).$$

For f sufficiently smooth, we have:

$$\frac{d}{dt} \int_{x_1}^{x_2} q(x, t) dx = - \int_{x_1}^{x_2} \frac{\partial}{\partial x} f(q(x, t)) dx,$$

which we can write as

$$\int_{x_1}^{x_2} \left[\frac{\partial}{\partial t} q(x, t) + \frac{\partial}{\partial x} f(q(x, t)) \right] dx = 0.$$

A 1D *quasilinear* system

$$q_t + A(q, x, t)q_x = 0$$

is *hyperbolic* at (q, x, t) if $A(q, x, t)$ is diagonalizable with real eigenvalues.

The 1D nonlinear conservation law

$$q_t + f(q)_x = 0$$

is hyperbolic if the Jacobian matrix $\frac{\partial f}{\partial q}$ is diagonalizable with real eigenvalues for each physically relevant q .

2. Riemann Problem/Solver

The hyperbolic equation with initial data

$$q_0(x) = \begin{cases} q_l & x < 0 \\ q_r & x > 0 \end{cases}$$

is known as the Riemann problem.

For the linear constant-coefficient system, the solution is

$$\begin{aligned} q(x, t) &= q_l + \sum_{p: \lambda^p < x/t} [l^p(q_r - q_l)] r^p \\ &= q_r - \sum_{p: \lambda^p \geq x/t} [l^p(q_r - q_l)] r^p \end{aligned}$$

Consider the linear hyperbolic IVP

$$\begin{cases} q_t + Aq_x = 0, \\ q(x, 0) = q_0(x) \end{cases}$$

Then we can write $A = R\Lambda R^{-1}$, where $R \in \mathbb{R}^{m \times m}$ is the matrix of eigenvectors and $\Lambda \in \mathbb{R}^{m \times m}$ is the matrix of eigenvalues. Making the substitution $q = Rw$, we get the decoupled system

$$w_t^p + \lambda^p w_x^p = 0, \quad p = 1 \dots m.$$

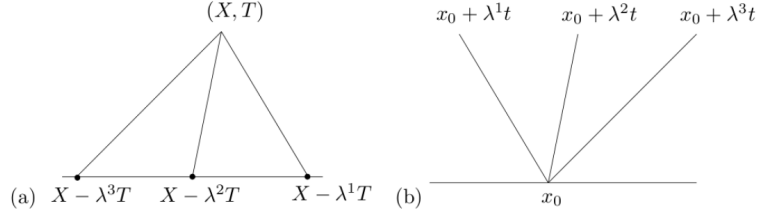


Fig. 3.2. For a typical hyperbolic system of three equations with $\lambda^1 < 0 < \lambda^2 < \lambda^3$, (a) shows the domain of dependence of the point (X, T) , and (b) shows the range of influence of the point x_0 .

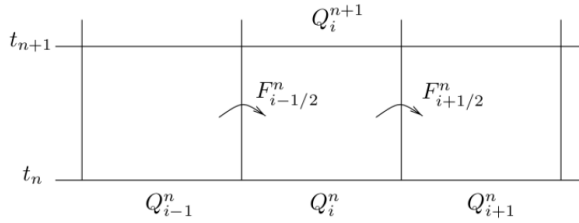
(R. Leveque, 2002)

5.3.1 2.有限体积法(Finite Volume Method)

Denote cells $C_i = (x_{i-1/2}, x_{i+1/2})$ and mean values on cells

$$Q_i^n \approx \frac{1}{|C_i|} \int_{C_i} q(x, t_n) dx.$$

FVM update Q_i^{n+1} based on the fluxes F^n between the cells



FVM scheme for 1D conservation law 积分形式的守恒律 (Remember that $C_i := [x_{i-1/2}, x_{i+1/2}]$):

$$\frac{d}{dt} \int_{x_{i-1/2}}^{x_{i+1/2}} q(x, t) dx = f(q(x_{i-1/2}, t)) - f(q(x_{i+1/2}, t)).$$

时间方向从 t_n 到 t_{n+1} 积分后同除以 Δx : 根据平均流量 Q 和流通量 F 的定义:

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} (F_{i+1/2}^n - F_{i-1/2}^n),$$

$$\begin{aligned} \frac{1}{\Delta x} \int_{C_i} q(x, t_{n+1}) dx &= \frac{1}{\Delta x} \int_{C_i} q(x, t_n) dx \\ &\quad - \frac{1}{\Delta x} \left[\int_{t_n}^{t_{n+1}} f(q(x_{i+1/2}, t)) dt - \int_{t_n}^{t_{n+1}} f(q(x_{i-1/2}, t)) dt \right]. \end{aligned}$$

这里 $F_{i-\frac{1}{2}} \approx \frac{1}{\Delta t} \int_{t_n}^{t_{n+1}} f(q(x_{i-1/2}, t)) dt$.

数值流通量 (Numerical flux)

For a hyperbolic problem, information propagates at a finite speed. So it is reasonable to assume that we can obtain $F_{i-1/2}^n$ using only the values Q_{i-1}^n and Q_i^n :

$$F_{i-1/2}^n = \mathcal{F}(Q_{i-1}^n, Q_i^n)$$

where \mathcal{F} is some *numerical flux function*. Then our numerical method becomes

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} [\mathcal{F}(Q_i^n, Q_{i+1}^n) - \mathcal{F}(Q_{i-1}^n, Q_i^n)].$$

FVM的收敛性 (Convergence)

We say that the numerical solution for a hyperbolic equation is convergent in the meaning of $\Delta x \rightarrow 0$ and $\Delta t \rightarrow 0$, it requires

The method be *consistent*, which promises the local truncation error goes to 0 as $\Delta t \rightarrow 0$. The method be *stable*, which means any small error in each timestep is under control (will not grow too fast)

相容性 (Consistency)

Denote the numerical method as $A^{n+1} = \mathcal{N}(Q^n)$ and the exact value as q^n and q^{n+1} . Then the local truncation error is defined as

$$\tau = \frac{\mathcal{N}(q^n) - q^{n+1}}{\Delta t}$$

We say that the method is *consistent* if τ vanished as $\Delta t \rightarrow 0$ for all smooth $q(x, t)$ satisfying the differential equation. It is usually straightforward when Taylor expansions are used.

稳定性 (Stability)

Courant-Friedrichs-Levy condition: the numerical domain of dependence contains the true domain of dependence domain of the PDE, at least in the limit as $\Delta t, \Delta x \rightarrow 0$

For a hyperbolic system with characteristic wave speeds λ^p ,

$$\frac{\Delta x}{\Delta t} \geq \max_p |\lambda^p|, \quad p = 1, \dots, m.$$

This condition is necessary but not sufficient !

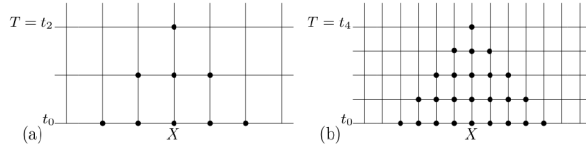


Fig. 4.3. (a) Numerical domain of dependence of a grid point when using a three-point explicit finite difference method, with mesh spacing Δx^a . (b) On a finer grid with mesh spacing $\Delta x^b = \frac{1}{2} \Delta x^a$.

通量(Flux)函数

To do the calculation,

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} (F_{i+\frac{1}{2}}^n - F_{i-\frac{1}{2}}^n),$$

the key step is to compute the numerical flux term

- j2- \bar{c} unstable: $\mathcal{F}(Q_{i-1}^n, Q_{i+1}^n) = \frac{1}{2} [f(Q_{i-1}^n) + f(Q_i^n)]$
- j3- \bar{c} stable: looking into the direction from which the flow come from(upwind), for e.g. $q_t + \lambda q_x = 0$ with $\lambda > 0$, yields

$$Q_i^{n+1} = Q_i^n - \lambda \frac{\Delta t}{\Delta x} (Q_i^n - Q_{i-1}^n) \quad (5.14)$$

Roe 的方案 Recall the numerical method for Conservation Law

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} [\mathcal{F}(Q_i^n, Q_{i+1}^n) - \mathcal{F}(Q_i^n, Q_{i-1}^n)],$$

A linearized choice of the numerical flux based on the Godunov's method for the nonlinear problems. Define $|A| = R|\Sigma|R^{-1}$, where $|\Sigma| = \text{diag}(|\lambda^p|)$, then we can derive the Roe's flux as

$$F_{i-\frac{1}{2}}^n = \frac{1}{2} [f(Q_{i-1}) + f(Q_i)] - \frac{1}{2} |A| [Q_{i-1} + Q_i]$$

Remark: In this sense, R is properly chosen, such that A is a good enough approximation to nonlinear functional \mathcal{F} .

Godunov 的方案 **Remark:** Evolve step (2) requires solving the Riemann problem.

Recall the solution to the Riemann problem form a linear system

$$Q_i - Q_{i-1} = \sum_{p=1}^m [l^p (Q_{i+1} - Q_i)] r^p = \sum_{p=1}^m \mathcal{W}_{i-\frac{1}{2}}^p$$

If Δt is small enough, waves from adjacent cells do not interact!

Godunov's method for General Conservation Laws 最后通过如下“迎风”组合获得流通量表达式

$$F_{i-\frac{1}{2}}^n = f(Q_{i-1}) + \sum_{p=1}^m (\lambda^p)^- \mathcal{W}_{i-\frac{1}{2}}^p,$$

or

$$F_{i-\frac{1}{2}}^n = f(Q_i) + \sum_{p=1}^m (\lambda^p)^+ \mathcal{W}_{i-\frac{1}{2}}^p,$$

The following *REA algorithm* was proposed by Godunov (1959):

1. **Reconstruct** a piecewise polynomial function $\tilde{q}^n(x, t_n)$ from the cell averages Q_i^n . In the simplest case, $\tilde{q}^n(x, t_n)$ is piecewise constant on each grid cell:

$$\tilde{q}^n(x, t_n) = Q_i^n, \quad \text{for all } x \in C_i.$$

2. **Evolve** the hyperbolic equation with this initial data to obtain $\tilde{q}^n(x, t_{n+1})$.
3. **Average** this function over each grid cell to obtain new cell averages

$$Q_i^{n+1} = \frac{1}{\Delta x} \int_{C_i} \tilde{q}^n(x, t_{n+1}) dx.$$

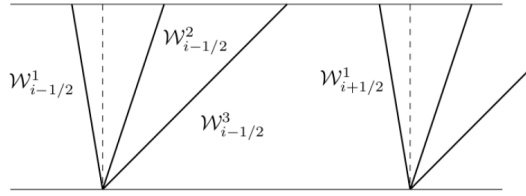


Fig. 4.7. An illustration of the process of Algorithm 4.1 for the case of a linear system of three equations. The Riemann problem is solved at each cell interface, and the wave structure is used to determine the exact solution time Δt later. The wave $W_{i-1/2}^2$, for example, has moved a distance $\lambda^2 \Delta t$ into the cell.

where $\lambda^+ = \max(\lambda, 0)$ and $\lambda^- = \min(\lambda, 0)$ is an upwind choice.

Total Variation Diminision(TVD) 方案 Recall the numerical method for Conservation Law

$$Q_i^{n+1} = Q_i^n - \frac{\Delta t}{\Delta x} [\mathcal{F}(Q_i^n, Q_{i+1}^n) - \mathcal{F}(Q_i^n, Q_{i+1}^n)],$$

where $\mathcal{F}(Q_i^n, Q_{i+1}^n) \approx F_{i+\frac{1}{2}}^n = h(Q_{i+\frac{1}{2}}^-, Q_{i+\frac{1}{2}}^+)$.

TVD: It is required that the numerical flux function $h(\cdot, \cdot)$ is monotone(Lipschitz continuous, monotone, $h(a, a) = a$)

Example

$$h(a, b) = 0.5(f(a) + f(b) - \alpha(b - a)),$$

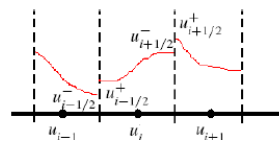
where $\alpha = \max_u |f'(u)|$

(Weighted) Essentially Non-Oscillatory((W)ENO) 方案 The main concept of (W)ENO is where $\{u_i\}_{i=0}^n$ are the given **cell average** of a function $q(x)$.

1. 2- ϵ Construct polynomials $p_i(x)$ of degree $k-1$, for each cell C_i , such that it is a k -th order accurate approximation to the function $q(x)$, which means

$$p_i(x) = q(x) + \mathcal{O}(\Delta^k) \quad \forall x \in C_i, i = 0, 1, \dots, N$$

2. 3- ϵ Evaluate u at each cell interface($u_{i+1/2}^-$ and $u_{i+1/2}^+$)



Use ENO/WENO to compute $u_{i+1/2}^{\pm}$

$$u_{i+1/2}^- = p_i(x_{i+1/2}) = v_i(u_{i-r}, \dots, u_{i+s})$$

$$u_{i+1/2}^+ = p_{i+1}(x_{i+1/2}) = v_{i+1}(u_{i-r}, \dots, u_{i+s})$$

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5.3.2 谱方法(Spectral methods)

Lloyd N. Trefethen:

Spectral methods are one of the "big three" technologies for the numerical solution of PDEs, which came into their own roughly in successive decades:

- 1950s: 有限差分方法
- 1960s: 有限元方法
- 1970s: 谱方法

Fast PDE Solver: $u_t + c(x)u_x = 0$

```
% p6.m - variable coefficient wave equation
% Grid, variable coefficient, and initial data:
N = 128; h = 2*pi/N; x = h*(1:N); t = 0; dt = h/4;
c = .2 + sin(x-1).^2;
v = exp(-100*(x-1).^2); vold = exp(-100*(x-.2*dt-1).^2);

% Time-stepping by leap frog formula:
tmax = 8; tplot = .15; clf, drawnow
plotgap = round(tplot/dt); dt = tplot/plotgap;
nplots = round(tmax/tplot);
data = [v; zeros(nplots,N)]; tdata = t;
for i = 1:nplots
    for n = 1:plotgap
        t = t+dt;
        v_hat = fft(v);
        w_hat = 1i*[0:N/2-1 0 -N/2+1:-1] .* v_hat;
        w = real(ifft(w_hat));
        vnew = vold - 2*dt*c.*w; vold = v; v = vnew;
    end
    data(i+1,:) = v; tdata = [tdata; t];
end
waterfall(x,tdata,data), view(10,70), colormap([0 0 0])
axis([0 2*pi 0 tmax 0 5]), ylabel t, zlabel u, grid off
```

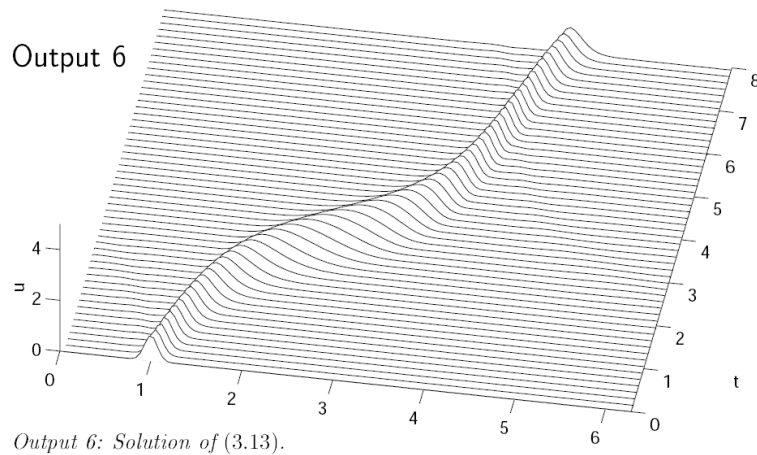
Remark: The examples and figures are from: Trefethen, spectral method in matlab.

4.Maxwell's Equation The governing equations for Electrodynamics are Understanding FDTD: <http://www.eecs.wsu.edu/~schneidj/ufdtd/>

Yee's grid The main concept of the Finite Difference Time Domain(FDTD) method is to define different component of the Electric field $\mathbf{E} := (\mathbf{E}_x, \mathbf{E}_y, \mathbf{E}_z)$ and the magnetic field $\mathbf{H} := (H_x, H_y, H_z)$ at different surface of the rectangular grid, which is very convenient when discretizing the $\nabla \times$ operator using finite difference method.

Perfect Match Layer(PML) A widely used boundary condition in practical calculation for wave scattering problem in the recent twenty years.

Textbook for Computational Electrodynamics



$$\begin{aligned}
 \nabla \cdot \mathbf{D} &= \rho_v, \\
 \nabla \cdot \mathbf{B} &= 0, \\
 \nabla \times \mathbf{E} &= -\frac{\partial \mathbf{B}}{\partial t}, \\
 \nabla \times \mathbf{H} &= \mathbf{J} + \frac{\partial \mathbf{D}}{\partial t}.
 \end{aligned}$$

- A. Taflov and S. C. Hagness, Computational Electrodynamics: The Finite-Difference Time-Domain Method, 3rd ed.

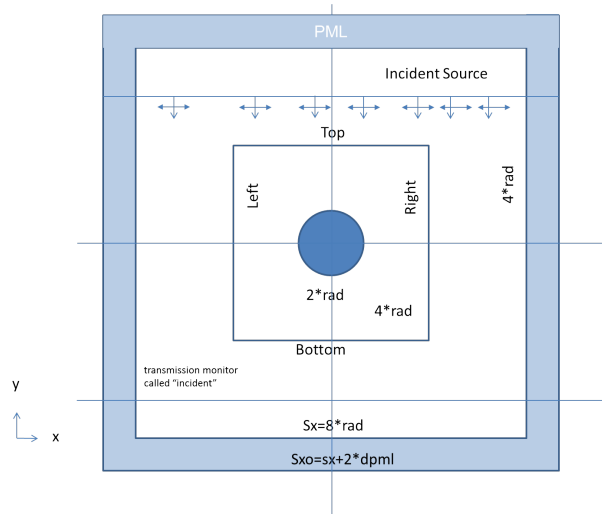
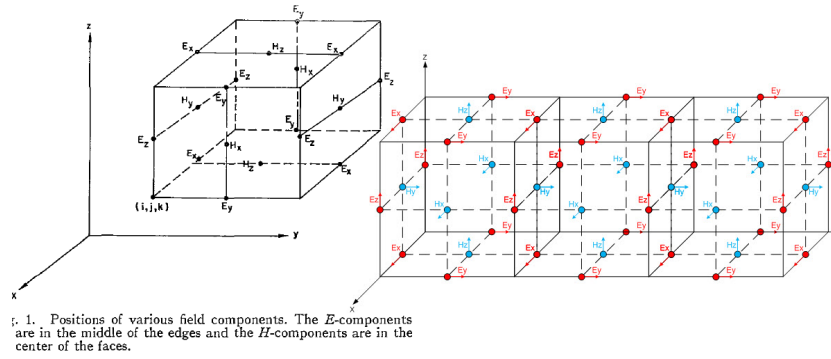
软件包与参考教材

- K. W. Morton and D.F. Mayers: Numerical Solution of Partial Differential Equations (李治平等 中译)
- 陆金甫, 关治: 偏微分方程数值解法
- Lloyd N. Trefethen: Finite Difference and Spectral Methods for Ordinary and Partial Differential Equations, 1996.

Please do not reproduce this text; all rights are reserved. Copies can be obtained for \$22 or £15 each (including shipping) by contacting me directly. Professors who are using the book in class may contact me to obtain solutions to most of the problems. [This is no longer true – the book is freely available online at <http://web.comlab.ox.ac.uk/oucl/work/nick.trefethen/pdetext.html>.]

Nick Trefethen
July 1996

<http://people.maths.ox.ac.uk/trefethen/pdetext.html>



Exercise

1. Convection-Diffusion equation

$$e \frac{\partial u}{\partial t} = \nabla \cdot (b \nabla u - \mathbf{a} u) + cu + d,$$

with proper parameter, boundary and initial conditions

2. Further numerically study of the quenching phenomenon (quenching1d.pdf)
3. Analyze of a two dimensional finite difference scheme (FDscheme2D.pdf)
4. **E. F. Toro**, Riemann Solvers and Numerical Methods for Fluid Dynamics, 3rd, Springer-Verlag.
5. **Ferziger Peric**: Computational Methods for Fluid Dynamics, 3rd eds.

[11,644 total Google Scholar \(GS\) citations](#) for the three editions of *Computational Electrodynamics: The Finite-Difference Time-Domain Method* (Books #2, #4, and #5) as of 7/23/2013.

This ranks 7th on the [Google Scholar search](#) conducted in September 2012 by the Institute of Optics of the University of Rochester for the all-time most-cited books in physics.

1. K. R. Umashankar and A. Taflove, *Computational Electromagnetics: Integral Equation Approach*. Norwood, MA: Artech House, 1993.
2. A. Taflove, *Computational Electrodynamics: The Finite-Difference Time-Domain Method*. Norwood, MA: Artech House, 1995. (Download the [1995 book front matter.pdf](#))
3. A. Taflove, ed., *Advances in Computational Electrodynamics: The Finite-Difference Time-Domain Method*. Norwood, MA: Artech House, 1998. (Download the [1998 book front matter.pdf](#))
4. A. Taflove and S. C. Hagness, *Computational Electrodynamics: The Finite-Difference Time-Domain Method, 2nd ed.* Norwood, MA: Artech House, 2000. (Download the [2000 book front matter.pdf](#))
5. A. Taflove and S. C. Hagness, *Computational Electrodynamics: The Finite-Difference Time-Domain Method, 3rd ed.* Norwood, MA: Artech House, 2005. (Download the [2005 book front matter.pdf](#) ; visit the [Publisher's webpage](#) ; read the book reviews: [IEEE EMC review.pdf](#) and [IEEE APS review.pdf](#))
6. A. Taflove, A. Oskooi, and S. G. Johnson, eds., *Advances in FDTD Computational Electrodynamics: Photonics and Nanotechnology*. Norwood, MA: Artech House, 2013. (Download the [2013 book front matter.pdf](#) ; visit the [Publisher's webpage](#))