

4.2 Neumann Boundary Conditions.

$$\begin{array}{l} \text{I.C.} \quad \left\{ \begin{array}{l} u_t = \sigma u_{xx}, \quad 0 < x < 1, \quad t > 0, \\ u(x, 0) = \phi(x), \quad 0 \leq x \leq 1 \end{array} \right. \\ \text{B.C's} \quad \left\{ \begin{array}{l} u(0, t) = f(t), \\ u_x(1, t) = g(t), \end{array} \right. \quad t > 0 \end{array}$$

↓
Neumann condition.

Let's use the weighted average scheme

$$\begin{aligned} -r\theta u_{j-1}^{n+1} + (1+2r\theta) u_j^{n+1} - r\theta u_{j+1}^{n+1} &= \\ = r(1-\theta) u_{j-1}^n + [1-2r(1-\theta)] u_j^n + r(1-\theta) u_{j+1}^n \end{aligned} \quad (1.1)$$

$$\theta \in [0, 1], \quad r = \sigma \Delta t / \Delta x^2.$$

(I) We already studied the case of ~~periodic~~ initial and boundary conditions over the interval $[0, 2\pi]$

and we were able to prove

a) Scheme (1.1) is unconditionally stable in L_2 norm

when $1/2 \leq \theta \leq 1$, i.e., $\|\tilde{u}^n\|_2 \leq \|\tilde{u}^0\|_2$, for all r .

b) Scheme (1.1) is absolutely stable, i.e., $\|\tilde{u}^n\|_2 \leq \|\tilde{u}^0\|_2$

if $r \leq 1/2(1-2\theta)$, when $0 \leq \theta < 1/2$.

(II) We also obtained a matrix representation for this scheme for homogeneous Dirichlet Condition

(III) Now, if there is a Neumann condition at $x=1$:

$$\frac{\partial u}{\partial x}(1, t) = g(t)$$

and Dirichlet condition at $x=0$:

$$u(0, t) = f(t).$$

and the I.C. $u(x, 0) = \phi(x), \quad 0 \leq x \leq 1.$

Then, $U_j^0 = \phi(x_j), \quad j = 0, 1, \dots, J.$

For the subsequent level in times $t_n = n \Delta t, \quad n = 1, 2, \dots$

Eqn. (1.1). is still the same for $j = 1, 2, \dots, J-1$

But now, for $j = J-1$

$$\underbrace{j = J-1}_{\text{new unknown}} \left[-r\theta U_{J-2}^{n+1} + (1+2r\theta)U_{J-1}^{n+1} - r\theta \underbrace{U_J^{n+1}}_{\text{new unknown}} = \right. \quad (2.1) \\ \left. r(1-\theta)U_{J-2}^n + [1-2r(1-\theta)]U_{J-1}^n + r(1-\theta)U_J^n \right]$$

The term U_J^n is known from the previous time step

and U_J^{n+1} is a new unknown. Since, we have $J-1$ eqns. and J unknowns one more equation is needed.

Alternatives:

a) Approximate (Backward 2 points)

$$(u_x)_J^{n+1} \approx \frac{u_J^{n+1} - u_{J-1}^{n+1}}{\Delta x} = g^{n+1}$$

Then $\Rightarrow u_J^{n+1} = u_{J-1}^{n+1} + \Delta x g^{n+1}$ New equ.

Disadvantage: Approx. $O(\Delta x)$ lower order than scheme.

b) Approximate (Backward 3 points)

$$(u_x)_J^{n+1} = \frac{u_{J-2}^{n+1} - 4u_{J-1}^{n+1} + 3u_J^{n+1}}{2\Delta x} + O(\Delta x^2).$$

O.K. approx. of u_x same order as scheme.

New equ.

$$u_J^{n+1} = \frac{1}{3} [4u_{J-1}^{n+1} - u_{J-2}^{n+1} + 2\Delta x g^{n+1}]$$

c) Approximation (Centered 3 points)

$$(u_x)_J^{n+1} = \frac{u_{J+1}^{n+1} - u_{J-1}^{n+1}}{2\Delta x} + O(\Delta x^2)$$

$\Rightarrow$ New Equ.

$$\boxed{u_{J+1}^{n+1} = u_{J-1}^{n+1} + 2\Delta x g^{n+1}} \quad (3.1)$$

Disadvantage:

Add. one more unknown U_{J+1}^{n+1} Value of discrete approximation at a fictitious point (x_{J+1}, t_{n+1}) , also called Ghost Point.

Now, we have J equations and $J+1$ unknowns.

$$U_j^{n+1}, \quad j = 1, 2, \dots, J, J+1.$$

We can add one more equation if we replace $j \rightarrow J$ in (2.1). In fact,

$$\begin{aligned} -r\theta U_{J-1}^{n+1} + (1+2r\theta)U_J^{n+1} - r\theta U_{J+1}^{n+1} = \\ r(1-\theta)U_{J-1}^n + [1-2r(1-\theta)]U_J^n + r(1-\theta)U_{J+1}^n \end{aligned}$$

New unknown U_{J+1}^n

Another new unknown appears U_{J+1}^n . Now, we have $J+1$ eqns. and $J+2$ unknowns. An additional eqn. can be obtained from (3.1) again (4.1)

replacing $n+1 \rightarrow n$

$$U_{J+1}^n = U_{J-1}^n + 2\Delta x g^n \quad (4.2)$$

Summarizing:

System of equations for Neumann condition

$$\frac{\partial u}{\partial x}(L, t) = g(t), \text{ at } x = L$$

and Dirichlet cond. at $x = 0$

$$u(0, t) = f(t).$$

$$j=1: (1+2r\theta)U_1^{n+1} - r\theta U_2^{n+1} = r(1-\theta)\overset{f^n}{U_0^n} + [1-2r(1-\theta)]U_1^n + r(1-\theta)U_2^n + r\theta \boxed{U_0^{n+1} = f^{n+1}} \text{ Known B.C.}$$

$$j=2, \dots, J-1: -r\theta U_{j-1}^{n+1} + (1+2r\theta)U_j^{n+1} - r\theta U_{j+1}^{n+1} = r(1-\theta)U_{j-1}^n + [1-2r(1-\theta)]U_j^n + r(1-\theta)U_{j+1}^n$$

$$j=J: -r\theta U_{J-1}^{n+1} + (1+2r\theta)U_J^{n+1} - r\theta \overset{\text{New fictitious}}{U_{J+1}^{n+1}} = r(1-\theta)U_{J-1}^n + [1-2r(1-\theta)]U_J^n + r(1-\theta) \overset{\text{New fictitious}}{U_{J+1}^n}$$

Two more equation from approx. of Neumann B.C.

$$U_{J+1}^n = U_{J-1}^n + 2\Delta x (U_x)_{J-1}^n \overset{= g^n}{=}$$

$$U_{J+1}^{n+1} = U_{J-1}^{n+1} + 2\Delta x (U_x)_{J-1}^{n+1} \overset{= g^{n+1}}{=}$$

Centered
Finite
Diff. Approx.

$J+2$ equations and $J+2$ unknowns.

(5)

The last 3 equations are combined in one and the values U_{J+1}^n and U_{J+1}^{n+1} at fictitious points are eliminated. So,

$$J = J: -2r\theta U_{J-1}^{n+1} + (1+2r\theta)U_J^{n+1} = 2r(1-\theta)U_{J-1}^n + \quad (5.1) \\ + [1-2r(1-\theta)]U_J^n + 2r(1-\theta)\Delta x g^n + 2r\theta\Delta x g^{n+1}$$

Now, we have J equations with

$U_1^{n+1}, \dots, U_J^{n+1}$ J unknowns.

Assuming $f(t) = U(0,t) = 0$ and $g(t) = U_x(1,t) = 0$

The linear system to be solved reduces to

$$[I - r\theta C] \tilde{U}^{n+1} = [I + r(1-\theta)C] \tilde{U}^n, \quad (5.2)$$

where

$$\tilde{U}^n = \begin{bmatrix} U_1^n \\ \vdots \\ U_J^n \end{bmatrix},$$

$$C = \begin{bmatrix} -2 & 1 & 0 & 0 & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ & \ddots & \ddots & \ddots & \ddots & \vdots \\ & & & 1 & -2 & 1 \\ & & & & 2 & -2 \end{bmatrix}_{J \times J}.$$

Also (5.2) can also be written as

$$\tilde{U}^{n+1} = [I - r\theta C]^{-1} [I + r(1-\theta)C] \tilde{U}^n = L_{\Delta} \tilde{U}^n$$

We want to study the stability of this numerical (6.1) Scheme using matrix analysis.

Useful Linear Algebra results:

Theorem.

If A, B ($n \times n$) matrices are similar, then they have same characteristic polynomials and hence same eigenvalues.

Proof.

$$\text{If } B = P^{-1}AP$$

$$\Rightarrow B - \lambda I = P^{-1}AP - \lambda P^{-1}P = P^{-1}(AP - \lambda P) = P^{-1}(A - \lambda I)P$$

$$\Rightarrow \det(B - \lambda I) = \det(P^{-1}) \det(A - \lambda I) \det(P)$$

$$= \det(A - \lambda I) \det(P^{-1}) \det(P)$$

$$= \det(A - \lambda I) \det(P^{-1}P) = \det(A - \lambda I).$$

Theorem.

If L_Δ is similar to a symmetric matrix Q

and $L_\Delta = P^{-1} Q P.$

and $\|P\|$ and $\|P^{-1}\|$ are uniformly bounded.

The condition (Von Neumann condition)

$$\rho(L_\Delta) \leq 1 + \hat{C} \Delta t, \quad \Delta t \leq \Delta t^*,$$

$$\Delta x \leq \Delta x^*$$

and $\hat{C}$ independent of $\Delta t, \Delta x$ $n \Delta t \leq T$

is sufficient for stability in the L_2 -norm.

Proof.

Notice

If $L_\Delta = P^{-1} Q P$, Q Symm

$$\begin{aligned} L_\Delta^2 &= (P^{-1} Q P)(P^{-1} Q P) = (P^{-1} Q)(P P^{-1}) P \\ &= P^{-1} Q^2 P \end{aligned}$$

In general,

$$L_\Delta^n = P^{-1} Q^n P.$$

Then,

$$\begin{aligned}
 \|L_\Delta^n\|_2 &\leq \|P^{-1} Q^n P\| \leq \|P^{-1}\|_2 \|Q^n\|_2 \|P\|_2 \leq \|P^{-1}\|_2 \|Q\|_2^n \|P\|_2 \\
 &\stackrel{\text{Symm } Q}{=} \|P^{-1}\| \|P\| \rho^n(Q) \stackrel{\text{Prev. Thm}}{=} \|P^{-1}\| \rho^n(L_\Delta) \|P\| \\
 &\leq K \rho^n(L_\Delta) \stackrel{\text{Hip.}}{\leq} K (1+c\Delta t)^n \leq K e^{c\Delta t} \\
 &\leq K e^{cT} = \hat{C}
 \end{aligned}$$

Thus, $\|L_\Delta^n\|_2 \leq \hat{C}$ and FDM is stable.

So our goal in what follows is to apply this theorem to show stability in the L_2 -norm for the weighted average scheme.

Back to stability analysis of the weighted average scheme

$$\vec{U}^{n+1} = L_{\Delta} \vec{U}^n$$

Where

$$L_{\Delta} = [I - r\theta C]^{-1} [I + r(1-\theta)C]$$

According to previous theorem, we need to prove:

① L_{Δ} is similar to a symmetric matrix.

② $\rho(L_{\Delta}) \leq 1 + C\Delta t$, $\Delta x \leq \Delta x^*$
 $\Delta t \leq \Delta t^*$
 $n \rightarrow \infty, n\Delta t \leq T$.

We will actually prove a stronger condition

$$\rho(L_{\Delta}) \leq 1.$$

Proof of ① requires to find P, P^{-1} and $\tilde{L}_{\Delta}$ such that

$$L_{\Delta} = P^{-1} \tilde{L}_{\Delta} P \quad \text{and} \quad (\tilde{L}_{\Delta})^T = \tilde{L}_{\Delta}.$$

$$\text{or } \tilde{L}_{\Delta} = \tilde{D}^{-1} L_{\Delta} \tilde{D}, \quad \text{where } \tilde{D}^{-1} = P, \quad \tilde{D} = P^{-1}$$

Lemma 1

If D is invertible

$$\tilde{L}_\Delta = D^{-1} L_\Delta D = (I - r\theta \tilde{C})^{-1} (I + r(1-\theta)\tilde{C})$$

where $\tilde{C} = D^{-1} C D$.

Proof.

$$\begin{aligned} \tilde{L}_\Delta &= D^{-1} L_\Delta D = (D^{-1} [I - r\theta C] D)^{-1} (D^{-1} (I + r(1-\theta)C) D) \\ &= [D^{-1} [I - r\theta C] D]^{-1} (I + r(1-\theta) D^{-1} C D) \\ &= (I - r\theta D^{-1} C D)^{-1} (I + r(1-\theta) D^{-1} C D) \end{aligned}$$

If $\tilde{C} = D^{-1} C D$ then,

$$\tilde{L}_\Delta = [I - r\theta \tilde{C}]^{-1} [I + r(1-\theta)\tilde{C}]$$

Lemma 2. If $\tilde{C}$ is symm. ($\tilde{C}^T = \tilde{C}$)

then, $[I - r\theta \tilde{C}]$ and $[I + r(1-\theta)\tilde{C}]$ are also symm.

Proof. $[I - r\theta \tilde{C}]^T = I - r\theta \tilde{C}^T = [I - r\theta \tilde{C}]$

Similarly $[I + r(1-\theta)\tilde{C}]^T = [I + r(1-\theta)\tilde{C}]$

Theorem. If $\tilde{C} = D^{-1}CD$ is symmetric $\Rightarrow \tilde{L}_\Delta$ is symmetric.

Proof. Calling $A \equiv [I - r\theta\tilde{C}]$, $B \equiv [I + r(1-\theta)\tilde{C}]$

$$\tilde{L}_\Delta = A^{-1}B$$

Therefore,

$$\tilde{L}_\Delta^T = (A^{-1}B)^T = B^T (A^{-1})^T = B^T (A^T)^{-1} \stackrel{\text{lemma 2}}{=} B A^{-1}$$

So for showing symm. is enough to prove that

$$A^{-1}B = BA^{-1} \quad (\text{commute}).$$

We first notice that $AB = BA$
by direct product of matrices.

$$\text{Now, } (BA)^{-1}(BA) = I \Rightarrow (AB)^{-1}(BA) = I$$

$$\text{Thus, } B^{-1}(A^{-1}BA) = I \Rightarrow A^{-1}BA = B$$

$$\text{Then, } (A^{-1}BA)A^{-1} = BA^{-1}$$

$$\text{or } A^{-1}B = (A^{-1}B)(AA^{-1}) = BA^{-1} \quad \checkmark$$

which is equivalent to

$$\tilde{L}_\Delta = \tilde{L}_\Delta^T \quad \checkmark$$

Similarly, if

C were symm. $\Rightarrow L_{\Delta}$ would be symm.

but C is not symm. for Neumann boundary condition.

However, according to our previous theorems and lemmas

If we can find D invertible such that

$$\tilde{C} = D^{-1} C D$$

is symmetric, then $\tilde{L}_{\Delta} = D^{-1} L_{\Delta} D$ is also symmetric.

Corollary. $\tilde{L}_{\Delta} = D^{-1} L_{\Delta} D$, where $D = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & & \ddots & & \vdots \\ 0 & & & 1 & 0 \\ 0 & & & 0 & \sqrt{2} \end{bmatrix}$

is symmetric.

Proof. According to previous theorem, it is enough to show that $\tilde{C} = D^{-1} C D$ is symm.

In fact,

$$\tilde{C} = D^{-1} C D = \begin{bmatrix} -2 & 1 & 0 & \dots & 0 \\ 1 & -2 & 1 & \dots & 0 \\ 0 & & \ddots & & \vdots \\ \vdots & & & 1 & -2 \\ 0 & & & 1 & \sqrt{2} \\ 0 & & & \sqrt{2} & -2 \end{bmatrix} \text{ is clearly symm.}$$

$$\tilde{C}^T = \tilde{C}.$$

In fact,

CD

$$\begin{bmatrix} -2 & 1 & 0 & 0 & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ & & \ddots & & & \\ & & & 1 & -2 & 1 \\ & & & & 2 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 0 \\ 1 \\ \sqrt{2} \end{bmatrix}$$

$$= \begin{bmatrix} -2 & 1 & 0 & 0 & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ & & \ddots & & & \\ & & & 1 & -2 & \sqrt{2} \\ & & & & 2 & -2\sqrt{2} \end{bmatrix}$$

 $D^{-1}(CD)$

$$\begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \\ \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} -2 & 1 & 0 & 0 & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ & & \ddots & & & \\ & & & 1 & -2 & \sqrt{2} \\ & & & & 2 & -2\sqrt{2} \end{bmatrix} = \begin{bmatrix} -2 & 1 & 0 & \dots & 0 \\ 1 & -2 & 1 & & \\ & & \ddots & & \\ & & & 1 & -2 & \sqrt{2} \\ & & & 0 & \sqrt{2} & -2 \end{bmatrix}$$

To prove stability of

$$\hat{U}^{n+1} = L_{\Delta} \hat{U}^n$$

where $L_{\Delta} \equiv [I - r\theta]^{-1} [I + r(1-\theta)C]$

We still need to show that

$$\rho(L_{\Delta}) \leq 1 + c\Delta t$$

We can actually show

$$\rho(L_{\Delta}) \leq 1.$$

Lemma 3.

If λ is an eigenvalue of L_{Δ} , i.e., there is $\bar{z} \neq \vec{0}$

such that $L_{\Delta} \bar{z} = \lambda \bar{z}$, or $[L_{\Delta} - \lambda I] \bar{z} = \vec{0}$.

then

$$\lambda = 1 + \frac{\mu r}{1 - \mu r \theta}$$

where μ is an eigenvalue of C , i.e.,

$$C \bar{v} = \mu \bar{v}, \quad \bar{v} \neq \vec{0}.$$

Proof. If $L_{\Delta} \vec{z} = \lambda \vec{z}$

then

$$[I - r\theta C]^{-1} [I + r(1-\theta)C] \vec{z} = \lambda \vec{z}$$

Thus,

$$[I + r(1-\theta)C] \vec{z} = \lambda [I - r\theta C] \vec{z}$$

$$\Rightarrow ([I + r(1-\theta)C] - \lambda [I - r\theta C]) \vec{z} = 0$$

$$\text{or } ([r(1-\theta) + \lambda r\theta]C - (\lambda - 1)I) \vec{z} = 0$$

$$\Rightarrow 0 \vec{z} = \left(\frac{\lambda - 1}{r(1-\theta) + \lambda r\theta} \right) \vec{z}$$

$$\Rightarrow \mu = \frac{\lambda - 1}{r(1-\theta) + \lambda r\theta} \text{ is an eigenvalue of } G. \quad (14.1)$$

Solving for λ , we will get the eigenvalues of L_{Δ} in terms of the eigenvalues of G .

$$\text{From (14.1)} \quad \lambda - 1 = r(1-\theta)\mu + \lambda r\theta\mu$$

$$\Rightarrow \lambda(1 - \mu r\theta) = 1 + r(1-\theta)\mu = \mu r + 1 - r\theta\mu.$$

$$\Rightarrow \boxed{\lambda = 1 + \frac{\mu r}{1 - \mu r\theta}} \quad \checkmark \quad (14.2)$$

Lemma 4.

The eigenvalues of G and corresponding eigenvectors $\vec{z}^k$ are

$$\mu_k = -4 \sin^2 \left(\frac{k\pi}{4J} \right), \quad k = 2m-1, \\ m = 0, 1, \dots, J-1$$

$$\underline{\vec{z}_j^k = \sin \left(\frac{k\pi j}{2J} \right)}$$

Theorem.

The spectral radius

$$\rho(L_\Delta) \leq 1,$$

1) if $\frac{1}{2} \leq \theta \leq 1$, for all values of $r = \frac{\sigma \Delta t}{\Delta x^2}$.

2) if $0 \leq \theta < \frac{1}{2}$, for $r \leq \frac{1}{2(1-2\theta)}$

Proof.

From the previous Lemmas 3, 4

$$\rho(L_\Delta) \leq 1 \Leftrightarrow |\lambda_k| \leq 1 \Leftrightarrow -1 \leq 1 - \frac{4r \sin^2 \left(\frac{k\pi}{4J} \right)}{1 + 4r\theta \sin^2 \left(\frac{k\pi}{J} \right)} \leq 1$$

$$\Leftrightarrow \frac{4r \sin^2(k\pi/J)}{1 + 4r\theta \sin^2(k\pi/J)} \leq 2$$

$$\Leftrightarrow 4r \sin^2(k\pi/J) \leq 2 + 8r\theta \sin^2(k\pi/J) / 2$$

$$2r(1-2\theta) \sin^2(k\pi/J) \leq 1. \quad (16.1)$$

To satisfy (16.1) is sufficient

$$2r(1-2\theta) \leq 1 \quad (16.2)$$

For this to happen two cases arise:

$$1) \text{ If } \frac{1}{2} \leq \theta \leq 1 \Rightarrow 1-2\theta \leq 0 \\ \Rightarrow (16.2) \text{ is verified for any } r.$$

$$2) \text{ If } 0 \leq \theta < \frac{1}{2} \Rightarrow 1-2\theta > 0 \\ \text{and } (16.2) \text{ is verified only if} \\ r \leq \frac{1}{2(1-2\theta)}.$$

Corollary. The weighted average scheme ^(WAS) is

1) Unconditionally stable (no restriction in $\Delta t, \Delta x$)
if $\frac{1}{2} \leq \theta \leq 1$.

2) If $0 \leq \theta < \frac{1}{2}$ then W.A.S. is stable
if $r = \frac{\sigma \Delta t}{\Delta x^2} \leq \frac{1}{2(1-2\theta)}.$