

HW #6

6a 1. $T^{n+1} = AT^n$, $T_j^{n+1} = T_j^n + \frac{\kappa \Delta t}{(\Delta x)^2} (T_{j+1}^n - 2T_j^n + T_{j-1}^n)$

In lecture we set $\alpha = \frac{\kappa \Delta t}{(\Delta x)^2} \Rightarrow T_j^{n+1} = T_j^n + \alpha (T_{j+1}^n - 2T_j^n + T_{j-1}^n)$

$$A = \begin{pmatrix} 0 & \alpha & 1-2\alpha & \alpha & 0 & \dots \\ 0 & \dots & 0 & \alpha & 1-2\alpha & \alpha & 0 & \dots & 0 \\ \dots & 0 & \alpha & 1-2\alpha & \alpha & 0 & \dots \end{pmatrix} \begin{pmatrix} T_0^n \\ T_1^n \\ \vdots \\ T_j^n \\ \vdots \\ T_{j-1}^n \\ T_j^n \end{pmatrix}$$

First/Last rows & cols ^{can} change due to boundary conditions. Ignoring the boundary conditions, A is a tridiagonal matrix of $[\alpha, 1-2\alpha, \alpha]$

2. $T_j^{n+1} = T_j^n + \frac{\kappa \Delta t}{(\Delta x)^2} (T_{j+1}^n - 2T_j^n + T_{j-1}^n)$, $T_j^n = \sum^N(k) e^{ik_j \Delta x}$

$$\sum^N(k) e^{ik_j \Delta x} = \sum^N(k) e^{ik_j \Delta x} + \frac{\kappa \Delta t}{(\Delta x)^2} \left(\sum^N(k) e^{ik_{j+1} \Delta x} - 2 \sum^N(k) e^{ik_j \Delta x} + \sum^N(k) e^{ik_{j-1} \Delta x} \right)$$

$$\Rightarrow \xi(k) = 1 + \frac{\kappa \Delta t}{(\Delta x)^2} (e^{iK \Delta x} - 2 + e^{-iK \Delta x}), \quad \left(\frac{e^{ix} + e^{-ix}}{2i} \right)^2 = \sin^2(x)$$

$$\Rightarrow \xi(k) = 1 - 4 \frac{\kappa \Delta t}{(\Delta x)^2} \sin^2\left(\frac{K \Delta x}{2}\right)$$

3. We need $\sin^2\left(\frac{K \Delta x}{2}\right) = 1$

$$\Rightarrow 1 \geq |\xi(k)| = 1 - 4 \frac{\kappa \Delta t}{(\Delta x)^2} \sin^2\left(\frac{K \Delta x}{2}\right) \Rightarrow 1 \geq 1 - 4 \frac{\kappa \Delta t}{(\Delta x)^2}$$

$$\Rightarrow 2 \geq 4 \frac{\kappa \Delta t}{(\Delta x)^2} \Rightarrow \Delta t = \frac{1}{2} \frac{(\Delta x)^2}{\kappa \Delta t}$$

4. Since Δt correlates with $(\Delta x)^2$, you would need to reduce Δt by a factor of $\frac{4}{\pi}$ in order to double the resolution.

6b

1. Implicit $\Rightarrow T^{N+1} = A^{-1} T^N$

$$\Rightarrow T_j^N = -\alpha T_{j+1}^{N+1} + (1 + 2\alpha) T_j^{N+1} - \alpha T_{j-1}^{N+1}, \text{ where } \alpha = \frac{K\Delta t}{(\Delta x)^2}$$

$$A = \begin{pmatrix} 1+\alpha & -\alpha & 0 & \dots & 0 \\ -\alpha & 1+\alpha & \alpha & 0 & \dots \\ 0 & -\alpha & 1+\alpha & 0 & \dots \\ 0 & 0 & -\alpha & 1+\alpha & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & -\alpha & 1+\alpha \\ \dots & 0 & 0 & -\alpha & 1+\alpha \end{pmatrix}$$

Tridiagonal matrix of $[-\alpha, 1+\alpha, \alpha]$,

2.

$$\begin{aligned} \sum_{j=0}^{N+1} \xi_j^{N+1} e^{ik_j \Delta x} &= \sum_{j=0}^N \xi_j^N e^{ik_j \Delta x} + \frac{K\Delta t}{(\Delta x)^2} \left(\sum_{j=0}^N \xi_j^N e^{ik_{j+1} \Delta x} - 2 \sum_{j=0}^N \xi_j^N e^{ik_j \Delta x} + \sum_{j=0}^N \xi_j^N e^{ik_{j-1} \Delta x} \right) \\ \Rightarrow \mathcal{Z} &= \xi^N(k) + \frac{K\Delta t}{(\Delta x)^2} \left(e^{ik\Delta x} - 2 + e^{-ik\Delta x} \right) = \xi^N(k) + \frac{K\Delta t}{(\Delta x)^2} \left(e^{\frac{ik\Delta x}{2}} - e^{-\frac{ik\Delta x}{2}} \right) \\ \Rightarrow \xi^N(k) &= \frac{1}{1 + 4 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{k\Delta x}{2}\right)} \Rightarrow \boxed{\xi(k) = \frac{1}{1 + 4 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{k\Delta x}{2}\right)}} \end{aligned}$$

3. Since $\left| \xi(k) \right| = \frac{1}{1 + 4 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{k\Delta x}{2}\right)} \leq 1,$

The implicit method is unconditionally stable for any Δt .

4. $T_j^N = \sum_{k=0}^N \xi(k) e^{ik_j \Delta x}, \quad \xi(k) = \frac{1}{1 + 4 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{k\Delta x}{2}\right)}$

As $\Delta t \rightarrow \infty, \xi(k) \rightarrow 0 \Rightarrow \boxed{T_j^N \rightarrow 0}$

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6c. 1. $T_j^{N+1} = T_j^N + \frac{1}{2} \left(\frac{K\Delta t}{(\Delta x)^2} \left(\underset{\substack{\uparrow \\ \text{explicit}}}{T_j^N} + \underset{\substack{\uparrow \\ \text{implicit}}}{T_j^{N+1}} \right) \right)$ (We take a half step explicitly and implicitly)

$$\Rightarrow T_j^{N+1} = T_j^N + \frac{1}{2} \frac{K\Delta t}{(\Delta x)^2} (T_{j+1}^N - 2T_j^N + T_{j-1}^N + T_{j+1}^{N+1} - 2T_j^{N+1} + T_{j-1}^{N+1})$$

$$\Rightarrow \hat{T}(k) e^{ik_j \Delta x} = \hat{T}(k) e^{ik_j \Delta x} + \frac{1}{2} \frac{K\Delta t}{(\Delta x)^2} (\hat{T}(k) e^{ik(j+1)\Delta x} - 2\hat{T}(k) e^{ik_j \Delta x} + \hat{T}(k) e^{ik(j-1)\Delta x} + \hat{T}(k) e^{ik(j+1)\Delta x} - 2\hat{T}(k) e^{ik_j \Delta x} + \hat{T}(k) e^{ik(j-1)\Delta x})$$

$$\Rightarrow 1 = \hat{T}^{-1}(k) + \frac{1}{2} \frac{K\Delta t}{(\Delta x)^2} \left[\hat{T}^{-1}(k) (e^{ik\Delta x} - 2 + e^{-ik\Delta x}) + (e^{ik\Delta x} - 2 + e^{-ik\Delta x}) \right]$$

$$\Rightarrow \hat{T}^{-1}(k) \left(1 - 2 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{K\Delta x}{2}\right) \right) = 1 + 2 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{K\Delta x}{2}\right)$$

$$\Rightarrow \hat{T}(k) = \frac{1 - 2 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{K\Delta x}{2}\right)}{1 + 2 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{K\Delta x}{2}\right)}$$

$$2. \hat{T}(k) = \frac{1 - 2 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{K\Delta x}{2}\right)}{1 + 2 \frac{K\Delta t}{(\Delta x)^2} \sin^2\left(\frac{K\Delta x}{2}\right)} \leq 1,$$

so similarly to the implicit method, the crank-nicolson is stable for any Δt .

6d. All in the code.

6e. 1. $u(x,t) = \frac{4u_0}{\pi} \sum_{n=1}^{\infty} \frac{\sin((2n-1)\pi x)}{2n-1} \exp(-(2n-1)^2 \pi^2 t)$

$$\Rightarrow u(x,t) = \frac{4u_0}{\pi} \sum_{n=0}^{\infty} \frac{\sin((2n+1)\pi x)}{2n+1} \exp(-(2n+1)^2 \pi^2 t)$$

2. Seems to be provided.

3. Also seems to be provided

4. $\Delta t \leq \frac{(\Delta x)^2}{2K}$, where $\Delta x = \frac{1}{20}$, $K=1$

$$\Rightarrow \Delta t \leq 1.25 \cdot 10^{-3}, \text{ so it will take } \frac{1}{1.25 \cdot 10^{-3}} = 800 \text{ iterations}$$

$$\Delta t \leq \frac{(\Delta x)^2}{2K}, \text{ where } \Delta x = \frac{1}{40}, K=1$$

$$\Delta t \leq 3.125 \cdot 10^{-4}$$

$\times 4$

3200 iterations

Discussion separately.

$J=20$ ftcs vs implicit vs. CN

$t=1.25 \times 10^{-3}$

For ftcs , the rms difference increases linearly with time to about -5.5 . Results are very similar to the analytical results.

For implicit, the rms difference increases slower than ftcs to about -4.4 , with similar results to the analytical solution.

For CN, the rms difference increases at a rate between ftcs and implicit to about -4.6 , with similar results to the analytical solution.

$J=40$

$t=1.25 \times 10^{-3}$

For ftcs , the method does not produce useful results since it is unstable, since we doubled J , we would need to divide Δt by a factor of 4 to stay stable.

Both the implicit and CN similarly ran with the rms difference being about -4.5 , but with twice the spatial points.

$J=40$

$t=3.125 \times 10^{-4}$

For ftcs , the rms error increases linearly over time to about -6 . Results are very similar to the analytical solution.

The implicit and CN methods ran very similarly with rms differences about -5 .

All three methods produced results very similar to analytical results.