

HW6 Report

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May 2021

Problems 1 and 2:

Both problem 1 and problem 2 are theoretical and are appended in a handwritten format to the end of this report.

Problem 3:

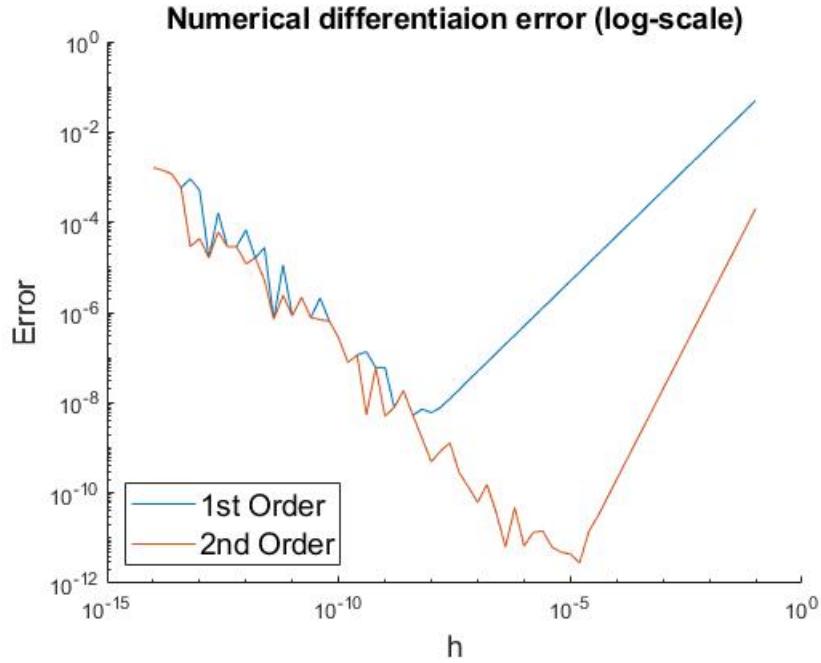
In this problem, we use numerical differentiation (first and second order) to approximate

$$\frac{d}{dx} \sin(x), \quad x = 1.45.$$

The purpose of this is to illustrate that there is an optimal value of h to achieve minimum error, and that a higher order method results in less error. We compare our numerical derivatives against the values of the actual derivative, $\cos(x)$. Below, we provide a loglog plot of both

$$E_{T,1}(h) = |q_1(h) - \cos(x)|, \quad E_{T,2}(h) = |q_2(h) - \cos(x)|,$$

versus h , where $q_1(h)$ and $q_2(h)$ are the first and second order approximations of the derivative, respectively.



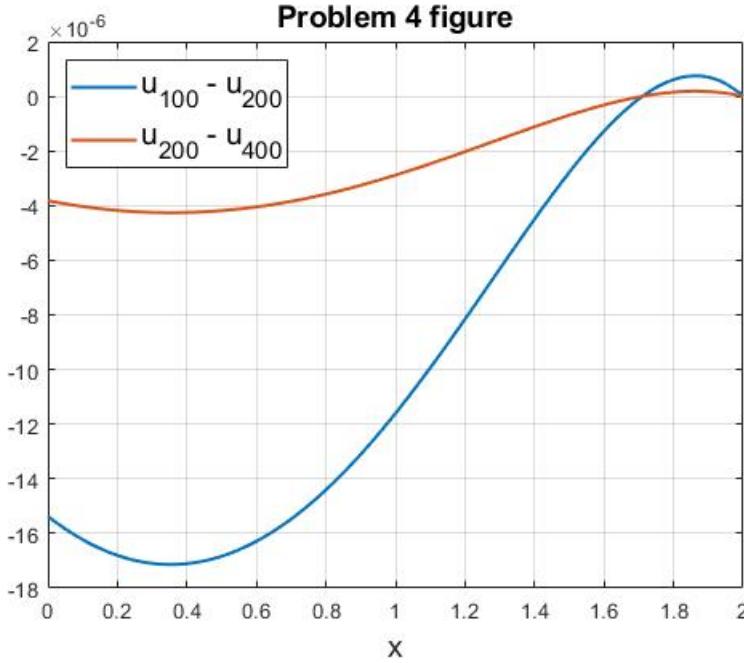
Problem 4:

In this problem, we continue with the same IBVP from problem 6 in assignment 5,

$$\begin{cases} u_t = u_{xx}, & x \in (0, L), t > 0 \\ u(x, 0) = p(x), & x \in (0, L) \\ u_x(0, t) - \alpha u(0, t) = 0., & u(L, t) = q(t) \end{cases}$$

with $L = 2$, $\alpha = 0.4$, $p(x) = (1 - 0.5x)^2$, $q(t) = 2 \sin^2(t)$.

In this problem, we solve the IBVP with differing spatial grids, namely when $N = 100$, $N = 200$, and $N = 400$. The method we use is the FTCS method, and we solve to $T = 3$ with $\Delta t = 10^{-5}$. We then use the spline function in MATLAB to map the solution for each value of N to the grid $x = [0 : 0.002 : 1]^*L$. Below we plot $(u_{\{N=100\}} - u_{\{N=200\}})$ vs x and then $(u_{\{N=200\}} - u_{\{N=400\}})$ vs x .



Problem 5:

In this problem, we solve the 2D IBVP problem

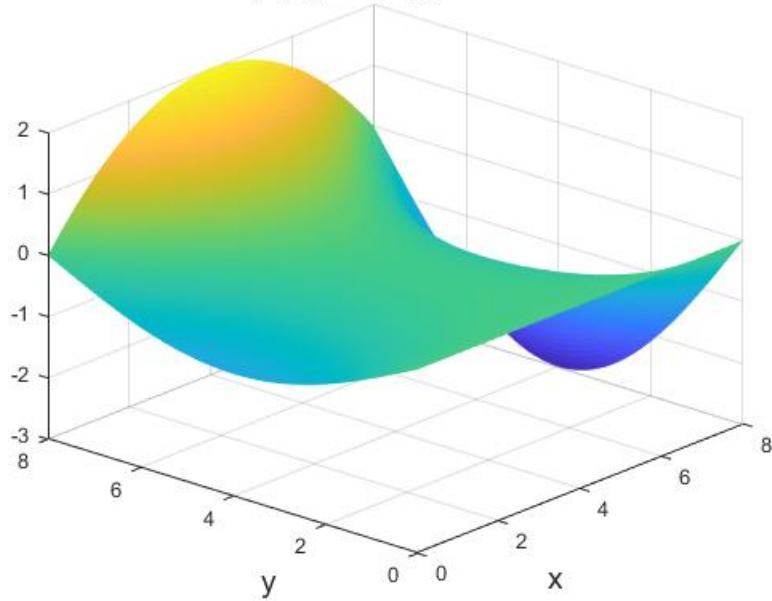
$$\begin{cases} u_t = u_{xx} + u_{yy}, & (x, y) \in (0, 8) \times (0, 8), t > 0 \\ u(x, y, 0) = f(x, y) \\ u(0, y, t) = g_L(y, t), \quad u(8, y, t) = g_R(y, t) \\ u(x, 0, t) = g_B(x, t), \quad u(x, 8, t) = g_T(x, t) \end{cases}$$

where

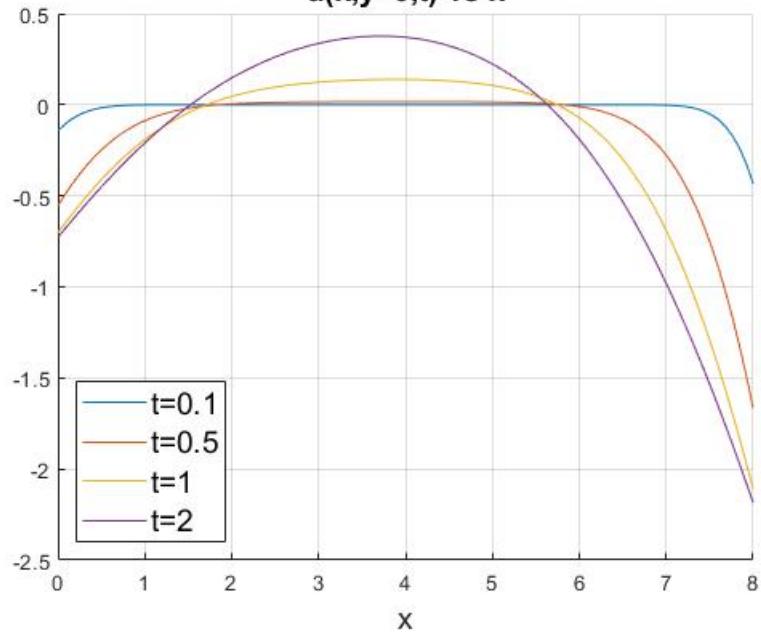
$$\begin{aligned} f(x, y) &= 0 \\ g_L(y, t) &= -\sin(\pi y/8)\tanh(2t), \quad g_R(y, t) = -3\sin(\pi y/8)\tanh(2t) \\ g_B(x, t) &= 0, \quad g_T(x, t) = x(1 - x/8)\tanh(2t). \end{aligned}$$

We solve to $T = 2$ using the 2D FTCS method, with $\Delta x = \Delta y = 0.08$ and $\Delta t = 1.25 \times 10^{-3}$. Below we plot $u(x, 6)$ vs x at $t = 0.1, 0.5, 0.12$ in one figure, and $u(x, y, t)$ vs (x, y) at $T = 2$ as a surface in another figure.

$u(x,y,t)$ vs (x,y) at $T = 2$



$u(x,y=6,t)$ vs x



Problem 6:

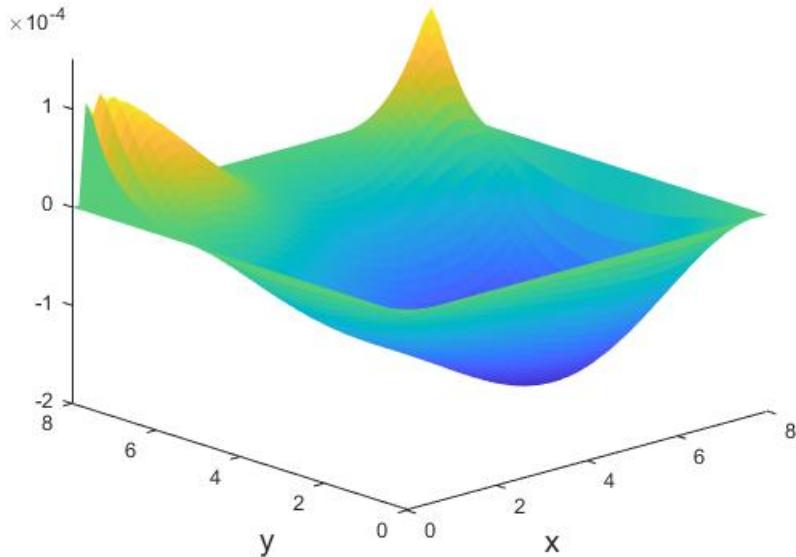
This problem is a continuation of the problem above. Using $\Delta x = \Delta y = 0.08$ and $\Delta t = 1.25 \times 10^{-3}$, we first solve the IBVP to $T = 20$. For each time level, we calculate the max value of $|u_t|$ at all of the internal points. We store the maximum at each t_n in the vector

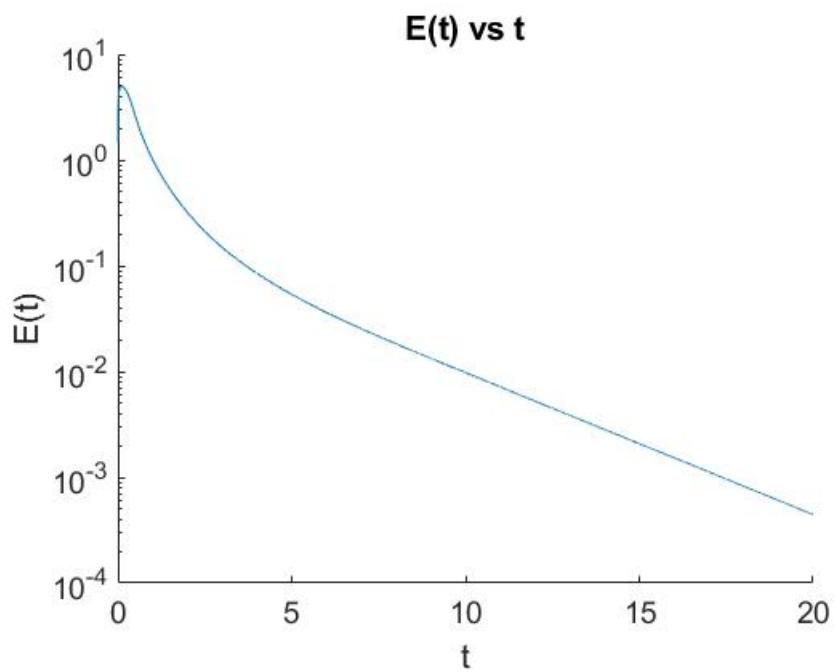
$$E(t_n) = \max \frac{|u_{i,j}^{n+1} - u_{i,j}^n|}{\Delta t}, \quad 1 \leq i \leq N_x - 1, \quad 1 \leq j \leq N_y - 1.$$

We then plot $E(t)$ against t , where $E(t)$ is on a logarithmic scale (t on a linear scale).

We also re-solve the IBVP using $\Delta x = \Delta y = 0.16$ and $\Delta t = 1.25 \times 10^{-3}$, using the new solution to calculate $u_{\{\Delta=0.16\}} - u_{\{\Delta=0.08\}}$ for $T = 20$, and then plot the result vs (x, y) as a surface.

$u_{\{\Delta=0.16\}} - u_{\{\Delta=0.08\}}$ at $T = 20$.





Problem 1 (Theoretical)

Consider the Lax-Friedrichs method for the general case ($a > 0$ or $a < 0$)

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

P1 Part 1: Carry out von Neumann stability analysis.

Part 2: Use Taylor expansion to find the local truncation error e_i^n .

Convert r back to $\Delta t/(\Delta x)$. Find coefficients of $(\Delta t)^2$, $(\Delta t)(\Delta x)$ and $(\Delta x)^2$ in e_i^n .

Part 3: Answer the two questions below.

For fixed $\frac{\Delta t}{\Delta x} = r$, write $\frac{e_i^n}{\Delta t}$ in terms of Δx only. Do we have $\lim_{\Delta x \rightarrow 0} \frac{e_i^n}{\Delta t} = 0$?

For fixed $\frac{\Delta t}{(\Delta x)^2} = c$, write $\frac{e_i^n}{\Delta t}$ in terms of Δx only. Do we have $\lim_{\Delta x \rightarrow 0} \frac{e_i^n}{\Delta t} = 0$?

Part 1: We have

$$u_i^{n+1} = \frac{u_{i+1}^n + u_{i-1}^n}{2} - \frac{ar}{2} (u_{i+1}^n - u_{i-1}^n), \quad r = \frac{\Delta t}{\Delta x} \quad (a < 0 \text{ or } a > 0)$$

First, write

$$L_{num}(u^{n+1}) = L_{num}(u^n)$$

⇒ Already in this form.

$$\text{We try } u_i^n = p^n e^{\sqrt{-\xi} i \Delta x}$$

$$\begin{aligned} \Rightarrow p^{n+1} e^{\sqrt{-\xi} i \Delta x} &= \frac{1}{2} \left[p^n e^{\sqrt{-\xi} (i+1) \Delta x} + p^n e^{\sqrt{-\xi} (i-1) \Delta x} \right] \\ &\quad - \frac{ar}{2} \left[p^n e^{\sqrt{-\xi} (i+1) \Delta x} - p^n e^{\sqrt{-\xi} (i-1) \Delta x} \right] \end{aligned}$$

Divide by $p^n e^{\sqrt{-\xi} i \Delta x}$:

$$p = \frac{1}{2} \left[e^{\sqrt{-\xi} \Delta x} + e^{-\sqrt{-\xi} \Delta x} \right] - \frac{ar}{2} \left[e^{\sqrt{-\xi} \Delta x} - e^{-\sqrt{-\xi} \Delta x} \right]$$

$$= \cosh(\sqrt{-1}\xi\Delta x) - \arcsinh(\sqrt{-1}\xi\Delta x)$$

$$= \cos(\xi\Delta x) - \arcsin(\xi\Delta x),$$

since $\cosh(i\alpha) = \cos(\alpha)$ and $\sinh(i\alpha) = i\sin(\alpha)$.

Considering $|\rho|^2$, we have

$$\begin{aligned} |\rho|^2 &= [\cos(\xi\Delta x) - i\arcsin(\xi\Delta x)] [\cos(\xi\Delta x) + i\arcsin(\xi\Delta x)] \\ &= \cos^2(\xi\Delta x) - i\arcsin(\xi\Delta x)\sin(\xi\Delta x) + i\arcsin(\xi\Delta x)\sin(\xi\Delta x) \\ &\quad + a^2 r^2 \sin^2(\xi\Delta x) \\ &= \cos^2(\xi\Delta x) + a^2 r^2 \sin^2(\xi\Delta x) \end{aligned}$$

\Rightarrow We need $|\rho| \leq 1 + C\Delta t$ for small Δt and all ξ .

$$|\rho|^2 = \cos^2(\xi\Delta x) + a^2 r^2 \sin^2(\xi\Delta x)$$

$$\Rightarrow |\rho| = \sqrt{\cos^2(\xi\Delta x) + (\ar)^2 \sin^2(\xi\Delta x)}$$

$$|\rho| = \sqrt{1 - \sin^2(\xi\Delta x) + (\ar)^2 \sin^2(\xi\Delta x)}$$

$$= \sqrt{1 + (a^2 r^2 - 1) \sin^2(\xi\Delta x)}$$

Case 1:

Let $\ar > 1$. Then, $a^2 r^2 > 1$.

Set $K = a^2 r^2 - 1 > 0$, giving us

$$|\rho| = \sqrt{1 + k \sin^2(\xi \Delta x)}' > 1.$$

Case 2:

Now, let $|ar| < -1$, $\Rightarrow a^2 r^2 > 1$ again,
giving $|\rho| > 1$.

Case 3:

Hence, let $|ar| \leq 1$, $\Rightarrow a^2 r^2 \leq 1$

$$\Rightarrow k := a^2 r^2 - 1 \leq 0$$

$$\Rightarrow |\rho| = \sqrt{1 + k \sin^2(\xi \Delta x)} \leq 1.$$

Thus, $|\rho| \leq 1$ requires $|ar| \leq 1$.

Therefore, the Lax-Friedrichs method is stable
for $|ar| \leq 1$.

Part 2: Ack:

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

We wts $e_i^n(\Delta x, \Delta t) = \Delta t \cdot O(\Delta t + \Delta x)$.

We know

$$e_i^n(\Delta x, \Delta t) = u(x_i, t_{n+1}) - L_{num}(u(x_i, t_n))$$

$$\begin{aligned} \Rightarrow e_i^n(\Delta x, \Delta t) &= u(x_i, t_{n+1}) - \frac{1}{2} [u(x_{i+1}, t_n) + u(x_{i-1}, t_n)] \\ &\quad + \frac{ar}{2} [u(x_{i+1}, t_n) - u(x_{i-1}, t_n)] \end{aligned}$$

$$+ \frac{\alpha r}{2} [u(x_{i+1}, t_n) - u(x_{i-1}, t_n)]$$

$$\Rightarrow e_i''(\Delta x, \Delta t) = \underbrace{u(x_i, t_n + \Delta t)}_{①} - \frac{1}{2} \left[\underbrace{u(x_i + \Delta x, t_n) + u(x_i - \Delta x, t_n)}_{②} \right]$$

$$+ \frac{\alpha r}{2} \left[\underbrace{u(x_i + \Delta x, t_n)}_{③} - u(x_i - \Delta x, t_n) \right]$$

Expanding ①, ②, ③ around (x_i, t_n) : Let U denote $u|_{(x_i, t_n)}$.

$$①: u(x_i, t_n + \Delta t) = U + U_t \Delta t + \frac{1}{2} U_{tt} \Delta t^2 + \mathcal{O}(\Delta t^3)$$

$$②: u(x_i + \Delta x, t_n) + u(x_i - \Delta x, t_n)$$

$$= U + U_x \cancel{\Delta x} + \frac{1}{2} U_{xx} \Delta x^2 + \mathcal{O}(\Delta x^3) + U - U_x \cancel{\Delta x} + \frac{1}{2} U_{xx} \Delta x + \mathcal{O}(\Delta x^4)$$

$$= 2U + U_{xx} \Delta x^2 + \mathcal{O}(\Delta x^4)$$

↑
odd powers of Δx cancel.

$$③: u(x_i + \Delta x, t_n) - u(x_i - \Delta x, t_n)$$

$$= U + U_x \Delta x + \frac{1}{2} U_{xx} \cancel{\Delta x^2} + \mathcal{O}(\Delta x^3) - \left[U - U_x \Delta x + \frac{1}{2} U_{xx} \cancel{\Delta x^2} + \mathcal{O}(\Delta x^3) \right]$$

$$= 2U_x \Delta x + \mathcal{O}(\Delta x^3)$$

Plugging in these expansions...

$$e_i''(\Delta x, \Delta t) = \underbrace{u(x_i, t_n + \Delta t)}_{①} - \frac{1}{2} \left[\underbrace{u(x_i + \Delta x, t_n) + u(x_i - \Delta x, t_n)}_{②} \right]$$

$$+ \frac{\alpha r}{2} \left[\underbrace{u(x_i + \Delta x, t_n)}_{③} - u(x_i - \Delta x, t_n) \right]$$

$$\Rightarrow e_i''(\Delta x, \Delta t) = U + U_t \Delta t + \frac{1}{2} U_{tt} \Delta t^2 + \mathcal{O}(\Delta t^3)$$

$$- \frac{1}{2} \left[2U + U_{xx} \Delta x^2 + \mathcal{O}(\Delta x^4) \right] + \frac{\alpha}{\pi} \cdot \frac{\Delta t}{\Delta x} \left[2U_x \Delta x + \mathcal{O}(\Delta x^3) \right]$$

$$= (U_t + \alpha U_x) \Delta t + \frac{1}{2} U_{tt} \Delta t^2 - \frac{1}{2} U_{xx} \Delta x^2$$

$$= (U_t + aU_x)\Delta t + \frac{1}{2}U_{tt}\Delta t^2 - \frac{1}{2}U_{xx}\Delta x^2 \\ + O(\Delta t^3) + O(\Delta x^4) + O(\Delta x^2)\Delta t$$

From the model equation for hyperbolic PDEs:

$$U_t + aU_x = 0$$

$$\Rightarrow e_i^n(\Delta x, \Delta t) = \frac{1}{2}[U_{tt}\Delta t^2 - U_{xx}\Delta x^2] + \Delta t O(\Delta t^2 + \Delta x^2) + O(\Delta x^4).$$

Part 3: We have that

$$e_i^n = \frac{1}{2}U_{tt}\Delta t^2 - \frac{1}{2}U_{xx}\Delta x^2 + \Delta t O(\Delta t^2 + \Delta x^2) + O(\Delta x^4)$$

$$\Rightarrow \frac{e_i^n}{\Delta t} = \frac{1}{2}U_{tt}\Delta t - \frac{1}{2}U_{xx}\frac{\Delta x^2}{\Delta t} + O(\Delta t^2 + \Delta x^2) + \frac{1}{\Delta t}O(\Delta x^4)$$

For fixed $r = \frac{\Delta t}{\Delta x}$:

$$r = \frac{\Delta t}{\Delta x} : \Rightarrow \Delta t = \Delta x \cdot r \Rightarrow \Delta t^2 = \Delta x^2 r^2. \quad \text{Thus, using } U_{tt} = a^2 U_{xx}$$

$$\frac{e_i^n}{\Delta t} = \frac{a^2 r^2}{2} U_{xx} \Delta x - \frac{1}{2} U_{xx} \Delta x r^{-1} + O(r^2 \Delta x^2 + \Delta x^2) + \frac{1}{r} O(\Delta x^3)$$

Hence,

$$\lim_{\Delta x \rightarrow 0} \frac{e_i^n}{\Delta t} = 0.$$

For fixed $c = \frac{\Delta t}{\Delta x^2}$, we have:

$$c = \frac{\Delta t}{\Delta x^2} \Rightarrow \Delta t = \Delta x^2 c \Rightarrow \Delta t^2 = c^2 \Delta x^4. \quad \text{Thus,}$$

$$\frac{e_i^n}{\Delta t} = \frac{1}{2} u_{tt} \Delta t - \frac{1}{2} u_{xx} \frac{\Delta x^2}{\Delta t} + O(\Delta t^2 + \Delta x^2) + \frac{1}{\Delta t} O(\Delta x^4)$$

$$\Rightarrow \frac{e_i^n}{\Delta t} = \frac{a^2 c}{2} u_{xx} \Delta x^2 - \frac{1}{2c} u_{xx} + O(c^2 \Delta x^4) + \frac{1}{c} O(\Delta x^2)$$

We can see that

$$\lim_{\Delta x \rightarrow 0} \frac{e_i^n}{\Delta t} = -\frac{1}{2c} \cdot u_{xx} \neq 0.$$

Problem 2 (Theoretical)

Carry out von Neumann stability analysis on each of the methods below

- i) the BTCS method for the general case ($a > 0$ or $a < 0$)

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

- ii) the implicit upwind method for the case of $a > 0$

$$u_i^{n+1} = u_i^n - ar (u_i^{n+1} - u_{i-1}^{n+1}), \quad r = \frac{\Delta t}{\Delta x}$$

P2 Hint: Examine $|1/\rho|^2$.

i) We need

$$L_{num}(u^{n+1}) = L_{num}(u^n)$$

$$u^{n+1} \quad \dots \quad u^n / \dots \quad \overset{n+1}{\dots} \quad \overset{n+1}{\dots} \quad | \quad \dots \quad \overset{n}{\dots}$$

$$\Rightarrow u_i^{n+1} + \frac{ar}{2} (u_{i+1}^{n+1} - u_{i-1}^{n+1}) = u_i^n$$

We try $u_i^n = p^n e^{\sqrt{1-\epsilon} i \Delta x}$:

$$p^{n+1} e^{-\sqrt{1-\epsilon} i \Delta x} + \frac{ar}{2} \left[p^{n+1} e^{\sqrt{1-\epsilon} (i+1) \Delta x} - p^{n+1} e^{\sqrt{1-\epsilon} (i-1) \Delta x} \right] = p^n e^{\sqrt{1-\epsilon} i \Delta x}$$

Dividing by $p^n e^{\sqrt{1-\epsilon} i \Delta x}$:

$$p + \frac{ar}{2} \left[p e^{\sqrt{1-\epsilon} \Delta x} - p e^{-\sqrt{1-\epsilon} \Delta x} \right] = 1$$

$$\Rightarrow p \left[1 + \operatorname{arsinh}(\sqrt{1-\epsilon} \Delta x) \right] = 1$$

Using $\sinh(i\alpha) = i \sin(\alpha)$:

$$p \left[1 + i \operatorname{arsin}(\epsilon \Delta x) \right] = 1$$

Hence,

$$p = \frac{1}{1 + i \operatorname{arsin}(\epsilon \Delta x)} \quad |z| = \sqrt{a^2 + b^2}$$

$$\Rightarrow |p| = \frac{1}{\sqrt{1 + a^2 r^2 \sin^2(\epsilon \Delta x)}}$$

Notice that $\sqrt{1 + a^2 r^2 \sin^2(\epsilon \Delta x)} > 0$.

Hence $|p| \leq 1$ regardless of ar .

Hence, $|r| \leq 1$ regardless of ar.
 \therefore BTCS is unconditionally stable.

ii) We need

$$L_{\text{num}}(u^{n+1}) = L_{\text{num}}(u^n),$$

$$\Rightarrow u_i^{n+1} + ar(u_i^{n+1} - u_{i-1}^{n+1}) = u_i^n.$$

We try $u_i^n = p^n e^{\sqrt{-1}\xi i \Delta x}$:

$$p^{n+1} e^{\sqrt{-1}\xi i \Delta x} + ar \left[p^{n+1} e^{\sqrt{-1}\xi i \Delta x} - p^{n+1} e^{\sqrt{-1}\xi (i-1) \Delta x} \right] = p^n e^{\sqrt{-1}\xi i \Delta x}$$

Dividing by $p^n e^{\sqrt{-1}\xi i \Delta x}$:

$$p + ar \left[p - p e^{-\sqrt{-1}\xi \Delta x} \right] = 1$$

$$\Rightarrow p \left[1 + ar(1 - e^{-\sqrt{-1}\xi \Delta x}) \right] = 1$$

$$\text{We have } e^{-\sqrt{-1}\xi \Delta x} = \cos(\xi \Delta x) - i \sin(\xi \Delta x)$$

$$\Rightarrow \rho = \frac{1}{1 + ar[1 - \cos(\xi \Delta x) + i \sin(\xi \Delta x)]}$$

$$\Rightarrow \rho = \frac{1}{1 + ar - \arccos(\xi \Delta x) + i \arcsin(\xi \Delta x)}$$

Hence,

$$|\rho| = \sqrt{\frac{1}{[1 + ar - \arccos(\xi \Delta x)]^2 + \arcsin^2(\xi \Delta x)}}$$

Working in the denominator:

$$\begin{aligned} (1 + ar - \arccos(\xi \Delta x))^2 &= 1 + ar - \arccos(\xi \Delta x) + ar + a^2 r^2 \\ &\quad - a^2 r^2 \cos(\xi \Delta x) - \arccos(\xi \Delta x) \\ &\quad - a^2 r^2 \cos(\xi \Delta x) + a^2 r^2 \cos^2(\xi \Delta x) \\ &= \underbrace{(1 + ar)^2}_{1 + 2ar + a^2 r^2} - 2\arccos(\xi \Delta x) - 2a^2 r^2 \cos(\xi \Delta x) + a^2 r^2 \cos^2(\xi \Delta x) \end{aligned}$$

Hence, considering the hint, we have

$$\begin{aligned} \frac{1}{|\rho|^2} &= \frac{1 + 2ar + a^2 r^2 - 2\arccos(\xi \Delta x) - 2a^2 r^2 \cos(\xi \Delta x) + a^2 r^2 \cos^2(\xi \Delta x)}{1 + 2ar + a^2 r^2 - 2\arccos(\xi \Delta x) - 2a^2 r^2 \cos(\xi \Delta x) + a^2 r^2 \sin^2(\xi \Delta x)} \\ &= 1 + 2ar + 2a^2 r^2 - 2ar \cos(\xi \Delta x) - 2a^2 r^2 \cos(\xi \Delta x) \\ &= 1 + 2ar(1 + ar - \cos(\xi \Delta x) - \arccos(\xi \Delta x)) \end{aligned}$$

$$\begin{aligned}
 &= 1 + 2ar \left[1 + ar - (1 + ar) \cos(\frac{r}{\Delta x}) \right] \\
 &= 1 + 2ar \left[(1 + ar)(1 - \cos(\frac{r}{\Delta x})) \right] \\
 &= 1 + \left[2(ar + a^2 r^2)(1 - \cos(\frac{r}{\Delta x})) \right]
 \end{aligned}$$

In Upwind method, $a > 0$.

$$\Rightarrow \frac{1}{|r|} \geq 1, \text{ hence } |r| \leq 1.$$

\therefore the implicit upwind method is unconditionally stable.