

equations we are interested in some very basic equations namely

$$u_t + a(uv)_x = 0$$

$$v_t + b(uv)_x = 0$$

For stability we are interested in the difference between the exact solution and the numerical solution produced by a finite precision algorithm. We are going to investigate this using Von Nuemann stability which is best for linear equations so we linearise.

1 Stability of Methods for Linearised Equations

Linearising means we assume the form of u and v so that

$$u = u_0 + u_1(x, t) + \dots, v = v_0 + v_1(x, t) + \dots$$

where the $u_i \gg u_{i+1}$ we then neglect all terms far smaller than u_1 so that

$$(u)_t + aU(v)_x + aV(u)_x = 0$$

$$(v)_t + bU(v)_x + bV(u)_x = 0$$

2 Silly FDM

Approximating all derivatives by centered FD methods gives

$$u_j^{n+1} = u_j^{n-1} - a \frac{\Delta t}{\Delta x} (U [v_{j+1}^n - v_{j-1}^n] + V [u_{j+1}^n - u_{j-1}^n])$$

$$v_j^{n+1} = v_j^{n-1} - b \frac{\Delta t}{\Delta x} (U [v_{j+1}^n - v_{j-1}^n] + V [u_{j+1}^n - u_{j-1}^n])$$

We assume $u_j^n = u(t)e^{ikx_j}$, $v_j^n = v(t)e^{ilx_j}$ Thus

$$u_j^{n+1} = u_j^{n-1} - a \frac{\Delta t}{\Delta x} (U [e^{il\Delta x} - e^{-il\Delta x}] v_j^n + V [e^{ik\Delta x} - e^{-ik\Delta x}] u_j^n)$$

using trig relations

$$\begin{aligned}
u_j^{n+1} &= u_j^{n-1} - a \frac{\Delta t}{\Delta x} (U 2i \sin(l\Delta x) v_j^n + V 2i \sin(k\Delta x) u_j^n) \\
u_j^{n+1} &= u_j^{n-1} - \frac{2ia\Delta t}{\Delta x} (U \sin(l\Delta x) v_j^n + V \sin(k\Delta x) u_j^n) \\
v_j^{n+1} &= v_j^{n-1} - \frac{2ib\Delta t}{\Delta x} (U \sin(l\Delta x) v_j^n + V \sin(k\Delta x) u_j^n)
\end{aligned}$$

We can write this in matrix form as

$$\begin{bmatrix} u_j^{n+1} \\ v_j^{n+1} \\ u_j^n \\ v_j^n \end{bmatrix} = \begin{bmatrix} -\frac{2ia\Delta t}{\Delta x} V \sin(l\Delta x) & -\frac{2ia\Delta t}{\Delta x} U \sin(k\Delta x) & 1 & 0 \\ -\frac{2ib\Delta t}{\Delta x} V \sin(l\Delta x) & -\frac{2ib\Delta t}{\Delta x} U \sin(k\Delta x) & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} u_j^n \\ v_j^n \\ u_j^{n-1} \\ v_j^{n-1} \end{bmatrix}$$

The easy situation is when u and v are the same and $a = b$ in which case

$$\begin{bmatrix} u_j^{n+1} \\ v_j^{n+1} \\ u_j^n \\ v_j^n \end{bmatrix} = \begin{bmatrix} -\frac{2ia\Delta t}{\Delta x} U \sin(k\Delta x) & -\frac{2ia\Delta t}{\Delta x} U \sin(k\Delta x) & 1 & 0 \\ -\frac{2ia\Delta t}{\Delta x} U \sin(k\Delta x) & -\frac{2ia\Delta t}{\Delta x} U \sin(k\Delta x) & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} u_j^n \\ v_j^n \\ u_j^{n-1} \\ v_j^{n-1} \end{bmatrix}$$

Again we are interested in the 2 norm which in this case since the matrix is symmetric is the largest eigenvalue. We also have the eigenvalues for a matrix of the form

$$\begin{bmatrix} a & a & 1 & 0 \\ a & a & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

are $\lambda = \pm 1, a \pm \sqrt{a^2 + 1}$

But since the largest of these must be greater than or equal to 1 we are basically done, and this scheme is unconditionally unstable. We should however check that we don't have equality

if

$$a \pm \sqrt{a^2 + 1} = 1$$

then $a = 0$ which implies $\frac{2ia\Delta t}{\Delta x} U \sin(k\Delta x)$ which is not possible by choosing $\Delta t, \Delta x$ for general U and a