1 Von Neumann analysis

We will analyse the advection equation to determine the required step-sizes for our solver. The advection equation is

$$\frac{\partial u(t,x)}{\partial t} = -a \frac{\partial u(t,x)}{\partial x},\tag{1}$$

where u(t,x) is the analytical solution to the equation and a is a constant. This is discretized over a grid such that $y(t_n,x_j)=y_{n,j}$, where $t_n=n\Delta t$ and $x_j=j\Delta x$ for $n,j\in\mathbb{N}$, is our numerical solution. Then the advection equation becomes

$$\left[\frac{\partial y}{\partial t}\right]_{n,j} = -a \left[\frac{\partial y}{\partial x}\right]_{n,j}.$$
 (2)

The numerical solution

$$y_{n,j} = u(t_n, x_j) + \epsilon_{n,j},\tag{3}$$

where $\epsilon_{n,j}$ is the round-off error. The round-off error must also satisfy the discretized equation and this gives us that

$$\left[\frac{\partial \epsilon}{\partial t}\right]_{n,j} = -a \left[\frac{\partial \epsilon}{\partial x}\right]_{n,j}.$$
 (4)

We expand the round-off error as a fourier series

$$\epsilon(t_n, x_j) = \sum_m E_m(t_n) e^{ik_m j\Delta x},\tag{5}$$

where k_m is the wavenumber and $E(t_n)$ is the time-dependent amplitude of the error. When inserting this into our differential equation we get a linear difference equation, meaning that each of the terms behave like the entire series so we can consider the growth of only one term.

$$\epsilon_m(t_n, x_j) = E_m(t_n)e^{ik_m j\Delta x}. (6)$$

We will show the calculations using the first-order upwind scheme with the second-order Runge-Kutta scheme. Since this should be true for any m we remove the subscipt, define $\beta \equiv k\Delta x$ and get that the spacial derivative is

$$\begin{split} \left[\frac{\partial \epsilon}{\partial x}\right]_{n,j} &= \frac{\epsilon_{n,j} - \epsilon_{n,j-1}}{\Delta x} \\ &= \frac{E(t_n)e^{i\beta j} - E(t_n)e^{i\beta (j-1)}}{\Delta x} \\ &= E(t_n)e^{i\beta j}\frac{1 - e^{-i\beta}}{\Delta x}. \end{split}$$

This gives us that the advection equation 4 becomes

$$\left[\frac{\partial \epsilon}{\partial t}\right]_{n,j} = e^{i\beta j} \left[\frac{\partial E(t_n)}{\partial t}\right]_{n,j} = -aE(t_n)e^{i\beta j} \frac{1 - e^{-i\beta}}{\Delta x} \tag{7}$$

$$\left[\frac{\partial E(t_n)}{\partial t}\right]_{n,j} = -aE(t_n)\frac{1 - e^{-i\beta}}{\Delta x}.$$
 (8)

We define $\lambda = -\frac{a}{\Delta x} \left(1 - e^{-i\beta}\right)$ which gives us

$$\mu = \Delta t \lambda = -C \left(1 - e^{-i\beta} \right), \tag{9}$$

where $C \equiv a\Delta t/\Delta x$ is the Courant number. This means that the differential equation for the time-dependent error is

$$\left[\frac{\partial E(t_n)}{\partial t}\right]_{n,j} = \lambda E(t_n). \tag{10}$$

Using this with the second order Runge-Kutta scheme, the slopes for the time-dependent error is

$$k_1 = \lambda E_n,$$

 $k_2 = \lambda \left(E_n + \frac{\Delta t}{2} k_1 \right) = E_n \left(\lambda + \frac{\Delta t \lambda^2}{2} \right).$

And the next time-step for the error is

$$E_{n+1} = E_n + \Delta t \left(\frac{k_1}{2} + \frac{k_2}{2} \right)$$
$$= E_n \left(1 + \Delta t \lambda + \frac{1}{2} (\Delta t \lambda)^2 \right)$$
$$= E_n \left(1 + \mu + \frac{1}{2} \mu^2 \right).$$

This gives us the amplification factor

$$g = \frac{E_{n+1}}{E_n} = \left(1 + \mu + \frac{1}{2}\mu^2\right). \tag{11}$$

We require $|g| \leq 1$, meaning that the time-dependent error does not grow in time. If this is any bigger than 1 the error will grow exponentially, meaning our numerical solution will not be stable.

- Table for all schemes used
- Images of analysis from jupyter file