

# hw6\_MK

May 12, 2022

0.0.1 1.

$$U = [U_0, U_1, \dots, U_m]^T$$

(a)

$$D_- = \frac{1}{h} \begin{pmatrix} 1 & & & & -1 \\ -1 & 1 & & & \\ & -1 & 1 & & \\ & & -1 & 1 & \\ & & & -1 & 1 \end{pmatrix}$$

$$D_-^2 = \frac{1}{h^2} \begin{pmatrix} 1 \cdot 1 & & & & (-1) \cdot (-1) & & (-1) \cdot 1 + 1 \cdot (-1) \\ (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 & & & & & (-1) \cdot (-1) \\ & (-1) \cdot 1 + 1 \cdot (-1) & (-1) \cdot (-1) & & & & \\ & & (-1) \cdot (-1) & 1 \cdot 1 & & & \\ & & & (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 & & \\ & & & & (-1) \cdot (-1) & (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 \end{pmatrix}$$

$$D_-^2 = \frac{1}{h^2} \begin{pmatrix} 1 & & & 1 & -2 \\ -2 & 1 & & & 1 \\ 1 & -2 & 1 & & \\ & 1 & -2 & 1 & \\ & & 1 & -2 & 1 \end{pmatrix}$$

Hence the Taylor series method gives:

$$U_j^{n+1} = U_j^n - \frac{ak}{h}(U_j^n - U_{j-1}^n) + \frac{(ak)^2}{2h^2}(U_j^n - 2U_{j-1}^n + U_{j-2}^n)$$

where the index  $j$  runs from 0 to  $m$  with addition of indices performed mod  $m + 1$  to incorporate the periodic boundary conditions.

(b) Beam-Warming method:

$$U_j^{n+1} = U_j^n - \frac{ak}{2h}(3U_j^n - 4U_{j-1}^n + U_{j-2}^n) + \frac{(ak)^2}{2h^2}(U_j^n - 2U_{j-1}^n + U_{j-2}^n)$$

Subtracting the method from part (a) we get:

$$\tau^n = -\frac{ak}{2h}U_j^n + \frac{ak}{h}U_{j-1}^n - \frac{ak}{2h}U_{j-2}^n$$

$$\tau^n = -\frac{ak}{2h}(U_j^n - 2U_{j-1}^n + U_{j-2}^n)$$

Hence it is **first order accurate** compared to the Beam-Warming method.

(c)

$$D_+ = \frac{1}{h} \begin{pmatrix} -1 & 1 & & & \\ & -1 & 1 & & \\ & & -1 & 1 & \\ & & & -1 & 1 \\ 1 & & & & -1 \end{pmatrix}$$

$$D_+ + D_- = \frac{1}{h} \begin{pmatrix} & 1 & & & -1 \\ -1 & & 1 & & \\ & -1 & & 1 & \\ & & -1 & & 1 \\ 1 & & & -1 & \end{pmatrix}$$

$$D_+ D_- = \frac{1}{h^2} \begin{pmatrix} (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 & & & (-1) \cdot (-1) \\ (-1) \cdot (-1) & (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 & & \\ & (-1) \cdot (-1) & (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 & \\ & & (-1) \cdot (-1) & (-1) \cdot 1 + 1 \cdot (-1) & 1 \cdot 1 \\ 1 \cdot 1 & & & (-1) \cdot (-1) & (-1) \cdot 1 + 1 \cdot (-1) \end{pmatrix}$$

$$D_+ D_- = \frac{1}{h^2} \begin{pmatrix} -2 & 1 & & & 1 \\ 1 & -2 & 1 & & \\ & 1 & -2 & 1 & \\ & & 1 & -2 & 1 \\ 1 & & & 1 & -2 \end{pmatrix}$$

Hence the method gives:

$$U_j^{n+1} = U_j^n - \frac{ak}{2h}(U_{j+1}^n - U_{j-1}^n) + \frac{(ak)^2}{2h^2}(U_{j+1}^n - 2U_j^n + U_{j-1}^n)$$

which is exactly the standard **Lax-Wendroff method**.

0.0.2 2.

(a)

$$U_j^{n+1} = U_j^n - \frac{ak}{2h}(U_j^n - U_{j-1}^n + U_j^{n+1} - U_{j-1}^{n+1})$$

The trapezoidal method form:

$$\frac{U_j^{n+1} - U_j^n}{k} = \frac{1}{2}((-a) \frac{U_j^n - U_{j-1}^n}{h} + (-a) \frac{U_j^{n+1} - U_{j-1}^{n+1}}{h})$$

$$\frac{U^{n+1} - U^n}{k} = \frac{1}{2}(AU^n + AU^{n+1})$$

Hence in  $U'(t) = AU(t)$ :

$$A = -\frac{a}{h} \begin{pmatrix} 1 & & & & -1 \\ -1 & 1 & & & \\ & -1 & 1 & & \\ & & -1 & 1 & \\ & & & -1 & 1 \end{pmatrix}$$

Since the index  $j$  runs from 1 to  $m + 1$  with addition of indices performed mod  $m + 1$  to incorporate the periodic boundary conditions  $A$  considers boundary conditions too. (This is illustrated for a grid with  $m + 1 = 5$  unknowns and  $h = 1/5$ .)

(b) Notice that  $A = A_- + A_+$  where:

$$A_- = -\frac{a}{2h} \begin{pmatrix} & 1 & & & -1 \\ -1 & & 1 & & \\ & -1 & & 1 & \\ & & -1 & & 1 \\ 1 & & & -1 & \end{pmatrix}$$

$$A_+ = \frac{a}{2h} \begin{pmatrix} -2 & 1 & & & 1 \\ 1 & -2 & 1 & & \\ & 1 & -2 & 1 & \\ & & 1 & -2 & 1 \\ 1 & & & 1 & -2 \end{pmatrix}$$

It is of the form shown in (10.15) of the textbook with  $\epsilon = (ah)/2$ . Hence according to the textbook, eigenvalues of  $A$  will be of the form:

$$\mu_p = -\frac{ia}{h} \sin(2\pi ph) - \frac{2\epsilon}{h^2} (1 - \cos(2\pi ph))$$

$$\mu_p = -\frac{ia}{h} \sin(2\pi ph) - \frac{ah}{h^2} (1 - \cos(2\pi ph))$$

$$\mu_p = -\frac{ia}{h} \sin(2\pi ph) - \frac{a}{h} (1 - \cos(2\pi ph))$$

For the trapezoidal rule:

$$\operatorname{Re}(z) \leq 0$$

where  $z = k\mu_p$  and hence:

$$\frac{ak}{h} (\cos(2\pi ph) - 1) \leq 0$$

Since  $\cos(2\pi ph) \leq 1$  for all  $p$  and  $h$ , we have  $\cos(2\pi ph) - 1 \leq 0$ . So the method is stable only when:

$$\frac{ak}{h} \geq 0$$

So since  $k, h > 0$  when  $a > 0$ , the method is **stable** for all Courant numbers, and when  $a < 0$ , it is **not stable** for all Courant numbers.

(c) Replacing  $U_j^n$  and  $ak/h$  by  $g(\xi)^n e^{i\xi jh}$  and  $\nu$  in (6):

$$g(\xi)^{n+1} e^{i\xi jh} = g(\xi)^n e^{i\xi jh} - \frac{\nu}{2} (g(\xi)^n e^{i\xi jh} - g(\xi)^n e^{i\xi(j-1)h} + g(\xi)^{n+1} e^{i\xi jh} - g(\xi)^{n+1} e^{i\xi(j-1)h})$$

$$g(\xi) = 1 - \frac{\nu}{2} (1 - e^{-i\xi h} + g(\xi) - g(\xi) e^{-i\xi h})$$

$$g(\xi) = \frac{1 + \frac{\nu}{2}(e^{-i\xi h} - 1)}{1 - \frac{\nu}{2}(e^{-i\xi h} - 1)}$$

$|g(\xi)| \leq 1$  only when  $\operatorname{Re}(\nu(e^{-i\xi h} - 1)) \leq 0$ . Hence:

$$\nu(\cos(\xi h) - 1) \leq 0$$

Since  $\cos(\xi h) - 1 \leq 0$  for all  $\xi$  and  $h$ , the method is stable only if:

$$\nu \geq 0$$

(d) An implicit method such as (6) satisfies the CFL condition for any time step  $k$ . In this case the numerical domain of dependence is the positive half of the entire real line because the lower two-diagonal system couples together half of all points in such a manner that the solution at each point depends on the data at half of all points (i.e., half of the inverse of a lower two-diagonal is dense). Hence:

$$\frac{ak}{h} \geq 0$$

(e) We have a system of the form  $U'(t) = AU(t) + g(t)$ , where:

$$A = -\frac{a}{h} \begin{pmatrix} 1 & & & & \\ -1 & 1 & & & \\ & -1 & 1 & & \\ & & -1 & 1 & \\ & & & -1 & 1 \end{pmatrix}$$

$$g(t) = \begin{pmatrix} g_0(t)a/h \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

According to the textbook,  $A$  is a defective Jordan block with all its eigenvalues at the point  $-a/h$ .  $A$  is highly nonnormal. It is essentially a Jordan block of the sort discussed in Section D.5.1 of the textbook, and on a fine grid its  $\epsilon$ -pseudospectra roughly fill up the circle of radius  $a/h$  about  $-a/h$ , even for very small  $\epsilon$ . This is a case where we need to apply a more stringent requirement than simply requiring that  $k\lambda$  be inside the stability region for all eigenvalues; we also need to require that

$$\operatorname{dist}(k\lambda_\epsilon, S) \leq C\epsilon$$

where  $S$  is the stability region for Trapezoidal method:

$$S = \{z : \text{Re}(z) \leq 0\}$$

Hence:

$$\frac{ak}{h} \geq 0$$

For a CFL condition using the same reasoning as in the previous part on  $A_\epsilon$ , since half of the  $\epsilon$ -pseudoinverse of a lower two-diagonal is dense we have:

$$\frac{ak}{h} \geq 0$$

### 0.0.3 3.

Beam-Warming method:

$$U_j^{n+1} = U_j^n - \frac{ak}{2h}(3U_j^n - 4U_{j-1}^n + U_{j-2}^n) + \frac{(ak)^2}{2h^2}(U_j^n - 2U_{j-1}^n + U_{j-2}^n)$$

Truncation error:

$$\tau^n = \frac{U_j^{n+1} - U_j^n}{k} + \frac{a}{2h}(3U_j^n - 4U_{j-1}^n + U_{j-2}^n) - \frac{a^2k}{2h^2}(U_j^n - 2U_{j-1}^n + U_{j-2}^n)$$

From a second order accurate Taylor series expansion in time:

$$u(x, t+k) = u(x, t) + ku_t(x, t) + \frac{1}{2}k^2u_{tt}(x, t)$$

Using the advection equation replace  $u_t$  by  $-au_x$  and  $u_{tt}$  by  $a^2u_{xx}$ :

$$u(x, t+k) = u(x, t) - kau_x(x, t) + \frac{1}{2}k^2a^2u_{xx}(x, t)$$

From a second order accurate Taylor series expansion in space:

$$u(x-h, t) = u(x, t) - hu_x(x, t) + \frac{1}{2}h^2u_{xx}(x, t)$$

$$u(x-2h, t) = u(x, t) - 2hu_x(x, t) + 2h^2u_{xx}(x, t)$$

Hence:

$$3u(x, t) - 4u(x-h, t) + u(x-2h, t) = 3u(x, t) - 4u(x, t) + 4hu_x(x, t) - 2h^2u_{xx}(x, t) + u(x, t) - 2hu_x(x, t) + 2h^2u_{xx}(x, t)$$

$$\frac{1}{2h}(3u(x, t) - 4u(x-h, t) + u(x-2h, t)) = u_x(x, t)$$

Therefore from Taylor series we will get:

$$u(x, t+k) = u(x, t) - \frac{ak}{2h}(3u(x, t) - 4u(x-h, t) + u(x-2h, t)) + \frac{1}{2}k^2a^2u_{xx}(x, t)$$

Relabelling and rearranging terms and using second order centered difference for  $u_{xx}$  we will get:

$$\frac{U_j^{n+1} - U_j^n}{k} = -\frac{a}{2h}(3U_j^n - 4U_{j-1}^n + U_{j-2}^n) + \frac{a^2k}{2h^2}(U_{j+1}^n - 2U_j^n + U_{j-1}^n)$$

Replacing this in the truncation error we get:

$$\begin{aligned}\tau^n &= -\frac{a}{2h}(3U_j^n - 4U_{j-1}^n + U_{j-2}^n) + \frac{a^2k}{2h^2}(U_{j+1}^n - 2U_j^n + U_{j-1}^n) + \\ &\quad + \frac{a}{2h}(3U_j^n - 4U_{j-1}^n + U_{j-2}^n) - \frac{a^2k}{2h^2}(U_j^n - 2U_{j-1}^n + U_{j-2}^n) \\ \tau^n &= \frac{a^2k}{2h^2}(U_{j+1}^n - 3U_j^n + 3U_{j-1}^n + U_{j-2}^n)\end{aligned}$$

Since the last term is just an average of two second order centered difference approximations in space and since initially Taylor series was a second order in time, we can conclude that the method is of a **second order**.

Inserting the formula  $v(x, t)$  into the difference equation:

$$v(x, t+k) = v(x, t) - \frac{ak}{2h}(3v(x, t) - 4v(x-h, t) + v(x-2h, t)) + \frac{a^2k^2}{2h^2}(v(x, t) - 2v(x-h, t) + v(x-2h, t))$$

Form above:

$$\frac{1}{2h}(3v(x, t) - 4v(x-h, t) + v(x-2h, t)) = v_x - \frac{1}{3}h^2v_{xxx}$$

Expanding these terms in Taylor series about  $(x, t)$  and simplifying gives:

$$\begin{aligned}v_t + \frac{1}{2}kv_{tt} + \frac{1}{6}k^2v_{ttt} + a(v_x - \frac{1}{3}h^2v_{xxx}) - \frac{1}{2}a^2k(v_{xx} - hv_{xxx}) &= 0 \\ v_t + av_x = -\frac{1}{2}kv_{tt} - \frac{1}{6}k^2v_{ttt} + \frac{1}{3}ah^2v_{xxx} + \frac{1}{2}a^2kv_{xx} - \frac{1}{2}a^2khv_{xxx}\end{aligned}$$

If we keep the  $O(k)$  terms:

$$v_t + av_x = -\frac{1}{2}kv_{tt} + \frac{1}{2}a^2kv_{xx}$$

Then taking derivative with respect to  $t$ :

$$v_{tt} = -av_{xt} - \frac{1}{2}kv_{ttt} + \frac{1}{2}a^2kv_{xxt}$$

With respect to  $t$ :

$$v_{ttt} = -av_{xtt} + O(k)$$

With respect to  $x$ :

$$v_{ttx} = -av_{xtx} + O(k)$$

With respect to  $x$  the initial equation:

$$v_{tx} = -av_{xx} - \frac{1}{2}kv_{ttx} + \frac{1}{2}a^2kv_{xxx}$$

With respect to  $x$  one more time:

$$v_{txx} = -av_{xxx} + O(k)$$

Combining these gives

$$v_{tt} = a^2v_{xx} + \frac{1}{2}akv_{ttt} - \frac{1}{2}a^3kv_{xxx} - \frac{1}{2}kv_{ttt} + \frac{1}{2}a^2kv_{xxt}$$

Replacing  $v_{xxt}$ :

$$v_{tt} = a^2v_{xx} - \frac{1}{2}a^2kv_{xxt} - \frac{1}{2}a^3kv_{xxx} - \frac{1}{2}kv_{ttt} + \frac{1}{2}a^2kv_{xxt} + O(k^2)$$

$$v_{tt} = a^2v_{xx} - \frac{1}{2}a^3kv_{xxx} - \frac{1}{2}kv_{ttt} + O(k^2)$$

Inserting this in the initial equation:

$$v_t + av_x = -\frac{1}{2}k(a^2v_{xx} - \frac{1}{2}a^3kv_{xxx} - \frac{1}{2}kv_{ttt}) - \frac{1}{6}k^2v_{ttt} + \frac{1}{3}ah^2v_{xxx} + \frac{1}{2}a^2kv_{xx} - \frac{1}{2}a^2khv_{xxx} + O(k^3)$$

$$v_t + av_x = \frac{1}{4}a^3k^2v_{xxx} + \frac{1}{12}k^2v_{ttt} + \frac{1}{3}ah^2v_{xxx} - \frac{1}{2}a^2khv_{xxx} + O(k^3)$$

From previous results:

$$v_{ttt} = -av_{xtt} + O(k) = a^2v_{xtx} + O(k) = -a^3v_{xxx} + O(k)$$

Finally inserting it in the previous equation:

$$v_t + av_x = \frac{1}{6}a^3k^2v_{xxx} + \frac{1}{3}ah^2v_{xxx} - \frac{1}{2}a^2khv_{xxx} + O(k^3)$$

$$v_t + av_x = \frac{1}{6}ah^2(2 - \frac{3ak}{h} + \frac{a^2k^2}{h^2})v_{xxx} + O(k^3)$$

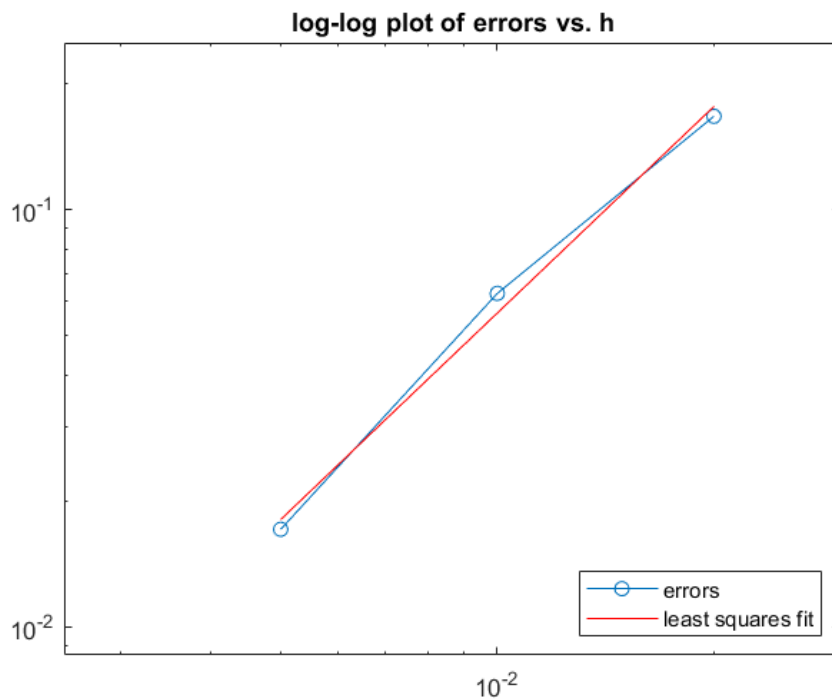
which is exactly (10.47) and of third order accuracy.

#### 0.0.4 4.

(a) The results of `error_table(hvals, E)`:

h	error	ratio	observed order
0.02000	1.66622e-01	NaN	NaN
0.01000	6.27543e-02	2.65514	1.40879
0.00500	1.70902e-02	3.67195	1.87654

The results of `error_loglog(hvals, E)`:  
Least squares fit gives  $E(h) = 108.653 * h^{1.64267}$



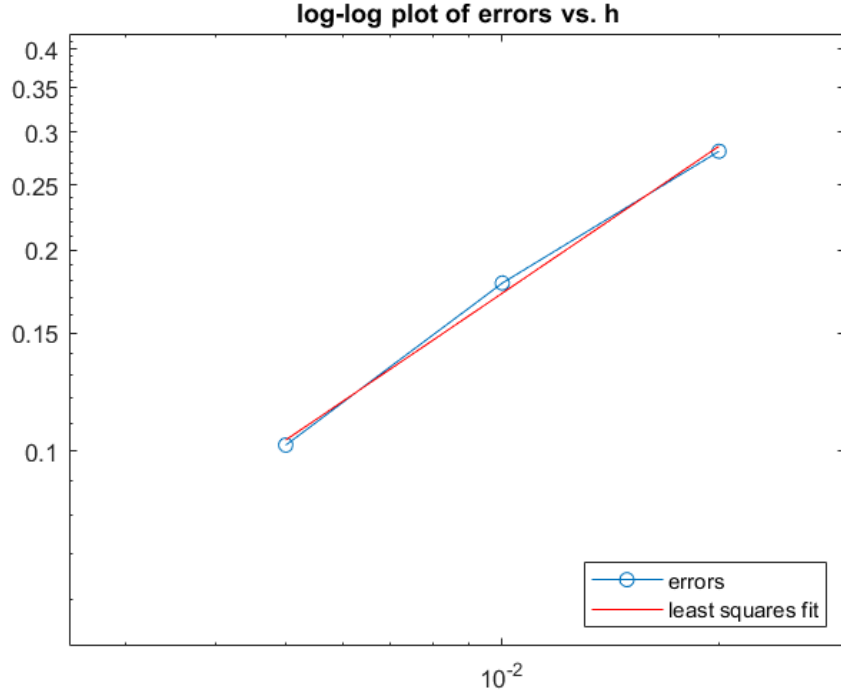
Corresponding plot:  
Indeed since  $1.64267 \sim 2$ , second-order accuracy is observed.

**(b)** The results of `error_table(hvals, E)`:

h	error	ratio	observed order
0.02000	2.81062e-01	NaN	NaN
0.01000	1.78422e-01	1.57526	0.65559
0.00500	1.02013e-01	1.74902	0.80654

The results of `error_loglog(hvals, E)`:  
Least squares fit gives  $E(h) = 4.99386 * h^{0.731067}$





Corresponding plot:

Indeed since  $0.731067 \sim 1$ , first-order accuracy is observed.

(c) The results of `error_table(hvals, E)` as we decrease  $k$ :

h	error	ratio	observed order
0.02000	2.81062e-01	NaN	NaN
0.02000	4.95204e-01	0.56757	-Inf
0.02000	5.47426e-01	0.90460	-Inf

Indeed accuracy is decreased as  $k \rightarrow 0$ .

The modified equation for the upwind method:

$$v_t + av_x = \frac{1}{2}ah\left(1 - \frac{ak}{h}\right)v_{xx}$$

Since we are solving the advection equation  $v_t + av_x = 0$  we can consider the RHS to be an error term:

$$E(k) = \frac{1}{2}ah\left(1 - \frac{ak}{h}\right)v_{xx}$$

Let's say  $k = ch$  then:

$$E(ch) = \frac{1}{2}ah\left(1 - \frac{ach}{h}\right)v_{xx}$$

$$E(ch) = \frac{1}{2}ah(1 - ac)v_{xx}$$

Notice that for reasonable  $a, c$  values  $(1 - ac) \in [0, 1]$  and as  $c \rightarrow 0$  we have  $(1 - ac) \rightarrow 1$ . In other words as  $k \rightarrow 0$  our error increases causing the mentioned paradox.

0.0.5 5.

$$\begin{aligned} e^{\frac{Ak}{2}} e^{Bk} e^{\frac{Ak}{2}} &= (I + \frac{1}{2}kA + \frac{1}{8}k^2A^2 + \dots)(I + kB + \frac{1}{2}k^2B^2 + \dots)(I + \frac{1}{2}kA + \frac{1}{8}k^2A^2 + \dots) \\ &= (I + \frac{1}{2}kA + \frac{1}{8}k^2A^2 + kB + \frac{1}{2}k^2AB + \frac{1}{2}k^2B^2 + \dots)(I + \frac{1}{2}kA + \frac{1}{8}k^2A^2 + \dots) \\ &= I + \frac{1}{2}kA + \frac{1}{8}k^2A^2 + kB + \frac{1}{2}k^2AB + \frac{1}{2}k^2B^2 + \frac{1}{2}kA + \frac{1}{4}k^2A^2 + \frac{1}{2}k^2BA + \frac{1}{8}k^2A^2 + \dots \\ &= I + k(A + B) + \frac{1}{2}k^2(A^2 + AB + BA + B^2) + \dots \end{aligned}$$

Since Taylor expansions are exactly the same up to quadratic term, we can conclude that the Strang splitting is second order accurate on the problem (11.18).