

CAAM 336 · DIFFERENTIAL EQUATIONS

Problem Set 8 · Solutions

Posted Monday 22 October 2012. Due Monday 29 October 2012, 5pm. Corrected 25 October 2012.

1. [50 points: 18 points for (a); 12 points for (b); 10 points each for (c) and (d)]

Consider the following three matrices:

$$(i) \quad \mathbf{A} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad (ii) \quad \mathbf{A} = \begin{bmatrix} -50 & 49 \\ 49 & -50 \end{bmatrix} \quad (iii) \quad \mathbf{A} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}.$$

- (a) For each of the matrices (i)–(iii), compute the matrix exponential $e^{t\mathbf{A}}$.

You may use `eig` for the eigenvalues and eigenvectors, but please construct the matrix exponential “by hand” (not with `expm`). For diagonalizable $\mathbf{A} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^{-1}$, recall the formula $e^{t\mathbf{A}} = \mathbf{V}e^{t\mathbf{\Lambda}}\mathbf{V}^{-1}$. If you encounter a complex eigenvalue $\lambda = \alpha + i\beta$, you may use the formula

$$e^\lambda = e^{\alpha+i\beta} = e^\alpha(\cos(\beta) + i\sin(\beta)).$$

- (b) Use your answers in part (a) to explain the behavior of solutions $\mathbf{x}(t)$ to the differential equation $\mathbf{x}'(t) = \mathbf{A}\mathbf{x}(t)$ as $t \rightarrow \infty$, given that $\mathbf{x}(0) = [2, 0]^T$ (e.g., specify and explain exponential growth, exponential decay, or neither) for each of the three matrices (i)–(iii).
- (c) For the matrix (ii), describe how large one can choose the time step Δt so that the forward Euler method applied to $\mathbf{x}'(t) = \mathbf{A}\mathbf{x}(t)$,

$$\mathbf{x}_{k+1} = \mathbf{x}_k + \Delta t \mathbf{A} \mathbf{x}_k,$$

will produce a solution that qualitatively matches the behavior of the true solution (i.e., the approximations \mathbf{x}_k should grow, decay, or remain of the same size as the true solution does).

Answer the same question for the backward Euler method

$$\mathbf{x}_{k+1} = \mathbf{x}_k + \Delta t \mathbf{A} \mathbf{x}_{k+1}.$$

- (d) For the matrix in (iii), describe how the forward Euler method behaves *for all* Δt as $k \rightarrow \infty$ for $\mathbf{x}(0) = [1, 1]^T$. Now describe how the backward Euler method must behave as $k \rightarrow \infty$ for the same matrix and initial condition.

Solution.

[GRADERS: it is acceptable for students to use MATLAB to compute eigendecompositions, but they should not simply use the `expm` command. In particular, only give half credit if students computed $e^{t\mathbf{A}}$ for a fixed value of t . The correct answer should depend on the variable t .]

- (a) We compute the matrix exponentials for each matrix in turn.

- (i) Note that $\det(\lambda\mathbf{I} - \mathbf{A}) = \lambda^2 - 1 = (\lambda + 1)(\lambda - 1)$ and hence the eigenvalues of \mathbf{A} are $\lambda_1 = -1$ and $\lambda_2 = 1$. The corresponding (normalized) orthogonal eigenvectors are

$$\mathbf{u}_1 = \frac{\sqrt{2}}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad \mathbf{u}_2 = \frac{\sqrt{2}}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

As \mathbf{A} is symmetric, if we set $\mathbf{U} = [\mathbf{u}_1 \ \mathbf{u}_2]$ and $\mathbf{\Lambda} = \text{diag}(\lambda_1, \lambda_2)$, we have $\mathbf{A} = \mathbf{U}\mathbf{\Lambda}\mathbf{U}^*$ and

$$e^{t\mathbf{A}} = \mathbf{U}e^{t\mathbf{\Lambda}}\mathbf{U}^* = \mathbf{U} \begin{bmatrix} e^{-t} & 0 \\ 0 & e^t \end{bmatrix} \mathbf{U}^*.$$

Multiplying this out gives

$$e^{t\mathbf{A}} = \frac{1}{2} \begin{bmatrix} e^t + e^{-t} & e^t - e^{-t} \\ e^t - e^{-t} & e^t + e^{-t} \end{bmatrix}.$$

- (ii) If we denote the matrix in part (a) as \mathbf{A}_1 , then we find that the \mathbf{A} in part (b) can be written as $\mathbf{A} = -50\mathbf{I} + 49\mathbf{A}_1$, from which it follows that \mathbf{A} has eigenvalues $\lambda_1 = -99$ and $\lambda_2 = -1$ with the same eigenvectors as in part (a):

$$\mathbf{u}_1 = \frac{\sqrt{2}}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad \mathbf{u}_2 = \frac{\sqrt{2}}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

Again \mathbf{A} is symmetric, and we have that

$$e^{t\mathbf{A}} = \mathbf{U}e^{t\mathbf{\Lambda}}\mathbf{U}^* = \frac{1}{2} \begin{bmatrix} e^{-t} + e^{-99t} & e^{-t} - e^{-99t} \\ e^{-t} - e^{-99t} & e^{-t} + e^{-99t} \end{bmatrix}.$$

- (iii) [GRADERS: please be a bit more lenient with this problem, as this \mathbf{A} is nonsymmetric, a case we didn't dwell excessively on in class.]

The characteristic polynomial is $\det(\lambda\mathbf{I} - \mathbf{A}) = \lambda^2 + 1 = (\lambda - i)(\lambda + i)$, where $i^2 = -1$. Hence the eigenvalues are $\lambda_1 = -i$ and $\lambda_2 = i$. The corresponding normalized eigenvectors are

$$\mathbf{v}_1 = \frac{\sqrt{2}}{2} \begin{bmatrix} 1 \\ -i \end{bmatrix}, \quad \mathbf{v}_2 = \frac{\sqrt{2}}{2} \begin{bmatrix} 1 \\ i \end{bmatrix}.$$

Since \mathbf{A} is not symmetric we write $\mathbf{A} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^{-1}$, where $\mathbf{V} = [\mathbf{v}_1 \ \mathbf{v}_2]$ and $\mathbf{\Lambda} = \text{diag}(\lambda_1, \lambda_2)$, and the matrix exponential takes the form

$$e^{t\mathbf{A}} = \mathbf{V}e^{t\mathbf{\Lambda}}\mathbf{V}^{-1} = \frac{1}{2} \begin{bmatrix} e^{it} + e^{-it} & i(e^{-it} - e^{it}) \\ i(e^{it} - e^{-it}) & e^{it} - e^{-it} \end{bmatrix}.$$

Note that for real numbers t ,

$$e^{it} = \cos(t) + i \sin(t)$$

and

$$e^{-it} = \cos(-t) + i \sin(-t) = \cos(-t) - i \sin(t),$$

and hence one could arrive at the simple formula (not required):

$$e^{t\mathbf{A}} = \begin{bmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{bmatrix}.$$

Alternatively, one can note that $\mathbf{A}^2 = -\mathbf{I}$, $\mathbf{A}^3 = -\mathbf{A}$, $\mathbf{A}^4 = \mathbf{I}$, ..., to obtain from the series expression

$$e^{t\mathbf{A}} = \mathbf{I} + t\mathbf{A} + \frac{1}{2}t^2\mathbf{A}^2 + \frac{1}{3!}t^3\mathbf{A}^3 + \dots$$

that

$$e^{t\mathbf{A}} = \begin{bmatrix} 1 - \frac{1}{2}t^2 + \frac{1}{4!}t^4 - \frac{1}{6!}t^6 + \dots & t - \frac{1}{3!}t^3 + \frac{1}{5!}t^5 - \frac{1}{7!}t^7 + \dots \\ -t + \frac{1}{3!}t^3 - \frac{1}{5!}t^5 + \frac{1}{7!}t^7 - \dots & 1 - \frac{1}{2}t^2 + \frac{1}{4!}t^4 - \frac{1}{6!}t^6 + \dots \end{bmatrix} = \begin{bmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{bmatrix}.$$

Here we have spotted that the entries in this matrix are the Taylor series for $\sin(t)$ and $\cos(t)$.

- (b) The behavior of $\mathbf{x}(t) = e^{t\mathbf{A}}\mathbf{x}(0)$ for $\mathbf{x}(0) = [2, 0]^T$ depends on the matrix.

- (i) For the specified initial condition, we have

$$\mathbf{x}(t) = e^{t\mathbf{A}}\mathbf{x}(0) = \begin{bmatrix} e^t + e^{-t} \\ e^t - e^{-t} \end{bmatrix}.$$

Thus, as $t \rightarrow \infty$, the solution $\mathbf{x}(t)$ blows up. (In fact, it behaves like $e^t[1 \ 1]^T$.)

- (ii) For the given $\mathbf{x}(0)$, we have

$$\mathbf{x}(t) = e^{t\mathbf{A}}\mathbf{x}(0) = \begin{bmatrix} e^{-t} + e^{-99t} \\ e^{-t} - e^{-99t} \end{bmatrix},$$

and hence $\mathbf{x}(t) \rightarrow \mathbf{0}$ as $t \rightarrow \infty$. This must be true since both eigenvalues of \mathbf{A} are negative.

(iii) Notice that

$$\mathbf{x}(t) = e^{t\mathbf{A}}\mathbf{x}(0) = 2 \begin{bmatrix} \cos(t) \\ -\sin(t) \end{bmatrix},$$

so $\mathbf{x}(t)$ neither grows nor decays. (In fact, $\|\mathbf{x}(t)\|$ is constant!)

- (c) The eigenvalues for the matrix given by (ii) are $\lambda_1 = -99$ and $\lambda_2 = -1$. Thus the solution to the equation $d\mathbf{x}/dt = \mathbf{A}\mathbf{x}$ will decay to zero as $t \rightarrow \infty$ for all choices of initial condition $\mathbf{x}(0)$.

We wish to choose the step size Δt for the forward Euler method so that the iterates \mathbf{x}_k decay to zero as $k \rightarrow \infty$. For this equation

$$\mathbf{x}_k = (\mathbf{I} + \Delta t \mathbf{A})^k \mathbf{x}_0,$$

and so we need all eigenvalues of the symmetric matrix $\mathbf{I} + \Delta t \mathbf{A}$ to be less than one in magnitude. The eigenvalues of $\mathbf{I} + \Delta t \mathbf{A}$ are simply

$$\mu_1 = 1 + \Delta t \lambda_1 = 1 - 99\Delta t, \quad \mu_2 = 1 + \Delta t \lambda_2 = 1 - \Delta t.$$

For all $0 < \Delta t < 2$ we have $|\mu_2| < 1$, but to get $|\mu_1| < 1$ we have a stricter requirement:

$$0 < \Delta t < 2/99.$$

(Alternatively, one can simply look for Δt such that $\Delta t \lambda_1, \Delta t \lambda_2 \in (-2, 0)$.)

For the backward Euler method, we have

$$\mathbf{x}_k = (\mathbf{I} - \Delta t \mathbf{A})^{-k} \mathbf{x}_0,$$

and we need all eigenvalues of $(\mathbf{I} - \Delta t \mathbf{A})^{-1}$ to be less than one in magnitude. Those eigenvalues are

$$\mu_1 = \frac{1}{1 - \Delta t \lambda_1} = \frac{1}{1 + 99\Delta t}, \quad \mu_2 = \frac{1}{1 - \Delta t \lambda_2} = \frac{1}{1 + \Delta t}.$$

These values are less than one in magnitude for all $\Delta t > 0$, so there is no restriction on Δt to obtain $\mathbf{x}_k \rightarrow 0$ as $k \rightarrow \infty$.

- (d) The diagonalizable matrix \mathbf{A} given in (iii) has eigenvalues $\lambda_{\pm} = \pm i$. It follows that the forward Euler iterations, given by

$$\mathbf{x}_k = (\mathbf{I} + \Delta t \mathbf{A})^k \mathbf{x}(0)$$

will behave as $k \rightarrow \infty$ like eigenvalues of $(\mathbf{I} + \Delta t \mathbf{A})^k$, i.e., like $(1 + \Delta t \lambda_{\pm})^k$. Since

$$|1 + \Delta t \lambda_{\pm}| = |1 \pm i\Delta t| = \sqrt{1 + (\Delta t)^2} > 1,$$

we conclude that the forward Euler iterates will always blow up as $k \rightarrow \infty$ for any choice of $\Delta t > 0$.

On the other hand, the backward Euler iterates,

$$\mathbf{x}_k = (\mathbf{I} - \Delta t \mathbf{A})^{-k} \mathbf{x}(0)$$

will behave as $k \rightarrow \infty$ like eigenvalues of $(\mathbf{I} - \Delta t \mathbf{A})^{-k}$, i.e., like $(1 - \Delta t \lambda_{\pm})^{-k}$. Since

$$\left| \frac{1}{1 - \Delta t \lambda_{\pm}} \right| = \frac{1}{|1 \mp i\Delta t|} = \frac{1}{\sqrt{1 + (\Delta t)^2}} < 1,$$

we conclude that the backward Euler iterates will always decay as $k \rightarrow \infty$ for any choice of $\Delta t > 0$.

2. [50 points: 12 points for (a) and (b); 16 points for (c); 10 points for (d)]

There exist a host of alternatives to the forward and backward Euler methods for approximating the solution of the differential equation $\mathbf{x}'(t) = \mathbf{A}\mathbf{x}(t)$. For example, the *trapezoid method* has the form

$$\mathbf{x}_{k+1} = \mathbf{x}_k + \frac{1}{2} \Delta t \mathbf{A}(\mathbf{x}_k + \mathbf{x}_{k+1}),$$

where $\Delta t > 0$ is the time-step, and $\mathbf{x}_k \approx \mathbf{x}(t_k)$ for $t_k = k\Delta t$.

- (a) Like backward Euler, the trapezoid method is an *implicit* technique: \mathbf{x}_{k+1} appears on both the right and left hand side of the formula above that defines it. Describe how to find \mathbf{x}_{k+1} given \mathbf{x}_k . In particular, what linear system of algebraic equations needs to be solved at each step? (For comparison, the backward Euler method requires the solution of the system $(\mathbf{I} - \Delta t \mathbf{A})\mathbf{x}_{k+1} = \mathbf{x}_k$ for the unknown \mathbf{x}_{k+1} at each step.)
- (b) Consider the matrix and initial condition

$$\mathbf{A} = \begin{bmatrix} -1 & 10 \\ 0 & -2 \end{bmatrix}, \quad \mathbf{x}(0) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

- Approximate the solution to $\mathbf{x}'(t) = \mathbf{A}\mathbf{x}(t)$ on the interval $t \in [0, 5]$ for time step $\Delta t = .05$. Produce a `semilogy` plot showing $t_k = k\Delta t$ versus $\|\mathbf{x}_k\|$ for $k = 0, \dots, 100$. (Use the `norm` command in MATLAB to compute $\|\mathbf{x}_k\|$.)
- (c) We wish to understand how the error in our approximation at time $t = 1$ improves as we run the simulation with smaller and smaller Δt values. Produce a `loglog` plot showing Δt versus the error in the trapezoid rule and backward Euler approximations for the matrix and initial condition in part (b) at time $t = 1$. To compute the error, first find the exact solution $\mathbf{x}(1) = e^{\mathbf{A}}\mathbf{x}(0)$ using the `expm` command, then compute the norms $\|\hat{\mathbf{x}} - \mathbf{x}(1)\|$, where $\hat{\mathbf{x}}$ denotes your approximation to $\mathbf{x}(1)$ from the trapezoid or backward Euler methods. Start your plot with $\Delta t = 1/2$ and use sufficiently many smaller values of Δt to make the trend in your plot clear. For which method does the error decay more rapidly as $\Delta t \rightarrow 0$?
- (d) Forward Euler iterates can be written as $\mathbf{x}_k = (\mathbf{I} + \Delta t \mathbf{A})^k \mathbf{x}_0$, while backward Euler iterates can be written as $\mathbf{x}_k = (\mathbf{I} - \Delta t \mathbf{A})^{-k} \mathbf{x}_0$.
- Work out a similar formula for the iterates \mathbf{x}_k generated by the trapezoid method.
- Suppose $\mathbf{A} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^T$ is symmetric, and all of its eigenvalues λ_j , $j = 1, \dots, n$, are negative. How must you choose the time step Δt to ensure that the iterates \mathbf{x}_k generated by the trapezoid method converge to zero, $\|\mathbf{x}_k\| \rightarrow 0$, as $k \rightarrow \infty$?

Solution.

- (a) Given the trapezoid rule

$$\mathbf{x}_{k+1} = \mathbf{x}_k + \frac{1}{2}\Delta t \mathbf{A}(\mathbf{x}_k + \mathbf{x}_{k+1}),$$

group all terms involving \mathbf{x}_{k+1} on the left and \mathbf{x}_k on the right to obtain

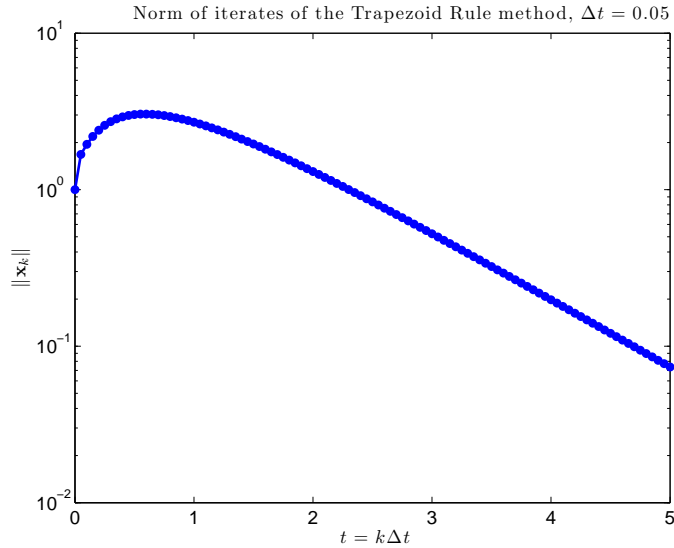
$$(\mathbf{I} - \frac{1}{2}\Delta t \mathbf{A})\mathbf{x}_{k+1} = (\mathbf{I} + \frac{1}{2}\Delta t \mathbf{A})\mathbf{x}_k.$$

One can solve then solve this system for \mathbf{x}_{k+1} (e.g., using MATLAB's 'backslash' command). Alternatively, invert the matrix on the left to obtain a formula for \mathbf{x}_{k+1} :

$$\mathbf{x}_{k+1} = (\mathbf{I} - \frac{1}{2}\Delta t \mathbf{A})^{-1}(\mathbf{I} + \frac{1}{2}\Delta t \mathbf{A})\mathbf{x}_k.$$

(Either of these forms is acceptable for full credit.)

- (b) The plot below shows the growth of norm of the solution \mathbf{x}_k as a function of k for $\Delta t = .05$. Though both eigenvalues of \mathbf{A} are negative, there is transient growth in \mathbf{x}_k before the eventual decay.



```

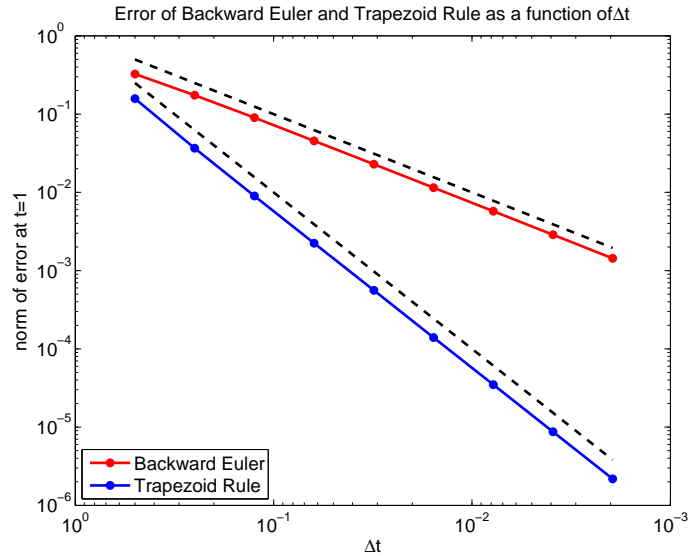
A = [-1 10; 0 -2];
x0 = [1;1];

tfinal = 5;
dt = .05;
I = eye(2);
x_trap = x0;
normx = 1;
for k=1:tfinal/dt
    x_trap = (I-.5*dt*A)\(I+.5*dt*A)*x_trap; % trapezoid rule
    normx = [normx;norm(x_trap)];
end
figure(1), clf
semilogy(0:dt:tfinal, normx, 'b.-','linewidth',2,'markersize',20)
xlim([0 tfinal])
set(gca,'fontsize',14)
xlabel('$t = k\{\Delta\}t$', 'fontsize',14,'interpreter','latex')
ylabel('$\| \mathbf{x}_k \|$', 'fontsize',14,'interpreter','latex')
title('Norm of iterates of the Trapezoid Rule method, $\Delta\{t\} = 0.05$',...
      'fontsize',14,'interpreter','latex')
print -depsc2 traprule1.eps

```

- (c) The following plot shows the error in the backward Euler and trapezoid rule computations as a function of the step size Δt . (Note use of the `set('gca','Xdir','reverse')` command to reverse the direction of the horizontal axis so that Δt decreases from left to right.

[GRADERS: Please deduct 7 points if the two lines have the same slope. This error most likely comes from students comparing the approximate solution at time $t = 1 \pm \Delta t$ to the true solution at $t = 1$.]



The black dashed lines show Δt and $(\Delta t)^2$: Note that the backward Euler error decreases at the rate Δt , while the trapezoid rule decreases at the rate $(\Delta t)^2$. Thus, as Δt is cut in half, the trapezoid rule error is quartered. Thus, the trapezoid rule is considerably better.

(Though not necessary for this problem, a complete analysis would also consider the amount of work required at each step. On that count the trapezoid rule is a bit more expensive, as it requires an extra matrix-vector multiplication at each step.)

```
A = [-1 10; 0 -2];
x0 = [1;1];

dtvec = 2.^-[1:9]';
tfinal = 1;
I = eye(2);
err_euler = zeros(size(dtvec));
err_trap = zeros(size(dtvec));

for j = 1:length(dtvec)
    dt = dtvec(j);
    x_euler = x0; x_trap = x0;
    normx = 1;
    for k=1:tfinal/dt
        x_euler = (I-dt*A)\x_euler; % backward Euler
        x_trap = (I-.5*dt*A)\(I+.5*dt*A)*x_trap; % trapezoid rule
    end
    x_true = expm(A*tfinal)*x0;
    err_euler(j) = norm(x_euler-x_true);
    err_trap(j) = norm(x_trap-x_true);
end
figure(1), clf
loglog(dtvec, err_euler, 'r.-','linewidth',2,'markersize',20), hold on
loglog(dtvec, err_trap, 'b.-','linewidth',2,'markersize',20)
loglog(dtvec, dtvec, 'k--','linewidth',2)
loglog(dtvec, dtvec.^2, 'k--','linewidth',2)
legend('Backward Euler', 'Trapezoid Rule', 3)
set(gca,'fontsize',14,'xdir','reverse')
xlabel('{\Delta}t'), ylabel('norm of error at t=1')
title('Error of Backward Euler and Trapezoid Rule as a function of {\Delta}t')
print -depsc2 traprule2.eps
```

- (d) The formula in part (a) enables the computation we need to make for this part. First, by applying k steps of the trapezoid method, we have the formula

$$\mathbf{x}_k = (\mathbf{I} - \frac{1}{2}\Delta t \mathbf{A})^{-k} (\mathbf{I} + \frac{1}{2}\Delta t \mathbf{A})^k \mathbf{x}_0.$$

Next, the problem asks how the method will behave as $k \rightarrow \infty$ if \mathbf{A} is symmetric with all eigenvalues negative. In this case write $BA = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^T$, so that

$$\begin{aligned}\mathbf{x}_k &= (\mathbf{I} - \tfrac{1}{2}\Delta t\mathbf{A})^{-k}(\mathbf{I} + \tfrac{1}{2}\Delta t\mathbf{A})^k \mathbf{x}_0 \\ &= \mathbf{V}(\mathbf{I} - \tfrac{1}{2}\Delta t\mathbf{\Lambda})^{-k}(\mathbf{I} + \tfrac{1}{2}\Delta t\mathbf{\Lambda})^k \mathbf{V}^T \mathbf{x}_0.\end{aligned}$$

The diagonal entries in

$$(\mathbf{I} - \tfrac{1}{2}\Delta t\mathbf{\Lambda})^{-k}(\mathbf{I} + \tfrac{1}{2}\Delta t\mathbf{\Lambda})^k$$

have the form

$$\frac{(1 + \tfrac{1}{2}\Delta t\lambda)^k}{(1 - \tfrac{1}{2}\Delta t\lambda)^k} = \left(\frac{1 + \tfrac{1}{2}\Delta t\lambda}{1 - \tfrac{1}{2}\Delta t\lambda} \right)^k,$$

and so the behavior of \mathbf{x}_k as $k \rightarrow \infty$ will be controlled by

$$\left| \frac{1 + \tfrac{1}{2}\Delta t\lambda}{1 - \tfrac{1}{2}\Delta t\lambda} \right|.$$

If this quantity is less than one for all eigenvalues λ of \mathbf{A} , then $\mathbf{x}_k \rightarrow \mathbf{0}$ as $k \rightarrow \infty$. If any of these quantities is larger than one, there exist initial conditions for which $\|\mathbf{x}_k\| \rightarrow \infty$ as $k \rightarrow \infty$.

[GRADERS: Please deduct 3 points if the student does not make the following key observation.]

We have not yet used the fact that $\lambda < 0$. What do we learn from this? If $\lambda < 0$, then

$$|1 - \tfrac{1}{2}\Delta t\lambda| = 1 + \tfrac{1}{2}\Delta t|\lambda| > |1 + \tfrac{1}{2}\Delta t\lambda|,$$

and so

$$\left| \frac{1 + \tfrac{1}{2}\Delta t\lambda}{1 - \tfrac{1}{2}\Delta t\lambda} \right| < 1.$$

Hence, $\mathbf{x}_k \rightarrow \mathbf{0}$ for all initial conditions.
