Solutions:

1. Root Finding.

(a) Let the root be \( x = \alpha \). We subtract \( \alpha \) from both sides of \( x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \) and write
\[
x_n - \alpha = \varepsilon_n
to obtain \( \varepsilon_{n+1} = \varepsilon_n - \frac{f(\alpha + \varepsilon_n)}{f'(\alpha + \varepsilon_n)} = \varepsilon_n - \frac{f(\alpha) + f'(\alpha) \varepsilon_n + O(\varepsilon_n^2)}{f'(\alpha) + O(\varepsilon_n)} \). Using \( f(\alpha) = 0 \), the RHS simplifies to \( \varepsilon_{n+1} - (\varepsilon_n + O(\varepsilon_n^2)) = O(\varepsilon_n^2) \).

(b) Let the update in the iteration step be \( x_{n+1} - x_n = \Delta x_n \). We want to achieve
\[
0 = f(x_n + \Delta x_n) = f(x_n) + \Delta x_n f'(x_n) + \frac{1}{2} (\Delta x_n)^2 f''(x_n) + \ldots \tag{1}
\]
Ignoring the last term and then solving for \( \Delta x_n \) gives the standard Newton iteration: \( \Delta x_n = -\frac{f(x_n)}{f'(x_n)} \). Since this is a very good approximation to the ideal \( \Delta x_n \), substituting this into the last term of (1) is much better than ignoring it. Doing this and then solving equation (1) again for \( \Delta x_n \) gives the desired formula.


(a) On each subinterval, the interpolation error in the trapezoidal approximation can (by the error formula for polynomial interpolation, in case of a linear function) be estimated by \(|f(x) - p_1(x)| \leq \frac{\varepsilon_n}{2} (x - x_{k+1})^2 f''(\xi) = O(h^2) \) (since both \( |x - x_k| \) and \( |x - x_{k+1}| \) are of size \( O(h) \)). Over an interval length of \( b - a \), the integration error can then be no more than \((b - a) \cdot O(h^2)\), which still is \( O(h^2) \).

(b) In the expansion \( e^{\cos x} = \sum_{k=0}^{\infty} a_k \cos kx \), the trapezoidal rule is exact for the first (constant) term and then gives the correct (zero) value for all further cosine modes until we reach the first mode (following the constant one) that takes the value one at all the node points. This mode gives the erroneous result of \( 2\pi \cdot a_n \), which will dominate the total error. With \( n = 6 \), the error will thus be approximately \( \frac{2\pi^2}{11520} = \pi / 11520 \approx 3 \cdot 10^{-4} \).
3. Interpolation/Approximation.

(a) \( p_n(x) = \sum_{k=0}^{n} L_k(x) f_k \), where \( L_k(x) = \frac{(x-x_0) \cdots (x-x_{k-1})(x-x_{k+1}) \cdots (x-x_n)}{(x_k-x_0) \cdots (x_k-x_{k-1})(x_k-x_{k+1}) \cdots (x_k-x_n)} \).

(b) Suppose there are two different polynomials \( p_n(x) \) and \( q_n(x) \) that both take the values \( f_k \) at node locations \( x_k, \ k = 0, 1, \ldots, n \). The difference \( p_n(x) - q_n(x) \) is again a polynomial of degree \( n \) but with \( n+1 \) zeros, showing that it must be identically zero, in conflict with the assumption that \( p_n(x) \) and \( q_n(x) \) were different.

(c) Each of the following three approaches will show that, for \( n + 1 \) nodes, the polynomial degree will be \( 2n + 1 \).

(i) **Direct solution of linear system:** Let the Hermite polynomial be \( H_{2n+1}(x) = a_0 + a_1 x + a_2 x^2 + \ldots + a_{2n+1} x^{2n+1} \). Imposing all the \( 2n + 2 \) requirements gives a square \((2n+2) \times (2n+2)\) linear system of the following structure for the coefficients:

\[
\begin{bmatrix}
1 & x_0 & x_0^2 & x_0^3 & \cdots \\
0 & 1 & 2x_0 & 3x_0^2 & \cdots \\
0 & 1 & 2x_1 & 3x_1^2 & \cdots \\
& & & & \\
& & & & \\
\end{bmatrix}
\begin{bmatrix}
a_0 \\
a_1 \\
\vdots \\
a_{2n+1} \\
\end{bmatrix}
= \begin{bmatrix}
f_0 \\
f_1 \\
\vdots \\
f_{2n+1} \\
\end{bmatrix}.
\]

(ii) **Based on Lagrange interpolation:** With \( L_k(x) \) denoting the Legendre kernel, the polynomials \( h_i(x) = (x-x_i) L_i(x)^2 \) and \( h_i(x) = [1 - 2L_i'(x_i)(x-x_i)] L_i(x)^2 \) will have the properties that \( h_i'(x_j) = h_i(x_j) = 0, \ 0 \leq i, j \leq n \) and \( h_i(x_j) = h_i'(x_j) = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases} \). The Hermite interpolation polynomial is then given by

\( H(x) = \sum_{i=0}^{n} h_i(x) f_i + \sum_{i=0}^{n} h_i(x) f_i' \).

(iii) **Based on Newton interpolation:** One can generalize the standard divided difference layout for Newton’s interpolation method by duplicating each node information (location and function value) and in the next column insert the derivative information. Following that, one proceeds as usual. In case of just two nodes, the divided difference tables for regular interpolation and for Hermite interpolation will take the forms:

\[
\begin{array}{cccc}
x_0 & f_0 & f'_0 \\
\bullet & f_0 & \bullet \\
\bullet & \bullet & \bullet \\
x_1 & f_1 & f'_1 \\
x_1 & f_1 & \\
\end{array}
\]

where the entries marked by dots are completed in the regular manner. In either case, the interpolation polynomial would be created from the resulting numbers in the top right diagonal. This Hermite extension works by the principle that each node can be seen as the limit of two nodes coming together, with function values approaching each other in just the right way that the derivative information becomes obeyed.
4. Linear algebra

(a) Since $A$ is an antisymmetric matrix, its eigenvalues are purely imaginary, or zero. Since it is a matrix with real entries, the roots of the characteristic polynomial come in pairs (if they are complex-valued). For odd-sized matrix these two conditions force at least one of the eigenvalues to be zero.

(b) For even-sized matrix the product of a pair of complex-valued eigenvalues is always positive and the conclusion follows.

(c) For non-zero eigenvalues the limit matrix will have a block-diagonal structure with two-by-two block size.

5. ODEs

(a) A general multistep method with $s$ steps to solve

\[
\begin{aligned}
    \frac{dy}{dt} &= f(t, y) \\
    y(0) &= y_0
\end{aligned}
\]

is of the form

\[
\sum_{m=0}^{s} a_m y_{n+m} = h \sum_{m=0}^{s} b_m f(t_{n+m}, y_{n+m}),
\]

$n = 0, 1, 2, \ldots$, where $a_s = 1$. For the test problem

\[
\begin{aligned}
    y' &= \lambda y \\
    y(0) &= y_0,
\end{aligned}
\]

consider solutions to the recurrence

\[
\sum_{m=0}^{s} a_m y_{n+m} = \lambda h \sum_{m=0}^{s} b_m y_{n+m}, \quad n = 0, 1, 2, \ldots
\]

The region of absolute stability of a method is a domain $\lambda h \in D \subset \mathbb{C}$ such that the solution of the corresponding linear recurrence is bounded.

The method is stable if zero belongs to the region of absolute stability

Replacing all computed values $y_n$ by the values of the solution $y(t_n)$ and using the ODE to evaluate local error, we define the order of the method as $p \geq 1$ (if it exists) such that

\[
\sum_{m=0}^{s} a_m y(t + mh) - h \sum_{m=0}^{s} b_m y'(t + mh) = O(h^{p+1})
\]

as $h \to 0$. In fact, for a multistep method, one can find the order by taking $y$ to be a polynomial and finding the largest degree for which

\[
\sum_{m=0}^{s} a_m y(t + mh) - h \sum_{m=0}^{s} b_m y'(t + mh) = 0.
\]

The method is consistent if its order $p \geq 1$. These easily verified algebraic properties of the method, namely its consistency and stability, imply its convergence (a property which otherwise requires a fairly lengthy derivation).
(b) For an explicit multistep method, the equation for the roots of the characteristic polynomial has the form
\[ \phi(u) = u^8 + \text{lower order terms} = 0. \]
Since the polynomial can be written in terms of its roots as
\[ \phi(u) = (u - u_1) (u - u_2) \ldots (u - u_s), \]
and in the region of absolute stability all roots \(|u_k| \leq 1\), we conclude that, in that region, all coefficients of the polynomial \(\phi\) are bounded (independently of \(\lambda h\)). However, if the region of absolute stability is unbounded, then some of the coefficients of \(\phi\) will become arbitrarily large since they exhibit linear dependence on \(\lambda h\).

6. PDEs

Lax-Wendro method is second order in both space and time variables. To see that, we write
\[ \frac{u_{j}^{n+1} - u_{j}^{n}}{h_t} = -c \frac{u_{j+1}^{n} - u_{j-1}^{n}}{2h_x} + \frac{1}{2} c^2 h_t \frac{u_{j+1}^{n} - 2u_{j}^{n} + u_{j-1}^{n}}{h_x^2} \]
and consider
\[ \psi(t, x) = \frac{u(t + h_t, x) - u(t, x)}{h_t} + c \frac{u(t, x + h_x) - u(t, x - h_x)}{2h_x} \]
\[ + \frac{1}{2} c^2 h_t \frac{u(t, x + h_x) - 2u(t, x) + u(t, x - h_x)}{h_x^2}. \]
We have
\[ \frac{u(t + h_t, x) - u(t, x)}{h_t} = u_t + \frac{1}{2} u_{tt} h_t + O(h_t^2) \]
\[ \frac{u(t, x + h_x) - u(t, x - h_x)}{2h_x} = u_x(t, x) + O(h_x^2) \]
and obtain
\[ \psi(t, x) = u_t + \frac{1}{2} u_{tt} h_t + cu_x(t, x) - \frac{1}{2} c^2 u_{xx}(t, x) h_t + O(h_t^2) + O(h_x^2). \]
From the equation, we have
\[ u_x = -\frac{1}{c} u_t \text{ and } u_{xx} = \frac{1}{c^2} u_{tt}, \]
so that
\[ \psi(t, x) = u_t + \frac{1}{2} u_{tt} h_t - u_t - \frac{1}{2} u_{tt} h_t + O(h_t^2) + O(h_x^2) = O(h_t^2) + O(h_x^2). \]
For stability analysis, we write the scheme as
\[ u_{j}^{n+1} = A u_{j+1}^{n} - B u_{j}^{n} + C u_{j-1}^{n}, \]
where
\[ A = \frac{1}{2} (c\mu)^2 - \frac{1}{2} c\mu, \quad B = (c\mu)^2 - 1 \quad \text{and} \quad C = \frac{1}{2} (c\mu)^2 + \frac{1}{2} c\mu. \]

Using \( \{ e^{ijkh_x} \}_{j=0}^{N-1} \) as an eigenvector (with index \( k = 0, \ldots, N-1 \)), we compute
\[
A e^{i(j+1)kh_x} - B e^{ijkh_x} + C e^{i(j-1)kh_x} = e^{ijkh_x} \left( A e^{ikh_x} - B + C e^{-ikh_x} \right)
= e^{ijkh_x} \left[ 1 - (c\mu)^2 + (c\mu)^2 \cos(kh_x) - ic\mu \sin(kh_x) \right]
\]
Computing the absolute value of the eigenvalue \( \lambda_k = 1 - (c\mu)^2 + (c\mu)^2 \cos(kh_x) - ic\mu \sin(kh_x) \), we have
\[
|\lambda_k|^2 = \left[ 1 - (c\mu)^2 + (c\mu)^2 \cos^2(kh_x) \right] + (c\mu)^2 \sin^2(kh_x)
= \left[ 1 - (c\mu)^2 \sin^2(kh_x) \right] + (c\mu)^2 \sin^2(kh_x).
\]
Setting \( a = (c\mu)^2 \), \( a > 0 \) and \( x = \sin^2(kh_x) \), \( 0 \leq x \leq 1 \), as a function of \( x \) we have \( (1-ax)^2 + ax = 1 - ax + a^2x^2 \). The condition \( a \leq 1 \) implies that
\[
1 - ax + a^2x^2 \leq 1.
\]
Thus, we obtain stability under the CFL condition \( c\mu \leq 1 \) or \( h_t/h_x \leq 1/c \).