

**AME 60614: Numerical Methods
Fall 2021**

Problem Set 4

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1 Modified Wavenumber Analysis

$$\begin{aligned}\frac{\partial \phi}{\partial t} &= \alpha \frac{\partial^2 \phi}{\partial x^2} \\ \phi_j &= \psi(t) e^{ikx_j} \\ \frac{d\psi}{dt} &= -\alpha k^2 \psi\end{aligned}$$

Considering the second-order one-sided scheme,

$$\begin{aligned}\frac{d\phi_j}{dt} &= \frac{\alpha}{\Delta x^2} (-\phi_{j+3} + 4\phi_{j+2} - 5\phi_{j+1} + 2\phi_j) \\ &= \frac{\alpha}{\Delta x^2} (-\psi e^{ikx_j} e^{ik3\Delta x} + 4\psi e^{ikx_j} e^{ik2\Delta x} - 5\psi e^{ikx_j} e^{ik\Delta x} + 2\psi e^{ikx_j}) \\ &= \frac{\alpha\phi}{\Delta x^2} (-e^{ik3\Delta x} + 4e^{ik2\Delta x} - 5e^{ik\Delta x} + 2) \\ &= \frac{\alpha\phi}{\Delta x^2} (-\cos 3k\Delta x - i\sin 3k\Delta x + 4\cos 2k\Delta x + 4i\sin 2k\Delta x - 5\cos k\Delta x - 5i\sin k\Delta x + 2) \\ &= \frac{\alpha}{\Delta x^2} [(2 - \cos 3k\Delta x + 4\cos 2k\Delta x - 5\cos k\Delta x) - i(\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x)] \phi \\ &= -\frac{\alpha}{\Delta x^2} [(-2 + \cos 3k\Delta x - 4\cos 2k\Delta x + 5\cos k\Delta x) + i(\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x)] \phi \\ -\alpha k'^2 \phi &= -\frac{\alpha}{\Delta x^2} [(-2 + \cos 3k\Delta x - 4\cos 2k\Delta x + 5\cos k\Delta x) + i(\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x)] \phi \\ -\alpha k'^2 &= -\frac{\alpha}{\Delta x^2} [(-2 + \cos 3k\Delta x - 4\cos 2k\Delta x + 5\cos k\Delta x) + i(\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x)] \\ k'^2 &= \frac{1}{\Delta x^2} [(-2 + \cos 3k\Delta x - 4\cos 2k\Delta x + 5\cos k\Delta x) + i(\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x)] \\ k'^2 \Delta x^2 &= (-2 + \cos 3k\Delta x - 4\cos 2k\Delta x + 5\cos k\Delta x) + i(\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x)\end{aligned}$$

k'^2 is a complex number and it could be written as $k'^2 = k_R'^2 + ik_I'^2$.

$$\begin{aligned}k'^2 &= k_R'^2 + ik_I'^2 \\ k_R'^2 &= \frac{1}{\Delta x^2} (-2 + \cos 3k\Delta x - 4\cos 2k\Delta x + 5\cos k\Delta x) \\ k_I'^2 &= \frac{1}{\Delta x^2} (\sin 3k\Delta x - 4\sin 2k\Delta x - 5\sin k\Delta x) \\ \phi &= \psi_0 e^{-\alpha k_R'^2 t} e^{ik(x - \alpha k_I'^2 t/k)}\end{aligned}$$

To analyze the numerical stability, considering the factor $e^{-\alpha k_R'^2 t}$ for amplitude. If it could be larger than 1, then the numerical solution will blow up, which means numerical instability. So the plot of $k\Delta x - k_R'^2 \Delta x^2$ is shown in Fig. 1.

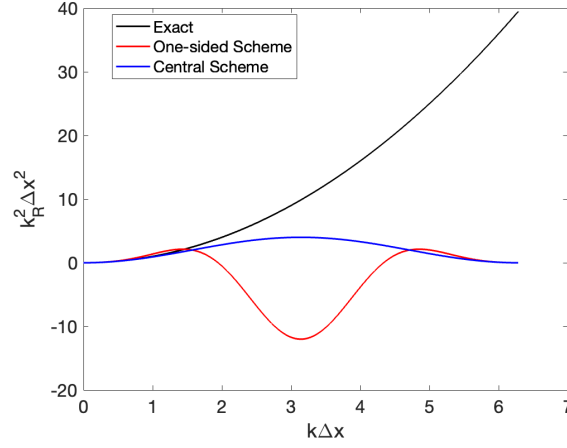


Figure 1: $e^{-\alpha k_R'^2 t}$ plot of one-sided scheme and central scheme.

For central scheme, real component of modified wavenumber is positive, which is the same as exact wavenumber. So the factor would not exceed 1 and the numerical stability is ensured. But for one-sided scheme, real component of modified wavenumber could be negative, making numerical solutions grow unbounded. So the one-sided scheme on the diffusion equation would lead to numerical instability.

2 One-Dimensional Diffusion Equation

2.1 a

2.1.1 Direct Derivation

The time step could still be uniform, so the forward-time scheme,

$$\frac{\partial u}{\partial t} = \frac{u_j^{n+1} - u_j^n}{\Delta t}$$

For non-uniform grid,

$$\begin{aligned} u_{j+1} &= u_j + \frac{\partial u}{\partial x}(x_{j+1} - x_j) + \frac{\partial^2 u}{\partial x^2} \frac{(x_{j+1} - x_j)^2}{2} \\ u_{j-1} &= u_j + \frac{\partial u}{\partial x}(x_{j-1} - x_j) + \frac{\partial^2 u}{\partial x^2} \frac{(x_{j-1} - x_j)^2}{2} \\ \frac{u_{j+1}}{x_{j+1} - x_j} &= \frac{u_j}{x_{j+1} - x_j} + \frac{\partial u}{\partial x} + \frac{\partial^2 u}{\partial x^2} \frac{(x_{j+1} - x_j)}{2} \\ \frac{u_{j-1}}{x_j - x_{j-1}} &= \frac{u_j}{x_j - x_{j-1}} - \frac{\partial u}{\partial x} + \frac{\partial^2 u}{\partial x^2} \frac{(x_{j-1} - x_j)}{2} \\ \frac{u_{j+1}}{x_{j+1} - x_j} + \frac{u_{j-1}}{x_j - x_{j-1}} &= \frac{u_j}{x_{j+1} - x_j} + \frac{u_j}{x_j - x_{j-1}} + \frac{\partial^2 u}{\partial x^2} \frac{(x_{j+1} - x_{j-1})}{2} \\ &\Rightarrow \frac{\partial^2 u}{\partial x^2} = \frac{2}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} u_{j+1} \\ &\quad - \left(\frac{2}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} + \frac{2}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} \right) u_j + \frac{2}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} u_{j-1} \\ &= \frac{2}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} u_{j+1} - \frac{2}{(x_{j+1} - x_j)(x_j - x_{j-1})} u_j + \frac{2}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} u_{j-1} \end{aligned}$$

Combing them for the FTCS scheme for non-uniform grid.

$$\begin{aligned} \frac{\partial u}{\partial t} &= \alpha \frac{\partial^2 u}{\partial x^2} \\ \frac{u_j^{n+1} - u_j^n}{\Delta t} &= \frac{2\alpha}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} u_{j+1} - \frac{2\alpha}{(x_{j+1} - x_j)(x_j - x_{j-1})} u_j + \frac{2\alpha}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} u_{j-1} \\ u_j^{n+1} &= u_j \\ &\quad + \alpha \Delta t \left(\frac{2}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} u_{j+1} - \frac{2}{(x_{j+1} - x_j)(x_j - x_{j-1})} u_j + \frac{2}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} u_{j-1} \right) \end{aligned}$$

Now considering periodic boundary conditions, $u_0 = u_N$. So in practice the point x_0 could be seen as x_N . That is, the left point of x_0 could be x_{N-1} and the right point of x_N could be x_1 . And to keep $u_0 = u_N$, only one of u_0 and u_N will be used in simulation. For convenience of notation, in matlab, u_N is used. In that case, since $x_j = \frac{\tanh(5 \frac{j-N/2}{N/2})}{\tanh 5}$ is an odd function, the dx is symmetric about the grid center. Therefore

$$x_{N+1} - x_N = x_1 - x_0.$$

$$\begin{aligned}\frac{\partial^2 u_1}{\partial x^2} &= \frac{2}{(x_2 - x_0)(x_2 - x_1)} u_2 - \frac{2}{(x_2 - x_1)(x_1 - x_0)} u_1 + \frac{2}{(x_2 - x_0)(x_1 - x_0)} u_0 \\ &= \frac{2}{(x_2 - x_0)(x_2 - x_1)} u_2 - \frac{2}{(x_2 - x_1)(x_1 - x_0)} u_1 + \frac{2}{(x_2 - x_0)(x_1 - x_0)} u_N \\ \frac{\partial^2 u_N}{\partial x^2} &= \frac{2}{(x_{N+1} - x_{N-1})(x_{N+1} - x_N)} u_{N+1} - \frac{2}{(x_{N+1} - x_N)(x_N - x_{N-1})} u_N + \frac{2}{(x_{N+1} - x_{N-1})(x_N - x_{N-1})} u_{N-1} \\ &= \frac{2}{(x_{N+1} - x_{N-1})(x_{N+1} - x_N)} u_1 - \frac{2}{(x_{N+1} - x_N)(x_N - x_{N-1})} u_N + \frac{2}{(x_{N+1} - x_{N-1})(x_N - x_{N-1})} u_{N-1}\end{aligned}$$

$$x_{N+1} - x_N = x_1 - x_0$$

$$x_{N+1} - x_{N-1} = x_1 - x_0 + x_N - x_{N-1} = 2(x_1 - x_0)$$

So the matrix form of the scheme is,

$$\begin{bmatrix} u_1^{n+1} \\ u_2^{n+1} \\ \dots \\ u_{N-1}^{n+1} \\ u_N^{n+1} \end{bmatrix} = I \begin{bmatrix} u_1 \\ u_2 \\ \dots \\ u_{N-1} \\ u_N \end{bmatrix} + 2\alpha\Delta t A$$

Matrix A is shown below,

$$\begin{bmatrix} -\frac{1}{(x_2-x_1)(x_1-x_0)} & \frac{1}{(x_2-x_0)(x_2-x_1)} & \dots & \dots & \frac{1}{(x_2-x_0)(x_1-x_0)} \\ \frac{1}{(x_3-x_1)(x_2-x_1)} & -\frac{1}{(x_3-x_2)(x_2-x_1)} & \frac{1}{(x_3-x_1)(x_3-x_2)} & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & \dots & \frac{1}{(x_N-x_{N-2})(x_{N-1}-x_{N-2})} & -\frac{1}{(x_N-x_{N-1})(x_{N-1}-x_{N-2})} & \frac{1}{(x_N-x_{N-2})(x_N-x_{N-1})} \\ \frac{1}{2(x_1-x_0)^2} & \dots & \dots & \frac{1}{2(x_1-x_0)(x_N-x_{N-1})} & -\frac{1}{(x_1-x_0)(x_N-x_{N-1})} \end{bmatrix}$$

With $p2_1.m$, eigenvalues of matrix A are calculated numerically.

$$\frac{\lambda_{max}}{\lambda_{min}} = \frac{7.0991359e + 9}{5.4662397e - 11} \approx 1.2987239e + 20$$

2.1.2 Coordinate Transformation

Considering coordinate transformation,

$$\begin{aligned}\zeta &= 5 \frac{j - N/2}{N/2} \\ x &= \frac{\tanh \zeta}{\tanh 5} \\ \Rightarrow \zeta &= \tanh^{-1}(x \tanh 5) = g(x)\end{aligned}$$

With chain rule, derivatives under new coordinate system could be derived.

$$\begin{aligned}\frac{du}{dx} &= \frac{d\zeta}{dx} \frac{du}{d\zeta} = g' \frac{du}{d\zeta} \\ \frac{d^2u}{dx^2} &= \frac{d}{dx} \left[g' \frac{du}{d\zeta} \right] = g'' \frac{du}{d\zeta} + (g')^2 \frac{d^2u}{d\zeta^2}\end{aligned}$$

Substituting ζ and g and taking $\tanh 5 = a$.

$$\begin{aligned}
 g' &= \frac{a}{1 - a^2 x^2} \\
 g'' &= \frac{2a^3 x}{(1 - a^2 x^2)^2} \\
 \frac{du}{dx} &= \frac{a}{1 - a^2 x^2} \frac{u_{j+1} - u_{j-1}}{2\Delta\zeta} \\
 \frac{d^2 u}{dx^2} &= \frac{2a^3 x}{(1 - a^2 x^2)^2} \frac{u_{j+1} - u_{j-1}}{2\Delta\zeta} + \left(\frac{a}{1 - a^2 x^2} \right)^2 \frac{u_{j+1} - 2u_j + u_{j-1}}{\Delta\zeta^2} \\
 &= \frac{a^3 x}{(1 - a^2 x^2)^2} \frac{u_{j+1} - u_{j-1}}{\Delta\zeta} + \frac{a^2}{(1 - a^2 x^2)^2} \frac{u_{j+1} - 2u_j + u_{j-1}}{\Delta\zeta^2} \\
 &= \frac{a^2}{(1 - a^2 x^2)^2 \Delta\zeta} \left[\left(\frac{1}{\Delta\zeta} - ax \right) u_{j-1} - \frac{2}{\Delta\zeta} u_j + \left(\frac{1}{\Delta\zeta} + ax \right) u_{j+1} \right]
 \end{aligned}$$

Therefore A is still a tridiagonal matrix. Since $1 - a^2 x^2$ is changing with j , the factor of A is $\frac{\alpha a^2 \Delta t}{\Delta\zeta}$.

$\frac{1}{(1 - a^2 x^2)^2} \left(\frac{1}{\Delta\zeta} - ax \right)$, $-\frac{1}{(1 - a^2 x^2)^2} \frac{2}{\Delta\zeta}$, $\frac{1}{(1 - a^2 x^2)^2} \left(\frac{1}{\Delta\zeta} + ax \right)$ construct the three diagonals of A . With $p2_2.m$, eigenvalues of matrix A are calculated numerically.

$$\frac{\lambda_{max}}{\lambda_{min}} = \frac{1.2031589e + 9}{3.7371737e - 11} \approx 3.2194353e + 19$$

Therefore, for both methods, the system is stiff.

2.2 b

For FTCS scheme, Explicit Euler is used for time forward. Then the stability limit of EE could be recalled.

$$\Delta t_{max} = \frac{2}{\alpha |\lambda_{max}|} \Rightarrow \alpha \Delta t_{max} = \frac{2}{|\lambda_{max}|}$$

By $p2_3.m$, the $\alpha \Delta t_{max}$ from the two methods are shown in Fig. 2.

Through coordinate transformation, the required Δt_{max} is larger than direct derivation.

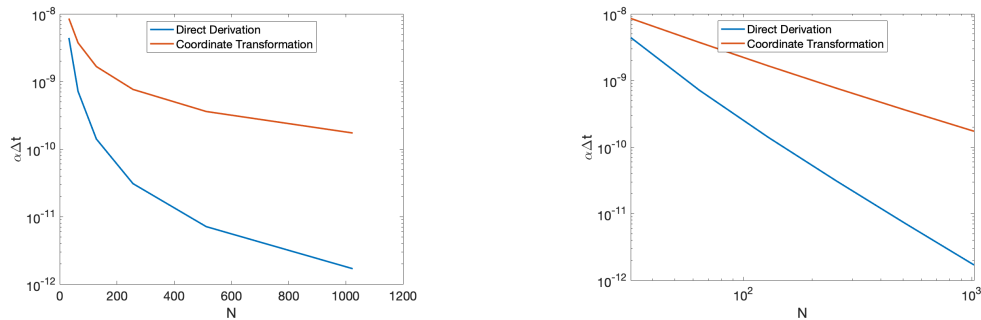


Figure 2: N - $\alpha \Delta t_{max}$ plot of two methods.

2.3 c

Considering the direct derivation,

$$\begin{aligned}
u_j^{n+1} &= u_j \\
&+ \alpha \Delta t \left(\frac{2}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} u_{j+1} - \frac{2}{(x_{j+1} - x_j)(x_j - x_{j-1})} u_j + \frac{2}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} u_{j-1} \right) \\
\sigma &= 1 \\
&+ \alpha \Delta t \left(\frac{2}{(x_{j+1} - x_{j-1})(x_{j+1} - x_j)} e^{ikx_{j+1}} - \frac{2}{(x_{j+1} - x_j)(x_j - x_{j-1})} + \frac{2}{(x_{j+1} - x_{j-1})(x_j - x_{j-1})} e^{ikx_{j-1}} \right) \\
|\sigma| &< \left| 1 + \alpha \Delta t \left(\frac{2}{2\Delta x_{min}^2} e^{ikx_{j+1}} - \frac{2}{\Delta x_{min}^2} + \frac{2}{2\Delta x_{min}^2} e^{ikx_{j-1}} \right) \right| \\
|\sigma| &< \left| 1 + \frac{2\alpha \Delta t}{\Delta x_{min}^2} (\cos \Delta x - 1) \right| \\
\Rightarrow \alpha \Delta t &\leq \frac{\Delta x_{min}^2}{2}
\end{aligned}$$

Considering the coordinate transformation method,

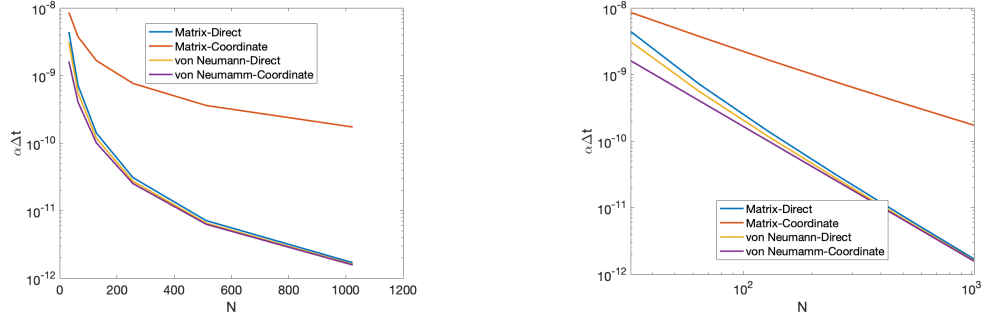
$$\begin{aligned}
u_j^{n+1} &= u_j^n + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta} \left[\left(\frac{1}{\Delta \zeta} - ax \right) u_{j-1} - \frac{2}{\Delta \zeta} u_j + \left(\frac{1}{\Delta \zeta} + ax \right) u_{j+1} \right] \\
\sigma^{n+1} e^{ik\zeta_j} &= \sigma^n e^{ik\zeta_j} + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta} \left[\left(\frac{1}{\Delta \zeta} - ax \right) \sigma^n e^{ik\zeta_{j-1}} - \frac{2}{\Delta \zeta} \sigma^n e^{ik\zeta_j} + \left(\frac{1}{\Delta \zeta} + ax \right) \sigma^n e^{ik\zeta_{j+1}} \right] \\
\sigma &= 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta} \left[\left(\frac{1}{\Delta \zeta} - ax \right) e^{-ik\Delta \zeta} - \frac{2}{\Delta \zeta} + \left(\frac{1}{\Delta \zeta} + ax \right) e^{ik\Delta \zeta} \right] \\
|\sigma| &= \left| 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta} \left[\left(\frac{1}{\Delta \zeta} - ax \right) (\cos \Delta \zeta - i \sin \Delta \zeta) - \frac{2}{\Delta \zeta} + \left(\frac{1}{\Delta \zeta} + ax \right) (\cos \Delta \zeta + i \sin \Delta \zeta) \right] \right| \\
|\sigma| &= \left| 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta^2} [(1 - ax\Delta \zeta) (\cos \Delta \zeta - i \sin \Delta \zeta) - 2 + (1 + ax\Delta \zeta) (\cos \Delta \zeta + i \sin \Delta \zeta)] \right|
\end{aligned}$$

$\Delta \zeta$ is small, so $|\sigma|$ could be approximated and simplified.

$$\begin{aligned}
|\sigma| &\approx \left| 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta^2} [(\cos \Delta \zeta - i \sin \Delta \zeta) - 2 + (\cos \Delta \zeta + i \sin \Delta \zeta)] \right| \\
|\sigma| &= \left| 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2 x^2)^2 \Delta \zeta^2} [2 \cos \Delta \zeta - 2] \right|
\end{aligned}$$

Obviously when $|x| = 1$, $|\sigma|$ would be the greatest.

$$\begin{aligned}
|\sigma| &= \left| 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2)^2 \Delta \zeta^2} [2 \cos \Delta \zeta - 2] \right| < 1 \\
-1 &< 1 + \frac{a^2 \alpha \Delta t}{(1 - a^2)^2 \Delta \zeta^2} [2 \cos \Delta \zeta - 2] < 1 \\
\frac{a^2 \alpha \Delta t}{(1 - a^2)^2 \Delta \zeta^2} &< \frac{1}{1 - \cos \Delta \zeta} \leq \frac{1}{2} \\
\alpha \Delta t &\leq \frac{(1 - a^2)^2 \Delta \zeta^2}{2a^2}
\end{aligned}$$

Figure 3: N - $\alpha\Delta t_{max}$ plot of four methods.

By $p2_4.m$, the $\alpha\Delta t_{max}$ from the four methods are shown in Fig. 3.

Since the idea of approximation is similar, $\alpha\Delta t_{max}$ from the two von Neumann stability are close. But under the amplification effects of approximation, $\alpha\Delta t_{max}$ from von Neumann stability are smaller than $\alpha\Delta t_{max}$ from matrix stability. And for direct derivation, the $|\lambda_{max}|$ comes from Δx_{min} closely, so the $\alpha\Delta t_{max}$ from von Neumann stability are also close to $\alpha\Delta t_{max}$ from matrix stability of direct derivation, which is only slightly larger.

3 Multi-Dimensional Convection Equation

3.1 a

To retain the initial energy, when there is no dissipation in the convection equation, that means the dissipation error should be eliminated. So modified wavenumber stability analysis is used to analyze the dissipation. Since same scheme will be used on different spatial directions, considering one-dimensional case is enough for modified wavenumber stability analysis.

$$\begin{aligned}
 u &= \psi_0 e^{ik(x-ct)} \\
 \frac{\partial u}{\partial t} &= -c \frac{\partial u}{\partial x} = -c \\
 u &= \psi_0 e^{ck'_I t} e^{ik\left(x - \frac{k'_R}{k} ct\right)}
 \end{aligned}$$

The factor $e^{ck'_I t}$ will lead to dissipation error. So the spatial scheme should not have imaginary component in the modified wavenumber. For example, central-difference scheme do not have such term. In that problem there is no other need for higher-order spatial scheme. So second-order central difference scheme for spatial difference is enough.

For choice of time scheme, here is a new requirement. To retain initial energy, dissipation error is eliminated, there will be only dispersion error in the scheme, which means purely imaginary eigenvalues. So naturally, the time scheme chosen must has imaginary-axis stability. That is why Explicit Euler is unconditional unstable for convection equation.

To compare discretization and maximum time step, von Neumann stability analysis is used based on second-

order central difference scheme. Taking $\Delta x = \Delta y = h$. For explicit methods,

$$\begin{aligned}\frac{\partial u}{\partial t} &= -\frac{1}{2h}(u_{i+1,j} - u_{i-1,j}) - \frac{1}{h}(u_{i,j+1} - u_{i,j-1}) \\ \Delta t &\leq \frac{2.83}{|\lambda_{max}|} = \frac{2.83}{|\frac{1}{h}i \cos(\frac{\pi i}{N}) + \frac{2}{h}i \cos(\frac{\pi j}{N})|} \leq \frac{2.83h}{3} \text{ for RK4} \\ (c_x + c_y) \frac{\Delta t}{h} &\leq 1 \Rightarrow \Delta t \leq \frac{h}{3} \text{ for Lax-Wendroff}\end{aligned}$$

From cost and Δt_{max} , RK4 will be better than Lax-Wendroff.

For Implicit Euler,

$$\begin{aligned}\frac{u_j^{n+1} - u_j^n}{\Delta t} &= -\frac{1}{2h}(u_{i+1,j}^{n+1} - u_{i-1,j}^{n+1}) - \frac{1}{h}(u_{i,j+1}^{n+1} - u_{i,j-1}^{n+1}) \\ u_j^n &= u_j^{n+1} + \frac{\Delta t}{2h}(u_{i+1,j}^{n+1} - u_{i-1,j}^{n+1}) + \frac{\Delta t}{h}(u_{i,j+1}^{n+1} - u_{i,j-1}^{n+1}) \\ 1 &= \sigma + \frac{3\Delta t}{2h}\sigma(e^{ikh} - e^{-ikh}) \\ 1 &= \sigma + \frac{3\Delta t}{h}i\sigma \sin kh \\ \sigma &= \frac{1}{1 + i\frac{3\Delta t}{h}\sin kh}\end{aligned}$$

Implicit Euler is unconditionally stable, but since $|\sigma| < 1$, even though central difference eliminates dissipation, implicit euler will import extra dissipation.

For trapizoid method, or so-called Crank-Nicolson method in diffusion equation,

$$\begin{aligned}\frac{u_j^{n+1} - u_j^n}{\Delta t} &= -\frac{1}{4h}(u_{i+1,j}^{n+1} - u_{i-1,j}^{n+1}) - \frac{1}{4h}(u_{i+1,j}^n - u_{i-1,j}^n) - \frac{1}{2h}(u_{i,j+1}^{n+1} - u_{i,j-1}^{n+1}) - \frac{1}{2h}(u_{i,j+1}^n - u_{i,j-1}^n) \\ \frac{\sigma - 1}{\Delta t} &= -\frac{3}{4h}\sigma(e^{ikh} - e^{-ikh}) - \frac{3}{4h}(e^{ikh} - e^{-ikh}) \\ \sigma &= \frac{1 - \frac{3\Delta t}{2h}i \sin kh}{1 + \frac{3\Delta t}{2h}i \sin kh}\end{aligned}$$

For trapizoid, $|\sigma| = 1$. So trapizoid method with central difference scheme will not import any dissipation and it is unconditionally stable. In summary, central difference with trapizoid method, or Crank-Nicolson method should be the best choice. Second-order accuracy in time and space will not be too expensive and it will not import any dissipation. So trapizoid is better than implicit euler.

For different schemes, than bad methods on fine grid should be faster good numerical methods on coarse grid. Generally, better scheme brings larger computation cost, compared to the tolerance it brings. So when the two schemes both meet the requirements, scheme with not such high order of accuracy will be more practical to implement.

3.2 b

In that problem, one cycle time could be defined as $\frac{2\pi}{c_{min}} = 2\pi$. So ten cycles actually requires total time of 20π . And 2-D implicit method is expensive and hard to implement, firstly RK4 is used in this problem. With $p3_1.m$, the energy reduces to less than 99% quickly. So explicit scheme also imports dissipation.

Therefore the trapizoid method is the only choice to retain such initial energy for such a long cycle time. In that case, the solution after ten cycles should be very close to the initial condition. The scheme is implemented with *p3_2.m*. Since it is unconditionally stable, the time step could be not so small. Considering the huge cost for solving huge matrices, the time step is firstly taken 0.05. Then I changed it to 0.01.

To implement such method with 2-D dimension, I took a cubersome method. The 2-D u_{ij} is streched to a one dimensional vector, from a 512×512 matriix to a 512^2 vector. Then the coefficient matrix becomes $512^2 \times 512^2$ and is something like penta-diagonal matrix with some modification because of periodic conditions.

$$\begin{aligned} \frac{u_j^{n+1} - u_j^n}{\Delta t} &= -\frac{1}{4h} (u_{i+1,j}^{n+1} - u_{i-1,j}^{n+1}) - \frac{1}{4h} (u_{i+1,j}^n - u_{i-1,j}^n) - \frac{1}{2h} (u_{i,j+1}^{n+1} - u_{i,j-1}^{n+1}) - \frac{1}{2h} (u_{i,j+1}^n - u_{i,j-1}^n) \\ u_j^{n+1} + \frac{1}{4h} (u_{i+1,j}^{n+1} - u_{i-1,j}^{n+1}) + \frac{1}{2h} (u_{i,j+1}^{n+1} - u_{i,j-1}^{n+1}) &= u_j^n - \frac{1}{4h} (u_{i+1,j}^n - u_{i-1,j}^n) - \frac{1}{2h} (u_{i,j+1}^n - u_{i,j-1}^n) \\ &\Rightarrow Au^{n+1} = Bu^n \end{aligned}$$

A and B are both matrices mentioned above, something like penta-diagonal matrix with some modification because of periodic conditions. After those are established, simulation could be processed by time steps. From *p3_2.m*, the code took 4073 seconds for $\Delta t = 0.05$ and 14371 second for $\Delta t = 0.01$. Both of them retained 99.99999999999999% of initial energy. The output of Matlab is shown in Fig. 4.

```
Command Window
Elapsed time is 4073.047486 seconds.
2.979106174591453

2.979106174591453

0.9999999999999999

4073.1029
fx >>

Command Window
Elapsed time is 14370.847783 seconds.
2.979106174591453

2.979106174591451

0.9999999999999992

14370.8788
fx >>
```

Figure 4: Matlab output of trapizoid method with central difference scheme.