
CS 714

HOMEWORK 3

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<https://github.com/Oafish1/CSC-714>

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1 Problem A

1.1 Part a

1.1.1 Part i

Let's take the derivative of $u(x(t), t)$ with respect to t . We get

$$\frac{d}{dt}u(x(t), t) = \partial_t u(x(t), t) + \partial_x u(x(t), t)x'(t)$$

Since u is constant in the characteristic curve, we know $\frac{d}{dt}u(x(t), t) = 0$. Thus,

$$\partial_t u(x(t), t) + \partial_x u(x(t), t)x'(t) = 0$$

1.1.2 Part ii

Using the advection equation and our result from Part i, we have

$$\begin{aligned}\partial_t u(x(t), t) + a\partial_x u(x(t), t) &= 0 \\ \partial_t u(x(t), t) + \partial_x u(x(t), t)x'(t) &= 0\end{aligned}$$

Then, subtracting the two equations

$$\begin{aligned}x'(t) - a &= 0 \\ x'(t) &= a\end{aligned}$$

1.1.3 Part iii

We now have

$$\begin{aligned}\frac{d}{dt}u(x(t), t) &= 0 \\ u(x, 0) &= u_0(x)\end{aligned}$$

So, $u(x(t), t)$ does not change with t and

$$u(x(t), t) = u_0(x_0)$$

1.1.4 Part iv

We now know the solution to the function along the characteristic curve for any x_0 . So, if we can solve for x_0 , we can solve for $u(x, t) = u_0(x_0)$.

We know that, for the advection equation, $a = x'(t)$ is a constant. So, $x_0 = x - at$ and

$$u(x, t) = u_0(x - at)$$

1.2 Part b

1.2.1 Part i

In this case, $u_0(x)$ should be absolutely or square integrable, L^1 and/or L^2 .

As is common with functions involving variable powers of e , we have regularity $\hat{u}_0(x) \in \mathcal{C}^\infty$.

As the magnitude of ξ approaches infinity, if $u_0(x)$ is absolutely integrable, $\hat{u}(\xi) \rightarrow 0$ by the *Riemann-Lebesgue Lemma*. If $u_0(x) \in L^2$, $\hat{u}_0(\xi) \rightarrow 0$ as $\xi \rightarrow \infty$ as well.

1.2.2 Part ii

We have

$$v(x) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} \hat{v}(\xi) d\xi$$

We can take the derivative

$$\begin{aligned} v'(x) &= \frac{1}{2\pi} \int_{\mathbb{R}} \left(\frac{d}{dx} e^{i\xi x} \right) \hat{v}(\xi) + e^{i\xi x} \left(\frac{d}{dx} \hat{v}(\xi) \right) d\xi \\ &= \frac{1}{2\pi} \int_{\mathbb{R}} i\xi e^{i\xi x} \hat{v}(\xi) d\xi \end{aligned}$$

Thus,

$$v'(x) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} i\xi \hat{v}(\xi) d\xi$$

1.2.3 Part iii

We can say

$$\begin{aligned} \mathcal{F}[v(x+h)](\xi) &= \int_{\mathbb{R}} e^{-i\xi(x+h)} v(x+h) dx \\ &= e^{-i\xi h} \int_{\mathbb{R}} e^{-i\xi x} v(x+h) dx \\ &= e^{-i\xi h} \hat{v}(\xi) \end{aligned}$$

The last step above is due to the integration. Since the function is evaluated for \mathbb{R} , translation does not affect the result.

Thus,

$$\mathcal{F}[v(\cdot + h)](\xi) = e^{-i\xi h} \hat{v}(\xi)$$

1.2.4 Part iv

Define

$$\hat{u}(\xi, t) = \int_{\mathbb{R}} e^{-i\xi x} u(x, t) dx$$

Then,

$$u(x, t) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} \hat{u}(\xi, t) d\xi$$

Using the same method as in Part ii, we can obtain

$$\partial_x u(x, t) = \frac{1}{2\pi} \int_{\mathbb{R}} i\xi e^{i\xi x} \hat{u}(\xi, t) d\xi$$

Note $\partial_t u(x, t) = -a \partial_x u(x, t)$. Then,

$$\partial_t u(x, t) = \frac{-1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} a i \xi \hat{u}(\xi, t) d\xi$$

Inverting this, by our definition,

$$\mathcal{F}[\partial_t u(x, t)](\xi, t) = -ai\xi \hat{u}(\xi, t)$$

Notice $\mathcal{F}[\partial_x v(x)](\xi) = \partial_x \hat{v}(x)(\xi)$.

Thus,

$$\partial_t \hat{u}(\xi, t) = -ai\xi \hat{u}(\xi, t)$$

1.2.5 Part v

Since a is constant, we know $x = x_0 + at$. Then, by Part iii, we can say

$$\begin{aligned}\mathcal{F}[u(x_0 + at)](\xi) &= e^{-ai\xi t} \mathcal{F}[u(x_0)](\xi) \\ \hat{u}(\xi, t) &= e^{-ai\xi t} \hat{u}_0(\xi)\end{aligned}$$

1.2.6 Part vi

If $u_0(\xi) \in L^2(\mathbb{R})$, we can determine $\hat{u}(\xi, t) = e^{-ai\xi t} \hat{u}_0(\xi)$ for any fixed t . Notice $\hat{u}(\xi, t) \in L^2(\mathbb{R})$.

Then, by the definition of Lebesgue spaces,

$$\int_{\mathbb{R}} |\hat{u}(\xi, t)|^2 d\xi < \infty$$

By Plancherel's theorem, we then know

$$\int_{\mathbb{R}} |u(x, t)|^2 dx < \infty$$

1.2.7 Part vii

We have

$$\frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi(x-at)} \hat{u}_0(\xi) d\xi$$

Which we can simplify using our result in Part v

$$\begin{aligned}\frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi(x-at)} \hat{u}_0(\xi) d\xi &= \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} e^{-i\xi at} \hat{u}_0(\xi) d\xi \\ &= \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi x} \hat{u}(\xi, t) d\xi \\ &= u(x, t)\end{aligned}$$

Thus,

$$u(x, t) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi(x-at)} \hat{u}_0(\xi) d\xi$$

1.2.8 Part iix

We can see, by our definition of the Fourier Transform,

$$\frac{1}{2\pi} \int_{\mathbb{R}} e^{i\xi(x-at)} \hat{u}_0(\xi) d\xi = u_0(x - at)$$

Then, from our result in Part vii,

$$u(x, t) = u_0(x - at)$$

2 Problem B

2.1 Part a

We are given

$$u_t = \pm u_{xxxx}$$

Which provides

$$\hat{u}_t(\xi, t) = \pm i^4 \xi^4 \hat{u}(\xi, t) = \pm \xi^4 \hat{u}(\xi, t)$$

Notice that, if we choose +, the magnitude for frequency ξ will continue to grow over time for all $|\xi| > 1$. So, we need to choose − for stable solutions. For an example, look below.

2.1.1 Example

Consider the Explicit Euler

$$\hat{u}(\xi, t + \Delta t) = \hat{u}(\xi, t) \pm h \xi^4 \hat{u}(\xi, t)$$

Then,

$$\hat{u}(\xi, t + \Delta t) = (1 \pm \Delta t \xi^4) \hat{u}(\xi, t)$$

Therefore, we want $|1 \pm \Delta t \xi^4| \leq 1$ for stability. We know $\Delta t \xi^4 \geq 0$ for all ξ . Thus, we want to choose $u_t = -u_{xxxx}$ as our PDE.

Keep in mind that it is impossible to have $|1 \pm \Delta t \xi^4| \leq 1$ for all ξ as ξ is unbounded by nature. However, also keep in mind that u_{xxxx} is not discretized here and that higher frequencies are often ignored for high Δx . Then, when Δx shrinks, so should Δt so that $|1 \pm \Delta t \xi^4| \leq 1$ for relevant ξ . For an illustration of this, take a look at Part b and Part c which document the relationship in more detail.

2.2 Part b

With the Explicit Euler Method, we acquire

$$u^{t+1} = u^t - (\Delta t) u_{xxxx}$$

For some finite difference discretization, we will obtain

$$U^{t+1} = U^t - \left(\frac{\Delta t}{\Delta x^4} \right) A U^t$$

Which becomes

$$U^{t+1} = \left(I - \left(\frac{\Delta t}{\Delta x^4} \right) A \right) U^t$$

$$U^t = B^t U^0$$

Suppose B has eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ and corresponding eigenvectors v_1, v_2, \dots, v_n that form a basis.

Then, given $E^t = B^t E^0$ from above,

$$E^0 = \sum_i^n c_i v_i$$

$$E^t \leq \rho(B)^t E^0$$

We then want $\rho(B) \leq 1$. Note that the eigenvalues of B are the eigenvalues of A scaled by $-\left(\frac{\Delta t}{\Delta x^4}\right)$ plus 1. Say A has minimal and maximal eigenvalues γ, Γ . Then

$$\rho(B) \leq 1$$

$$\left| 1 - \frac{\Delta t}{\Delta x^4} \gamma \right|, \left| 1 - \frac{\Delta t}{\Delta x^4} \Gamma \right| \leq 1$$

$$0 \leq \frac{\Delta t}{\Delta x^4} \gamma, \frac{\Delta t}{\Delta x^4} \Gamma \leq 2$$

Which implies

$$\frac{\Delta t}{\Delta x^4} \leq \min\left(\frac{2}{\gamma}, \frac{2}{\Gamma}\right) \quad (1)$$

Note, for either $\gamma, \Gamma = 0$, the condition on γ, Γ is met. Thus, 1 implies stability. Notice that this is actually equivalent to the CFL condition.

2.3 Part c

Given

$$\frac{\Delta t}{\Delta x^4} \leq \min\left(\frac{2}{\gamma}, \frac{2}{\Gamma}\right)$$

We know, in general, if we scale Δx by q , then we need to scale Δt by q^4 ; this is impractical.

To remedy this, we can use a method with a larger stability region to make the stability condition more lenient. Examples include the Trapezoidal Rule and Backward Euler.

3 Problem C

3.1 Part a

For the 2D wave equation discretized with a second-order accurate scheme in time alone, we have

$$\begin{aligned} u_{tt} &= \Delta u \\ u^{t+1} - 2u^t + u^{t-1} &= h_t^2 \Delta u^t \\ (u + u_t h_t + u_{tt} \frac{h_t^2}{2} + \dots) - 2u + (u - u_t h_t + u_{tt} \frac{h_t^2}{2} + \dots) &= h_t^2 \Delta u \\ u_{tt} + u_{tttt} \frac{h_t^2}{12} + O(h_t^4) &= \Delta u \end{aligned}$$

For u sufficiently smooth we can say (Piazza @59)

$$\begin{aligned} u_{tttt} &= \partial_{tt} u_{tt} \\ &= \partial_{tt} (u_{xx} + u_{yy}) \\ &= (\partial_{xx} + \partial_{yy}) u_{tt} \\ &= (\partial_{xx} + \partial_{yy}) \Delta u \\ &= \Delta^2 u \end{aligned}$$

Then,

$$u_{tt} + \Delta^2 u \frac{h_t^2}{12} + O(h_t^4) = \Delta u$$

So,

$$u_{tt} = \Delta u + \Delta^2 u \frac{h_t^2}{12}$$

is our modified equation. By adding $\Delta^2 u \frac{h_t^2}{12}$ to the right-hand side of the equation, we are making our schemes fourth-order accurate. While we will be using a Chebyshev grid for the actual implementation, the general process that shows the accuracy of the scheme can be seen in the below example.

3.1.1 Example

Using a second-order time discretization on our new PDE, we will have

$$\begin{aligned} u_{tt} + \Delta^2 u \frac{h_t^2}{12} + O(h_t^4) &= \Delta u + \Delta^2 u \frac{h_t^2}{12} \\ u_{tt} + O(h_t^4) &= \Delta u \end{aligned}$$

Then, for some approximation of Δu and $\Delta^2 u$, we can come up with the method

$$\begin{aligned}\hat{u}_{tt} &= \hat{\Delta} u + \frac{h_t^2}{12} \hat{\Delta}^2 u \\ h_t^2 \hat{u}_{tt} &= h_t^2 \hat{\Delta} u + \frac{h_t^4}{12} \hat{\Delta}^2 u \\ h_t^2 \hat{u}_{tt} &= h_t^2 \hat{\Delta} u + \frac{h_t^4}{12} \hat{\Delta}^2 u\end{aligned}$$

To use this, suppose $h_t^2 \hat{u}_{tt} = u^{t+1} - 2u^t + u^{t-1}$. Solving for u^{t+1} provides the method that will be used in Part c,

$$u^{t+1} = 2u^t - u^{t-1} + h_t^2 \hat{\Delta} u + \frac{h_t^4}{12} \hat{\Delta}^2 u$$

3.2 Part b

We can create a fourth-order forward difference approximation of u^t for $t = 0$ and solve for u^1 .

Let's start by solving a five-point stencil.

$$\frac{\alpha_0 u_0 + \alpha_1 u_1 + \alpha_2 u_2 + \alpha_3 u_3 + \alpha_4 u_4}{\Delta t}$$

Looking at the Taylor Series we get

$$u_t = u_0 + u'_0(t\Delta t) + u''_0 \frac{(t\Delta t)^2}{2} + u'''_0 \frac{(t\Delta t)^3}{6} + u''''_0 \frac{(t\Delta t)^4}{24} + O(\Delta t^5)$$

We want

$$u'_0(\Delta t) + O(\Delta t^5)$$

So, we then need to solve

$$\begin{aligned}\alpha_0 + \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 &= 0 \\ \alpha_1 + 2\alpha_2 + 3\alpha_3 + 4\alpha_4 &= 1 \\ \alpha_1 + 4\alpha_2 + 9\alpha_3 + 16\alpha_4 &= 0 \\ \alpha_1 + 8\alpha_2 + 27\alpha_3 + 64\alpha_4 &= 0 \\ \alpha_1 + 16\alpha_2 + 81\alpha_3 + 256\alpha_4 &= 0\end{aligned}$$

Solving this provides

$$\alpha_0, \alpha_1, \alpha_2, \alpha_3, \alpha_4 = -\frac{25}{12}, 4, -3, \frac{4}{3}, -\frac{1}{4}$$

We can then solve for $t = 1$ by solving for u_1 in $\frac{\partial}{\partial t} u(x, y, 0) = f(x)f(y)$ using our approximation.

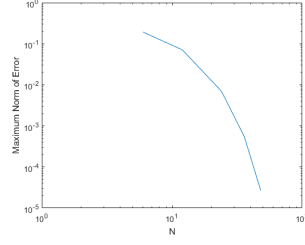
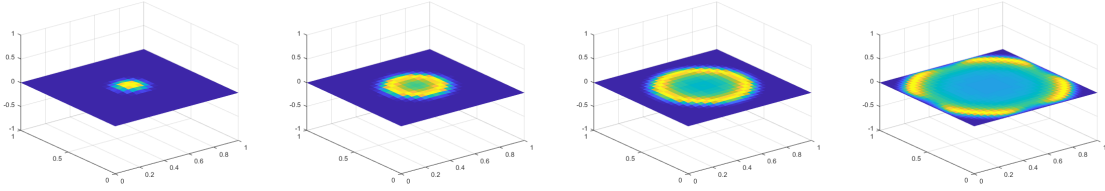
$$u_1 = \frac{1}{4} \left(\frac{25}{12} u_0 + 3u_2 - \frac{4}{3} u_3 + \frac{1}{4} u_4 + f(x)f(y)\Delta t \right)$$

Implementing this will involve initializing a 3D matrix with all zeros, applying the above, then applying our scheme from Part a. This will be repeated until convergence.

3.3 Part c

In solving our modified PDE, using a Chebyshev grid, we can use a differentiation matrix [1] or spectral differentiation [2]. In this case, the results are indistinguishable. The methods used for computing second derivatives are taken from the cited chapters.

The log-log plot of N vs Max Norm of the Error can be seen in figure 1. This does appear to be at least fourth-order accurate, if not better. Keep in mind that, in this implementation, $h_t \propto h^2$. Some samples can also be seen in figure 2.

Figure 1: N vs Max Norm of the Error. Note that the lesser N are very small (*Spectral_Heat.m*).

 Figure 2: Samples for $N = 36$ at $t = .16, .39, .63, 1.18$


3.4 Part d

From the general equation

$$y''(t) = \gamma y$$

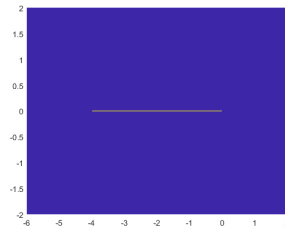
We were able to determine in HW2 that

$$\pi(\xi, z) = \rho(\xi) - \lambda(\Delta t)^2 \sigma(\xi) = \xi^2 - (2 + z)\xi + 1 \text{ for } z = \lambda(\Delta t)^2$$

Solving for ξ , we determined

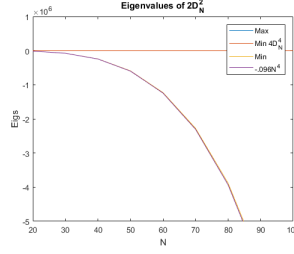
$$\xi = \frac{2 + z \pm \sqrt{z^2 + 4z}}{2}$$

And $|\xi| \leq 1$, $|\xi| < 1$ for non-simple roots, implies absolute stability. This lends the region shown in figure 3.

 Figure 3: Region of absolute stability on the complex plane (*Stability.m*)


So, $\lambda(\Delta t)^2 \in (-4, 0) \in \mathbb{R}$. Notice the open interval. This is because the boundary values of z produce identical, non-simple roots.

The eigenvalues of our second differentiation matrix are strictly negative and real with the largest (in magnitude) being proportional to N^4 . In particular, $-.048N^4$ [3]. However, note that we are actually using this matrix twice in calculating the laplacian, so we have $\approx -.096N^4$ as our maximal eigenvalue. Keep in mind that, in calculating the Biharmonic/Square Laplacian, we are also using \hat{D}_N^4 which has negligible eigenvalues in comparison, even though it is used four times per calculation. All of this can be seen in figure 4.

Figure 4: N vs Minimal and Maximal Eigenvalues (*Cheb_Eig.m*). Max and $4D_N^4$ overlap on the plot.


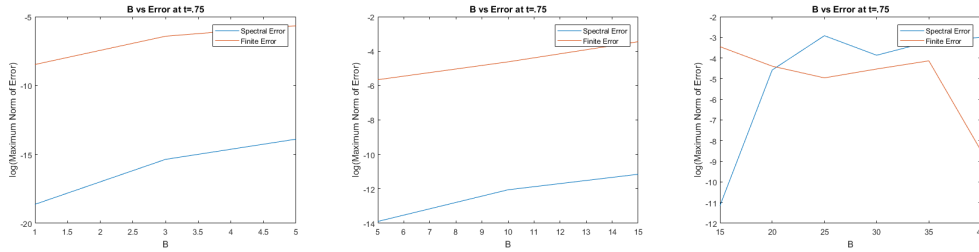
Combining this with our previous calculation provides our CFL condition.

$$-.096N^4(\Delta t)^2 \gtrsim -4$$

$$\frac{\Delta t}{(\Delta x, y)^2} \lesssim 6.45$$

3.5 Part e

The comparison was performed with $N = 32$ and $N_{\text{fine}} = 64$. Each method was compared to its respective fine counterpart for every given B . If this was not done, the program might favor one algorithm's accuracy over the other, even if interpolation is used. The results can be seen in figure 5

 Figure 5: B vs Maximal Error (*Comparison.m*)


Spectral accuracy was noticeably affected by the change in B . This makes sense, since, for high frequency sin waves, the lack of coverage in the middle of the graph when using a chebyshev grid could be a problem – especially considering the max norm. FD accuracy became unstable for larger B , perhaps this is a result of N_{fine} needing to be larger than for spectral accuracy.

The graph suggests that, for accuracy of 10^{-3} , spectral methods require roughly $\frac{N}{25} \approx 1.28$ average points per wavelength while FD methods require $\frac{N}{15} \approx 2.14$.

4 Problem D

To see why this is true, consider

$$|\hat{f}(\xi)| = \|f\|_{TV} |\xi|^{-1}$$

for some frequency ξ . Then, the magnitude of ξ is equivalent to the total variance of the function divided by the frequency ξ . So, ξ accounts for all the variance in f .

Now, consider

$$|\hat{f}(\xi)| \geq \|f\|_{TV} |\xi|^{-1}$$

This then implies that the variance in ξ exceeds that in f . This is impossible since the wave with frequency ξ is a component of f by the definition of the Fourier Transform. So,

$$|\hat{f}(\xi)| \leq \|f\|_{TV} |\xi|^{-1}$$

References

- [1] *Chebyshev Differentiation Matrices*, chapter 6, pages 51–59.
- [2] *Chebyshev Series and the FFT*, chapter 8, pages 75–86.
- [3] *Time-Stepping and Stability Regions*, chapter 10, pages 101–114.