

Computational Physics

PHYS 6260

Solving Partial differential equations (PDEs)

Announcements:

- HW3: Due Friday 1/31

We will cover these topics

- Relaxation method

Faster methods

Over-relaxation

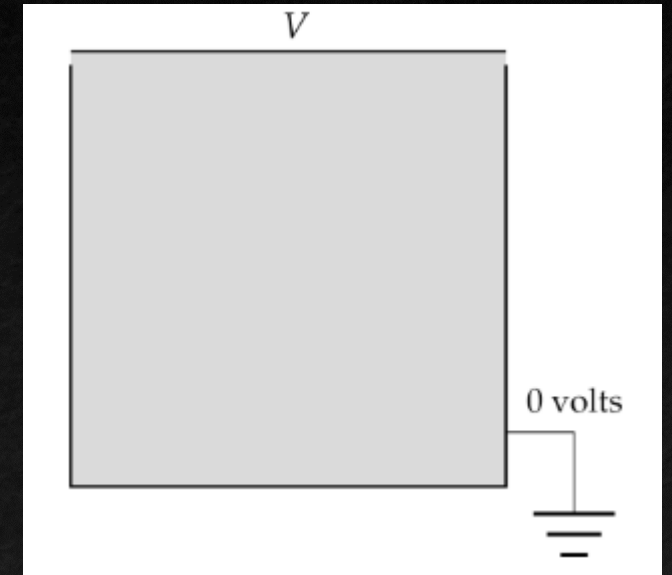
Gauss-Seidel method

Initial value problems

Lecture Outline

Introduction

- Some of the most interesting physics problems involve PDEs
 - Wave eqn, heat eqn, diffusion eqn, Laplace/Poisson eqn, Maxwell's eqns, Schrödinger eqn
- PDEs are straightforward to solve but are computationally intense because of their 3D nature
- First we'll cover boundary value problems (BVP) today
- Next lecture we'll start to cover initial value problems (e.g. a string being plucked)
- BVPs have constraints on their boundaries while the interior obeys a set of PDEs

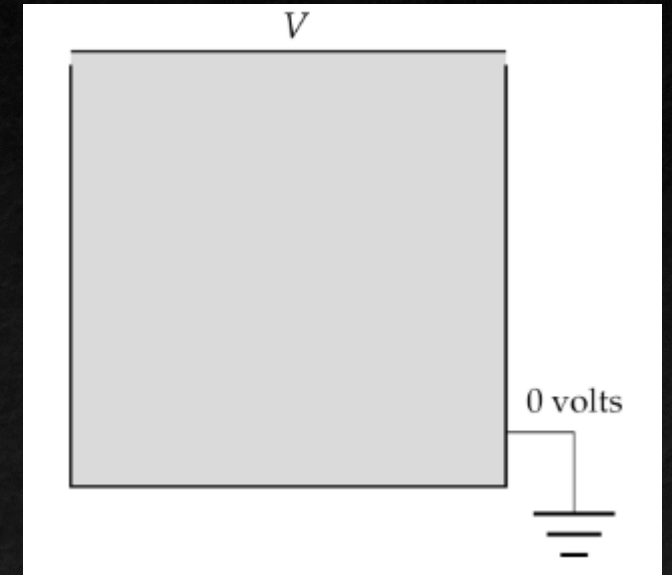


Introduction

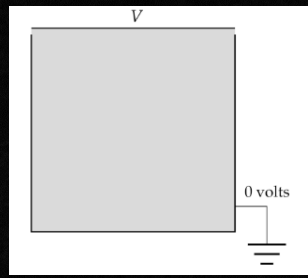
- BVPs have constraints on their boundaries while the interior obeys a set of PDEs
- Consider an empty conducting box with a voltage V on the top wall while the other 5 walls are grounded
- What is the electric potential V inside the box?
- We have Gauss's Law $\nabla \cdot \vec{E} = 0$ and $\vec{E} = -\nabla\phi$, giving

$$\nabla^2\phi = \frac{\partial^2\phi}{\partial x^2} + \frac{\partial^2\phi}{\partial y^2} + \frac{\partial^2\phi}{\partial z^2} = 0$$

- This is time-independent (electrostatics) that makes the solution much easier



Relaxation method: BC problems



- A fundamental technique to solve PDEs is the *method of finite differences*
- Let's consider our electrostatics problem in 2D

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} = 0$$

- We want to solve for ϕ given the boundary conditions
- Set up a grid of points where we'll solve for ϕ
- Because it's a box, we'll use Cartesian coordinates
- We can use central differencing to numerically calculate the partials

$$\begin{aligned}\frac{\partial^2 \phi}{\partial x^2} &= \frac{1}{a^2} [\phi(x + a, y) + \phi(x - a, y) - 2\phi(x, y)] \\ \frac{\partial^2 \phi}{\partial y^2} &= \frac{1}{a^2} [\phi(x, y + a) + \phi(x, y - a) - 2\phi(x, y)]\end{aligned}$$

Relaxation method: BC problems

$$\frac{\partial^2 \phi}{\partial^2 x} = \frac{1}{a^2} [\phi(x+a, y) + \phi(x-a, y) - 2\phi(x, y)]$$
$$\frac{\partial^2 \phi}{\partial^2 y} = \frac{1}{a^2} [\phi(x, y+a) + \phi(x, y-a) - 2\phi(x, y)]$$

- We insert these expressions into the Laplace equation and obtain

$$\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a) - 4\phi(x, y) = 0.$$

- Solving for the potential at (x,y), we find

$$\phi(x, y) = \frac{1}{4} [\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a)],$$

- This is just an average of the surrounding points
- Known as the **Jacobi method**

Relaxation method: BC problems

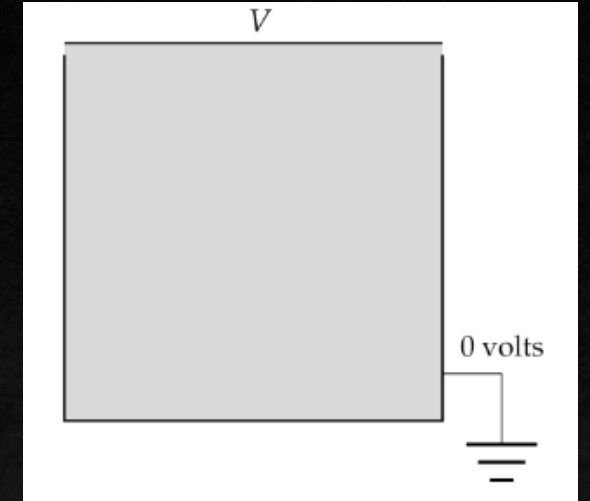
$$\phi(x, y) = \frac{1}{4} [\phi(x + a, y) + \phi(x - a, y) + \phi(x, y + a) + \phi(x, y - a)],$$

- How do we solve this system of equations?
 - Gaussian elimination or LU decomposition is too difficult with so many terms
- We can use the relaxation method
 - Choose an initial guess
 - Iteratively update the solution
 - Halt when a desired accuracy is reached (that is, the change between current and updated solutions)
- The Jacobi method is always stable and will converge to the correct solution

Jacobi method: Laplace's Eqn

In-class problem

- Let's solve this example problem for a top voltage $V = 1 \text{ V}$
- Download the skeleton code `06_laplace0.py`
- The box is 1 meter on a side
- We set up a 101^2 grid (including the boundaries) to have a 1cm spacing
- Use the Jacobi method to solve for the potential inside the box



$$\phi(x, y) = \frac{1}{4} [\phi(x + a, y) + \phi(x - a, y) + \phi(x, y + a) + \phi(x, y - a)],$$

Relaxation method: BC problems

- This example took over 700 iterations to converge to a solution, coming from the 2nd derivative being only 1st-order accurate
- We can use higher-order estimates (involving more neighboring grid points) that are more accurate and converge in fewer iterations. However this requires more calculations
- As in any physics problem, you want to choose a grid that aligns with the system's symmetry or boundary

Jacobi method: Poisson Eqn

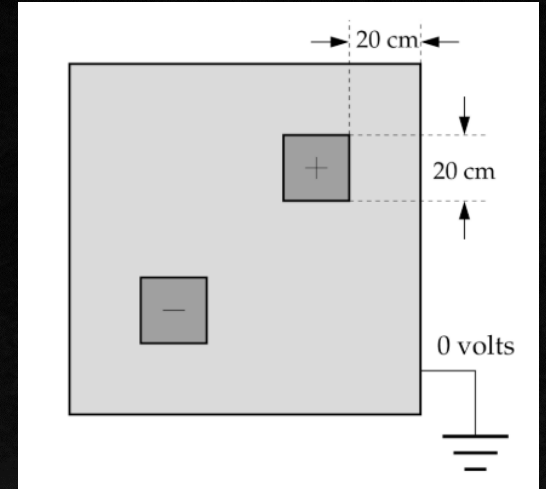
In-class problem

- Let's look at a more complex electrostatics problem
- Consider a conducting and grounded 1m box
- There are two $(20\text{cm})^2$ squares with a charge density of $\pm 1 \text{ C m}^{-2}$. Their lower left corners are at $(x,y) = (0.2, 0.2) \text{ m}$ and $(0.6, 0.6) \text{ m}$. The system is shown above.
- We need to solve the Poisson equation

$$\nabla^2 \phi = -\frac{\rho}{\epsilon_0}$$

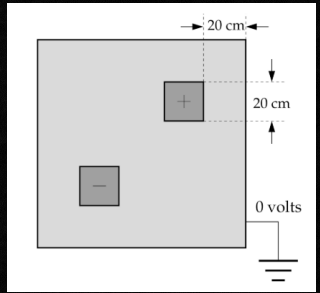
- We can use finite differencing to express the Poisson equation as

$$\frac{1}{a^2} [\phi(x+a, y) + \phi(x-a, y) + \phi(x, y+a) + \phi(x, y-a) - 4\phi(x, y)] = -\frac{\rho(x, y)}{\epsilon_0}.$$



Jacobi method: Poisson Eqn

In-class problem



$$\frac{1}{a^2} [\phi(x + a, y) + \phi(x - a, y) + \phi(x, y + a) + \phi(x, y - a) - 4\phi(x, y)] = -\frac{\rho(x, y)}{\epsilon_0}.$$

- Solving for the potential, we find

$$\phi(x, y) = \frac{1}{4} [\phi(x + a, y) + \phi(x - a, y) + \phi(x, y + a) + \phi(x, y - a)] + \frac{a^2}{4\epsilon_0} \rho(x, y).$$

- The last term is known as the source/sink term that supplies the charge density for the potential
- Modify the skeleton code `06_poisson0.py` and use the Jacobi method to calculate the potential

Over-relaxation

- The Jacobi method is straightforward but it converges slowly
- We can accelerate the convergence by overshooting the solution
 - For example at some point, suppose our initial guess is 0.1, but the solution is 0.5
 - In the first iteration, the solution is updated from 0.1 to 0.3
 - We can overshoot a little and go to 0.4
- We don't know the solution, but the method of over-relaxation will converge with less iterations
- Let's consider the Laplace equation again, where we're updating the solution by some amount $\Delta\phi$

$$\phi'(x, y) = \phi(x, y) + \Delta\phi(x, y),$$

Over-relaxation

$$\phi'(x, y) = \phi(x, y) + \Delta\phi(x, y),$$

- Define a set of over-relaxed values

$$\phi_{\omega}(x, y) = \phi(x, y) + (1 + \omega)\Delta\phi(x, y),$$

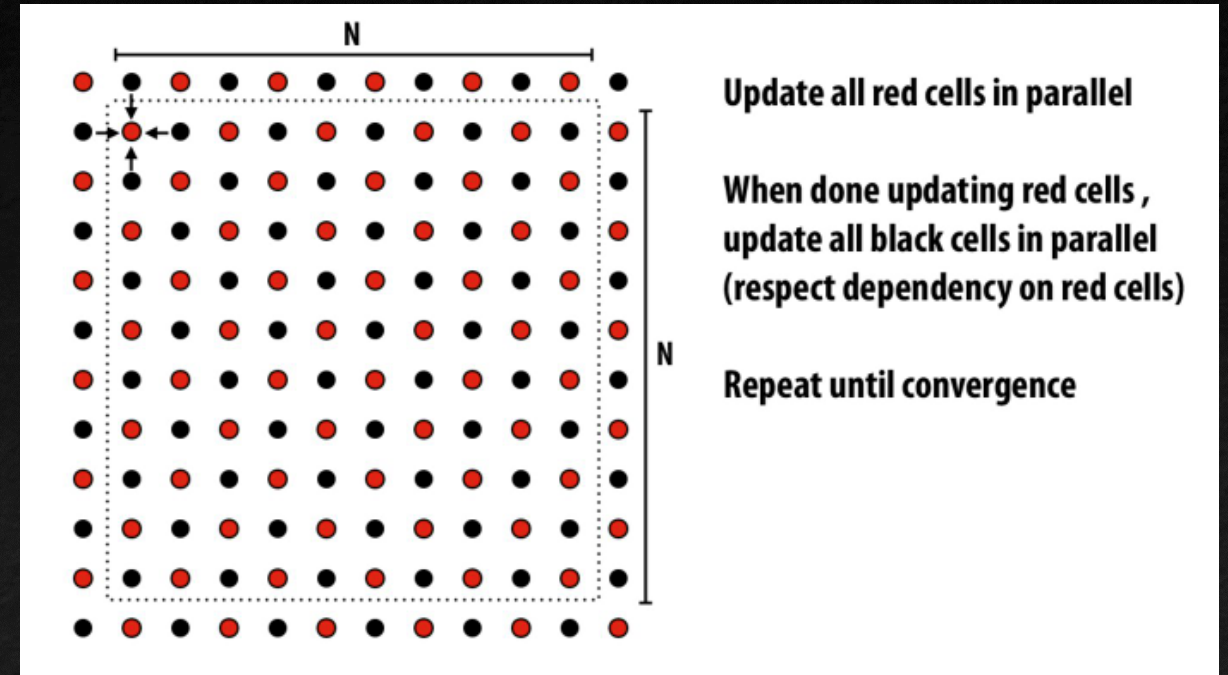
- Here $\omega > 0$, meaning that we let the solution change a little more than the Jacobi method
- Now substitute $\Delta\phi$ from its definition into the over-relaxed solution to obtain

$$\begin{aligned}\phi_{\omega}(x, y) &= \phi(x, y) + (1 + \omega)[\phi'(x, y) - \phi(x, y)] \\ &= (1 + \omega)\phi'(x, y) - \omega\phi(x, y) \\ &= \frac{1 + \omega}{4} [\phi(x + a, y) + \phi(x - a, y) + \phi(x, y + a) + \phi(x, y - a)] - \omega\phi(x, y)\end{aligned}$$

- The over-relaxation method is very similar to the Jacobi method, but we have to choose some ω value
 - No exact method to choose ω , it is usually stable for <1
 - If too large, the method will oscillate around the true solution
 - In the previous example, $\omega = 0.9$ accelerates the convergence tremendously by 10-20 times

Gauss-Seidel method

- A widely-used “numerical trick” to speed up the Jacobi method is to **use the updated values as they are calculated**. This is known as the Gauss-Seidel method.
- In practice, this means we will have one array for the solution and replace the values as they are calculated
- To the right, the red-black variant is shown
- We can also use over-relaxation in combination with the Gauss-Seidel method
- One benefit is that the over-relaxed Gauss-Seidel method is **formally stable and never diverges**



Initial value problems: Overview

- Boundary value problems are only one type of PDE that are time-independent
- There are many problems that evolve and are time-dependent
- These are called initial value problems
- Goal: Predict their future variation

- A simple example is the diffusion (heat) equation in one dimension

$$\frac{\partial \phi}{\partial t} = D \frac{\partial^2 \phi}{\partial x^2}$$

- Here D is the diffusion coefficient

Initial value problems: Overview

$$\frac{\partial \phi}{\partial t} = D \frac{\partial^2 \phi}{\partial x^2}$$

- The variable $\phi(x, t)$ depends on space and time
- Can think of this as a 2D problem, like our previous example using the 2D Laplace equation
- Now we can consider a space-time grid, where both space and time are discretized
- We have an initial system setup, but we don't know the endpoint
- Goal: Predict how it behaves afterward
- The only feasible method is based on forward integration

The FTCS Method

$$\frac{\partial \phi}{\partial t} = D \frac{\partial^2 \phi}{\partial x^2}$$

- Just like how we represented partial derivatives in boundary value problems, we can do the same with the time derivative but forward in time
- We can write the right-hand side as

$$\frac{\partial^2 \phi}{\partial x^2} = \frac{1}{a^2} [\phi(x + a, t) + \phi(x - a, t) - 2\phi(x, t)]$$

- That gives

$$\frac{\partial \phi}{\partial t} = \frac{D}{a^2} [\phi(x + a, t) + \phi(x - a, t) - 2\phi(x, t)]$$

- We can think of the variables at different points in space as separate variables
- Meaning that we have a set of simultaneous ODEs
- On a single compute core, it's no problem to solve thousands or even millions of them

The FTCS Method

$$\frac{\partial \phi}{\partial t} = D \frac{\partial^2 \phi}{\partial x^2}$$

- Now we need to choose a method to solve the left-hand side, $\partial \phi / dt$
- The right-hand side is only first-order accurate
- So there's little benefit to represent the left-hand side with a higher-order method
- We choose Euler's method to advance the system forward in time
 - RK methods would be a waste because the increased accuracy would be lost to the RHS 2nd order error

$$\begin{aligned}\phi(t + h) &\simeq \phi(t) + h \frac{d\phi}{dt} \\ &\simeq \phi(x, t) + h \frac{D}{a^2} [\phi(x + a, t) + \phi(x - a, t) - 2\phi(x, t)]\end{aligned}$$

- This is known as the **forward-time, centered-space (FTCS)** method for solving PDEs

The Heat Equation: FTCS

In-class example

$$\begin{aligned}\phi(t+h) &\simeq \phi(t) + h \frac{d\phi}{dt} \\ &\simeq \phi(x,t) + h \frac{D}{a^2} [\phi(x+a,t) + \phi(x-a,t) - 2\phi(x,t)]\end{aligned}$$

- The flat base of a container made of 1cm thick stainless steel is initially at a uniform temperature of 20C
- The contained is placed in a cold bath at 0C and filled with soup at 80C (assume constant w/ time)
- **Goal: Calculate the temperature profile from the hot → cold side w.r.t. position and time**
- Thermal conduction is governed by the diffusion equation (shown above)
- Divide the x-axis into 100 equal grid integrals
- Boundary conditions: hot and cold bath temperatures
- Initial condtions: 20C everywhere except boundaries
- Heat diffusion coefficient for stainless steel: $D = 4.25 \times 10^{-6} \text{ m}^2 \text{ s}^{-1}$
- Make a temperature profile plot at $t = (0.01, 0.1, 0.4, 1, \text{ and } 10)$ seconds all on the same graph



Numerical stability

- The FTCS method works well for the diffusion equation
- But there are many other equations where it's **stable for only some conditions**
- Consider the wave equation

$$\frac{\partial^2 \phi}{dx^2} - \frac{1}{v^2} \frac{\partial^2 \phi}{dt^2} = 0$$

- To solve this, we would use central spatial differencing, like before
- But for the 2nd order time derivative, we can write two 1st order ODEs

$$\frac{\partial \phi}{\partial t} = \psi(x, t), \quad \frac{\partial \psi}{\partial t} = \frac{v^2}{a^2} [\phi(x + a, t) + \phi(x - a, t) - 2\phi(x, t)]$$

Numerical stability

$$\frac{\partial \phi}{\partial t} = \psi(x, t), \quad \frac{\partial \psi}{\partial t} = \frac{v^2}{a^2} [\phi(x + a, t) + \phi(x - a, t) - 2\phi(x, t)]$$

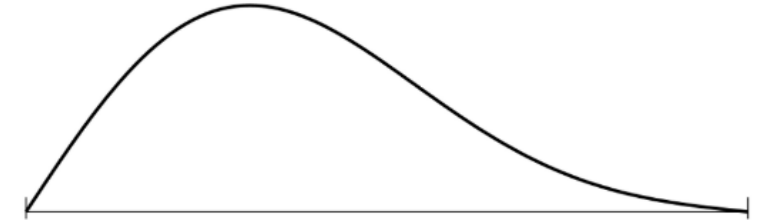
- Apply Euler's method to both variables

$$\phi(x, t + h) = \phi(x, t) + h\psi(x, t)$$

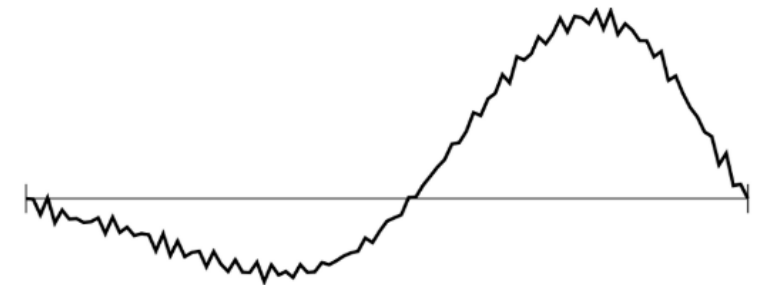
$$\psi(x, t + h) = \phi(x, t) + h\frac{v^2}{a^2} [\phi(x + a, t) + \phi(x - a, t) - 2\phi(x, t)]$$

- Evolve the system with some timestep h

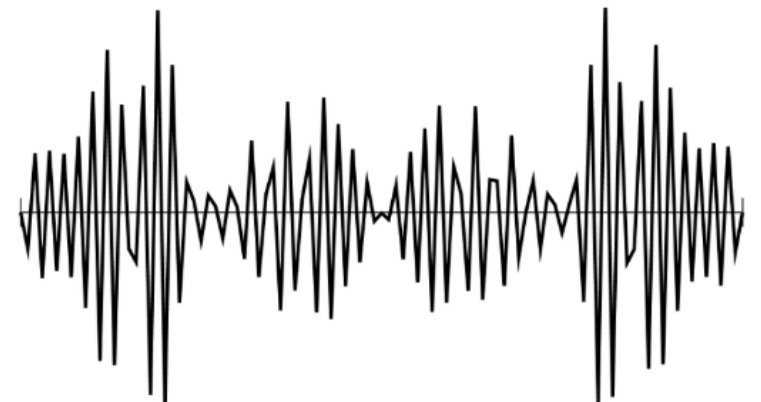
System at time $t = 2$ ms



System at time $t = 50$ ms



System at time $t = 100$ ms



Numerical stability

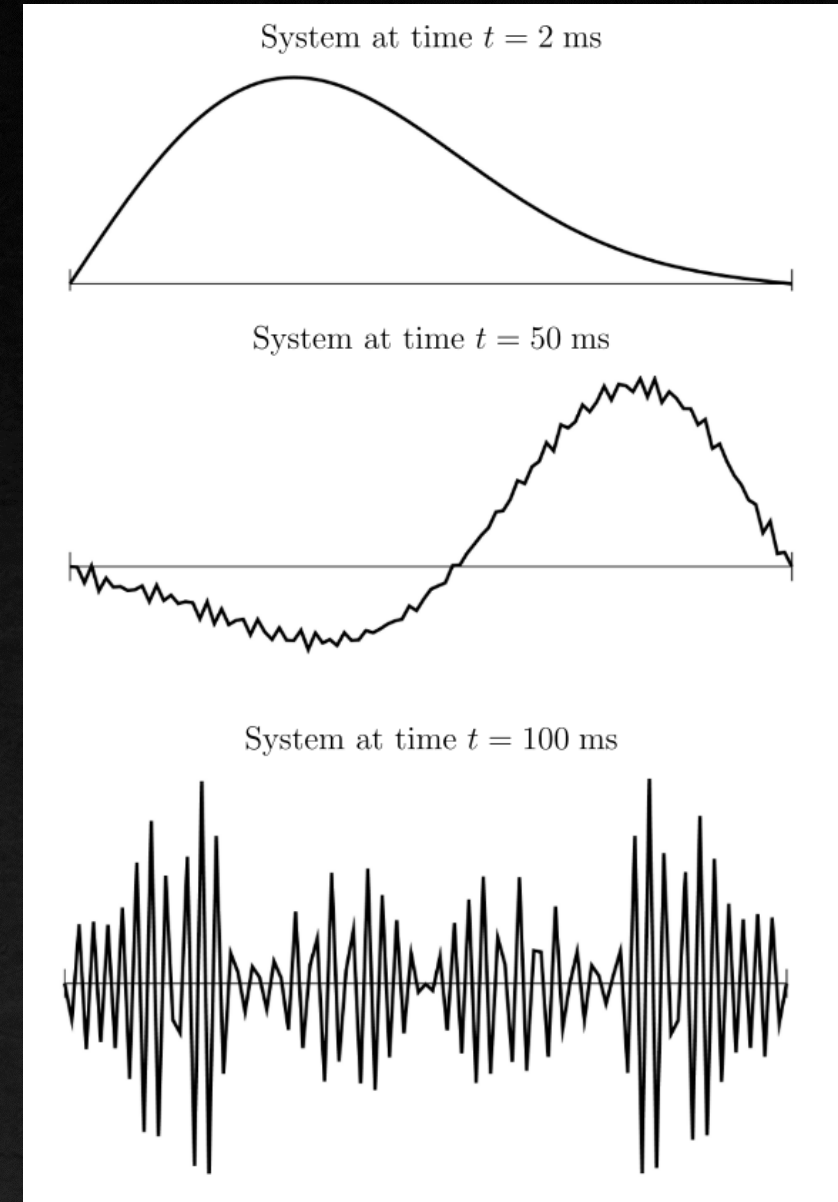
- This is not roundoff or truncation error
- The errors will accumulate, eventually causing the calculation to crash

- We can perform a **von Neumann stability analysis**

- Here we express the solution as a Fourier series

$$\phi(x, t) = \sum_k c_k(t) e^{ikx}$$

- Given such a solution, we can see how each term changes in the next timestep
- *Note: We cannot perform this for non-linear equations*



Numerical stability

- Let's plug this solution into the FTCS diffusion equation

$$\begin{aligned}\phi(x, t + h) &= c_k(t)e^{ikx} + h\frac{D}{a^2} [e^{ik(x+a)} + e^{ik(x-a)} - 2e^{ikx}] \\ &= \left[1 + h\frac{D}{a^2} (e^{ika} + e^{-ika} - 2) \right] c_k(t)e^{ikx} \\ &= \left[1 - h\frac{4D}{a^2} \sin^2 \left(\frac{ka}{2} \right) \right] c_k(t)e^{ikx},\end{aligned}$$

- Here we've used $e^{i\theta} + e^{-i\theta} = 2$ and $1 - \cos\theta = 2\sin^2 \left(\frac{\theta}{2} \right)$

Numerical stability

- Notice that each Fourier coefficient is independent of each other, evolving as

$$c_k(t + h) = \left[1 - h \frac{4D}{a^2} \sin^2 \left(\frac{ka}{2} \right) \right] c_k(t).$$

- The solution will be unstable if these grow with them, otherwise it's stable
- The largest the \sin^2 term can be is 1, so the solution is stable when $h \leq \frac{a^2}{4D}$
- If the timestep is larger than this, the solution can diverge

Numerical stability: Wave eqn

- Let's apply the von Neumann analysis to the wave equation (two 1st order ODEs)

$$\begin{pmatrix} \phi(x, t) \\ \psi(x, t) \end{pmatrix} = \begin{pmatrix} c_\phi(t) \\ c_\psi(t) \end{pmatrix} e^{ikx}$$

- Making the same Fourier series substitution, we find

$$\begin{aligned} c_\phi(x + h) &= c_\phi(t) + hc_\psi(t), \\ c_\psi(x + h) &= c_\psi(t) - hc_\phi(t) \frac{4v^2}{a^2} \sin^2 \frac{ka}{2}. \end{aligned}$$

Numerical stability: Wave eqn

$$\begin{aligned}c_\phi(x+h) &= c_\phi(t) + hc_\psi(t), \\c_\psi(x+h) &= c_\psi(t) - hc_\phi(t) \frac{4v^2}{a^2} \sin^2 \frac{ka}{2}.\end{aligned}$$

- Write this in vector form as $\mathbf{c}(t+h) = \mathbf{A}\mathbf{c}(t)$, where

$$\mathbf{A} = \begin{pmatrix} 1 & h \\ -hr^2 & 1 \end{pmatrix} \quad \text{with} \quad r = \frac{2v}{a} \sin \frac{ka}{2}$$

- We can write $\mathbf{c}(t)$ as a linear combination of two eigenvectors \mathbf{v}_1 and \mathbf{v}_2 of \mathbf{A}

$$\mathbf{c}(t+h) = \mathbf{A}(\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2) = \alpha_1 \lambda_1 \mathbf{v}_1 + \alpha_2 \lambda_2 \mathbf{v}_2,$$

- After m timesteps, we would have

$$\mathbf{c}(t+h) = \alpha_1 \lambda_1^m \mathbf{v}_1 + \alpha_2 \lambda_2^m \mathbf{v}_2,$$

Numerical stability: Wave eqn

- The solution is stable if both eigenvalues $(\lambda_1, \lambda_2) \leq 1$
- The eigenvalues are given by the determinant equation $A - \lambda I = 0$
- Using A for the wave equation (see above), both solutions have the same magnitude and **always greater than 1**

$$|\lambda| = \sqrt{1 + h^2 r^2} = \sqrt{1 + \frac{4h^2 v^2}{a^2} \sin^2 \frac{ka}{2}},$$

- Meaning that **the numerical solution with FTCS will always diverge**
- We need to find another technique to solve the wave equation