



LECTURE 4-2

DIFFUSION EQUATION

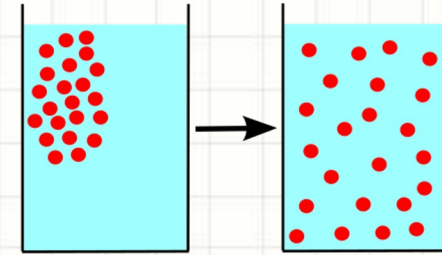
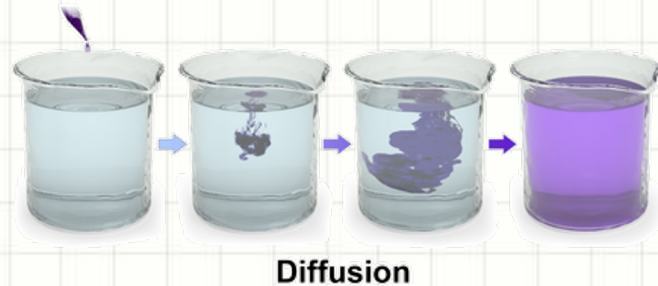
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Outline

- Diffusion Equation
- Analytic Solution in 1D
- 1st Order Method
- 2nd Order Method
- Alternating-Direction Implicit Method for 2D (and higher).

Reference: [Numerical Recipes in C](#)

Diffusion Equation



- Fick's law of diffusion

$$\mathbf{\Gamma} = -D\nabla u$$

$\mathbf{\Gamma}$ is the particle flux

- Flux conservation

$$\frac{\partial u}{\partial t} + \nabla \cdot \mathbf{\Gamma} = 0$$

- Fick's second law of diffusion: combining two equation above

$$\frac{\partial u}{\partial t} = D\nabla^2 u$$

- When there exist particle source, above equation is generalized to

$$\frac{\partial u}{\partial t} = D\nabla^2 u + S(\mathbf{x})$$

Analytic Solution in 1D

- Separation the variables as $u = T(t)S(x)$, where T and S are time-dependent-only and position-dependent-only functions.
- Put this into the diffusion equation to get

$$\frac{1}{D} \frac{T'}{T} = \frac{S''}{S} = -k^2$$

- The solution for S is

$$S = A \sin kx + B \cos kx$$

- The boundary condition is $S = 0$ at $x=0$ and L , where L is the length of the system. From this, $k = \frac{N\pi}{L}$ and $B=0$, where N is an integer.

- The time-part solution is, $T \propto e^{-t/\tau}$, where $\tau = \frac{1}{N^2\pi^2} \frac{L^2}{D}$, is the characteristic time of diffusion (i.e. how fast the particles diffuse).
- The solution for a specific initial mode k ,

$$u = u_0 e^{-t/\tau} \sin kx$$

- When the system starts with more general profile of n ,

$$u = \sum_{N=1}^{\infty} A_N e^{-t/\tau_N} \sin k_N x$$

- Note that higher harmonics diffuse faster.

Methods of 1st Order in Time

FTCS method

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} = D \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{\Delta x^2}$$

$$u_j^{n+1} = u_j^n + \alpha [u_{j+1}^n - 2u_j^n + u_{j-1}^n] \quad \alpha = \frac{D\Delta t}{\Delta x^2}$$

Amplification factor from von Neumann stability analysis

$$\xi = 1 - \frac{4D\Delta t}{(\Delta x)^2} \sin^2\left(\frac{k\Delta x}{2}\right)$$

Stability condition

$$\frac{2D\Delta t}{(\Delta x)^2} \leq 1$$

BTCS method

$$u_j^{n+1} = u_j^n + \alpha [u_{j+1}^{n+1} - 2u_j^{n+1} + u_{j-1}^{n+1}]$$

Amplification factor

$$\xi = \frac{1}{1 + 4\alpha \sin^2\left(\frac{k\Delta x}{2}\right)}$$

Unconditionally stable

Methods of 1st Order in Time

Example 1. Write a FTCS diffusion solver. Set the initial shape of u to $\sin(\pi x/L)$, where L is the system length. Run it with $x_{\max}=1$, $dx=0.01$, $dt=0.001$, $D=0.05$, $i_{\max}=500$.

Answer: Look at the code 'diffusion-1d-ftcs.py'.

Example 2. Add some higher harmonics to the initial condition of u . Observe the long-time behavior of u . How does the shape of u change? Explain it in terms of the characteristic diffusion time.

Answer: Look at the code 'diffusion-1d-ftcs-HH.py'.

Crank-Nicolson Method

- CN method is an implicit, second-order-in-time method.

$$\frac{u_j^{n+1} - u_j^n}{\Delta t} = \frac{D}{2} \left[\frac{(u_{j+1}^{n+1} - 2u_j^{n+1} + u_{j-1}^{n+1}) + (u_{j+1}^n - 2u_j^n + u_{j-1}^n)}{(\Delta x)^2} \right]$$

- Average of $D\nabla^2 u$ at n and $n+1$ steps gives the effect of evaluating $\partial u / \partial t$ at $n+1/2$ step. Hence the left-hand-side is the 3-point derivative with the 2nd-order.
- The amplification factor:

$$\xi = \frac{1 - 2\alpha \sin^2\left(\frac{k\Delta x}{2}\right)}{1 + 2\alpha \sin^2\left(\frac{k\Delta x}{2}\right)} \quad \text{with} \quad \alpha = \frac{D\Delta t}{\Delta x^2} \quad \text{Unconditionally stable!}$$

- The discretized equation is

$$\beta u_j^{n+1} - \alpha u_{j+1}^{n+1} - \alpha u_{j-1}^{n+1} = \alpha u_{j+1}^n + \alpha u_{j-1}^n + \gamma u_j^n \equiv f_j^n$$

$$\text{with } \beta = 2(1 + \alpha) \text{ and } \gamma = 2(1 - \alpha)$$

Alternating-Direction Implicit Method

- To solve more than 2D systems, the ADI (alternating-direction implicit) method is quite powerful.
- For the first half-step, advance in one direction implicitly, keeping the other. In the next half-step, advance the other, keeping the first direction.

$$\overset{\text{advance x-direction}}{u_{j,l}^{n+1/2}} - \overset{\text{keep y-direction}}{u_{j,l}^n} = \frac{\alpha}{2} \left[\overset{\text{advance x-direction}}{u_{j+1,l}^{n+1/2}} - 2\overset{\text{advance x-direction}}{u_{j,l}^{n+1/2}} + \overset{\text{advance x-direction}}{u_{j-1,l}^{n+1/2}} + \overset{\text{keep y-direction}}{u_{j,l+1}^n} - 2\overset{\text{keep y-direction}}{u_{j,l}^n} + \overset{\text{keep y-direction}}{u_{j,l-1}^n} \right]$$

$$\overset{\text{keep x-direction}}{u_{j,l}^{n+1}} - \overset{\text{advance y-direction}}{u_{j,l}^{n+1/2}} = \frac{\alpha}{2} \left[\overset{\text{keep x-direction}}{u_{j+1,l}^{n+1/2}} - 2\overset{\text{keep x-direction}}{u_{j,l}^{n+1/2}} + \overset{\text{keep x-direction}}{u_{j-1,l}^{n+1/2}} + \overset{\text{advance y-direction}}{u_{j,l+1}^{n+1}} - 2\overset{\text{advance y-direction}}{u_{j,l}^{n+1}} + \overset{\text{advance y-direction}}{u_{j,l-1}^{n+1}} \right]$$

- The discrete equations used in the code are

$$\begin{aligned} \beta u_{j,l}^{n+1/2} - \alpha u_{j+1,l}^{n+1/2} - \alpha u_{j-1,l}^{n+1/2} &= \alpha u_{j,l+1}^n + \alpha u_{j,l-1}^n + \gamma u_{j,l}^n = f_{j,l}^n \\ \beta u_{j,l}^{n+1} - \alpha u_{j,l+1}^{n+1} - \alpha u_{j,l-1}^{n+1} &= \alpha u_{j+1,l}^{n+1/2} + \alpha u_{j-1,l}^{n+1/2} + \gamma u_{j,l}^{n+1/2} = f_{j,l}^{n+1/2} \end{aligned}$$

$$\beta = 2(1 + \alpha) \quad \gamma = 2(1 - \alpha)$$

- This method is unconditionally stable.
- The ADI method can be applied to the relaxation solver of the Poisson equation.

2D Diffusion in Python

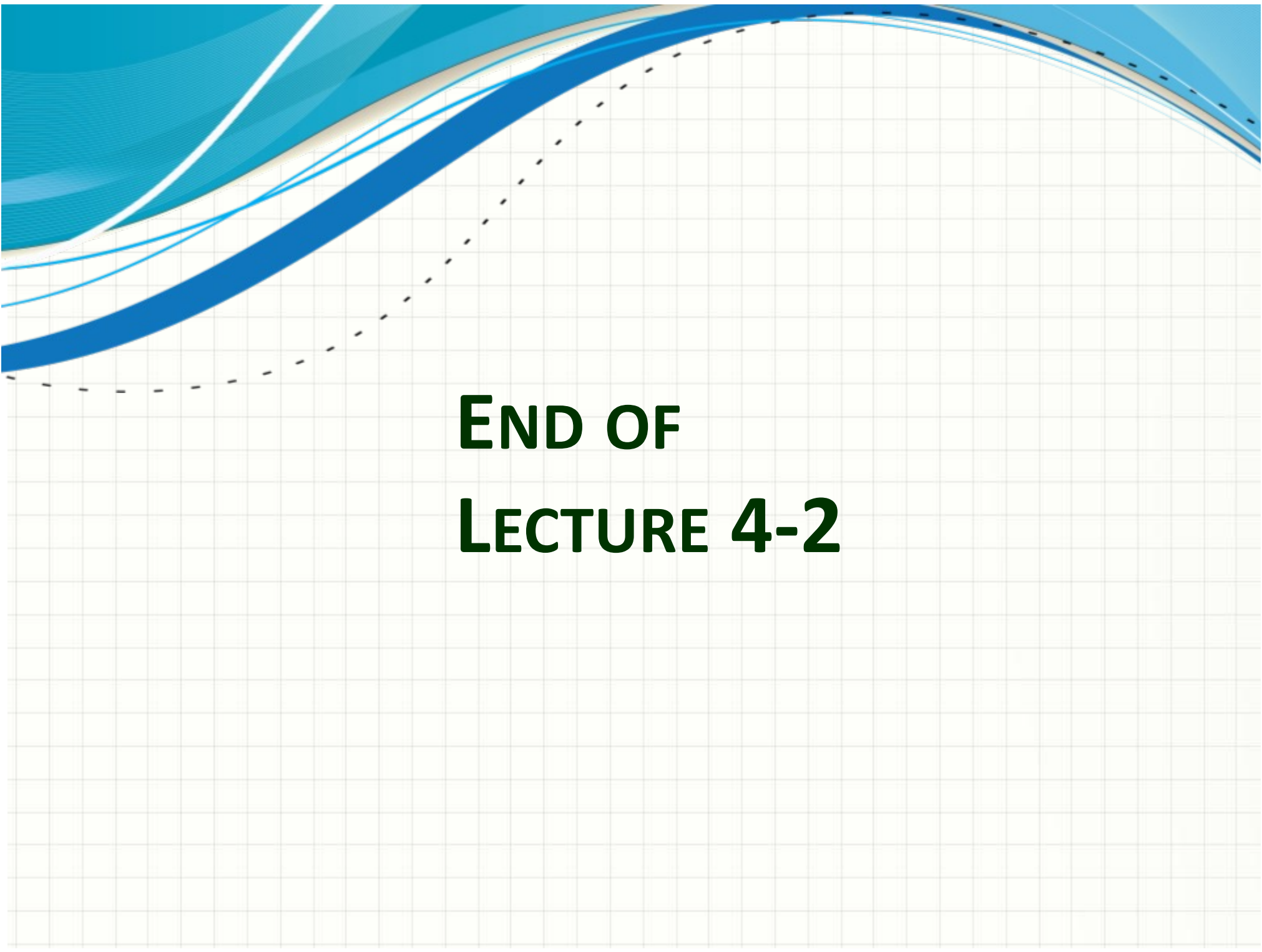
Example 3. Create a python script for 2D simulation of the diffusion. Initialize u by $\sin(\pi x/L_x) \sin(\pi y/L_y)$ and its harmonics.

Answer: Look at the code 'diffusion-2d-adi.py'. Main part of the code is,

```
...
while i<imax:
    f[1:-1,1:-1]=g*u[1:-1,1:-1]+a*u[0:-2,1:-1]+a*u[2:,1:-1] # fn
    Half-step
    advance in x
    j=0
    while j<Nx-1: f[1:-1,j+1] -= f[1:-1,j]*(ll/d[j]); j+=1 # f-transform by diagonalization
    j=Nx-1
    while j>0: u[1:-1,j] = (f[1:-1,j] - uu*u[1:-1,j+1])/d[j]; j -=1 # backsubstitution for un+1/2

    f[1:-1,1:-1]=g*u[1:-1,1:-1]+a*u[1:-1,0:-2]+a*u[1:-1,2:] # fn+1/2
    Another half-step
    advance in y
    l=0
    while l<Ny-1: f[l+1,1:-1] -= f[l,1:-1]*(ll/d[l]); l+=1 # f-transform
    l=Ny-1
    while l>0: u[l,1:-1] = (f[l,1:-1] - uu*u[l+1,1:-1])/d[l]; l -=1 # backsubsti. for un+1

    i+=1
```



**END OF
LECTURE 4-2**