

Monday

Lecture #1

19-09-16

Numerical Analysis

Num. Sol. of ODEs
 Num. Sol of PDEs

Norm:-

$$\|f(x)\|_1 = \int_0^{\infty} |f(x)| dx$$

$$\|f(x)\|_2 = \sqrt{\int_0^{\infty} |f(x)|^2 dx}$$

$$\|f(x)\|_{\infty} = \max_{()} |f(x)|$$

$$\int_a^b f(x) dx = \sum_{i=1}^N df(x_i)$$

$$\frac{du_{\text{dependent}}}{dt} = f(t)_{\text{independent}}$$

$$\frac{du_1}{dt} = f_1(t, u_1, u_2)$$

$$\frac{du_2}{dt} = f_2(t, u_1, u_2)$$

$$\frac{d^2y}{dx^2} = f(x, y)$$

$y(a) = \alpha, y(b) = \beta, a \neq b$ BVP

otherwise IVP

$$\frac{\partial u}{\partial t} + a \frac{\partial u}{\partial x} = c$$

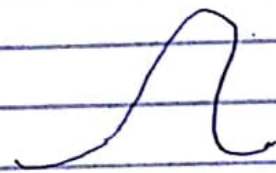
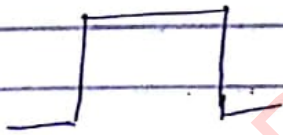
$$\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2}$$

Taylor Expansion
 not used for
 polynomial
 interpolation.

Mathematical Modeling and Simulation:-

- (1) Define the physical problem.
- (2) Create a mathematical model.
 - (i) system of PDEs, ODEs and algebraic Equations.
 - (ii) Define initial and boundary conditions to get a well posed problem.
- (3) Create a discrete (numerical) model
 - (i) discretize the domain \Rightarrow generate the grid.
 - (ii) solve the discrete system.
- (4) Analyze errors in discrete system
 - (i) consistency, stability and convergence.

well posed
not well posed



PDEs:-

Mathematical models of continuous physical phenomena in which the dependant variable, say u is, a function of more than one independent variables, say t (time) and x (e.g. spatial position)

Well posed Problem:-

- (1) Solution exists
- (2) Solution unique.

(3) solution depends continuously on data.
 i.e. small changes in the data produces small changes in the solution (i.e. the solution is continuously differentiable).

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Lecture #02

20-09-16

Our Aim:-

To develop efficient algorithms for computing numerical solutions of PDEs, especially practical problems.

Beyond learning basic techniques, it is crucial to understand that how algorithms are constructed, why they work and what are their limitations.

Source of Errors in Mathematical

Model:-

(1) Model Error:-

inexactness of the mathematical model for underline physical problem

(2) Data Error:-

of parameters entering the model. Error in measurement

(3) Round off error:-

computer arithmetic errors in

(4) Truncation Error:-

numerical method used to solve the full mathematical system. Errors in

Errors Estimates:-

mathematical information about the accuracy of calculated numerical solution.

u_n, u_e

$$\text{error} = \|u_e - u_n\| \quad \checkmark$$

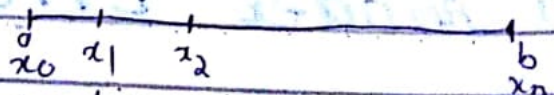
Numerical solutions of PDEs:-

are discretization parameters n and h

main

$u_n \quad n \rightarrow \infty$ } numerical

$u_h \quad h \rightarrow 0$ } approximation



n = number of discretization points.

h = step size

classical Problems of numerical Analysis:-

- (1) Existence and uniqueness:- The solution U_h exists and unique.
- (2) **Stability**:- The solution U_h remains bounded for $h \rightarrow 0$ (in clear sense).
- (3) **Consistency**:- A small residuum (remaining part of Taylor expansion) is obtained after substituting the exact solution u of the differential equation in the discretization formulation (the numerical method formulation). Moreover this residuum tends to zero $h \rightarrow 0$.
- (4) **Convergence**:- For $h \rightarrow 0$, the discrete solution U_h tends (converges) to continuous solution u .

PDEs:-

Linear General PDE:-

$$a u_x + b u_y + c u + d = 0 \text{ (1st order)}$$

a, b, c, d are either constant or function of independent variables x and y .

$$a = a(u, u_x, u_y) \text{ Hyperbolic PDE.}$$

Linear PDE of order 2:-

$$a u_{xx} + b u_{xy} + c u_{yy} + d u_x + e u_y + f u + g = 0$$

Classification:-

(1) Hyperbolic PDEs:-

$$U_{tt} = c^2 U_{xx} \text{ (wave equation)}$$

(NTS)

$$u_t + a u_x = 0$$

$$\frac{du}{dt} = \frac{\partial u}{\partial t} \frac{\partial t}{\partial t} + \frac{\partial u}{\partial x} \frac{dx}{dt}$$

$$= 0$$

(2) Parabolic PDE:-

$$b^2 - 4ac = 0$$

$$u_t = a u_{xx} \text{ (Heat Equation)}$$

(3) elliptic PDE:-

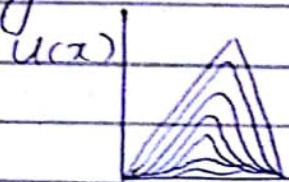
$$b^2 - 4ac < 0$$

$$\nabla^2 u = 0 \text{ Laplace Equation}$$

$$u_{xx} + u_{yy} = 0$$

(1) Hyperbolic PDEs describe time-dependent conservative physical processes, such as convection, that are not evolving towards steady state.

(2) Parabolic PDEs describe time-dependent dissipative physical processes, such as diffusion, that are evolving towards a steady state.



(3) Elliptic PDEs describe processes that have already reached steady state and hence are time-independent.

$$(1) u_t + a u_x = 0 \rightarrow \text{IVP \& BVP}$$

$$(2) u_t = a u_{xx}$$

$$(3) u_{xx} + u_{yy} = 0 \rightarrow \text{BVP}$$

Mixed Equations:-

$$\frac{\partial^2 u}{\partial x^2} = x \frac{\partial^2 u}{\partial y^2}$$

* This Equation is elliptic when

x < 0

* parabolic if $x = 0$ * hyperbolic if $x > 0$.

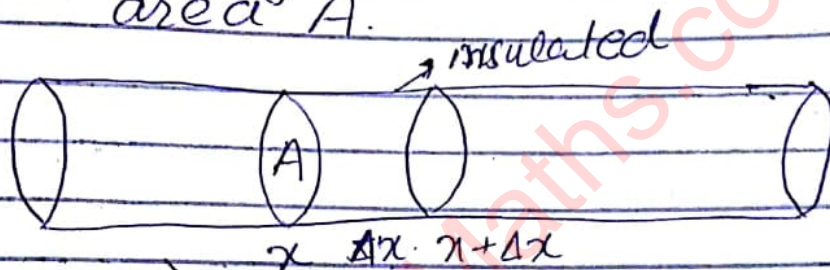
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Lecture #03

26-09-16

Heat conduction in a

rod:- consider a rod of constant length L and cross section area A .



Let $e(x,t)$ = thermal energy density
= amount of energy per unit volume.

Rod surface is insulated \Rightarrow NO energy loss across the surface.

Then the total energy in the slice = $e(t,x)A\Delta x$.

The rate of change of heat energy is given by $\frac{\partial}{\partial t}(e(x,t))A\Delta x$. because of conservation law of heat energy

Rate of change of total energy per unit time = Heat generated per unit time + Heat energy flowing across the boundaries per unit time

Let $\phi \rightarrow$ Heat flux. (amount of thermal energy per unit time flowing to the right per unit surface per unit time)

$S(x,t)$ denote heat energy generated per unit volume per unit time

$$\frac{\partial}{\partial t} (e(x,t) A \Delta x) = -[\phi(x+\Delta x, t) - \phi(x, t)] A + s(x,t) A \Delta x$$

$$\Rightarrow \frac{\partial}{\partial t} (e(x,t)) = -\frac{\phi(x+\Delta x, t) - \phi(x, t)}{\Delta x} + s(x,t)$$

$$\lim_{\Delta x \rightarrow 0} \frac{\partial}{\partial t} (e(x,t)) = -\lim_{\Delta x \rightarrow 0} \frac{\phi(x+\Delta x, t) - \phi(x, t)}{\Delta x} + \lim_{\Delta x \rightarrow 0} s(x,t)$$

$$\frac{\partial}{\partial t} (e(x,t)) = -\frac{\partial \phi(x,t)}{\partial x} + s(x,t)$$

Moreover

$$e(x,t) = c(x,t) \rho(x) u(x,t)$$

$c(x,t)$ → specific heat (Heat energy to be supplied to a unit mass to raise its temperature by one degree ($J/K \cdot kg$))
 Fourier law:

$$\phi(x,t) \propto -\frac{\partial u}{\partial x} = -k_T \frac{\partial u}{\partial x}$$

$$\frac{\partial}{\partial t} [c(x) \rho(x) u(x,t)] = \frac{\partial}{\partial x} (k_T \frac{\partial u}{\partial x}) + s(x,t)$$

in a special case c, ρ and k_T are constants.

$$\frac{\partial}{\partial t} (u(x,t)) = a \frac{\partial^2 u}{\partial x^2} + Q(x,t) \quad \left| \quad a = \frac{k_T}{\rho c} \right.$$

$$\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2} + Q \quad \left| \quad Q = \frac{s(x,t)}{\rho c} \right.$$

parabolic

Equation

initial condition $u(x, 0) = f(x)$

Boundary conditions

(i) Prescribed temperature (Dirichlet Bcs)
 $u(0, t) = p(t), u(L, t) = q(t)$

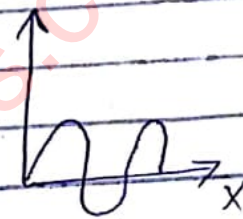
(ii) Insulate boundary conditions
 $\frac{\partial u}{\partial n}(0, t) = 0$

$$\frac{\partial u}{\partial n} = -\frac{\partial u}{\partial x} \text{ at } x=0$$

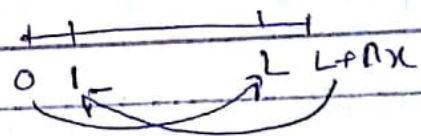
$$\frac{\partial u}{\partial n} = \frac{\partial u}{\partial x} \text{ at } x=L$$

(iii) periodic boundary condition
 $u(0, t) = u(L, t)$

Problem:-

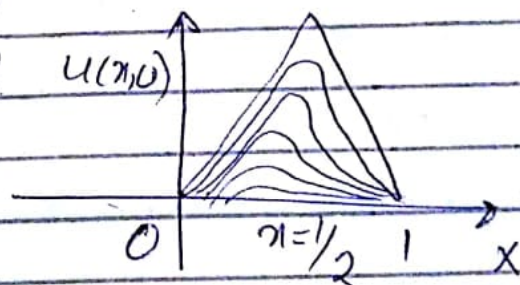
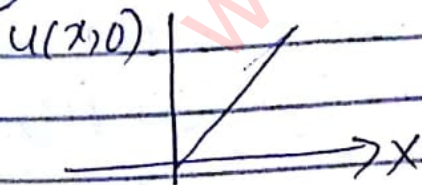


$$u(x, 0) = \begin{cases} 2x, & 0 \leq x < 1/2 \\ 2(1-x), & 1/2 \leq x < 1 \end{cases}$$



$$u(0, t) = 0 = u(L, t)$$

if $u(0, t), u(L, t) = 1$



Finite Difference Method for parabolic Equation:-

$$\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2}, \quad 0 \leq x \leq 1$$

$$u(x, 0) = u_0(x), \quad u(0, t) = 0, \quad u(1, t) = 0$$

$\Delta x = h \rightarrow$ space step
 $\Delta t = k \rightarrow$ timestep

(10)

$$u(x+h, t) = u(x, t) + h \frac{\partial u}{\partial x} + \frac{h^2}{2!} \frac{\partial^2 u}{\partial x^2} + \frac{h^3}{3!} \frac{\partial^3 u}{\partial x^3} + \dots$$

$$u(x-h, t) = u(x, t) - h \frac{\partial u}{\partial x} + \frac{h^2}{2!} \frac{\partial^2 u}{\partial x^2} - \frac{h^3}{3!} \frac{\partial^3 u}{\partial x^3} + \dots$$

$$\frac{u(x+h, t) - u(x, t)}{h} = \frac{\partial u}{\partial x} + \frac{h}{2!} \frac{\partial^2 u}{\partial x^2} + \frac{h^2}{3!} \frac{\partial^3 u}{\partial x^3} + \dots$$

$$= \frac{\partial u}{\partial x} + O(h)$$

(forward difference formula)

(1st order accurate)

$$\frac{u(x, t) - u(x-h, t)}{h} = \frac{\partial u}{\partial x} + O(h)$$

(backward diff formula) (1st order accurate)

$$u(x+h, t) + u(x-h, t) = 2u(x, t) + \frac{h^2}{2!} \frac{\partial^2 u}{\partial x^2} + O(h^4)$$

$$u(x+h, t) - 2u(x, t) + u(x-h, t) = \frac{\partial^2 u}{\partial x^2} + O(h^2)$$

(Central difference Method) (2nd order accurate)

$$u(x+h, t) - u(x-h, t) = 2h \frac{\partial u}{\partial x} + O(h^3)$$

$$\frac{u(x+h, t) - u(x-h, t)}{2h} = \frac{\partial u}{\partial x} + O(h^2)$$

central difference formula
 1st order accurate

Example:- $\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2}$ (ii) $0 \leq x \leq 1$

$u(0, t) = 0, u(1, t) = 0$

$$\frac{u(t+k, x) - u(t, x)}{k} = (a) \frac{u(t, x-h) - 2u(t, x) + u(t, x+h)}{h^2}$$

$$\frac{u_j^{n+1} - u_j^n}{k} = (a) \frac{u_{j-1}^n - 2u_j^n + u_{j+1}^n}{h^2}$$

N = number of discretization points

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

let $\lambda = \frac{k}{h^2}$

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

$$= (1 - 2a\lambda) u_j^n + a\lambda [u_{j-1}^n + u_{j+1}^n]$$

$h = \frac{b-a}{N}$

$x_j = jh$

$j = 0, 1, 2, \dots, N$

$t_n = nk$

$n = 0, 1, \dots, M$

$t \rightarrow t_n$

$t+k \rightarrow t_{n+1}$

$x \rightarrow x_j$

$x+h \rightarrow x_{j+1}$

$x-h \rightarrow x_{j-1}$

Explicit Scheme:-

Each value at

time level t_{n+1} can be independently calculated from values at time level t_n .

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Lecture #04

27-09-16

Finite Difference Method for parabolic Equations:-

(Heat Equation) $\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2}$

$u(0,t) = 0 = u(1,t) \rightarrow$ BCs

$u(x,0) = u_0(x) \rightarrow$ IC

$0 < x < 1$

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

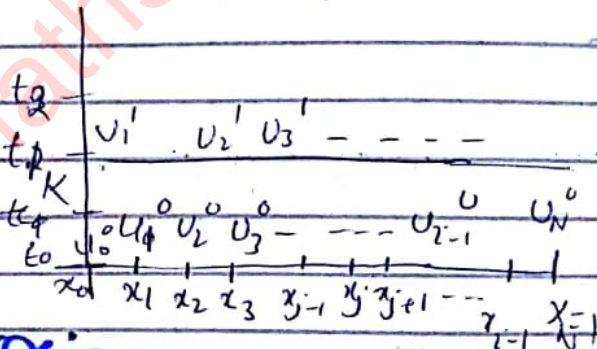
$= (1 - 2a\lambda) u_j^n + a\lambda [u_{j-1}^n + u_{j+1}^n]$

(Discrete solution)

explicit method

$O(k, h^2)$

First order in time and 2nd order in space.



Truncation Error:-

$L(u, x, t, k, h) =$ Truncation error produced by difference method

$\frac{\partial u}{\partial t} - a \frac{\partial^2 u}{\partial x^2} = 0$

$T.E = \frac{u(x, t+k) - u(x, t)}{k} - a \left[\frac{u(x-h, t) - 2u(x, t) + u(x+h, t)}{h^2} \right]$

$u(x, t+k) = u(x, t) + k u_t + \frac{k^2}{2!} u_{tt} + \frac{k^3}{3!} u_{ttt} + \frac{k^4}{4!} u_{tttt} + O(k^5)$

$u(x+h, t) = u(x, t) + h u_x + \frac{h^2}{2!} u_{xx} + \frac{h^3}{3!} u_{xxx} + \frac{h^4}{4!} u_{xxxx} + O(h^5)$

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$$\frac{u(x, t+k) - u(x, t)}{k} = u_t + \frac{k}{2} u_{tt} + \frac{k^2}{3!} u_{ttt} + \frac{k^3}{4!} u_{tttt} + O(k^4)$$

$$u(x-h, t) + u(x+h, t) = 2u + h^2 u_{xx} + \frac{2h^4}{4!} u_{xxxx} + O(h^6)$$

$$\frac{u(x-h, t) - 2u(x, t) + u(x+h, t))}{h^2} = u_{xx} + \frac{2}{4!} h^2 u_{xxxx} + O(h^4)$$

$$T.E = u_t + \frac{k}{2} u_{tt} + \frac{k^2}{6} u_{ttt} + \frac{k^3}{4!} u_{tttt} + O(k^4)$$

$$-a \left[u_{xx} + \frac{h^2}{12} u_{xxxx} + O(h^4) \right]$$

$$= (u_t - a u_{xx}) + \frac{k}{2} u_{tt} + O(k^3) - \frac{a h^2}{12} u_{xxxx} + O(h^4)$$

$$= O(k, h^2)$$

$$= \frac{k}{2} u_{tt} - \frac{a h^2}{12} u_{xxxx} + O(k^2, h^4) = O(k, h^2)$$

$$u_t = a u_{xx} \Rightarrow u_{tt} = a u_{xxt} = a (u_t)_{xx}$$

$$= a^2 u_{xxxx}$$

$$L(T.E) = \frac{k}{2} \cdot a^2 u_{xxxx} - a \frac{h^2}{12} u_{xxxx} + O(k^2, h^4)$$

$$= a \frac{k}{2} \left(\frac{a k}{h^2} - \frac{1}{6} \right) u_{xxxx} + O(k^2, h^4)$$

$$= a \frac{h^2}{2} \left(a \lambda - \frac{1}{6} \right) u_{xxxx} + O(k^2, h^4)$$

$$\text{if } \lambda = \frac{1}{6a} \Rightarrow T.E \sim O(k^2, h^4)$$

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$$U_j^{n+1} = (1 - 2a\lambda) U_j^n + a\lambda [U_{j-1}^n + U_{j+1}^n]$$

$$1 - 2a\lambda > 0$$

$$2a\lambda < 1$$

$$\lambda < \frac{1}{2a}$$

$$\frac{k}{h^2} < \frac{1}{2a}$$

$$k < \frac{h^2}{2a} \rightarrow \text{stability condition}$$

$$T, E \sim O(k^2, h^4)$$

(Courant-Friedrichs-Lewy condition)

Convergence:-

The scheme is convergent

if as $k \rightarrow 0, h \rightarrow 0$

$\forall (x^*, t^*)$, when $x_j \rightarrow x^*, t_n \rightarrow t^*$

numerical solution $U_j^n \rightarrow U(x^*, t^*) \rightarrow$ exact analytical.

$$\textcircled{1} \rightarrow \text{let } e_j^n = u(x_j, t_n) - U_j^n$$

=(Analytical sol. - numerical sol)

$$L_j^n = \frac{u(x_j, t_{n+1}) - u(x_j, t_n)}{k} - a \frac{u(x_{j-1}, t_n) - 2u(x_j, t_n) + u(x_{j+1}, t_n))}{h^2}$$

$$\textcircled{2} \rightarrow u(x_j, t_{n+1}) = (1 - 2a\lambda) u(x_j, t_n) + a\lambda [u(x_{j-1}, t_n) + u(x_{j+1}, t_n)] + k L_j^n$$

$$\textcircled{3} \quad U_j^{n+1} = (1 - 2a\lambda) U_j^n + a\lambda [U_{j-1}^n + U_{j+1}^n] + k L_j^n$$

②-③ by using ①

$$|e_j^{n+1}| = |(1 - 2a\lambda) e_j^n + a\lambda [e_{j-1}^n + e_{j+1}^n]|$$

$$\leq |1 - 2a\lambda| |e_j^n| + |a\lambda| (|e_{j-1}^n| + |e_{j+1}^n|) + k |L_j^n|$$

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$$\max_{1 \leq j \leq N} |e_j^{n+1}| \leq |1 - 2\alpha\lambda| \max_{1 \leq j \leq N} |e_j^n| + |\alpha\lambda|$$

$$\left(\max_{1 \leq j \leq N} |e_j^{n+1}| + \max_{1 \leq j \leq N} |e_j^n| \right) + k \max_{1 \leq j \leq N} |L_j^n|$$

$$E^{n+1} = \max_j |e_j^{n+1}| \quad \forall j$$

$$E^{n+1} \leq |1 - 2\alpha\lambda| E^n + |\alpha\lambda| (E^n + E^n) + k\tilde{L}$$

$$2\alpha\lambda < 1 \Rightarrow \lambda < \frac{1}{2\alpha} \Rightarrow k < \frac{h^2}{2\alpha}$$

then

$$E^{n+1} \leq (1 - 2\alpha\lambda) E^n + 2\alpha\lambda E^n + k\tilde{L}$$

(stability condition)

$$= E^n + k\tilde{L}$$

$$|x| = \begin{cases} x & x \geq 0 \\ -x & x < 0 \end{cases}$$

$$E^{n+1} \leq E^n + k\tilde{L}$$

$$E^n \leq E^{n-1} + k\tilde{L}$$

$$\leq E^{n-2} + 2k\tilde{L}$$

$$\leq E^0 + nk\tilde{L}$$

$$\text{if } E^0 = 0 \text{ then } E^n \leq nk\tilde{L} = \text{trunc} \tilde{L}$$

$$\text{if } k \rightarrow 0 \Rightarrow E^n \rightarrow 0$$

$$h \rightarrow 0 \tilde{L} \rightarrow 0$$

Monday (16)

Lecture #05

03-10-16

Stability Analysis:-

Fourier series:-

$$v_t = v_{xx}, \quad x \in \mathbb{R}, t \geq 0 \rightarrow \text{①}$$

$$v(x, 0) = f(x), \quad v(0, t) = 0 = v(\infty, t)$$

$$\hat{v}_t(\omega, t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-i\omega x} v_t(x, t) dx$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-i\omega x} v_{xx}(x, t) dx$$

Integrating by parts and terms at $x=0$ & ∞ are zero.

$$= -\frac{\omega^2}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{i\omega x} v(x, t) dx$$

$$= -\omega^2 \hat{v}(\omega, t) \quad (\text{PDE transform into ODE})$$

$$\hat{v}_t + \omega^2 \hat{v}(\omega, t) = 0 \rightarrow \text{(ODE)}$$

$$v(x, t) = \int_{-\infty}^{\infty} e^{i\omega x} \hat{v}(\omega, t) d\omega \leftarrow \text{inverse Fourier transform}$$

$$\text{Parseval's identity } \|\hat{v}\|_2 = \|v\|_2$$

$$\|\cdot\|_2 \rightarrow L_2(\mathbb{R})$$

$$\text{Continuous case} \rightarrow \hat{u}(\omega, t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-i\omega x} u(x, t) dx$$

$$\text{Discrete case} \rightarrow \hat{u}(f) = \frac{1}{\sqrt{2\pi}} \sum_{m=-\infty}^{\infty} e^{-imf} u_m, \quad f \in [-\pi, \pi]$$

Definition:-

The two level scheme $u = Q_n u_n$, $(n \geq 1)$ is said to be stable with respect

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to a norm $\|\cdot\|$ if there exist positive constant k and h , and non-negative constant β s.t

$$\|u^n\| \leq k e^{\beta t} \|u_0\|$$

also $\|\hat{u}^n\| \leq k e^{\beta t} \|\hat{u}_0\|$

Thus, both Fourier transform and Parseval's Identity are the two basic tools for stability analysis.

Explicit Finite Difference Method for Heat Equation:-

$$u_t = \alpha u_{xx} \quad u(x,0) = u_0(x), \quad u(0,t) = 0 = u(L,t)$$

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

$$\alpha \lambda < 1/2 \Rightarrow k < \frac{h^2}{2\alpha} \quad \text{CFL-condition}$$

$$u(\xi)^{n+1} = \frac{1}{\sqrt{2\pi}} \sum_{j=-\infty}^{\infty} e^{-i j \xi} u_j^{n+1} \quad \xi \in [-\pi, \pi]$$

$$\frac{1}{\sqrt{2\pi}} \sum_{j=-\infty}^{\infty} u_j^{n+1} e^{-i j \xi} = \frac{1}{\sqrt{2\pi}} \sum_{j=-\infty}^{\infty} e^{-i j \xi} \left\{ u_j^n + \alpha \lambda [u_{j-1}^n - 2u_j^n + u_{j+1}^n] \right\}$$

$$u^{n+1}(\xi) = (1 - 2\alpha\lambda) u^n(\xi) + \frac{\alpha\lambda}{\sqrt{2\pi}} \sum_{j=-\infty}^{\infty} e^{-i j \xi} [u_{j-1}^n + u_{j+1}^n]$$

substitute $m = j \mp 1 \Rightarrow j = m \pm 1$

$$u^{n+1}(\xi) = (1 - 2\alpha\lambda) u^n(\xi) + \alpha\lambda \frac{1}{\sqrt{2\pi}} \left[\sum_{m=-\infty}^{\infty} e^{-i(m+1)\xi} u_m^n + \sum_{m=-\infty}^{\infty} e^{-i(m-1)\xi} u_m^n \right]$$

$$+ \sum_{m=-\infty}^{\infty} u_m^n e^{-i(m-1)\xi}$$

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$$= (1 - 2a\lambda) \hat{u}^n(s) + a\lambda \left[\frac{e^{-is}}{\sqrt{2\pi}} \sum_{m=-\infty}^{\infty} + \frac{e^{is}}{\sqrt{2\pi}} \right] u_m^n e^{-im\gamma}$$

$$\hat{u}^{n+1}(s) = (1 - 2a\lambda) \hat{u}^n(s) + a\lambda (e^{-is} + e^{is}) \hat{u}^n(s)$$

$$\cos \gamma = \frac{1 - 2\sin^2 \frac{\gamma}{2}}{2} \quad = (1 - 2a\lambda) \hat{u}^n(s) + 2a\lambda \cos \gamma \hat{u}^n(s)$$

$$2\sin^2 \frac{\gamma}{2} = 1 - \cos \gamma$$

$$= 1 - 2a\lambda (1 - \cos \gamma) \hat{u}^n(s)$$

$$\hat{u}^{n+1}(s) = \left[1 - 4a\lambda \sin^2 \frac{\gamma}{2} \right] \hat{u}^n(s)$$

$$\hat{u}^{n+1}(s) = \left[1 - 4a\lambda \sin^2 \frac{\gamma}{2} \right] \hat{u}^n(s)$$

$$\hat{u}^{n+1}(s) = \left[1 - 4a\lambda \sin^2 \frac{\gamma}{2} \right] \hat{u}^{n-1}(s)$$

$$= \left[1 - 4a\lambda \sin^2 \frac{\gamma}{2} \right]^n \hat{u}^0(s)$$

Damping factor D

Stability $|D| < 1$

This means

$$\left| 1 - 4a\lambda \sin^2 \frac{\gamma}{2} \right| < 1 \quad \text{Stability Condition}$$

$$-1 < 1 - 4a\lambda \sin^2 \frac{\gamma}{2} < 1$$

$$\text{L.H.S} \Rightarrow 4a\lambda \sin^2 \frac{\gamma}{2} < 2$$

$$a\lambda < \frac{1}{2\sin^2 \frac{\gamma}{2}}$$

$$, \gamma \in (-\pi, \pi)$$

for $\gamma = \pm\pi$

$$a\lambda < 1/2$$

R.H.S

$$4a\lambda \sin^2 \frac{\xi}{2} > 0$$

for $\xi = \pm \pi$

$$a\lambda > 0$$

Thus $0 < a\lambda < \frac{1}{2}$

$$\|U^n\| \leq \|U^0\|$$

Exercise:-

$$u_t = a u_{xx}$$

$$u(x,0) = u_0(x), u(0,t) = 0 = u(1,t)$$

implicit scheme

$$u_j^{n+1} - u_j^n = a \left[\frac{u_{j-1}^{n+1} - 2u_j^{n+1} + u_{j+1}^{n+1}}{h^2} \right]$$

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

$$-a\lambda u_{j-1}^{n+1} + (1+2a\lambda)u_j^{n+1} - a\lambda u_{j+1}^{n+1} = u_j^n, j=1, \dots, N-1$$

$$\begin{bmatrix} 1+2a\lambda & -a\lambda & 0 & \dots & 0 \\ -a\lambda & 1+2a\lambda & -a\lambda & \dots & 0 \\ & & & \ddots & \\ & & & & 1+2a\lambda \end{bmatrix} \begin{bmatrix} u_1^{n+1} \\ u_2^{n+1} \\ \vdots \\ u_{N-1}^{n+1} \end{bmatrix} = \begin{bmatrix} u_1^n \\ u_2^n \\ \vdots \\ u_{N-1}^n \end{bmatrix}$$

put $j=1$

$$-a\lambda u_0^{n+1} + (1+2a\lambda)u_1^{n+1} - a\lambda u_2^{n+1} = u_1^n$$

$j=2$

$$-a\lambda u_1^{n+1} + (1+2a\lambda)u_2^{n+1} - a\lambda u_3^{n+1} = u_2^n$$

$$j=N-1 \quad -a\lambda u_{N-2}^{n+1} + (1+2a\lambda)u_{N-1}^{n+1} - a\lambda u_N^{n+1} = u_{N-1}^n$$

$$e^{-ij} = \cos \xi - \xi \sin \xi$$

Matrix becomes (20)

$$AU^{n+1} = B^n$$

$$\underline{U}^{n+1} = \underline{A}^{-1} \underline{B}^n$$

$$-a\lambda U_{j-1}^{n+1} + (1+2a\lambda)U_j^{n+1} - a\lambda U_{j+1}^{n+1} = U_j^n$$

fourier transform

$$-a\lambda e^{-i\xi} \hat{U}^{n+1}(\xi) + (1+2a\lambda)\hat{U}^{n+1}(\xi) - a\lambda e^{i\xi} \hat{U}^{n+1}(\xi) = \hat{U}^n(\xi)$$

$$-2a\lambda \cos \xi \hat{U}^{n+1}(\xi) + (1+2a\lambda)\hat{U}^{n+1}(\xi) = \hat{U}^n(\xi)$$

$$[2a\lambda(1-\cos \xi) + 1] \hat{U}^{n+1}(\xi) = \hat{U}^n(\xi)$$

$$\left[1 + 4a\lambda \sin^2 \frac{\xi}{2}\right] \hat{U}^{n+1}(\xi) = \hat{U}^n(\xi)$$

$$\hat{U}^{n+1}(\xi) = \frac{1}{1 + 4a\lambda \sin^2 \frac{\xi}{2}} \hat{U}^n(\xi)$$

$$\|\hat{U}^{n+1}(\xi)\| = \frac{1}{1 + 4a\lambda \sin^2 \frac{\xi}{2}} \|\hat{U}^n(\xi)\|$$

$$|D| = \frac{1}{1 + 4a\lambda \sin^2 \frac{\xi}{2}} \leq 1 \quad \forall \xi \in [-\pi, \pi]$$

\Rightarrow unconditionally stable

$\xi \in [-\pi, \pi]$

Lecture #06 Tuesday

(21)

04-10-16

Example:-

$$u_t = u_{xx}$$

$$u(x, 0) = f(x) = \begin{cases} 2x & 0 \leq x \leq 1/2 \\ 2-2x & 1/2 \leq x \leq 1 \end{cases}$$

$$u(0, t) = 0 = u(1, t)$$

Exact sol:- (By separation of variables)

$$u(x, t) = \sum_{n=1}^{\infty} C_n e^{-(n\pi)^2 t} \sin(n\pi x)$$

$$C_n = 2 \int_0^1 f(x) \sin(n\pi x) dx$$

$$= \frac{8}{n^2 \pi^2} \sin\left(\frac{n\pi}{2}\right), n=1, 2, \dots, \infty$$

Remaining

$$U_j^{n+1} = U_j^n + \nu [U_{j-1}^n - 2U_j^n + U_{j+1}^n]$$

$$\nu = 0.48$$

$$\nu = 1 \cdot \frac{\Delta x}{h^2}, \nu < 1/2$$

$$N = 20$$

$$h^2 = \frac{1-0}{2} = 0.05$$

$$K = \nu h^2 = 0.48 \times (0.05)^2 = 0.0012$$

Boundary

$$x_0 = 0, x_1 = 0.05, x_2 = 0.1, x_3 = 0.15, \dots, x_{20} = 1.0 \rightarrow \text{Boundary}$$

$$u(x, 0) = \begin{cases} 2x & 0 \leq x \leq 0.5 \\ 2(1-x) & 0.5 \leq x \leq 1 \end{cases}$$

$$u_0^0 = 0, u_1^0 = 0.1, u_2^0 = 0.2, u_3^0 = 0.3, \dots$$

$$u_0^n = 0$$

$$u_{20}^n = 0$$

$$n = 0, 1, \dots$$

$$\dots u_{10}^0 = 1, u_{11}^0 = 0.9, \dots, u_{20}^0 = 0$$

1-01-20

(22)

for $j=1$

$$\begin{aligned}
 u_1^1 &= u_1^0 + \nu [u_0^0 - 2u_1^0 + u_2^0] \\
 &= 0.1 + .48 [0 - 2(0.1) + 0.2] \\
 &= 0.1
 \end{aligned}$$

for $j=2$

$$u_2^1 = u_2^0 + \nu (u_1^0 - 2u_2^0 + u_3^0) = 0.2$$

for $j=3$

$$\begin{aligned}
 u_3^1 &= u_3^0 + \nu (u_2^0 - 2u_3^0 + u_4^0) \\
 &= 0.3 + .48 (.2 - 2(.3) + u_4^0) \\
 &= 0.3
 \end{aligned}$$

$$\|U^E - U^N\|_2 = \sqrt{\sum_{i=1}^{N-1} |U_i^E - U_i^N|^2}$$

MATLAB.

$$u_t = a u_{xx} \quad 0 \leq x \leq 1$$

$$u(x, 0) = \sin(\pi x)$$

$$u(0, t) = 0 = u(1, t)$$

$$a = 1$$

$$h = 0.02$$

$$\nu = \tau \frac{h^2}{2}$$

$$k = \tau \frac{h^2}{a}$$

$x = (0:h:1)$ is space variable.

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

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

(23)

$$= \gamma_2 u_j^n + \gamma [u_{j-1}^n + u_{j+1}^n]$$

Exact solution
 $= \sin(\bar{n}x) e^{-a\bar{n}^2 t}$

MONDAY.

Lecture # 07

17-10-16

- FDM $u_t = a u_{xx}$
- (1) Explicit Method: - conditionally stable (i.e. CFL condition needed) time step is usually small ($O(h^2)$)
- (2) Implicit Method: - unconditionally stable (no time-step restriction). T.E $\sim (k, h^2)$

θ-Method:-

$$\frac{u_j^{n+1} - u_j^n}{k} = \frac{a}{h^2} \left[(1-\theta) (u_{j-1}^n - 2u_j^n + u_{j+1}^n) + \theta (u_{j-1}^{n+1} - 2u_j^{n+1} + u_{j+1}^{n+1}) \right] \quad \theta \in [0, 1]$$

θ = 0, explicit method

θ = 1, implicit method

θ = 1/2, Crank-Nicholson method.

Crank - Nicholson Method :-

$$u_j^{n+1} - u_j^n = \frac{a\lambda}{2} \left[u_{j-1}^n - 2u_j^n + u_{j+1}^n + u_{j-1}^{n+1} - 2u_j^{n+1} + u_{j+1}^{n+1} \right] \quad \lambda = \frac{k}{h^2}$$

(24)

$$-\frac{\gamma}{2} U_{j-1}^{n+1} + (1+\gamma) U_j^{n+1} - \frac{\gamma}{2} U_{j+1}^{n+1} = \frac{\gamma}{2} U_{j-1}^n + (1-\gamma) U_j^n + \frac{\gamma}{2} U_{j+1}^n$$

$$\begin{bmatrix} 1+\gamma & -\gamma/2 & 0 \\ -\gamma/2 & 1+\gamma & -\gamma/2 \\ 0 & -\gamma/2 & 1+\gamma \end{bmatrix} \begin{bmatrix} U_1^{n+1} \\ \vdots \\ U_N^{n+1} \end{bmatrix} = \begin{bmatrix} \gamma/2 & 0 & 0 \\ \gamma/2 & 1-\gamma & 0 \\ 0 & \gamma/2 & 1-\gamma \end{bmatrix} \begin{bmatrix} U_1^n \\ \vdots \\ U_N^n \end{bmatrix}$$

$$\Rightarrow \underline{U}^{n+1} = \underline{A}^{-1} \underline{B} \underline{U}^n$$

Truncation Error:-

$$U_t = a U_{t+\Delta t}$$

Crank Nicolson Method:-

$$\frac{U_j^{n+1} - U_j^n}{\Delta t} = \frac{\gamma}{2} \left[U_{j-1}^n - 2U_j^n + U_{j+1}^n + U_{j-1}^{n+1} - 2U_j^{n+1} + U_{j+1}^{n+1} \right]$$

(25)

$$T.E = U(t+k, x) - U(t, x)$$

$$= \frac{\delta}{2} \left[U(t, x-h) - 2U(t, x) + U(t, x+h) \right]$$

$$+ U(t+k, x+h) - 2U(t+k, x-h)$$

$$- 2U(t+k, x) + U(t+k, x+h)]$$

$$\frac{\delta^2}{h}$$

$$U(t+k, x) = U + kU_t + \frac{k^2}{2!} U_{tt} + \frac{k^3}{3!} U_{ttt} + \frac{k^4}{4!} U_{tttt} + O(k^5)$$

$$U(t, x+h) = U + hU_x + \frac{h^2}{2!} U_{xx} + \frac{h^3}{3!} U_{xxx} + \frac{h^4}{4!} U_{xxxx} + O(h^5)$$

$$U(t+k, x) - U(t-k, x) = 2kU_t + \frac{2k^3}{3!} U_{ttt} + O(k^5)$$

$$\frac{U(t+k, x) - U(t-k, x)}{2k} = U_t + \frac{k^2}{3!} U_{ttt} + O(k^4)$$

$$U(t+k, x) + U(t-k, x) = 2U + k^2 U_{tt} + \frac{2k^4}{4!} U_{tttt} + O(k^6)$$

$$\frac{U(t+k, x) - U(t-k, x)}{2k} = U_t + \frac{2k^4}{4!} U_{tttt} + O(k^6)$$

(26)

$$U(t+k, n+h) - 2U(t+k, n) + U(t+k, n-h) \\ = h^2 U_{xx}(t+k, n) + \frac{2h^4}{4!} U_{xxxx}(t+k, n) + O(h^6)$$

$$U(t+k, n+h) - 2U(t+k, n) + U(t+k, n-h) \\ = h^2 U_{xx}(t+k, n) + \frac{2h^4}{4!} U_{xxxx}(t+k, n) + O(h^6)$$

$$= h^2 [U_{xx}(t, n) + k U_{xxt}(t, n) + O(k^2)] \\ + \frac{2h^4}{24!} [U_{xxxx}(t, n) + k U_{xxxxx}(t, n) + O(k^2)]$$

$$= h^2 U_{xx}(t, n) + O(h^4)$$

$$F.E = U_t + O(k^2) - \alpha U_{xx} + O(h^2)$$

$$= (U_t - \alpha U_{xx}) + O(k^2, h^2)$$

$$\stackrel{\approx 0}{=} O(k^2, h^2)$$

$$\approx O(k^2, h^2)$$

$$U_j^{n+1} - U_j^n = \frac{\alpha}{2} [U_{j-1}^{n+1} - 2U_j^{n+1} + U_{j+1}^{n+1} \\ + U_{j-1}^n - 2U_j^n + U_{j+1}^n]$$

$$+ U_{j-1}^n - 2U_j^n + U_{j+1}^n]$$

(27)

von-Neuman stability Analysis:

$$\begin{aligned} \hat{U}^{\hat{n}+1}(\xi) - \hat{U}^{\hat{n}}(\xi) &= \frac{\gamma}{2} \left[e^{-i\xi} \hat{U}^{\hat{n}+1}(\xi) \right. \\ &\quad \left. - 2 \hat{U}^{\hat{n}+1}(\xi) + e^{i\xi} \hat{U}^{\hat{n}+1}(\xi) + e^{-i\xi} \hat{U}^{\hat{n}}(\xi) - 2 \hat{U}^{\hat{n}}(\xi) \right. \\ &\quad \left. + e^{i\xi} \hat{U}^{\hat{n}}(\xi) \right] \\ &= \frac{\gamma}{2} \left[(2 \cos \xi - 2) \hat{U}^{\hat{n}+1}(\xi) + (2 \cos \xi - 2) \hat{U}^{\hat{n}}(\xi) \right. \\ &\quad \left. - 4 \sin^2 \frac{\xi}{2} \hat{U}^{\hat{n}+1}(\xi) + 4 \sin^2 \frac{\xi}{2} \hat{U}^{\hat{n}}(\xi) \right] \end{aligned}$$

$$= \left(1 + 2\gamma \sin^2 \frac{\xi}{2} \right) \hat{U}^{\hat{n}+1}(\xi)$$

$$= \left(1 - 2\gamma \sin^2 \frac{\xi}{2} \right) \hat{U}^{\hat{n}}(\xi)$$

$$\hat{U}^{\hat{n}+1}(\xi) = \left(\frac{1 - 2\gamma \sin^2 \frac{\xi}{2}}{1 + 2\gamma \sin^2 \frac{\xi}{2}} \right) \hat{U}^{\hat{n}}(\xi)$$

$$\hat{U}^{\hat{n}}(\xi) = \left(\frac{1 - 2\gamma \sin^2 \frac{\xi}{2}}{1 + 2\gamma \sin^2 \frac{\xi}{2}} \right)^{\hat{n}} \hat{U}^0(\xi)$$

$$|D| = \left| \frac{1 - 2\gamma \sin^2 \frac{\xi}{2}}{1 + 2\gamma \sin^2 \frac{\xi}{2}} \right| \leq 1$$

unconditionally stable

By θ -method (28)

$$U_j^{n+1} - U_j^n = \tau \left[\theta (U_{j-1}^{n+1} - 2U_j^{n+1} + U_{j+1}^{n+1}) + (1-\theta) (U_{j-1}^n - 2U_j^n + U_{j+1}^n) \right]$$

$$U^{n+1}(\xi) = \frac{1 - 4\tau(1-\theta)\sin^2 \xi/2}{1 + 4\tau\theta \sin^2 \xi/2}$$

if $1/2 \leq \theta \leq 1$ implicit method

$$(\theta \in [1/2, 1])$$

$0 \leq \theta < 1/2$ explicit method ($\theta \in [0, 1/2)$)

for $0 \leq \theta \leq 1/2$

|D| ≤ 1

$$-1 < \frac{1 - 4\tau(1-\theta)\sin^2 \xi/2}{1 + 4\tau\theta \sin^2 \xi/2} < 1$$

R.H.S $1 - 4\tau \sin^2 \xi/2 < 1$

$-4\tau \sin^2 \xi/2 \geq 0$ always hold.

L.H.S

$$0 < \frac{2 - 4\tau(1-2\theta)\sin^2 \xi/2}{2}$$

if $0 < \theta < 1/2$

$$\frac{1}{2}(1-2\theta) > 0$$

$$\tau \leq \frac{1}{2}$$

$$2(1-2\theta)\sin^2(\xi/2)$$

for $\xi = \pm \pi$

$$\gamma \leq \frac{1}{2(1-2\alpha)}$$

(1) Richardson Method:-

$$u_t = \alpha u_{xx}$$

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

(2) Dufort - Method:-

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

(1) Truncation - Error

(2) von - Neuman Stability (Assignment)

Tuesday

Lecture #08

18-10-16.

DuFort - Frankel method for
Heat Equation:-

$$u_t = \alpha u_{xx}$$

$$T.E = \frac{u_j^{n+1} - u_j^{n-1}}{2k} - \frac{\alpha}{h^2} [u_{j-1}^n - u_j^{n-1}$$

$$- u_j^n + u_{j+1}^n] \neq 0$$

$$= \frac{\alpha k^2}{h^2} u_{ttt} + O(k^2, h^2)$$

$$= \alpha^2 \lambda u_{ttt} + O(k^2, h^4)$$

The method is only consistent if

$k \rightarrow 0$ faster than h^2 .

otherwise the method is inconsistent

because the method solves the

following PDE

$$u_t - a u_{xx} + a \eta u u_x = 0 \quad \eta = \frac{k}{h} \rightarrow 0$$

if $k \rightarrow 0$
faster than h

Stability Analysis:-

$$\frac{u_j^{n+1} - u_j^{n-1}}{2k} = \frac{a}{h^2} \left[u_{j-1}^n u_j^{n+1} - u_j^{n-1} + u_{j+1}^n \right] \neq 0$$

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

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

$$\hat{u}^{n+1}(\xi) = \frac{1}{1+2\tau} \left[(1-2\tau) \hat{u}^{n-1}(\xi) + 4\tau \cos \xi \hat{u}^n(\xi) \right]$$

$$\hat{u}^n(\xi) = \hat{u}^n(\xi)$$

$$\begin{pmatrix} \hat{u}^{n+1}(\xi) \\ \hat{u}^n(\xi) \end{pmatrix} = \begin{pmatrix} \frac{4\tau \cos \xi}{1+2\tau} & \frac{1-2\tau}{1+2\tau} \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \hat{u}^n(\xi) \\ \hat{u}^{n-1}(\xi) \end{pmatrix}$$

$$|M - \lambda I| = 0$$

M.

(3)

$$X^{k+1} = M X^{(k)}$$

↑
iteration matrix.

$$\lambda^2 - \frac{4r \cos \theta}{1+2r} \lambda - \frac{1-2r}{1+2r} = 0.$$

$$\lambda_{\pm} = \frac{4r \cos \theta}{1+2r} \pm \sqrt{\frac{16r^2 \cos^2 \theta}{(1+2r)^2} + \frac{4(1-2r)}{1+2r}}$$

$$= \frac{2r \cos \theta \pm \sqrt{1 - 4r^2 \sin^2 \theta}}{1+2r}$$

if $1 - 4r^2 \sin^2 \theta < 0$

$$\lambda_{\pm} = \frac{1}{1+2r} \left(2r \cos \theta \pm \sqrt{4r^2 \sin^2 \theta - 1} \right)$$

$$|\lambda_{\pm}| = \sqrt{\frac{1}{(1+2r)^2} (2r \pm 1)^2}$$

$$|\lambda_{\pm}| = \frac{2r+1}{1+2r} = 1$$

if $1 - 4r^2 \sin^2 \theta > 0$

$$r^2 > \frac{1}{4 \sin^2 \theta}$$

$$r > \frac{1}{2} \Rightarrow 2r > 1$$

$$|\lambda_{\pm}|^2 = \frac{4r^2}{(1+2r)^2} \cos^2 \theta + \frac{4r^2 \sin^2 \theta - 1}{(1+2r)^2}$$

$$\Rightarrow \frac{4r^2 - 1}{(1+2r)^2} = \frac{2r-1}{1+2r}$$

if $1 - 4r^2 \sin^2 \theta \geq 0 \Rightarrow |1 - 4r^2 \sin^2 \theta| \leq 1$

$$|\lambda + 1| \leq \frac{1}{1+2r} (2r|\cos \theta| + \sqrt{1 - 4r^2 \sin^2 \theta})$$

$$\leq \frac{1}{1+2r} (2r \cdot 1 + 1)$$

This is unconditionally stable.

Monday

Lecture # 09

24-10-16.

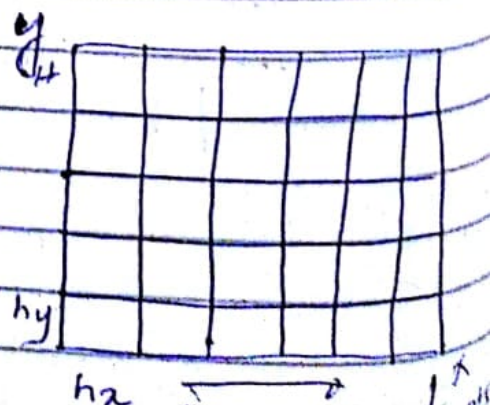
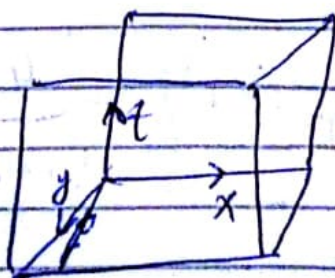
Two-Dimensional Heat Equation:-

$$u_t = a(u_{xx} + u_{yy}), \quad 0 \leq x \leq L, \quad 0 \leq y \leq H$$

B.C $u(x, y, t) = g(x, y, t)$ on boundary

I.C $u(x, y, 0) = f(x, y), \quad 0 \leq x \leq L, \quad 0 \leq y \leq H$

$$h_x = \frac{L-0}{N}, \quad h_y = \frac{H-0}{M}$$



(33)

$$u_t = a(u_{xx} + u_{yy})$$

Explicit method time

$$u(x_i, y_j, t_n) = u_{i,j}^n$$

$$x_i = i h_x, \quad y_j = j h_y$$

$$t_n = n k$$

$$i, j, n = 0, 1, 2, \dots$$

$$\frac{u_{i,j}^{n+1} - u_{i,j}^n}{k} = a \left[\frac{u_{i-1,j}^n - 2u_{i,j}^n + u_{i+1,j}^n}{h_x^2} \right.$$

$$\left. + \frac{u_{i,j-1}^n - 2u_{i,j}^n + u_{i,j+1}^n}{h_y^2} \right]$$

$$u_{i,j}^{n+1} - u_{i,j}^n = r_x (u_{i-1,j}^n - 2u_{i,j}^n + u_{i+1,j}^n)$$

$$+ r_y (u_{i,j-1}^n - 2u_{i,j}^n + u_{i,j+1}^n)$$

$$u_{i,j}^{n+1} = r_x u_{i-1,j}^n + (1 - 2r_x - 2r_y) u_{i,j}^n$$

$$+ r_x u_{i+1,j}^n + r_y u_{i,j-1}^n + r_y u_{i,j+1}^n$$

$$i = 1, 2, \dots, N-1$$

$$j = 1, 2, \dots, M-1$$

$$T.E.N(k, h_x, h_y)$$

Stability Analysis:-

$$\hat{u}(\xi, \eta) \text{ and } \xi, \eta \in [-\pi, \pi]$$

$$\hat{u}^{n+1}(\xi, \eta) = r_x e^{-i\xi h_x} \hat{u}^n(\xi, \eta) + (1 - 2r_x - 2r_y) \hat{u}^n(\xi, \eta)$$

$$+ r_x e^{i\xi h_x} \hat{u}^n(\xi, \eta) + r_y e^{-i\eta h_y} \hat{u}^n(\xi, \eta) + r_y e^{i\eta h_y} \hat{u}^n(\xi, \eta)$$

(34)

$$2rx \cos \theta + (1 - rx - 2ry + 2ry \cosh \eta) u^n(s, \eta)$$

$$= (-rx \sin^2 \frac{\theta}{2} - ry \cosh^2 \frac{\eta}{2} + 1) u^n(s, \eta)$$

By induction

$$u^n(s, \eta) = (1 - 4rx \sin^2 \frac{\theta}{2} - 4ry \cosh^2 \frac{\eta}{2})^n$$

$$|D| < 1$$

$$|1 - 4rx \sin^2 \frac{\theta}{2} - 4ry \cosh^2 \frac{\eta}{2}| < 1$$

$$-1 < 1 - 4rx \sin^2 \frac{\theta}{2} + ry \cosh^2 \frac{\eta}{2} < 1$$

L.H.S

$$4rx \sin^2 \frac{\theta}{2} + 4ry \cosh^2 \frac{\eta}{2} < 2$$

$$r \neq u = \frac{a}{h}$$

$$4rx + 4ry < 2$$

$$r + ry < \frac{1}{2}$$

$$\frac{ak}{h^2} < \frac{1}{4} \checkmark$$

$$k < \frac{h^2}{4a}$$

$$r_x = \frac{a}{h^2}$$

$$r_y = \frac{ak}{h^2}$$

$$r = \frac{ak}{h^2}$$

$$h_x = h_y = h$$

$$r < \frac{1}{2}$$

$$r < \frac{1}{2^2}$$

$$r < \frac{1}{2^d}$$

Implicit Method:- Crank-Nicolson method

$$u_t = a(u_{xx} + v_{yy})$$

$$\frac{u_{ij}^{n+1} - u_{ij}^n}{k} = a \left[\frac{\delta x^2 u_{ij}^n + \delta x^2 u_{ij}^{n+1}}{2h_x^2} + \frac{\delta y^2 u_{ij}^n + \delta y^2 u_{ij}^{n+1}}{2h_y^2} \right]$$

$$\delta x^2 u_{ij} = u_{i-1,j} - 2u_{ij} + u_{i+1,j}$$

$$\delta y^2 u_{ij} = u_{i,j-1} - 2u_{ij} + u_{i,j+1}$$

This method is

- (1) unconditionally stable
- (2) 2nd order accurate in space & time
- (3) Large banded matrix (Band is larger)
- (4) If a is constant then the coefficient matrix is larger and we need to factor the matrix once

To overcome these problems we can use Alternating direction method (ADI)

ADI - Method:-

The idea is to alternate direction and thus solve one dimensional problem at each time step

The first step is to keep y -fixed.

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$$U_{ij}^{n+1/2} - U_{ij}^n = a \left(\frac{\delta x^2 U_{ij}^{n+1/2}}{h_x^2} + \frac{\delta y^2 U_{ij}^n}{h_y^2} \right)$$

on the second step we keep x fixed

$$\frac{U_{ij}^{n+1} - U_{ij}^{n+1/2}}{\tau/2} = a \left(\frac{\delta x^2 U_{ij}^{n+1/2}}{h_x^2} + \frac{\delta y^2 U_{ij}^n}{h_y^2} \right)$$

$$U_{ij}^{n+1/2} = U_{ij}^n + \frac{\tau a}{2} \left(U_{i-1,j}^{n+1/2} + 2U_{ij}^{n+1/2} \right)$$

$$+ \tau a \left(\frac{U_{i,j-1}^n - 2U_{ij}^n + U_{i,j+1}^n}{2} \right)$$

$$= \frac{1}{2} \tau a U_{i-1,j}^{n+1/2} + (1 + \tau a) U_{ij}^{n+1/2} - \frac{1}{2} \tau a U_{i+1,j}^{n+1/2}$$

$$- \frac{1}{2} \tau a \delta y U_{i,j-1}^n + (1 - \tau a \delta y) U_{ij}^n$$

$$+ \frac{1}{2} \tau a \delta y U_{i,j+1}^n$$

$$\underline{A} U^{n+1/2} = \underline{B} U^n$$

so we have a tri-diagonal system and every time step we have to order the unknowns

differentially at every timestep. The method is 2nd order in

time and space and it is unconditionally stable, since the

denominator is always larger than numerator

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High order Difference Formulas:-

$$T_{\pm h} u = u(x \pm h)$$

$$\Delta_+ u = u(x+h) - u(x)$$

$$\Delta_- u = u(x) - u(x-h)$$

$$\Delta_c u = u(x+h) - u(x-h)$$

$$\ln(x) = (x-1) - \frac{(x-1)^2}{2} + \frac{(x-1)^3}{3} - \dots$$

$$\ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \dots$$

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$$

$$\begin{aligned} T_h u(x) &= u(x+h) - u(x) \\ &+ u(x) \\ &= \Delta_h u(x) + u(x) \\ &= (1 + \Delta_h) u(x) \end{aligned}$$

$$\begin{aligned} T_h u(x) &= u(x+h) \\ &= u(x) + hu_x + \frac{h^2}{2!} u_{xx} + \dots \\ &= (1 + h \frac{\partial}{\partial x} + \frac{h^2}{2!} \frac{\partial^2}{\partial x^2} + \dots) u(x) \\ &= e^{h \frac{\partial}{\partial x}} u(x) \end{aligned}$$

$$e^{h \frac{\partial}{\partial x}} = (1 + \Delta_h)$$

$$T_h = e^{h \frac{\partial}{\partial x}}$$

$$h \frac{\partial}{\partial x} = \ln(1 + \Delta_h)$$

$$= \Delta_h - \frac{\Delta_h^2}{2} + \frac{\Delta_h^3}{3} - \dots$$

$$\begin{aligned} \Delta_h^2 u &= \Delta_h(u(x+h) - u(x)) \\ &= u(x+2h) - u(x+h) - u(x+h) + u(x) \end{aligned}$$

$$\frac{\partial}{\partial x} = \frac{\Delta_h}{h} - \frac{\Delta_h^2}{2h} + \frac{\Delta_h^3}{3h} - \dots$$

$$\frac{\partial y}{\partial x} = \frac{1}{h} \Delta_h u - \frac{1}{2h} \Delta_h^2 u + \frac{1}{3h} \Delta_h^3 u - \dots$$

Tuesday (38)

25-10-16

Lecture #10

Formal methods for higher order difference formula:-

$$\ln(x) = (x-1) - \frac{(x-1)^2}{2} + \frac{(x-1)^3}{3} - \frac{(x-1)^4}{4} + \dots$$

$$\ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{4} + \dots$$

$$T_h u(x) = u(x+h), \quad T_h u(x) = u(x-h)$$

$$T_h u(x) = u(x+h) = u(x) + hu(x) + \frac{h^2}{2!} u''(x) + \dots$$

$$= \left[1 + h \frac{\partial}{\partial x} + \frac{h^2}{2!} \frac{\partial^2}{\partial x^2} + \dots \right] u(x)$$

$$= e^{h \frac{\partial}{\partial x}} u(x) \quad \text{--- (2)}$$

$$T_h u(x) = u(x+h) = u(x) + \Delta_h u(x)$$

$$= \Delta_h u(x) + u(x)$$

$$= (1 + \Delta_h) u(x) \quad \text{--- (2)}$$

Comparing (1) & (2)

$$e^{h \frac{\partial}{\partial x}} = (1 + \Delta_h) \Rightarrow \frac{\partial}{\partial x} = \frac{1}{h} \ln(1 + \Delta_h)$$

$$= \frac{1}{h} \Delta_h - \frac{\Delta_h^2}{2h} + \frac{\Delta_h^3}{3h} - \frac{\Delta_h^4}{4h} + \dots$$

$$\frac{\partial}{\partial x} u(x) = \left[\frac{\Delta_h}{h} - \frac{\Delta_h^2}{2h} + \frac{\Delta_h^3}{3h} - \frac{\Delta_h^4}{4h} + \dots \right] u(x)$$

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First order approximation of U_x .

$$U_x = \frac{\Delta h}{h} U(x) = \frac{U(x+h) - U(x)}{h} \quad \text{forward diff} \\ + O(h)$$

2nd order

$$U_x = \frac{\Delta h}{h} U(x) - \frac{\Delta^2 h}{2h} U(x) \\ = \frac{U(x+h) - U(x)}{h} - \frac{1}{2h} [U(x) - 2U(x+h) + U(x+2h)] \\ \Delta^2 h U(x) = \Delta h [U(x+h) - U(x)] \left[\begin{array}{l} U_x = -3U(x) + 4U(x+h) \\ - \frac{U(x+2h)}{2h} + O(h) \end{array} \right] \\ = U(x+2h) - U(x+h) - U(x+h) + U(x) \\ = U(x+2h) - 2U(x+h) + U(x)$$

Analogously for $\Delta_- h$ \Rightarrow Higher order backward difference formula.

$$\Delta_- U(x) = U(x+h) - U(x-h) \quad (\text{Do yourself})$$

$$= T_h U(x) - T_{-h} U(x)$$

$$= (T_h - T_{-h}) U(x)$$

$$= (e^{h\partial_x} - e^{-h\partial_x}) U(x)$$

$$= 2 \sinh h(h\partial_x) U(x)$$

$$h\partial_x = \text{arc sinh} \left(\frac{\Delta_-}{2} \right)$$

$$\text{arc sinh } x = x - \frac{1}{2 \cdot 3} x^3 + \frac{1 \cdot 3}{2 \cdot 4 \cdot 5} x^5 + \dots \\ = \sum_{k=0}^{\infty} (-1)^k \frac{(2k)!}{2^{2k} (k!)^2 (2k+1)} x^{2k+1}$$

$$\partial x = \frac{\Delta c}{2h} - \frac{1}{6h} \left(\frac{\Delta c}{2}\right)^3 + \frac{3}{40} \left(\frac{\Delta c}{2}\right)^5 + \dots$$

Finite difference methods for linear advection equation:-

$$u_t + a u_x = 0$$

$$0 < f(x) = f(x)$$

$$\frac{du}{dt} = \frac{\partial u}{\partial t} + \frac{\partial u}{\partial x} \frac{dx}{dt}$$

↓
source of sink

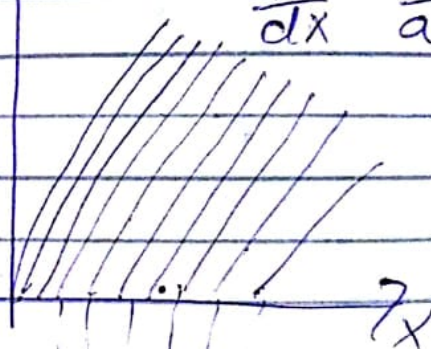
$$= \frac{\partial u}{\partial t} + \frac{dx}{dt} \frac{\partial u}{\partial x}$$

$$= u_t + a u_x = 0$$

$$u_t + a u_x = f(x, t) \delta(x - x_0) t \uparrow$$

$$\frac{dt}{dx} = \frac{1}{a}$$

- transport eq.
- convection eq.
- Advection eq.



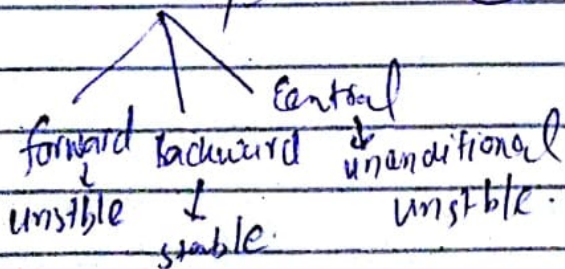
I.C $u_t + a u_x = 0$

$$u(0, t) = 0 = u(b, t) \quad x \in [a, b]$$

$$u(x, 0) = u_0(x)$$

$$t \in (0, 1)$$

- ① Explicit ② implicit ③ Crank Nicolson



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ADVECTION

Finite difference methods for linear advection equation:

Lecture # 11
31-10-16.

$$u_t + a u_x = 0$$

$$u(x, 0) = f(x)$$

$$\frac{\partial u}{\partial x} \Big|_{x=0} = 0, \quad \frac{\partial u}{\partial x} \Big|_{x=L} = 0$$

FDMs:-

Explicit method:

$$\frac{u_j^{n+1} - u_j^n}{k} = -a \begin{cases} \frac{u_j^n - u_{j-1}^n}{h} & \text{Backward diff.} \\ \frac{u_{j+1}^n - u_j^n}{h} & \text{forward diff.} \end{cases}$$

$$u_j^{n+1} = u_j^n - a\lambda \begin{cases} u_j^n - u_{j-1}^n \\ u_{j+1}^n - u_j^n \end{cases}$$

$$u_j^{n+1} = \begin{cases} (1 - a\lambda) u_j^n + a\lambda u_{j-1}^n \\ (1 + a\lambda) u_j^n - a\lambda u_{j+1}^n \end{cases}$$

if $a > 0 \Rightarrow$ Backward diff. is stable and forward diff. is unstable.

if $a < 0 \Rightarrow$ Backward diff. unstable and forward is stable.

T.E $\sim O(k, h)$ for both forward and backward diff formulas

Stability :-

~~T.E $\sim O(k, h)$~~

$$u_j^{n+1} = (1 - a\lambda) u_j^n + a\lambda u_{j-1}^n$$

$$\begin{aligned}
 \textcircled{42} \\
 u^{n+1}(\xi) &= (1-a\lambda)u^n(\xi) + a\lambda e^{-i\xi} u^n(\xi) \\
 &= (1-a\lambda) + a\lambda (\cos\xi - i\sin\xi) \Big] u^n(\xi) \\
 &= \left\{ (1-a\lambda) + a\lambda \cos\xi \right\} - i a\lambda \sin\xi \Big] u^n(\xi)
 \end{aligned}$$

By induction

$$u^n(\xi) = \left\{ (1-a\lambda) + a\lambda \cos\xi \right\}^n - i a\lambda \sin\xi \Big] u^0(\xi)$$

$$\begin{aligned}
 |D|^2 &= (1-a\lambda)^2 + a^2 \lambda^2 \cos^2\xi + 2a\lambda(1-a\lambda)\cos\xi \\
 &\quad + a^2 \lambda^2 \sin^2\xi
 \end{aligned}$$

$$= (1-a\lambda)^2 + a^2 \lambda^2 + 2a\lambda(1-a\lambda)\cos\xi$$

let $a\lambda = R$

$$|D(\xi)|^2 = (1-R)^2 + R^2 + 2R(1-R)\cos\xi$$

$$|D^2(\xi)|' = -2R(1-R)\sin\xi = 0$$

$$\Rightarrow \xi = -\pi, 0, \pi$$

for $\xi = 0$

$$|D|^2 = (1-R)^2 + R^2 + 2R(1-R)$$

$$= 1$$

for $\xi = \pm \pi$

$$|D(\xi)|^2 = (1-R)^2 + R^2 - 2R(1-R)$$

$$= (1-2R)^2$$

$$|D(\xi)| = |1-2R| < 1$$

$$-1 < 1-2R < 1$$

$$R > 0, \quad 2R < 2, \quad \Rightarrow R < 1$$

$$\Rightarrow 0 < R < 1$$

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$$\Rightarrow a k / h < 1 \Rightarrow k \leq h/a$$

Central difference (EXPLICIT).

$$\frac{U_j^{n+1} - U_j^n}{k} = -\frac{a}{2h} [U_{j+1}^n - U_{j-1}^n] \sim O(k, h^2)$$

stability :-

$$U^{n+1}(\xi) - U^n(\xi) = -\frac{a\lambda}{2} [e^{i\xi} - e^{-i\xi}] U^n(\xi)$$

$$= -i a \lambda \sin \xi U^n(\xi)$$

$$U^n(\xi) = [1 - i a \lambda \sin \xi]^n U^0(\xi)$$

$$D = 1 - i a \lambda \sin \xi$$

$$|D|^2 = 1 + a^2 \lambda^2 \sin^2 \xi \leq 1 \quad \forall \xi \in [-\pi, \pi]$$

unconditionally unstable.

Q:- can we stabilize this method?

Ans:- Yes, there are two ways
(Lax - Friedrich Method)

- (1) $U(t, x)$ Modification s.t in truncation error the coefficient of U_{xx} is less than or equal to one (≤ 1)
- (2) one can also represent (formulate) the central difference method as a method with inherent diffusion (Lax, Wendroff method)

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Method 1:-

$$\frac{U_j^{n+1} - U_j^n}{k} = -\frac{a}{2h} \left[U_{j+1}^n - U_{j-1}^n \right]$$

$$U_j^{n+1} = \frac{U_{j+1}^n + U_{j-1}^n}{2}$$

$$U_j^{n+1} = \frac{U_{j-1}^n + U_{j+1}^n}{2} - \frac{a\tau}{2} \left[U_{j+1}^n - U_{j-1}^n \right]$$

Lax - Friedrich: Method

and T.E $\sim O(k, h)$
stable conditionally.

Method 2:-

$$\left[\delta_k + \delta_h^c - \frac{ka^2}{2!} \delta_h \delta \right] U(x, t) = 0$$

$$\frac{U_j^{n+1} - U_j^n}{k} + \frac{a}{2h} \left[U_{j+1}^n - U_{j-1}^n \right] - \frac{ka^2}{2h^2} \left[U_{j+1}^n - 2U_j^n + U_{j-1}^n \right] = 0$$

$$U_j^{n+1} = U_j^n - \frac{a\tau}{2} \left[U_{j+1}^n - U_{j-1}^n \right]$$

$$+ \frac{a^2\tau^2}{2} \left[U_{j-1}^n - 2U_j^n + U_{j+1}^n \right] \text{ Lax - Wendroff method.}$$

T.E $\sim O(k, h^2)$

stable (conditionally)!

(1) Implicit method:-

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$$\frac{u_j^{n+1} - u_j^n}{h} = -a\lambda [u_j^{n+1} - u_{j-1}^{n+1}]$$

unconditionally stable

$$u_j^{n+1} - u_j^n = -a\lambda [u_{j+1}^{n+1} - u_j^{n+1}]$$

Implicit Backward diff:

$$u_j^{n+1} - u_j^n = -a\lambda [u_j^{n+1} - u_{j-1}^{n+1}]$$

$$|D|^2 = \frac{1}{1 + 4R(R+1)\sin^2 \frac{\xi}{2}} \leq 1 \quad \forall \xi$$

unconditionally stable. $R = a\lambda = \frac{ak}{h^2 \tau}$

Implicit forward diff:

$$u_j^{n+1} - u_j^n = -a\lambda [u_{j+1}^{n+1} - u_j^{n+1}]$$

$$|D|^2 = \frac{1}{1 + 4R(R-1)\sin^2 \frac{\xi}{2}} < 1$$

for stability $R \geq 1$

The method is unstable when

$$0 < R < 1$$

Implicit central difference.

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

unconditionally stable.

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General 3-point method ($x-h, x, x+h$)

method of unknown coefficients

$$u(x, t+k) = C_0 u(x+h, t) + C_1 u(x, t) + C_{-1} u(x-h, t)$$

$$u + k u_t + \frac{k^2}{2!} u_{tt} + \frac{k^3}{3!} u_{ttt} + O(k^3)$$

$$= C_0 \left[u + h u_x + \frac{h^2}{2!} u_{xx} + O(h^3) \right] + C_1 u$$

$$+ C_{-1} \left[u - h u_x + \frac{h^2}{2!} u_{xx} - O(h^3) \right]$$

$$= (C_0 + C_1 + C_{-1}) u + h (C_1 - C_{-1}) u_x + \frac{h^2}{2!} (C_0 + C_{-1})$$

$$u_{xx} + O(h^3)$$

on comparing the coefficients, we obtain

$$C_{-1} + C_0 + C_1 = 1$$

$$C_1 - C_{-1} = a \Delta t$$

$$C_0 + C_{-1} = a^2 \Delta t^2$$

$$\begin{cases} u_t + a u_x = 0 \\ u_t = -a u_x \\ u_{tt} = -a (u_t)_x \\ = -a^2 u_{xx} \end{cases}$$

order? (i) zeroth-order method

$$C_{-1} + C_0 + C_1 = 1 \Rightarrow \Delta \text{-free parameter}$$

(ii) 1st order.

u varies linearly

$$C_{-1} + C_0 + C_1 = 1$$

$$C_1 - C_{-1} = a \Delta t$$

} one-free parameter

$C_0 = 0 \Rightarrow$ Lax Friedrich's method

$C_1 = 0 \Rightarrow$ Backward difference method

$C_{-1} = 0 \Rightarrow$ forward difference method

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2nd order accurate

3 - equation for unknowns.

$$C_1 = -\frac{\lambda a}{2} + \frac{\lambda^2 a^2}{2}$$

$$C_1 = \frac{\lambda a}{2} + \frac{\lambda^2 a^2}{2}$$

$$C_0 = 1 - \lambda^2 a^2$$

Remark:- one can use more points to get higher order methods

$$u(x, t+k) = C_2 u(x+2h, t) + C_1 u(x+h, t) + C_0 u(x, t) + C_1 u(x-h, t) + C_2 u(x-2h, t)$$

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Lecture #12

07-11-16

$$u_t + au_x = 0, \quad a = 1$$

$$u(x, 0) = \sin(2\pi x)$$

$$u(0, t) = 0 \quad \frac{\partial u}{\partial x}(1, t) = 0$$

$$h = 0.2, \quad \lambda = \frac{ak}{h} = 0.1 \Rightarrow k = 0.02$$

$$t_{\max} = 0.06 \quad N = 5$$

$$n = 3$$

$$t_{\max} = 0.06$$

$$u_0^n = 0 = u_5^n$$

we need three iterations to reach t_{\max}

Explicit method

Backward

$$u_j^{n+1} = u_j^n - a\lambda [u_j^n - u_{j-1}^n]$$

$$n = 0$$

$$u_j^1 = u_j^0 - a\lambda [u_j^0 - u_{j-1}^0], \quad j = 1, 2, 3, 4$$

$$u_1^1 = u_1^0 - 0.1 [u_1^0 - u_0^0], \quad u_0^0 = 0$$

$$= 0.95 - 0.1 [0.95 - 0], \quad u_1^0 = 0.95$$

$$= 0.85$$

$$u_2^0 = 0.587$$

$$u_2^1 = 0.587 - 0.1 [0.587 - 0.95]$$

$$= 0.623$$

$$u_3^0 = 0.586$$

$$u_3^1 = -0.47$$

$$u_4^0 = -0.95$$

$$u_4^1 = 0.0914$$

$$u_5^0 = 0$$

$$u_5^1 = u_4^1$$

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$$n=1$$

$$U_1^2 = U_1^1 - 0.1 [U_1^1 - U_0^1]$$

$$U_2^2 = U_2^1 - 0.1 [U_2^1 - U_1^1]$$

$$U_3^2 = U_3^1 - 0.1 [U_3^1 - U_2^1]$$

$$U_4^2 = U_4^1 - 0.1 [U_4^1 - U_3^1]$$

$$U_5^2 = U_4^2$$

$$U_0^1 = 0$$

$$U_1^1 = .85$$

$$U_2^1 = .623$$

$$U_3^1 = -0.47$$

$$U_4^1 = -0.98$$

$$U_5^1 = -0.98$$

Exact:

$$U_0^e = \sin(2\pi(-0.06)) =$$

$$U_1^e = \sin(2\pi(0.2 - 0.06)) =$$

$$U_2^e = \sin(2\pi)(0.4 - 0.06) =$$

$$U_3^e = \sin(2\pi)(0.6 - 0.06) =$$

$$U_4^e = \sin(2\pi)(.8 - 0.06) =$$

X	U^N	U^e	Error $ U^e - U^N $
0			

(50)

$$u_t + a u_x = 0$$

$$a \neq 0$$

$$u(t+k, x) = C_1 u(t, x-h) + C_0 u(t, x) + C_2 u(t, x+h)$$

①

$$+ C_1 u(t, x+h)$$

$$u(t+k, x) = u(t, x) - \frac{v}{2} [u(t, x+h) - u(t, x-h)]$$

②

(Alternative method)

$$+ \frac{Q}{2} [u(t, x+h) - 2u(t, x) + u(t, x-h)]$$

(Coefficient of numerical diffusion)

2nd derivative (numerical diffusion)

$$C_1 = \frac{Q-v}{2}, C_0 = 1-Q, C_2 = \frac{Q+v}{2}$$

By comparing eq ① and eq ②

$$C_1 + C_0 + C_2 = 1$$

$$Q - C_1 = -v = -\lambda a$$

$$C_1 + C_2 = Q = a^2 \lambda^2 = v^2$$

$$u(t+k, x) = \left[(1-Q) + \frac{Q-v}{2} \tau_h + \frac{Q+v}{2} \tau_{-h} \right] u(t, x)$$

$$\hat{u}(t+k, \xi) = \left[(1-Q) + \frac{Q-v}{2} \hat{0}^{\xi} + \frac{Q+v}{2} \hat{0}^{\xi} \right] \hat{u}(t, \xi) \Big|_{\tau_{\pm h}} \tau u = u(t, x \pm h)$$

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$$D = (1-Q) + \frac{Q-v}{2} e^{i\delta} + \frac{Q+v}{2} e^{-i\delta}$$

$$= 1 - i2v \sin \frac{\delta}{2} \left[\frac{1}{2} \cos \frac{\delta}{2} - iQ \sin \frac{\delta}{2} \right]$$

$$= 1 - 2Q \sin^2 \frac{\delta}{2} - i2v \sin \frac{\delta}{2} \cos \frac{\delta}{2}$$

Stability $\Rightarrow |D|^2 < 1$

$$1 - 4Q \sin^2 \frac{\delta}{2} + 4Q^2 \sin^4 \frac{\delta}{2}$$

$$+ 4v^2 \sin^4 \frac{\delta}{2} (1 - \sin^2 \frac{\delta}{2}) \leq 1$$

$$-Q + Q^2 \sin^2 \frac{\delta}{2} + v^2 (1 - \sin^2 \frac{\delta}{2}) \leq 0$$

$$\delta = 0, \pi, 2\pi$$

$$\delta = 0 \quad v^2 - Q \leq 0 \Rightarrow Q \geq v^2 > 0$$

$$\delta = \pi \Rightarrow v^2 - Q + Q^2 - v^2 \leq 0$$

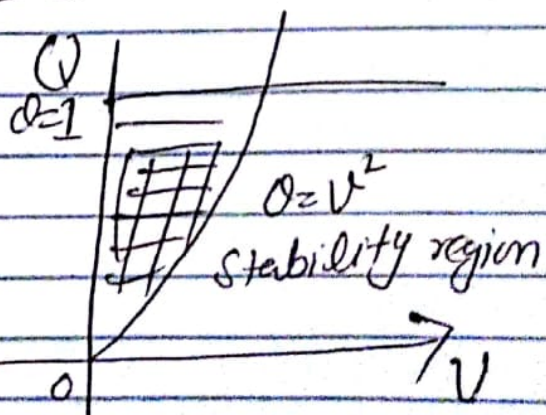
$$\Rightarrow Q(Q-1) \leq 0$$

$$\Rightarrow Q \leq 1$$

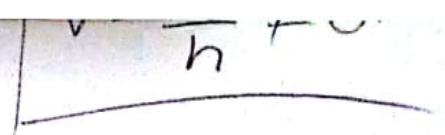
$$y_j^{n+1} = y_j^n + \frac{\Delta t}{\Delta x} \left[v_{j+1/2}^n - v_{j-1/2}^n \right]$$

$$+ \frac{\Delta t}{\Delta x} \left[v_{j-1/2}^n - 2v_j^n + v_{j+1/2}^n \right]$$

if $Q \leq 1$
Lax-Friedrichs method \rightarrow stable



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Backward diff	$\theta = v$	stable.
Forward diff	$\theta = -v$	unstable.
Central diff	$\theta = 0$	unstable

Lax-wandroff : $\theta = v^2$

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Lecture 13

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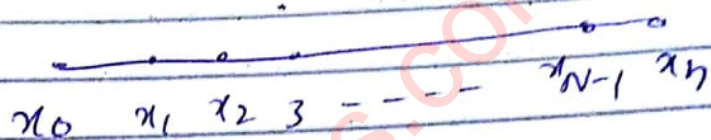
(ELLIPTIC PDES) 14-11-2016

Finite Difference method
for 1D - Poisson equation:-

BVP

$$-\frac{\partial^2 u}{\partial x^2} = f(x) \text{ in } \Omega = (0,1)$$

$$u(0) = 0 = u(1) \rightarrow \text{Dirichlet BCS}$$



$$u_i \approx u(x_i), \quad f_i \approx f(x_i), \quad x_i = ih, \quad h = \frac{1-0}{N} = \frac{1}{N}$$

central difference:-

$$-\frac{u_{i-1} - 2u_i + u_{i+1}}{h^2} = f_i$$

$$\forall i=1, 2, \dots, N-1, \quad \text{BCS.}$$

Result:- The original PDE is replaced by a linear system for model values

1-D Poisson Eq

$$i=1 \quad - \quad \left[\frac{u_0 - 2u_1 + u_2}{h^2} = f_1 \right]$$

$$i=2 \quad - \quad \left[\frac{u_1 - 2u_2 + u_3}{h^2} = f_2 \right]$$

$$i=N-1 \quad - \quad \left[\frac{u_{N-2} - 2u_{N-1} + u_N}{h^2} = f_{N-1} \right]$$

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matrix form:

$$\underline{AU} = \underline{F}$$

$$A = \frac{1}{h^2} \begin{bmatrix} 2 & -1 & 0 & 0 & \dots & 0 \\ -1 & 2 & -1 & 0 & \dots & 0 \\ & & \ddots & & & \\ & & & 0 & & -1 & 2 & -1 \\ & & & & & & & -1 & 2 \end{bmatrix}$$

$$U = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_{N-1} \end{bmatrix}, \quad F = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_{N-1} \end{bmatrix}$$

Matrix A is SPD because for any $\underline{v} = \{v_1, v_2, \dots, v_{N-1}\}^T$

$$\underline{v}^T A \underline{v} = \frac{1}{h^2} \left[v_1^2 + \sum_{i=2}^{N-1} (v_i - v_{i-1})^2 + v_{N-1}^2 \right]$$

> 0 for $\underline{v} \neq \underline{0}$

u exist and is unique.

A is SPD $\Rightarrow A$ is invertible

$$\Rightarrow \underline{u} = \underline{A}^{-1} \underline{F}$$

if $N \leq 100$ then direct method is applied. otherwise iterative solvers should be used, such as Jacobi method, Gauss sidal method and SOR - method.

(55)

Iterative Solution Method:-

The iterative method proceeds by starting with a guess $u_i^{(0)}$ which in general will not satisfy eq. (1). The guess is then updated to enforce the equality at point i .

$$u_i^{(n+1)} = -u_{i-1}^{(n)} - u_{i+1}^{(n)} + h^2 f_i$$

where the superscript n is the iteration number. This update is repeated until a convergence criterion is reached, for example the correction drops below a specified tolerance:

$$\max_{1 \leq i \leq N-1} |G_i^{(n)}| = \max_{1 \leq i \leq N-1} |u_i^{(n+1)} - u_i^{(n)}| \leq \epsilon$$

Theoretical analysis guarantees that the process will eventually reach a solution, and that the number of iterations needed to reach convergence scales as:

$$K \gg \frac{\ln \epsilon}{\ln(1 - \sin^2 \frac{\pi}{2(N-1)})} \sim \frac{-2(N-1)^2}{\pi^2} \ln \epsilon$$

The above can be used as a rough estimate to bound the iteration count.

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Lecture 14

15-11-2016

FDM of elliptic PDEs

Example:-

$$u_{xx} = f(x)$$

$$u(0) = 0 = u(1)$$

$$h = 1/5 = 0.2$$

$$-u_{i-1} + 2u_i - u_{i+1} = f_i$$

$$i=1 \quad \frac{h^2}{h^2} u_0 + 2u_1 - u_2 = h^2 f_1$$

$$i=2 \quad -u_1 + 2u_2 - u_3 = h^2 f_2$$

$$i=3 \quad -u_2 + 2u_3 - u_4 = h^2 f_3$$

$$i=4 \quad -u_3 + 2u_4 - u_5 = h^2 f_4$$

$$\begin{bmatrix} 2 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \end{bmatrix} = \frac{h^2}{h} \begin{bmatrix} f_1 \\ f_2 \\ f_3 \\ f_4 \end{bmatrix}$$

other types of Boundary

Conditions:- $-u_{xx} = f(x), x \in (0,1)$

$$u(0) = 0, \quad \frac{\partial u}{\partial x}(1) = 0$$

central difference $u_{N+1} - u_{N-1}$

$$\boxed{u_{N+1} = u_{N-1}} \quad \frac{2h}{2h} = 0$$

$$-u_{i-1} + 2u_i - u_{i+1} = f_i \quad i=1, \dots, N-1$$

$$i=1 \quad \frac{u_0 + 2u_1 - u_2}{h^2} = f_1$$

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$$i=N \quad \frac{-u_{N-1} + 2u_N - u_{N+1}}{h^2} = f_N$$

$$\frac{-2u_{N-1} + 2u_N}{h^2} = f_N$$

$$\Rightarrow \frac{-u_{N-1} + u_N}{h^2} = \frac{1}{2} f_N$$

$$\underline{A} \underline{U} = \underline{F}$$

$$\frac{1}{h^2} \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & 0 & 0 \\ & & & & \\ & & & 2 & 0 \\ & & & -1 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ \vdots \\ \vdots \\ \vdots \\ u_N \end{bmatrix} = \begin{bmatrix} \frac{1}{2} f_1 \\ \vdots \\ \vdots \\ \vdots \\ \frac{1}{2} f_N \end{bmatrix}$$

$$\frac{\partial u}{\partial x}(0) = 0 \Rightarrow \frac{u_1 - u_0}{2h} = 0 \Rightarrow u_1 = u_0$$

$$-u_{xx} = f(x), \quad x \in (0, 1)$$

$$u(0) = g_0, \quad \frac{\partial u}{\partial x}(1) = g_1$$

$$i=1 \quad \frac{-u_0 + 2u_1 - u_2}{h^2} = f_1$$

$$\frac{2u_1 - u_2}{h^2} = f_1 + \frac{g_0}{h^2}$$

$$\frac{u_{N+1} - u_{N-1}}{2h} = g_1$$

$$u_{N+1} = u_{N-1} + 2hg_1$$

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$$i=N \left| \frac{-u_{N-1} + u_N}{h^2} = \frac{1}{2} f_N + \frac{g_1}{h} \right|$$

Assignment:-

Solve $-u_{xx} = f(x)$

$$f(x) = -6x + 12x^2$$

(i) $u(0) = 0 = u(1)$

(ii) $u(0) = 0, \frac{\partial u}{\partial x}(1) = 0$

(iii) $u(0) = 0.1, \frac{\partial u}{\partial x} = 0.2$

$$h = 1/5 = 0.2$$

Do it manually and through program.

2D-Poisson Equation:-

BVP $-\frac{\partial^2 u}{\partial x^2} - \frac{\partial^2 u}{\partial y^2} = f$ in

$$\Omega = [0, 1] \times [0, 1]$$

$$h_x = h_y = h$$

$$u_{ij} \approx u(x_i, y_j), f_{ij} = f(x_i, y_j)$$

$$(x_i, y_j) = (ih, jh), i, j = 0, 1, \dots, N$$

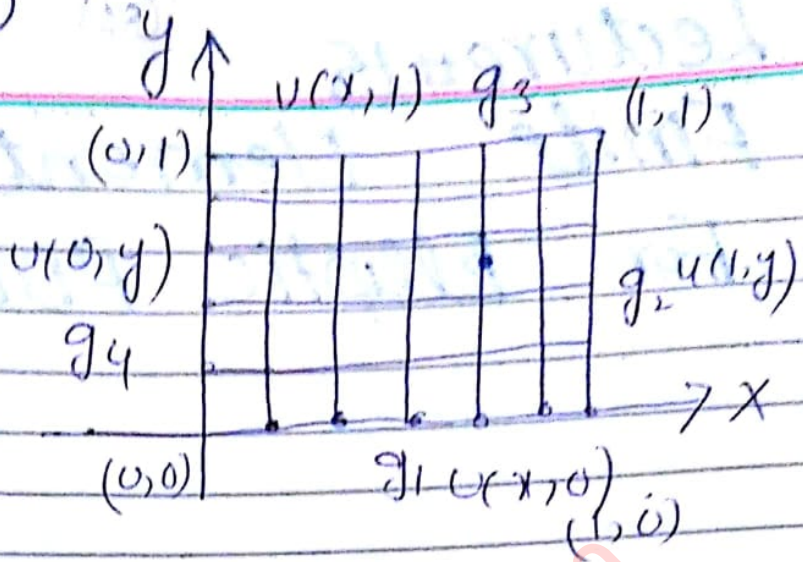
$$u(x, 0) = g_1 = 0$$

$$u(1, y) = g_2 = 0$$

$$u(x, 1) = g_3 = 0$$

$$u(0, y) = g_4 = 0$$

(59)



$$\frac{-u_{i+1,j} + 2u_{i,j} - u_{i-1,j} - u_{i,j+1} + 2u_{i,j} - u_{i,j-1}}{h^2} = f_{i,j}$$

$$\frac{u_{i+1,j} - u_{i,j-1} + 4u_{i,j} - u_{i,j+1} - u_{i+1,j}}{h^2} = f_{i,j}$$

$$\forall i,j = 1, 2, \dots, N-1$$

$$\underline{A} \underline{u} = \underline{F} \quad \underline{A} \in \mathbb{R}^{(N+1)^2 \times (N-1)^2}$$

$$\underline{u}, \underline{F} \in \mathbb{R}^{(N-1)^2}$$

$$\underline{u} = \begin{bmatrix} u_{1,1}, u_{2,1}, u_{3,1}, \dots, u_{N+1,1} \\ u_{2,1}, u_{2,2}, \dots, u_{N,1,j}, \dots, u_{N-1,j-1} \end{bmatrix}^T$$

row by row

$$\underline{A} = \begin{bmatrix} \underline{B} & -\underline{I} & \underline{O} \\ -\underline{I} & \underline{B} & -\underline{I} \\ \underline{O} & -\underline{I} & \underline{B} \end{bmatrix}$$

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21-11-2016

Lecture #15

Finite Element Method

Method

For Elliptic PDE's:-

The FEM was created to solve complicated eq. of elasticity and structured mechanic usually modeled by elliptic type PDE's with complicated geometric. It has been developed for other application as well

$$\text{our model: 1D: } -u_{xx} = f(x), 0 < x < L$$

$$2D: -(u_{xx} + u_{yy}) = f(x, y)$$

$x, y \in \Omega \rightarrow$ domain in (x, y) plane

$$u(x, y) = u_0(x, y) \quad \forall x, y \in \partial\Omega$$

Ω interior $\partial\Omega$ Boundary of Ω

$$\frac{\partial u}{\partial t} + \frac{\partial f(x)}{\partial x} = 0 \rightarrow (\text{strong formulation})$$

$$\int_{\Omega} u dx - f(x) dV = 0 \quad \text{weak formulation}$$

or $\int_{\Omega} u dx - f(x) dV = 0$ its solution is weak solution.

The FEM approximation the

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unknown solution function u by a linear combination of basis functions each with finite (local) support, hence finite elements. The support of a function is a region where function is non-zero.

$$u(x) \approx U_h(x) = \sum_{i=1}^{m+1} u_i \phi_i(x)$$

\hookrightarrow unknown

The basis function $\phi_i(x)$ (finite element) are chosen (given) in advance. The residual is defined to be $r(x) = u''(x) + f(x)$.

The coefficient u_i are obtained by requirements on residual.

1) Galerkin: The residual is orthogonal to the basis functions.

2) Collocation: The residual is zero on n discrete points.

3) Rayleigh - Ritz: The PDE is transformed into equivalent minimization problem (Euler-Lagrange Equation).

Galerkin FEM:-

1) Poisson $-u_{xx} = f(x)$, $0 < x < 1$
 $u(0) = 0 = u(1)$.

1) Weak formulation:-

This can be done by multiplying a test function $\phi(x)$ on both sides of PDE

$$v(x) \leq \phi(x) \Rightarrow \int_0^1 v(x) \phi(x) dx = 0$$

$$-\int_0^1 u_{xx} \phi(x) dx = \int_0^1 f(x) \phi(x) dx$$

Integration by parts $\phi(0) = 0 = \phi(1)$.

$$-u(x) \phi(x) \Big|_0^1 + \int_0^1 u_x \phi_x(x) dx = \int_0^1 f(x) \phi(x) dx$$

$$\Rightarrow \int_0^1 u'(x) \phi'(x) dx = \int_0^1 f(x) \phi(x) dx$$

↓
weak formulation

2) Discrete your Domain: $(0 \leq x \leq 1)$

$$x_i = ih \quad i = 0, 1, \dots, n$$

constant $\leftarrow h = \frac{1-0}{N}$

intervals $[x_{i-1}, x_i]$ $i = 1, 2, \dots, N$

↓
line element. Equidistant (uniform mesh)

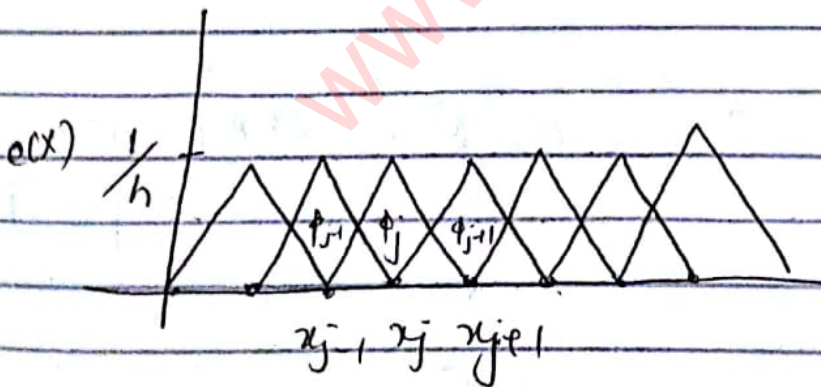
(3)

(3) Let $\phi_j(x) = e_j \rightarrow$ (Basis function with compact support)

Hat Function:

$$e_j(x) = \begin{cases} \frac{x - x_{j-1}}{h} & \checkmark \quad x_{j-1} \leq x \leq x_j \\ \frac{x_{j+1} - x}{h} & \text{if } x_j \leq x \leq x_{j+1} \\ 0 & \text{otherwise} \end{cases}$$

$$e_j'(x) = \begin{cases} 1/h & \text{if } x_{j-1} \leq x \leq x_j \\ -1/h & x_j \leq x \leq x_{j+1} \\ 0 & \text{otherwise} \end{cases}$$



4) Approximate the solution as a linear combination of basis function

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$$U_h(x) = \sum_{i=1}^{n-1} u_i e_i(x), \text{ where } u_i \text{ are unknown}$$

replace $\phi(x) = e_j(x)$,

$$\int_0^1 u_h'(x) \cdot e_j'(x) dx = \int_0^1 f(x) e_j(x) dx$$

$j = 1, 2, \dots, n-1$

$$\int_0^1 \sum_{i=1}^{n-1} u_i' e_i'(x) e_j'(x) dx = \int_0^1 f(x) e_j(x) dx$$

$$\sum_{i=1}^{n-1} u_i \int_0^1 e_i'(x) e_j'(x) dx = \int_0^1 f(x) e_j(x) dx$$

$j = 1, \dots, n-1$

$$\left(\int_0^1 e_1' e_1' dx \right) u_1 + \left(\int_0^1 e_1' e_2' dx \right) u_2 + \dots$$

$$+ \left(\int_0^1 e_1' e_{n-1}' dx \right) u_{n-1} = \int_0^1 f e_1 dx$$

$$\left(\int_0^1 e_2' e_1' dx \right) u_1 + \dots + \left(\int_0^1 e_2' e_{n-1}' dx \right) u_{n-1}$$

$$= \int_0^1 f e_2 dx$$

$$\left(\int_0^1 e_{n-1}' e_1' dx \right) u_1 + \dots + \left(\int_0^1 e_{n-1}' e_{n-1}' dx \right) u_{n-1}$$

$$= \int_0^1 f e_{n-1} dx$$

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$$\begin{bmatrix} a(e_1, e_1) & a(e_1, e_2) & \dots & a(e_1, e_{n-1}) \\ \vdots & \vdots & & \vdots \\ a(e_{n-1}, e_1) & a(e_{n-1}, e_2) & \dots & a(e_{n-1}, e_{n-1}) \end{bmatrix} \begin{bmatrix} u_1 \\ \vdots \\ u_{n-1} \end{bmatrix} = \begin{bmatrix} f, e_1 \\ \vdots \\ f, e_{n-1} \end{bmatrix}$$

$$a(e_i, e_j) = \int_0^1 e_i' e_j' dx, (f, e_j) = \int_0^1 f e_j dx$$

Galerkin FEM

$$\int_0^1 e_i' e_j' dx?$$

$$\begin{aligned} i=j \quad \int_0^1 e_i' e_j' dx &= \int_0^1 (e_j')^2 dx \\ &= \int_{x_{j-1}}^{x_{j+1}} (e_j')^2 dx = \int_{x_{j-1}}^{x_j} (e_j')^2 dx + \int_{x_j}^{x_{j+1}} (e_j')^2 dx \\ &= \frac{h}{h^2} + \frac{h}{h^2} = \frac{2}{h} \end{aligned}$$

$$\begin{aligned} i=j-1 \quad \int_0^1 e_{j-1}' e_j' dx &= \int_{x_{j-1}}^{x_j} e_{j-1}' e_j' dx + \int_{x_j}^{x_{j+1}} e_{j-1}' e_j' dx \\ &= \left(-\frac{1}{h^2}\right) \cdot h = -\frac{1}{h} \end{aligned}$$

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$$z = j+1$$

$$\int_0^1 e^{j+1} e_j' dx = \int_{x_j-1}^{x_j} e^{j+1} e_j' dx + \int_y^{x_j} e^{j+1} e_j' dx$$

$$= 0 + \frac{1}{h} (h) = \frac{1}{h}$$

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22-11-16.

$$\int_0^1 e_i' e_j' dx?$$

$$\text{for } i=j \quad \int_0^1 e_j' e_j' dx = \int_{x_{j-1}}^{x_j} |e_j'|^2 dx$$

$$= \int_{x_{j-1}}^{x_j} |e_j'|^2 dx + \int_{x_j}^{x_{j+1}} |e_j'|^2 dx$$

$$= \frac{1}{h} + \frac{1}{h} = \frac{2}{h}$$

for $i=j-1$

$$\int_0^1 e_{j-1}' e_j' dx = \int_{x_{j-1}}^{x_j} e_{j-1}' e_j' dx + \int_{x_j}^{x_{j+1}} e_{j-1}' e_j' dx$$

$$= -\frac{1}{h^2} \cdot h = -\frac{1}{h}$$

for $i=j+1$

$$\int_0^1 e_j' e_{j+1}' dx = \int_{x_{j-1}}^{x_j} e_j' e_{j+1}' dx + \int_{x_j}^{x_{j+1}} e_j' e_{j+1}' dx$$

$$= 0 + \left(-\frac{1}{h^2} \cdot h\right) = -\frac{1}{h}$$

$$A = \frac{1}{h} \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & 2 & -1 & \dots & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}$$

(Tri-diagonal)

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$$A = \frac{1}{h} \begin{bmatrix} 2 & -1 & 0 & \dots \\ & & & \end{bmatrix} \quad \text{Ans}$$

Both FEM and FDM agree with each other because

- 1) Mesh is equidistant (uniform).
- 2) Linear basis function (piecewise linear).

Advantages of FEM:-

- 1) Mesh is non uniform.
- 2) irregular Geometry (Multi-dimensional problems).

Summary of Galerkin method:-

- 1) Basis functions ϕ_j can be chosen to fit the requirement of the problem. Common choices are low order (multidimensional) polynomials.
- 2) The necessary integrals can be easily calculated for a standard element and the whole process is easily automated.
- 3) Much easier to generalize for arbitrary geometry and non-uniform irregular meshes than FDM.

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4. Finite elements are very flexible and easily adaptable
 5. This made the FEM as, most popular method

Example 1 -

$$-u_{xx} = x, u(0) = 0 = u(1)$$

$$(i) \quad u(x) = u_1 x(1-x) \quad \phi(x) = x(1-x)$$

$$(ii) \quad u(x) = u_1 x(1-x) + u_2 x^2(1-x)$$

Use Galerkin method to find u_1, u_2

$$(i) \quad \int_0^1 (u_{xx} + x) \cdot \phi(x) dx = 0$$

$$\int_0^1 [u_1(-x) + x] x(1-x) dx = 0$$

$$\Rightarrow \int_0^1 [-2u_1 + x] x(1-x) dx = 0$$

$$\Rightarrow \int_0^1 [-2u_1 x + 2u_1 x^2 + x^2 - x^3] dx = 0$$

$$-2u_1 \frac{x^2}{2} \Big|_0^1 + 2u_1 \frac{x^3}{3} \Big|_0^1 + \frac{x^3}{3} \Big|_0^1 - \frac{x^4}{4} \Big|_0^1 = 0$$

$$-u_1 + \frac{2}{3}u_1 + \frac{1}{3} - \frac{1}{4} = 0$$

$$\frac{-3u_1 + 2u_1}{3} = \frac{1}{4} - \frac{1}{3}$$

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$$-\frac{u_1}{3} =$$

$$\frac{3-4}{12}$$

$$-\frac{u_1}{3} = \frac{+1}{12}$$

$$u_1 = \frac{3}{12}$$

$$u_1 = \frac{1}{4}$$

$$(ii) \int_0^1 x(x) \phi_1 dx = 0 \quad \int_0^1 x(x) \phi_2 dx = 0$$

$$(a) \int_0^1 u' \phi_1'(x) dx = \int_0^1 f(x) \phi_1'(x) dx$$

$$(b) \int_0^1 u' \phi_2'(x) dx = \int_0^1 f(x) \phi_2'(x) dx$$

Example:-

$$-u_{xx} = f(x)$$

$$u(0) = 0, u(1) = 1$$

$$\int_0^1 u' \phi_1'(x) dx = \int_0^1 f(x) \phi_1'(x) dx \quad ?$$

$$w(x) = u(x) - x \quad \checkmark$$

$$w(0) = 0 - 0 = 0$$

$$w(1) = 1 - 1 = 0$$

$$-w_{xx} = f(x)$$

at the end $u(x) = w(x) + x$

(72)

$$a(u, u) = \int_0^1 |u'(x)|^2 dx > 0 \quad \forall u \neq 0$$

$\Rightarrow A$ is a positive definite matrix

(symmetric $a(u, v) = a(v, u)$ and positive eigen values)

(The system of equations always give a unique solution.)

Proof:-

Symmetry

$$(A_h)_{jk} = \int_0^1 \frac{\partial \phi_j}{\partial x}(x) \frac{\partial \phi_k}{\partial x} dx = \int_0^1 \frac{\partial \phi_k}{\partial x} \frac{\partial \phi_j}{\partial x} dx = (A_h)_{kj}$$

\Rightarrow symmetric

positive definite matrix.

For any vector $v \in \mathbb{R}^{n-1}$ (not 0)

$$v^T A v = \sum_{j, k=1}^{n-1} v_j v_k \int_0^1 \frac{\partial \phi_j}{\partial x} \frac{\partial \phi_k}{\partial x} dx$$

$$= \int_0^1 \left(\sum_{j=1}^{n-1} v_j \frac{\partial \phi_j}{\partial x} \right) \left(\sum_{k=1}^{n-1} v_k \frac{\partial \phi_k}{\partial x} \right) dx$$

$$= \int_0^1 \left(\sum_{j=1}^{n-1} v_j \frac{\partial \phi_j}{\partial x} \right)^2 dx > 0$$

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stability $\int_0^1 \left(\frac{\partial u_h}{\partial x} \right)^2 dx = \int_0^1 \left(\sum_{j=1}^{m-1} u_j^h \frac{\partial \phi_j}{\partial x} \right)^2 dx$

$$u_h(x) = \sum_{j=1}^{m-1} u_j \phi_j(x)$$

$$= U^T (AU) = U^T F_h.$$

$$= \sum_{j=1}^{m-1} \int_0^1 u_j^h \phi_j(x) f(x) dx.$$

$$= \int_0^1 u^h f(x) dx \rightarrow \textcircled{1}$$

$$\left[\int_a^b f(x) g(x) dx \right] \leq \left[\int_a^b |f(x)|^2 dx \right]^{1/2} \left[\int_a^b |g(x)|^2 dx \right]^{1/2}$$

$$\times \left[\int_a^b |g(x)|^2 dx \right]^{1/2}$$

(1) \Rightarrow

$$\int_0^1 \left(\frac{\partial u}{\partial x} \right)^2 dx = \int_0^1 u(x) f(x) dx.$$

$$\leq \left[\int_0^1 |u(x)|^2 dx \right]^{1/2} \left[\int_0^1 |f(x)|^2 dx \right]^{1/2}$$

 $\textcircled{2}$

Also

(Cauchy-Schwarz)

$$\int_0^1 |u(x)|^2 \leq \frac{1}{4} \int_0^1 \left(\frac{\partial u}{\partial x} \right)^2 dx \text{ for}$$

function with BCS $u(0) = 0 = u(1)$
Poincaré inequality.

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$$\int_0^1 |u(x)| dx \leq \frac{1}{4} \int_0^1 \left(\frac{\partial u}{\partial x}\right)^2 dx \leq \frac{1}{4} \sqrt{\int_0^1 |u(x)|^2 dx} \sqrt{\int_0^1 |f(x)|^2 dx}$$

$$\sqrt{\int_0^1 |u(x)|^2 dx} \sqrt{\int_0^1 |u(x)|^2 dx} \leq \frac{1}{4} \sqrt{\int_0^1 |u(x)|^2 dx} \sqrt{\int_0^1 |f(x)|^2 dx}$$

$$\Rightarrow \int_0^1 |u(x)|^2 dx \leq \frac{1}{4} \int_0^1 |f(x)|^2 dx$$

$$\|u(x)\|_2 \leq \frac{1}{4} \|f(x)\|_2$$

we get an estimate which is independent from h and L_2 norm.

28-11-2016

Exercise:-

use Galerkin method to solve steady state $au_x - bu_{xx} = 0$

$$u(0) = 0, \quad u(1) = 1$$

$$a = 5 \quad \text{--- (1)}$$

$$b = 1, 0.5, 0.1$$

$$\{ u_x - u_{xx} = 0 \} \quad \text{--- (2)}$$

(15)

$w = U - x$

$w(0) = 0$

$w(1) = 0$

$u = w + x$

(2) $\Rightarrow \sum (w_{x+1}) - w_{xx} = 0$

$\sum w_x + \epsilon - w_{xx} = 0 \quad \text{--- (3)}$

$w(0) = 0 = w(1)$

Galerkin method: - multiply with test function and integrate over $[0, 1]$

$\int_0^1 \epsilon w_x \phi(x) + \int_0^1 \epsilon \phi(x) - \int_0^1 w_{xx} \phi(x) = 0$

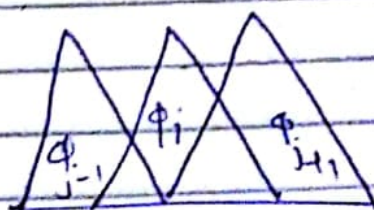
$\phi(x) = e_j$
 $w_h(x) = \sum_{i=1}^{n-1} w_i e_i(x)$ integrate by parts

$\sum_{i=1}^{n-1} \epsilon w_i \int_0^1 e_i e_j dx + \epsilon \int_0^1 e_j dx - \sum_{i=1}^{n-1} w_i \int_0^1 e_i e_j dx = 0$

B Eh A

$A = \frac{1}{h} \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & 2 & -1 \\ \dots & \dots & \dots & -1 & 2 \end{bmatrix}$

$\int_0^1 e_i e_j dx = ?$



for $i=j$

$\int_0^1 e_j e_j dx = \int_{x_{j-1}}^{x_j} e_j e_j dx + \int_{x_j}^{x_{j+1}} e_j e_j dx$

(76)

$$= \int_{x_{j-1}}^{x_j} \frac{1}{h} \left(\frac{x - x_{j-1}}{h} \right) dx + \int_{x_j}^{x_{j+1}} \left(-\frac{1}{h} \right) \frac{x_{j+1} - x}{h} dx$$

$$= \frac{1}{h^2} \left. \frac{(x - x_{j-1})^2}{2} \right|_{x_{j-1}}^{x_j} + \frac{1}{h^2} \left. \frac{(x_{j+1} - x)^2}{2} \right|_{x_j}^{x_{j+1}}$$

$$= \frac{1}{2} - \frac{1}{2} = 0$$

for $i = j$

$$= \int_{x_{j+1}}^{x_j} -\frac{1}{h} \left(\frac{x - x_{j-1}}{h} \right) dx + \int_{x_j}^{x_{j+1}} \frac{x_{j+1} - x}{h} dx$$

$$= -\frac{1}{h^2} \left. \frac{(x - x_{j-1})^2}{2} \right|_{x_{j+1}}^{x_j} - \frac{1}{2}$$

for $i = j+1$

$$= \int_{x_{j-1}}^{x_j} -\frac{1}{h} \left(\frac{x - x_{j-1}}{h} \right) dx + \int_{x_j}^{x_{j+1}} \frac{1}{h} \left(\frac{x_{j+1} - x}{h} \right) dx$$

$$= 0 + \left(\frac{1}{h^2} \right) - \left. \frac{(x_{j+1} - x)^2}{2} \right|_{x_j}^{x_{j+1}}$$

$$B = \begin{bmatrix} 0 & \epsilon/2 & \dots & 0 \\ -\epsilon/2 & 0 & \epsilon/2 & \dots & 0 \\ 0 & 0 & 0 & -\epsilon/2 & 0 \end{bmatrix} \quad (77)$$

$$A+B = \begin{bmatrix} 2/h & -1/h + \epsilon/2 & 0 & \dots & 0 \\ -1/h - \epsilon/2 & 2/h & -1/h + \epsilon/2 & \dots & \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & -1/h - \epsilon/2 & 2/h \end{bmatrix} = \tilde{A}$$

$$\int_0^1 e_j dx = \int_{x_j-1}^{x_j} \left(\frac{x - x_j - 1}{h} \right) dx + \int_{x_j}^{x_{j+1}} \left(\frac{x_{j+1} - x}{h} \right) dx$$

$$= \frac{1}{h} \left(\frac{x - x_j - 1}{2} \right) \Big|_{x_j-1}^{x_j} + \frac{1}{h} \left(\frac{x_{j+1} - x}{2} \right) \Big|_{x_j}^{x_{j+1}}$$

$$= \frac{h}{2} + \frac{h}{2} = h$$

$$AW = F$$

$$\begin{bmatrix} 2/h & -1/h + \epsilon/2 & 0 & \dots & 0 \\ -1/h - \epsilon/2 & 2/h & -1/h + \epsilon/2 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & -1/h - \epsilon/2 & 2/h \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_{n-1} \end{bmatrix} = \begin{bmatrix} -\epsilon h \\ \epsilon h \\ \vdots \\ \epsilon h \end{bmatrix}$$

Finally -
 $u = w + x$ final solution

(78)

Exact solution

$$u(x) = \frac{e^x}{1 - e^x}$$

The maximum principle holds if Ch (Peclet no.) does not exceed unity. When $Ch \leq 1$, if Ch exceeds unity, the solution has boundary layer which can not be resolved by standard Galerkin method. In that case some stabilization technique is needed. In this problem for $b=1$ & $b=0.5$ the maximum principle holds.

Finite Element method for Heat equation:-

$$u_t - a u_{xx} = 0, \quad t > 0,$$

$$x \in [b_1, b_2],$$

$$u(b_1) = 0 = u(b_2).$$

a70

$$u(x, 0) = u_0(x)$$

$$\int_{b_1}^{b_2} u_t \phi(x) dx - a \int_{b_1}^{b_2} u_{xx} \phi(x) dx = 0$$

$$\Rightarrow \int_{b_1}^{b_2} u_t \phi(x) dx + a \int_{b_1}^{b_2} u_x \phi_x(x) dx = 0$$

Discretization

$$b_1 = x_0 < x_1 < \dots < x_n = b_2$$

$$q_j = e_j$$

(79)

$$u_h(x, t) = \sum_{i=1}^{n-1} u_i e_i(x)$$

$$\int_{b_1}^{b_2} \sum_{i=1}^{n-1} (u_i)_t e_i e_j dx + a \int_{b_1}^{b_2} \sum_{i=1}^{n-1} u_i e_i e_j' dx = 0$$

$$\Rightarrow \sum_{i=1}^{n-1} (u_i)_t \int_{b_1}^{b_2} e_i e_j dx + a \sum_{i=1}^{n-1} u_i \int_{b_1}^{b_2} e_i e_j' dx = 0$$

Mass matrix; M

A
stiffness matrix.

$$M = \begin{bmatrix} \int_{b_1}^{b_2} e_1 e_1 dx & \int_{b_1}^{b_2} e_1 e_2 dx & \dots \\ \vdots & \vdots & \vdots \\ \int_{b_1}^{b_2} e_{n-1} e_1 dx & \int_{b_1}^{b_2} e_{n-1} e_2 dx & \dots \end{bmatrix}$$

$$A = \begin{bmatrix} \int_{b_1}^{b_2} e_1 e_1' dx & \int_{b_1}^{b_2} e_1 e_2' dx & \dots \\ \vdots & \vdots & \vdots \\ \int_{b_1}^{b_2} e_{n-1} e_1' dx & \int_{b_1}^{b_2} e_{n-1} e_2' dx & \dots \end{bmatrix}$$

$$\underline{M} \underline{U}_t + a \underline{A} \underline{U} = \underline{0}$$

$$i=j \int_{b_1}^{b_2} e_i e_i dx = \int_{x_{j-1}}^{x_{j+1}} e_j^2 dx = \int_{x_{j-1}}^{x_j} e_j^2 dx + \int_{x_j}^{x_{j+1}} e_j^2 dx$$

$$= \int_{x_{j-1}}^{x_j} \frac{(x - x_{j-1})^2}{h^2} dx + \int_{x_j}^{x_{j+1}} \frac{(x_{j+1} - x)^2}{h^2} dx$$

(80)

$$= \frac{h}{3} + \frac{h}{3} = \frac{2h}{3}$$

for $i = j+1$

$$\int_{b_1}^{b_2} e_{j+1} e_j dx = \int_{x_{j-1}}^{x_{j+1}} e_{j+1} e_j dx = \int_{x_{j-1}}^{x_j} e_{j+1} e_j dx + \int_{x_j}^{x_{j+1}} e_{j+1} e_j dx$$

$$= \int_{x_j}^{x_{j+1}} \left(\frac{x - x_j}{h} \right) \left(\frac{x_{j+1} - x}{h} \right) dx$$

$$= \frac{h}{6}$$

$$\int_{b_1}^{b_2} e_{j-1} e_j dx = \frac{h}{6}$$

$$\underline{M} = h \begin{bmatrix} \frac{2}{3} & \frac{1}{6} & 0 & \dots & 0 \\ \frac{1}{6} & \frac{2}{3} & \frac{1}{6} & \dots & \\ \dots & 0 & \dots & \frac{1}{6} & \frac{2}{3} \end{bmatrix}$$

$$\underline{M} \underline{U} + a \underline{A} \underline{U} = \underline{0}$$

System of ODEs.

approximate derivative by
implicit Euler method

$$\underline{M} \underline{U}^{n+1} - \underline{U}^n + a \underline{A} \underline{U}^{n+1} = \underline{0}$$

(21)

$$\left(\frac{1}{k} \underline{M} + a \underline{A} \right) \underline{U}^{n+1} = \frac{1}{k} \underline{M} \underline{U}^n$$

$$u_t - a u_{xx} = 0$$

$$\underline{M} \underline{U} + a \underline{A} \underline{U} = 0$$

$$\underline{M} = h \begin{bmatrix} 2/3 & 1/6 & 0 & \dots & 0 \\ 1/6 & \dots & 2/3 & 1/6 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & \dots & \dots & 1/6 & 2/3 & \dots \end{bmatrix}$$

$$\underline{A} = \frac{1}{h} \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots \\ 0 & \dots & \dots & -1 & 2 \end{bmatrix}$$

Discretization in terms of space variable only. while time derivative is kept continuous.

Any ODE solve can be applied to solve the resulting ODE system

(B2)

Here as an example, we apply implicit Euler method

$$\text{mass matrix} \leftarrow \frac{M}{h} \underline{U}^{n+1} - \underline{U}^n + \alpha \overset{\text{stiffness matrix}}{A} \underline{U}^{n+1} = 0$$

$$\left(\frac{1}{h} \underline{M} + \alpha \underline{A} \right) \underline{U}^{n+1} = \frac{1}{h} \underline{M} \underline{U}^n$$

$$\underline{U}^{n+1} = \underline{B}^{-1} \frac{1}{h} \underline{M} \underline{U}^n$$

sometimes one would like to solve instead of $\underline{M} \underline{U}^n + \alpha \underline{A} \underline{U} = 0$, the system of equations $\underline{U}^n = \underline{B} \underline{U}$. This could be obtained by lumping of mass matrix.

Trapezoidal rule:

$$Q_n(g) = \int_a^b I_g(g(x)) dx$$

we use trapezoidal rule on the left integration

(83)

$$\int_a^b f(x) dx = \frac{b-a}{2} [f(a) + f(b)]$$

$$= \frac{h}{2} [0+1]$$

$$b_2 = h/2$$

$$\Rightarrow U^t Q_h(e_i e_j) + a \sum_{i=1}^{n-1} \pi_i \int_{b_1}^{b_2} e_i' e_j' dx = 0$$

mass matrix will be approximated as

$$\tilde{M} = (Q_h(e_i e_j))$$

Calculation of $Q_h(e_i e_j)$

if $i=j$

$$Q_h(e_j^2) = \int_{b_1}^{b_2} e_j^2 dx$$

$$\frac{x_i - x_{j-1}}{h} \left. \begin{array}{l} x_j \leq x < x_{j+1} \\ x_{j+1} - x \end{array} \right\}$$

$$\frac{x_{j+1} - x}{h} \left. \begin{array}{l} x_j \leq x < x_{j+1} \\ x_{j+1} - x \end{array} \right\}$$

0 otherwise

$$= \int_{x_{j-1}}^{x_j} e_j^2 dx + \int_{x_j}^{x_{j+1}} e_j^2 dx$$

$$= h/2 [0+1] + h/2 [1+0]$$

$$= h$$

if $i=j-1$

$$\int_{b_1}^{b_2} e_{j-1} e_j dx = \int_{x_{j-1}}^{x_{j+1}} e_{j-1} e_j dx$$

$$= 0 \text{ due to}$$

trapezoidal rule.

if $i=j+1 \Rightarrow Q_h(e_j e_{j+1}) = 0$

$$\tilde{M} = h \begin{bmatrix} 1 & & & \\ & 1 & & 0 \\ & & 1 & \\ & & & 1 \end{bmatrix} = hI$$

(B4)

$$\tilde{M}U_t + aAU = 0$$

$$hIU_t + aAU = 0$$

$$U_t = -\frac{a}{h}AU$$

$$U_t = f(u)$$

$$U_t = \underline{\underline{\beta U}}$$

$$U_t - aU_{xx} = 0$$

$$U_t - a \left[\frac{U_{j-1} - 2U_j + U_{j+1}}{h^2} \right] = 0, \quad \forall j=1, \dots, n-1$$

$$U_t \mp \frac{a}{h}AU = 0 \quad \text{FDM}$$

B/c of Equidistant + Lumping

\Rightarrow FDM and FEM are equidistant

Reaction - Diffusion Equation

$$U_t = aU_{xx} + f(x, U, t), \quad t \geq 0,$$

$$x \in [b_1, b_2]$$

$$U(x, 0) = U_0(x), \quad U(b_1) = 0 = U(b_2)$$

Galerkin method $\int_{b_1}^{b_2} u \phi dx = -a \int_{b_1}^{b_2} u_x \phi dx = \int_{b_1}^{b_2} f(x, u, t) \phi dx$

$\phi = e_j$

$$\sum_{i=1}^{n-1} (u_i)_t \int_{b_1}^{b_2} e_i e_j dx = -a \sum_{i=1}^{n-1} \int_{b_1}^{b_2} e_i e_j dx + \int_{b_1}^{b_2} f(x, u, t) e_j dx$$

$I_3 \quad \quad \quad I_3 \quad \quad \quad I_3$

$$\int_{b_1}^{b_2} f(x, u, t) e_j dx ?$$

If f is non-linear, then I_3 is complicated.

$$I_3 \approx Q(f e_j) = \int_{b_1}^{b_2} (f e_j) dx$$

$$= f(x_j, u_j, t) \int_{b_1}^{b_2} e_j dx$$

$$= h f(x_j, u_j, t)$$

$$\underline{M} \underline{u}_t + a \underline{A} \underline{u} = \underline{h} \underline{f} \quad \text{Non-linear ODE.}$$

(86)

General Boundary Conditions:-

$$-(p(x)u_x)_x + qu = f, \quad a \leq x \leq b$$

$$u(a) = 0, \quad \alpha u(b) + \beta u'(b) = \gamma,$$

$$\beta \neq 0, \quad \alpha/\beta \neq 0$$

where α, β, γ are known constants, but $u(b)$ and $u'(b)$ are unknown's.

Galerkin method :- Multiply with test function and integrate
Integration by parts.

$$-p(b)u'(b)\phi(b) + p(a)u'(a)\phi(a) + \int_a^b (p'u'\phi' + qu\phi) dx = \int_a^b f\phi dx.$$

$$-p(b)\phi(b) \cdot \frac{\gamma - \alpha u(b)}{\beta} + \int_a^b (p'u'\phi' + qu\phi) dx = \int_a^b f\phi dx$$

$$\Rightarrow \int_a^b (p'u'\phi' + qu\phi) dx + \frac{\alpha}{\beta} p(b)u(b)\phi(b) = \int_a^b f\phi dx + \frac{\gamma}{\beta} p(b)\phi(b)$$

This is the weak form (variational form) of the Sturm-Liouville problem with given BCs

define $a(u, \phi) = \int_a^b (p'u'\phi' + qu\phi) dx + \frac{\alpha}{\beta} p(b)u(b)\phi(b)$

(87)

$$L(\phi) = \int_a^b f\phi dx + \alpha/\beta \cdot p(b)\phi(b)$$

we can prove that

$$(1) a(u, \phi) = a(\phi, u) \Rightarrow \text{symmetric}$$

(2) $a(u, \phi)$ is a bilinear form, i.e.

$$a(\alpha u + \beta w, \phi) = \alpha a(u, \phi) + \beta a(w, \phi)$$

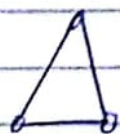
$$a(u, \alpha \phi + \beta w) = \alpha a(u, \phi) + \beta a(u, w)$$

(3) $a(u, \phi)$ is an inner product. The every norm

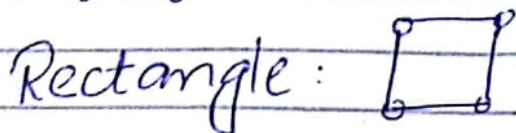
$$\|u\|_a = \sqrt{a(u, u)} = \left\{ \int_a^b (pu^2 + qu^2) dx + \alpha/\beta p(b) u^2(b) \right\}^{1/2}$$

Now we can see why we need to have $\beta \neq 0$ & $\alpha/\beta \neq 0$

Two Dimensional FEM:-



$$P(x, y) = ax + by + c$$



Bilinear function

$$P(x, y) = (ax + b)(cy + d)$$

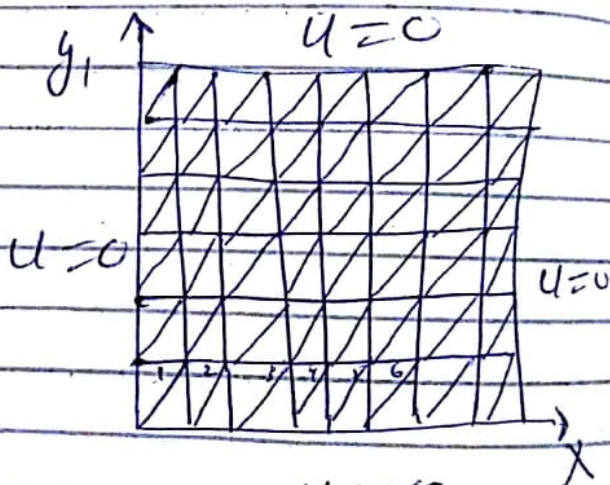
$$= acxy + adx + bcy + bd$$

$$= exy + fx + gy + h$$

Example:-

$$-\Delta u = 1 \quad \Omega = [0, 1] \times [0, 1]$$

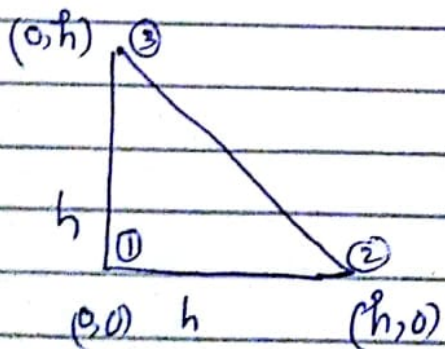
$$\frac{\partial u}{\partial \Omega} = 0$$



Numbering in the interior nodal points

$$x_1, x_2, \dots, x_{n^2}$$

$$h = \frac{1}{n+1}$$



$$(1) \quad \phi_1 = \frac{h-x-y}{h}$$

$$(2) \quad \phi_2 = \frac{x}{h}$$

$$(3) \quad \phi_3 = \frac{y}{h}$$

$$\sum \phi_i = 1$$

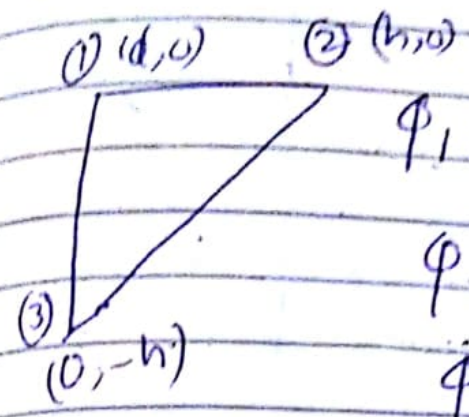
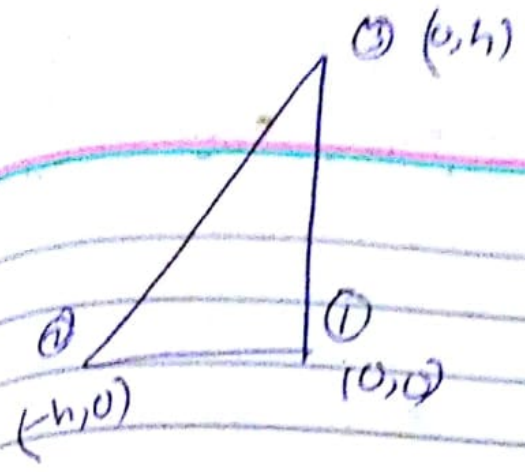
$$h_x = h_y = h$$

(89)

(1) $\phi_1 = \frac{h+x-y}{h}$

(2) $\phi_2 = -x/h$

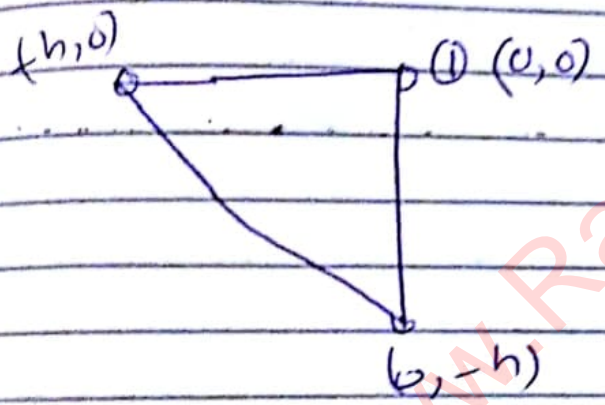
(3) $= y/h$



$\phi_1 = \frac{h-x+y}{h}$

$\phi_2 = x/h$

$\phi_3 = -y/h$



$\phi_1 = \frac{h+x+y}{h}$

$\phi_2 = -x/h$

$\phi_3 = -y/h$

Example:-

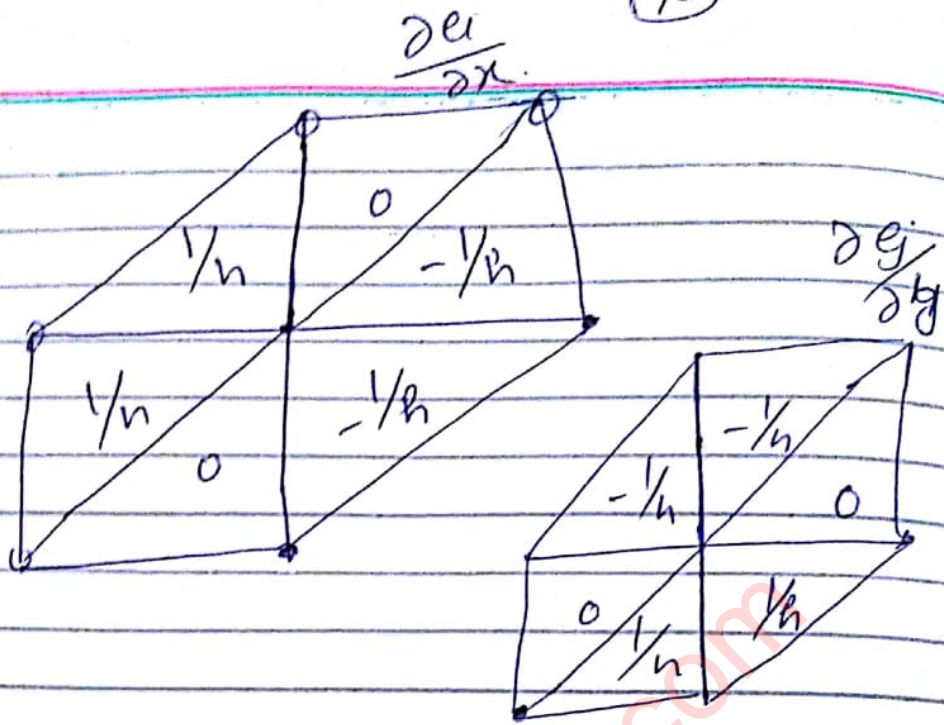
$-\Delta u = 1$

$-u_{xx} - u_{yy} = 1$

$\sum u_i \int \nabla e_i \nabla e_j dx = \int e_j dx$

$\nabla e_i \nabla e_j = \frac{\partial e_i}{\partial x} \frac{\partial e_j}{\partial x} + \frac{\partial e_i}{\partial y} \frac{\partial e_j}{\partial y}$

(90)



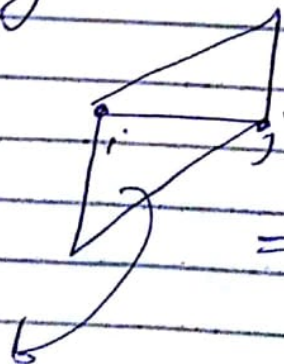
$$\int_{\Omega} \frac{\partial e_j}{\partial x} \cdot \frac{\partial e_i}{\partial x} dx = \frac{4}{h^2} \cdot \frac{h^2}{2} = 2$$

area of triangle

$$\int_{\Omega} \frac{\partial e_j}{\partial y} \cdot \frac{\partial e_j}{\partial y} = 2$$

$$i \neq j \int_{\Omega} \nabla e_i \cdot \nabla e_j dx = 4$$

$i \neq j$



$$\nabla e_i \cdot \nabla e_j = \frac{\partial e_i}{\partial x} \frac{\partial e_j}{\partial x} + \frac{\partial e_i}{\partial y} \frac{\partial e_j}{\partial y}$$

$$= (-1/h)(1/h) + 0(-1/h)$$

$$(-1/h)(1/h) + 0(-1/h) = -1/h^2$$

$$= -1/h^2$$

(91)

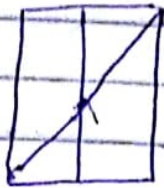
$$\nabla e_i \nabla e_j = \frac{\partial e_i}{\partial x} \frac{\partial e_j}{\partial x} + \dots$$

$$\int \nabla e_j \nabla e_j dx = -\frac{1}{h^2} + \frac{1}{h^2} \times \frac{h^2}{2}$$

$$= -\frac{2}{h^2} \cdot \frac{h^2}{2}$$

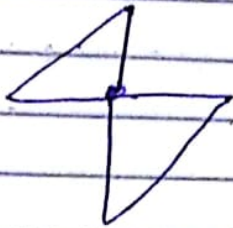
$$= -1$$

(2)



= 0

(3)



= -1

$$\int \nabla e_i \nabla e_j dx = \begin{cases} 4 & i=j \\ -1 & \text{diagonal} \\ 0 & \text{otherwise} \end{cases}$$

$$A = \begin{matrix} \tilde{A} & I \\ I & \tilde{A} \end{matrix}$$

$$0 \quad \tilde{A}$$

$$\tilde{A} = \begin{bmatrix} 4 & -1 & 0 & 0 & \dots \\ -1 & 4 & -1 & 0 & \dots \\ & & 0 & -1 & 4 \end{bmatrix}$$

(92)

$$\textcircled{1} \int_0^h \int_0^x e^y dy dx$$

$$= \int_0^h \int_0^x 1 \cdot e^y dy dx = \int_0^h \left[\frac{e^y}{1} \right]_0^x dx$$

$$= -\frac{h^2}{3}$$

$$\textcircled{2} \int_0^h \int_0^x \frac{-y}{h} dy dx = -\frac{h^2}{6}$$

$$\textcircled{3} \int_0^h \int_0^x \frac{h+x-y}{h} dy dx = \frac{2}{3} h^2$$

$$\textcircled{4} \int_0^h \int_0^x \frac{x}{h} dy dx = \frac{h^2}{3}$$

$$\textcircled{5} \int_0^h \int_0^x \frac{y}{h} dy dx = \frac{h^2}{6}$$

$$\textcircled{6} \int_0^h \int_0^x \frac{h-x+y}{h} dy dx = \frac{h^2}{3}$$

At the end

$$\int_{\Omega} e^y dx = \frac{-h^2}{3} + \frac{h^2}{3} - \frac{h^2}{6} + \frac{h^2}{6} + \frac{h^2}{3} + \frac{2h^2}{3}$$

$$= \frac{h^2 + 2h^2}{3}$$

$$= \frac{h^2}{3}$$

Do yourself

(93)

13-12-2016

Rayleigh-Ritz method for
elliptic PDE, &:-

$$-u_{xx} = f, \quad 0 < x < 1$$

www.RanaMaths.com

(94)

19-12-2016.

Method of characteristics:-

It is a technique for solving PDEs, especially first order PDEs.

This method can be applied to hyperbolic PDEs.

This method reduces given PDE to an ODE, which can be then integrated easily.

Example:-

$$u_t + a(u)u_x = 0, \text{ quasi-linear}$$

$$\frac{\partial u}{\partial t} + \frac{\partial f(u)}{\partial x} = 0$$

$$\frac{\partial u}{\partial t} + \frac{\partial f}{\partial u} \cdot \frac{\partial u}{\partial x} = 0$$

$$\frac{du}{dt} = \frac{\partial u}{\partial t} \cdot \frac{dt}{dt} + \frac{\partial u}{\partial x} \cdot \frac{dx}{dt}$$

$$= \frac{\partial u}{\partial t} + a(u) \frac{\partial u}{\partial x} = 0$$

$$\Rightarrow \frac{du}{dt} = 0 \Rightarrow u(x,t) = u_0(x_0) \text{ (constant)} \quad x(0) = x_0$$

$$\Rightarrow u(x,t) = u_0(x - a(u_0)t)$$

$$\frac{dx}{dt} = a(u) = a(u_0)$$

$$x = a(u_0)t + C$$

$$x_0 = C$$

$$x = a(u_0)t + x_0$$

$$\Rightarrow x_0 = x - a(u_0)t$$

(95)

7

$$\frac{\partial u}{\partial t} + a \frac{\partial u}{\partial x} = 0, a > 0$$

$$u_0(x) = \begin{cases} \frac{1}{2}x, & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\frac{dx}{dt} = a$$

$$\Rightarrow x = at + C \Rightarrow u(x)$$

$$x_0 = x - at$$

$$u(x, 0) = u_0$$

$$u(x, t) = u_0(x_0) = \begin{cases} \frac{1}{2}x_0, & 0 < x_0 < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$u(x, t) = \begin{cases} \frac{1}{2}(x - at), & 0 < x - at < 1 \\ 0 & \text{otherwise} \end{cases}$$

final solution.

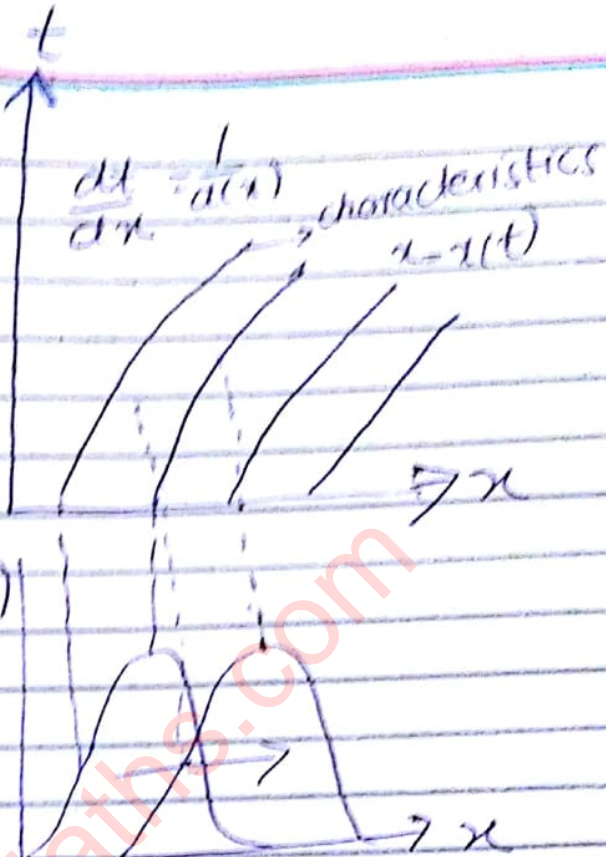
Example 1-

$$u_t - 2u u_x = e^{2x}, \quad u(x, 0) = f(x)$$

Solve by M.C.

$$\frac{du}{dt} = \frac{\partial u}{\partial t} + \frac{\partial u}{\partial x} \frac{dx}{dt} = \frac{\partial u}{\partial t} - 2 \frac{\partial u}{\partial x} = e^{2x}$$

$$\Rightarrow \boxed{\frac{du}{dt} = e^{2x}}$$



(96)

$$\frac{du}{dt} = e^{\frac{2(x_0 - 2t)}{4}}$$

$$= e^{\frac{2x_0 - 4t}{4}}$$

$$u(x, t) = \frac{e^{\frac{2x_0 - 4t}{4}}}{4} + C$$

$$u(x, 0) = \frac{1}{4} e^{\frac{2x_0}{4}} + C$$

$$u(x_0, 0) = \frac{1}{4} e^{\frac{2x_0}{4}} + C$$

$$f(x) = C = \frac{1}{4} e^{\frac{2x_0}{4}}$$

$$C = f(x_0) + \frac{1}{4} e^{\frac{2x_0}{4}}$$

$$u(x, t) = \frac{1}{4} e^{\frac{2x_0 - 4t}{4}} + f(x) + \frac{1}{4} e^{\frac{2x_0}{4}}$$

$$u(x, t) = \frac{1}{4} e^{\frac{2x}{4}} + f(x + 2t) + \frac{1}{4} e^{\frac{2x + 4t}{4}}$$

Example:-

$$u_t + 2u_x = \cos(x)$$

$$u(x, 0) = e^x$$

$$\frac{dx}{dt} = -2, x(0) = x_0$$

$$x = -2t + C$$

$$x_0 = +C$$

$$C = x_0$$

$$x = -2t + x_0$$

$$x_0 = x + 2t$$

or

$$x = x_0 - 2t$$

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$$\frac{du}{dt} = \frac{\partial u}{\partial t} + \frac{\partial u}{\partial x} \cdot \frac{\partial x}{\partial t}$$

$$= \frac{\partial u}{\partial t} + 2 \frac{\partial u}{\partial x} = \cos(x)$$

$$\frac{du}{dt} = \cos(x)$$

$$\frac{dx}{dt} = 2$$

$$x = 2t + C$$

$$x(0) = 0 + C$$

$$C = x_0$$

$$x = 2t + x_0$$

$$x_0 = x - 2t$$

$$\frac{du}{dt} = \cos(x_0 + 2t)$$

$$u(x, t) = -\sin(x) + C$$

$$u(x, 0) = -\sin(x) + C$$

$$e^x = -\sin(x) + C$$

$$C = e^x + \sin(x)$$

$$u(x, t) = -\sin(x) + e^x + \sin(x)$$

$$u = \frac{1}{2} (-\sin(x_0 + 2t)) + C$$

$$u(x, t) = -\frac{1}{2} \sin(x_0 + 2t) + C$$

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$$u(x_0, 0) = -\frac{1}{2} \sin(x_0 + 2(0)) + C$$

$$0 = -\frac{1}{2} \sin(x_0) + C$$

$$C = \frac{1}{2} \sin(x_0)$$

$$u(x_0, t) = +\frac{1}{2} \sin(x_0 + 2t) + e^{-\frac{x_0}{2}} \sin(x_0)$$

$$u(x, t) = +\frac{1}{2} \sin(x - 2t + 2t) + e^{-\frac{x-2t}{2}} \sin(x - 2t)$$

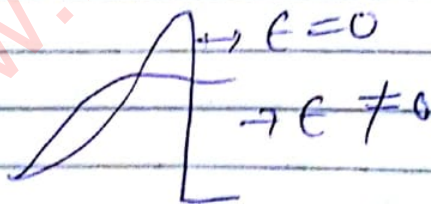
$$= \frac{1}{2} \sin(x - 2t) + e^{-\frac{x-2t}{2}} \sin(x)$$

Simple non linear PDE :-

non linear
hyperbolic

$$u_t + uu_x = 0$$

Burger Equation



$$a(u) = u$$

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

implicit Sol.

$$\frac{\partial u}{\partial x} = \frac{\partial u}{\partial \xi} \cdot \frac{\partial \xi}{\partial x}$$

$$u_x = \frac{\partial u}{\partial \xi} \cdot \frac{\partial \xi}{\partial x}$$

$$u_x = u' (1 - ut)$$

$$u_t = u' (-u_x t - u)$$

$$u_x = \frac{u'}{1 + ut}$$

$$u_t = \frac{-uu'}{1 + ut}$$

21-11-08

(99)

$$u_t + u u_x = 0$$

$$- \frac{u u'}{1+u} + \frac{u u'}{1+u} = 0$$

Example:-

$$u_t + u u_x = 0 \Rightarrow \begin{cases} u(x,0) = u_0(x) \\ \dots \end{cases}$$

$$u(x,0) = u_0(x) = \begin{cases} 0 & x < 0 \\ x_0 & 0 < x < 1 \\ 1 & x > 1 \end{cases}$$

$$\frac{dx}{dt} = u_0(x)$$

$$x(0) = x_0$$

$$\frac{dx}{dt} = 0$$

$$x = \frac{dx}{dt} t + C$$

$$x_0 = x - 0$$

$$\frac{dx}{dt} = x_0$$

$$x = x_0 t + x_0$$

$$= x_0 (1+t)$$

$$x_0 = \frac{x}{1+t}$$

$$\frac{dx}{dt} = 1, \quad x_0 = x - t$$

$$u(x,t) = \begin{cases} 0, & x < t \\ \frac{x}{1+t}, & 0 \leq x \leq 1+t \\ 1, & x > 1+t \end{cases}$$

$$x(0) = x_0$$

(100)

20-12-16

Example: -

$u_t + uu_x$ → Burger Eq (inviscid)

$$u(x, 0) = u_0(x) = \begin{cases} 1 & x < 0 \\ 1-x & 0 \leq x \leq 1 \\ 0 & x > 1 \end{cases}$$

$$\frac{dx}{dt} = u_0(x_0) = \begin{cases} 1 & x_0 < 0 \\ 1-x_0 & 0 \leq x_0 \leq 1 \\ 0 & x_0 > 1 \end{cases}$$

$$\frac{dx}{dt} = 1 \Rightarrow x = t + C \Rightarrow x_0 = x - t$$

$$\frac{dx}{dt} = 1 - x_0 \Rightarrow x = (1 - x_0)t + C \Rightarrow x = (1 - x_0)t + x_0$$

$$x_0 = \frac{x - t}{1 - t}$$

$$\frac{dx}{dt} = 0 \Rightarrow x = x_0$$

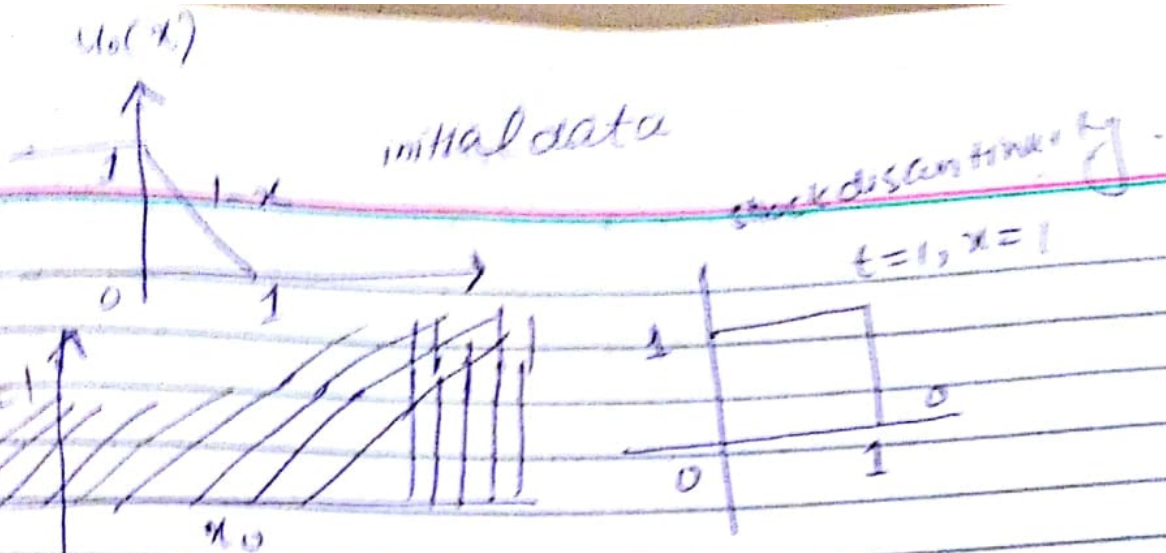
$$\Rightarrow u(t, x) = u_0(x_0) = \begin{cases} 1 & x < t \\ \frac{1-x}{1-t} & 0 < \frac{x-t}{1-t} < 1 \\ 0 & (t \leq x \leq 1) \\ 0 & x > 1 \end{cases}$$

The solution has a pole (singularity)

at $t=1$

Thus above solution is only valid for $0 \leq t < 1$

For $t=1$: The characteristics intersect each other, the slope of solution is infinite.



- ① shock wave
- ② refraction wave
- ③ contact wave.

$$u_t + uu_x = 0$$

$$\Rightarrow u_t + \left(\frac{1}{2}u^2\right)_x = 0$$

$$u_t + f(u)_x = 0 \rightarrow \text{Non-linear hyperbolic equation}$$

$$\int_{x_L}^{x_R} u_t dx + \int_{x_L}^{x_R} f(u)_x dx = 0$$

$$\Rightarrow \frac{\partial}{\partial t} \int_{x_L}^{x_R} u dx = - (f(u_R) - f(u_L))$$

$$\frac{1}{\delta t} (u_L - u_R) \cdot \delta x = f(u_L) - f(u_R)$$

$$X_s = \frac{\delta x}{\delta t} = \frac{f(u_L) - f(u_R)}{u_L - u_R} = \frac{1}{2} (u_L + u_R)$$

Shock speed for Burgers Eq

(102)

$$\dot{x}_s = \frac{1}{2}(U_L + U_R)$$

$$= \frac{1}{2}(1+0)$$

$$x_s = \frac{1}{2}t + C$$

$$x_s(1) = 1$$

$$1 = \frac{1}{2} + C \Rightarrow C = \frac{1}{2}$$

$$x_s = \frac{t+1}{2} \quad \text{for } t \geq 1$$

$$u(t, x) = \begin{cases} 0 & x < x_s \\ 1 & x > x_s \end{cases}$$

Example:-

$$u_t + 200u_x = 0$$

$$u(x, 0) = U_0(x) = \begin{cases} 1 & x < 0 \\ \frac{1-x}{2} & 0 \leq x \leq 1 \\ 0 & x > 1 \end{cases}$$

$$\frac{dx}{dt} = 200 U_0(x_0) = \begin{cases} 1 & x_0 < 0 \\ 1-x_0 & 0 \leq x_0 \leq 1 \\ 0 & x_0 > 1 \end{cases}$$

$$\frac{dx}{dt} = 1, \quad x = t + C \Rightarrow x_0 = x - t$$

$$\frac{dx}{dt} = 1 - x, \quad x = (1-x)t + C$$

(102)

$$x_0 = (1-x_0)t + c$$

$$x_0 = -t + x_0 t$$

$$x_0 = t(x_0 - 1)$$

$$x_0 = (1-x)t + (x_0 - 1)t$$

$$x_0 = c$$

$$x = (1-x)t + x_0$$

~~$$x_0 = x - t + t x$$~~

$$x_0 = x - t + t x$$

$$= x - t + t x$$

$$x_0 = x - t + t x$$

$$\frac{dx}{dt} = 0$$

$$x = c$$

$$x = x_0$$

$$x_s = \frac{f(v_L) - f(v_R)}{v_L - v_R}$$

$$= \frac{1}{2} + 0$$

$$x_s = \frac{1}{2} \quad \text{Same as above}$$

(104)

Example

$$u_t + uu_x = 0$$

$$\frac{\partial u}{\partial t} = u(x, 0) = \begin{cases} 2 & x < 0 \\ 2-x & 0 \leq x < 2 \\ 0 & x \geq 2 \end{cases}$$

$$u(t, x) = \begin{cases} 2 & x < 2t \\ \frac{2-x}{1-t} & 2t \leq x \leq 2 \\ 0 & x \geq 2 \end{cases}$$