

# MAST90026 Computational Differential Equations: Week 1

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# Goal of the subject

Cover numerical techniques for solving

- Boundary value problems (BVPs) for ODEs
- Simple PDEs
  - 2D Elliptic PDES
  - 1+1 (1 space dimension, 1 time dimension) parabolic, hyperbolic
- Sparse solvers for linear systems

# Model equation

Convection-diffusion equation:

$$u_t + \vec{v} \cdot \nabla u = \nabla \cdot (D \nabla u) + f.$$

$$\nabla = \left( \frac{\partial}{\partial x}, \frac{\partial}{\partial y} \right)$$

$\vec{v}, D > 0, f$  given functions of  $(x, y)$

Scalar, linear parabolic equation

## Model equation

Limiting cases  $\vec{v} = \vec{0}$ : Heat equation

$$u_t = \nabla \cdot (\mathbf{D} \nabla u) + f.$$

1D heat equation:

$$u_t = Du_{xx} + f.$$

## Model equation

Limiting cases  $D = 0$ : Advection/transport/1-way wave equation

$$u_t + \vec{v} \cdot \nabla u = f.$$

1D hyperbolic equation:

$$u_t + vu_x + f.$$

# Model equation

Limiting cases  $u_t = 0$  (steady state): elliptic equation

$$\vec{v} \cdot \nabla u = \nabla \cdot (\mathbf{D} \nabla u) + f.$$

Further  $\vec{v} = \vec{0} \rightarrow$  Poisson equation

$$-\nabla \cdot (\mathbf{D} \nabla u) = f.$$

1D elliptic equation

$$\nu u_x = \mathbf{D} u_{xx} + f.$$

# Issues

- How do we iterate in time (for IVPs)?
- How do we represent the solution in space (for BVPs/PDEs)?
- How do we handle higher dimensionality?
- How do we handle non-linearities?

## 2nd order BVPs

Linear BVP

$$-u'' + p(x)u' + q(x)u = r(x), x \in [a, b]$$

with linear separated boundary condition (Robin boundary condition)

$$A_{11}u(a) + A_{12}u'(a) = \beta_1,$$

$$A_{21}u(b) + A_{22}u'(b) = \beta_2.$$

## More general 2nd order BVPs

General form

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

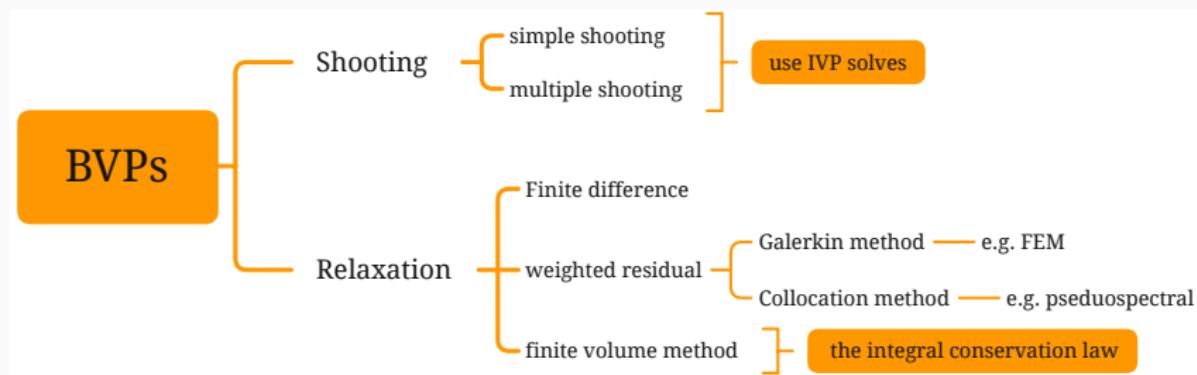
with separated boundary condition at  $x = a$  and  $x = b$

$$g_1(u(a), u'(a)) = B_1,$$

$$g_2(u(b), u'(b)) = \beta_2.$$

Note linear BVPs could have 0, 1, or  $\infty$  many solutions.

# Existing methods for boundary value problems



## General principle

Consider

$$Lu = r(x)$$

where  $L$  is a linear operator involving derivatives.

The main difference between finite difference methods(FDMs) and weighted residual methods (WRMs):

1. For FDM, approximate the differential operator  $L$ .
2. For WRMs, approximate the solution  $u$ .

## Finite difference formulae

Idea: Approximate  $u'$  and  $u''$ .

Only approximate derivatives at selected points:  $h = \frac{b-a}{N+1}$  and  $x_j = a + jh$



Can also use nonuniform meshes  $\{x_j\}$  with  $h_j = x_{j+1} - x_j$ .

## Some examples

For  $u'$ :

- Forward difference:  $u'(x) = \frac{u(x+h) - u(x)}{h} + \text{error.}$
- Backward difference:  $u'(x) = \frac{u(x) - u(x-h)}{h} + \text{error.}$
- Central difference:  $u'(x) = \frac{u(x+h) - u(x-h)}{2h} + \text{error.}$

For  $u''$ :

- Central difference:  $u''(x) = \frac{u(x+h) - 2u(x) + u(x-h)}{h^2} + \text{error.}$

## General finite difference scheme

In general, approximate  $u^{(k)}$  as

$$u^{(k)}(x) = \sum_{i=1}^n c_i u(x + h_i) + \text{error}.$$

Derivation:

- The method of underdetermined coefficient (Taylor series expansion)
- Interpolation and then differentiation

Error for interpolating using  $n$  points is  $\mathcal{O}(h^n)$ . The worst-case error for approximating  $u^{(k)}$  is  $\mathcal{O}(h^{n-k})$ .

## Discretization error

Central finite difference error:

$$\begin{aligned}& \frac{u(x+h) - u(x-h)}{2h} \\&= \frac{1}{2h} \left( u(x) + hu'(x) + \frac{1}{2}h^2 u''(x) + O(h^3) \right) - \\&\quad \frac{1}{2h} \left( u(x) - hu'(x) + \frac{1}{2}h^2 u''(x) + O(h^3) \right) \\&= \frac{2hu'(x) + O(h^3)}{2h} \\&= u'(x) + O(h^2),\end{aligned}$$

Similarly, the error for FD, BD, CD for  $u''$ .

## FDM for BVPs

Consider

$$\begin{aligned} u''(x) + p(x)u'(x) + q(x)u(x) &= r(x), \\ u(a) = \alpha, u(b) = \beta. \end{aligned}$$

Consider  $h = \frac{b-a}{N+1}$  and  $x_j = a + jh$ :



**Idea:** Use central difference approximation for  $u'$  and  $u''$  at  $x_j$ :

$$\frac{u_{j+1} - 2u_j + u_{j-1}}{h^2} + p(x_j) \frac{u_{j+1} - u_{j-1}}{2h} + q(x_j)u_j = r(x_j).$$

## Linear system for FDM

Let  $p(x_j) = p_j$ ,  $q(x_j) = q_j$  and  $r(x_j) = r_j$ .

$$u_0 = \alpha,$$

$$\frac{u_0 - 2u_1 + u_2}{h^2} + p_1 \frac{u_2 - u_0}{2h} + q_1 u_1 = r_1,$$

$$\frac{u_1 - 2u_2 + u_3}{h^2} + p_2 \frac{u_3 - u_1}{2h} + q_2 u_2 = r_2,$$

⋮

$$\frac{u_{N-1} - 2u_N + u_{N+1}}{h^2} + p_N \frac{u_{N+1} - u_{N-1}}{2h} + q_N u_N = r_N,$$

$$u_{N+1} = \beta.$$

## Simplification of Linear system for FDM

Change the second and second last equation to:

$$\begin{aligned}\frac{-2u_1 + u_2}{h^2} + p_1 \frac{u_2}{2h} + q_1 u_1 &= r_1 - \frac{\alpha}{h^2} + p_1 \frac{\alpha}{2h}, \\ \frac{u_{N-1} - 2u_N}{h^2} + p_N \frac{-u_{N-1}}{2h} + q_N u_N &= r_N - \frac{\beta}{h^2} - p_N \frac{\beta}{2h}.\end{aligned}$$

## Matrix form

$$\begin{aligned}
 & \frac{1}{h^2} \begin{pmatrix} -2 + h^2 q_1 & 1 + h \frac{p_1}{2} & 0 & 0 \\ 1 - h \frac{p_2}{2} & \ddots & \ddots & 0 \\ 0 & \ddots & \ddots & 1 + h \frac{p_{N-1}}{2} \\ 0 & 0 & 1 - h \frac{p_N}{2} & -2 + h^2 q_N \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_N \end{pmatrix} \\
 &= \begin{pmatrix} r_1 - \frac{\alpha}{h^2} + p_1 \frac{\alpha}{2h} \\ r_2 \\ \vdots \\ r_N - \frac{\beta}{h^2} - p_N \frac{\beta}{2h} \end{pmatrix}
 \end{aligned}$$

## Questions to ask

1. Does this have a solution?
2. How hard is it to solve?
3. How well does  $\mathbf{u}$  approximate  $u(x)$  for a given  $h$ ?
4. Does  $u_j \rightarrow u(x_j)$  as  $h \rightarrow 0$ ?

## Measure the error

The grid function (i.e. it is only defined on the grid/mesh) norms:

$$\|\Sigma\|_1 = h \sum |\Sigma_i|,$$

$$\|\Sigma\|_2 = \left( h \sum |\Sigma_i|^2 \right)^{\frac{1}{2}},$$

$$\|\Sigma\|_\infty = \max |\Sigma_i|.$$

Define the global error:

$$\|E\| = \|\mathbf{u} - \hat{\mathbf{u}}\|.$$

# Some terminologies

## Definition : Convergence

A method with global error  $\|E\| = O(h^p)$  as  $h \rightarrow 0$  is convergent of order  $p$ .

## Definition : Local Truncation Error (LTE)

The local truncation error  $\tau_j$  is the residual at grid point  $x_j$  when the true solution is put into the finite difference formula.

## Definition : Consistency

A method is consistent of order  $p$  if  $\|\tau\| = O(h^p)$  as  $h \rightarrow 0$ .

## LTE for our system

$$\begin{aligned}\tau_j &= \frac{1}{h^2} (u(x_{j+1}) - 2u(x_j) + u(x_{j-1})) + p(x_j) \frac{u(x_{j+1}) - u(x_{j-1})}{2h} + \\ &\quad q(x_j)u(x_j) - r(x_j) \\ &= u''(x_j) + p(x_j)u'(x_j) + q(x_j)u(x_j) - r(x_j) + Ch^2u^{(3)} + O(h^4) \\ &= Ch^2u^{(3)}(\xi) + O(h^4),\end{aligned}$$

## Error equation

Original scheme

$$A\mathbf{u} = \mathbf{b}.$$

Consistency

$$A\hat{\mathbf{u}} = \mathbf{b} + \tau.$$

The global error  $\mathbf{E} = \mathbf{u} - \hat{\mathbf{u}}$  satisfies

$$A(\mathbf{u} - \hat{\mathbf{u}}) = -\tau$$

# Convergence

Error equation

$$\mathbf{E} = -A^{-1}\tau.$$

Taking norms

$$\begin{aligned}\|\mathbf{E}\| &= \| -A^{-1}\tau \| \\ &\leq \|A^{-1}\| \|\tau\| \\ &\leq \|A^{-1}\| Ch^2.\end{aligned}$$

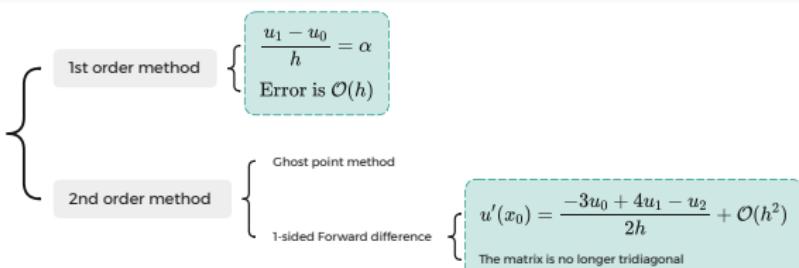
# Summary

Consistency + Stability  $\Rightarrow$  Convergence

# Other type boundary condition

Taking  $u'(x_0) = \alpha$  as an example.

## Handling Nuemann BCs



## Ghost point method



Apply ODE at  $x_0$  and use central difference on BC to eliminate  $u_{-1}$ :

$$\frac{u_{-1} - 2u_0 + u_1}{h^2} + p_0 \left( \frac{u_1 - u_{-1}}{2h} \right) + q_0 u_0 = r_0$$
$$\frac{u_1 - u_{-1}}{2h} = \alpha$$

$$\Rightarrow \frac{-2u_0 + 2u_1}{h^2} + q_0 u_0 = r_0 - p_0 \alpha + \frac{2\alpha}{h}$$