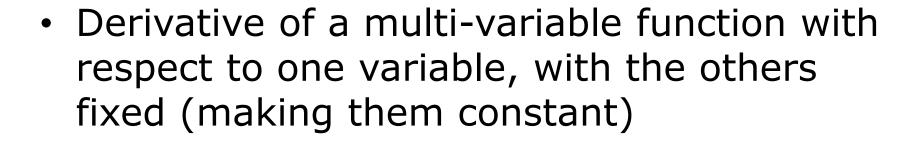
# Partial Differential Equations

## IPCST Seoul National University

#### **Partial Derivative**



• Ex.) 
$$f(x,y) = x^2 + xy + y^2$$
  $f_x \text{ or } f_{,x} \equiv \partial f/\partial x$   $f_{xx} \text{ or } f_{,xx} \equiv \partial^2 f/\partial x^2$   $f_{xx} \equiv \partial^2 f/\partial x^2$   $f_{xx} \equiv \partial^2 f/\partial x^2$   $f_{xx} \equiv \partial^2 f/\partial x$   $f_{xx} \equiv \partial^2 f/\partial x$   $f_{xx} \equiv \partial^2 f/\partial x$ 

### **Partial Differential Equation**



- Differential equation × multivariable function(s) × partial derivatives
- Variables in PDEs
  - Continuous variables
  - Two or more independent variables
    - State variables: functions of independent variables.

# 参考: Partial Differential Equation

#### Examples

- 1-D advection equation (or flow equation)
  - $cu_x + u_t = 0$
- 1-D wave equation
  - $c^2 u_{xx} u_{tt} = 0$
- 1-D diffusion equation (heat equation)
  - $u_t = Du_{xx}$
- 2-D Poisson's equation
  - $u_{xx} + u_{yy} = f(x,y)$
- 2-D Helmholtz equation
  - $u_{xx} + u_{yy} + k^2 u = 0$

# 参考: Partial Differential Equation

#### Examples

- 1-D Klein-Gordon equation
  - $\psi_{tt}$   $\psi_{xx}$  +  $m^2\psi$  = 0
- 1-D incompressible Navier-Stokes equation

• 
$$u_t + u \cdot u_x - v^2 u_{xx} = -p_x/\rho$$

- Burgers' equation
  - $u_t + u \cdot u_x = v^2 u_{xx}$
- Inviscid Burgers' equation
  - $u_t + u \cdot u_x = 0$
- Black-Scholes equation
  - $V_t + \frac{1}{2}\sigma^2 s^2 V_{ss} + rsV_s rV = 0$

# /参考: Laplace Operator (= Laplacian)

$$\Delta f = \nabla^2 f = \nabla \cdot \nabla f$$

$$\nabla f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial^2 f}{\partial z^2}.$$

$$\Delta f = \frac{1}{\rho} \frac{\partial}{\partial \rho} \left( \rho \frac{\partial f}{\partial \rho} \right) + \frac{1}{\rho^2} \frac{\partial^2 f}{\partial \varphi^2} + \frac{\partial^2 f}{\partial z^2}.$$

$$\Delta f = \frac{1}{r^2} \frac{\partial}{\partial r} \left( r^2 \frac{\partial f}{\partial r} \right) + \frac{1}{r^2 \sin \theta} \frac{\partial}{\partial \theta} \left( \sin \theta \frac{\partial f}{\partial \theta} \right) + \frac{1}{r^2 \sin^2 \theta} \frac{\partial^2 f}{\partial \varphi^2}.$$

## Types of 2<sup>nd</sup> order PDEs



- Elliptic equations
  - Poisson, Laplace (f = 0)

$$\triangle u = f(\vec{x})$$

- Parabolic equations
  - Heat, dispersion

$$\frac{\partial u}{\partial t} - \alpha \nabla^2 u = 0$$

- Hyperbolic equations
  - Advection(convection), wave

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

- Mixed type
  - Euler-Tricomi

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

### Types of 2<sup>nd</sup> order PDEs

• 
$$Au_{xx} + 2Bu_{xy} + Cu_{yy} = F(u_x, u_y, u, ....)$$

- Elliptic equations if  $AC B^2 > 0$ 
  - $u_{xx} + u_{yy} = f(x,y)$
- Parabolic equations if  $AC B^2 = 0$ 
  - $u_t = Du_{xx}$
- Hyperbolic equations if  $AC B^2 < 0$ 
  - $c^2 u_{xx} u_{tt} = 0$  ( =  $(c \partial_x + \partial_t)(c \partial_x \partial_t)u$  )
  - $cu_x + u_t = 0$

#### **Initial Conditions of PDEs**

- Elliptic equations
  - Static. Steady-state solution. No initial condition is needed.
- Parabolic or Hyperbolic equations
  - Initial-boundary-value problems (IBVPs)
  - Initial condition is defined as a function of space variables.

## **Boundary Conditions of PDEs**



- Periodic boundary condition
  - If period = L,
    - u(x+L) = u(x)
- B. C. for elliptic or parabolic PDEs
  - Dirichlet B. C.
    - u(boundary) = value
  - Neumann B. C.
    - $\partial u(\text{boundary})/\partial x = \text{value}$
    - 2-D or 3-D: derivative normal to the boundary
  - Robin B. C.
    - · Linear combination of Dirichlet & Neumann BCs.

# **Boundary Conditions of PDEs**



- B. C. for hyperbolic PDEs
  - Cauchy B. C.
    - Both the function and derivative
    - For the 2<sup>nd</sup> order PDE with single initial condition
      - Dirichlet or Neumann BCs are applicable, but the solution is not unique.
  - If there are double initial conditions for every position, the solution is unique even with Dirichlet or Neumann BCs.
    - Cauchy problem: the 2<sup>nd</sup> order PDE with initial state distribution u(t=0) and initial change rate distribution  $\partial_t u(t=0)$



# 參考: Usual Modeling with PDEs



- 1. Selection of state variables from a phenomenological observation
- 2. Modeling of the interaction between the inner system (the main model) and the outer environment
- 3. Finding the equilibrium, conservation, and/or balance equations
- Constitutive or material modeling (equivalent to approximation of the interaction among individuals in microscopic modeling)
- 5. Deriving a set of equations describing the evolution of the state variable(s) in time and space

# 參考: Usual Modeling with PDEs

- Constitutive Relation
  - A relation between two physical quantities
  - Types of constitutive relations
    - 1) Definitions or physical laws
    - Phenomenological or empirical
    - Derived from first-principles or microscopic model calculations
- Systems may be composed by different interconnected systems. One needs different models for each sub-system and compatibility conditions (as I. C. and/or B. C.) between contiguous systems.



## Dismantling a PDE into ODEs



- Methods to solve a PDE by transforming it into a system of ODEs or ODE + PDE.
  - Method of characteristics (for hyperbolic equations)
  - Separation of variables
  - Method of lines

# 參考: Method of Characteristics

- Characteristic: a curve of singularities characterizing a hyperbolic equation
  - Ex.) simple case advection equation
    - $cu_x + u_t = 0$  (c: constant)

```
Since c is the advection velocity,

dx/dt = c \rightarrow x = ct + x_0 \rightarrow x_0 = x - ct
du/dt = (\partial u/\partial x)(dx/dt) + (\partial u/\partial t) = cu_x + u_t = 0
\rightarrow u = f(x_0)
\therefore u = f(x - ct) : general solution
```

• Ex.) 
$$\frac{\partial^2 u}{\partial t^2} = \triangle u$$
 (2-D)

Let u = T(t)V(x,y)

Then,

$$\frac{\frac{d^2}{dt^2}T(t)}{T(t)} = \frac{\nabla^2 V(x,y)}{V(x,y)} = -\lambda$$

 $\checkmark$   $\lambda \ge 0$  for steady state solutions

$$\frac{\frac{d^2}{dt^2}T(t)}{T(t)} = -\lambda$$

- $\lambda > 0$ : Let  $\lambda = k^2$  $T(t) = A\cos(kt) + B\sin(kt)$
- $\lambda < 0$ : Let  $\lambda = -\kappa^2$  $T(t) = A \exp(-\kappa t) + B \exp(\kappa t)$

- Separation of space variables depends on the system's symmetry
  - Cuboid → Cartesian coordinates
    - V(x, y, z) = X(x) Y(y) Z(z)
  - Cylinder → cylindrical coordinates
    - $V(r, \theta, z) = R(r) \Theta(\theta) Z(z)$
  - Sphere → spherical coordinates
    - $V(r, \theta, \varphi) = R(r) \Theta(\theta) \Phi(\varphi)$
  - Irregular shape: Go numerical!

- Because the time variable part is usually easily separated, the method of variable separation can be used with other methods.
  - Time part: ODE (often analytically solvable)
  - Space part: a numerical method

### **Superposition Principle**

- If f and g are solutions of a homogeneous PDE, u = af + bg is also a solution. (a,b): constants)
  - A PDE is a homogeneous PDE if u = 0 is its solution. Otherwise, it is an inhomogeneous PDE.
- If h is a solution of an inhomogeneous PDE, u = af + bg + h is also a solution.
- \* This principle is used for separation of variables before initial conditions are applied.

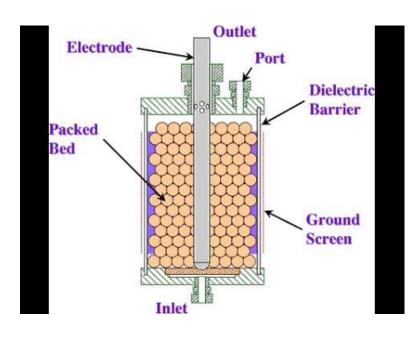
#### **Method of Lines**

- 1. Discretizing space (usually by finite difference methods)
- PDE → ODEs on a grid: an ODE for each grid point → system of ODEs
  - Method of lines can be used to construct or analyze a numerical method by leaving the time variable continuous.
- 3. Solve the system of ODEs by an ODE solver





- Example: Bioreactor model
  - J. Chem. Technol. Biotechnol. **74**, 78 (1999); Automatica **28**, 873 (1992)



- A fixed bed bioreactor
- Two reactions
  - 1. Growth:substrate + biomass→ entrapped on a bed
  - 2. Death of micro-organisms
- State variables
  - X: biomass concentration
  - S: substrate concentration

#### **Method of Lines**



J. Chem. Technol. Biotechnol. 74, 78 (1999); Automatica 28, 873 (1992)

$$\partial X/\partial t = \mu(X, S)X - k_{\rm d}X$$

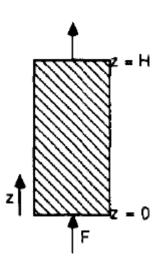
$$\partial S/\partial t = -(F/A)\partial S/\partial z - k_{Y}\mu(X, S)X$$



$$dX_i/dt = \mu(X_i, S_i)X_i - k_d X_i$$

$$dS_{i}/dt = -(F/A)(S_{i} - S_{i-1})/\Delta z - k_{Y} \mu(X_{i}, S_{i})X_{i}$$

$$S_i = S(z_i, t), \quad X_i = X(z_i, t), \quad z_i = i \Delta z,$$
 
$$S_0 = S_{\rm in}$$



**F** : flow

A: area of

cross-section



## **Numerical Methods for PDEs**



- Traditional space discretization
- Finite Difference Method (FDM)
  - Numerical differentiation on a uniform mesh
- Finite Element Method (FEM)
  - Sub-domain division → Element equation over each sub-domain → Connecting solutions
- Finite Volume Method (FVM)
  - PDEs are recast in a conservative form by using the divergence theorem and are solved over discrete volumes. (Application in fluid dynamics)



## **Numerical Methods for PDEs**



- Others
- Lattice Boltzmann Method
  - Transforming PDEs to integral equations. Matrix-free.
- Boundary Element Method
  - FEM on boundaries.
- Spectral Element Method
- Differential Quadrature Method

#### **Simple FDM**

- Example methods of applying finite differences directly
  - Euler method
    - Time derivative → Forward or Backward 2-point
  - Five-point stencil
    - 2D space Laplacian → Central 3-point
  - Leapfrog method
    - Time derivative → Central 3-point
    - Space derivative → Central 3-point



Forward Euler method

$$\frac{\partial u}{\partial t} \to \frac{u(\vec{x}, t + \delta) - u(\vec{x}, t)}{\delta}$$

•  $\delta$ : time interval ( $\Delta t$ )

Ex.) 
$$\partial_t u = \triangle u \rightarrow u(\vec{x}, t + \delta) = u(\vec{x}, t) + \delta \triangle_h u(\vec{x}, t)$$

- h: space interval  $(\Delta x)$
- This method is explicit.
- However, parabolic PDEs are often too stiff to apply this method.



- Forward Euler method
  - 1-D diffusion equation

$$u_t = Du_{xx}$$

(uniform grids)

• Let 
$$U_i^k = u(x_i, t_k)$$
 where  $x_i = ih + x_0, t_k = k\delta + t_0$ 

$$u_t \to \frac{U_i^{k+1} - U_i^k}{\delta}$$

$$u_{xx} \to \frac{U_{i+1}^k - 2U_i^k + U_{i-1}^k}{h^2}$$

$$u_t = Du_{xx} \to \frac{U_i^{k+1} - U_i^k}{\delta} = D \frac{U_{i+1}^k - 2U_i^k + U_{i-1}^k}{h^2}$$

• Let 
$$\gamma = D\delta/h^2$$
, then  $U_i^{k+1} = \gamma U_{i+1}^k + (1-2\gamma)U_i^k + \gamma U_{i-1}^k$ 



1-D diffusion equation

$$u_t = Du_{xx}$$

- Let the domain = [0, L]
- Dirichlet boundary conditions

$$u(0) = c \rightarrow U_0^k = c$$
  
$$u(L) = c \rightarrow U_N^k = c$$

- Neumann boundary conditions
  - 1) Forward FD or Backward FD
  - 2) Ghost boundary points



- Forward Euler method
  - Discrete maximum principle

$$\max_{i} \left| U_i^{k+1} \right| \le \max_{i} \left| U_i^k \right| \quad \text{for every } k$$

→ (sufficient) Stability condition for 1-D diffusion eq.

$$\gamma \le \frac{1}{2} \to 2D\delta \le h^2$$

- Algorithm for 1-D diffusion equation
  - 1. Setting parameters
  - 2. Setting the initial condition & time-invariant Dirichlet BCs.
  - 3. Loop: applying  $U_i^{k+1} = \gamma U_{i+1}^k + (1-2\gamma)U_i^k + \gamma U_{i-1}^k$  with Neumann & time-variant BCs.

#### Do It Yourself

Make your code to solve

$$u_t = 4u_{xx} + 1, \qquad 0 < x < 1, t > 0$$

- Initial condition: u(x,0) = 0, 0 < x < 1
- Boundary conditions:  $u(t,0) = u(t,1) = 0, t \ge 0$  with the forward Euler method.
  - Wen Shen 11.5-2
  - The main formula is a little bit changed.
- Check out the results every time step.
  - [option]: numbers, graphs, animation

## Five-point Stencil (2D)

5-point discrete Laplacian (2D)

$$\triangle u = u_{xx} + u_{yy}$$
- Let  $U_{i,j} = u(x_i, y_j)$  where  $x_i = ih, y_j = jh$ 

$$u_{xx, h} = (U_{i-1,j} - 2U_{i,j} + U_{i+1,j})/h^2$$

$$u_{yy, h} = (U_{i,j-1} - 2U_{i,j} + U_{i,j+1})/h^2$$

$$\triangle_h u = u_{xx, h} + u_{yy, h}$$

$$= (U_{i-1,j} + U_{i+1,j} + U_{i,j-1} + U_{i,j+1} - 4U_{i,j})/h^2$$



Simple advection equation

$$\partial_x u + \partial_t u = 0$$

- Backward central (or BTCS)
  - Let  $U_i^k = u(x_i, t_k)$  where  $x_i = ih$ ,  $t_k = k\delta$
  - Time: backward difference

• 
$$\partial_{t,\delta} U_i^k = (U_i^k - U_i^{k-1})/\delta$$

Space: central difference

• 
$$\partial_{x,h} U_i^k = (U_{i+1}^k - U_{i-1}^k)/(2h)$$

where  $\gamma = \delta/(2h)$ 



Backward central (or BTCS)

$$\gamma u_{i+1}^{k+1} + u_i^{k+1} - \gamma u_{i-1}^{k+1} = u_i^{k}$$

- Matrix-vector form:  $\mathbf{M}\mathbf{u}^{k+1} = \mathbf{u}^k$ 

$$\mathbf{M}\mathbf{u}^{k+1} = \mathbf{u}^k$$

In case of Dirichlet B. C., 0 at the ends,

$$\mathbf{M} = \begin{pmatrix} 1 & \gamma & 0 & 0 & \cdots & 0 \\ -\gamma & 1 & \gamma & 0 & \cdots & 0 \\ 0 & -\gamma & 1 & \gamma & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ \cdots & \cdots & -\gamma & 1 & \gamma & 0 \\ 0 & \cdots & 0 & -\gamma & 1 & \gamma \\ 0 & \cdots & 0 & 0 & -\gamma & 1 \end{pmatrix}$$



Backward central (or BTCS)

$$\gamma u_{i+1}^{k+1} + u_i^{k+1} - \gamma u_{i-1}^{k+1} = u_i^{k}$$

- Algorithm
  - 1. Setting parameters
  - 2. Setting **u**<sup>0</sup> & time-invariant Dirichlet BCs.
  - 3. If BCs have no variation in time, set M under Neumann BCs before the main loop  $\rightarrow$  find M<sup>-1</sup>
  - 4. Loop: solving  $\mathbf{M}\mathbf{u}^{k+1} = \mathbf{u}^k$  with time-variant BCs. The different form of the PDE can change  $\mathbf{M}$ .
- Stability
  - Unconditionally stable because  $\|\mathbf{M}\| > 1$



$$\partial_x u + \partial_t u = 0$$

- Leapfrog method
  - Let  $U_i^k = u(x_i, t_k)$  where  $x_i = ih$ ,  $t_k = k\delta$
  - Time: central difference

• 
$$\partial_{t,\delta} U_i^k = (U_i^{k+1} - U_i^{k-1})/(2\delta)$$

Space: central difference

• 
$$\partial_{x,h} U_i^k = (U_{i+1}^k - U_{i-1}^k)/(2h)$$

$$\rightarrow U_i^{k+1} = U_i^{k-1} - \gamma (U_{i+1}^k - U_{i-1}^k)$$

where  $\gamma = \delta/h$ 

## FDM for 1-D Hyperbolic PDE



- Leapfrog method
  - Stability condition:  $\gamma \leq 1$
  - Algorithm
    - 1. Setting parameters
    - 2. Setting the initial condition & time-invariant Dirichlet BCs.
    - Beginning with the Euler method
    - 4. Loop: applying  $U_i^{k+1} = U_i^{k-1} \gamma(U_{i+1}^k U_{i-1}^k)$  with Neumann & time-variant Dirichlet BCs.

### 參考: Multi-Grid Technique

- Use of different scales of discretization
- Ex.) use of coarse and fine grids
  - Coarse grid: smooth function
  - Fine grid: highly oscillating function
  - A coarse grid can be used for correction of a fine-grid solution.
- This can be applied to any discretizing method (FDM, FEM, .....).



### 參考: Multi-Grid Technique



- Algorithm framework
  - 1. Smoothing: rough calculation on a fine grid
  - 2. Restriction: fine  $\rightarrow$  coarse (error transfer)
  - 3. Calculation on the coarse grid
  - 4. Interpolation or prolongation: coarse → fine
  - 5. Further calculation on the fine grid

- BVP on domain → BVPs on subdomains
  - In addition to boundary conditions, we need conditions at interfaces or in overlapping regions
- Usefulness
  - 1. Efficient parallel computing
  - 2. It is often useful to use different time steps or grids on different subdomains.
  - More often used for FEM than FDM

- Overlap conditions
  - 1. Domains overlap
  - 2. Domains do not overlap, but they are appended with buffer regions
  - 3. Without buffer regions, domains intersect only along an interface

- Simple example: FDM of 1-D heat eq.
  - $\partial_t u = \partial_x^2 u$
  - u(x, 0) = f(x); u(0, t) = u(1, t) = 0
  - Let  $U_i^n \equiv u(x_i, t_n)$  where  $x_i = ih$ ,  $t_n = n\delta$
  - Assume each subdomain ranges from one interface point to the next interface point. Then,
    - $U_i^n = 0$
    - $\partial_{t,\delta} U_i^n = \partial_{x,h}^2 U_i^{n-1}$
    - $\partial_{t,\delta} U_i^n = \partial^2_{x,h} U_i^n$

at boundary points

at interface points

at interior points

Ref.) C. N. Dawson *et al.*, "A Finite Difference Domain Decomposition Algorithm for Numerical Solution of the Heat Equation", Math. Comput. **57**, 63 (1991).

Simple example: FDM of 1-D heat eq.

• 
$$U_i^n = 0$$

at boundary points

• 
$$\partial_{t,\delta} U_i^n = \partial_{x,h}^2 U_i^{n-1}$$

at interface points

• 
$$\partial_{t,\delta} U_i^n = \partial_{x,h}^2 U_i^n$$

at interior points

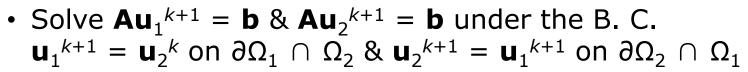
where 
$$\partial_{t, \delta} U_i^n = (U_i^n - U_i^{n-1})/\delta$$
,  
 $\partial_{x, h}^2 U_i^n = (U_{i-1}^n - 2U_i^n + U_{i+1}^n)/h^2$ 

- Explicit for interface and implicit for interior
  - After computing the interface values, the interior values in each subdomain are computed.

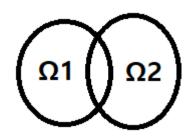


• 
$$\Omega = \Omega_1 \cup \Omega_2$$

- Schwarz iteration
  - PDE  $\rightarrow$  Au = b form
  - Supposing  $\mathbf{A}\mathbf{u}_1^k = \mathbf{b} \ \& \ \mathbf{A}\mathbf{u}_2^k = \mathbf{b}$ ,



 Convergence depends on boundary conditions and size of the overlapping region



#### References

- Wikipedia
- Wen Shen,
   An Introduction to Numerical Computation
- G. B. Arfken & H. J. Weber
   Mathematical Methods for Physicists

#### References

- C. Moler,
   Numerical Computing with MATLAB
- E. Weinan,
   Principles of Multiscale Modeling
- G. D. Smith,
   Numerical Solution of Partial Differential
   Equations: Finite Difference Methods

#### References

- A. Borzì, "Introduction to Multigrid Methods"
- N. Bellomo et al., "Lecture Notes on Mathematical Modelling in Applied Sciences"

### Investigation

About the Navier-Stokes equation

About the Black-Scholes equation

About the Euler-Tricomi equation

Not about numerical methods for these equations.