

Lecture 6b: Solution Strategies for Distributed Parameter Models

Rafiqul Gani (Plus material from Ian Cameron)

PSE for SPEED Skyttemosen 6, DK-3450 Allerod, Denmark rgani2018@gmail.com

www.pseforspeed.com



Contents

- Partial differential equation types
- Initial and boundary conditions
- Finite difference methods
- Method of lines
- Method of weighted residuals
 - ♦ collocation
 - orthogonal collocation



Parabolic

$$\frac{\partial c}{\partial t} = D\left(\frac{\partial^2 c}{\partial x^2}\right)$$

Elliptic

$$0 = D\left(\frac{\partial^2 c}{\partial x^2}\right)$$

Hyperbolic

$$\frac{\partial^2 c}{\partial t^2} = D\left(\frac{\partial^2 c}{\partial x^2}\right)$$

Initial and boundary conditions

Initial conditions needed for time varying problems

$$u(x,y,z,t) = f(u)$$
 at $t = 0$

Boundary conditions needed to set conditions on the balance volume surfaces. 3 main types exist:

- Dirichlet
- Neumann
- Robbins



Boundary Conditions



Dirichlet condition

$$\Phi = f(x, y)$$
 on \sum_{1}

Neumann condition

$$\frac{\partial \Phi}{\partial n} = g(x, y) \text{ on } \sum_{2}^{n} g(x, y)$$

Robbins (third) condition

$$a(x, y)\Phi + \beta(x, y)\frac{\partial \Phi}{\partial n} = \gamma(x, y) \text{ on } \sum_{3}$$



Solution methods for DPS models



Finite difference (FD)

Finite element (FE)

Weighted residuals (WR)

Method of lines (MOL) Shooting methods (SM)

Own reading: orthogonal collocation methods

Finite Difference Methods (FDM)

Popular and simple approach

- Seeks to replace derivative terms with finite difference approximations (FDA)
- Leads to large sets of algebraic equations (difference equations)
- Handles most problems with accuracy adjustable via grid spacing or order of FDA

Finite Difference Approximations (1)

✤ Based on a grid or mesh over the 1D or 2D domain

- Mesh can have equal or non-equal spacing
- Uses various manipulations of the Taylor Series to generate FDAs

Finite Difference Approximations (2)

Taylor series

$$u(x_{i} + \Delta x) = u(x_{i}) + \Delta x u'(x_{i}) + \frac{1}{2} \Delta x^{2} u''(x_{i}) + \frac{1}{6} \Delta x^{3} u'''(x_{i}) + K$$
$$u(x_{i} - \Delta x) = u(x_{i}) - \Delta x u'(x_{i}) + \frac{1}{2} \Delta x^{2} u''(x_{i}) - \frac{1}{6} \Delta x^{3} u'''(x_{i}) + K$$

First order approximations

$$u'_{i} = \frac{du(x_{i})}{dx} \simeq \frac{u_{i+1} - u_{i}}{\Delta x} + 0(\Delta x)$$
$$u'_{i} = \frac{du(x_{i})}{dx} \simeq \frac{u_{i} - u_{i-1}}{\Delta x} + 0(\Delta x)$$

Second order approximations

$$u'_{i} = \frac{du(x_{i})}{dx} \simeq \frac{u_{i+1} - u_{i-1}}{2\Delta x} + 0(\Delta x^{2})$$
$$u''_{i} = \frac{d^{2}u(x_{i})}{dx^{2}} \simeq \frac{u_{i+1} - 2u_{i} + u_{i-1}}{\Delta x^{2}} + 0(\Delta x^{2})$$

FDM - Application (1)

Unsteady diffusion $\frac{\partial c}{\partial t} = k \left(\frac{\partial^2 c}{\partial x^2} \right)$ u=0 u **=**0 u _{i, j+1} unknown values j+1 $_{\Delta}^{\mathsf{T}}$ BCs: u i-1,j u_{i,j} u i+1,,j known values j u(0,t) = 1; u(1,t) = 0**←** ∆x -ICs: ► X i-1 i+1 i. u(x,0) = 2x; $0 < x \le \frac{1}{2}$ $u(x,0) = 2(1-x); \frac{1}{2} < x < 1$ u=f(x)

FDM - Application (2)

gives a set of algebraic equations to be solved explicitly

*Apply boundary conditions to difference equations *Severe restrictions on *r* due to numerical stability of $0 < r \le \frac{1}{2}$

FDM – Crank-Nicholson methods

- Introduce some implicit character to FDA of governing equation
- Improve numerical stability
- Three main methods
 - fully explicit method (cf. Euler)
 - Crank-Nicholson (original) method
 - fully implicit method (cf. Backward Euler)

Crank-Nicholson mesh

Lecture 6b: Advanced Computer Aided Modelling

Crank-Nicolson application

Governing equation:

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

C-N-FDA:

$$\frac{u_{i,j+1} - u_{i,j}}{\Delta t} = \beta \left(\frac{u_{i+1,j+1} - 2u_{i,j+1} + u_{i-1,j+1}}{\Delta x^2} \right) + (1 - \beta) \left(\frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{\Delta x^2} \right)$$

Stable for:

$$r = \frac{\Delta t}{\Delta x^2} \le \frac{0.5}{(1 - 2\beta)}$$


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C-N - Stability Regions
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Lecture 6b: Advanced Computer Aided Modelling

Solving the finite difference equations

*****Obtain matrix form: $Au_{j+1} = f(u_j)$ (This is equivalent to the system: Ax=b)

Solve using linear matrix methods

- Matlab solution: $x=A\setminus b$
- Matlab factorization: [L,U]=lu(A);

$$y=L\setminus b$$
; $x=U\setminus y$;

✤Iterate if nonlinear (F(x)=0)

- Matlab solution: e.g. 'fsolve' routine

C-N solution example

for $\beta = \frac{1}{2}$ get: $-u_{i-1, j+1} + 4u_{i, j+1} - u_{i+1, j+1} = u_{i-1, j} + u_{i+1, j}$

FD methods for Parabolic Equations

FDM example - Parabolic Equation

♦Governing equation:

$$\frac{\partial u}{\partial t} = k \left(\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$

*****Explicit FDA:

$$\frac{u_{i,j,n+1} - u_{i,j,n}}{\Delta t} = k \left(\frac{u_{i-1,j,n} - 2u_{i,j,n} + u_{i+1,j,n}}{\Delta x^2} \right) + k \left(\frac{u_{i,j-1,n} - 2u_{i,j,n} + u_{i,j+1,n}}{\Delta y^2} \right)$$

Stability limit is:

$$k \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} \right) \Delta t \le \frac{1}{2}$$

Handling Derivative Boundary Conditions

Neumann condition

$$\frac{\partial T}{\partial y} = -T \quad \text{on} \quad y = 1$$

Create "false" boundary

 $\mathbf{\mathbf{\hat{v}}}$ Write central difference approximation for BC at j=N.

Eliminate fictitious value T_{i,N+1} by substitution of FDA of BC into FDA of equation.

Ν

N-1

Method of Lines (MOL)

Key features:

Converts parabolic PDEs to ODEs
 Discretizes the spatial variable(s) by using finite difference approximations
 Can result in sets of stiff ODEs
 Simple computer implementation

Method of Lines - application

Governing equation:

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

Discretize spatial terms to get:

$$\frac{du_i}{dt} = f_i(u) \qquad i = 0(1)N$$

Apply appropriate boundary conditions (at i=0, i=N)

Method of Lines - example

Problem:

$$\frac{\partial c}{\partial t} = \alpha \nabla^2 c + \beta R(c)$$

$$\frac{\partial c}{\partial r} = 0 \text{ at } r = 0; \quad \frac{\partial c}{\partial r} = Bi_w(c(1, z) - c_w(Z))$$

MOL gives:

$$\frac{dc_i}{dt} = \alpha \left(\frac{c_{i-1} - 2c_i + c_{i+1}}{\Delta r^2} + \left(\frac{\alpha - 1}{r_i} \right) \frac{c_{i+1} - c_{i-1}}{2\Delta r} \right) + \beta R_i$$

BCs give:

$$\frac{c_1 - c_{-1}}{2\Delta r} = 0 \quad ; \quad \frac{c_{N+1} - c_{N-1}}{2\Delta r} = Bi_w (c_N - c_w)$$
$$c_{-1} = c_1 \qquad ; \quad c_{N+1} = c_{N-1} + 2\Delta r Bi_w (c_N - c_w)$$

Method of Lines – example (cont.)

Final equation set:

$$\frac{dc_0}{dt} = \frac{2a\alpha}{\Delta r^2}(c_1 - c_0) + \beta R_c$$

$$\frac{dc_{i}}{dt} = \alpha \left(\frac{c_{i-1} - 2c_{i} + c_{i+1}}{\Delta r^{2}} + \left(\frac{\alpha - 1}{r_{i}} \right) \frac{c_{i+1} - c_{i-1}}{2\Delta r} \right) + \beta R_{i} \quad i = 1(1)N - 1$$

$$\frac{dc_N}{dt} = \alpha \left(\frac{2c_{N-1} + Bi_w 2\Delta r(c_N - c_w) - 2c_N}{\Delta r^2} + (a-1)(2\Delta r Bi_w(c_N - c_w)) \right) + \beta R_N$$

Method of Weighted Residuals (MOWR)

Key characteristics:

It is a polynomial approximation method with the following steps:

- choose a polynomial form
- fit the BCs to the polynomial
- substitute approximate solution into PDE
- distribute the error by making residual zero:

•collocation, least squares

•Galerkin, subdomain, moments

It is a powerful technique with accurate results from low order polynomials

MOWR - Application Example (1)

Trial function: ϕ_i

Fit BCs:

Problem:

Terms:

$$\phi_{N} = \sum_{i=0}^{N+1} c_{i} x^{i} \qquad \text{(power series in x)}$$

$$\phi_{N}(0) = 0 \rightarrow c_{0} = 0 \qquad \text{BC at x=0}$$

$$\phi_{N}(1) = 1 \rightarrow \sum_{i=1}^{N+1} c_{i} = 1 \qquad \text{BC at x=1}$$

$$\phi_{N} = x + \sum_{i=1}^{N} A_{i} \left(x^{i+1} - x \right) \qquad \text{modified trial function}$$

$$\phi_{1} = x + A_{1} \left(x^{2} - x \right)$$

$$\phi_{1}' = 1 + A_{1} \left(2x - 1 \right) \qquad \text{trial functions for N=1}$$

 $\frac{d}{dx}\left[(1+\alpha u)\frac{du}{dx}\right] = (1+\alpha u)\frac{d^2u}{dx^2} + \alpha \left(\frac{du}{dx}\right)^2 = 0 ; \quad u(0) = 0, \quad u(1) = 1$

MOWR - Application Example (2)

Form the residual equation (substitute trial function):

$$R(x,\phi_1) = \{1 + \alpha [x + A_1(x^2 - x)]\} 2A_1 + \alpha [1 + A_1(2x - 1)]^2$$

for collocation at $x = \frac{1}{2}$, $\alpha = 1$ residual $R(\frac{1}{2}, \phi_1)=0$ is:

$$-\frac{1}{2}A_1^2 + 3A_1 + 1 = 0$$
$$A_1 = -0.3166$$

Approximate first order solution is therefore:

$$\phi_1 = x - 0.3166(x^2 - x)$$
 $\varphi_1 = 0.5795; at, x = 0.5; (true = 0.5811)$

MOWR - Orthogonal Collocation

Similar to standard collocation except polynomials are orthogonal with predefined roots

Solve directly for solution y(x) and collocation point x;

Use of symmetric or non-symmetric polynomials depending on whether problem has symmetry, e.g. $\frac{\partial c}{\partial y} = 0$ at x = 0

Handles major geometries easily (built into polynomials)

- planar
- cylindrical
- spherical

Orthogonal Collocation with Symmetry

Problem: Diffusion and reaction in a planar catalyst pellet

$$\frac{d^2 y}{dx^2} = \phi^2 R(y) \qquad ; \ R(y) = y^2$$

BCs $\frac{dy}{dx} = 0 \quad at \ x = 0 \ ; \ y(1) = 1$

Residual equations obtained by substituting for terms as:

For N=1:
$$\sum_{i=1}^{N+1} B_{ji} y_i = \phi^2 R(y_j)$$
 $j = 1(1)N$

BC: $y_2 = 1$

Solution:

$$y_1 = \frac{-2.5 + \sqrt{6.25 + 10\phi^2}}{2\phi^2}$$

 $B_{11}y_1 + B_{12}y_2 = \phi^2 y_1^2$ $B_{11} = -2.5 ; B_{12} = 2.5$

These are pre-calculated coefficients

Example: OC without symmetry

Problem:

$$\frac{d}{dx} \begin{bmatrix} c(u) \frac{du}{dx} \end{bmatrix} = 0$$
$$u(0) = 0 \quad u(1) = 1$$
$$c(u) = 1 + u$$

$$(1+u)\frac{d^2u}{dx^2} + \left(\frac{du}{dx}\right)^2 = 0$$

Residual equations: $(1+u_j)\sum_{i=1}^{N+2} B_{ji}u_i + \left[\sum_{i=1}^{N+2} A_{ji}u_i\right]^2 = 0$ j = 2(1)(N+1)

BCs: $u_1 = 0$ $u_{N+2} = 1$

Solution is:

$$(1+u_2)\sum_{i=1}^{3} B_{2i}u_i + \left(\sum_{i=1}^{3} A_{2i}u_i\right)^2 = 0 \quad j=2$$

$$(1+u_2)(4u_1 - 8u_2 + 4u_3) + (-u_1 + u_3)^2 = 0$$

$$u_1 = 0 \quad , \quad u_3 = 1 \quad , \quad u_2 = 0.579$$

Orthogonal collocation on elements

- Useful technique for problems with sharp fronts
- Extension of standard OC method
- Easily implemented technique (convert PDEs and ODEs with polynomial coeffi cients <u>A</u> as a function of time; solve ODEs)
- Allows low order polynomials on the elements
- Computationally efficient

Example: OC for parabolic PDE

Problem:
$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x} \left(D(u) \frac{\partial u}{\partial x} \right) + R(u)$$

BCs:
$$u(0,t) = 1$$
; $u(1,t) = 0$ ICs: $u(x,0) = 0$

Residual equations:
$$\frac{du_j}{dt} = \sum_{i=1}^{N+2} A_{j,i} D(u_i) \sum_{z=1}^{N+2} A_{i,z} u_z + R(u_j)$$
 $j = 1,..., N$

BCs:
$$u_1(t) = 1$$
; $u_{N+2}(t) = 0$

Solve as a set of N ODEs

Modelling exercise – 5c: Solution of PDEs

Solve the following problems with MoT * 2nd order PDE * Unsteady state heat-transfer

* 1st- & 2nd-order PDAE

Model equations will be given in class